

5.

$$(a_1, a_2) + (0, 0) = (a_1, a_2)$$

$$(t+0)(a_1, a_2) = t(a_1, a_2) + 0(a_1, a_2)$$

$$t[(a_1, a_2) + (0, 0)] = t(a_1, a_2)$$

$$t(a_1, a_2) = t(a_1, a_2) + 0(a_1, a_2)$$

$$t(a_1, a_2) + t(0, 0) = t(a_1, a_2)$$

Suppose  $t(a_1, a_2) \neq (ta_1, ta_2)$ , then

$$0(a_1, a_2) \neq (0a_1, 0a_2) = (0, 0)$$

$$\Rightarrow t(a_1, a_2) + 0(a_1, a_2) \neq t(a_1, a_2) + (0, 0)$$

$$\Rightarrow (t+0)(a_1, a_2) \neq t(a_1, a_2)$$

$\Rightarrow$

By the property of a vector space  $V$  over a field  $F$  that:

(VS 8) For  $a, b \in F$  and  $x = (a_1, a_2) \in V$ ,  $(a+b)x = ax + bx$ ;

Let  $t \in F$ ,

$$\overrightarrow{AB} = \overrightarrow{DC} \quad \overrightarrow{DA} = \overrightarrow{CB}$$

$$(b_1 - a_1, b_2 - a_2) = (c_1 - d_1, c_2 - d_2)$$

$$b_1 - a_1 = c_1 - d_1 \quad b_2 - a_2 = c_2 - d_2$$

$$b_1 + d_1 = a_1 + c_1 \quad b_2 + d_2 = a_2 + c_2$$

$$(x+y)^{2n} = \sum_{i=0}^{2n} \binom{2n}{i} x^{2n-i} y^i$$

$$\frac{2n}{2} = n \quad n+1$$

$n-1$  in front  $(n+1)-1 = n$   
 $2n-n = n$  behind  $2n+1-(n+1) = n$

Total number of terms:  $2n+1$

Given any sequence with an odd number of terms, i.e.  $2n+1$  number of terms, the middle term is the  $(n+1)$ th term as the number of terms before and after it are equal, i.e.:

Number of terms before:  $(n+1)-1 = n$

Number of terms after:  $2n+1-(n+1) = n$

$$\binom{2n}{i} x^{2n-i} y^i = \binom{n+1}{i} x^{n-i} y^{n+1-i}$$

$$\forall x \forall y (x, y \in V \Rightarrow \exists! z (x+y=z)) \iff \forall x \forall y (x, y \in V \Rightarrow [\exists z (x+y=z) \wedge \forall z_1, \forall z_2 (x+y=z_1 \wedge x+y=z_2 \Rightarrow z_1=z_2)])$$

$$\exists z [x+y=z \wedge \forall \xi (x+y=\xi \Rightarrow z=\xi)]$$

$$\text{If } (u_1+W) + (v_1+W) = \overset{(a+b)+W}{z_1} \text{ and } (u_1+W) + (v_2+W) = \overset{(c+d)+W}{z_2}, \\ = (u_1+v_1)+W \qquad = (u_1+v_2)+W$$

$$x R y_1 \wedge x R y_2 \Rightarrow y_1 = y_2$$

$$4 \sqrt{12} \wedge 4 \sqrt{12} = 2$$

$$x_1 = x_2 \Rightarrow (x_1 R y_1 \wedge x_2 R y_2 \Rightarrow y_1 = y_2)$$

$$\text{Conclude: } z_1 = z_2$$

$$(a+b)+W = (c+d)+W$$

If  $u+W$  is a subspace of  $V$ ; then  $u+W$  is closed under addition (A1) and for all elements of  $u+W$ , there exists an additive inverse  $-(u+W) \in u+W$ . Plus,  $\vec{0} \in V$ ;

$$\text{For } w = \vec{0}, \quad u + \vec{0} \in u+W$$

$$u + (-u) = \vec{0} \in u+W$$

If  $u \notin W$  and  $v \in V$ , then

$$u+W \neq W$$

$x+y$  unique

$$x_1+y_1 = x_2+y_2$$

$$x_1 = x_2 \text{ and } y_1 = y_2$$

$$au_1 + w = a_1u_1 + a_2w_1 + a_3u_1 + (-a_4w_1)$$

$$= a_1u_1 + a_2w_1$$

$$= w + au_1$$

$$\{u_1 + v_2 + w_1 + w_2\}$$

$$a_4 + w_1$$

$$a_1u_1 + a_2w_1 + a_3u_1 + a_4(-w_1)$$

$$v_1 - v_1' \in W \quad v_2 - v_2' \in W$$

$$v_1' - v_1 \in W \quad v_2' - v_2 \in W$$

$$v_1 + W_1 \quad v_2 + W_2$$

$$v_1 + W_2$$

$$u + (-u) = -u + (-(-u)) = \vec{0}$$

$$-u + (-u) + (-(-u)) = -u$$

$$\vec{0} + (-(-u)) = u$$

$$u = -(-u)$$

$$v_2 + W = \{v_2 + w \mid v_2 \in V \text{ and } w \in W\}$$

$$= \{v_1 + v_2 + (-v_2) \mid v_1, v_2 \in V \text{ and } w \in W\}$$

$$= \{v_1 + w \mid v_1 \in V \text{ and } w \in W\}$$

$$= v_1 + W$$

$$a(u + \frac{1}{a}w)$$

$$au + w$$

$$(v_1 + v_2) + (-(-v_3 + v_4)) \in W$$

$$(v_1 + v_2) + W = (v_3 + v_4) + W$$

$$(v_1 + W) + (v_2 + W) = (v_3 + W) + (v_4 + W)$$

$$\{v_1 + W\} + \{v_2 + W\} = \{v_3 + W\} + \{v_4 + W\}$$

$$\{v_1 + v_2 + w \mid w \in W\} = \{v_3 + v_4 + w \mid w \in W\}$$

$$v_1 + v_2 + w_0 = v_3 + v_4 + w_0$$

$$(v_1 + v_2) + (w_0) + (-(-v_3 + v_4)) = (v_3 + v_4) + (-(-v_3 + v_4)) + (v_1 + v_2) + w_0$$

$$(1+1)(v_1 + v_2) + (-(-v_3 + v_4)) = (v_1 + v_2)$$

What we want:  $v_1 + W = v_3 + W$  and  $v_2 + W = v_4 + W$   
 OR  $v_1 + W = v_4 + W$  and  $v_2 + W = v_3 + W$

$$(A^t)_{ij} = A_{j,i} \quad (-A)_{ij} = -(A_{ij}) \quad (C^t)_{ij} = C_{j,i} = C_{ij}$$

$$A_{j,i} = -(A_{i,j})$$

$$A_{i,j} + A_{j,i} = 0$$

$$X \in M_{n \times n}(F) \iff X \in W_1 + W_2 = \{y + z \mid y \in W_1 \text{ and } z \in W_2\}$$

Since  $W_1$  and  $W_2$  are subspaces of  $M_{n \times n}(F)$ , and  $M_{n \times n}(F)$  is a vector space, hence closed under addition (A1);  $A+C \in M_{n \times n}(F)$ .

Conversely, if  $X \in M_{n \times n}(F)$ , then we can construct an  $E \in W_1$  and  $F \in W_2$  such that  $X = E + F$ :

$$\{t \mid F(\psi(x))\}$$

$$\text{ran } \psi = \{t \mid \exists x(x \psi t)\} = A/\sim$$

$$\text{dom } \hat{F} = \{t \mid \exists z(t \hat{F} z)\} = A/\sim$$

$$F^{-1}(b) = \{t \mid b F^{-1} t\} = \{t \mid b F t\} = F^{-1}(b)$$

$$(\text{ran } \psi) \cap (\text{dom } \hat{F}) = A/\sim = \{t \mid t F b\} = \{t \mid t F B\}$$

$$\{t \mid \exists x(x \psi t)\} \cap \{t \mid \exists z(z \hat{F} t)\} = A/\sim$$

$$\{t \mid \exists x \exists z(x \psi t \wedge z \hat{F} t)\} = A/\sim$$

$$\text{dom } (F \circ \psi) = A/\sim$$

$$x(\hat{F} \circ \psi)y \quad x \psi u \wedge u \hat{F} v$$

$$\text{dom } (\hat{F} \circ \psi) = \{x \mid \exists u \exists v(x \psi u \wedge u \hat{F} v)\}$$

$$= \{x \mid \exists u(x \psi u)\} \cap \{x \mid \exists v(u \hat{F} v)\}$$

$$= \{x \mid \exists u(x \psi u)\}$$

$$= \text{ran } \psi \cap \text{dom } \hat{F}$$

$$x \in A \implies \langle x, [x]_n \rangle \in \psi$$

$$[x]_n \in A$$

$$\begin{aligned} i_{ij} - x_{jii} &= -F_{ij} + x_{ij} \\ (1+1)F_{ij} &= x_{ij} + x_{jii} \\ F_{ij} &= \frac{x_{ij} + x_{jii}}{1+1} \end{aligned}$$

$$E_{ij} = F_{ij} - x_{jii}$$

$$E_{ij} = -F_{ij} + x_{ij}$$

$$-E_{ij} + F_{ij} = x_{jii} \implies F_{ij} = E_{ij} + x_{jii}$$

$$E_{ij} + F_{ij} = x_{ij} \implies F_{ij} = x_{ij} - E_{ij}$$

$$E_{ij} + x_{jii} = x_{ij} - E_{ij}$$

$$(1+1)E_{ij} = x_{ij} - x_{jii}$$

$$E_{ij} = \frac{x_{ij} - x_{jii}}{1+1}$$

|          |          |          |          |           |          |          |          |
|----------|----------|----------|----------|-----------|----------|----------|----------|
| $x_{11}$ | $x_{12}$ | $x_{13}$ | $x_{14}$ | $E_{11}$  | $e$      | $F_{11}$ | $f$      |
| $x_{21}$ | $x_{22}$ | $x_{23}$ | $x_{24}$ | $-E_{11}$ | $E_{22}$ | $f$      | $F_{22}$ |
| $x_{31}$ | $x_{32}$ | $x_{33}$ | $x_{34}$ | $E_{33}$  | $E_{44}$ | $f$      | $F_{33}$ |
| $x_{41}$ | $x_{42}$ | $x_{43}$ | $x_{44}$ | $E_{44}$  | $E_{33}$ | $f$      | $F_{44}$ |

$V = W_1 \oplus W_2$  iff ( $W_1$  and  $W_2$  are subspaces of  $V$ )  $W_1 \cap W_2 = \emptyset$  and  $W_1 + W_2 = V$

$\Rightarrow$ )  $V = W_1 \oplus W_2$  implies that there exists a unique  $x_1 \in W_1$  and  $x_2 \in W_2$  such that  $x_1 + x_2 \in V$ :

Immediately means that  $W_1 \cap W_2 = \emptyset$  and  $W_1 + W_2 = V$   
 $= \{x_1 + x_2 \mid x_1 \in W_1 \text{ and } x_2 \in W_2\}$

Let  $y_1 \in W_1$  and  $y_2 \in W_2$ ; then  $y_1 + y_2 \in V$ .

$$x_1 + x_2 = y_1 + y_2 \Rightarrow x_1 = y_1 \text{ and } x_2 = y_2$$

①  $x_1 = y_1 + y_2$  X    ②  $x_2 = y_1 + y_2$  X    ③  $x_1 = y_1$  and  $x_2 = y_2$     ④ ...

$\Leftarrow$ ) Assume every vector in  $V$  can be uniquely written as  $x_1 + x_2$ , where  $x_1 \in W_1$  and  $x_2 \in W_2$ :

Then if  $x_1 + x_2 = y_1 + y_2$ , then  $x_1 = y_1$  and  $x_2 = y_2$ .

$z \in W_1$  and  $z \in W_2$

$$z + x_2 = y_1 + z$$

$$\begin{aligned}
 (F^{-1})^{-1} &= \{ \langle x, y \rangle \mid x(F^{-1})^{-1}y \} \\
 &= \{ \langle x, y \rangle \mid yF^{-1}x \} \\
 &= \{ \langle x, y \rangle \mid xFy \} \\
 &= F
 \end{aligned}$$

$$\begin{aligned}
 F &= \{ \langle x, y \rangle \mid xFy \} \\
 &= \{ \langle x, y \rangle \mid yF^{-1}x \} \\
 &= \{ \langle x, y \rangle \mid x(F^{-1})^{-1}y \}
 \end{aligned}$$

$$\exists x (\langle [x]_n, b_1 \rangle \neq \langle x, b_2 \rangle) \\
 [x]_n \neq x \text{ OR } b_1 \neq b_2$$

$$\begin{aligned}
 \text{ran}(\hat{G} \circ \varphi) &= \{ t \mid \exists u \exists v (u \varphi v \wedge v \hat{G} t) \} \\
 &= \{ t \mid \varphi(u) \in \text{dom } \hat{G} \wedge t \in \text{ran } \hat{G} \} \\
 &= \{ t \in \text{ran } \hat{G} \mid \varphi(u) \in \text{dom } \hat{G} \} \\
 &= \{ t \in \text{ran } \hat{G} \mid \varphi(u) \in A/n \} \\
 &\subset \text{ran } \hat{G} \neq \text{ran } \hat{F}
 \end{aligned}$$

$$\forall t (t \in e \iff t \in X \wedge x \in X)$$

$$\forall x \forall y \forall z \forall t [t \in e \iff (t \in X \wedge x \in X \wedge x \sim t)]$$

For all  $x, y$ :

$$\begin{aligned}
 x(F^{-1})^{-1}y &\iff yF^{-1}x \\
 &\iff xFy
 \end{aligned}$$

Assume  $F$  is single-valued, i.e.  $(x_1 F y \wedge x_2 F y) \implies x_1 = x_2$

$$\hat{F}^{-1} = \{ \langle y, x \rangle \mid x F y \}$$

then,  $(y F^{-1} x_1 \wedge y F^{-1} x_2) \implies x_1 = x_2$

$$A/\sim = \{[x] \mid x \in A\}$$

$$y \in A/\sim \iff \exists x (x \in A \wedge [x] = y)$$

$$\iff \exists x (x \in A \wedge \{t \mid x \sim t\} = y)$$

$$\iff \exists x (x \in A \wedge t \in y)$$

$$t \in [x] \iff \exists t (x \in A \wedge x \sim t)$$

$$\exists x (x \in A \wedge [t \in y] \iff \exists x (x \in A \wedge x \sim t))$$

$$\exists x (x \in A \wedge [t \in y] \iff x \sim t)$$

$$\langle u, v \rangle \dot{Q} \langle x, y \rangle \iff u + v = x + y$$

$$2v + 2x = 2y + 2u$$

$$(u + y) + (2v + 2x) = (x + v) + (2y + 2u)$$

$$(u + 2v) + (y + 2x) = (x + y) + (v + 2u)$$

$$u + 2v + y + 2x = x + y + v + 2u$$

$$\langle u + 2v, v + 2u \rangle \dot{Q} \langle x + 2y, y + 2x \rangle \quad (u + 2v) + (v + 2u) = (x + 2y) + (y + 2x)$$

$$\forall t (t \in e \iff t \in A \wedge x \sim t)$$

$$\exists x [x \in A \wedge (P(x, t) \iff \exists u [x \in A \wedge Q(x, t)])]$$

$$\exists x [x \in A \wedge (P(x, t) \iff [x \in A \wedge Q(x, t)])]$$

$$\exists x [P(x) \wedge (A \wedge t) \iff \dots]$$

$$\forall x (P(x) \iff Q(x))$$

$$\forall x (P(x)) \iff \forall x (Q(x))$$

$$\forall y [y \in A/\sim \iff \exists x (\varphi(x) = y)]$$

$$\forall x ([x] \in A/\sim \iff \varphi(x) = [x])$$

$$\forall y (\exists x (\varphi(x) = y = [x]) \implies y \in A/\sim)$$

$$\forall x (\varphi(x) = [x] \implies [x] \in A/\sim)$$

$$\forall x ([x] \in A/\sim \implies x \in A \implies \varphi(x) = [x])$$

$$\forall y [y \in A/\sim \iff (y \in \mathcal{P}(\text{ran } \varphi) \wedge \forall x [x \in y \iff \exists t (x \in A \wedge x \sim t)])]$$

$$x_1 + 2x_2 - x_3 + x_4 = 5$$

$$2x_2 - 2x_3 - 4x_4 = 1 \implies x_2 - x_3 - 2x_4 = \frac{1}{2}$$

$$-x_2 + x_3 + 2x_4 = -2$$

$$0 = -\frac{3}{2}$$

$$\text{Span}(S_1) + \text{Span}(S_2) = \{x+y \mid x \in \text{Span}(S_1) \text{ and } y \in \text{Span}(S_2)\}$$

$$\begin{aligned} x_1 + 2x_2 + 6x_3 &= -1 \checkmark & x_1 + 2x_2 + 6x_3 &= -1 \\ -3x_2 - 11x_3 &= 10 \checkmark & x_2 + 4x_3 &= -4 \implies x_2 = -28 \\ -5x_2 - 17x_3 &= 18 \checkmark & x_3 &= 6 \\ x_2 + 4x_3 &= -4 \checkmark & -5x_2 - 14x_3 &= 18 \end{aligned}$$

$\implies x_1 = 19$

$26 = 18$   
Not possible

$$a(x^3 + x^2 + x + 1) + b(x^2 + x + 1) + c(x + 1) = 2x^3 - x^2 + x + 3$$

$$ax^3 + (a+b)x^2 + (a+b+c)x + a+b+c = 2x^3 - x^2 + x + 3$$

$$\begin{aligned} a &= 2 \\ a+b &= -1 \\ a+b+c &= 1 \\ a+b+c &= 3 \end{aligned}$$

$$\begin{aligned} \begin{pmatrix} 1 & 2 \\ -3 & 4 \end{pmatrix} &= a \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + c \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \\ &= \begin{pmatrix} a & 0 \\ -a & 0 \end{pmatrix} + \begin{pmatrix} 0 & b \\ 0 & b \end{pmatrix} + \begin{pmatrix} c & c \\ 0 & 0 \end{pmatrix} \\ &= \begin{pmatrix} a+c & b+c \\ -a & b \end{pmatrix} \end{aligned}$$

$$\begin{aligned} a+c &= 1 \\ b+c &= 2 \\ -a &= -3 \implies a=3 \\ b &= 4 \end{aligned}$$

$$\begin{aligned} 3+c &= 1 \implies 3-2=1 \\ 4+c &= 2 \implies c=-2 \end{aligned}$$

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$$\text{Span}(S) = \left\{ a_1 u_1 + a_2 u_2 + \dots + a_n u_n \mid a_1, a_2, \dots, a_n \in F \text{ and } u_1, u_2, \dots, u_n \in S \right\}$$

(A1)

$$\sum_{k=0}^n a_k i_k + \sum_{k=0}^m b_k j_k \Rightarrow \text{Sum of } n+m \text{ terms}$$

Finite sum of multiples of vectors in  $S \Rightarrow$  By definition in  $\text{Span}(S)$ .

$a_i, b_j \in F$  since fields are closed under addition.

(M1)

$$c \sum_{k=0}^n a_k i_k = \sum_{k=0}^n c(a_k i_k) = \sum_{k=0}^n (ca_k) i_k \in \text{Span}(S)$$

$$\text{Span}(S) = \left\{ \sum_{i=1}^n a_i v_i \mid a_i \in F \text{ and } v_i \in S, \text{ for all natural numbers } i \text{ and } n \text{ such that } 1 \leq i \leq n \right\}$$

Exists some vector  $s \in \text{Span}(S)$  such that  $\sum_{i=1}^n a_i v_i = \sum_{i=1}^m b_i u_i = s$

Info:  $\sum_{i=1}^n a_i v_i = \vec{0} \implies a_i = 0$  for all  $i$

(which also means  $\vec{0} \notin S$ )

Show:  $n = m$ ; for all  $i$ ,  $a_i = b_i$  and  $u_i = v_i$

Proof: If  $\vec{0} \neq s$ ,  
 $\sum_{i=1}^n a_i v_i + \left( - \sum_{i=1}^m b_i u_i \right) = \vec{0}$  where  $a_i \neq 0$  and  $b_i \neq 0$

$$\sum_{i=1}^{\max(n,m)} a_i v_i - b_i u_i = \vec{0} \quad \text{where we define } a_i = 0 \text{ and } v_i = 0 \text{ if } i > n$$

$$b_i = 0 \text{ and } u_i = 0 \text{ if } i > m$$

$n = m$  For all  $j$ , there exists a  $k$  such that

$$a_j v_j - b_k u_k = \vec{0}$$

Simultaneously,

$$a_j v_j + (-b_k) u_k = \vec{0} \implies a_j = b_k = 0$$

$$a_j v_j = b_k u_k$$

$a_j = b_k = 0$  OR  $a_j \neq 0$  and  $b_k \neq 0$  but  $a_j \neq b_k$ .

$$\implies u_k = (b_k^{-1} \cdot a_j) (v_j)$$

$$\alpha v_i + \beta u_j = \vec{0}$$

$$\alpha v_i + \beta (b_k^{-1} \cdot a_j) (u_j) = \vec{0}$$

OR  $a_j = b_k \neq 0$

$$\implies v_j = u_k$$

If there exists  $u_k$  such that  $u_k \in \text{Span}\{u_1, u_2, \dots\}$  then

$$a_1 v_1 + a_2 v_2 + \dots + a_n v_n + b_k u_k = \vec{0}$$

$$\implies a_1, a_2, \dots, a_n, b_k = \vec{0}$$

$v \in \text{Span}(S_1 \cap S_2) \implies$  exists  $u_1, u_2, \dots, u_n \in S_1 \cap S_2$  and  $a_1, a_2, \dots, a_n \in \mathbb{F}$  such that  $v = \sum_{i=1}^n a_i u_i$

$\implies v \in S_1$  and  $v \in S_2$ .

$v \in \text{Span}(S_1) \cap \text{Span}(S_2) \implies v \in \text{Span}(S_1)$  and  $v \in \text{Span}(S_2)$

$\sum_{i=0}^n a_i x_i \in \text{Span}(S_1)$  and  $\sum_{i=0}^m b_i y_i \in \text{Span}(S_2)$ , where  $a_1, a_2, \dots, a_n, b_1, b_2, \dots, b_m \in \mathbb{F}$ ;  
 $x_1, x_2, \dots, x_n \in \text{Span}(S_1)$   
 and  
 $y_1, y_2, \dots, y_m \in \text{Span}(S_2)$

Basically,  $\sum_{i=0}^n a_i x_i = \sum_{i=0}^m b_i y_i = v \in \text{Span}(S_1) \cap \text{Span}(S_2)$ .

However, it may not be the case that  $n=m$  or  $a_i = b_i$  for all  $i$  or  $x_i = y_i$  for all  $i$ .  
 Just as long as they add up to be equivalent.

Example:

$V = \mathbb{R}^2$  over  $\mathbb{R}$

$$\begin{aligned} \text{Span}(\{(0,1), (2,0)\} \cap \{(0,1), (0,2)\}) &= \text{Span}(\{(0,1)\}) \\ &= \{a(0,1) \mid a \in \mathbb{R}\} \\ &= \{(0, a) \mid a \in \mathbb{R}\} \end{aligned}$$

$$\begin{aligned} \text{Span}(\{(0,1), (2,0)\}) \cap \text{Span}(\{(0,1), (0,2)\}) &= \{(2b, a) \mid a, b \in \mathbb{R}\} \cap \\ &\quad \{(0, c+2d) \mid c, d \in \mathbb{R}\} \\ &= \{(0, e) \mid e \in \mathbb{R}\} \end{aligned}$$

$$\forall x \forall y (x, y \in A \wedge x < y \implies f(x) < f(y))$$

$$\forall x \forall y (x, y \in A \wedge x = y \implies f(x) = f(y) \in A)$$

$$\forall x \forall y [(f(x) > f(y) \vee f(x) = f(y)) \implies (x > y \wedge x = y)]$$

$$\text{all } x, y \in A; \quad f(x) < f(y) \quad \text{OR} \quad \underline{f(x) = f(y)} \quad \text{OR} \quad f(y) < f(x)$$

Assume  $x \neq y$ . Then,

1.  $x < y$ , implying  $f(x) < f(y)$

2.  $y < x$ , implying  $f(y) < f(x)$

$$f(x) < f(y)$$

$$\text{Span}\left(\bigcup_{i \in I} S_i\right) = W \text{ where } I \text{ is a finite set and } S_i = S_j \text{ iff } i = j$$

$$v_1, v_2, \dots, v_n \in S_i \implies a_1 v_1 + a_2 v_2 + \dots + a_n v_n = \vec{0}$$

$$c_1 v_1, c_2 v_2, \dots, c_n v_n \in W \quad \text{Span}(\{v_1, v_2, \dots, v_n\}) = \text{Span}(\{c_1 v_1, c_2 v_2, \dots, c_n v_n\})$$

If  $F$  is a finite field of characteristic  $p$ ,

$$F = \{0, 1, 1+1, \dots, \sum_{i=1}^{p-1} 1+1\} \quad ? \text{ Probably not}$$

$$\forall x (x \in A \implies \neg x R x)$$

$$\forall x [(x R y \wedge y R z) \implies x R z] \quad (x R y \wedge y R x) \implies x R x$$

$$\forall x \forall y [x \neq y \implies ((x R y \wedge \neg y R x) \vee (\neg x R y \wedge y R x))]$$

If the field  $F$  is finite, then it has a nonzero field characteristic.

The statement is true for 0:  $0^3 = 0^3 + 0^3 = 0$ !

Assume it is true for some  $n \in \mathbb{N}$ , then it is true in the  $(n+1)$ th case as well, because:

$$(n+1)^3 = \sum_{i=1}^{n+1} i^3 - \sum_{i=1}^n i^3 = \sum_{i=1}^n i^3 - \sum_{i=1}^{n-1} i^3 + (n+1)^3 - n^3$$

$$\begin{aligned} n^3 &= \sum_{i=1}^n i^3 - \sum_{i=1}^{n-1} i^3 \\ &= \left[ \frac{n(n+1)}{2} \right]^2 - \left[ \frac{(n-1)(n-1+1)}{2} \right]^2 \end{aligned}$$

$$\langle a_1, b_1 \rangle <_L \langle a_2, b_2 \rangle \wedge \langle a_2, b_2 \rangle <_L \langle a_3, b_3 \rangle$$

$$\iff [a_1 <_A a_2 \vee (a_1 = a_2 \wedge b_1 <_B b_2)] \wedge [a_2 <_A a_3 \vee (a_2 = a_3 \wedge b_2 <_B b_3)]$$

1.  $a_1 <_A a_2$  and  $a_2 <_A a_3 \Rightarrow a_1 <_A a_3$  by the transitivity of the linear ordering  $<_A$ . Therefore,  $\langle a_1, b_1 \rangle <_L \langle a_3, b_3 \rangle$ .

2.  $a_1 <_A a_2$  and  $(a_2 = a_3 \wedge b_2 <_B b_3) \Rightarrow a_1 <_A a_3$ . So,  $\langle a_1, b_1 \rangle <_L \langle a_3, b_3 \rangle$ .

3.  $(a_1 = a_2 \wedge b_1 <_B b_2)$  and  $a_2 <_A a_3 \Rightarrow a_1 <_A a_3$ . Hence,  $\langle a_1, b_1 \rangle <_L \langle a_3, b_3 \rangle$ .

4.  $(a_1 = a_2 \wedge b_1 <_B b_2)$  and  $(a_2 = a_3 \wedge b_2 <_B b_3) \Rightarrow a_1 = a_3 \wedge b_1 <_B b_3$  by the transitivity of the linear ordering  $<_B$ . Accordingly,  $\langle a_1, b_1 \rangle <_L \langle a_3, b_3 \rangle$ .

$$\{(0,0), (0,1), (0,2), (1,0), (1,1), (1,2), (2,0), (2,1), (2,2)\} \iff (x \in A \vee y \in B) \wedge [(x \in A \wedge y \in A) \vee (x \in B \wedge y \in C)]$$

$$(x \in A \wedge y \in A) \vee (x \in B \wedge y \in C)$$

$$(x \in A \vee x \in B) \wedge (y \in A \vee y \in C)$$

Set of all equivalence relations on  $A$ : Let's call it  $S$ .

$$S = \left\{ \left\{ \{x, x\} \mid x \in A \right\} \cup \left\{ \{(x, y), (z, z)\} \mid x, y \in A \wedge x \neq z \wedge y \neq z \right\} \mid z \in A \right\}$$

$$\forall A \exists S \forall K K \in S \iff K \in \mathcal{P}(A \times A) \wedge \forall k k \in K \Rightarrow \exists a \exists b (a, b \in A \wedge \langle a, b \rangle \in k)$$

$$\forall z_1, z_2, \dots, z_n \forall x \forall y \left( \begin{array}{l} \exists k k \in K \wedge \\ z_1, z_2, \dots, z_n, x, y \in A \wedge x \neq z_1 \wedge x \neq z_2 \wedge \dots \wedge x \neq z_n \\ \wedge y \neq z_1 \wedge y \neq z_2 \wedge \dots \wedge y \neq z_n \end{array} \right)$$

$$\Rightarrow (x, x), (y, y), (x, y), (y, x), (z_1, z_1), (z_2, z_2), \dots, (z_n, z_n) \in k$$

$$\begin{aligned} 2a_1 + a_2 + a_3 &= 1 \\ -a_1 - a_2 + a_3 &= -2 \\ 4a_1 + 3a_2 - a_3 &= -1 \end{aligned}$$

$$\begin{aligned} -a_2 + 3a_3 &= -1 \\ -a_1 - a_2 + a_3 &= -2 \\ -a_2 + 3a_3 &= -9 \end{aligned}$$

$$\begin{aligned} |a_1 + a_2 - a_3| &= 2 \\ |a_2 - 3a_3| &= 1 \\ a_2 - 3a_3 &= -9 \end{aligned}$$

contradiction:  $1 \neq -9$  ✓

$$x_1 = x_2 \Rightarrow f(x_1) = f(x_2)$$

$$xRy_1 \wedge xRy_2 \Rightarrow y_1 = y_2$$

$$y_1 = f(x) \wedge y_2 = f(x) \Rightarrow y_1 = y_2$$

$$x_1 = x_2 \wedge y_1 = f(x_1) \wedge y_2 = f(x_2) \Rightarrow y_1 = y_2$$

$$x_1 = x_2 \Rightarrow f(x_1) = f(x_2)$$

$$\begin{aligned} \exists u_1, \exists u_2, \dots, \exists u_n \exists a_1, \exists a_2, \dots, \exists a_n & \left( \begin{array}{l} u_1, u_2, \dots, u_n \in S \\ \wedge a_1, a_2, \dots, a_n \in F \end{array} \wedge (a_1 \neq 0 \vee a_2 \neq 0 \vee \dots \vee a_n \neq 0) \wedge a_1 u_1 + a_2 u_2 + \dots + a_n u_n = \vec{0} \right) \\ \forall u_1, \forall u_2, \dots, \forall u_n \forall a_1, \forall a_2, \dots, \forall a_n & \left( \begin{array}{l} u_1, u_2, \dots, u_n \notin S \vee \\ \vee a_1, a_2, \dots, a_n \notin F \end{array} \vee (a_1 = 0 \vee a_2 = 0 \vee \dots \vee a_n = 0) \vee \neg (a_1 u_1 + a_2 u_2 + \dots + a_n u_n = \vec{0}) \right) \\ \vec{0} & \Rightarrow \left( \begin{array}{l} u_1, u_2, \dots, u_n \notin S \vee \\ \vee a_1, a_2, \dots, a_n \notin F \end{array} \right) \end{aligned}$$



$$\sum_{i=1}^m \left[ \sum_{n=1}^m (b_n \cdot a_{i,n}) x^i \right]$$

For all  $i$ , coefficient of  $x^i$  is 0.  
 i.e. For each and every  $i$ ,

$$b_n \cdot a_{i,n} = 0 \text{ for all } n.$$

$$\sum_{n=1}^m \left( b_n \cdot \sum_{i=1}^m a_{i,n} x^i \right) = 0$$

$$\sum_{n=1}^m \left[ \sum_{i=1}^m (b_n \cdot a_{i,n}) x^i \right] = 0$$

$$\sum_{n=1}^m \left[ \sum_{i=1}^m (b_n \cdot a_{i,n}) x^i \right] = 0 \text{ unless } a_{i,n} = 0 \text{ if } i > n$$

$$\forall i \forall n (b(n) \cdot a(i,n) = 0) \iff \forall i \forall n (b(n) = 0 \vee a(i,n) = 0)$$

$$n \cup \{n\} = m \cup \{m\}$$

$$\forall x (x \in n \vee x \in \{n\}) \iff x \in m \vee x \in \{m\}$$

$$\forall x (x \in n \vee x = n) \iff x \in m \vee x = m$$

$$[(x \in n \vee x = n) \wedge (x \in m \vee x = m)] \vee [(x \neq n \wedge x \neq n) \wedge (x \neq m \wedge x \neq m)]$$

$$[(x \in n \vee x = n) \wedge x \in m] \vee [(x \in n \vee x = n) \wedge x = m]$$

$$(x \in n \wedge x \in m) \vee (x = n \wedge x \in m) \vee (x \in n \wedge x = m) \vee (x = n \wedge x = m)$$

$$\underline{n \in m \vee n = m}$$

$$m \in n \vee m = n$$

Assume  $n \in m$ , then  $n \neq m$ .

$$\{n\} \cap \{\{n\}\} = \emptyset$$

$$n \in \{n\} \wedge n \in \{\{n\}\} \iff n \in \emptyset$$

$$n = \{\{n\}\}$$

$$\{n\} \in n$$

$$n = \{m\} \wedge \{n\} = m$$

$$n \in n \wedge n \in m \quad n \in m$$

$$n \in \{n\} \in n$$

$$n \in m \cap n$$

$$\{\{n\}\} \cap \{n\} = \emptyset$$

$$x \in \{\{n\}\} \wedge x \in \{n\} \iff x \in \emptyset$$

$$x = \{\{n\}\} \wedge x = n \iff x \in \emptyset$$

$\emptyset \in \emptyset$  False  
 Let  $x \neq \emptyset$ ; then there exists some set  $y \neq \emptyset$  such that  $y \in x$   
 $\emptyset \in x$   
 $x = \{\emptyset, y, \dots\}$

$$x \in \cup x$$

$$x \in x$$

$$\{x\} \subseteq x$$

$$x = x$$

$$\forall y (y \in x \iff y \in x)$$

$$y \in x \implies y \in$$

$$n \cap \{n\} = \emptyset$$

$$n \in n \wedge n = n \iff n \in \emptyset$$

$$n \in n \iff n \in \emptyset$$

$$\{n\} \in n \wedge \{n\} \in \{n\} \iff n \in \emptyset$$

$$\exists x (x \in a \Rightarrow x \in a^+) \Leftrightarrow a^+ \neq \emptyset$$

$$\neg [\forall x (x \in A \Leftrightarrow x \in \emptyset)]$$

$$[(\neg P \vee \neg Q) \wedge (P \vee Q)] \Leftrightarrow [(P \Rightarrow Q) \wedge (P \vee Q)]$$

$$\exists x [(x \notin A \vee x \notin \emptyset) \wedge (x \in A \vee x \in \emptyset)]$$

$$(x \in \omega' \Leftrightarrow [x \in \omega \wedge \exists y (y \in \omega \wedge y^+ = x)])$$

$$\neg [\exists A \exists n (A \text{ is inductive} \wedge n \in \omega \wedge |A| = n)]$$

$$\forall A \forall n (A \text{ is not inductive} \vee n \notin \omega \vee |A| \neq n)$$



$$n^+ = \{x \in \omega \mid \dots\}$$

$$\mathcal{P}(n^+) = \mathcal{P}(m^+)$$

$$\mathcal{P}(n \cup \{n\}) = \mathcal{P}(m \cup \{m\})$$

$$cu + dv = \vec{0} \implies c = d = 0$$

$$\text{Show: } a_1(u+v) + a_2(u-v) = \vec{0} \implies a_1 = a_2 = 0$$

$$\text{Let } a_2 = \frac{c-d}{1+1}, a_1 = (-a_2 = d+a_2)$$

$$(a_1+a_2)u + (a_1-a_2)v = \vec{0}$$

$$c = d + (1+1)a_2$$

$$a_1 + a_2 = 0 \quad a_1 - a_2 = 0$$

$$(1+1)a_1 = 0$$

$$a_1 = a_2$$

$$c = d + c - d = c$$

$$\begin{aligned} 1+1 &= 0 \text{ OR } a_1 = 0 \\ \text{N.A.} & \quad a_2 = 0 \end{aligned}$$

$$d - b = (1+1)a_3$$

$$a_1 = b - a_2$$

$$(1+1)a_2 = d + b - c$$

$$d = b + (1+1)a_3$$

$$b = a_1 + a_2$$

$$d = (1+1)a_2 + c - b$$

$$a_2 + c - b = a_3$$

$$d = a_2 + a_3$$

$$a_1(u+v) + a_2(u-v) = \vec{0} \implies a_1 = a_2 = 0$$

$$a_2 = c - b - a_3$$

$$a_1 = b - a_2$$

$$\implies (a_1+a_2)u + (a_1-a_2)v = \vec{0} \implies a_1 = a_2 = 0$$

$$b = c - a_2 - a_3$$

$$b = a_1 + a_2$$

$$b_1 u + b_2 v = \vec{0}$$

$$(-a_2 - a_3) = a_1 + a_2$$

$$c = a_1 + (1+1)a_2 + a_3$$

$$b_1(\vec{u} + \vec{v}) + b_2(\vec{u} + \vec{w}) + b_3(\vec{v} + \vec{w}) = \vec{0} \implies a_1 u + a_2 v + a_3 w = \vec{0} \implies a_1 = a_2 = a_3 = 0$$

$$(b_1 + b_2)u + (b_1 + b_3)v + (b_2 + b_3)w = \vec{0}$$

$$b_1 + b_2 = 0$$

$$b_1 + b_3 = 0$$

$$b_2 + b_3 = 0$$

$$a_1 = b_1 + b_2 \quad a_2 = b_1 + b_3 \quad a_3 = b_2 + b_3$$

$$b_1 = a_1 - b_2$$

$$a_2 = a_1 - b_2 + b_3$$

$$a_3 = a_1 + (1+1)b_3$$

$$b_2 + b_3 = a_2 - a_1$$

$$b_3 = \frac{a_3 - a_1}{1+1}$$

$$b_2 = a_2 - a_1 - \frac{a_3 - a_1}{1+1}$$

$$b = a_1 + a_2$$

$$c = a_1 + a_3$$

$$d = a_2 + a_3$$

$$a_1 = b - a_2$$

$$c = b - a_2 + a_3$$

$$d = (1+1)a_2 + c - b$$

$$a_3 = c - b + a_2$$

$$a_2 = \frac{d + b - c}{1+1}$$

$$a_1 - b_2$$

$$\frac{a_1 - a_1}{1+1}$$

$$\begin{bmatrix} a & b \\ c & d \\ e & f \end{bmatrix}$$

$$\begin{bmatrix} 0 & b_1 \\ c_1 & d_1 \\ e_1 & f_1 \end{bmatrix},$$

$$\begin{bmatrix} a_2 & 0 \\ c_2 & d_2 \\ e_2 & f_2 \end{bmatrix},$$

$$\begin{bmatrix} a_3 & b_3 \\ 0 & d_3 \\ e_3 & f_3 \end{bmatrix},$$

...

$$\begin{bmatrix} a_6 & b_6 \\ c_6 & d_6 \\ e_6 & 0 \end{bmatrix}$$

T

$$\bullet \begin{bmatrix} 0 & c_1 & e_1 \\ b_1 & d_1 & f_1 \end{bmatrix},$$

$$\begin{bmatrix} a_2 & c_2 & e_2 \\ 0 & d_2 & f_2 \end{bmatrix},$$

$$\begin{bmatrix} a_3 & 0 & e_3 \\ b_3 & d_3 & f_3 \end{bmatrix},$$

...

$$\begin{bmatrix} a_6 & c_6 & e_6 \\ b_6 & d_6 & 0 \end{bmatrix}$$

$$a e^{rt} + b e^{st} = 0$$

$$a e^{rt} = -b e^{st}$$

$$a = -b e^{(s-r)t}$$

$$e^{\ln b}$$

$$a e^{rt} + b e^{st} = c e^{Rt} + d e^{St}$$

Let  $V$  be a vector space over the field  $F$  and  $u_1, u_2, \dots, u_n$  be distinct vectors in  $V$ .

$\Rightarrow$  Assume  $\beta = \{u_1, u_2, \dots, u_k\}$  is a basis for  $V$ .

$\Rightarrow \text{Span } \beta = V$

$\Rightarrow \beta$  is linearly independent

If there are some natural  $n \leq k$  and  $m \leq k$

with  $a_1, a_2, \dots, a_n \in F$  <sup>not zero</sup> and  $\bar{a}_1, \bar{a}_2, \dots, \bar{a}_m \in F$  <sup>not zero</sup> such that  $\sum_{i=1}^n a_i v_i = \sum_{i=1}^m \bar{a}_i \bar{v}_i = v$ , then

$$\sum_{i=1}^n a_i v_i - \sum_{i=1}^m \bar{a}_i \bar{v}_i = \vec{0}$$

each  $v_i$  is distinct & each  $\bar{v}_i$  is distinct

Show  $a_i = \bar{a}_j$  &  $v_i = \bar{v}_j$  &  $i=j$

Know  $\vec{0} \notin \beta$ , left  $\beta$  not linearly independent.

$\Rightarrow$  Either  $n=m$  or  $n \neq m$ . If  $n \neq m$ , suppose  $n > m$  wlog.

Since ... distinct ..., at most  $m$  pairs of vectors  $v_i$  and  $\bar{v}_i$  with  $v_i = \bar{v}_i$ .

$$\sum_{i \in A} a_i u_i = \sum_{i \in B} b_i u_i = v$$

Conversely, suppose each  $v \in V$  can be uniquely expressed as a linear combination of vectors of  $\beta$ .

$$v = \sum_{i=1}^k a_i u_i$$

$$\sum_{i=1}^k a_i u_i = \vec{0}$$

$\Rightarrow$  the scalars  $a_1, a_2, \dots, a_k$  are unique

One possible combination of scalars is  $a_1 = a_2 = \dots = a_k = 0$ .

Since unique  $\Rightarrow$  the above is the only possible combination

$\Rightarrow$  All reps are trivial.  
 $\Rightarrow \beta$  is linearly independent

$$\sum_{i \in A \setminus B} (a_i - b_i) u_i + \sum_{i \in A \cap B} a_i u_i + \sum_{i \in B \setminus A} (-b_i) u_i = \vec{0}$$

If  $A \neq B$ , then  $A \setminus B$  and/or  $B \setminus A$  is nonempty.

Let  $k$  in the above nonempty set, then  $a_k \neq 0$

$\Rightarrow$  so there is a nontrivial representation of  $\vec{0}$  contradicting  $\beta$  lin independent.

$\Rightarrow A=B$  must be true.

$$\sum_{i \in A \setminus B} (a_i - b_i) u_i = \vec{0}$$

Since  $u_i \neq \vec{0}$  for all  $i$ ,  $a_i - b_i = 0$  and the above rep must be trivial  $a_i = b_i$ .

Assume  $S_1$  and  $S_2$  are disjoint linearly independent subsets of  $V$  so that  $S_1 \cup S_2$  is linearly dependent. Then there exists some natural  $k$  with  $u_1, u_2, \dots, u_k \in S_1 \cup S_2$  and scalars  $a_1, a_2, \dots, a_k \in F$  such that

$$a_1 u_1 + a_2 u_2 + \dots + a_k u_k = \vec{0}.$$

In other words, since each  $u_i \in S_1 \cup S_2$  is either in  $S_1$  or  $S_2$ , this means that  $u_1, u_2, \dots, u_m \in S_1$ ;  $u_{m+1}, u_{m+2}, \dots, u_n \in S_2$  and  $a_1, a_2, \dots, a_m, \dots, a_n$

$$a_1 u_1 + a_2 u_2 + \dots + a_m u_m + a_{m+1} u_{m+1} + \dots + a_n u_n = \vec{0}$$

for some natural  $m$  and  $n$ . Equivalently, we know that

$$a_1 u_1 + a_2 u_2 + \dots + a_m u_m = (-a_{m+1}) u_{m+1} + (-a_{m+2}) u_{m+2} + \dots + (-a_n) u_n.$$

Thus, as  $\sum_{i=1}^m a_i u_i \in \text{Span}(S_1)$  and  $\sum_{i=m+1}^n (-a_i) u_i \in \text{Span}(S_2)$ , their intersection  $\text{Span}(S_1) \cap \text{Span}(S_2)$  must contain this vector as well. Hence,  $\text{Span}(S_1) \cap \text{Span}(S_2) \neq \{\vec{0}\}$  because  $\sum_{i=1}^m a_i u_i = \sum_{i=m+1}^n (-a_i) u_i \neq \vec{0}$  by virtue of  $S_1$  and  $S_2$  being linearly independent.

(Conversely, suppose  $\text{Span}(S_1) \cap \text{Span}(S_2) \neq \{\vec{0}\}$  where  $S_1$  and  $S_2$  are still disjoint linearly independent subsets of the vector space  $V$  over the field  $F$ . Consequently, we see that there exists the natural  $m$  and  $n$  in the following way:

$$\sum_{i=1}^m a_i u_i = \sum_{i=1}^n b_i v_i \quad \text{where } a_1, a_2, \dots, a_m, b_1, b_2, \dots, b_n \in F, u_1, u_2, \dots, u_m \in S_1, \text{ and } v_1, v_2, \dots, v_n \in S_2.$$

Thus we simply reverse what we previously did:

$$\sum_{i=1}^m a_i u_i - \sum_{i=1}^n b_i v_i = \vec{0}$$

By the simple reindexing of letting  $-b_i = a_{m+i}$  and  $v_i = u_{m+i}$ , we get that

$$\sum_{i=1}^{m+n} a_i u_i = \vec{0}.$$

Therefore, there indeed exists a nontrivial representation of  $\vec{0}$  as a linear combination of vectors in  $S_1 \cup S_2$ . Hence,  $S_1 \cup S_2$  is

linearly dependent.  $S_1 \cup S_2$  is linearly independent if and only if  $\text{Span}(S_1) \cap \text{Span}(S_2) = \{\vec{0}\}$ .  $\square$

$\text{Span}(S) = W$

~~$\forall S (S \subset J \Rightarrow \text{Span}(S) \neq W)$~~

$\Rightarrow J$  linearly independent

$J \cap W \Rightarrow \text{Span}(S) \neq W$   
 ~~$\exists S (S \subset J \text{ and } \text{Span}(S) = W)$~~

$J \subseteq W$   
 (lid)  
 $J \cup \{v\}$  is lid iff  $v \notin \text{Span}(S)$   
 $\Rightarrow \text{Span}(J \cup \{v\}) =$

Assume  $J \subset W$   
 Assume  $\text{Span}(J) = W$  and for all  $S \subset J$ ,  $\text{Span}(S) \neq W$ ;  
 $w \in W \Rightarrow w = \sum_{i=1}^n a_i u_i$   
 $\bar{w} \neq \sum_{i=1}^n a_i u_i$

Show for all  $a_1, a_2, \dots, a_n \in F$  and  $u_1, u_2, \dots, u_n \in J$  and  $k \in \mathbb{N}$ ;

$\sum_{i=1}^k a_i u_i \neq \vec{0}$

Suppose otherwise:

exists  $m$   
 with  $\sum_{j=1}^m b_j u_j = \vec{0}$

$0x^2 + 2(x) - (Lx) = 0$

~~$S \subset J$~~



$$\left. \begin{array}{l} \text{Span}(G) = V \quad G \subseteq V \quad m \text{ vectors} \\ L \text{ linearly independent} \quad L \subseteq V \quad n \text{ vectors} \\ \Rightarrow \sum_{i=1}^n a_i u_i = \vec{0} \\ \quad \rightarrow a_i = 0 \text{ for all } i=1, \dots, n \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} m \leq n \quad \& \quad \text{exists } H \text{ with } \text{Span}(L \cup H) = V \\ H \subseteq G \quad n-m \text{ vectors} \end{array} \right.$$

Some subset  $\beta$  of  $G$  containing  $k \leq m$  vectors is a basis of  $V$ .

Since  $\text{span}(G) = V$  and  $L \subseteq V$ ,  $L \subseteq \text{span}(G) \Rightarrow$  Every vector in  $L$  can be written as a linear combination of vectors in  $G$ .

If  $n > m$ , then for any two vectors of  $L$ ,  $l_1 \neq l_2$ ,

$$l_1 = \sum_{i=1}^m a_i u_i$$

$$l_2 = \sum_{i=1}^m b_i u_i$$

where  $a_k \neq b_k$  for some natural  $k \leq m$ .

$$l_1 - l_2 = \sum_{i=1}^m (a_i - b_i) u_i = \vec{0}$$

Since  $a_k \neq b_k$  the coefficient of  $u_k$ , i.e.  $a_k - b_k$ , must also be nonzero.  $\rightarrow$  nontrivial rep of  $\vec{0}$  as vectors in  $L$ .

This contradicts our assumption that  $L$  is linearly independent.

If  $m = n$ ,  $\emptyset \subseteq G$ .

By Theorem 1.5, since  $L \subseteq V$ ,  $\text{span}(L) \subseteq V$ .

$$\sum_{i=1}^k c_i v_i = \vec{0} \Rightarrow c_i = 0 \text{ for all } i \leq k$$

$$\vec{v} = \sum_{i=1}^k c_i v_i$$

$$l_j = \sum_{i=1}^k a_{ij} v_i \quad \text{where } v_i \in \beta$$

$$\sum_{j=1}^n \left( b_j \cdot \sum_{i=1}^k a_{ij} v_i \right)$$

Let  $V$  be a vector space over the field  $F$  and  $u_1, u_2, \dots, u_n$  be distinct vectors in  $V$ .

First assume that  $B = \{u_1, u_2, \dots, u_n\}$  is a basis of  $V$ . By definition, every  $v \in V$  can be expressed as a linear combination of vectors of  $B$  because  $B$  generates  $V$ . The trickier part is to prove the uniqueness of such a linear combination. When there are two linear combinations that are identical to some  $v \in V$ , this means that there are some subsets  $A$  and  $B$  of  $\mathbb{N}$  containing natural numbers less than  $n$  so that

$$\sum_{i \in A} a_i u_i = \sum_{i \in B} b_i u_i = v;$$

$$\sum_{i \in A \setminus B} (a_i - b_i) u_i + \sum_{i \in A \cap B} a_i u_i + \sum_{i \in B \setminus A} (-b_i) u_i = \vec{0}. \quad (1)$$

Either  $A=B$  or  $A \neq B$  must hold. Consider  $A \neq B$ : then  $A \setminus B$  ~~and~~ <sup>or</sup>  $B \setminus A$  is nonempty, i.e. there is some natural  $k$  in ~~one~~ <sup>precisely</sup> one of the aforementioned sets, with  $a_k \neq 0$  or  $b_k \neq 0$ . Notice that coefficient of  $u_k$  must thus be nonzero. In other words, there would be a non-trivial representation of  $\vec{0}$  as a linear combination of vectors in  $B$ , this would contradict our assumption that  $B$  is a basis of  $V$  — that is,  $B$  is linearly independent. Hence, it must be that  $A=B$ . Consequently, we can state equation (1) now as

$$\sum_{i \in A \setminus B} (a_i - b_i) u_i = \vec{0}.$$

Again, by virtue of the fact that the basis  $B$  is linearly independent, this must be a trivial representation of  $\vec{0}$  (as vectors in  $B$ ). So,  $a_i = b_i$ . Which means that  $\sum_{i \in A} a_i u_i$  is the exact same representation of the vector  $v \in V$  as  $\sum_{i \in B} b_i u_i$ .

Conversely, now suppose that each  $v \in V$  can be uniquely expressed as a linear combination of vectors of  $B$ . Therefore, the zero vector can also be uniquely written as  $\sum_{i \in A} a_i u_i$  for some natural  $k$ . Since  $a_1 = a_2 = \dots = a_k = 0$  is clearly one such possible combination of scalars and uniqueness is presumed, this must be the only possible combination (of coefficients), which is trivial. Thereupon,  $B$  is a basis of  $V$ .

Therefore,  $B = \{u_1, u_2, \dots, u_n\}$  is a basis for  $V$  if and only if each  $v \in V$  can be uniquely expressed as a linear combination of vectors of  $B$ .

Q.E.D.  $\square$

1-(a) False. It is its own basis. ✓

(b) True. In fact, a basis must have cardinality less than or equal to that generating set. ✓

(c) False. Counterexample:  $P(\mathbb{R})$ . ✓

(d) False.  $\{\hat{i}, \hat{j}\}$  and  $\{2\hat{i}, 2\hat{j}\}$  are bases of  $\mathbb{R}^2$ . ✓

(e) True. See Corollary 1 of the Replacement Theorem. ✓

(f) False. Since  $\{1, x, x^2, \dots, x^n\}$  is a basis of  $P_n(F)$ , so  $\dim P_n(F) = n+1$ . ✓

(g) False.  $\dim M_{m \times n}(F) = m \cdot n$  because  $\{A^{ij} \in M_{m \times n}(F) \mid 0 \leq i \leq m \text{ and } 0 \leq j \leq n\}$  <sup>the set</sup> where  $A^{ij}$  is the matrix with  $A^{ij}_{ij} = 1$  and  $A^{ij}_{kl} = 0$  otherwise. <sup>is a basis,</sup> ✓

(h) True. As  $S_2$  generates  $V$ ,  $|S_2| \geq \dim(V) \geq |S_1|$ . ✓

(i) False. That is true iff  $S$  is a basis of  $V$ . Counterexample: Suppose  $V = \mathbb{R}^2$  and  $S = \{\hat{i}, 2\hat{i}, \hat{k}\}$ , then  $2(\hat{i}) + \hat{k} = (2\hat{i}) + \hat{k}$ . ✓

(j) True. Follows from Theorem 1.11. ✓

(k) True. A subspace  $S$  of dimension 0 must be spanned by  $\emptyset$  so  $S = \{0\}$ . Similarly, a subspace  $S$  of dimension  $n = \dim V$  must be  $V$  itself. ✓

(l) True. See Corollary 2 of the Replacement Theorem. ✓

$\rightarrow \subseteq V$   $\rightarrow \subseteq V$   $\rightarrow$  vectors.  $\} \Rightarrow \} m \leq n$

$W$  being finite dimensional is an immediate because otherwise there exists a linearly independent subset of  $V$  with more than  $\dim(V)$  vectors, a contradiction. For the same reason, we can conclude  $\dim(W) \leq \dim(V)$ . Let  $\beta$  be a basis of  $W$ , then  $|\beta| = \dim(W) \leq \dim(V)$ . Hence,  $\beta$  is also a basis of  $V$ .

As such,  $V=W$  follows. □

Example 18 (check)

$$p(-1, 0, 1, 0, 0) + q(-1, 0, 0, 0, 1) + r(0, 1, 0, 1, 0) = (a, b, c, d, e)$$

$$(-p-q, r, p, r, q) = (a, b, c, d, e)$$

$$\Rightarrow \begin{cases} a = -p-q \\ b = r \\ c = p \\ d = r \\ e = q \end{cases} \quad \begin{cases} p = -a-q \\ p = -a-e \\ r = b = d \\ q = e \end{cases}$$

From here, we notice that if  $(a, b, c, d, e) = \mathbf{0}$ , then  $p = q = r = 0$ . Hence, this set is indeed linearly independent.

Now fix  $p := -(a+e)$ ,  $q := e$  and  $r := b$  so  $(a, b, c, d, e)$  is an arbitrary vector in  $W$ . Which means  $b=d$  and  $c=-a-e$  by the construction of  $W$ . Therefore,  $\{(-1, 0, 1, 0, 0), (-1, 0, 0, 0, 1), (0, 1, 0, 1, 0)\}$  generates  $W$ :

$$\begin{aligned} (-p)(-1, 0, 1, 0, 0) + q(-1, 0, 0, 0, 1) + r(0, 1, 0, 1, 0) &= (-p-q, r, p, r, q) \\ &= (a+e-e, b, -(a+e), b, e) \\ &= (a, b, c, d, e) \end{aligned}$$

Indeed, we have now shown that it is a basis of  $W$  <sup>with cardinality 3</sup>. Consequently,  $\dim(W) = 3$ .

2. (c)

$$a(1, 2, -1) + b(1, 0, 2) + c(2, 1, 1) = (x, y, z)$$

$$(a+b+2c, 2a+c, -a+2b+c) = (x, y, z)$$

$$\Rightarrow \begin{cases} a+b+2c = x & \text{--- (1)} \\ 2a+c = y & \text{--- (2)} \\ -a+2b+c = z & \text{--- (3)} \end{cases}$$

(3)+(1):

$$\begin{aligned} 3b+3c &= x+z \\ b+c &= \frac{1}{3}x + \frac{1}{3}z & \text{--- (3')} \end{aligned}$$

(1)-(3'):

$$a+c = \frac{2}{3}x - \frac{1}{3}z & \text{--- (1')}$$

$$\Rightarrow \begin{cases} a+c = \frac{2}{3}x - \frac{1}{3}z & \text{--- (1')} \\ 2a+c = y & \text{--- (2)} \\ b+c = \frac{1}{3}x + \frac{1}{3}z & \text{--- (3')} \end{cases}$$

$$\Rightarrow \begin{cases} a+c = \frac{2}{3}x - \frac{1}{3}z & \text{--- (1')} \\ a+b+c = \frac{1}{3}x + y + \frac{1}{3}z & \text{--- (2')} \\ b+c = \frac{1}{3}x + \frac{1}{3}z & \text{--- (3')} \end{cases}$$

$$\Rightarrow \begin{cases} c = \frac{4}{3}x - y - \frac{2}{3}z \\ a = -\frac{1}{3}x + y + \frac{1}{3}z \\ b = -x + y + z \end{cases}$$

Once more, this suffices to show that it is indeed a basis.

2. (a)

$$a(1, 0, -1) + b(2, 5, 1) + c(0, -4, 3) = (x, y, z)$$

$$(a+2b, 5b-4c, -a+b+3c) = (x, y, z)$$

$$a+2b = x$$

$$5b-4c = y$$

$$-a+b+3c = z$$

$$a+2b = x$$

$$b - \frac{4}{5}c = \frac{1}{5}y$$

$$3b+3c = x+z$$

$$a+2b = x$$

$$b - \frac{4}{5}c = \frac{1}{5}y$$

$$\frac{27}{5}c = x - \frac{1}{5}y + z$$

$$\begin{aligned} a &= \frac{14}{15}x - \frac{1}{5}y - \frac{1}{15}z \\ b &= \frac{4}{15}x + \frac{1}{5}y + \frac{2}{15}z \\ c &= \frac{5}{27}x - \frac{1}{45}y + \frac{1}{27}z \end{aligned}$$

Therefore, since  $\text{span}\{(1, 0, -1), (2, 5, 1), (0, -4, 3)\} = \mathbb{R}^3$ , and this set contains 3 =  $\dim \mathbb{R}^3$  vectors, by Corollary 2(a) of the Replacement Theorem, it is a basis for  $\mathbb{R}^3$ .

(b)  $a(2, -4, 1) + b(0, 3, -1) + c(6, 0, -1) = (x, y, z)$

$$(2a+6c, -4a+3b, -b-c) = (x, y, z)$$

$$\Rightarrow \begin{cases} 2a + 6c = x \\ -4a + 3b = y \\ -b - c = z \end{cases} \Rightarrow \begin{cases} a + 3c = \frac{1}{2}x \\ b + 4c = \frac{2}{3}x + \frac{1}{3}y \\ c = \frac{2}{9}x + \frac{1}{9}y + \frac{1}{3}z \end{cases} \Rightarrow \begin{cases} a = -\frac{1}{9}x - \frac{1}{3}y - z \\ b = -\frac{2}{9}x - \frac{1}{9}y - \frac{4}{3}z \\ c = \frac{2}{9}x + \frac{1}{9}y + \frac{1}{3}z \end{cases}$$

Again, this proves that  $\{(2, -4, 1), (0, 3, -1), (6, 0, -1)\}$  is a basis of  $\mathbb{R}^3$ .

2.(e)

$$a(1, -3, -2) + b(-3, 1, 3) + c(-2, -10, -2) = (x, y, z)$$

$$(a - 3b - 2c, -3a + b - 10c, -2a + 3b - 2c) = (x, y, z)$$

$$\Rightarrow \begin{cases} a - 3b - 2c = x & (1) \\ -3a + b - 10c = y & (2) \\ -2a + 3b - 2c = z & (3) \end{cases}$$

(1) + 3(2):

$$-5a - 22c = x + 2y$$

$$a + \frac{22}{5}c = -\frac{1}{5}x - \frac{2}{5}y \quad (1')$$

(2) + 3(1):

$$b - 10c + \frac{60}{5}c = y - \frac{3}{5}x - \frac{6}{5}y$$

$$b + \frac{16}{5}c = -\frac{3}{5}x - \frac{1}{5}y \quad (2')$$

(3') - 3(2')

$$-\frac{14}{5}c = \frac{7}{5}x - \frac{1}{5}y + z$$

$$c = -\frac{1}{2}x + \frac{1}{14}y - \frac{5}{14}z \quad (3'')$$

(3) + 2(1')

$$3b - 2c + \frac{44}{5}c = -\frac{2}{5}x - \frac{4}{5}y + z \quad (3')$$

$$\underbrace{3b - 2c + \frac{44}{5}c}_{3b + \frac{34}{5}c} = -\frac{2}{5}x - \frac{4}{5}y + z \quad (3')$$

$$\Rightarrow \begin{cases} a + \frac{22}{5}c = -\frac{1}{5}x - \frac{2}{5}y & (1'') \\ b + \frac{16}{5}c = -\frac{3}{5}x - \frac{1}{5}y & (2'') \\ c = -\frac{1}{2}x + \frac{1}{14}y - \frac{5}{14}z & (3'') \end{cases}$$

It is clear that (1) -  $\frac{22}{5}$ (3'') and (2') -  $\frac{16}{5}$ (3'') gives us our desired equations for a and b in terms of x, y, z.

Hence, we see that this is also a basis of  $\mathbb{R}^3$ .

X

$$a = -2x - \frac{5}{7}y + \frac{11}{7}z$$

$$b = -\frac{11}{5}x - \frac{3}{7}y + \frac{8}{7}z$$

If  $(x, y, z) = (1, 0, 0)$ ,  $a = -2$ ,  $b = -\frac{11}{5}$ ,  $c = -\frac{1}{2}$

plugging into (1):  $\frac{21}{5} = 1$

(2):  $\frac{61}{5} = 0$

(3):  $-\frac{8}{5} = 0$

2.(d)

$$a(-1, 3, 1) + b(2, -4, -3) + c(-3, 8, 2) = (x, y, z)$$

$$(-a + 2b - 3c, 3a - 4b + 8c, a - 3b + 2c) = (x, y, z)$$

$$\Rightarrow \begin{cases} -a + 2b - 3c = x & \text{--- (1)} \\ 3a - 4b + 8c = y & \text{--- (2)} \\ a - 3b + 2c = z & \text{--- (3)} \end{cases}$$

(2) + 2(1):

$$a + 2c = 2x + y \quad \text{--- (2')}$$

(3) + (1):

$$-b - c = x + z$$

$$b + c = -x - z \quad \text{--- (3')}$$

2(3') - (1):

$$a + 5c = x + 2z \quad \text{--- (1')}$$

(1') - (2'):

$$a - 3c = -x - y + 2z$$

$$c = -\frac{1}{3}x - \frac{1}{3}y + \frac{1}{3}z \quad \text{--- (1'')$$

$$\Rightarrow \begin{cases} c = -\frac{1}{3}x - \frac{1}{3}y + \frac{1}{3}z & \text{--- (1'')} \\ a + 2c = 2x + y & \text{--- (2')} \\ b + c = -x - z & \text{--- (3')} \end{cases}$$

(2') - 2(1'') and (3') - (1''):

$$\Rightarrow \begin{cases} c = -\frac{1}{3}x - \frac{1}{3}y + \frac{1}{3}z \\ a = \frac{5}{3}x + \frac{5}{3}y - \frac{4}{3}z \\ b = -\frac{2}{3}x + \frac{1}{3}y - \frac{5}{3}z \end{cases}$$

Thus, this is yet again a basis of  $\mathbb{R}^3$ .



10. (b) Let  $c_0 = -4$ ,  $c_1 = 1$ ,  $c_2 = 3$

$$f_0(x) = \frac{(x-1)(x-3)}{(-4-1)(-4-3)} = \frac{1}{35}(x^2 - 4x + 3), \quad g(c_0) = 24$$

$$f_1(x) = \frac{(x+4)(x-3)}{(1+4)(1-3)} = -\frac{1}{10}(x^2 + x - 12), \quad g(c_1) = 9$$

$$f_2(x) = \frac{(x+4)(x-1)}{(3+4)(3-1)} = \frac{1}{14}(x^2 + 3x - 4), \quad g(c_2) = 3$$

The required polynomial  $g: \mathbb{R} \rightarrow \mathbb{R}$  is given by

$$g(x) = \sum_{i=0}^2 g(c_i) f_i(x)$$

$$= \frac{24}{35}(x^2 - 4x + 3) - \frac{9}{10}(x^2 + x - 12) + \frac{3}{14}(x^2 + 3x - 4)$$

$$= -3x + 12$$

check:  $g(c_0) = 24$

$g(c_1) = 9$

$g(c_2) = 3$  ✓

4. NO, <sup>by corollary 2 of the Replacement Theorem</sup> they do not generate  $P_3(\mathbb{R})$  because  $\dim(P_3(\mathbb{R})) = 3+1 = 4$  while there are only 3 vectors provided.

5. Since  $\dim \mathbb{R}^3 = 3$ , no linearly independent subset of  $\mathbb{R}^3$  can have more than 3 vectors. <sup>by the replacement theorem</sup> Thus, the provided set containing 4 vectors must not be linearly independent.

7. 
$$\frac{22}{57} u_1 + \frac{1}{3} u_2 + \frac{2}{57} u_5 = (1, 0, 0)$$
$$-\frac{2}{57} u_1 + \frac{1}{3} u_2 + \frac{5}{57} u_5 = (0, 1, 0)$$
$$\frac{7}{57} u_1 + \frac{1}{3} u_2 + \frac{11}{57} u_5 = (0, 0, 1)$$
 Since the standard basis is contained in  $\text{span}\{u_1, u_2, u_5\}$  <sup>of  $\dim \mathbb{R}^3 = 3$  vectors</sup>  $\text{span}\{u_1, u_2, u_5\} = \mathbb{R}^3$  so that <sup>the set</sup>  $\{u_1, u_2, u_5\}$  forms a basis for  $\mathbb{R}^3$ .

10. (a) let  $c_0 = -2$ ,  $c_1 = -1$ , and  $c_2 = 1$ :  
$$f_0(x) = \frac{(x+1)(x-1)}{(-2+1)(-2-1)} = \frac{1}{3}(x^2-1), \quad g(c_0) = -6$$
$$f_1(x) = \frac{(x+2)(x-1)}{(-1+2)(-1-1)} = -\frac{1}{2}(x^2+x-2), \quad g(c_1) = 5$$
$$f_2(x) = \frac{(x+2)(x+1)}{(1+2)(1+1)} = \frac{1}{6}(x^2+3x+2), \quad g(c_2) = 3$$

The required polynomial  $g: \mathbb{R} \rightarrow \mathbb{R}$  is thus given by

$$g(x) = \sum_{i=0}^2 g(c_i) f_i(x)$$
$$= -2(x^2-1) - \frac{5}{2}(x^2+x-2) + \frac{1}{6}(x^2+3x+2)$$
$$= -4x^2 - x + 8$$

check:  $g(-2) = -6$   
 $g(-1) = 5$   
 $g(1) = 3$  ✓

10. (d) Let  $c_0 = -3$ ,  $c_1 = -2$ ,  $c_2 = 0$ ,  $c_3 = 1$ :

$$f_0(x) = \frac{(x+2)(x-0)(x-1)}{(-3+2)(-3-0)(-3-1)} = -\frac{1}{12}(x^3+x^2-2x), \quad g(c_0) = g(-3) = -30$$

$$f_1(x) = \frac{(x^2+x-2)x}{(-2+3)(-2)(-2-1)} = \frac{1}{6}(x^3+2x^2-3x), \quad g(c_1) = g(-2) = 7$$

$$f_2(x) = \frac{(x+3)(x+2)(x-1)}{(3)(2)(-1)} = -\frac{1}{6}(x^3+4x^2+x-6), \quad g(c_2) = g(0) = 15$$

$$f_3(x) = \frac{(x^2+5x+6)(x-1)}{(1+3)(1+2)(1)} = \frac{1}{12}(x^3+5x^2+6x), \quad g(c_3) = g(1) = 10$$

Thus, the required function  $g: \mathbb{R} \rightarrow \mathbb{R}$  is given by

$$g(x) = \sum_{i=0}^3 g(c_i) f_i(x)$$

$$= \frac{5}{2}(x^3+x^2-2x) + \frac{7}{6}(x^3+2x^2-3x) - \frac{5}{2}(x^3+4x^2+x-6) + \frac{5}{6}(x^3+5x^2+6x)$$

$$= 2x^3 - x^2 - 6x + 15$$

check:  $g(c_0) = -30$

$$g(c_1) = 7$$

$$g(c_2) = 15$$

$$g(c_3) = 10$$

$$10. (c) \text{ let } c_0 = -2, c_1 = -1, c_2 = 1, c_3 = 3$$

$$f_0(x) = \frac{(x+1)(x-1)(x-3)}{(-2+1)(-2-1)(-2-3)} = -\frac{1}{5}(x^3 - 3x^2 - x + 3), \quad g(c_0) = 3$$

$$(x^2-1)(x-3) = x^3 - 3x^2 - x + 3$$

$$f_1(x) = \frac{(x+2)(x-1)(x-3)}{(-1+2)(-1-1)(-1-3)} = \frac{1}{8}(x^3 - 2x^2 - 5x + 6), \quad g(c_1) = -6$$

$$(x^2+x-2)(x-3) = x^3 - 3x^2 + x^2 - 3x - 2x + 6 = x^3 - 2x^2 - 5x + 6$$

$$f_2(x) = \frac{(x+2)(x+1)(x-3)}{(1+2)(1+1)(1-3)} = -\frac{1}{12}(x^3 - 7x - 6), \quad g(c_2) = 0$$

$$(x^2+3x+2)(x-3) = x^3 - 3x^2 + 3x^2 - 9x + 2x - 6 = x^3 - 7x - 6$$

$$f_3(x) = \frac{(x+2)(x+1)(x-1)}{(3+2)(3+1)(3-1)} = \frac{1}{40}(x^3 + 2x^2 - x - 2), \quad g(c_3) = -2$$

$$(x+2)(x^2-1) = x^3 - x + 2x^2 - 2$$

Hence, the required function  $g: \mathbb{R} \rightarrow \mathbb{R}$  is given by

$$g(x) = \sum_{i=0}^3 g(c_i) f_i(x)$$

$$= -\frac{1}{5}(x^3 - 3x^2 - x + 3) - \frac{3}{4}(x^3 - 2x^2 - 5x + 6) - \frac{1}{12}(x^3 + 2x^2 - x - 2)$$

$$= -x^3 + 2x^2 + 4x - 5$$

$$\text{check: } g(c_0) = 3$$

$$g(c_1) = -6$$

$$g(c_2) = 0$$

$$g(c_3) = -2 \quad \checkmark$$

15 - Ideas

$$\{A_{ij} \mid i \neq j\} \cup \{B_{ij} \mid j = i+1\} \cup \{B^{n-1}\}$$

$$\sum_{i=1}^{n-1} c_i B^{i+1} + c_n B^{n-1} = M, \quad \sum_{i=1}^n M_{ii} = 0$$

$$\Rightarrow \begin{cases} c_1 - c_n = M_{11} \\ -c_{k-1} + c_k = M_{kk}, \quad k \geq 2 \end{cases}$$

$$[q]_{km} = \sum_{i=1}^n M_{ii} + \sum_{j=1}^k M_{jj} = c_k \quad \text{if } k \neq n$$

$$\sum_{i=1}^n M_{ii} = c_n$$

$$c_1 = M_{11}$$

$$\begin{cases} M_{11} & \text{if } k=1 \\ M_{kk} & \\ M_{nn} & \\ c_1 & \text{if } k=1 \\ -c_{k-1} + c_k & \text{if } k < n \\ c_n & \text{if } k=n \end{cases}$$

$$\sum_{i=1}^n M_{ii} + M_{11} - \sum_{i=1}^n M_{ii} = M_{11} \quad \checkmark$$

$$-\sum_{i=1}^n M_{ii} - \sum_{j=1}^{k-1} M_{jj} + \sum_{i=1}^n M_{ii} + \sum_{j=1}^k M_{jj} = M_{kk}$$

$$\begin{aligned} c_1 &= a \\ -c_1 + c_2 &= b & c_1 &= a \\ -c_2 + c_3 &= c & c_2 &= a+b \\ -c_3 &= d & c_3 &= a+b+c \\ & & c_4 &= -d \end{aligned}$$

$$\begin{pmatrix} a & b & c & d \end{pmatrix} = c_1 \begin{pmatrix} 1 & -1 & 0 & 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 & 1 & -1 & 0 \end{pmatrix} + c_3 \begin{pmatrix} 0 & 0 & 1 & -1 \end{pmatrix} + c_4 \begin{pmatrix} 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\begin{aligned} c_1 - c_4 &= a \\ -c_1 + c_2 &= b \\ -c_2 + c_3 &= c \\ -c_3 + c_4 &= d \end{aligned}$$

$$\begin{aligned} c_1 - c_4 &= a \\ c_2 - c_4 &= a+b \\ c_3 - c_4 &= a+b+c \\ c_4 &= a+b+c+d \end{aligned}$$

$$\begin{aligned} c_1 &= 2a + b + c + d \\ c_2 &= 2a + 2b + c + d \\ c_3 &= 2a + 2b + 2c + d \\ c_4 &= a + b + c + d \end{aligned}$$

Ideas

17.  $M^t = -M$

$M_{ji} = -M_{ij}$

let  $S^{ij}$  be the matrix having  $\mathbb{1}$  and  $-\mathbb{1}$  as its  $(i,j)$ th and  $(j,i)$ th entry respectively. If  $i=j$  let  $(j,i)$ th also be  $\mathbb{1}$

$\beta := \{S^{ij} \mid i \leq j\}$

Linear independent

$$\sum_{j=1}^n \sum_{i=1}^j G_{ij} S_{k_1, k_2}^{ij} = \begin{cases} M_{k_1, k_2} S^{k_1, k_2} & M_{k_1, k_2} \\ M_{k_1, k_2} S^{k_1, k_2} & M_{k_1, k_2} S^{k_1, k_2} \end{cases}$$

$$13. \begin{cases} x_1 - 2x_2 + x_3 = 0 & \text{--- (1)} \\ 2x_1 - 3x_2 + x_3 = 0 & \text{--- (2)} \end{cases}$$

$$(2) - 2(1): \\ (-3 + 4)x_2 + (1 - 2)x_3 = 0 \\ x_2 - x_3 = 0$$

$$\begin{cases} x_1 - 2x_2 + x_3 = 0 & \text{--- (1')} \\ x_2 - x_3 = 0 & \text{--- (2')} \end{cases}$$

$$\begin{cases} x_1 = x_2 \\ x_2 = x_3 \end{cases}$$

Therefore, this subspace is given by  $\{(x, y, z) \in \mathbb{R}^3 \mid x = y = z\}$ . Accordingly,  $\{(1, 1, 1)\}$  is clearly a basis for it.

15. Let  $A^{ij}$  be the  $n \times n$  matrix with the  $(i, j)$ th entry being  $\mathbb{1}$  and all other entries  $0$ . Also let  $B^i$  be the  $n \times n$  matrix with the  $(i, i)$ th and  $(i+1, i+1)$ th entries be  $\mathbb{1}$  and  $-\mathbb{1}$  respectively,  $0$  everywhere else. Lastly, suppose  $M$  is a  $n \times n$  matrix with trace  $0$  and all non-diagonal entries being  $0$  too. By defining the constants  $c_k := \sum_{i=1}^k M_{ii}$ , we see that for any diagonal matrix  $M$  of trace  $0$ :

$$\sum_{i=1}^{n-1} c_i B_{kk}^i = \begin{cases} c_1 & \text{if } k=1, \\ -c_{k-1} + c_k & \text{if } 1 < k < n, \\ -c_{n-1} & \text{if } k=n. \end{cases}$$

$$= \begin{cases} M_{11} & \text{if } k=1 \\ M_{kk} & \text{if } 1 < k < n, \\ M_{nn} & \text{if } k=n \end{cases}$$

$$= M_{kk}$$

because  $-\sum_{i=1}^{n-1} M_{ii} = M_{nn}$  follows from  $\text{trace}(M) = 0$ .

11. For any representation of  $\underline{0}$  in  $\{\underline{u+v}, a\underline{u}\}$ , that is:

$$c_1(\underline{u+v}) + c_2(a\underline{u}) = \underline{0}$$

$$(c_1 + c_2 a)\underline{u} + c_1\underline{v} = \underline{0}$$

~~we know that  $c_1 + c_2 a = 0$  and  $c_1 = 0$  as  $\{\underline{u}, \underline{v}\}$  is a basis for  $V$ . As  $a, b$  are nonzero,  $c_1$  and  $c_2$  must be zero instead~~

We see that  $\{\underline{u+v}, a\underline{u}\}$  generates  $V$  because for any vector  $d_1\underline{u} + d_2\underline{v}$  in  $V$ ,

$$\begin{aligned} d_2(\underline{u+v}) + \left(\frac{d_1-d_2}{a}\right)(a\underline{u}) &= (d_1-d_2)\underline{u} + d_2\underline{u} + d_2\underline{v} \\ &= d_1\underline{u} + d_2\underline{v}. \end{aligned}$$

As such,  $\{\underline{u+v}, a\underline{u}\}$  indeed forms a valid basis for  $V$  with dimension 2.

Similarly, any vector  $d_1\underline{u} + d_2\underline{v}$  in  $V$  can be written as  $\left(\frac{d_1}{a}\right)(a\underline{u}) + \left(\frac{d_2}{b}\right)(b\underline{v})$ . Hence, forming a basis for  $V$  again.





20. (a) Define the function  $f: \mathbb{N}_0 \rightarrow S$  recursively by  $f(m)$  being any element of  $S$ -span  $f[[m]]$  if it is nonempty, otherwise fix it to be some extraneous object  $e$  not belonging to  $S$ . All  $f(m)$  must be in  $S$  given  $m \leq n$ , lest there exists some least  $m \leq n$  for which  $S$ -span  $f[[m]] = \emptyset$ . This means  $S \subseteq \text{span } f[[m]]$  so  $\text{span } S = \text{span } f[[m]] = V$ , but the generating set  $f[[m]]$  having  $m < n$  members contradicts  $\dim V = n$ . Similarly, we see that  $f[[m]]$  must always be linearly independent (given it is a subset of  $S$ ).

$$\sum_{k=0}^m c_k f(k) = \underline{0}$$

(certainly tells us all constants  $c_k = 0$  because  $\sum_{k=0}^{m-1} d_k f(k) = f(m)$  for no constants  $d_k$ . Hence, it follows that  $f[[n]]$  is a basis of  $V$  since it is a linearly independent set of  $n$  vectors. □

(b) In the above proof, we already constructed a subset of  $S$  containing exactly  $n$  vectors. This suffices to show  $|S| \geq n$ . □

25.  $\dim Z = m+n$  because

$$\sum_{i=1}^m c_i (v_i, \underline{0}) + \sum_{i=1}^n d_i (\underline{0}, w_i) = (\underline{0}, \underline{0})$$

tells us

$$\sum_{i=1}^m c_i v_i = \underline{0} \quad \text{and} \quad \sum_{i=1}^n d_i w_i = \underline{0}$$

So that all  $c_i$  and  $d_i$  must be 0 given  $\{v_i | i \leq m\}$  and  $\{w_i | i \leq n\}$  are bases of  $V$  and  $W$  respectively. It is clear that this set of vectors  $\{(v_i, \underline{0}) | i \leq m\} \cup \{(\underline{0}, w_i) | i \leq n\}$  spans  $Z$ . Hence it is a basis of  $Z$  with cardinality  $m+n$ .

15. We also see that all our  $B^i$  are linearly independent, since

$$\sum_{i=1}^{n-1} c_i B^i = \underline{0}$$

$$\underline{0} = \begin{cases} c_1 & \text{if } k=1, \\ -c_{k-1} + c_k & \text{if } 1 < k < n, \\ -c_{n-1} & \text{if } k=n. \end{cases}$$

Immediately,  $c_1 = c_{n-1} = 0$ . Assuming  $c_{k-1} = 0$ , then  $c_k = 0$  too for all intermediate cases. So,  $c_k = 0$  for any  $1 \leq k \leq n$  as required.

Now, it is clear that  $\{A^{ij} \mid i \neq j\} \cup \{B^i \mid i \leq n-1\}$  is a linearly independent subset of  $W$  that spans  $W$ . In other words, this

is a basis for  $W$ . Accordingly,  $\dim(W) = (n^2 - n) + (n-1) = n^2 - 1$

17. Let  $S^{ij}$  be the  $n \times n$  matrix with its  $(i, i)$ th entry being  $\mathbb{1}$ , and if  $i \neq j$ , its  $(j, j)$  being  $-\mathbb{1}$ . It is straight forward to show that  $\{S^{ij} \mid i \leq j\}$  is a basis for  $W$ . As such,  $\dim(W) = \frac{1}{2}n(n+1)$ . □

18. Suppose  $\sigma_k$  is the sequence in  $F$  with  $\sigma_k(k) = \mathbb{1}$  and  $\sigma_k(n) = 0$  when  $n \neq k$ . Now, it is clear that the set  $\{\sigma_k \mid k \in \mathbb{N}\}$  is a basis for  $W$ , because a linear combination of these vectors only involves some  $n$  number of them by definition. And it follows from our construction of  $\sigma_k$  that any resultant vector will only have  $n$  number of nonzero terms. □

$n = n$  for which  $S$ -span  $f \Pi_m \pi - \alpha$  ... object not belonging to  $S$ . All  $f(m)$  ... being any element of  $S$ -span  $f[\alpha_1]$  if it is nonzero.

29. (a) Ideas

$$\dim(W_1 + W_2) \leq \dim(W_1) + \dim(W_2) \quad \begin{matrix} n_1 \leq m_1 \\ n_2 \leq m_2 \end{matrix}$$

$\beta \quad \beta_1 \quad \beta_2 \quad \alpha$

$$\beta_1 \cap W_2 \subseteq W_1 \cap W_2$$

$$(\beta - W_2) \cup (\beta_1 \cap W_2) \\ \cup (\beta_1 \cap W_2) \cup \alpha$$

$\text{span}(\beta_1 \cap W_2) = W_1 \cap W_2$   
 test  $\text{span}((\beta_1 \cap W_2) \cup \alpha) = W_1 \cap W_2$

but then  $\alpha \in \text{span}(\beta_1)$ . In fact, the coefficient of  $\beta_1 - W_2$  must be 0 for any rep of  $\alpha \in W_1 \cap W_2$  because  $(\beta_1 - W_1) \cap (W_1 \cap W_2) = \emptyset$ . (So if  $\alpha = \sum c_i \beta_i + \sum d_j w_j$ ,  $\alpha \in W_1 \cap W_2$  implies  $\sum c_i \beta_i \in W_1 \cap W_2$  and  $\sum d_j w_j \in W_1 \cap W_2$ .)

Proof (without seeing author's hint)

Suppose  $\beta_1$  is a basis for  $W_1$ . We see that  $\beta_1 \cap W_2$  must be a basis of  $W_1 \cap W_2$ , test  $\text{span}(\beta_1 \cap W_2) \subseteq W_1 \cap W_2$  so there exists some  $\alpha \in W_1 \cap W_2$  that extends  $\beta_1 \cap W_2$  to generate  $W_1 \cap W_2$  (i.e.  $\text{span}((\beta_1 \cap W_2) \cup \alpha) = W_1 \cap W_2$ ). But then  $\alpha \in \text{span}(\beta_1) (= W_1 \cap W_2)$  such that  $\alpha \in \text{span}(\beta_1 - W_2)$ .

Thus,  $\text{span}(\beta_1 \cap W_2) = W_1 \cap W_2$  after all. Consequently,  $\dim(W_1 + W_2) = \dim(W_1) + \dim(W_2) - \dim(W_1 \cap W_2)$  is easy to see.

(b) When  $V$  is the direct sum of  $W_1$  and  $W_2$ ,  $W_1 \cap W_2 = \{0\}$  so from (a) we indeed have that  $\dim(V) = \dim(W_1) + \dim(W_2)$ . The converse holds trivially.

33. (a) By definition,  $W_1 \cap W_2 = \{0\}$  so  $\beta_1 \cap \beta_2 = \emptyset$  is immediate. It is trivial to see that  $\beta_1 \cup \beta_2$  forms a basis for  $V$ .  
 (b) If  $\beta_1 \cup \beta_2$  forms a basis for  $V$ , then  $\dim(V) = \dim(W_1) + \dim(W_2)$ . So the result holds from 29. (a).

35. (a) Consider any member of  $V/W$ , i.e.  $u + W$  for some  $u \in V$ . If  $u \in W$ ,  $\sum_{i=1}^n c_i u_i + W = u + W = W$ . Otherwise,  $u \in V - W$  so  $u = \sum_{i=1}^n c_i u_i + w$  for some constants  $c_i$ . Hence,  $u + W = \sum_{i=1}^n c_i (u_i + W)$ . Telling us that  $\text{span}\{u_1 + W, \dots, u_n + W\} = V/W$ . We see that the zero vector of  $V/W$  is just  $W$ , thus if  $\sum_{i=1}^n c_i (u_i + W) = W$ ,  $\sum_{i=1}^n c_i u_i = 0$ . As such,  $c_i = 0$  by the linear independence of  $\{u_1, u_2, \dots, u_n\}$ . Therefore,  $\{u_1 + W, \dots, u_n + W\}$  is also linearly independent and therefore a basis of  $V/W$ .

(b)  $\dim(V/W) = \dim(V) - \dim(W)$  by the above result.

26. It has dimension  $n$  as the set  $\{x_i - a_i + a \mid i \leq n\}$  is a basis for the mentioned subspace.

27.  $\dim(W_1 \cap P_n(F)) = m$ , where  $n=2m$  in the case of  $n$  being even and  $n=2m+1$  if  $n$  odd.  
and  $\dim(W_2 \cap P_n(F)) = n-m$

(b) 28. Ideas

$$\beta_{\mathbb{C}} = \{v_1, v_2, \dots, v_n\}$$

$$\beta_{\mathbb{R}} = \{v_1, v_2, \dots, v_n, iv_1, iv_2, \dots, iv_n\}$$

Proof  
Let  $\beta_{\mathbb{C}} := \{v_k \mid k \leq n\}$  be a basis of  $V$  over  $\mathbb{C}$ . We shall see that  $\beta_{\mathbb{R}} := \{v_k \mid k \leq n\} \cup \{iv_k \mid k \leq n\}$  forms a basis for  $V$  over  $\mathbb{R}$  (with

tells the same operations of addition, multiplication, of scalar).

$$\sum_{k=1}^n c_k v_k + \sum_{k=1}^n d_k (iv_k) = 0,$$

$$\sum_{k=1}^n (c_k + d_k i) v_k = 0,$$

We see that  $c_k + d_k i$  must be 0 so  $c_k = d_k = 0$ . By the equality of the two sums above, it is easy to see that  $\text{span}(\beta_{\mathbb{R}}) = \text{span}(\beta_{\mathbb{C}}) = V$ . As such,  $\beta_{\mathbb{R}}$  is indeed a basis for  $V$  over  $\mathbb{R}$  as we claimed, and now, the dimension of  $V$  over  $\mathbb{R}$  must be  $2n$ . □

# Exercises

- 1. (a) False ✓
- (b) False ✓
- (c) False ✓
- (d) True ✓
- (e) True ✓
- (f) True ✓

2. Let  $(k_n)$  be the sequence with  $k_k = 1$  and  $k_n = 0$  if  $n \neq k$ . The set of all such sequences  $(k_n)$  is an infinite linearly independent subset of the space of convergent sequences. This suffices to complete the proof (the rest is trivial). □
3. We shall see that the set of all powers of  $\pi$ , namely  $\{\pi^n \mid n \in \mathbb{N}_0\}$ , is a linearly independent subset of  $V$ .  $\{1\}$  is trivially linearly independent. So suppose  $\{\pi^n \mid n \leq k\}$  is linearly independent for a  $k \in \mathbb{N}_0$ . Then, so must  $\{\pi^n \mid n \leq k+1\}$  lest the transcendence of  $\pi$  is contradicted. Therefore, by induction, this is true of every  $k \in \mathbb{N}_0$ . Consequently, it is easy to see that  $\{\pi^n \mid n \in \mathbb{N}_0\}$  must be linearly independent. □
- As such,  $V$  must be infinite-dimensional because all finite dimensional vector spaces only contain finite linearly independent subsets.

### Self-Proof of Theorem 1.12

Suppose, for the sake of contradiction, that  $\beta$  does not generate  $V$ . Then pick any element  $u$  of  $S - \text{span}(\beta) \neq \emptyset$ , then  $\beta \cup \{u\}$  must be linearly independent, a contradiction. Therefore,  $\beta$  must generate  $V$ . Consequently, it is a basis for  $V$ . □

### Self-Proof of Theorem 1.13

Let  $\mathcal{F}$  be the set of linearly independent subsets of  $V^n$  containing  $S$  if they are nonempty and  $\mathcal{C}$  any chain in  $\mathcal{F}$ . Then we claim  $\cup \mathcal{C} \in \mathcal{F}$ . If not, there exists some vectors  $u_1, u_2, \dots, u_n \in \cup \mathcal{C}$  and constants  $d_1, d_2, \dots, d_n \in \mathbb{F}$  (one of which is nonzero) so  $\sum_{i=1}^n d_i u_i = 0$ . Now, select a member  $A_i \in \mathcal{C}$  for each  $1 \leq i \leq n$  such that  $u_i \in A_i$ . It is clear that there must be some largest (wrt the  $\subseteq$  relation) set  $A_k$  for which  $u_i \in A_k$  for all  $1 \leq i \leq n$ . Since  $A_k \in \mathcal{F}$ , it is linearly independent, a contradiction. As such, we can be certain that  $\cup \mathcal{C} \in \mathcal{F}$ . By the Hausdorff Maximal Principle, there is some maximal linearly independent subset of  $V$  that contains  $S$ . □

LMAO even the use of  $A_i$  and  $A_k$  is the same as the author's

### Corollary

$\mathcal{B}$  is a linearly independent subset of  $V$ . So, from Theorem 1.13 we can be certain of the existence of a maximally independent subset of  $V$ . Which, by Theorem 1.12, must be a basis for  $V$ .

False ✓  
False ✓

4. By Theorem 1.15, for any basis  $\beta_W$  for  $W$ , there exists an extended basis  $\beta_V$  for  $V$  with  $\beta_W \subseteq \beta_V$ . a nonzero vector  $\vec{v}$ , some other vectors  $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n$  and

5. Assume, for the sake of contradiction, that even though  $\beta$  is a basis for  $V$ , there exists  $\vec{v}_{n+1}, \vec{v}_{n+2}, \dots, \vec{v}_{n+m}$  in  $\beta$  with nonzero scalars  $c_1, c_2, \dots, c_{n+m}$  such that  $\sum_{i=1}^n c_i \vec{u}_i - \sum_{i=n+1}^{n+m} c_i \vec{v}_i = \vec{v}$ . But now,

$$\sum_{i=1}^n c_i \vec{u}_i - \sum_{i=n+1}^{n+m} c_i \vec{v}_i = \vec{0},$$

which is a nontrivial representation of  $\vec{0}$  in terms of vectors in  $\beta$ . As such,  $\beta$  is not linearly independent, a contradiction. To summarize, we have shown that if  $\beta$  is a basis for  $V$ , then <sup>the</sup> unique representation of vectors in  $V$  is guaranteed. (Conversely, now suppose unique representation holds for every vector in  $V$ . Then,  $\text{span}(\beta) = V$  is immediate, and more importantly,  $\vec{0}$  has a unique representation too — that must be the trivial representation (empty sum of vectors). Consequently,  $\beta$  is linearly independent. Therefore the biconditional holds.

6. A similar procedure as used in the proof of Theorem 1.13 can be applied.

7. We claim that  $\bigcup_{i=1}^r [\beta - \text{span}(S_i)] \cup S_i$  gives the desired basis for  $V$ . (Clearly,  $\beta \subseteq \text{span}(S)$  so  $\text{span}(S) = V$  is guaranteed. For any vectors  $\vec{u}_i \in \beta - \text{span}(S_i)$  and  $\vec{v}_i \in S_i$  with corresponding constants  $c_i$ , when

$$\sum_{i=1}^n c_i \vec{u}_i + \sum_{i=n+1}^{n+m} c_i \vec{v}_i = \vec{0},$$

$$\sum_{i=1}^n c_i \vec{u}_i = - \sum_{i=n+1}^{n+m} c_i \vec{v}_i.$$

We shall see that  $\sum_{i=1}^n c_i \vec{u}_i \notin \text{span}(S_i)$  whilst  $\sum_{i=n+1}^{n+m} (-c_i) \vec{v}_i \in \text{span}(S_i)$ , all  $c_i$  must be 0.

$v = c_{n+1} \vec{v}_{n+1} + \dots + c_{n+m} \vec{v}_{n+m}$   
 $\downarrow$   
 $\text{span}(\beta \cap \text{span}(S_i))$  since  $\vec{u}_i \in \beta - \text{span}(S_i)$

7. By the previous exercise, taking  $S_2 := \beta U S_1$ , we have a basis  $\mathcal{T}$  for  $U$  that contains  $S_1$ . And so, the desired  $S$  is simply  $\mathcal{T} - S_1$ .



16. Let  $f$  be any polynomial in  $P(\mathbb{R})$  given by  $f(x) = \sum_{i=0}^n c_i x^i$  for some natural  $n$ . Then,  $T\left(\sum_{i=0}^n \frac{c_i}{i+1} x^{i+1}\right) = \sum_{i=0}^n c_i x^i = f(x)$ . So, surjectivity is certain. However,  $T$  is clearly not injective since  $T(f(x)+1) = T(f(x))$  even though  $f(x)+1 \neq f(x)$ .

17. (a) Let  $\beta$  be a basis for  $V$ . Thus, since  $R(T) = \text{span } T[\beta]$ ,  $\text{rank}(T)$  is maximally  $\dim(V) < \dim(W)$  so that  $R(T) \subset W$ .  
 (b) Similarly, when  $\beta$  is a basis for  $V$ , some subset  $\gamma$  of  $T[\beta]$  forms a basis of  $R(T)$ , where in fact,  $\gamma \subset T[\beta]$  must hold left  $\dim(V) = \dim(W)$ .  
 Which entails that there exists some  $x \in \beta$  with  $T(x)$  (capable of being expressed as  $T(\text{some linear combination of vectors in } \gamma)$ , removing any hope that  $T$  is injective.

~~18. Define  $T(x,y) = (x-y, x-y)$ , then  $T(c(x_1+y_1), c(x_2+y_2)) = T(cx_1+cy_1, cx_2+cy_2) = T(cx_1+cx_2, cy_1+cy_2) = T(c(x_1+x_2), c(y_1+y_2)) = T(cx_1+cx_2, cy_1+cy_2) = T(cx_1+cx_2, cy_1+cy_2)$ , thus linearity holds. Now notice that  $T(x,y) = (0,0)$  implies  $x=y$  so  $\{(1,1)\}$  is a basis for  $N(T)$  and hence no  $\rightarrow$  Oh wow, read wrongly~~

18.  $T_0(x,y) = (0,0)$  clearly gives such a linear transformation.

19. Let  $V=W=\mathbb{R}^2$ ,  $T, U: V \rightarrow W$  be defined by  $T(x,y) = (x,y)$  and  $U(x,y) = 2(x,y)$ . Linearity is clear. We see that  $N(T) = N(U) = \{(0,0)\}$ .  
 Similarly, notice that if  $2(x,y) \in R(U)$ ,  $2(x,y) \in R(T)$  because  $T(2x, 2y) = 2(x,y)$ ; and when  $(x,y) \in R(T)$ ,  $(x,y) \in R(U)$  since  $U(\frac{1}{2}(x,y)) = (x,y)$ .  
 As such,  $R(T) = R(U)$  as desired.

21. (a) We notice that

$$T(c(a_n) + (b_n)) = (ca_{n+1} + b_{n+1}) = cT(a_n) + T(b_n), \text{ and } U(c(a_n) + (b_n)) = (ca_{n+1} + b_{n+1}) = cU(a_n) + U(b_n).$$

Hence,  $T$  and  $U$  are linear.

(b) For any sequence  $(a_n) \in V$ ,  $T(a_{n-1}) = (a_n)$  (where  $(a_{n-1})$  is the sequence  $(0, a_1, a_2, a_3, \dots)$ ) so that  $T$  is surjective. Notice that although  $(a_n) = (0, 1, 1, 1, \dots) \neq (1, 1, 1, 1, \dots) = (b_n)$ ,  $T(a_n) = T(b_n)$  such that  $T$  is not injective for sure.

(c) Clearly,  $U$  is injective. Non-surjectivity holds easily as  $(a_n) \notin R(U)$  for any sequence  $(a_n)$  with  $a_1 \neq 0$ .

22.

Ideas

$$T(1,0,0) = a, \quad T(0,1,0) = b, \quad T(0,0,1) = c$$

Proof

Define the scalars  $a := T(1,0,0)$ ,  $b := T(0,1,0)$ ,  $c := T(0,0,1)$ . Now,  $T(x,y,z) = xT(1,0,0) + yT(0,1,0) + zT(0,0,1) = ax + by + cz$ .  
 For  $T: \mathbb{F}^n \rightarrow \mathbb{F}^m$ , the same procedure can be repeated. Finally, consider  $T: \mathbb{F}^n \rightarrow \mathbb{F}^m$ . Define the  $j^{\text{th}}$  coordinate of  $T(e_i)$  to be  $c_{ij}$ . We see that  
 given  $x \in \mathbb{F}^n$ , we have some constants  $s_i$  for which

$$\begin{aligned} T(x) &= \sum_{i=1}^n s_i T(e_i) \\ &= \sum_{i=1}^n \left( s_i \cdot \sum_{j=1}^m c_{ij} e_j \right) \\ &= \sum_{j=1}^m \left( \sum_{i=1}^n s_i c_{ij} \right) e_j. \end{aligned}$$

This is our analogue, which has now been proven too.

13. The nullspace of  $T$  can be

1. A point (0-dimensional)
2. A line (1-dimensional) cutting through  $(0,0,0)$
3. A plane (2-dimensional) cutting through  $(0,0,0)$
4. The whole of  $\mathbb{R}^3$  (3-dimensional)

And in the language of exercise 22, these occur when

1.  $a = b = c = 0$
2. At least one of  $a, b, c$  are non zero
3. At least 2 of  $a, b, c$  are non zero
4. All of  $a, b, c$  are non zero

respectively.

# Self-Proof of Theorem 2.6.

We see that the transformation  $T: V \rightarrow W$  defined by  $T(\sum_{i=1}^n c_i v_i) = \sum_{i=1}^n c_i w_i$  must be linear by its construction. To prove uniqueness, suppose that  $U: V \rightarrow W$  is a linear transformation with  $U(v_i) = w_i$ . Then it is clear that  $U(\sum_{i=1}^n c_i v_i) = \sum_{i=1}^n c_i w_i = T(\sum_{i=1}^n c_i v_i)$ . That is,  $U = T$ . As such, there indeed only exists exactly one such linear transformation  $T$ . □

$$T(cx, y) \neq cT(x, y)$$

$$T(x_1 + x_2, y_1 + y_2) = T(x_1, y_1) + T(x_2, y_2) \neq cT(x_1, y_1) + T(x_2, y_2)$$

## Exercises

1. (a) T
- (b) F ✓
- (c) T (F) *more info T might not be linear*
- (d) T ✓
- (e) F ✓
- (f) F ✓
- (g) T ✓
- (h) F ✓

9. (a) counterexample:  $10T(10, 1) = (10, 10) \neq (1, 10) = T(100, 10)$ .

(b) counterexample:  $T(1, 1) + T(2, 1) = (1, 1) + (2, 4) = (3, 5) \neq (3, 9) = T(3, 1) = T((1, 1) + (2, 1))$ .

$$\begin{aligned} a(1, 1) + b(2, 3) &= (8, 11) \\ a + 2b &= 8 & a + 3b &= 11 \\ a &= 2, & b &= 3 \end{aligned}$$

11. Notice that  $(1, 1), (1, 0) \in \text{span}\{(1, 1), (2, 3)\}$  so  $\{(1, 1), (2, 3)\}$  forms a basis for  $\mathbb{R}^2$ . From Theorem 2.6, it follows that there exists a linear transformation  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^3$  with  $T(1, 1) = (1, 0, 2)$  and  $T(2, 3) = (1, -1, 4)$ . Thus,  $T(8, 11) = 2T(1, 1) + 3T(2, 3) = 2(1, 0, 2) + 3(1, -1, 4) = (5, -3, 16)$ . □

Let  $\beta$  be a basis for  $V$ . Thus, since  $R(T) = \text{span } T[\beta]$ ,  $\text{rank}(T)$  is maximally  $\dim(V) < \dim(W)$  so that  $R(T) \subset W$ .

13. Assume, for the sake of contradiction, that  $S$  is linearly dependent. Then,  $\sum_{i=0}^k c_i v_i = \underline{0}$  for some constants  $c_i$ , at least one of which is nonzero.

We see that this gives rise to a nontrivial representation of the zero vector from  $W$ :

$$\sum_{i=0}^k c_i w_i = T\left(\sum_{i=0}^k c_i v_i\right) = T(\underline{0}) = \underline{0}.$$

However, this contradicts the initial condition that  $\{w_i | 1 \leq i \leq k\}$  is linearly independent. Its such, we can conclude that, in fact,  $S$  being linearly independent must hold true. □

14. (a) Suppose  $T$  is injective, then  $\text{nullity}(T) = 0$  so that the result easily follows. (conversely, consider a linear transformation  $T: V \rightarrow W$  that carries linearly independent subsets of  $V$  into linearly independent subsets of  $W$ . Given any nonzero vector  $v \in V$ ,  $\{v\}$  is linearly independent so  $\{T(v)\}$  is too. Telling us  $T(v) \neq \underline{0}$ , and critically,  $\text{nullity}(T) = 0$ . Hence,  $T$  is injective. Thus, the biconditional holds. □

(b) In the case that  $T[S]$  is linearly independent, exercise 13 of this section immediately informs us that  $S$  is linearly independent. (conversely, when  $S$  is linearly independent, it follows from (a) that  $T[S]$  is linearly independent. □

(c) From (b),  $T[\beta]$  is a linearly independent subset of  $W$  containing  $n$  vectors. This must be a basis of  $W$  since the injectivity of  $T$  tells us  $\text{nullity}(T) = 0$  and its surjectivity says  $R(T) = \dim(W)$ , so that  $\dim(V) = \dim(W) = n$ . □

15. Notice  $cT(f(x)) + T(g(x)) = c \int_0^x f(t) dt + \int_0^x g(t) dt = \int_0^x (cf(t) + g(t)) dt = T(cf(x) + g(x))$ , so linearity of  $T$  follows. Suppose  $T(f(x)) = T(g(x))$  for any  $x \in \mathbb{R}$ . Then,  $\int_0^x f(t) dt = \int_0^x g(t) dt$  such that  $f(x) = g(x)$  as long as  $x \in \mathbb{R}$ . In other words,  $f = g$  and injectivity holds. □

Clearly,  $T$  cannot be surjective since  $T(f(x)) = x f(x)$  lets us know that no constant polynomials exist in  $R(T)$ , lest  $f(x) = \frac{c}{x}$  which is impossible. □

is clear that  $T(\sum_{i=1}^m c_i v_i) = \sum_{i=1}^m c_i w_i = T(\sum_{i=1}^m c_i v_i)$ . That is,  $1_{\mathcal{R}(T)}$ . To prove uniqueness, suppose that  $\gamma$  is another basis for  $\mathcal{R}(T)$ . Then for any  $w \in \mathcal{R}(T)$ , we can express it as a linear combination of  $\gamma$ 's. For some constants  $c_i$ :

$$\sum_{i=1}^n c_i T(v_i) = T(u_j),$$

thus by linearity  $T(\sum_{i=1}^m c_i v_i - u_j) = 0$ .

Let  $\dim(V) = m$  and  $|\mathcal{I}| = n$ . We see that the subset  $\eta := \{\sum_{i=1}^m c_{ij} v_i - u_j \mid j \in \mathcal{I}\}$  must be linearly independent. For any constants  $d_j$  with

$$\sum_{j=1}^{m-n} d_j (\sum_{i=1}^m c_{ij} v_i - u_j) = 0,$$

$$\left[ \sum_{i=1}^m v_i \cdot \left( \sum_{j=1}^{m-n} d_j c_{ij} \right) \right] + \sum_{j=1}^{m-n} (-d_j) u_j = 0,$$

As  $\beta$  is a basis for  $V$ , each  $d_j$  is certainly 0. Similarly,  $\eta$  must also generate  $N(T)$ . For any  $w \in V$ , we can express it as a linear combination of vectors in  $\beta$  for some constants  $a_i$  and  $d_j$ , that is,

$$w = \sum_{i=1}^m a_i v_i + \sum_{j=1}^{m-n} d_j u_j,$$

$$= \sum_{j=1}^{m-n} d_j u_j + \left[ \sum_{i=1}^m v_i \cdot \left( \sum_{j=1}^{m-n} (-d_j) c_{ij} \right) \right] + \left[ \sum_{i=1}^m v_i \cdot \left( \sum_{j=1}^{m-n} d_j c_{ij} \right) \right] + \sum_{i=1}^m a_i v_i.$$

As such, it again follows from linearity that given  $T(w) = 0$ ,

$$T(w) = \sum_{j=1}^{m-n} (-d_j) T(\sum_{i=1}^m c_{ij} v_i - u_j) + \sum_{i=1}^m (a_i + \sum_{j=1}^{m-n} d_j c_{ij}) T(v_i) = 0.$$

Since the  $T(v_i)$ 's come together to form a basis  $\gamma$  for  $W$ , each  $a_i + \sum_{j=1}^{m-n} d_j c_{ij} = 0$ . So,  $w = \sum_{j=1}^{m-n} (-d_j) (\sum_{i=1}^m c_{ij} v_i - u_j)$ .  
 Indeed,  $\eta$  generates  $N(T)$  such that  $\eta$  is now a basis for  $N(T)$ . Consequently, we notice that  $\text{rank}(T) = |\mathcal{I}| = n$  and  $\text{nullity}(T) = |\mathcal{I}^c| = m - n$  where  $\dim(V) = m$  so,  
 $\dim(V) = \text{rank}(T) + \text{nullity}(T)$



## Self - Proof of Theorem 2.4

Assume  $v_1$  and  $v_2$  are distinct vectors in  $V$ , then  $v_2 - v_1 \neq \underline{0}$ . Then,  $T(v_2 - v_1) \neq \underline{0}$  as  $N(T) = \{\underline{0}\}$  so that  $T(v_2) \neq T(v_1) + T(v_2 - v_1) = T(v_1)$ . □

# Self-Proof of the Dimension Theorem

Ideas

Choose any nonzero vector  $T(v_i) \in R(T)$ , where  $v_i \in \beta$ . Then there exists a basis  $\gamma \subseteq T[\beta]$  for  $R(T)$ .

$$T\left(\sum_{i=1}^n c_i v_i\right) = T(\underline{u})$$

$$T\left(\sum_{i=1}^n c_i v_i - \underline{u}\right) = \underline{0}$$

$$\sum_{j=1}^{m-n} d_j \left(\sum_{i=1}^n c_{ij} v_i - u_j\right) = \underline{0}$$

$$\left[\sum_{i=1}^n v_i \cdot \left(\sum_{j=1}^{m-n} d_j c_{ij}\right)\right] + \sum_{j=1}^{m-n} (-d_j) u_j = \underline{0}$$

As  $\beta$  is a basis for  $V$ , each  $d_j = 0$ .

Therefore the subset  $\left\{\sum_{i=1}^n c_{ij} v_i - u_j \mid j \leq m-n\right\}$  of  $V$  forms a basis for  $N(T)$ .

(Still need prove generation tho)

$$T(\underline{w}) = \underline{0}$$

$$\underline{w} = \sum_{i=1}^n a_i v_i + \sum_{j=1}^{m-n} d_j u_j$$

$$= \sum_{j=1}^{m-n} d_j u_j + \left[\sum_{i=1}^n v_i \cdot \left(\sum_{j=1}^{m-n} (-d_j) c_{ij}\right)\right] + \left[\sum_{i=1}^n v_i \cdot \left(\sum_{j=1}^{m-n} d_j c_{ij}\right)\right] + \sum_{i=1}^n a_i v_i$$

$$T(\underline{w}) = T\left[\sum_{j=1}^{m-n} (-d_j) \left(\sum_{i=1}^n c_{ij} v_i - u_j\right)\right] + T\left[\sum_{i=1}^n v_i \cdot \left(a_i + \sum_{j=1}^{m-n} d_j c_{ij}\right)\right]$$

$$\Rightarrow \sum_{i=1}^n \left(a_i + \sum_{j=1}^{m-n} d_j c_{ij}\right) T(v_i) = \underline{0}$$

As  $T(v_i) \in \gamma$ ,  $T(v_i) \neq \underline{0}$  from linear independence. So,

$$a_i + \sum_{j=1}^{m-n} d_j c_{ij} = 0$$

$$\text{Hence, } \underline{w} = \sum_{j=1}^{m-n} (-d_j) \left(\sum_{i=1}^n c_{ij} v_i - u_j\right)$$

Consequently, we get that  
 $\dim(R(T)) = |\gamma| = n$  and  
 $\dim(N(T)) = |\eta| = m-n$  so that  
 $\dim(V) = \dim(R(T)) + \dim(N(T))$ .

□

thus be linear.

$$\sum_{j=1}^n c_j T(\underline{v}_j) = T(\underline{u}_j),$$

$\underline{u}_j \in \beta$  but  $T(\underline{u}_j) \notin \delta$ , we can express it as

Self-Proof of Theorem 2.1

Firstly,  $\underline{0}$  must be in both  $N(T)$  and  $R(T)$  as  $T(\underline{0}) = \underline{0}$   $T(\underline{x}) = \underline{0}$  for any nonzero vector  $\underline{x} \in V$  (if  $V = \{\underline{0}\}$ ,  $T(\underline{0} + \underline{0}) = T(\underline{0})$  tells us  $T(\underline{0}) = \underline{0}$ )

Secondly, suppose  $\underline{u}_1, \underline{u}_2 \in N(T)$  and  $T(\underline{v}_1), T(\underline{v}_2) \in R(T)$ . Then, it is clear that  $T(\underline{u}_1 + \underline{u}_2) = T(\underline{u}_1) + T(\underline{u}_2) = \underline{0} + \underline{0} = \underline{0}$  and  $T(\underline{v}_1) + T(\underline{v}_2) =$

$T(\underline{v}_1 + \underline{v}_2)$  so that  $\underline{u}_1 + \underline{u}_2 \in N(T)$  and  $T(\underline{v}_1) + T(\underline{v}_2) \in R(T)$ . Lastly, closure under scalar multiplication follows in a similar way. Thus,  $N(T)$

and  $R(T)$  are subspaces of  $V$  and  $W$  respectively. □

Self-Proof of Theorem 2.2

Let  $\underline{u}$  be any member of  $V$ . Then it can be expressed as a linear combination of vectors in  $\beta$ , i.e.  $\underline{u} = \sum_{i=1}^n c_i \underline{v}_i$  for some constants  $c_i$ .

thus,  $T(\underline{u}) = T(\sum_{i=1}^n c_i \underline{v}_i) = \sum_{i=1}^n c_i T(\underline{v}_i)$  by property 4 of linear transformations, so  $R(T) = \text{span}(\{T(\underline{v}_1), T(\underline{v}_2), \dots, T(\underline{v}_n)\})$  is (clearly) □

Example 1

$$T(cf + g) = \begin{pmatrix} cf(1) + g(1) - cf(2) - g(2) & 0 \\ 0 & cf(0) + g(0) \end{pmatrix}$$

$$cT(f) + T(g) = \begin{pmatrix} cf(1) - cf(2) & 0 \\ 0 & cf(0) \end{pmatrix} + \begin{pmatrix} g(1) - g(2) & 0 \\ 0 & cf(0) + g(0) \end{pmatrix}$$

Thus,  $T(cf + g) = cT(f) + T(g)$  so  $T$  is indeed linear.



24. Consider  $K \neq \emptyset$  because otherwise the result is immediately vacuously true. Let  $s, r \in K$ . We see that  $r-s \in N(T)$  because  $T(r) = T(r-s) + T(s)$ . Thus, for any  $r \in K$ ,  $r = (r-s) + s \in \{s\} + N(T)$ . In other words,  $S \subseteq \{s\} + N(T)$ . The converse, that  $\{s\} + N(T) \subseteq S$  is trivial as given  $r \in N(T)$ ,  $T(s+r) = T(s) + T(r) = b$ . Therefore, equality holds.  $\square$

35. ~~Define the transformation  $T: V \rightarrow W$  by  $T(x) = f(x)$  for all  $x \in B$~~  Let  $u, v \in V$ , there exists linear combinations of vectors in  $B$ ;  $u = \sum_{i=1}^n a_i x_i$  and  $v = \sum_{i=1}^n b_i x_i$  for some  $x_i \in B$  and  $a_i, b_i \in F$ , by exercise 5 of chapter 1 section 1.7. Now define the transformation  $T: V \rightarrow W$  by  $T(x) = \sum_{i=1}^n a_i f(x_i)$ . Given any constant  $c \in F$ ,  $(T(u) + T(v)) = c \sum_{i=1}^n a_i f(x_i) + \sum_{i=1}^n b_i f(x_i) = T(cu + v)$  is clear. Hence,  $T$  is linear. Suppose  $U: V \rightarrow W$  is another linear transformation with  $U(x) = f(x)$  for all  $x \in B$ . Notice  $U(u) = \sum_{i=1}^n a_i f(x_i) = T(u)$  for each  $u \in V$ . Hence,  $T = U$  which establishes uniqueness.  $\checkmark$   $\square$

38.  $T(\frac{p}{q}x + y) = T(\frac{p}{q}x) + T(y)$   $T(\frac{p}{q}x) = \frac{p}{q}T(x)$  (suppose wlog  $q > p$ )  $T(-x) + T(x) = T(x-x) = 0$   
 $T(\frac{1}{q}x) = \frac{1}{q}T(x)$   $T(-x) = -T(x)$

Proof: Let  $p$  and  $q$  be natural numbers. We see that  $pT(x) = T(px) = T(q(\frac{p}{q}x)) = qT(\frac{p}{q}x)$  by additivity so  $\frac{p}{q}T(x) = T(\frac{p}{q}x)$ . Furthermore,  $T(x) + T(-x) = 0$ , telling us  $T(-x) = -T(x)$ . This suffices to show  $T(\frac{p}{q}x) = \frac{p}{q}T(x)$  for every rational number  $\frac{p}{q} \in \mathbb{Q}$ .

39. Since  $T((a+bi)(c+di)) = T((ac-bd) + (ad+bc)i) = ac-bd + (ad+bc)i \neq ac+bd - (bc-ad)i = (a+bi)(c-di) = (a+bi)T(c+di)$  in general. So  $T$  is not linear.  $\square$

40.  $T(x+y) = T(x) + T(y)$  but  $T(cx) \neq cT(x)$  for some  $c, x \in \mathbb{R}$   
 neither  $T(x) = x+c$  linear  
 $T(2x) = (x+c) \neq (x+c) = cT(x)$   $T(x) = 0$   
 $T(x) = x$   
 $T(x) = qx$

$T(x) = e^x$   
 $T(cx) = e^{cx} \neq ce^x = cT(x)$   
 $T(x) = x^3$   
 $T(cx) = (cx)^3 \neq (x^3) = cT(x)$   
 $T(x+y) = (x+y)^3 \neq x^3 + y^3 = T(x) + T(y)$

2. (b) Let  $\beta$  be the standard ordered bases for  $\mathbb{R}^3$  and  $\mathbb{R}^2$  respectively;

$$T(1,0,0) = (2,1), \quad T(0,1,0) = (3,0), \quad T(0,0,1) = (-1,1).$$

Thus, we obtain

$$[T]_{\beta}^{\beta} = \begin{pmatrix} 2 & 3 & -1 \\ 1 & 0 & 1 \end{pmatrix}.$$

□

(closure under Addition: Since  $T+U: V \rightarrow W$ ,  $T+U \in \mathcal{L}(V,W)$ .

(closure under (scalar) Multiplication: Again, as  $\alpha T: V \rightarrow W$ ,  $\alpha T \in \mathcal{L}(V,W)$ .

(uniqueness of these functions hold true by the well-definedness of  $+$  in  $W$ )

(US 1) - (US 2) is immediate from the commutativity and associativity of  $+$  in  $W$ .

(US 3)  $(T+T_0)(x) = T(x) + T_0(x) = T(x) + \underline{0} = T(x)$  for all  $x \in V$  so  $T+T_0 = T$ .

(US 4)  $(T+(-\mathbb{1})T)(x) = T(x) - T(x) = \underline{0}$  again for each  $x \in V$ . Hence,  $(T-T) = T_0$ .

(US 5)  $\mathbb{1}T(x) = T(x)$  for any  $x \in V$  and  $T: V \rightarrow W$ .

(US 6) - (US 7) are clear from the associativity and distributivity of  $\cdot$  over  $+$  in  $W$ .

(US 8) is similarly straightforward from the properties of  $+$  and  $\cdot$  in  $W$ .

Consequently, we have verified  $\mathcal{L}(V,W)$  to indeed be a vector space.

□

8. Let  $x, y$  be vectors in  $V$  with the unique representations  $\sum_{i=1}^n a_i v_i$  and  $\sum_{i=1}^n b_i v_i$  respectively. Suppose  $c \in \mathbb{F}$ . Since  $\underline{x} + \underline{y} = \sum_{i=1}^n (a_i + b_i) v_i$ , we note that

$$T(\underline{x} + \underline{y}) = \left( [\underline{x} + \underline{y}]_{\beta} \right)_{i_1} = (a_i + b_i)_{i_1} = c \left( [x]_{\beta} \right)_{i_1} + \left( [y]_{\beta} \right)_{i_1} = c \left( T(\underline{x}) \right)_{i_1} + \left( T(\underline{y}) \right)_{i_1} \text{ for any } 1 \leq i_1 \leq n.$$

in other words,  $T$  is linear.

✓ should be ok

Example 3 Self Attempt

$T(1,0) = (1, 0, 2)$ ,  $T(0,1) = (3, 0, -4)$  so  $[T]_{\beta}^{\gamma} = \begin{pmatrix} 1 & 3 \\ 0 & 0 \\ 2 & -4 \end{pmatrix}$  where  $\beta$  and  $\gamma$  are ...

Self-Proof of Theorem 2.7

- (a) Notice that  $(aT+U)(u+v) := aT(u+v) + U(u+v) = (aT(u) + aT(v) + U(u) + U(v)) = (aT+U)(u) + (aT+U)(v)$  so  $aT+U$  is linear.
- (b) Just tedious verification.

Hint perhaps using  $[T+U]_{\beta}^{\gamma}$  is fine too since the author does it (was initially concerned with hidden notation)

Self-Proof of Theorem 2.8

- (a) Let  $\beta := \{u_1, u_2, u_3, \dots, u_n\}$  and  $\gamma := \{w_1, w_2, w_3, \dots, w_m\}$ , with  $T(u_j) = \sum_{i=1}^m a_{ij} w_i$  and  $U(u_j) = \sum_{i=1}^m b_{ij} w_i$ . We see that  $(T+U)(u_j) := T(u_j) + U(u_j) = \sum_{i=1}^m (a_{ij} + b_{ij}) w_i$ . As such,  $[T+U]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma} + [U]_{\beta}^{\gamma}$ . Therefore,  $[T+U]_{ij} = a_{ij} + b_{ij} = [T]_{ij} + [U]_{ij}$  (where  $[T+U]_{ij}$  is the  $(i,j)$ th entry of  $[T+U]_{\beta}^{\gamma}$  for all  $1 \leq i \leq m$  and  $1 \leq j \leq n$ ). This suffices to show  $[cT]_{\beta}^{\gamma} = c[T]_{\beta}^{\gamma}$ .
- (b) Similarly, we notice that given any scalar  $c \in \mathbb{F}$ ,  $(cT)(u_j) = \sum_{i=1}^m ca_{ij} w_i$ . Thus,  $[cT]_{ij} = ca_{ij} = c[T]_{ij}$ .

Exercise

- (a) True ✓
- (b) True ✓ (L.T.s are uniquely determined by their action on a basis (so the entries of the matrix w.r.t. the same))
- (c) False ✓  $[T]_{\beta}^{\gamma} = \left( \begin{matrix} & & \\ & & \\ & & \end{matrix} \right)_{|\beta|=m}$
- (d) True ✓
- (e) True ✓
- (f) False ✓

Matrix Rept of Linear Transformation

$$\begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} \rightarrow \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} a_1x + b_1y + c_1z \\ \vdots \\ a_3x + b_3y + c_3z \end{pmatrix}$$

$\sim \sum_{i=1}^m \gamma_i v_i$   
 $\sim T(u_j)$   
 $\sim U(u_j)$



12. Ideas

If upper triangular,

$$([T]_{\beta})_{ij} = 0 \text{ when } i > j \quad \text{i.e. } T(v_j) = \sum_{i=1}^n a_{ij} v_i \text{ where } a_{ij} = 0 \text{ when } i > j$$

$$= \sum_{i=1}^j a_{ij} v_i \quad \swarrow \text{so}$$

when  $T(v_j) \in \text{span}\{\dots\}$ ,

Proof

Suppose  $[T]_{\beta}$  is upper triangular. Then, by definition, <sup>there exist unique scalars</sup>  $a_{ij} = 0$  when  $i > j$  so  $T(v_j) = \sum_{i=1}^n a_{ij} v_i = \sum_{i=1}^j a_{ij} v_i$ , a linear combination of  $v_1, v_2, \dots, v_j$ . That is,  $T(v_j) \in \text{span}\{v_1, v_2, \dots, v_j\}$ . (Inversely, when  $T(v_j) \in \text{span}\{v_1, v_2, \dots, v_j\}$ ,  $T(v_j) = \sum_{i=1}^j a_{ij} v_i$  <sup>where the unique scalars  $a_{ij}$  ( $1 \leq i \leq j \leq n$ )</sup>. Further defining  $a_{ij} = 0$  for  $i > j$ , we have  $([T]_{\beta})_{ij} = a_{ij}$  by definition, which is just 0 if  $i > j$ . Whence,  $[T]_{\beta}$  is upper triangular indeed. □

11. Ideas

~~$T(W) \subseteq W$ , let  $\{\overset{\delta}{v_1}, v_2, \dots, v_n\}$  be an <sup>(ordered)</sup> basis of  $W$ , with  $\beta = \{v_1, v_2, \dots, v_n\}$  being an extension of  $\delta$  to a <sup>(ordered)</sup> basis for  $V$ .  $T(v_j) = \sum_{i=1}^n a_{ij} v_i$~~

$$([T]_{\beta})_{ij} = a_{ij} = 0 \text{ if } k < i \leq n \text{ and } 0 \leq j \leq k.$$

Proof

Let  $\{v_1, v_2, \dots, v_k\}$  be a basis for  $W$  and  $\beta = \{v_1, v_2, \dots, v_n\}$  be an extension of it to a basis for  $V$ . Given any  $v_j$ , there exists unique scalars  $a_{ij} \in \mathbb{F}$  with  $T(v_j) = \sum_{i=1}^n a_{ij} v_i$ . In fact, by the  $T$ -invariance of  $W$ , if  $k < i \leq n$  and  $0 \leq j \leq k$ ,  $a_{ij} = 0$  is certain. That is,  $([T]_{\beta})_{ij} = 0$  holds. □

2. (b) Let  $\beta$  be the standard ordered bases for  $\mathbb{R}^3$  and  $\mathbb{R}^2$  respectively.

16. (a) (Leaving the zero element  $T_0: V \rightarrow W$  is in  $S^0$ . Moreover,  $(T+U)(x) = T(x) + U(x) = 0$  and  $(cT)(x) = cT(x) = 0$  for  $T, U \in \mathcal{L}(V, W)$ ,  $c \in \mathbb{F}$  and all  $x \in V$ . So,  $(T+U), (cT) \in S^0$ . Accordingly,  $S^0$  is indeed a subspace of  $\mathcal{L}(V, W)$ .

(b) Let  $T \in S_2^0$ . Notice that if  $x \in S_1$ ,  $T(x) = 0$  as  $x \in S_2$ . Hence,  $T \in S_1^0$ . Thus,  $S_2^0 \subseteq S_1^0$  is proven.

(c) Since  $V_1 \cup V_2 \subseteq V_1 + V_2$ , it is straightforward from (b) that  $(V_1 \cup V_2)^0 \subseteq (V_1 \cup V_2)^0$ . Hence, assume  $T \in (V_1 \cup V_2)^0$  and  $x \in V_1 + V_2$ . Then for some  $v_1 \in V_1$  and  $v_2 \in V_2$ ,  $T(x) = T(v_1) + T(v_2) = 0$  so  $T \in (V_1 + V_2)^0$ . Similarly, for  $x \in V_1$  and  $x' \in V_2$ ,  $T(x) = T(x') = 0$  thus  $T \in V_1^0 \cap V_2^0$ . And when  $u \in V_1^0 \cap V_2^0$ , then for  $x \in V_1 \cup V_2$ ,  $u(x) = 0$  is clear. Hence,  $u \in (V_1 \cup V_2)^0$ . In other words,  $(V_1 \cup V_2)^0 = (V_1 + V_2)^0 = V_1^0 \cap V_2^0$ .

17. Ideas:  $\dim(V) = \dim(W)$  need not be finite! (Of course it must be finite lol) Let  $\beta = \{v_1, v_2, \dots, v_n\}$  be any basis of  $V$ . Exercise 20 of Section 1.6 tells us there exists a basis  $\gamma = \{w_1, w_2, \dots, w_k\}$  of  $W$ . wlog,  $w_j = T(v_j)$  for each  $1 \leq j \leq k$ .

By the dimension theorem,  $k = \text{rank}(T) \leq \dim(V) = \dim(W)$ . Exercise 20 of Section 1.6 tells us there exists a basis  $\gamma = \{w_1, w_2, \dots, w_k\}$  of  $W$ . wlog,  $w_j = T(v_j)$  for each  $1 \leq j \leq k$ .

$\subseteq T[\beta]$  for  $R(T)$ , which can be extended into a basis  $\gamma = \{w_1, w_2, \dots, w_k\}$  of  $W$ . wlog,  $w_j = T(v_j)$  for each  $1 \leq j \leq k$ .

Let  $\alpha$  be a basis for  $N(T)$ .  
 $= \{u_1, u_2, \dots, u_n\}$

$$\sum_{i=1}^n c_i u_i = 0$$

$$\sum_{i=1}^n c_i T(u_i) = 0$$

$$c_i = 0 \text{ for } 1 \leq i \leq k$$

$$T(u_j) = T(u_j)$$

Proof: Let  $n = \dim(V) = \dim(W)$  and  $k = \text{rank}(T)$ . By Theorem 2.2 and Exercise 20 of Section 1.6, there exists a basis  $\gamma = \{T(u_1), T(u_2), \dots, T(u_k)\}$  with  $\{u_1, u_2, \dots, u_k\} \subseteq V$  any linearly independent. Now let  $\alpha = \{u_{k+1}, u_{k+2}, \dots, u_n\}$  be a basis for  $N(T)$ . Given any representation of 0 in terms of  $\beta = \{u_1, u_2, \dots, u_n\}$ , i.e.  $\sum_{i=1}^n c_i u_i = 0$ , we have  $\sum_{i=1}^k c_i T(u_i) = 0$  by linearity so  $c_i = 0$  when  $1 \leq i \leq k$ . As such,  $\sum_{i=1}^n c_i u_i = \sum_{i=k+1}^n c_i u_i = 0$ , thus  $c_i = 0$  even if  $k < i \leq n$ . Accordingly,  $\beta$  is a basis for  $V$ . Consequently, for any extension of  $\{T(u_1), T(u_2), \dots, T(u_k)\}$  to a basis  $\gamma$  for  $W$ ,  $([T]_{\beta}^{\gamma})_{ij} = 0$  whenever  $i > j$  (because  $T(u_i)$  can be uniquely expressed in terms of  $\gamma$  as either itself if  $1 \leq j \leq k$ , or 0 if  $k < j \leq n$ ).



9. Let  $k \in \mathbb{R}$ ; notice  $T(k(a+bi) + (c+di)) = (ka+c) - (kb+d)i = kT(a+bi) + T(c+di)$ . This suffices to show  $T$  is linear.  
 $T(1) = 1$  and  $T(i) = -i$ . As such,  $[T]_{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$ .

10.  $([T]_{\beta})_{ij} = 1$  if  $i=j$  or  $i=j+1$ , otherwise it is simply 0. In matrix form, this is represented as

$$[T]_{\beta} = \begin{pmatrix} 1 & 0 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 1 \end{pmatrix}$$

IF  $T(v_j) := v_j + v_{j+1}$   
 Too late in the night read -1 into +1 lol.  
 But conceptually I think I'm fine.

For  $T(v_j) = v_j + v_{j-1}$ ,

$$\begin{pmatrix} 1 & 1 & 0 & \dots & 0 \\ 0 & 1 & 1 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix}$$

Yeah this way of writing these types of matrices is probably clearer.

Ight I'ma head out 😊

Spl/Proof of Theorem 2.9

Let  $x, y \in V$  and  $c \in \mathbb{F}$ . We see that  $U(T(cx+y)) = U(cT(x) + T(y)) = cU(T(x)) + U(T(y))$  by the linearity of  $U$  and  $T$ . Hence,  $UT$  is indeed linear.  
 $= c(U(T(x))) + (U(T(y)))$   
 $= c(UT)(x) + (UT)(y)$

Spl/Proof of Theorem 2.10

(a)  $(T(u_1 + u_2))(x) = T(u_1(x) + u_2(x)) = (Tu_1)(x) + (Tu_2)(x)$  and  $((u_1 + u_2)T)(x) = u_1(T(x)) + u_2(T(x)) = (u_1 T)(x) + (u_2 T)(x)$

(b)  $(T(u_1 u_2))(x) = T(u_1(u_2(x))) = ((Tu_1)u_2)(x)$

(c) Trivial as  $\mathcal{L}(V, W)$  is a vector space with  $I$  being its identity vector (Ex 6 of Section 2.2)

(d)  $a(u_1 u_2)(x) = (au_1)(u_2(x)) = a(u_1(u_2(x))) = u_1((au_2)(x))$

Choice of Basis in Theorem 2.11

Let

$$\begin{aligned} T(v_j) &= \sum_{i=1}^{\dim(W)} a_{ij} h_i \\ &= \sum_{i=1}^{\dim(W)} a_{ij} \left( \sum_{k=1}^{\dim(W)} b_{ki} w_k \right) \\ &= \sum_{k=1}^{\dim(W)} \left( \sum_{i=1}^{\dim(W)} a_{ij} b_{ki} \right) w_k \end{aligned}$$

$$([T]_{\alpha}^{\beta})_{kj} = \sum_{i=1}^{\dim(W)} a_{ij} b_{ki}$$

$$\begin{aligned} U(h_j) &= \sum_{k=1}^{\dim(W)} b_{kj} U(w_k) \\ &= \sum_{k=1}^{\dim(W)} b_{kj} \sum_{i=1}^{\dim(Z)} c_{ik} z_i \\ &= \sum_{i=1}^{\dim(Z)} \left( \sum_{k=1}^{\dim(W)} b_{kj} c_{ik} \right) z_i \end{aligned}$$

$$([U]_{\beta'}^{\delta})_{ij} = \sum_{k=1}^{\dim(W)} b_{kj} c_{ik}$$

$$= \sum_{j=1}^{\dim(W)} b_{jk} c_{ij}$$

$$([U]_{\beta}^{\delta} [T]_{\alpha}^{\beta})_{ij} = \sum_{k=1}^{\dim(W)} \overbrace{([U]_{\beta}^{\delta})_{ik}}^{c_{ik}} \overbrace{([T]_{\alpha}^{\beta})_{kj}}^{\sum_{i=1}^{\dim(W)} a_{ij} b_{ki}}$$

$$([U]_{\beta'}^{\delta} [T]_{\alpha}^{\beta'})_{ij} = \sum_{k=1}^{\dim(W)} \left( \sum_{j=1}^{\dim(W)} b_{jk} c_{ij} \right) a_{kj}$$

$i \rightarrow k$   
 $j \rightarrow j$   
 $k \rightarrow k$

Defn of matrix multiplication suffices to prove this lol



Self-Proof of Theorem 2.13

$$(a) (u_j)_{i2} = \sum_{k=1}^n A_{ik} B_{kj} = \sum_{k=1}^n A_{ik} (v_j)_k$$

$$(b) (v_j)_{i2} = B_{ij} = \left( \sum_{k=1}^n B_{ik} \cdot 0 \right) + B_{ij} \cdot \mathbf{1} = \sum_{k=1}^n B_{ik} (e_j)_{i2} \quad \square$$

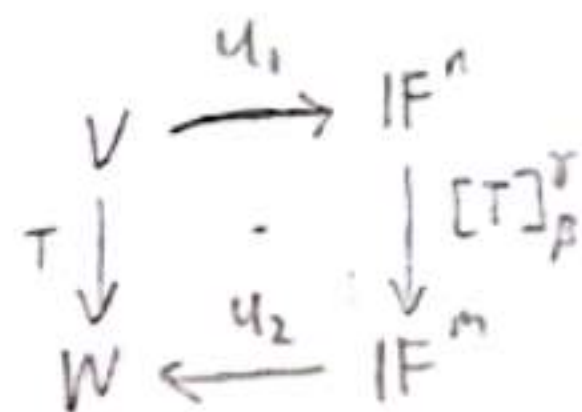
Self-Proof of Theorem 2.14

Idea

$$[T(u)]_\gamma = \left[ \sum_{i=1}^n a_i T(v_i) \right]_\gamma \quad \left( [T]_\beta^\gamma \right)_{ij} = b_{ij}$$

$$= \left[ \sum_{i=1}^n a_i \sum_{j=1}^m b_{ij} w_j \right]_\gamma \quad \left( [u]_\beta \right)_{i2} = a_i$$

$$\sum_{j=1}^m \left( \sum_{i=1}^n a_i b_{ij} \right) w_j \quad \left( [T]_\beta^\gamma [u]_\beta \right)_{i2} = \sum_{k=1}^n b_{ik} a_k$$



Proof

Assume  $n = \dim(V)$ ,  $m = \dim(W)$ . We know there exists scalars  $a_k$  and  $b_{ij}$  for which  $u = \sum_{k=1}^n a_k v_k$  and  $T(v_k) = \sum_{i=1}^m b_{ik} w_i$ . Thus, notice that

$$[T(u)]_\gamma = \left[ \sum_{k=1}^n a_k T(v_k) \right]_\gamma$$

$$= \left[ \sum_{k=1}^n a_k \sum_{i=1}^m b_{ik} w_i \right]_\gamma$$

$$= \left[ \sum_{i=1}^m \left( \sum_{k=1}^n a_k b_{ik} \right) w_i \right]_\gamma$$

whose  $(i, 1)$ th entry is just  $\sum_{k=1}^n b_{ik} a_k$ . Similarly, first noting  $\left( [T]_\beta^\gamma \right)_{ik} = b_{ik}$  and  $\left( [u]_\beta \right)_{i2} = a_i$ , we see that  $\left( [T]_\beta^\gamma [u]_\beta \right)_{i2} = \sum_{k=1}^n b_{ik} a_k$ . Hence, since  $[T(u)]_\gamma$  and  $[T]_\beta^\gamma [u]_\beta$  are both  $m \times 1$  column matrices / vectors in  $\mathbb{F}^m$  with identical entries, they are equivalent.  $\square$



# Self - Proof of Theorem 2.15 -

(b) If  $A = B$ ,  $L_A = L_B$  is clear. So, consider  $L_A \neq L_B$ . Then  $L_A(e_j) = Ae_j$  gives the  $j$ th column of  $A$  since  $(Ae_j)_i = \sum_{k=1}^n A_{ik}(e_j)_k = A_{ij}$ , which is also the  $j$ th column of  $B$ . Since this holds for any column of  $A$  and  $B$ ,  $A = B$ . Accordingly, the biconditional holds.

Alt  
Assume  $A \neq B$ . Then  $A_{ij} \neq B_{ij}$  for some natural  $i, j$ . Then  $L_A(e_j)$  again is the  $j$ th column of  $A$  which differs from that of  $B$  at least in its  $(i, 2)$ th entry. So,  $L_A \neq L_B$ .

(a) Again,  $L_A(e_j) = Ae_j$  is the  $j$ th column of  $A$  so it is just  $\sum_{i=1}^n A_{ij}e_i$ . Hence,  $[[L_A]_{\beta}]_{ij} = A_{ij}$ . Since they are both  $m \times n$  matrices, equality holds.

(c) This is just Theorem 2.12

(d) It follows from Theorem 2.14 that  ${}^{T(v)}[T(v)]_{\gamma} = [T]_{\beta}^{\gamma} [u]_{\beta} = L_{[T]_{\beta}^{\gamma}}(u)$ . Thus,  $T = L_{[T]_{\beta}^{\gamma}}$  where uniqueness must have hold by use of (b).

(e) Notice that

$$\begin{aligned} [L_{AE}(v)]_{ij} &= [(AE)v]_{ij} \\ &= \sum_{k=1}^n (AE)_{ik} v_{kj} \\ &= \sum_{k=1}^n \left( \sum_{l=1}^n A_{il} E_{lk} \right) v_{kj} \\ &= \sum_{l=1}^n A_{il} \left( \sum_{k=1}^n E_{lk} v_{kj} \right) \\ &= \sum_{k=1}^n A_{il} (v)_{kj} \\ &= [A(v)]_{ij} \\ &= [L_A L_E(v)]_{ij} \end{aligned}$$

Therefore,  $L_{AE} = L_A L_E$ .

Self - Proof of Theorem 2.16 / Ex. 19

See self - proof of Thm 2.15 (c).



Exercises

- 1. (a) False ✓
- (b) True ✓
- (c) False ✓
- (d) False (depends on basis) ✓
- (e) False ✓
- (f) False ✓
- (g) False,  $V$  is not necessarily some  $\mathbb{F}^n$ .
- (h) True False  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}^2 = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$
- (i) True ✓
- (j) True ✓

Uhh what was I thinking?  
 If you're representing the same basis vectors ( $Iv(u)=u$ ) in terms of the same basis vectors, we obviously should have the identity matrix  $I$ .

Idea 1 (f) let  $A$  be a  $m \times n$  matrix with  $A^2 = I$ ,

$$(Ae_j)_{i2} = \sum_{k=1}^n A_{ik} (e_j)_{k1} = A_{ij}$$

$$\begin{pmatrix} 0 \\ \vdots \\ A_{ij} \\ \vdots \\ 0 \end{pmatrix} \text{ in } \mathbb{F}^m$$

$$[(AA)(e_j)]_{i1} = [A(Ae_j)]_{i1} \quad \& \quad Ie_j = e_j$$

$$= \sum_{k=1}^n A_{ik} (Ae_j)_{k1} = A_{ij}^2$$

Case 1  $i=j$ :  $A_{ij}^2 = 1 \quad (= (Ie_j)_{i1})$   
 $A_{ij}^2 - 1 = 0$

Check:  
 $A_{ij} + A_{ij} - A_{ij} - 1 = A_{ij}^2 - 1$

$$(A_{ij} + 1)(A_{ij} - 1) = 0$$

$$A_{ij} + 1 = 0 \quad \text{or} \quad A_{ij} - 1 = 0 \quad (\text{fields have no zero divisors})$$

$$A_{ij} = -1 \quad \text{or} \quad A_{ij} = 1$$

Case 2  $i \neq j$ :  $A_{ij}^2 = 0, (= (Ie_j)_{i1})$   
 $A_{ij} = 0$

Therefore,  $A$  need not be  $\pm I$ . A suitable example is  $A_{11} = -1$  but agreeing with  $I$  everywhere else. Then  $[(AA)e_i]_{11} = A_{11}^2 = 1 = (e_i)_{11} = (Ie_i)_{11}$ , even though  $A \neq I$ .

choosing  $\mathbb{F} = \mathbb{R}$  and the matrix  $A$  to be such that

# Exercise 14

(a) We know that for some constants  $a_j$ ,  $z = \sum_{j=1}^n a_j e_j$ . So by the distributivity of matrix multiplication over sums (corollary of Thm 2.12) and Theorem 2.13 (b),

$$Bz = B\left(\sum_{j=1}^n a_j e_j\right) = \sum_{j=1}^n a_j B e_j = \sum_{j=1}^n a_j v_j.$$

(b) Ideas

$$u_j = A v_j = \left(\sum_{i=1}^n w_i\right) v_j = \sum_{i=1}^n w_i v_j$$

Now  $u_j = \sum_{i=1}^n a_i w_i$  with  $a_i = (v_j)_i$

$$(u_j)_{ij} = \sum_{k=1}^n (A_{ik})(B_{kj})$$

$$\left(\sum_{k=1}^n (v_j)_k w_k\right)_{ij} = \sum_{k=1}^n (v_j)_k (w_k)_{ij}$$

$$= \sum_{k=1}^n (v_j)_k (w_k)_{ij}$$

$$= \sum_{k=1}^n B_{kj} A_{ik}$$

$$= (u_j)_{ij}$$

$$u_j = \sum_{k=1}^n w_k$$

Proof  
Let  $w_k$  be the  $k$ th column of  $A$ . We see that  $\left(\sum_{k=1}^n (v_j)_k w_k\right)_{ij} = \sum_{k=1}^n (v_j)_k (w_k)_{ij} = \sum_{k=1}^n B_{kj} A_{ik} = (u_j)_{ij}$ . Therefore, the two  $n \times 1$  matrices,  $u_j$  and  $\sum_{k=1}^n (v_j)_k w_k$ , are equal.

(c) Ideas  
 $wA = (A^t w^t)^t = \left(\sum_{j=1}^n a_j (e_j)^t\right)^t$   $\alpha_j$   $j$ th column of  $A^t$   
 $\alpha_j$   $j$ th row of  $A$   
 $= \sum_{j=1}^n a_j x_j$

Let  $x_j$  be the  $j$ th row of  $A$ .

Proof  
Again,  $w = \sum_{j=1}^n a_j e_j$  for some constants  $a_j$ , so  $w^t = \sum_{j=1}^n a_j (e_j)^t$ . From (a) it now holds that  $wA = (A^t w^t)^t = \left(\sum_{j=1}^n a_j (e_j)^t\right)^t = \sum_{j=1}^n a_j x_j$ .

(d) notice the analogous result that  $\left(\sum_{k=1}^n (v_j)_k v_k\right)_{ij} = \sum_{k=1}^n (v_j)_k (v_k)_{ij} = \sum_{k=1}^n A_{jk} B_{ki} = (u_j)_{ij}$ . Hence, we similarly have that  $u_j = \sum_{k=1}^n (v_j)_k v_k$ .  
Letting  $x_j$  be the  $j$ th row of  $A$  as before, let also  $v_j$  and  $v_j$  be the respective  $j$ th rows of  $AB$  and  $B$ , we

$$2. (a) \quad A(2B+3C) = \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} \left[ \begin{pmatrix} 2 & 0 & -6 \\ 8 & 2 & 4 \end{pmatrix} + \begin{pmatrix} 3 & 3 & 12 \\ -3 & -6 & 0 \end{pmatrix} \right] = \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 5 & 3 & 6 \\ 5 & -4 & 4 \end{pmatrix} = \begin{pmatrix} 20 & -9 & 18 \\ 5 & 10 & 8 \end{pmatrix}$$

$$(AB)D = \left[ \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 1 & 0 & -3 \\ 4 & 1 & 2 \end{pmatrix} \right] \begin{pmatrix} 2 \\ -2 \\ 3 \end{pmatrix} = \begin{pmatrix} 13 & 3 & 3 \\ -2 & -1 & -8 \end{pmatrix} \begin{pmatrix} 2 \\ -2 \\ 3 \end{pmatrix} = \begin{pmatrix} 29 \\ -26 \end{pmatrix}$$

$$A(BD) = \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} \left[ \begin{pmatrix} 1 & 0 & -3 \\ 4 & 1 & 2 \end{pmatrix} \begin{pmatrix} 2 \\ -2 \\ 3 \end{pmatrix} \right] = \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} -1 \\ 12 \end{pmatrix} = \begin{pmatrix} 29 \\ -26 \end{pmatrix}$$

Indeed, we again notice  $(AB)D = A(BD)$ , the associativity of matrix multiplication.

$$(b) \quad A^t = \begin{pmatrix} 2 & -3 & 4 \\ 5 & 1 & 2 \end{pmatrix}, \quad A^t B = \begin{pmatrix} 2 & -3 & 4 \\ 5 & 1 & 2 \end{pmatrix} \begin{pmatrix} 3 & -2 & 0 \\ 1 & -1 & 4 \\ 5 & 5 & 3 \end{pmatrix} = \begin{pmatrix} 23 & 19 & 0 \\ 27 & -2 & 14 \end{pmatrix}$$

$$B C^t = \begin{pmatrix} 3 & -2 & 0 \\ 1 & -1 & 4 \\ 5 & 5 & 3 \end{pmatrix} \begin{pmatrix} 4 \\ 0 \\ 3 \end{pmatrix} = \begin{pmatrix} 12 \\ 16 \\ 29 \end{pmatrix}, \quad (B = (4 \ 0 \ 3)) \begin{pmatrix} 3 & -2 & 0 \\ 1 & -1 & 4 \\ 5 & 5 & 3 \end{pmatrix} = (27 \ 7 \ 9)$$

$$(A = (4 \ 0 \ 3)) \begin{pmatrix} 2 & 5 \\ -3 & 1 \\ 4 & 2 \end{pmatrix} = (20 \ 26)$$

12. (a) If  $T$  is not injective, then  $N(T) \neq \{0_V\}$  so  $T(0_V) = T(x)$  for some nonzero vector  $x$  in  $V$ . Thus,  $(UT)(0_V) = (UT)(x) = 0_Z$ .

OR

Since  $UT$  is injective,  $N(UT) = \{0_V\}$  so  $N(T) = \{0_V\}$ . As such  $T$  must be injective.

$U$  itself need not be injective, only the restriction of  $U$  to  $R(T)$ , i.e.  $U \upharpoonright R(T)$ . In fact, it is clear that assuming  $T$  injective,  $UT$  is injective iff  $U \upharpoonright R(T)$  is injective.

(b) If  $UT$  is surjective, then for each  $z \in Z$  there exists  $x \in V$  with  $(UT)(x) = U(T(x)) = z$ . So, for each  $z \in Z$  there is some vector  $y \in W$  with  $U(y) = z$ . Accordingly  $U$  must be surjective. On the other hand,  $T$  need not be surjective itself. In fact, for example  $T(x)$ , with  $U(y) = z$ . Accordingly  $U$  must be surjective. On the other hand,  $T$  need not be surjective itself. In fact,  $R(T) \geq \dim(Z) < n$  be enough. For instance, consider  $T: \mathbb{F}^2 \rightarrow \mathbb{F}^2$  and  $U: \mathbb{F}^2 \rightarrow \mathbb{F}$  defined by  $T(e_1) = e_1$  and  $T(e_2) = 0_{\mathbb{F}^2}$ ,  $U(e_1) = 1$  and  $U(e_2) = 0$ . Then  $(UT)(e_1) = 1$ , from which the surjectivity of  $UT$  is clear.

(c) Consider any sets  $A, B, C$  and bijections  $T: A \rightarrow B$  and  $U: B \rightarrow C$ . Given  $(UT)(a) = (UT)(a')$ ,  $T(a) = T(a')$  by  $U$ 's injectivity, so  $a = a'$  similarly by  $T$ 's injectivity. Any element  $c$  of  $C$  also has a preimage  $b \in B$  for which  $U(b) = c$  by  $U$ 's surjectivity. Lastly, from  $T$ 's surjectivity, it holds that  $b = T(a)$  for some  $a \in A$ . Therefore, the existence of an element  $a \in A$  for which  $(UT)(a) = c$  is guaranteed. As such,  $UT$  itself is bijective.  $\square$

13. We see that

$$\begin{aligned}\operatorname{tr}(AB) &= \sum_{i=1}^n (AB)_{ii} \\ &= \sum_{i=1}^n \sum_{k=1}^n A_{ik} B_{ki} \\ &= \sum_{k=1}^n \sum_{i=1}^n B_{ki} A_{ik} \\ &= \sum_{k=1}^n (BA)_{kk} \\ &= \operatorname{tr}(BA)\end{aligned}$$

and  $\operatorname{tr}(A) = \operatorname{tr}(A^t)$  follows easily from the fact that  $A_{ii} = (A^t)_{ii}$ .

□

15. Ideas (To see which presentation to use)

Let  $A$  be  $m \times n$ ,  $B$  be  $n \times r$ ;

$$\begin{aligned} \sum_{k=1}^n A_{ik} B_{kj} (u_p)_{i2} &= \sum_{k=1}^n A_{ik} (v_p)_{k2} = \sum_{k=1}^n A_{ik} \cdot \sum_{\ell=1}^r c_\ell (v_{j\ell})_{k2} \\ &= (\alpha_i v_p)_{i2} = \sum_{\ell=1}^r c_\ell \sum_{k=1}^n A_{ik} (v_{j\ell})_{k2} = \sum_{\ell=1}^r c_\ell (u_{j\ell})_{i2} \\ &= \sum_{\ell=1}^r c_\ell (\alpha_i v_{j\ell})_{i2} = \sum_{\ell=1}^r c_\ell (\alpha_i v_{j\ell})_{i2} \\ &= \sum_{\ell=1}^r c_\ell (u_{j\ell})_{i2} \end{aligned}$$

$$(u_{j\ell})_{i2} = \sum_{s=1}^n A_{is} \overbrace{B_{sj\ell}}^{(v_{j\ell})_{s2}}$$

(3) (Ans): Use Theorem 2.13(a).

$$\begin{aligned} u_p &= A v_p \\ &= A(c_1 v_{j1} + c_2 v_{j2} + \dots + c_k v_{jk}) \\ &= c_1 A v_{j1} + c_2 A v_{j2} + \dots + c_k A v_{jk} \\ &= c_1 u_{j1} + c_2 u_{j2} + \dots + c_k u_{jk} \end{aligned}$$

□

Proof  
Let  $A$  and  $B$  be  $m \times n$  and  $n \times r$  matrices respectively, so  $AB$  exists. We see that

$$(u_p)_{i2} = \sum_{s=1}^n A_{is} (v_p)_{s2} = \sum_{s=1}^n A_{is} \cdot \sum_{\ell=1}^r c_\ell (v_{j\ell})_{s2} = \sum_{\ell=1}^r c_\ell \sum_{s=1}^n A_{is} (v_{j\ell})_{s2} = \sum_{\ell=1}^r c_\ell (u_{j\ell})_{i2}$$

Therefore,  $u_p = \sum_{\ell=1}^r c_\ell u_{j\ell}$  indeed.

17. Ideas

When  $T = T^2$ ,  $T(x) = T(T(x))$

If some  $v_j \in \beta$  is not in  $R(T)$ ,

$W_1$  a subspace since  $H$  clearly closed under  $+$ ,  
and  $T(0) = 0$  so  $0 \in W_1$

$$W_1 \oplus N(T) = V$$

$W_1 \cap N(T) = \{0\}$  because for nonzero  $v \in N(T)$ ,  $T(v) = 0 \neq v$

$$\sum_{i=1}^{r \in R(T)} c_i T(v_i) = T(v_j)$$

$$\left( \sum_{i=1}^{r \in R(T)} c_i v_i \right) - v_j \in N(T)$$

$$T(w+n) = w$$

$$\begin{aligned} T(T(w+n)) &= T(w) \\ &= T(w+n) \end{aligned}$$

Proof

First assume that  $T = T^2$ . Then  $W_1 \cap N(T) = \{0\}$  is immediate. Let  $\alpha = \{v_1, v_2, \dots, v_r\}$  be a basis of  $R(T)$  and  $\beta = \{v_1, v_2, \dots, v_n\}$  its extension to  $V$ .

We see that if  $v_j \in \beta - \alpha$ ,  $T(v_j) = \sum_{i=1}^r c_i v_i$  for some scalars  $c_i$  so  $T(v_j) = \sum_{i=1}^r c_i T(v_i)$  and  $v_j - \sum_{i=1}^r c_i T(v_i) \in N(T)$ . Notice that for  $T(u) \in R(T)$ ,  $T(T(u)) = T(u) \in R(T)$ .

Thus,  $R(T) = W_1$ . Hence,  $W_1 \oplus N(T) = V$  since it contains  $\beta$ . By this fact, we know that for any  $u \in V$ , there exists  $y \in W_1$  and  $x \in N(T)$  with  $u = y + x$ .

Accordingly,  $T(u) = y$ . Therefore,  $T$  is indeed a projection on  $W_1$  along  $N(T)$ . Conversely, suppose  $T$  is a projection on  $W_1$  along  $N(T)$ . Then, again, for  $u \in V$  there is  $y \in W_1$  and  $x \in N(T)$  with  $u = y + x$ . As such,  $T(T(u)) = T(y) = T(y) + T(x) = T(y) = T(u)$ . Which informs us  $T = T^2$ .

16. (a) Ideas

rank(T) = rank(T^2) :  $R(T^2) \subseteq R(T)$   
 $\rightarrow R(T^2) = R(T)$

$\alpha = \{v_1, v_2, \dots, v_r\}$   
 $T(v_j) = \sum_{i=1}^n a_{ij} v_i$

$v_j = \sum_{i=1}^n a_i T(v_i)$      $v_j = \sum_{i=1}^n a_i T(v_i)$

False, say  $R(T) \cap N(T) \neq \{0\}$ , ... result ...  $v_j \in V$

$T(v_j) = \sum_{i=1}^n a_{ij} v_i$  for unique  $a_{ij}$

$\text{rank}(T^2) \leq \text{rank}(T) - 1$

$T(T(v_j)) = \sum_{i=1}^n a_{ij} T(v_i)$

$T(v_j) = \sum_{i=1}^n a_{ij} v_i$   
 $= \sum_{i=1}^n a_{ij} \sum_{k=1}^n b_{ki} T(v_k)$  as  $v_i \in R(T^2)$   
 $= \sum_{k=1}^n T\left(\sum_{i=1}^n a_{ij} b_{ki} v_k\right)$

$= \sum_{i=1}^n \left( a_{ij} \sum_{k=1}^n a_{ki} v_k \right)$   
 $= \sum_{k=1}^n \left( \sum_{i=1}^n a_{ij} a_{ki} \right) v_k$

Proof

Assume, for the sake of contradiction, that there exists some nonzero  $v \in R(T) \cap N(T)$ .  $\{v\}$  can be extended to form a basis  $\beta$  for  $R(T)$ . ~~which must also be a basis for  $R(T^2)$~~   
 Furthermore,  $R(T^2)$  being a subspace of  $R(T)$  having equal dimension tells us  $R(T^2) = R(T)$ . Since  $T[\beta]$  spans  $R(T^2)$ ,  $T(v) = 0$  means that  
 $\text{rank}(T^2) \leq \text{rank}(T) - 1$ . A contradiction.

Now let  $\{v_1, v_2, \dots, v_r\}$  be a basis for  $R(T)$ , and  $\beta = \{v_1, v_2, \dots, v_n\}$  being its extension to  $V$ . We know  $T(v_j) = \sum_{i=1}^n a_{ij} v_i$  for (unique) scalars  $a_{ij}$ . In fact, as  $v_i \in R(T)$ ,  
 $T(v_j) = \sum_{i=1}^n a_{ij} \sum_{k=1}^n b_{ki} T(v_k) = T\left(\sum_{k=1}^n \sum_{i=1}^n a_{ij} b_{ki} v_k\right)$ . So,  $v_j - \sum_{k=1}^n \sum_{i=1}^n a_{ij} b_{ki} v_k \in N(T)$ ; which is nonzero for any  $j > r$ . It is hence clear that  $V = R(T) \oplus N(T)$ .

(b) Ideas

$R(T^k) \oplus N(T^k)$

$R(T^k) \cap N(T^k) = \{0\}$

$R(T^k) + N(T^k) = V$   
 If  $A \neq \{0\}$ , exists nonzero  $v \in$  all  $R(T^k)$  and  $N(T^k)$

$A := \bigcap_{i=1}^{\infty} (R(T^i) \cap N(T^i))$

$N(T^2) - N(T) \subseteq R(T)$  as  $T^2: R(T) \rightarrow R(T)$

$R(T) \cap N(T) \supseteq R(T^2) \cap N(T^2) \supseteq \dots$

$R(T) \supseteq R(T^2) \supseteq R(T^3) \supseteq \dots$  nonincreasing sequence

$R(T^k) = R(T^{2k})$  by leastness.

$A$  being an intersection of subspaces is a vector space itself  
 Let  $\alpha$  be a basis of  $A$  containing  $v$ ,  $\beta$  its extension to  $V$ .

$\lambda = \dim(R^k)$  for some positive integer  $k$

of  $\{\dim(R^i) \mid i \in \mathbb{Z}^+\}$ .

~~Since  $V$  is finite dimensional,~~

Here must exist some least dimension  $\wedge$  of  $\{\dim(R^i) \mid i \in \mathbb{Z}^+\}$ . Therefore,  $\text{rank}(T^k) =$

Proof

Since  $R(T) \supseteq R(T^2) \supseteq R(T^3) \supseteq \dots$ , there must exist some least  $\text{rank}(T^k)$  of  $\{\text{rank}(T^i) \mid i \in \mathbb{Z}^+\}$  for some positive integer  $k$ . Therefore,  $\text{rank}(T^k) =$   
 $\text{rank}(T^{2k})$  so by (a),  $V = R(T^k) \oplus N(T^k)$ .



$$= \sum_{i=1}^n c_i(u_i)$$

When  $k=0$ ,  $[T^0]_{\beta} = I_V = I_n = ([T]_{\beta})^0$  (where  $n := \dim(V)$ ) as desired. Assume  $[T^k]_{\beta} = ([T]_{\beta})^k$ . By theorem 2.11,  $[T^{k+1}]_{\beta} = [T^k T]_{\beta} = [T^k]_{\beta} [T]_{\beta} = ([T]_{\beta})^k [T]_{\beta} = ([T]_{\beta})^{k+1}$ . Therefore,  $[T^k]_{\beta} = ([T]_{\beta})^k$  indeed holds for any nonnegative integer  $k$ . □



### Corollary 1

This is just a special case of Theorem 2.18

### Corollary 2

Again, the fact that  $A$  is invertible iff  $L_A$  is, is straight forward from Theorem 2.18. Since Theorem 2.15 says  $(L_A)(L_{A^{-1}}) = L_{AA^{-1}} = I$  (and the other way around), so  $(L_A)^{-1} = L_{A^{-1}}$ . [Or just notice  $(L_A)^{-1} = (L_A)^{-1}[(L_A)(L_{A^{-1}})] = [(L_A)^{-1}(L_A)](L_{A^{-1}}) = L_{A^{-1}}$ ]

### Exercise 13

For reflexivity, just take the identity map. Similarly, for symmetry and transitivity, we simply use the inverse and composition of isomorphisms, respectively.

### Example 3

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \quad \& \quad \begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

### Set Proof of Theorem 2.19

~~When there is an isomorphism  $T: V \rightarrow W$ , Theorem 2.4 tells us  $\text{nullity}(T) = 0$ . Furthermore,~~  
when  $V \cong W$ , the corollary to Theorem 2.17 tells us  $\dim(V) = \dim(W)$ . (conversely, consider  $\dim(V) = \dim(W)$ . Then let  $\beta := \{v_1, v_2, \dots, v_n\}$  and  $\gamma := \{w_1, w_2, \dots, w_n\}$  be bases of  $V$  and  $W$  respectively. Define the linear transformation  $T: V \rightarrow W$  by  $T(v_i) = w_i$ . Now,  $\text{rank}(T) = \dim(W) = \dim(V)$  is clear, so by Theorem 2.4,  $T$  is bijective. And hence, is an isomorphism.

### Self-Proof of Theorem 2.20

Idea:  $\Phi_{\beta}^{\gamma}(T+U) = (\Phi_{\beta}^{\gamma}(T) + \Phi_{\beta}^{\gamma}(U))$  by Theorem 2.8

Injectivity:  $\Phi_{\beta}^{\gamma}(T) = \Phi_{\beta}^{\gamma}(U)$   
 $T(v_j) = U(v_j)$  for any  $1 \leq j \leq n$   
 $T = U$

Surjectivity: Let  $A \in M_{mn}(F)$ . Define the l.t.  $T: V \rightarrow W$  by  
 $T(v_j) = \sum_{i=1}^m a_{ij} w_i$  (possible by Thm 2.6)

Proof  
By Theorem 2.8,  $\Phi_{\beta}^{\gamma}$  must be linear. (Injectivity must also hold because when  $\Phi_{\beta}^{\gamma}(T) = \Phi_{\beta}^{\gamma}(U)$ ,  $T(v_j) = U(v_j)$  for any  $1 \leq j \leq n$ , so  $T = U$  by Theorem 2.6. Similarly, surjectivity holds because of Theorem 2.6 asserting the existence of a linear transformation  $T: V \rightarrow W$  for which  $T(v_j) = \sum_{i=1}^m a_{ij} w_i$  given any matrix  $A \in M_{mn}(F)$ . Hence,  $\Phi_{\beta}^{\gamma}(T) = A$ . As such  $\Phi_{\beta}^{\gamma}$  is indeed an isomorphism.

Corollary  
 $M_{mn}(F)$  is clearly of dimension  $mn$  so by Theorem 2.19, so must  $\mathcal{L}(V, W)$ .

# Self-Proof of Theorem 2.17

Idea

$$T: V \rightarrow W \quad T^{-1}: W \rightarrow V$$

$$T^{-1}(cw_1 + w_2) = T^{-1}(cT(v_1) + T(v_2)) = T^{-1}(T(cv_1 + v_2)) = cv_1 + v_2 = cT^{-1}(w_1) + T^{-1}(w_2)$$

Corollary

~~If  $V$  is finite dimensional,~~ By Theorem 2.5,  $\dim(V) = \text{rank}(T) = \dim(W)$  as  $R(T) = W$ . □

Example 2

$$\begin{pmatrix} 5 & 7 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 3 & -7 \\ -2 & 5 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 3 & -7 \\ -2 & 5 \end{pmatrix} \begin{pmatrix} 5 & 7 \\ 2 & 3 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Thus, the inverse of  $\begin{pmatrix} 5 & 7 \\ 2 & 3 \end{pmatrix}$  is indeed  $\begin{pmatrix} 3 & -7 \\ -2 & 5 \end{pmatrix}$ .

# Self-Proof of Theorem 2.18

Idea

If  $T$  is invertible, let  $v_j$  be the  $j$ th column of  $[T]_{\beta}^{\gamma}$  and  $w_j$  the  $j$ th column of  $[T]_{\beta}^{\gamma}$ .  ~~$([T]_{\beta}^{\gamma} [T^{-1}]_{\beta}^{\gamma})_{ij} = \sum_{k=1}^n ([T]_{\beta}^{\gamma})_{ik} ([T^{-1}]_{\beta}^{\gamma})_{kj}$~~

$$u_j = [T]_{\beta}^{\gamma} v_j$$

Define  $U: W \rightarrow V$  with  $U(w_j) = \sum_{i=1}^n B_{ij} v_i$

$$T(U(w_j)) = \sum_{i=1}^n B_{ij} T(v_i) = \sum_{i=1}^n B_{ij} \sum_{k=1}^n A_{ik} w_k = \sum_{i=1}^n \left( \sum_{k=1}^n A_{ik} B_{ij} \right) w_i = \sum_{i=1}^n (I_n)_{ij} w_i = w_j$$

$$[T]_{\beta}^{\gamma} [T^{-1}]_{\beta}^{\gamma} = [TT^{-1}]_{\beta}^{\gamma} = [I_V]_{\beta}^{\gamma}$$

$$[T^{-1}]_{\beta}^{\gamma} [T]_{\beta}^{\gamma} = \dots = I_n$$

Proof

consider  $T$  being invertible.

Let  $\dim(V) = n$ . By Theorem 2.11,  $[T]_{\beta}^{\gamma} [T^{-1}]_{\beta}^{\gamma} = [TT^{-1}]_{\beta}^{\gamma} = [I_V]_{\beta}^{\gamma} = I_n$ . Similarly,  $[T^{-1}]_{\beta}^{\gamma} [T]_{\beta}^{\gamma} = I_n$ , so equality holds. Furthermore, we have now shown that  $[T^{-1}]_{\beta}^{\gamma} = ([T]_{\beta}^{\gamma})^{-1}$ . Conversely, when  $A := [T]_{\beta}^{\gamma}$  has some inverse  $B$ , define  $U: W \rightarrow V$  with  $U(w_j) = \sum_{i=1}^n B_{ij} v_i$  where

$$T(U(w_j)) = \sum_{k=1}^n B_{kj} T(v_k) = \sum_{k=1}^n B_{kj} \sum_{i=1}^n A_{ik} w_i = \sum_{i=1}^n \left( \sum_{k=1}^n A_{ik} B_{kj} \right) w_i = \sum_{i=1}^n (I_n)_{ij} w_i = w_j$$

It can be shown similarly that  $U(T(v_j)) = v_j$ . It is now clear that  $U = T^{-1}$ .

Note: Oops I thought we didn't have this fact (quite simple to prove anyways) but we do now on page 8! That  $[U]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma}$  implies  $U = T$ .

Proof of Theorem 2.21

Follows similarly as Theorem 2.20.

### Exercises

1. (a) ~~True~~ False, T is not said to be invertible / bijective. ✓

(b) True ✓

(c) False ✓

(d) False ✓

(e) True ✓

(f) False ✓

(g) True ✓

(h) True ✓

(i) True ✓

2. (c) Ideas:  
/ checking

$$3a_1 - 2a_3 = 3b_1 - 2b_3$$

$$3a_1 + 4a_2 = 3b_1 + 4b_2$$
$$a_1 = b_1$$

$$3a_1 - 2a_3 = \alpha$$
$$a_2 = \beta$$

$$3a_1 + 4a_2 = \gamma$$

$$\gamma - 4\beta - 2a_3 = \alpha$$
$$a_3 = \frac{\gamma - 4\beta - \alpha}{2}$$

$$3a_1 + 4\beta = \gamma$$
$$a_1 = \frac{\gamma - 4\beta}{3}$$

Proof  
By solving simultaneous equations, we can easily verify T to be bijective, i.e. invertible. So; yes T is invertible. ✓

(d) T cannot be invertible since  $T(1) = T(2) = 0$  tells us T is not injective. ✓

000  
This 4. Proof

(or)  
Age  
So

Exe  
For  
E

Exe  
For  
E

Exe  
For  
E

Exe  
For  
E

Exe  
For  
E

By the associativity of matrix multiplication shown in Theorem 2.16,  
 $(AB)(B^{-1}A^{-1}) = [(AB)B^{-1}]A^{-1} = [A(BB^{-1})]A^{-1} = AA^{-1} = I.$

It is clear that  $(AB)(B^{-1}A^{-1}) = (B^{-1}A^{-1})(AB) = I$  since  $(A^{-1})^{-1} = A$ . Hence,  $AB$  is indeed invertible, and in fact,  $(AB)^{-1} = B^{-1}A^{-1}$ .

5. From page 89, we know  $(AB)^t = B^t A^t$  for any suitable matrices  $A$  and  $B$  (for which  $AB$  exists). As such,  
 $A^t [(A^{-1})^t] = [A^{-1} A]^t = I$  and  $[(A^{-1})^t] A^t = (A A^{-1})^t = I.$

So,  $A^t$  is indeed invertible, with  $(A^t)^{-1} = (A^{-1})^t$ .  
6. If  $A$  is invertible and  $AB = 0$ ,  $A^{-1}(AB) = A^{-1}0 = 0$  so  $(A^{-1}A)B = B = 0$  by associativity (again, see Theorem 2.16). However,  $AA^{-1}$  is not  $0$ .

7. (a) Suppose, for the sake of contradiction, that  $A^2 = 0$  but  $A$  is invertible. Then,  $A^2 A^{-1} = A = 0 A^{-1} = 0$ . Thus, if  $AB = 0$  and  $B \neq 0$ , then  $A$  can't be invertible.  
(b) Assume  $A$  is invertible, and  $AB = 0$ . Then  $A^{-1}(AB) = I_n B = B = A^{-1}0 = 0$ . Thus, if  $AB = 0$  and  $B \neq 0$ , then  $A$  can't be invertible.

8. See the self-proof of these Corollaries.  
9. (a) If one of them, say  $A$ , isn't invertible, then by Theorem 2.5,  $L_A: \mathbb{F}^n \rightarrow \mathbb{F}^n$  is neither injective nor surjective. Therefore,  $L_{AB}$  cannot be surjective either. Thence implying  $AB$  is not invertible.

(b)  $(a_1 \ a_2 \ a_3) \begin{pmatrix} a_1^{-1} \\ a_2^{-1} \\ a_3^{-1} \end{pmatrix} = (1)$  which is its own inverse.

10. (a) Since  $AB = I_n$ , it is its own inverse so exercise 9 says  $A$  and  $B$  are invertible.  
(b) As  $A$  and  $B$  are invertible,  $(AB)B^{-1} = A = I_n B^{-1} = B^{-1}$ .

(c) For linear transformations  $T: V \rightarrow W$ ,  $U: W \rightarrow Z$ , let  $n = \dim(V) = \dim(Z)$  and  $m = \dim(W)$ . We cannot generalise this to  $m \neq n$  because (a) wouldn't always hold. As an example, let  $V = \mathbb{R}^3$ ,  $W = \mathbb{R}^4$ ,  $Z = \mathbb{R}^3$ , and  $T(a, b, c) = (a, b, c, 0)$ ,  $U(a, b, c, d) = (a, b, c)$ . Then  $UT = I_{\mathbb{R}^3}$ , which is its own inverse, while  $U$  is not invertible, for instance  $U(1, 1, 1, 0) = U(1, 1, 1, 1) = (1, 1, 1)$ . So consider  $n = \dim(V) = \dim(W) = \dim(Z)$  and let's assume  $V = Z$  so  $UT = I_V$ . So, as in (b),  $U = (UT)T^{-1} = I_V T^{-1} = T^{-1}$ . And  $U^{-1}$  and  $T^{-1}$  exists as shown above.

23. Notice  $T(c\sigma + \sigma') = \sum_{i=0}^n (c\sigma + \sigma')(i)x^i = c \sum_{i=0}^n \sigma(i)x^i + \sum_{i=0}^n \sigma'(i)x^i = cT(\sigma) + T(\sigma')$ , thus  $T$  is linear. Let  $Q(x) = \sum_{i=0}^n c_i x^i \in \mathcal{P}(F)$ . Thus, the sequence  $\sigma$  with  $\sigma(i) = c_i$  is the unique sequence so that  $T(\sigma) = Q(x)$ . Accordingly,  $T$  is invertible and hence an isomorphism.

22. Ideas

$$T(f) = T(f')$$

$$\sum_{i=0}^n a_i c_k^i = \sum_{i=0}^n b_i c_k^i \quad \text{for any } 0 \leq k \leq n$$

$$\sum_{i=0}^n (a_i - b_i) c_k^i = 0$$

$$a_{n+1} = \sum_{i=0}^{n+1} b_i c_k^{i-n-1} - \sum_{i=0}^n a_i c_k^{i-n-1} = \sum_{i=0}^n (b_i - a_i) c_k^{i-n-1} + b_{n+1}$$

$$a_{n+1} - b_{n+1} = \sum_{i=0}^n (b_i - a_i) c_k^{i-n-1}$$

$$(a_{n+1} - b_{n+1}) c_k^{n+1} = \sum_{i=0}^n (b_i - a_i) c_k^i$$

$$\sum_{i=0}^n (b_i - a_i) c_k^i = \sum_{i=0}^n (b_i - a_i) c_{k+1}^i$$

$$\sum_{i=0}^n (b_i - a_i) (c_k^i - c_{k+1}^i)$$

12. See self-proof of Theorem 2.21

15. Assume that  $T$  is an isomorphism. Thus, by Theorem 2.2,  $W = \text{span}(T[\beta])$ . Since  $\dim(V) = \dim(W) = n$ , we can be sure that  $T[\beta]$  is linearly independent, and hence, must be a basis for  $W$ . For the converse, see the proof / self-proof of Theorem 2.19.  $\square$

16. Linearity:  $\Phi(cA+B) = B^{-1}(cA+B)B = (cB^{-1}A + I_n)B = cB^{-1}AB + B = c\Phi(A) + \Phi(B)$

Injectivity: Presume  $\Phi(A) = \Phi(A')$ . Then,  $A = B(B^{-1}AB)B^{-1} = B(B^{-1}A'B)B^{-1} = A'$ .

Hence, by Theorem 2.5, bijectivity is certain.

Consequently,  $\Phi$  is indeed an isomorphism as expected.

17. (a) By Theorem 2.1,  $T[V_0] = \text{ran}(T|_{V_0})$  is a subspace of  $W$  (as  $T|_{V_0}$  is still linear).  $\square$

(b) Clearly, the linear transformation  $T|_{V_0}: V_0 \rightarrow T[V_0]$  is an isomorphism, so  $\dim(V_0) = \dim(T[V_0])$  by Theorem 2.19.  $\square$

20. First notice that  $\text{rank}(L_A \phi_\beta) = \text{rank}(L_A)$  since  $\text{ran}(L_A \phi_\beta) = \text{ran}(L_A)$  since  $\phi_\beta$  is an isomorphism. Similarly,  $\text{rank}(T \phi_\gamma) = \text{rank}(T)$  as exercise 17 tells us. Hence, from the commutative diagram that is Fig 2.2,  $\text{rank}(L_A) = \text{rank}(L_A \phi_\beta) = \text{rank}(\phi_\gamma T) = \text{rank}(T)$ .  $\square$

Using the rank-nullity / dimension theorem,  $\text{rank}(L_A) + \text{nullity}(L_A) = n = \text{rank}(T) + \text{nullity}(T)$  so  $\text{nullity}(T) = \text{nullity}(L_A)$ .

21. For  $\sum_{j=1}^m \sum_{i=1}^n c_{ij} T_{ij} = T_0$ ,  $\sum_{j=1}^m \sum_{i=1}^n c_{ij} T_{ij}(v_k) = \sum_{i=1}^n c_{ik} w_i = 0$  for each  $v_k$  so  $c_{ik}$  is always 0 by  $\gamma$ 's linear independence. Hence, so is  $\{T_{ij} \mid 1 \leq i \leq m, 1 \leq j \leq n\}$ . Let  $u \in \mathcal{L}(V, W)$ . For every  $v_k$ ,  $u(v_k) = \sum_{i=1}^n c_{ik} w_i = \sum_{j=1}^m \sum_{i=1}^n c_{ij} T_{ij}(v_k)$ . Therefore by Theorem 2.6,  $u = \sum_{j=1}^m \sum_{i=1}^n c_{ij} T_{ij}$ . As such,  $\{T_{ij} \mid 1 \leq i \leq m, 1 \leq j \leq n\}$  is indeed a basis for  $\mathcal{L}(V, W)$ . Since  $T_{ij}(v_j) = w_i$ , in the  $(i, j)$ th entry we have  $[T_{ij}]_\beta^\gamma$  being 1. Everywhere else, its entries are zero because  $T_{ij}(v_k) = 0_W$  if  $k \neq j$ . Accordingly,  $[T_{ij}]_\beta^\gamma = M^{ij}$  as expected. It is clear that  $\{M^{ij} \mid 1 \leq i \leq m, 1 \leq j \leq n\}$  is a basis for  $M_{m \times n}(\mathbb{F})$ . As such, exercise 15 tells us  $\Phi_\beta^\gamma$  is an isomorphism.

## 2.2. Ideas

When  $n=0$ , immediately true  
 Assume true for  $n$ , then

$$\frac{\sum_{i=0}^{n+1} a_i x^i}{x-r} = a_{n+1} x^n + \frac{\left[ (a_{n+1}r + a_n)x^n + \sum_{i=1}^n a_i x^i \right]}{x-r}$$

$$\frac{a_{n+1} x^{n+1} + a_n x^n + \dots + a_0}{x-r} - \frac{(a_{n+1} x^{n+1} - a_{n+1} r x^n)}{(a_{n+1} r + a_n) x^n + \dots + a_0}$$

$$= a_{n+1} x^n + q(x) + \frac{k}{x-r} \quad \text{by ih}$$

$$p(x) = (x-r)q(x) + k$$

$$p(c_i) = 0 \rightarrow p(x) = (x-r)q(x)$$

Proof

Lemma 1

Notice that for any zeroth degree polynomial  $c \in P_n(\mathbb{F})$  and any  $r \in \mathbb{F}$ ,  $c = 0(x-r) + c$  obviously. Now assume that for any  $p(x) \in P_n(\mathbb{F})$  of degree  $m \leq n$  and each  $r \in \mathbb{F}$ , there exists  $q(x) \in P_n(\mathbb{F})$  of degree  $m-1$  and scalar  $k \in \mathbb{F}$  so  $p(x) = (x-r)q(x) + k$ . As such, given any  $(m+1)$ th degree polynomial  $p(x) \in P_n(\mathbb{F})$ ,  $\diamond$

similarly we have  $p(x) = (a_{n+1} x^n)(x-r) + (a_{n+1}r + a_n)x^n + \sum_{i=1}^n a_i x^i = (a_{n+1} x^n + q(x))(x-r) + k$ , where  $\deg(a_{n+1} x^n + q(x)) = n$  as expected.

Lemma 2

Suppose  $p(x) \in P_n(\mathbb{F})$  and  $x=r$  is one of its roots. Thus, by Lemma 1,  $p(r) = (r-r)q(x) + k = 0$  so  $k=0$ . Implying that  $p(x) = (x-r)q(x)$ .

Linearity is clear from definition, as  $(af+g)(c_i) = af(c_i) + g(c_i)$ . Theorems 2.4 and 2.5 inform us that it suffices to show nullity  $(T) = 0$ . Accordingly, suppose  $T(f) = 0$ . Then given  $0 \leq i \leq n$ ,  $f(c_i) = 0$ . Therefore, from Lemma 2, it holds that  $f(x) / \prod_{i=0}^n (x-c_i) \in P_n(\mathbb{F})$ . As a result,  $\deg(f) = -1$  and  $f$  is the zero polynomial, lest  $\deg(f) - (n+1) \geq 0$  but then  $\deg(f) \geq n+1 > n$ , a contradiction. Hence, nullity  $(T) = 0$ . Which tells us that, indeed,  $T$  is an isomorphism.  $\square$

### Self-Proof of Theorem 2.2.2

(a) This holds by Theorem 2.18 since  $I_V$  has an inverse, namely itself.

(b) By Theorem 2.14,  $[v]_{\beta} = [I_V(v)]_{\beta} = [I_V]_{\beta}^{\beta} [v]_{\beta} = Q [v]_{\beta}$ .

Self-Proof of Theorem 2.2.3  
 By Theorems 2.11 and 2.18,  $[T]_{\beta'} = [I_V]_{\beta'}^{\beta'} [T]_{\beta}^{\beta} [I_V]_{\beta}^{\beta} = ([I_V]_{\beta'}^{\beta'})^{-1} [T]_{\beta}^{\beta} [I_V]_{\beta}^{\beta} = Q^{-1} T Q$ .

### Example 2

$T\begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \\ 4 \end{pmatrix} = 3\begin{pmatrix} 1 \\ 1 \end{pmatrix} - \begin{pmatrix} 1 \\ -1 \end{pmatrix}$  and  $T\begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} 4 \\ -2 \end{pmatrix} = -\begin{pmatrix} 1 \\ 1 \end{pmatrix} + 3\begin{pmatrix} 1 \\ -1 \end{pmatrix}$  so  $[T]_{\beta} = \begin{pmatrix} 3 & 1 \\ -1 & 3 \end{pmatrix}$  indeed.

$T\begin{pmatrix} 2 \\ 4 \end{pmatrix} = \begin{pmatrix} 2 \\ 14 \end{pmatrix} = 4\begin{pmatrix} 2 \\ 4 \end{pmatrix} - 2\begin{pmatrix} 3 \\ 1 \end{pmatrix}$  and  $T\begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 8 \\ 6 \end{pmatrix} = \begin{pmatrix} 2 \\ 4 \end{pmatrix} + 2\begin{pmatrix} 3 \\ 1 \end{pmatrix}$  hence  $[T]_{\beta'} = \begin{pmatrix} 4 & 1 \\ -2 & 2 \end{pmatrix}$ .

### Example 3

$T(0, 1) = \frac{1}{5} T(2(1, 2) + (-2, 1)) = \frac{2}{5} (1, 2) + \frac{1}{5} (2, -1) = \frac{1}{5} (4, 3)$

$T(1, 0) = \frac{1}{5} T(1(1, 2) - 2(-2, 1)) = \frac{1}{5} (1, 2) - \frac{2}{5} (2, -1) = \frac{1}{5} (-3, 4)$

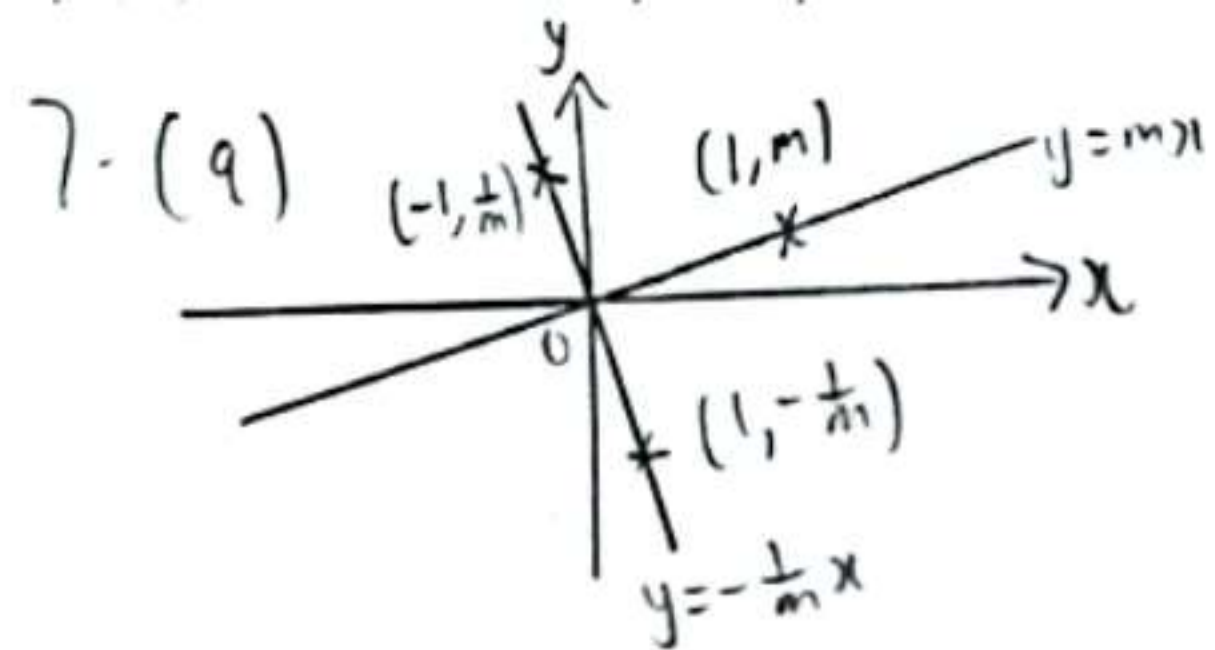
so  $[T]_{\beta} = \frac{1}{5} \begin{pmatrix} 4 & 3 \\ -3 & 4 \end{pmatrix}$



24. The equivalence relation  $\sim$  on  $V$  defined by  $u \sim v$  iff  $u-v \in N(T)$  is being induced, where  $T$  respects  $\sim$  (i.e. if  $u \sim v$  then  $T(u) = T(v)$ ), □

Hence, the necessary results are straight-forward from § Theorem I.

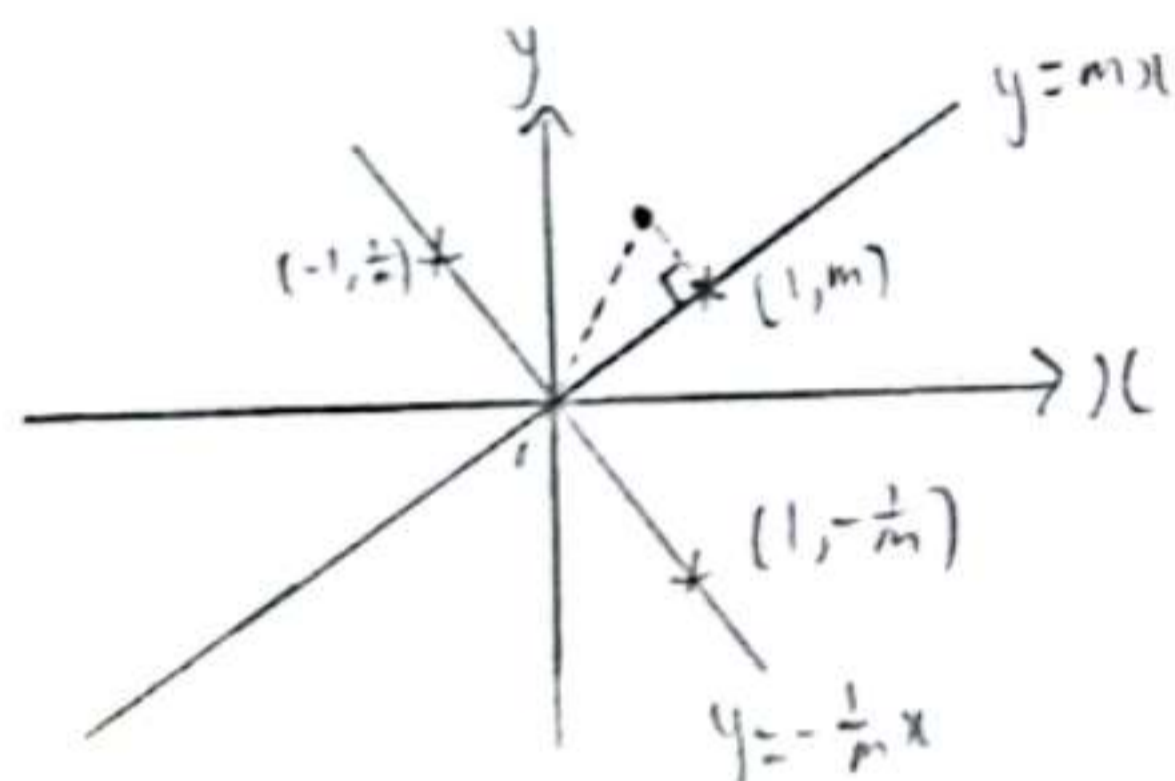
25. Notice  $\Psi(cf+g) = \sum_{s \in S, (cf+g)(s) \neq 0} (cf+g)(s) \cdot s = \sum_{s \in S, (cf+g)(s) \neq 0} cf(s)s + g(s)s = c \sum_{s \in S, f(s) \neq 0} f(s)s + \sum_{s \in S, (cf+g)(s) \neq 0} g(s)s = c\Psi(f) + \Psi(g)$  because if  $(cf+g)(s) = 0$ ,  $cf(s) = -g(s)$  so they still 'cancel' out even if  $f(s)$  or  $g(s)$  is nonzero. Hence, linearity is certain. As such, now consider  $v \in V$ . For some <sup>unique</sup> vectors  $u_i \in S$  and scalars  $c_i \in \mathbb{F}$ ,  $v = \sum_{i=1}^n c_i u_i$  by exercise 5 of section 1.7. Thus, define the unique function  $f: S \rightarrow \mathbb{F}$  with  $f(u_i) = c_i$  and  $f(w) = 0$  otherwise. Accordingly,  $\Psi(f) = \sum_{s \in S, f(s) \neq 0} f(s)s = \sum_{i=1}^n c_i u_i = v$ . Consequently,  $\Psi$  is indeed a bijection, therefore an isomorphism when paired with linearity. □



Define the linear transformation  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  representing the reflection of  $\mathbb{R}^2$  about  $L$  by  $T(1, m) = (1, m)$ , and  $T(-1, \frac{1}{m}) = (1, -\frac{1}{m})$ . In other words, we have that  $[T]_{\beta'} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$ , where  $\beta' := \{(1, m), (-1, \frac{1}{m})\}$ . And let  $\beta := \{(1, 0), (0, 1)\}$ . So the change of coordinate matrix  $Q$  that changes  $\beta'$ -coordinates to  $\beta$  coordinates is given by  $\begin{pmatrix} 1 & -1 \\ m & \frac{1}{m} \end{pmatrix}$ . Further, since  $(1, 0) = \frac{1}{m^2+1} [(1, m) + m^2(1, -\frac{1}{m})]$  and  $(0, 1) = \frac{m}{m^2+1} [(1, m) - (1, -\frac{1}{m})]$ ,  $Q^{-1} = \frac{1}{m^2+1} \begin{pmatrix} 1 & m \\ m^2 & -m \end{pmatrix}$ . Therefore, by Theorem 2-23,  $[T]_{\beta} = Q [T]_{\beta'} Q^{-1} = \frac{1}{m^2+1} \begin{pmatrix} 1 & -1 \\ -m & \frac{1}{m} \end{pmatrix} \begin{pmatrix} 1 & m \\ m^2 & -m \end{pmatrix} = \frac{1}{m^2+1} \begin{pmatrix} 1-m^2 & -2m \\ -2m & m^2-1 \end{pmatrix}$ .  
 Consequently,  $T(x, y) = \frac{1}{m^2+1} (-m^2x + 2my + x, m^2y + 2mx - y)$

To verify, let's compute  $T(1, m) = \frac{1}{m^2+1} (-m^2 + 2m^2 + 1, -m^3 - 2m + m) = (1, m)$   
 and  $T(1, -\frac{1}{m}) = \frac{1}{m^2+1} (-m^2 - 1, m + \frac{1}{m}) = (-1, \frac{1}{m})$  as expected.

(b)



The projection  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  on  $L$  along the line perpendicular to  $L$  is given by  $T(1, m) = (1, m)$  and  $T(1, -\frac{1}{m}) = (0, 0)$ . Notice  $T(1, 0) = T(\frac{1}{m^2+1}(1, m) + \frac{m^2}{m^2+1}(1, -\frac{1}{m})) = \frac{1}{m^2+1}(1, m)$  and  $T(0, 1) = T(\frac{m}{m^2+1}(1, m) - \frac{m}{m^2+1}(1, -\frac{1}{m})) = \frac{m}{m^2+1}(1, m)$ .  
 Hence,  $[T]_{\beta} = \frac{1}{m^2+1} \begin{pmatrix} 1 & m \\ m & m^2 \end{pmatrix}$ . In other words,  $T(x, y) = \frac{1}{m^2+1} (x + my, mx + m^2y)$ .

To check, again we compute  $T(1, m) = \frac{1}{m^2+1} (1 + m^2, m^2 + m^3) = (1, m)$  and  $T(1, -\frac{1}{m}) = \frac{1}{m^2+1} (1 - 1, m - m) = (0, 0)$  as necessary.

Exercises

1. (a) False, it should be  $[x's]_{\beta}$

(b) True

(c) True

(d) False

(e) True

2. (a)

Note  $(a_1, a_2) = a_1 e_1 + a_2 e_2$  and  $(b_1, b_2) = b_1 e_1 + b_2 e_2$ . Thus, the required matrix is  $Q = \begin{pmatrix} a_1 & b_1 \\ a_2 & b_2 \end{pmatrix}$ .

(c)

We see that  $e_1 = 3(2, 5) + 5(-1, -3)$  and  $e_2 = -(2, 5) + 2(-1, -3)$ . So, the change of coordinate matrix necessary is  $Q = \begin{pmatrix} 3 & -1 \\ 5 & -2 \end{pmatrix}$ .

5. First, to find the change of coordinate matrix  $Q$  that changes  $\beta'$ -coordinates to  $\beta$ -coordinates, notice that  $1+x = 1(1) + 1(x)$  and  $1-x = 1(1) + (-1)(x)$ . Therefore  $Q = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$ . Furthermore, since  $T(1) = 0$  and  $T(x) = -1$ ,  $[T]_{\beta} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ . Accordingly, Theorem 2.23 informs us that

$$[T]_{\beta'} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix}$$

To verify this result, we find that  $T(1+x) = 1 = \frac{1}{2}(1+x) + \frac{1}{2}(1-x)$  and  $T(1-x) = -1 = -\frac{1}{2}(1+x) - \frac{1}{2}(1-x)$  such that  $[T]_{\beta'} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix}$ .

6. (a)

Notice  $\begin{pmatrix} 1 & 3 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 4 \\ 2 \end{pmatrix} = 6 \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 2 \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and  $\begin{pmatrix} 1 & 3 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 4 \\ 3 \end{pmatrix} = 11 \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 4 \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ . Hence,  $[LA]_{\beta} = \begin{pmatrix} 6 & 11 \\ -2 & -4 \end{pmatrix}$ .

Similarly,  $Q = \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix}$ . Now we wish to find  $Q^{-1}$  so  $QQ^{-1} = \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a+c & b+d \\ a+2c & b+2d \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ . Solving the corresponding simultaneous equations, we find that  $a=2, b=c=-1$  and  $d=1$ . In other words,  $Q^{-1} = \begin{pmatrix} 2 & -1 \\ -1 & 1 \end{pmatrix}$  such that  $Q^{-1}AQ = \begin{pmatrix} 6 & 11 \\ -2 & -4 \end{pmatrix}$  as expected.

(c)

We see that  $Q = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \end{pmatrix}$ . We now want to see  $Q^{-1}$  with  $QQ^{-1} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \end{pmatrix} \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = \begin{pmatrix} a+d+g & b+e+h & c+f+i \\ a+d+2g & b+e+2h & c+f+2i \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ . Again solving the corresponding simultaneous equations,  $a=1, b=1, c=-1, d=1, e=-1, f=0, g=-1, h=0, i=1$ . Therefore,  $Q^{-1} = \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix}$  and

$$\text{now } [LA]_{\beta} = \begin{pmatrix} 1 & -1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 2 & -2 \\ -2 & -3 & -4 \end{pmatrix}$$

It just happened to just nice do the 2 parts where the ans are provided, not that I necessarily mind checking.

(b) For a linear operator  $T$  on a finite  $n$ -dimensional vector space  $V$ , we can define  $\text{tr}(T) := \text{tr}[T]_s$ , where  $s$  is the standard ordered basis for  $\mathbb{F}^n$ . Well-definedness comes by the fact that  $[T]_s$  is unique as the action of  $T$  on a fixed basis  $s$  is fixed.

1. (a) By Theorem 2.11,  $RQ = [I]_B^{\alpha} [I]_B^{\beta} = [I]_B^{\alpha}$ , which by definition is the change of coordinate matrix that changes  $\alpha$ -coordinates into  $B$ -coordinates.

(b) This is straightforward from Theorem 2.18.

2. Straight forward from Theorem 2.23.

3. Define the invertible linear transformation  $T: V \rightarrow V$  by  $T(x_j) = \sum_{i=1}^n Q_{ij} x_i = x'_j$ . Since  $T$  is a surjection, Theorem 2.2 tells us  $\text{span } \beta' = V$ . Furthermore,  $\text{rank}(T) = n$  so  $\beta'$  is certainly a basis for  $\beta'$ . Now,  $[I_V]_{\beta'}^{\beta} = Q$  too as a result.

4. Idea

$$V := \mathbb{F}^n \quad \beta := \{e_1, e_2, \dots, e_n\} \quad \beta' \text{ as in } B \quad Q$$

$$W := \mathbb{F}^m \quad \gamma := \{e_1, e_2, \dots, e_m\} \quad \gamma' \quad P$$

$$B = P^{-1}AQ$$

$$[T]_{\beta'}^{\gamma'} = [I_W]_{\gamma'}^{\gamma} [T]_{\beta}^{\alpha} [I_V]_{\beta'}^{\beta}$$

Proof

One possible choice is to let  $V := \mathbb{F}^n$ ,  $W := \mathbb{F}^m$ ,  $\beta := \{e_1, e_2, \dots, e_n\}$ , and  $\gamma := \{e_1, e_2, \dots, e_m\}$ . Define  $x'_j := \sum_{i=1}^n Q_{ij} e_i$  and  $\beta' := \{x'_j\}_{j=1}^n$ .  $y'_j := \sum_{i=1}^m P_{ij} e_i$  and  $\gamma' := \{y'_j \mid 1 \leq j \leq m\}$ . By exercise 14,  $Q$  is the change of coordinate matrix changing  $\beta'$ -coordinates to  $\beta$ -coordinates and  $P$  the change of coordinate matrix that changes  $\gamma'$  coordinates to  $\gamma$  coordinates. Now,  $[LA]_{\beta'}^{\gamma'} = [I_W]_{\gamma'}^{\gamma} [LA]_{\beta}^{\alpha} [I_V]_{\beta'}^{\beta} = P^{-1}AQ = B$ .

8. Ideas  $P^{-1} [T]_{\beta}^{\gamma} Q = [I_W]_{\gamma}^{\gamma'} [T]_{\beta}^{\gamma} [I_V]_{\beta'}^{\beta} = [I_W]_{\gamma}^{\gamma'} [T I_V]_{\beta'}^{\gamma} = [T]_{\beta'}^{\gamma'}$  by Theorem 2.11

Proof We notice that  $[T]_{\beta}^{\gamma'} = [I_W]_{\gamma}^{\gamma'} [T I_V]_{\beta'}^{\gamma} = [I_W]_{\gamma}^{\gamma'} [T]_{\beta}^{\gamma} [I_V]_{\beta'}^{\beta} = P^{-1} [T]_{\beta}^{\gamma} Q$  by Theorems 2.11 and 2.18. □

9. Reflexivity

This is immediate since  $B = I_n B I_n = I_n^{-1} B I_n$ , i.e.  $B$  is similar to itself.

Symmetry

When  $B$  is similar to  $A$ , that is  $B = Q^{-1} A Q$  for some invertible  $n \times n$  matrix  $Q$ ,  $A = Q B Q^{-1} = (Q^{-1})^{-1} B (Q^{-1})$ . So,  $A$  is similar to  $B$  too.

Transitivity

If  $A$  is similar to  $B$  and  $B$  is similar to  $C$ , in other words there exist <sup>non invertible</sup> matrices  $P$  and  $Q$  such that  $A = P^{-1} B P$  and  $B = Q^{-1} C Q$ , then since  $B = P A P^{-1}$ ,  $P A P^{-1} = Q^{-1} C Q$ . Therefore,  $A = P^{-1} Q^{-1} C Q P = (QP)^{-1} C (QP)$ . Which also means  $A$  is similar to  $C$ . □

With all three properties satisfied, similarity of  $n \times n$  matrices indeed induces an equivalence relation.

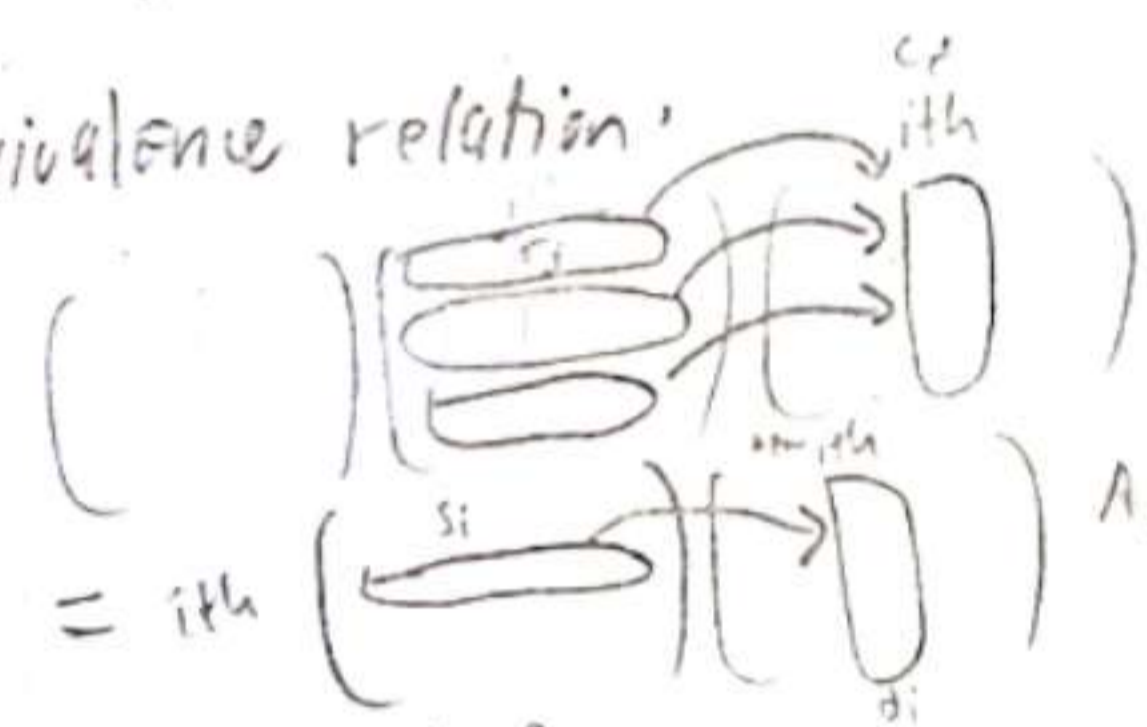
10. (a) Ideas

$$B_{ii} = \sum_{k=1}^n (Q^{-1})_{ik} (A Q)_{ki} = \sum_{k=1}^n (Q^{-1})_{ik} \sum_{\ell=1}^n A_{k\ell} Q_{\ell i} = \sum_{k=1}^n \sum_{\ell=1}^n (Q^{-1})_{ik} A_{k\ell} Q_{\ell i}$$

$$Q^{-1} Q = I_n \implies \sum_{k=1}^n (Q^{-1})_{ik} Q_{ki} = \mathbb{1}$$

$$\sum_{i=1}^n \sum_{k=1}^n \sum_{\ell=1}^n (Q^{-1})_{ik} A_{k\ell} Q_{\ell i} = \sum_{k=1}^n \sum_{\ell=1}^n \left( \sum_{i=1}^n Q_{\ell i} (Q^{-1})_{ik} \right) A_{k\ell} = \sum_{k=1}^n \sum_{\ell=1}^n A_{k\ell} (I_n)_{\ell k} = \sum_{k=1}^n (A I_n)_{kk} = \sum_{k=1}^n A_{kk}$$

$A_{ii} = \sum_{j=1}^n s_j d_j$   
 $\approx \sum_{j=1}^n s_j c_j$   
 $= \sum_{k=1}^n (A I_n)_{kk} = \sum_{k=1}^n A_{kk}$



Proof

Since  $A$  and  $B$  are similar,  $A = Q^{-1} B Q$  for some invertible  $n \times n$  matrix  $Q$ . Hence,  $\text{tr}(A) = \sum_{i=1}^n \sum_{k=1}^n (Q^{-1})_{ik} (B Q)_{ki} = \sum_{i=1}^n \sum_{k=1}^n (Q^{-1})_{ik} \sum_{\ell=1}^n B_{\ell k} Q_{\ell i} = \sum_{k=1}^n \sum_{\ell=1}^n B_{\ell k} (I_n)_{\ell k} = \sum_{k=1}^n (B I_n)_{kk} = \text{tr}(B)$ . □

Idea for alternate proof  $[L_B]_{\gamma} = [I_n]_s [L_B]_s [I_n]_r$

$\text{tr}[L_B]_{\gamma} = \text{tr}[L_B]_s = \text{tr}(B)$   
 $\gamma = \{w_1, \dots, w_n\}$ ,  $s = \{e_1, \dots, e_n\}$   
 $L_B(w_j) = \sum_{i=1}^n c_{ij} w_i$   $j^{\text{th}}$  of  $L_B(w_j)$  is  $c_{jj} w_j$



### Self-Proof of Theorem 2.26

Linearity is clear so since the previous lemma informs us that  $\text{nullity}(\Psi) = 0$ , which means  $\Psi$  is injective by Theorem 2.4. Furthermore, for any  $\alpha: V^* \rightarrow \mathbb{F}$ , letting  $\beta := \{v_1, v_2, \dots, v_n\}$  be a basis for  $V$ , we can define  $x := \sum_{i=1}^n \alpha(f_i) v_i$  such that  $\hat{\alpha}(f_i) = \alpha(f_i)$ . By Theorem 2.6,  $\hat{\alpha} = \alpha$ . In other words,  $\Psi(x) = \alpha$  and  $\Psi$  is surjective. Now, we can be certain it is an isomorphism.

Yeah or just use <sup>(directly)</sup> Thm 2.5, but this way is a bit funner as expected

### Self-Proof of Corollary

Let  $\beta^* := \{f_1, f_2, \dots, f_n\}$  be a basis for  $V^*$  and  $\beta^{**} := \{g_1, g_2, \dots, g_n\}$  the dual basis of  $\beta^*$ . Using Theorem 2.26, define  $u_i := \Psi^{-1}(g_i)$  for  $1 \leq i \leq n$  and 0 otherwise. which must be a basis for  $V$  by Theorem 2.2. Furthermore, we notice that since  $\Psi(u_i) = g_i$ ,  $f_j(u_i) = g_i(f_j)$ , which is  $\mathbb{1}$  if  $i=j$ . As such, it is clear that  $\beta^*$  is the dual basis of  $\beta := \{u_1, u_2, \dots, u_n\}$ .

# Self-Proof of Theorem 2.24

Idea 1

$$f(x_i) = f(x_i) \cdot \underbrace{1}_{f_i(x_i)} = \sum_{i=1}^n f_i(x_i) \cdot c_i = \sum_{i=1}^n f_i(x_i) f_i(c_i x_i)$$

Proof

Given  $v \in V$ , we see that for some scalars  $c_i \in \mathbb{F}$ ,  $f(v) = \sum_{i=1}^n f(c_i x_i) = \sum_{i=1}^n f_i(x_i) \cdot c_i = \sum_{i=1}^n f_i(x_i) f_i(c_i x_i) = \sum_{i=1}^n f_i(x_i) f_i(v)$ . Therefore,  $f = \sum_{i=1}^n f_i(x_i) f_i$  holds.  $\square$

# Self-Proof of Theorem 2.25

Idea  $T^t(g_j) = g_j T = \sum_{i=1}^n (g_j \circ T)(x_i) f_i$

$$([T^t]_{\gamma^*}^{\beta^*})_{ij} = (g_j \circ T)(v_i) \quad T(v_j) = \sum_{k=1}^n ([T]_{\beta}^{\alpha})_{kj} w_k$$

$$= g_j \left( \sum_{k=1}^n ([T]_{\beta}^{\alpha})_{kj} w_k \right) = ([T]_{\beta}^{\alpha})_{ji}$$

Proof

Let  $\beta := \{v_1, v_2, \dots, v_n\}$ ,  $\beta^* := \{f_1, f_2, \dots, f_n\}$ ,  $\gamma := \{w_1, w_2, \dots, w_m\}$ ,  $\gamma^* := \{g_1, g_2, \dots, g_m\}$ . By Theorem 2.24, linearity certainly holds since  $T^t(cg + g') = (cg + g')T = cgT + g'T$ .  
 Accordingly,  $([T^t]_{\gamma^*}^{\beta^*})_{ij} = (g_j \circ T)(v_i) = g_j \left( \sum_{k=1}^n ([T]_{\beta}^{\alpha})_{ki} w_k \right) = ([T]_{\beta}^{\alpha})_{ji}$ , so that  $[T^t]_{\gamma^*}^{\beta^*} = ([T]_{\beta}^{\alpha})^t$ .  $\square$

# Self-Proof of Lemma

Idea: If  $x \neq 0$ ,  $f_i(x) \neq 0$  for some  $i$ .  $f(x) = 0$  for all  $f \in V^*$

Proof: Let  $\beta := \{u_1, u_2, \dots, u_n\}$  be a basis for  $V$  and suppose  $x \in V$  is nonzero. Then there exist scalars  $c_i \in \mathbb{F}$  with  $x = \sum_{i=1}^n c_i u_i$  where some  $c_j \neq 0$ . Assume  $\hat{x}(f_j) = f_j(x) = c_j \neq 0$ .  
 The transformation  $\chi: V \rightarrow V^{**}$  defined by  $\chi(x) = \hat{x}$  is an isomorphism.  $\square$

Guess for what Theorem 2.26 is about: The transformation  $\chi: V \rightarrow V^{**}$  defined by  $\chi(x) = \hat{x}$  is an isomorphism.

Idea:  $V^{**} = \mathcal{L}(\mathcal{L}(V, \mathbb{F}), \mathbb{F})$

$$G \in V^{**} \Rightarrow G: V^* \rightarrow \mathbb{F}$$

for all  $f$ ,  $G(f) = c = f(x)$  for some  $x$ ?

Counter-arg not necessary. We can define  $G(f_i) = 0$  (where  $\mathbb{F} = \mathbb{R}$ ) so  $f_i(x) = 0$ .  
 $\hat{x} = 0$ -function  
 $\hat{x}(f) = 0$  for any  $f \in V^*$

$$\chi: V \rightarrow V^{**}, \chi(x) = \hat{x}$$

$$f: V \rightarrow \mathbb{F} \quad (\hat{x}(f)) = f(x+y) = f(x) + f(y) \text{ by linearity}$$

$$= c f(x) + d f(y) = c \hat{x}(f) + d \hat{y}(f)$$

By Lemma, nullity( $\chi$ ) = 0. Theorem 2.4 says  $\chi$  is injective.  
 let  $G: V^* \rightarrow \mathbb{F}$  want to find  $x$  so  $f_i(x) = G(f_i)$   
 $G(f_i) = c_i$   
 $x := \sum_{i=1}^n G(f_i) u_i$  so  $\hat{x}(f_i) = f_i(x) = G(f_i)$  surjectivity  $\checkmark$

# Exercises

1. (a) False ✓

(b) False. If  $\dim(V) = n$ , then a linear functional  $T: V \rightarrow \mathbb{F}$  would have a  $1 \times n$  matrix representation.

(c) False, true only if  $V$  is of finite dimension.

(d) False. Let  $V$  be the vector space of all sequences  $s: \mathbb{N} \rightarrow \mathbb{N}$  with a finite number of nonzero entries, over the field of itself,  $\mathbb{N}$ . Then it is clear that  $\dim(V) = \aleph_0$ . Since for finite spaces  $W$ ,  $\dim(W^*) = \dim(W) < \aleph_0$ ; for infinite spaces  $Z$ ,  $\dim(Z^*) > \dim(Z) \geq \aleph_0$ , so there exists no vector space whose dual is isomorphic to  $V$ .

→ "Assume all vector spaces are finite dimensional" oops

(e) False ✓

(f) True ✓

(g) True ✓

(h) False ✓

2. (a) This is a linear functional ✓

(c) This is also a linear functional ✓

(f) This is a linear functional as well ✓

3. (b)  $f_1(a+bx+cx^2) = a$ ,  $f_2(a+bx+cx^2) = b$ ,  $f_3(a+bx+cx^2) = c$

(a) Notice  
 $f_1(1,0,1) = f_1(e_1) + f_1(e_3) = 1$   
 $f_1(1,2,1) = f_1(e_1) + 2f_1(e_2) + f_1(e_3) = 0$   
 $f_1(0,0,1) = f_1(e_3) = 0$

Solving similarly for all 3, we have that  $f_1(x,y,z) = x - \frac{1}{2}y$ ,  $f_2(x,y,z) = \frac{1}{2}y$ ,  $f_3(x,y,z) = -x + z$  using a calculator.  
 (Just change the coefficients on the right hand side, [010] and [001] for  $f_2$  and  $f_3$ )



4. We know  $\dim(V^*) = \dim(V) = 3$  so proving  $\text{span}\{f_1, f_2, f_3\} = V^*$  suffices. To do this, we just need to show there exists linear functionals  $g_1, g_2, g_3 \in V^*$  with  $g_1(x, y, z) = x$ ,  $g_2(x, y, z) = y$ ,  $g_3(x, y, z) = z$ . Notice that for  $g_1$ ,

$$a(x-2y) + b(x+y+z) + c(y-3z) = x$$

$$(a+b)x + (-2a+b+c)y + (b-3c)z = x$$

Thus, by comparing,  $a+b=1$ ,  $-2a+b+c=0$ ,  $b-3c=0$ ; Solving these simultaneous equations, we have that  $g_1 = \frac{2}{5}f_1 + \frac{3}{5}f_2 + \frac{1}{5}f_3$ ,  $g_2 = -\frac{3}{10}f_1 + \frac{3}{10}f_2 + \frac{1}{10}f_3$ ,  $g_3 = -\frac{1}{10}f_1 + \frac{1}{10}f_2 - \frac{3}{10}f_3$  by repeating a similar procedure for  $g_2$  and  $g_3$ . Now, for each linear functional  $f$  on  $V$ ,  $f = f(1,0,0)g_1 + f(0,1,0)g_2 + f(0,0,1)g_3$  which can be expressed as a linear combination of  $f_1, f_2, f_3$ .

Accordingly,  $\{f_1, f_2, f_3\}$  is a basis for  $V^*$ .

$\{f_1, f_2, f_3\}$  is the dual basis of some  $\beta = \{(a_1, a_2, a_3), (b_1, b_2, b_3), (c_1, c_2, c_3)\}$ , so we have the following equations

$$a_1 - 2a_2 = 1 \quad b_1 + b_2 + b_3 = 1 \quad c_2 - 3c_3 = 1$$

$$2a_1 - 2a_2 = 0 \text{ (if } a_1 \neq 0) \quad 2b_1 + b_2 + b_3 = 0 \text{ (if } b_1 \neq 0) \quad 2(c_2 - 3c_3) = 0 \text{ (if } c_2 \neq 0)$$

$$b_1 + 2b_2 + b_3 = 0 \text{ (if } b_2 \neq 0)$$

Therefore,  $a_1 = a_2 = -1$ ,  $b_1 = b_2 = -1$  and  $b_3 = 3$ ,  $c_2 = -1$  and  $c_3 = -\frac{2}{3}$ . We need  $\beta$  to span  $\mathbb{R}^3$  so for any  $(x, \beta, \gamma) \in \mathbb{R}^3$  there should exist  $a, b, c \in \mathbb{R}$  with

$$a(-1, -1, a_3) + b(-1, -1, 3) + c(c_1, -1, -\frac{2}{3}) = (x, \beta, \gamma)$$

$$-a - b + ca_3 = x \quad -a - b - c = \beta \quad aa_3 + 3b - \frac{2}{3}c = \gamma$$

$$(c_1+1)c = x - \beta \quad a + b = \frac{\beta - x}{c_1+1} - \beta \quad aa_3 + \frac{3\beta - 3x}{c_1+1} - 3\beta - 3a + \frac{2}{3}(\frac{\beta - x}{c_1+1}) = \gamma$$

$$c = \frac{x - \beta}{c_1+1} \quad (a_3 - 3)a = \gamma + 3\beta + \frac{4}{3}(\frac{x - \beta}{c_1+1})$$

Hence, one possibility for  $\beta$  is  $\{(-1, -1, 1), (-1, -1, 3), (1, -1, -\frac{2}{3})\}$ . For which  $\{f_1, f_2, f_3\}$  is the dual basis of, by construction.  $\square$

7. (a)  $T^t(f)(a+bx) = f(T(a+bx)) = f(a-2a-2b, a+b) = f(-a-2b, a+b) = -a-2b-2(a+b) = -3a-4b$

(b) Let  $\gamma^* := \{g_1, g_2\}$  be the dual basis of  $\gamma := \{(1,0), (0,1)\}$  and  $\beta^* := \{f_1, f_2\}$  the dual basis of  $\beta := \{1, x\}$ .  
 There exists scalars  $a, b \in \mathbb{R}$  with  $T^t(g_1) = af_1 + bf_2$  with  $T^t(g_1)(1) = -1$  and  $T^t(g_1)(x) = -2$  so that  $a = -1, b = -2$ .

Similarly,  $T^t(g_2) = a'f_1 + b'f_2$  such that  $T^t(g_2)(1) = 1$  and  $T^t(g_2)(x) = 1$ . Therefore,  $a' = 1$  and  $b' = 1$ .

Consequently,  $[T^t]_{\beta^*}^{\gamma^*} = \begin{pmatrix} -1 & -2 \\ 1 & 1 \end{pmatrix}$ .

(c) Notice  $T(1) = (-1, 1)$  and  $T(x) = (-2, 1)$ . As such,  $[T]_{\beta}^{\gamma} = \begin{pmatrix} -1 & -2 \\ 1 & 1 \end{pmatrix}$ . Indeed, as we expected,  $\begin{pmatrix} -1 & -2 \\ 1 & 1 \end{pmatrix}^t = \begin{pmatrix} -1 & 1 \\ -2 & 1 \end{pmatrix}$ .

8. Ideas

$u := (a, b, c) \quad v := (d, e, f)$   
 $\alpha a + \beta d = 0 \quad \alpha b + \beta e = 0 \quad \alpha c + \beta f = 0 \Rightarrow \alpha = \beta = 0$   
 $\alpha = -\frac{\beta d}{a} \quad -\frac{\beta d}{a} + \beta e = 0$   
 $\alpha \beta e = \beta b d$   
 $a = \frac{bd}{e}$

$T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$   
 $T(r, s, t) := su + tv$   
 $\beta := \{u, v, w\}$

$f(su + tv) = 0$   
 $f(u) = f(v) := 0, \quad f(w) := 1$

$(\mathbb{R}^3)^* := \mathcal{L}(\mathbb{R}^3, \mathbb{R})$

required vector / L.F.A.L  $f: \mathbb{R}^3 \rightarrow \mathbb{R}$

Proof

Define the required linear functional on  $\mathbb{R}^3$  by  $f(u) = f(v) := 0$  and  $f(w) := 1$ , where  $w$  is a vector with  $\beta := \{u, v, w\}$  being a basis for  $\mathbb{R}^3$ .  
 Then, for every  $a, b, c \in \mathbb{R}$ ,  $f(au + bv + cw) = 0$  implies  $c = 0$ . Hence,  $N(f) = \{su + tv \mid s, t \in \mathbb{R}\}$ .



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9. Ideas

Assume  $T$  linear, want  $f_i: \mathbb{F}^n \rightarrow \mathbb{F}$

Define  $f_j(e_i) = ([T]_s^{s'})_{ji}$        $\sum_{j=1}^m f_j(\sum_{i=1}^n c_i e_i) =$

$T(\sum_{i=1}^n c_i e_i) = \sum_{i=1}^n c_i T(e_i) = \sum_{i=1}^n c_i \sum_{j=1}^m ([T]_s^{s'})_{ji} e_j$

or:  $f_j(x) := g_j(T(x))$

Proof

Assume that  $T$  is linear, then define the linear functional on  $\mathbb{F}^n$ ,  $f_j$ , by  $f_j(e_i) = ([T]_s^{s'})_{ji}$  where  $s$  and  $s'$  are the standard ordered bases for  $\mathbb{F}^n$  and  $\mathbb{F}^m$ , respectively. So,  $T(\sum_{i=1}^n c_i e_i) = \sum_{i=1}^n c_i \sum_{j=1}^m ([T]_s^{s'})_{ji} e_j = \sum_{j=1}^m (\sum_{i=1}^n c_i f_j(e_i)) e_j = \sum_{j=1}^m f_j(\sum_{i=1}^n c_i e_i) e_j = (f_1(\sum_{i=1}^n c_i e_i), f_2(\sum_{i=1}^n c_i e_i), \dots, f_m(\sum_{i=1}^n c_i e_i))$   
 Conversely, if  $T(x) = (f_1(x), f_2(x), \dots, f_m(x))$ ,  $T(ax+by) = \sum_{j=1}^m f_j(ax+by) e_j = a \sum_{j=1}^m f_j(x) e_j + b \sum_{j=1}^m f_j(y) e_j = aT(x) + bT(y)$  which means  $T$  is linear. □

So, the biconditional holds.

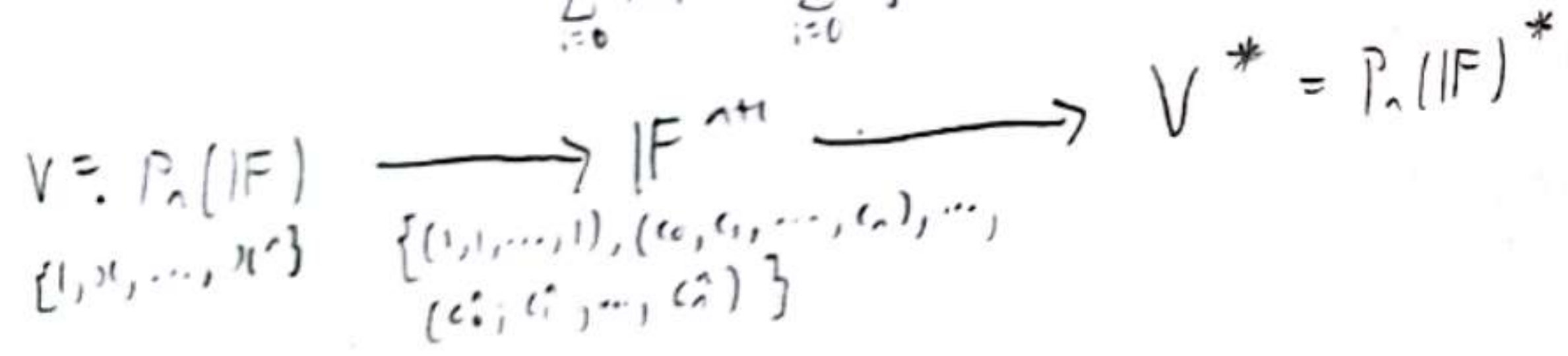
10. (a) Ideas

$g_i(x^j) = \delta_{ij}$        $f_i(x^j) = c_{ij}$        $\sum_{k=1}^n a_{ik} f_k(x^j) =$   
 $g_i = \sum_{k=1}^n a_{ik} f_k$        $\sum_{k=1}^n a_{ik} c_{jk} = \mathbb{1}$        $\sum_{k=1}^n a_{ik} c_{jk} = 0$  if  $i \neq j$   
 $\sigma(x^i) = c_{ij}$

$\sum_{i=0}^n c_i e_i \sim \sum_{i=0}^n c_i g_i =: \sigma$

$\sum_{i=1}^n a_i f_i = T_0$   
 $\sum_{i=1}^n a_i p(c_i) = 0$  for all  $p \in P_n(\mathbb{F})$   
 $\sum_{i=1}^n \sum_{j=1}^n a_i b_j c_i^j = 0$  for all  $b_j \in \mathbb{F}$

let  $k := |\{1 \leq i \leq n \mid a_i \neq 0\}|$ ,  $b_j :=$



$f(c_i) = f_i(f(x))$

Proof  
 Let  $\gamma := \{1, x, \dots, x^n\}$ ,  $\bar{\gamma} := \{e_1, e_2, \dots, e_{n+1}\}$  — the standard ordered bases for  $P_n(\mathbb{F})$  and  $\mathbb{F}^{n+1}$  — and  $\gamma^* := \{g_0, g_1, \dots, g_n\}$  be the dual basis of  $\gamma$ .  
 Notice that Exercise 22 of section 2.5 tells us that  $\bar{\gamma}' := \{(1, \dots, 1), (c_0, c_1, \dots, c_n), \dots, (c_0^k, c_1^k, \dots, c_n^k)\}$  is also a basis for  $\mathbb{F}^{n+1}$ . By defining the isomorphism  $T: \mathbb{F}^{n+1} \rightarrow V^*$  with  $T(e_{i+1}) = g_i$ , we see that  $T[\bar{\gamma}']$  forms a basis for  $V^*$ . (Consider any one of its members;  $T(\sum_{i=0}^n c_i^k e_{i+1}) = \sum_{i=0}^n c_i^k g_i$  and note  $\sum_{k=0}^n c_i^k g_k(x^i) = c_i^k x^i = f_j(x^i)$  for any  $c_i^k = 1$ ). As such,  $T[\bar{\gamma}'] = \{f_0, f_1, \dots, f_n\}$ . □

Ideas

$$T^t(f) := fT \quad T^{tt}(g) := gT^t$$

$$T^{tt}(g)(f) = gT^t(f) = g^f T$$

$$\psi_2 T(v) = T(\hat{v}) \quad W^* \rightarrow \mathbb{F} \quad T(\hat{v})(f) = f(T(v))$$

$$T^{tt} \psi_1(v) = T^{tt}(\hat{v}) = \hat{v} T^t \quad W^* \rightarrow V^* \rightarrow \mathbb{F}$$

$$\hat{v} T^t(f) = \hat{v}(fT) = fT(v) = T(v)(f)$$

Proof  $v \in V$  and  $f \in W^*$ ,  $[\hat{T}(v)](f) := f(T(v)) = fT(v) = \hat{v}(fT) = \hat{v} T^t(f)$ , so  $\psi_2 T(v) = T(\hat{v}) = \hat{v} T^t = T^{tt}(\hat{v}) = T^{tt} \psi_1(v) = T^{tt} \psi_1(v)$

And hence,  $\psi_2 T = T^{tt} \psi_1$ , as expected since the above holds for each  $v \in V$ . The diagram commutes, in other words. □

2. Ideas

$$\psi(v_i) = \hat{v}_i : V^* \rightarrow \mathbb{F}, \quad \hat{v}_i(h) = h(v_i)$$

$$\hat{v}_i(f_j) = f_j(v_i) = \delta_{ji} = \delta_{ij} = g_i(f_j) \quad f_i(v_j) = \delta_{ij}, \quad g_i(f_j) = \delta_{ij}$$

$$V \rightarrow \mathbb{F} \quad V^* \rightarrow \mathbb{F}$$

Proof Let  $\beta^* := \{f_1, f_2, \dots, f_n\}$  and  $\beta^{**} := \{g_1, g_2, \dots, g_n\}$  be the dual bases of  $\beta$  and  $\beta^*$  respectively. Consider any  $1 \leq i \leq n$  and  $1 \leq j \leq n$ , then  $[\psi(v_i)](f_j) := \hat{v}_i(f_j) := f_j(v_i) := \delta_{ji} = \delta_{ij} = g_i(f_j)$  so  $\psi(v_i) = g_i$ . (consequently,

3. (a) Since  $T_c(x) = 0$  for all  $x \in V$ ,  $T_0 \in S_0$ . Furthermore, for any two  $f, g \in S^0$ ,  $c \in \mathbb{F}$ , and  $x \in S$ ;  $(f+g)(x) = f(x) + g(x) = 0$  and  $(cf)(x) = c \cdot 0 = 0$ . Thus,  $f+g$  and  $cf$  are both in  $S^0$ . Hence  $S^0$  is a subspace of  $V^*$ .

(b) Ideas

$$x \in V-W \quad \beta := \{x, x_1, x_2, \dots, x_n\}$$

$$f_0 \in S$$

Proof Let  $\beta := \{x, x_1, x_2, \dots, x_n\}$  be a basis for  $V$  and  $\beta^* := \{f_0, f_1, \dots, f_n\}$  be its dual basis. Since  $x \in V-W$ ,  $f_0 \in W^0$  where  $f_0(x) := 1 \neq 0$ .  $\beta$  be a basis of  $V$  extend from  $S$   $\beta^*$  be its dual basis  $= \{h_1, h_2, \dots, h_m\}$

(c) Ideas

$$(S^0)^0 = \{f \in V^{**} \mid f(g) = 0 \text{ for all } g \in S^0\} \quad (\cong V^*) \quad \hat{x}(g) = g(x) = 0$$

$$f(h_{n+j}) = 0 \quad \text{for } 1 \leq j \leq m-n \quad \text{span}\{\hat{v}_i \mid 1 \leq i \leq n\}$$

$$f(g) = \sum_{i=1}^n c_i \hat{v}_i(g) = \sum_{i=1}^n c_i g(v_i) \quad h_j(v_i) = \delta_{ji}$$

$$\sum_{i=1}^n f(h_j) \hat{v}_i(h_j) = f(h_j)$$

Proof Let  $\{v_1, v_2, \dots, v_n\}$  be the least subset of  $S$  whose span includes  $S$ ,  $\beta := \{v_1, v_2, \dots, v_n\}$  be its extension to  $V$ , and  $\beta^* := \{h_1, h_2, \dots, h_m\}$  the dual basis of  $\beta$ . For  $f \in (S^0)^0$ , that is, any  $f: V^* \rightarrow \mathbb{F}$  with  $f(g) = 0$  for each  $g \in S^0$ ,  $\sum_{i=1}^n f(h_j) \hat{v}_i(h_j) = f(h_j)$  if  $1 \leq j \leq m$  and  $\sum_{i=1}^n f(h_j) \hat{v}_i(h_j) = 0 = f(h_j)$  otherwise. Hence,  $f = \sum_{j=1}^m f(h_j) \hat{v}_j$ . So, we have that  $\text{span}\{\psi(S)\} \cong (S^0)^0$ . (conversely, when  $x \in S$ ,  $\hat{x}(g) = g(x) = 0$  for every  $g \in S^0$ . Thus,  $\text{span}\{\psi(S)\} \subseteq (S^0)^0$ . And accordingly, equality holds.

10. (b) By the corollary to Theorem 2.2 (c), there exists a basis  $\beta := \{p_0(x), p_1(x), p_2(x), \dots, p_n(x)\}$  for  $\mathcal{P}_n(\mathbb{F})$ , where  $\{f_0, f_1, \dots, f_n\}$ , as defined in (a), is the dual basis of  $\beta$ . Accordingly,  $p_i(c_j) = f_j(p_i(x)) = \delta_{ij}$ . (Consider another basis  $\gamma := \{q_0(x), q_1(x), q_2(x), \dots, q_n(x)\}$  for which  $\{f_0, f_1, \dots, f_n\}$  is also the dual basis of  $\gamma$ . Then,  $p_i(c_j) = q_i(c_j) = \delta_{ij}$  for each  $0 \leq i \leq n$  and  $0 \leq j \leq n$ . ~~So~~ So  $\frac{p_i(x)}{\prod_{j \neq i} (x - c_j)} = \frac{q_i(x)}{\prod_{j \neq i} (x - c_j)} = k \in \mathbb{F}$  by Lemma 2 of exercise 2.2 in section 2.5. (In fact, as  $p_i(c_i) = q_i(c_i) = 1$ ,  $k = 1$ .) Hence,  $p_i = q_i$  for any  $0 \leq i \leq n$ , thereby establishing uniqueness.

(c) By defining the polynomial as suggested,  $q(x) = \sum_{j=0}^n a_j \delta_{ji} = a_i$  by (b). Uniqueness clearly holds from the fact that  $\beta$ , as defined above in (b), is a basis for  $\mathcal{P}_n(\mathbb{F})$ . Hence, we see that  $p(c_i) = a_i p_i(c_i) = a_i$ .

(d) Let  $p(x) \in V = \mathcal{P}_n(\mathbb{F})$ . Then there exist scalars  $a_i$  with  $p(x) = \sum_{i=0}^n a_i p_i(x)$  by virtue of  $\beta$  being a basis. Hence, we see that  $p(c_i) = a_i p_i(c_i) = a_i$ . Thus, it holds that  $p(x) = \sum_{i=0}^n p(c_i) p_i(x)$ .

(e) Ideas  
 $\sum_{i=0}^n p(c_i) \int_a^b p_i(t) dt = \int_a^b \sum_{i=0}^n p(c_i) p_i(t) dt = \int_a^b p(t) dt$   
 Direct  
 Simply notice that  $\int_a^b p(t) dt = \int_a^b \sum_{i=0}^n p(c_i) p_i(t) dt = \sum_{i=0}^n p(c_i) \int_a^b p_i(t) dt = \sum_{i=0}^n p(c_i) d_i$  by (d) and properties of the definite integral.

We see that when  $n=1$ ,  $d_0 = \int_a^b \frac{t-b}{a-b} dt = \left[ \frac{t^2 - 2bt}{2(a-b)} \right]_a^b = \frac{b-a}{2}$  and  $d_1 = \int_a^b \frac{t-a}{b-a} dt = \frac{b-a}{2}$ . So

$\int_a^b p(t) dt = p(a)d_0 + p(b)d_1 = \frac{b-a}{2} [p(a) + p(b)]$  [for  $p(t)$  of degree at most  $n=1$ ]  
 Similarly, in the case where  $n=2$ ,  $d_0 = \int_a^b \frac{t - \frac{a+b}{2}}{a - \frac{a+b}{2}} \cdot \frac{t-b}{a-b} dt = \frac{b-a}{6}$ ,  $d_1 = \int_a^b \frac{t-a}{\frac{a+b}{2} - a} \cdot \frac{t-b}{\frac{a+b}{2} - b} dt = \frac{2}{3}(b-a)$ , and  $d_2 = \int_a^b \frac{t-a}{b-a} \cdot \frac{t - \frac{a+b}{2}}{b - \frac{a+b}{2}} dt = \frac{b-a}{6}$   
 by wolfram alpha. Hence,  $\int_a^b p(t) dt = p(a)d_0 + p(c_1)d_1 + p(c_2)d_2 = \frac{b-a}{6} p(a) + \frac{2}{3}(b-a) p(\frac{a+b}{2}) + \frac{b-a}{6} p(b) = \frac{b-a}{6} [p(a) + 4p(\frac{a+b}{2}) + p(b)]$  as expected.  
 [for  $p(t)$  of degree at most  $n=2$ ]

(Newton-Cotes formulas)



(d) Ideas  
 Let  $W_1^0 = W_2^0$ , i.e.  $\{f: V \rightarrow \mathbb{F} \mid f(x) = 0 \forall x \in W_1\} = \{f: V \rightarrow \mathbb{F} \mid f(x) = 0 \forall x \in W_2\}$   
 If  $x \in W_1$ ,  $f(x) = 0$  for all  $f \in W_2^0 \Rightarrow$

OR: Suppose  $W_1 \neq W_2$ , and wlog, that  $\exists x (x \in W_1 \text{ and } x \notin W_2)$   
 $\beta := \{x, x_1, x_2, \dots, x_m\}$  a basis for  $V$   
 $f(x) := 1, f(x_i) := 0$   
 $\Rightarrow f(y)$  always 0 for  $y \in W_2$ ,  $f \in W_1$   
 $\Rightarrow f(x) = 1$ ,  $f \notin W_1$   
 its extension to a basis

Proof  
 Assume  $W_1 \neq W_2$ , and without loss of generality, that there exists  $x \in W_1 - W_2$ . Let  $\beta := \{x_1, x_2, \dots, x_m\}$  be a basis for  $W_2$ , and  $\beta' := \{x, x_1, x_2, \dots, x_m\}$  its extension to a basis for  $V$ .  
 Define  $f: V \rightarrow \mathbb{F}$  with  $f(x) = 1$  and  $f(x_i) = 0$  for every  $1 \leq i \leq m$ . As such,  $f \in W_2^0$  but  $f \notin W_1^0$ . Therefore,  $W_1^0 \neq W_2^0$ . The converse is trivial. Hence, the biconditional holds.

(e) By exercise 16(c), this must be true. □

14. Ideas  $W_m$   
 $\beta := \{v_1, v_2, \dots, v_m\}$  basis for  $W$ ,  $\beta^* := \{f_1, \dots, f_m\}$  its dual basis  
 $j \geq 1: f_{m+j} \in W^0$ , while  $1 \leq i \leq m: f_i \notin W^0$  since  $f_i(v_i) = 1 \neq 0$

Proof Let  $\{v_1, v_2, \dots, v_n\}$  be a basis for  $V$ ,  $\beta := \{v_1, v_2, \dots, v_m\}$  its extension to a basis for  $V$ , and  $\beta^* := \{f_1, f_2, \dots, f_m\}$  the dual basis of  $\beta$ . Then, for  $1 \leq j \leq m-n$  and  $1 \leq i \leq m$ ,  $f_{m+j} \in W^0$  while  $f_i \notin W^0$ . That is,  $\beta^* = \{f_1, f_2, \dots, f_m\}$  and  $\dim(W^0) = m-n$ . So,  $\dim(W) + \dim(W^0) = n + (m-n) = m = \dim(V)$ .  
 as  $f = \sum_{i=1}^m f(v_i) f_i$  for  $f \in W^0$  by Thm 2.24 □

15. Ideas  $T^t(f) = f \circ T = N(T^t)$   
 $f \circ T = T_0$   
 $f(T(v)) = 0 \forall v \in V$   
 $(R(T))^0 := \{f: W \rightarrow \mathbb{F} \mid \underbrace{f(w) = 0 \text{ for all } w \in R(T)}_{f(T(v)) = 0 \text{ for all } v \in V}\}$

Proof  
 When  $f \in N(T^t)$ , then  $f(T(v)) = 0$  for every  $v \in V$ . In other words,  $f(w) = 0$  given  $w \in R(T)$ . Thus,  $f \in (R(T))^0$  and implying  $N(T^t) \subseteq (R(T))^0$ . (conversely, say  $f \in (R(T))^0$ . Then, as aforementioned,  $f(T(v)) = 0$  for any  $v \in V$ . Accordingly,  $T^t(f) = f \circ T = T_0$ , telling us  $f \in N(T^t)$  and  $(R(T))^0 \subseteq N(T^t)$ .  
 Hence, equality holds. □

# 16. Ideas

$A$  is  $m \times n \iff A^t$  is  $n \times m$   
 $Ax, x = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}^n \quad A^t x, x = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}^m$

$L_A: \mathbb{F}^n \rightarrow \mathbb{F}^m \quad L_{A^t}: \mathbb{F}^m \rightarrow \mathbb{F}^n$

$[L_A^t]_S = A^t \quad L_A^t: (\mathbb{F}^n)^* \rightarrow (\mathbb{F}^m)^*$

$L_A^t(f) := f L_A \quad f(Ax)$   
 $f L_A(e_j) = f \left[ \sum_{i=1}^m (A^t)_{ij} e_i \right]$

Example

$$A = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix} \quad x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

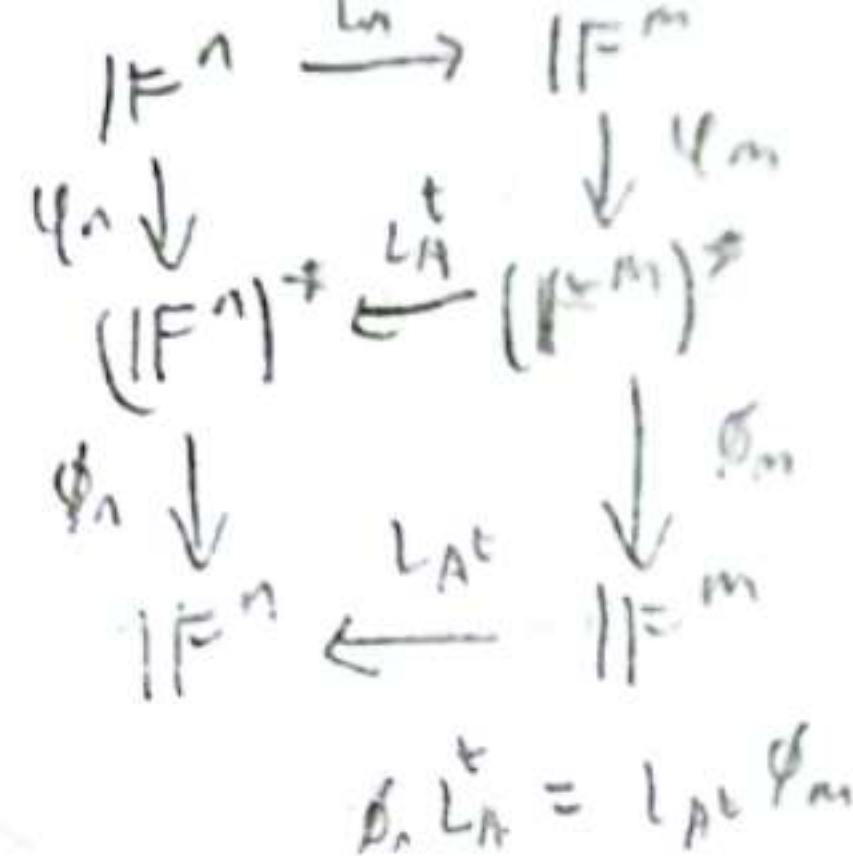
$f: \mathbb{F}^m \rightarrow \mathbb{F} \quad f \in R(L_A) = T_0$   
 $N(L_A^t) = (R(L_A))^0$

$\text{rank}(L_A) + \dim(R(L_A)^0) = m$

$\text{rank}(L_A) + \text{nullity}(L_A^t) = m$

$\text{rank}(L_A) + [m - \text{rank}(L_A^t)] = m$

$\text{rank}(L_A^t) = \text{rank}(L_A)$



$\phi_n L_A^t = L_{A^t} \phi_m$

## Proof

First, notice that since  $N(L_A^t) = (R(L_A))^0$  according to exercise 15, this means  $\text{rank}(L_A) + \text{nullity}(L_A^t) = m$  by exercise 14. As such,  $\text{rank}(L_A^t) = \text{rank}(L_A)$ . Furthermore, by letting  $S_n := \{e_i \mid 1 \leq i \leq n\}$  (the standard ordered basis for  $\mathbb{F}^n$ ), we notice  $\phi_{S_n} L_A^t = L_{A^t} \phi_{S_m}$  because  $[\phi_{S_n}^t]_S = A^t$ .  
 In other words,  $\text{rank}(L_A) = \text{rank}(L_A^t) = \text{rank}(A^t)$ . □

18. Ideas

$$\Phi(f) = f_0 \text{ (the } f_0: S \rightarrow \mathbb{F} \text{ with } f(s) = 0)$$

$$f_s = f_0$$

$$\forall t \in S, f_s(t) = 0$$

Proof  
 Suppose  $\Phi(f) = f_0$ , the zero function from  $f_0: S \rightarrow \mathbb{F}$  with  $f(t) = 0$ . Then  $f_s = f_0$  so, for each  $t \in S$ ,  $f_s(t) = 0$ . Since  $S$  is a basis for  $V$ ,  $f = T_0$ , the zero transformation  $T_0: V \rightarrow \mathbb{F}$ . By Theorems 2.4 & 2.5,  $\Phi$  is guaranteed to be an isomorphism.  $\square$

19. (a) Let  $\gamma := \{v_1, v_2, \dots, v_n\}$  be a basis for  $W$ ,  $\beta := \{v_1, v_2, \dots, v_m\}$  be its extension to a basis for  $V$ . Define  $f: V \rightarrow \mathbb{F}$  by  $f(v_i) := g(v_i)$  for  $1 \leq i \leq n$ ,  $f(v_{n+1}) = a$ , and  $f(v_i) = 0$  otherwise (for  $n+2 \leq i \leq m$ ). Then  $f(v_{n+1}) = a$  (the  $v_{n+1}$ 's existence is assured by  $W$  being a proper subspace of  $V$ , and  $f(x) = g(x)$  for all  $x \in W$  is clear.  $\checkmark$

(b) Let  $g(x)$  be the zero transformation from  $W$  into  $\mathbb{F}$ . Then set  $a := 1$  and define  $f$  and the relevant terms as above, in (a). It is clear that  $f(x) = 0$  for all  $x \in W$  but  $f(v_{n+1}) = 1$  tells us  $f$  is nonzero.  $\checkmark$

20. (a) If  $T$  is surjective, suppose  $T^t(f) = T^t(g)$ . Then for every  $w \in W$ , there exists  $v \in V$  so that  $f(w) = fT(v) = [T^t(f)](v) = [T^t(g)](v) = gT(v) = g(w)$ . Hence,  $f = g$  which informs us that  $T^t$  is injective. Now consider  $T$  not being surjective, i.e. there exists  $y \in W$  with  $y \notin T[V]$ . So, letting  $\beta := \{y, w_1, w_2, \dots, w_k\}$  be a basis for  $W$  (aligned by Exercise 7 of (1.7)), we can define  $f(y) = 1$  and  $f(w_i) = 0$  for all  $i \in \{1, \dots, k\}$ . Thus, for the zero linear functional  $g_0: W \rightarrow \mathbb{F}$ ,  $fT(v) = g_0T(v) = 0$  (a linear combination of vectors in  $\beta$ ) for any  $v \in V$ , even though  $f \neq g_0$  by virtue of  $f(y) = 1 \neq 0 = g_0(y)$ . Therefore, the biconditional must hold.  $\square$



17. Idea,  
W is T-invariant

First  
 $T: V \rightarrow V$ ,  $W \subseteq V$   $T^t: V^* \rightarrow V^*$   
 i.e.  $T[W] = W$ , show  $T^t[W^0] = W^0$

Let  $g \in W^0$ , i.e.  $g: V^* \rightarrow \mathbb{F}$  where  $g(w) = 0 \ \forall w \in W$ , find  $T^t(f) = fT = g$

$$g := \sum_{i=1}^m g(w_i) f_i = \sum_{i=1}^m g(w_i) f_i$$

$$f := \sum_{i=1}^m c_i f_i$$

$$fT := \sum_{i=1}^m c_i f_i T$$

$$c_i f_i(T(w_i)) = g(w_i)$$

$$c_i c_i = g(w_i)$$

$$c_i = \frac{g(w_i)}{f_i(T(w_i))}$$

must be non-zero!

since  $W = \mathcal{R}(T|_W) = \text{span } T[W]$

$$W \xrightarrow{T} [W]$$

$$W^0 \xrightarrow{T^t} [W^0]$$

$\Rightarrow$  Deps mixed up the dots of T-invariance  
 Show  $T^t(f) \in W^0$ . Let  $f \in W^0$ ,  $T^t(f) = fT$   
 $\forall w \in W$   $fT(w) = 0$  as  $T(w) \in W$

$W^0$  T<sup>t</sup>-invariant

$$fT(w) = 0, \quad f := \sum_{i=1}^m f_i$$

$\Rightarrow T(w) \in W$  as if  $T(w) \notin W$  for one  $w$ , then  $f(T(w)) \neq 0$

Proof

Assume  $W$  is T-invariant and let  $f \in W^0$ . Then for each  $w \in W$ ; since  $T(w) \in W$ ,  $[T^t(f)](w) = fT(w) = 0$ . Hence,  $T^t(f) \in W^0$  and  $W^0$  is T<sup>t</sup>-invariant. Now let  $\gamma = \{\gamma_1, \gamma_2, \dots, \gamma_m\}$  be a basis for  $W$ ,  $\beta = \{\nu_1, \nu_2, \dots, \nu_m\}$  its extension to a basis for  $V$ , and  $\beta^* = \{f_1, f_2, \dots, f_m\}$  the dual basis of  $\beta$ .  
 Supposing  $W^0$  is T<sup>t</sup>-invariant, we can define  $f := \sum_{i=1}^m f_i$ . (Given  $w \in W$ ,  $fT(w) = 0$  so this means  $T(w)$  is a linear combination of  $\gamma$ . Thus,  $T(w) \in W$  and  $W$  is T-invariant. Therefore, the biconditional holds.

[The case where  $V$  and/or  $W$  are infinite dimensional follows similarly:  $f_\alpha T(w) = 0$  for  $k \leq \alpha \leq \lambda$ , where  $\gamma := \{\nu_\alpha | \alpha \in K\}$  and  $\beta := \{\nu_\alpha | \alpha \in \lambda\}$ ]

$$u_1 u_2 u_3 = u_1 u_3 u_2 = u_2 u_1 u_3 = u_3 u_2 u_1 = u_3 u_1 u_2$$

# 9. Ideas

when  $n=1$ , clear

Proof

Lemma: Order Invariance

We want to first prove that for any bijection  $q: \{1, 2, \dots, n\} \rightarrow \{1, 2, \dots, n\}$ , we have  $u_1 u_2 \dots u_n = u_{q(1)} u_{q(2)} \dots u_{q(n)}$ .  
 When  $n=1$ , the result follows immediately. Assume it holds for  $n \in \mathbb{N}$ . Then, for any bijection  $q: \{1, 2, \dots, n+1\} \rightarrow \{1, 2, \dots, n+1\}$ , with  $q(k) = k+1$  for some  $k$ , define the bijection  $p: \{1, 2, \dots, n\} \rightarrow q[\{1, 2, \dots, n\}]$  by having  $p(i)$  be the least  $q(i)$ , and  $p(n+1)$  be the least  $q(i) > p(n)$ , where  $1 \leq k \leq n$ . For the commutative linear operator on  $V$  given by  $T_i := U_{p(i)}, U_{q(1)} \dots U_{q(n)} = T_{p^{-1}(q(1))} \dots T_{p^{-1}(q(n))} = T_1 \dots T_n$  by the above assumption since  $p^{-1}q$  is a bijection from  $\{1, 2, \dots, n\}$  to  $\{1, 2, \dots, n\}$ . Thus, we now have that  $U_{q(1)} \dots U_{q(n+1)} = U_1 U_2 \dots U_{q(n+1)} U_{q(n+2)} \dots U_{q(n+1)} U_{n+1}$  by perewise commutativity. If  $q(n+1) = q(n)$ , the result is clear. Otherwise, repeat the above reasoning to obtain  $U_{q(1)} \dots U_{q(n+1)} = U_1 \dots U_{n+1}$ . So, the  $(n+1)$ th case must hold true. Accordingly, it is true for all  $n \in \mathbb{N}$  by induction.  $\square$

[Implicit assumption of general associativity in the qns and lemma above but that's straightforward.]

Consequently, when  $U_i(v) = 0_V$  for any  $v \in V$ ,  $(U_1 \dots U_n)(v) = (U_1 U_2 \dots U_{i-1} U_{i+1} U_{i+2} \dots U_n) U_i(v) = 0_V$  as expected. So, the claim that  $N(U_i) \subseteq N(U_1 U_2 \dots U_n)$  certainly is true.  $\square$

$$f_1(x_1, x_2) = x_1 + x_2$$

$$f_{n+1}(x_1, \dots, x_{n+1}) = f_n(x_1, \dots, x_n) + x_{n+1}$$

Self-check of Theorem 2.29

$$y''(t) = D(e^{ct}) = D(e^{at+ibt}) = D(e^{at}[\cos(bt) + i\sin(bt)]) = ae^{at}[\cos(bt) + i\sin(bt)] + e^{at}[-b\sin(bt) + ib\cos(bt)]$$

$$= e^{at}[a\cos(bt) - b\sin(bt) + ia\sin(bt) + ib\cos(bt)]$$

$$= (a+bi)e^{at}[\cos(bt) + i\sin(bt)]$$

$$= ce^{ct}$$

Self-check of Theorem 2.30

Idea

AP:  $p(t) = t + a_0 = 0$

$$y'' = -a_0 y' \quad y' = -a_0 y$$

$$= a_0^2 y$$

$$\Rightarrow -a_0 y' = a_0^2 y$$

$$y' = -a_0 y$$

$$\int -\frac{1}{a_0 y} y' dt = \int 1 dt$$

$$\int -\frac{1}{a_0 y} dy = t + c_0$$

$$-\frac{1}{a_0} \ln|y| = t + c_0$$

$$\ln|y| = -a_0 t + c_1$$

$$y = e^{-a_0 t + c_1}$$

$$= C e^{-a_0 t} \quad y' = -(a_0 e^{-a_0 t})$$

$$-(a_0 e^{-a_0 t}) + (a_0 e^{-a_0 t}) = 0$$

Proof  
 Notice  $y'' = -a_0 y' = -a_0(-a_0 y) = a_0^2 y$  so  $y' = -a_0 y$ . Thus, integrating both sides, we have  $\int -\frac{1}{a_0 y} y' dt = \int -\frac{1}{a_0 y} dy = \int 1 dt$ . which simplifies to  $y = C e^{-a_0 t}$ , for any real constant  $C \in \mathbb{R}$ . Conversely, if  $y = C e^{-a_0 t}$  for some  $C \in \mathbb{R}$ ,  $y' = -(a_0 C e^{-a_0 t})$  so  $y' + a_0 y = 0$  as expected. Hence, it is clear that  $\{e^{-a_0 t}\}$  is a basis and the solution space is of dimension 1.

Unfortunately only proves the result for functions  $\mathbb{R} \rightarrow \mathbb{R}$ . While we need it for functions  $\mathbb{R} \rightarrow \mathbb{C}$ .

(checking commutativity of operators  $D - c_i I$ : As expected,

$$(D - c_i I)(D - c_j I) = D^2 - D(c_j I) - (c_i I)D + (c_i I)(c_j I) = D^2 - D(c_j I) - D(c_i I) + (c_i I)(c_j I) = (D - c_i I)(D - c_j I).$$

**Self-Proof of Theorem 2.32** Let  $N_j$  be the number of times  $D - c_j I$  is repeated in the product  $(D - c_1 I) \cdots (D - c_n I)$

Idea:  
When  $n=1$ , clear

Assume true for  $n \in \mathbb{N}$ .

$$\underbrace{(D - c_1 I) \cdots (D - c_n I)}_{D_n} (D - c_{n+1} I)(y) = y_0$$

$$D \underbrace{(D - c_1 I) \cdots (D - c_n I)}_Z (y) = c_{n+1} \underbrace{(D - c_1 I) \cdots (D - c_n I)}_Z (y)$$

$$D(Z) = c_{n+1} Z$$

$$Z' - c_{n+1} Z = 0$$

$$Z = A e^{c_{n+1} t} \quad \text{for some } A \in \mathbb{C}$$

$$\Rightarrow \underbrace{(D - c_1 I) \cdots (D - c_n I)}_{r(D)} (y) = A e^{c_{n+1} t} \quad (\star)$$

$$\sum_{j=1}^k \sum_{i=0}^{m_j-1} a_{ij} t^i e^{c_j t}$$

Let  $u_p = \alpha e^{c_{n+1} t}$ ,  $u_p' = \alpha c_{n+1} e^{c_{n+1} t}$   
 $\alpha(c_{n+1} - c_n) e^{c_{n+1} t} = A e^{c_{n+1} t}$

let  $C := \prod_{i=1}^n \frac{A}{c_{n+1} - c_i}$

Just 1 possible soln

$$t^{m_1-1} e^{c_1 t} + a_1 e^{c_1 t} + a_2 t e^{c_2 t} = e^{c_1 t}$$

$$a_1 e^{c_1 t} + a_2 t e^{c_2 t} = e^{c_1 t}$$

$$a_1 + a_2 = 1$$

$$a_{ij} e^{c_j}$$

(claim: The solution space of  $q(D)(y)$  is  $\text{span}\{e^{c_i t}\}$  (if  $c_{n+1}$  is different from each  $c_i$ ,  $1 \leq i \leq n$ )

Suppose  $f$  is a solution, i.e.  $(D - c_1 I) \cdots (D - c_n I)(f) = 0$ . Then  $f + (e^{c_{n+1} t})$  must also be a soln. Hence,  $(D - c_1 I) \cdots (D - c_n I)(f + (e^{c_{n+1} t})) = 0$ . By assumption,  $f$  is a linear combination of members of  $\mathcal{S}$ .

$$(D - c_j I)(D - c_i I)(e^{c_i t}) = (D - c_i I)(D - c_i I)(e^{c_i t}) = (c_i - c_i) e^{c_i t} = 0$$

$$(D - c_i I)(e^{c_i t}) = c_i e^{c_i t} - c_i e^{c_i t} = 0$$

$$(D - c_i I)(t^{m_i-1} e^{c_i t}) = (m_i - 1) t^{m_i-2} e^{c_i t} + c_i t^{m_i-1} e^{c_i t} - c_i t^{m_i-1} e^{c_i t} = [(m_i - 1) t^{m_i-2} + (c_i - c_i)] e^{c_i t}$$

$$= [(m_i - 1) t^{m_i-2} + (c_i - c_i)] e^{c_i t}$$

$c_k$  repeated roots:

$$(D - c_1 I)^{m_1} \cdots (D - c_k I)^{m_k} = 0$$

$$(D - c_1 I)^{m_1-1} \cdots (D - c_k I)^{m_k} (y) = A e^{c_k t} \quad (\star)$$

If all repeated,

$$(D - c I)^{n+1}(y) = 0$$

$$(D - c I)(t^n e^{ct}) = n t^{n-1} e^{ct} + c t^n e^{ct} - c t^n e^{ct} = n t^{n-1} e^{ct}$$

$$(D - c I)^n (t^n e^{ct}) = n! e^{ct}$$

Suppose  $g$  is a soln. By  $(\star)$ ,  $(D - c I)^{n+1}(t^n e^{ct}) = 0$ . Again,  $g + t^n e^{ct}$  should be a soln.

$$(D - c I)(t^{m_i-1} e^{c_i t}) = (m_i - 1) t^{m_i-2} e^{c_i t} + c_i t^{m_i-1} e^{c_i t} - c_i t^{m_i-1} e^{c_i t}$$

## Self-Proof of Lemma 2

Idea:  $U$  not necessarily injective

When  $TU(v) = 0$ , either

$$U(v) = 0$$

$$v \in N(U)$$

$v$  linear comb of vectors  $v_i$

or  $U(v) \neq 0$

$$0 \neq U(v) \in N(T) \subseteq R(U) = V$$

$U(v)$  is a linear comb of vectors  $v_j$

$$\text{OR } \beta := \{v_i \mid \alpha \leq i \leq \alpha+n\}$$

$$U(v_i) = 0 \quad \alpha \leq i \leq \alpha+n$$

$$T(v_j) = 0 \quad \beta \leq j \leq \beta+m$$

$$U' := U|_{N(TU)}$$

$$N(TU) \subseteq R(U)$$

$$R(U') = \{U(v) \in V \mid TU(v) = 0\}$$

$$= \cancel{U[N(TU)]}$$

$$= N(T)$$

by surjectivity of  $U$

$$\gamma := \{u_1, u_2, \dots, u_n\}$$

$$S := \{v \in V \mid TU(v) = 0\}$$

$$= N(TU)$$

$$\text{rank}(U') + \text{nullity}(U') = \dim(U')$$

$$\text{nullity}(U') = \text{nullity}(U)$$

## Proof

As usual,  $U|_{N(TU)}$  denotes the restriction of  $U$  to  $N(TU)$ : We notice  $\text{rank}(U|_{N(TU)}) + \text{nullity}(U|_{N(TU)}) = \dim(N(TU))$ . Where  $R(U|_{N(TU)}) :=$

$\{U(v) \in V \mid TU(v) = 0\} = N(T)$  by  $U$ 's surjectivity so that  $\text{rank}(U|_{N(TU)}) = \text{nullity}(T)$ , and  $\text{nullity}(U|_{N(TU)}) = \text{nullity}(U)$  is clear.

Hence,  $\text{nullity}(TU) = \text{nullity}(T) + \text{nullity}(U)$  as expected.

Self-Proofs of Theorem 2.33 & 2.34 (including its lemma)

I accidentally proved both of these in my self-proof of Theorem 2.32 (a).

# Self-Proof of Theorem 2.32

Let  $m_j$  be the number of times  $D - c_j I$  is repeated in  $p(D)$ . We first claim that the set  $S_n := \{t^{ij} e^{c_j t} \in C^\infty \mid 1 \leq j \leq n \text{ \& \ } 1 \leq ij \leq m_j - 1\}$  is a basis for the nullspace of any  $n$ th order differential operator  $p(D) = (D - c_1 I)(D - c_2 I) \dots (D - c_n I)$ . When  $n=1$ , this is just Theorem 2.30. So, assume that this is true for any differential operator  $p(D)$  of order  $n-1$ . Then, for any differential operator  $p(D)$  of order  $n$ , suppose that it has some repeated roots. That is,  $p(D) = (D - c_1 I)^{m_1} (D - c_2 I)^{m_2} \dots (D - c_k I)^{m_k}$  for some naturals  $m_j$ . For  $p(D)(y) = 0$ , it simplifies to  $z' - c_1 z = 0$  by having  $z := (D - c_1 I)^{m_1 - 1} \dots (D - c_k I)^{m_k} (y)$ . Therefore, by Theorem 2.30,  $z = A e^{c_1 t}$  —  $(\star)$  for some  $A \in C$ .

Notice  $(D - c_1 I)(t^{m_1 - 1} e^{c_1 t}) = (m_1 - 1)t^{m_1 - 2} e^{c_1 t}$ . By repetition,  $(D - c_1 I)^{m_1 - 1}(t^{m_1 - 1} e^{c_1 t}) = (m_1 - 1)! e^{c_1 t}$ . Continuing,  $(D - c_j I)(t^{m_j - 1} e^{c_j t}) = (m_j - 1)! (c_1 - c_j) e^{c_j t}$ . Again repeating this,  $(D - c_1 I)^{m_1 - 1} (D - c_2 I)^{m_2} \dots (D - c_k I)^{m_k} (t^{m_1 - 1} e^{c_1 t}) = C e^{c_1 t}$ , where we define  $C := (m_1 - 1)! \prod_{j=1}^k (c_1 - c_j)^{m_j}$  for convenience. Hence,  $t^{m_1 - 1} e^{c_1 t}$  is a solution to  $(\star)$ . Furthermore, since  $C e^{c_1 \cdot 0} = C \neq 0$  in the case that  $t=0$ , it is certainly not the zero function. As such,  $t^{m_1 - 1} e^{c_1 t}$  cannot be expressed as a linear combination of functions in  $S_n$ , implying the linear independence of  $S_{n+1}$ .

Presume  $f$  is a solution to  $p(D)(y) = 0$ . Then,  $f$  satisfies  $(\star)$  for some value of  $A \in C$ , and so does  $\frac{A}{C} t^{m_1 - 1} e^{c_1 t}$  for the same value of  $A$  by the above result. Consequently,  $(D - c_1 I)^{m_1 - 1} (D - c_2 I)^{m_2} \dots (D - c_k I)^{m_k} (f - \frac{A}{C} t^{m_1 - 1} e^{c_1 t}) = 0$ . By our initial assumption,  $f - \frac{A}{C} t^{m_1 - 1} e^{c_1 t}$  is a linear combination of functions in  $S_n$ . Accordingly,  $f$  is a linear combination of functions in  $S_{n+1}$ . That is to say,  $\text{span}(S_{n+1}) = N(p(D))$ . Now,  $S_{n+1}$  is a basis for  $N(p(D))$ .

In other words, the initial claim is true of  $n+1$  too. Therefore, it is true for each  $n \in \mathbb{N}$  by induction.

Exercises

- 1. (a) True ✓
- (b) True ✓
- (c) False ✓
- (d) True x False "Any" solution!
- (e) True ✓
- (f) False (it is not given that  $p(t)$  is of degree  $k$ )
- (g) True ✓

$$\begin{aligned}
 a &= e \\
 ae^c &= e^e \\
 e^c &= e^{e-1} \\
 ae^{2c} &= e^{e^2} \\
 e^{2c} &= e^{e^2-1}
 \end{aligned}$$

2. (a) <sup>The statement is false.</sup> Counterexample: We know  $\text{span}\{e^{e^t}\}$  is a subspace of  $C^\infty$ , yet for any homogeneous linear differential equation of order 1, i.e.  $(D-cI)(y) = 0$  for some  $c \in \mathbb{C}$ , Theorem 2.30 says the solution space has basis  $e^{ct}$ . Since for any  $a, c \in \mathbb{C}$ :  $ae^0 = e$ ,  $ae^c = e^e$ , and  $ae^{2c} = e$  would imply  $e^c = e^{e-1}$  and  $e^{2c} = e^{e^2-1}$ , a contradiction as  $2e-1 \neq e^2-1$ , this means  $\text{span}\{e^{e^t}\}$  is the solution space of no homogeneous linear differential equation of order 1, or in fact of any finite order as  $\text{span}\{e^{e^t}\}$  is of dimension 1.

~~(b) This is true: simply take the differential equation  $D^3(y) = 0$ . Then Theorem 2.34 says it has basis  $\{$~~

(b) This is also false as Theorem 2.34 tells us that, for any homogeneous linear differential equation, if  $\{t, t^2\}$  is included in its solution space, so must  $\{1\}$

(c) Contrastingly, this is true. Consider any homogeneous linear differential equation  $p(D)y = 0$ . We notice  $p(D)(y') = p(D)(Dy) = D(p(D)(y)) = D(0) = 0$  which means  $y'$  is also a solution.

(d) Indeed, we see that  $p(D)q(D)(x+y) = p(D)q(D)(x) + p(D)q(D)(y) = q(D)(p(D)(x)) + p(D)(q(D)(y)) = 0$ . Hence, the statement must be true.

(e) Ideas / Sketch

$$(D-I)(D+I)(1) = (D^2 - I^2)(1) = D^2(1) - 1^2 = -1 \neq 0$$

$$\begin{matrix}
 e^t & e^{-t} \\
 e^{t-t} = 1
 \end{matrix}$$

Proof  
 This is false: case in point: consider  $p(D) := D-I$ ,  $q(D) := D+I$ , and naturally,  $x := e^t$  and  $y := e^{-t}$ . Then,  $p(D)q(D)(xy) = (D^2 - I^2)(e^t e^{-t}) = D^2(1) - I^2(1) = -1 \neq 0$ .



3. (e) The auxiliary polynomial here is  $t^3 - t^2 + 3t + 5 = (t+1)(t-(1+2i))(t-(1-2i))$ . Thus, it has roots  $t = -1, 1+2i, 1-2i$ .

Accordingly, our differential equation has basis  $\{e^{-t}, e^{(1+2i)t}, e^{(1-2i)t}\}$ . OR:  $\{e^{-t}, e^{t \cos(2t)}, e^{t \sin(2t)}\}$

4. (c) The corresponding differential equation has auxiliary polynomial  $t^3 + 6t^2 + 8t = t(t^2 + 6t + 8) = t(t+4)(t+2)$ , with roots  $t = 0, 2, 4$ . As such,  $N(D^3 + 6D^2 + 8D)$  has basis  $\{1, e^{2t}, e^{-4t}\}$ .

7. Ideas  $\frac{1}{2}e^{0i} \quad \frac{1}{2}e^{\frac{i\pi}{2}}$   
 $\frac{1}{2}(x+iy) \quad \frac{1}{2i}(x-y)$   
 $x+y \quad \frac{1}{i}(x-y) = -i(x-y)$   
 $= i(-x+y)$

$x \cdot i = 1$   
 $a+bi = \alpha(x+y) + i\beta(-x+y)$   
 $\alpha = \frac{a}{x+y} \quad \beta =$   
 $(x+y \neq 0 \text{ lest not basis})$

Proof  
 Let  $a+bi \in \mathbb{C}$ . So,  $\frac{2a}{x+y} \cdot \frac{1}{2}(x+y) + \frac{2b}{-x+y} \cdot \frac{1}{2i}(x-y) = a+bi$  because  $x$  and  $y$  are distinct and nonzero by the fact that  $\{x, y\}$  is a basis.

8. Ideas  
 $e^{(a+ib)t}$   
 $= e^{at} [\cos(b) + i \sin(b)]$   
 $= e^{at} \cos(b)$   
 $= e^{(a-ib)t}$   
 $= e^{at} [\cos(-b) + i \sin(-b)]$   
 $= e^{at} [\cos(b) - i \sin(b)]$

Proof  
 This follows easily from 7.

10. & 11. See the self-proofs of these Theorems.

$t+1 \mid t^3 - t^2 + 3t + 5$   
 $-(t^2 + t)$   
 $-2t^2 + 2t + 5$   
 $-(2t^2 - 2t)$   
 $4t + 5$

□  
 □  
 □



Index... this is true... 2.34... equation... the solut... de c... different...

12. Ideas  
 $V := N(p(D))$

$$g(t)h(t) = h(t)g(t)$$

$$y \in V: h(D)g(D)y = g(D)h(D)y = 0$$

$$\Rightarrow g(D)y \in N(h(D))$$

$$g(D)(V) \subseteq N(h(D))$$

$$h(D)y = 0$$

$$\Rightarrow h(D)g(D)y = g(D)h(D)y = 0$$

Proof

We see that  $h(D)g(D)[V] = g(D)h(D)[V] := p(D)[V] = \{0\}$  so  $g(D)[V] \subseteq N(h(D))$  is certain. Now, we claim that  $N(h(D)) = g(D)[V]$ . To show this, consider any basis vector  $t^k e^{ct}$  of  $N(h(D))$ . As in the self-proof of Theorem 2.32, we first notice  $(D-cI)^m(t^{n+m}e^{ct}) = A t^n e^{ct}$  for a constant  $A \in \mathbb{C}$  (namely,  $\frac{(n+m)!}{n!}$ ), and by setting  $x := (-1)^{n-k} \frac{(n-k)!}{(k-j)!} t^{k-j} e^{ct}$  for  $j \leq k$  that meets the criteria of  $g(D)x = t^k e^{ct}$

$$(D-b_j I) \left( \sum_{i=0}^k \alpha_{ki} t^i e^{ct} \right) = t^k e^{ct}$$

namely  $x := A^{-1} \sum_{i=0}^k \sum_{j=0}^i \sum_{l=0}^j \dots \sum_{r=0}^{i_1} \alpha_{ki_1} \alpha_{i_1 i_2} \alpha_{i_2 i_3} \dots \alpha_{i_{r-1} i_r} t^{i_r} e^{ct}$ , where  $k := n+m$  here and  $r$  is the number of roots  $b_j \neq c$  in  $p(t)$  and  $n$  the number of repeated roots  $b_j = c$  in  $p(t)$ . As such,  $N(h(D)) \subseteq g(D)[V]$ , indeed holds.

□

Max ideaation 1d1

consequently,  $N(h(D)) = g(D)[V]$ .

$$N(h(D)) \subseteq V \subseteq R(g(D))$$

$$A^{-1} \sum_{i=0}^k \sum_{j=0}^i \sum_{l=0}^j \dots \sum_{r=0}^{i_1} \alpha_{ki_1} \alpha_{i_1 i_2} \alpha_{i_2 i_3} \dots \alpha_{i_{r-1} i_r} t^{i_r} e^{ct}$$

$$t^n e^{ct}$$

$$(D-cI)(t^n e^{ct}) = t^n e^{ct}$$

$$(D-cI)(t^{n+m} e^{ct}) = (n+m)t^n e^{ct} + (n+m-1)t^{n+m-1} e^{ct} - t^{n+m} e^{ct}$$

$$(D-c_1 I)(x_i t^i e^{ct})$$

$$= i x_i t^{i-1} e^{ct} + x_i (c-c_1) t^i e^{ct}$$

$$x_n (c-c_1) t^n e^{ct} = t^n e^{ct}$$

$$x_n = (c-c_1)^{-1}$$

$$d = i \geq n-1:$$

$$\alpha_{i+1} t^i e^{ct} + \alpha_i (c-c_1) t^i e^{ct} = 0$$

$$(i+1)\alpha_{i+1} + \alpha_i (c-c_1) = 0$$

$$n x_n + \alpha_{n-1} (c-c_1) = 0$$

$$1 + \alpha_{n-1} (c-c_1) = 0$$

$$\alpha_i = \frac{-\alpha_{i+1}}{c-c_1}$$

$$\alpha_{n-1} = \frac{-1(c-c_1)^{-1}}{(c-c_1)} = -1(c-c_1)^{-2}$$

$$\alpha_{n-2} = -1(n-1)(c-c_1)^{-3}$$

conjecture:  $i = n - (n-i) \binom{n}{i}$

$$\alpha_i = (-1)^{n-i} \frac{n!}{i!} (c-c_1)^{-(n-i)}$$

checked with Wolfram

$$t^i e^{ct}$$

$$\downarrow$$

$$d_{i+1}$$

$$\downarrow$$

$$\sum \alpha_{i+1} \alpha_{ij}$$

13. (a) Idea:

$$(D - cI)(y) = x$$

$$y' - cy = x$$

$$(D - cI)(\alpha x + \beta x') = \alpha x^2 - c^2 x + \beta x'' - c\beta x'$$

$$(D - cI)(\alpha x + \beta x' + \gamma x'')$$

$$= \alpha x' - c^2 x$$

$$+ \beta x'' - c^2 \beta x'$$

$$+ \gamma x''' - c\gamma x''$$

$$(D - cI)\left(\sum_{i=0}^{\infty} c^{-i-1} x^{(i)}\right) = \sum_{i=0}^{\infty} c^{-i-1} x^{(i+1)} - c^{-i} x^{(i)}$$

$$(D - cI)(c^{-i} x^{(i)}) = c^{-i} x^{(i+1)} - x^{(i)}$$

$$= -c^{-i} x^{(i)}$$

$$= x$$

$$y' - cy = x^2$$

$$2\alpha t + \beta - (2\alpha t^2 - c\beta t - c\gamma) = t^2$$

$$-\alpha = 1$$

$$\alpha = -\frac{1}{c} = -c^{-1}$$

$$2\alpha - c\beta = 0$$

$$-2c^{-1} - c\beta = 0$$

$$-2 - c^2\beta = 0$$

$$\beta = -2c^{-2}$$

$$\beta - c\gamma = 0$$

$$\gamma = \beta c^{-1} = -2c^{-3}$$

For each  $i \geq 0$ ,  $c^{-i-1} x^{(i+1)} - c^{-i} x^{(i)} = 0$

$$y^{(n)} + a_{n-1} y^{(n-1)} + \dots + a_1 y' + a_0 y = x$$

$$(D - c_1 I)(D - c_2 I) \dots (D - c_n I) = x$$

Proof

We first focus on the case of a 1st order homogeneous <sup>linear</sup> differential equation: Notice that since  $c^{-i-1} x^{(i+1)} - c^{-i} x^{(i)} = 0$  for  $i \geq 0$  (where square brackets are used as normal brackets) to generalize this to all (homogeneous) linear differential equations of order any  $n \in \mathbb{N}$ .

(b) Suppose  $f$  is any solution to the nonhomogeneous linear differential equation  $p(D)y = x$ . Then, taking the difference of the equations, we have that  $p(D)(f - z) = 0$ . So,  $f = z + (f - z)$ , where  $f - z \in V$  follows from the previous line. That is, the set of all solutions to the nonhomogeneous equation is a subset of  $V$ .

Conversely,  $p(D)(z + y) = x + 0 = x$  as expected. Therefore, equality holds.



14. Ideas

$$\begin{aligned}
 & a e^{ct} \\
 & a e^0 = 0 \\
 & a = 0 \\
 & a e^{ct} + b e^{ct} \\
 & a c e^{ct} + b c e^{ct} + b t e^{ct} \\
 & (a+c)b e^{ct} + b t e^{ct} \\
 & a = 0 \quad a+c = 0 \\
 & \quad \quad \quad b = 0
 \end{aligned}$$

$$\begin{aligned}
 & a e^{ct} + b e^{ct} \\
 & a c_1 e^{ct} + b c_2 e^{ct} \\
 & a + b = 0 \quad a c_1 + b c_2 = 0 \\
 & b = -a \quad a(c_1 - c_2) = 0 \\
 & \quad \quad \quad a = b = 0
 \end{aligned}$$

$$\begin{aligned}
 p(D)(x) &= p(D)(a t^k e^{ct}) \\
 &= (e^{ct}) \text{ for some } c \in \mathbb{C} \\
 &= a k! (c-c_1)(c-c_2)\dots(c-c_n)
 \end{aligned}$$

$$a k t^{k-1} e^{ct} + a(c-c_1) t^k e^{ct} = a k t^{k-1} e^{ct}$$

$n=1$  trivial  
Assume true for  $n \in \mathbb{N}$

$$\begin{aligned}
 a t^k e^{ct} &= \sum_{i=0}^n a_i b_i, \quad k \leq n \\
 a k! &= 0 \\
 a &= 0
 \end{aligned}$$

by assumption &  $e^{c \cdot 0} = 1$

$$\begin{aligned}
 (D-cI)(a t^k e^{ct}) &= a k t^{k-1} e^{ct} + a c t^k e^{ct} - a c t^k e^{ct} = a k t^{k-1} e^{ct} \\
 (D-cI)(a t^k e^{ct}) &= a(c-c_1) e^{ct} \\
 (D-cI)(b t^k) &= a t^{(n)} - c t^k = 0 = a k t^{k-1} e^{ct}, \quad t=0
 \end{aligned}$$

Proof

When  $n=1$ , this follows from Theorem 2.30. So assume this is true for  $n \in \mathbb{N}$  and  $p(D)(y) = 0$  is an  $(n+1)$ th order homogeneous differential equation in which any solution  $x$  is such that  $x(t_0) = x'(t_0) = \dots = x^{(n)}(t_0) = 0$ . Then there exists  $q(D)$  with  $p(D) = (D-cI)q(D)$  for some  $c \in \mathbb{C}$ . Let  $\{b_i | 1 \leq i \leq n\}$  be the basis, as specified in Theorem 2.34, for the solution space of  $q(D)(y) = 0$ . For some  $k \leq n$ ,  $\{b_i | 1 \leq i \leq n\} \cup \{t^k e^{ct}\}$  is a basis for the solution space of  $p(D)(y) = 0$ . Given any solution  $x = \sum_{i=1}^n a_i b_i + a t^k e^{ct}$  as specified in our assumption, we notice  $0 = [(D-cI)^k(x)](t_0) = [(D-cI)^k(\sum_{i=1}^n a_i b_i)](t_0) + a k! e^{ct_0} = a k! e^{ct_0}$  by assumption. As such,  $a = 0$  by virtue of  $k! \neq 0$ . Therefore, the statement is true for every  $p(D)(y) = 0$  of order  $n+1$  too. By induction, this is true for all  $n \in \mathbb{N}$ .

Since the case of  $n=1$  follows from Theorem 2.30, assume this is true for  $n \in \mathbb{N}$  and that  $p(D)(y) = 0$  is an  $(n+1)$ th order homogeneous linear differential equation, with  $p(D) = (D-cI)q(D)$ ;  $\{b_i | 1 \leq i \leq n\}$  and  $\{b_i | 1 \leq i \leq n\} \cup \{t^k e^{ct}\}$  the basis for the solution spaces of  $q(D)(y) = 0$  and  $p(D)(y) = 0$ , as specified in Theorem 2.34. For any solution  $x$  of  $p(D)(x) = 0$  such that  $x(t_0) = x'(t_0) = \dots = x^{(n)}(t_0) = 0$ ,  $q(D)(x) = q(D)(\sum_{i=1}^n a_i b_i + a t^k e^{ct}) = (e^{ct})$  where  $c := a k! (c-c_1)(c-c_2)\dots(c-c_{n-k})$  for the roots  $c_i \neq c$  of  $q(t)$ . So, by construction,  $(e^{ct_0}) = 0$  so  $c = 0$ . Hence,  $a = 0$ . Accordingly, each  $a_i = 0$  by assumption, since  $a = 0$  means  $x$  is a solution to  $q(D)(x) = 0$  with  $x(t_0) = x'(t_0) = \dots = x^{(n-1)}(t_0) = 0$ . Thus, the statement is true for  $p(D)(y) = 0$  of order  $n+1$  too. By induction, this is true for all  $n \in \mathbb{N}$ . □

5. (a) Linearity is clear from definition while the second claim follows from exercise 14. Since  $N(\Phi) = 0$ ,  $\Phi$  is homeomorphism according to Theorem 2.5 because  $\dim(V) = \dim(\mathbb{C}^n) = n$  is already known.

(b) By (a), the required unique  $x \in V$  is given by  $\Phi^{-1} \begin{pmatrix} x(t_0) \\ x'(t_0) \\ \vdots \\ x^{(n-1)}(t_0) \end{pmatrix}$ .

9. Ideas / Thought Organisation

1. More generality = better, obviously.  $p^{(0)}(y) = 0$
2. The space of real solutions,  $S_{\mathbb{R}}$ , is a subspace of the solution space  $V \subseteq \mathbb{C}^{\infty}$  for any HLDE, so it is generated by the same basis  $\beta$  for  $V$  as in Theorem 2.34. In fact, letting  $c_1, c_2, \dots, c_k$  be roots of  $p(t)$  whose conjugates are also roots of  $p(t)$ , and  $N(c_i)$  the number of times  $\pm c_i$  is repeated in  $p(t)$ , the set  $\{t^{i_j} e^{t \operatorname{Re}(c_i)} (\cos(t \operatorname{Im}(c_i)) + t^{i_j} e^{t \operatorname{Re}(c_i)} \sin(t \operatorname{Im}(c_i))) \mid 1 \leq k \neq 0 \leq i_j \leq \min\{N(c_i), N(\bar{c}_i)\} - 1\}$  is a basis for  $S_{\mathbb{R}}$ .
3. Moreover, if the coefficients of the HLDE are real, all conjugate roots are also roots (of  $p(t)$ ) and occur the same number of times.   
 which is the case for physical motion

### Self-Proof of Theorem 3.13

Suppose  $x$  is a solution to  $(CA)x = Cb$ . Then  $Ax = C^{-1}(CA)x = C^{-1}(Cb) = b$ . Conversely, when  $x$  is a solution to  $Ax = b$ ,  $(CA)x = C(Ax) = Cb$  as expected. Thus, the biconditional holds. □

### Self-Proof of Corollary

Assume that  $(A'|b')$  is obtained from  $(A|b)$  by a finite number of elementary row operations. Let  $E_1, E_2, \dots, E_n$  be the associated elementary matrices and define  $M := E_1 E_2 \dots E_n$ , an invertible matrix. So, Exercise 15 of section 3.2 says  $(A'|b') = (MA|Mb)$ . That is,  $A' = MA$  and  $b' = Mb$ . Therefore, since the system  $(MA)x = Mb$  is equivalent to  $Ax = b$  by the above theorem, the system  $A'x = b'$  must be equivalent to  $Ax = b$ . □

### Self-Proof of Theorem 3.14



Let  $A$  be any  $n \times n$  matrix. Exercise 11 of section 3.1 says that we can transform  $A$  into an upper triangular matrix  $A''$  via elementary row operations. If  $n=1$ , the matrix  $A$  is already in its reduced row echelon form,  $A''$ . So assume the result is true of  $n \in \mathbb{N}$ . Define  $A''$  via elementary row operations. □

### Self-Proof of Theorem 3.15

(a) Let  $a_j$  be the  $j$ th column of  $A$  / of  $(A|b)$  and  $\lambda_i$  be the least natural so  $A_{i\lambda_i} \neq 0$  for each  $1 \leq i \leq r$ . By condition (b) & (c) of the reduced row echelon form,  $\{a_{\lambda_i} \mid 1 \leq i \leq r\}$  is linearly independent. In fact, as all other rows are zero, by condition (a),  $\text{rank}(A) = \text{rank}(A|b) = r$ . □

### (b) Ideas / thought organization

for every  $i \neq i'$ ,  $A_{i\lambda_i} = 0$  / columns  $a_{\lambda_i} = e_j$  for some  $1 \leq j \leq n$   
 $x_{\lambda_i} = -\sum_{j=\lambda_{i'}+1}^n A_{ij}x_j - b_{i\lambda_i}$  /  $A_{ij} = 0$  if  $j \in \{\lambda_i \mid 1 \leq i \leq r\}$

$$\sum_{i=1}^{n-r} t_i u_i = 0 \quad b_{\lambda_i} e_{\lambda_i}$$

Uh why did I tug this page again?

### Proof

Let  $x$  be a solution to  $Ax = b$ . For any  $1 \leq i \leq n-r$ , there exists  $j_i \in \{1, 2, \dots, n\} - \{\lambda_i \mid 1 \leq i \leq r\}$  with  $t_i = x_{j_i}$  so  $(u_i)_{j_i} = 1$  and  $(u_i)_{j_i'} = 0$  when  $i \neq i'$ . Hence, for any scalars  $a_i \in \mathbb{F}$ , if  $\sum_{i=1}^{n-r} a_i u_i = 0$ , then every  $a_i = 0$  (to have the  $j_i$ th entry be 0 for all  $i$ ). Hence, since the solution set is of rank  $n-r$  from (a), this means  $\{u_i \mid 1 \leq i \leq n-r\}$  is a basis for it as expected. Similarly, since we have that  $(x_0)_{\lambda_i} = b_{i\lambda_i}$  if  $1 \leq i \leq r$  and  $(x_0)_{\lambda_i} = 0$  otherwise, therefore  $Ax_0 = \sum_{i=1}^r (x_0)_{\lambda_i} a_{\lambda_i} = \sum_{i=1}^r b_{i\lambda_i} e_{\lambda_i} = b$  indeed. □

# Self-Proof of Theorem 3.1

Exercise 7  
IDEAL

$p_j := j$ th row of  $EA$ ,  $f_j := j$ th row of  $E$ ,  $a_j := j$ th row of  $A$ ,  $i_j := j$ th row of  $I_m$

(1)  $p_j = \sum_{k=1}^m (f_j)_k a_k = q_j = \begin{cases} a_j & \text{if } j \neq i_1, i_2 \\ \dots \end{cases}$

$$\begin{aligned} (EA)_{ij} &= \sum_{k=1}^m E_{ik} A_{kj} \\ \text{if } i \neq i_1, i_2 &= \sum_{k \in M - \{i_1, i_2\}} I_{ik} B_{kj} + \overbrace{I_{i i_1} B_{i_1 j}}^0 + \overbrace{I_{i i_2} B_{i_2 j}}^0 \\ &= I_{ii} B_{ij} = B_{ij} \\ \text{if } i = i_1 &= \sum_{k \in M - \{i_1, i_2\}} I_{i_1 k} B_{kj} + I_{i_1 i_1} B_{i_1 j} + \overbrace{I_{i_1 i_2} B_{i_2 j}}^0 \\ &= B_{i_1 j} \end{aligned}$$

**Proof**  
Let  $E$  and  $B$  be the  $n \times n$  matrices obtained from  $I$  and  $A$  respectively by means of a type 1 elementary operation. There exists distinct natural numbers  $i_1, i_2 = m$  with  $I_{i_1 k} = E_{i_1 k}$ ,  $I_{i_2 k} = E_{i_2 k}$ , and  $A_{i_1 k} = B_{i_2 k}$ ,  $A_{i_2 k} = B_{i_1 k}$  for all  $0 \leq k \leq m$ . Then, if  $i \neq i_1$  and  $i \neq i_2$ ,  
 $(EA)_{ij} := \sum_{k=1}^m E_{ik} A_{kj} = \sum_{k=1}^m I_{ik} A_{kj} = A_{ij} = B_{ij}$ . Even when  $i = i_1$  (or  $i = i_2$ ),  $(EA)_{i_1 j} = \sum_{k=1}^m E_{i_1 k} A_{kj} = E_{i_1 i_2} A_{i_2 j} = I_{i_1 i_2} B_{i_2 j} = B_{i_1 j}$ . Hence,  $EA = B$ .  
To get the same result for columns, just apply transposition. The rest follow similarly.

**Example 2**

$$EA = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 & 2 \\ 0 & 2 & 0 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 2 & 0 \\ 2 & 0 & 2 \\ 0 & 1 & 0 \end{pmatrix}$$

# Split-Proof of Theorem 3.2

Idea: Elementary matrix  $E$ ,

$$I_n = EF$$

$$F = E^{-1} \text{ by Ex 10.(b) of Section 2.4}$$

$$I = FE$$

Proof

Let  $E$  be an elementary  $n \times n$  matrix. By Theorem 3.1, there exists some  $n \times n$  matrix  $F$  at the same type with  $I = FE$  since  $I$  can be obtained from  $E$  by means of an elementary row operation. Hence,  $F = E^{-1}$  by exercise 10(b) of section 2.4. Same holds for elementary column operations. □

Exercises

1. (a) True ✓
- (b) False "scalar multiple"
- (c) True "multiplication of any row/column by  $\mathbb{I}$ "
- (d) False: Consider two elementary matrices  $E$  and  $F$  obtained by interchanging rows 1 and 2, and rows 3 and 4 of  $I$  respectively. Then  $EF$  cannot be expressed with only one elementary row operation, since 4 rows differ from  $I$ , while any single elementary row operation can only fix at most two of those. Theorem 3.1 says
- (e) True, Theorem 3.2
- (f) False, consider  $I_2 + I_2$
- (g) True ✓
- (h) False ✓
- (i) True ✓ Nice

$$\begin{pmatrix} 0 & 3 & 2 \\ 0 & 2 & 2 \\ 1 & -3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 3 & 2 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix}$$

2. An elementary operation that transforms  $A$  into  $B$  is one which adds  $-2$  of the 1st column to the 2nd. For  $B$  to  $C$ , add  $-1$  of the 1st row to the 2nd row. Finally, to convert  $C$  to  $I_3$ , the following elementary operations form one way:

- i. Multiplying the 2nd row by  $-\frac{1}{2}$ :  $\begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & 1 \\ 1 & -3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & 1 \\ 1 & -3 & 1 \end{pmatrix}$
  - ii. Adding  $-1$  of the 3rd row to the 1st row:  $\begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & 1 \\ 1 & -3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 2 \\ 0 & 1 & 1 \\ 1 & -3 & 1 \end{pmatrix}$
  - iii. Adding the 1st row to the 3rd row:  $\begin{pmatrix} 0 & 1 & 2 \\ 0 & 1 & 1 \\ 1 & -3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 2 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix}$
  - iv. Adding  $-2$  of the 2nd row to the 1st row:  $\begin{pmatrix} 0 & 1 & 2 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix}$
  - v. Add  $-1$  of the 1st row to the 2nd row:  $\begin{pmatrix} 0 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 3 \end{pmatrix}$
  - vi. Add  $-3$  of the 2nd row to the 3rd row:  $\begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}$
  - vii. Swap the 1st and 3rd rows:  $\begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
  - viii. Swap the 2nd and 3rd rows:  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$
- Indeed, we have  $I_3$  now.

3. (a) To obtain  $I_3$ , swap rows 1 and 3 of  $\begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ . Hence, its inverse is given by swapping rows 1 and 3 of  $I_3$ , i.e. itself. ✓

(b) To obtain  $I_3$ , multiply row 2 of  $\begin{pmatrix} 0 & 0 & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & 0 & 1 \end{pmatrix}$  by  $\frac{1}{3}$ . Thus, its inverse is  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ . ✓

(c) To get  $I_3$ , add 2 of row 1 to row 3 of  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix}$ . Therefore, its inverse is  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ . ✓

4. Let  $E$  be an elementary matrix.  
We evaluate case-wise by the type of  $E$ :

Type 1

Suppose row  $m$  is swapped with row  $n$  to obtain  $E$  from  $I$ . Then, swapping columns  $m$  and  $n$  of  $I$ , notice we get  $E$  still.

Type 2

Say, row  $m$  is multiplied by some scalar  $c \in \mathbb{F}$  in  $I$  to get  $E$ . Now we also have  $E$  by multiplying column  $m$  by  $c$  in  $I$ .

Type 3

Consider a scalar multiple  $c \in \mathbb{F}$  of a row  $m$  being added to row  $n$  to give  $E$  from  $I$ . It follows that  $E$  is also the result of adding  $c$  times of column  $m$  to column  $n$  of  $I$ .

Essentially, the arguments here boil down to  $I_{ij} = 1$  if  $i=j$ , and  $I_{ij} = 0$  otherwise. Indeed,  $E$  can be obtained in at least 2 ways.

5. Assume  $E$  is an elementary matrix. Evaluate case-wise again by type.

$$E_{mk} = I_{km} \quad (E^t)_{mk} = E_{km} = I_{km} \text{ or } I_{nm}$$

Type 1

Suppose row  $m$  is swapped with row  $n$  to obtain  $E$  from  $I$ . Then notice  $E$  is symmetric so  $E^t = E$ , because  $(E^t)_{mk} = E_{km}$ , which evaluates to  $I_{km} = E_{km}$  provided  $k \neq n$  and to  $I_{mm} = E_{nm} = E_{km}$  if  $k = n$ .

Type 2

Trivial.

Type 3

Consider a scalar multiple  $c \in \mathbb{F}$  of a row  $m$  being added to row  $n$  to give  $E$  from  $I$ . Now, by adding  $c$  times of row  $n$  to row  $m$  in  $I$ , we have  $E^t$ .

The converse is trivial since  $(E^t)^t = E$ .



6. Similar to 5.

7. See self-proof

8. Note that each elementary operation has an inverse of the same type for any suitable matrix E

9. Assume the elementary row operation  $R$  swaps rows  $m$  and  $n$ . Then, the following operations done in order (clearly results in swapping of rows  $m$  and  $n$ )

i. Add row  $n$  to row  $m$ ;  $E'_{mk} = E_{mk} + E_{nk}$  (&  $E'_{nk} = E_{nk}$ ), ✓

ii. Add  $-I$  of row  $m$  to row  $n$ ;  $E''_{nk} = E_{nk} - E'_{mk} = -E_{mk}$  (&  $E''_{ml} = E'_{ml}$ ), ✓

iii. Add row  $n$  to row  $m$ ;  $E'''_{mk} = E_{mk}$  (&  $E'''_{nk} = E_{nk}$ ), ✓

iv. Multiply row  $n$  by  $-I$ ;  $E''''_{nk} = E_{nk}$  &  $E''''_{nl} = E_{ml}$  (for all  $l$ ), so  $E'''' = R(E)$  as expected. ✓

(where the number of primes ('))  $p = E$  represents the matrix after applying the first  $p$  elementary row operations <sup>states above,</sup> successively on  $E$ .) ✓

10. Trivial since for any field  $F$  and  $x \in F$ ,  $(x^{-1})^{-1} = x$ .

11. Trivial since for any field  $F$  and  $x \in F$ ,  $-(-I)x = x$

12. Ideas

$$\begin{pmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ [c] & c_2 & c_3 \end{pmatrix}$$

$$\begin{pmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ 0 & c_2 & c_3 \end{pmatrix}$$

$$\begin{pmatrix} b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \\ 0 & c_2 & c_3 \end{pmatrix}$$

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \rightarrow \begin{pmatrix} a+c & b+d \\ c & d \end{pmatrix} \rightarrow \begin{pmatrix} a+c & b+d \\ a+2c & b+2d \end{pmatrix}$$

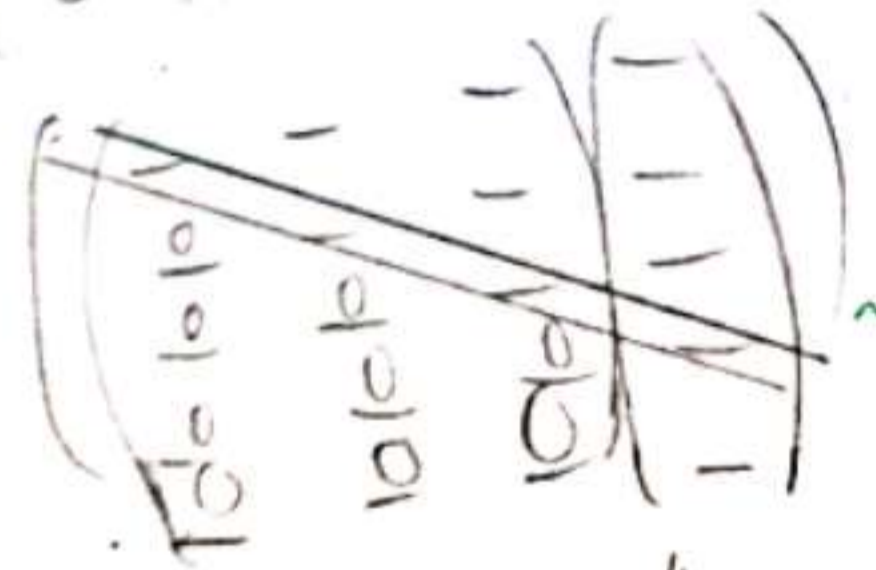
$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \xrightarrow{if a=c} \begin{pmatrix} a & b \\ 0 & d-c \end{pmatrix}$$

$$\xrightarrow{if a=0} \begin{pmatrix} c & d \\ 0 & b \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \checkmark \quad \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \checkmark \quad \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{pmatrix} \quad \begin{pmatrix} 0 & - \\ a_1 & - \end{pmatrix} \quad \begin{pmatrix} a_1 & b_1 \\ 0 & 1 \end{pmatrix}$$



~ *Probably better to say: This sequence of elementary row ops leads to an upper triangular matrix as they only involve its entries to with rows, which are all zero below the diagonal. An upper triangular matrix is constructed by construction. 2nd ed*

→ *Though the phrasing here isn't the best but I only did one iteration and no refinement.*

Proof.

When  $n=1$ , if ~~not~~ ~~these~~ ~~one~~ ~~entry~~ ~~of~~ ~~A~~ ~~is~~ ~~nonzero~~  $A_{a,1}, A_{a,2}, \dots, A_{a,k} \neq 0$  for some  $k \in \mathbb{Z}_0^+$  and  $a_i$ 's in  $F$ , then apply the row operations  $R_2, R_3, \dots, R_k(A) := \bar{A}$  forming an upper triangular matrix from  $A$ .

Assume this holds for  $n \in \mathbb{N}$ . For any  $m \times (n+1)$  matrix  $A_{n+1}$ , let  $A_n$  be the  $n \times n$  matrix defined by  $(A_n)_{ij} := (A_{n+1})_{ij}$ . So there is some  $k$  and sequence of elementary row operations  $O_1, \dots, O_k$  for  $1 \leq i \leq k$  that transforms  $A_n$  into an upper triangular matrix  $\bar{A}_n := O_1 O_2 \dots O_k(A_n)$ . Take the bottom right  $(m-n+1) \times 1$  matrix  $A_1$  defined by  $A_1$ , which from the  $n=1$  case above, we know exists a sequence of row operations  $R_i$  for  $1 \leq i \leq k'$  with  $R_1 R_2 \dots R_{k'}(A_1)$  being upper triangular. Notice this sequence of transformations keep  $\bar{A}_n$  upper triangular since  $(\bar{A}_n)_{ij} = 0$  for  $n+1 \leq i \leq m$  and  $1 \leq j \leq n$ .

Now,  $R_1 R_2 \dots R_{k'} O_1 O_2 \dots O_k(A)$  must be an upper triangular matrix. Consequently, induction tells us this is true for all  $n$  (and  $m$ ) in  $\mathbb{N}$ .

(2 by taking  $O_1 O_2 \dots O_k(A_{n+1})$ )

Self-Proof of Theorem 3.4

(a) By the invertibility of  $Q$ ,  $R(LA) = \mathbb{F}^n$  so  $R(LA) = R(LA) \cdot I_n$ . Hence,  $\text{rank}(AQ) = \text{rank}(A)$ .

(b) Follows similarly.

(c)  $\text{rank}(PAQ) = \text{rank}((PA)Q) \stackrel{(a)}{=} \text{rank}(PA) \stackrel{(b)}{=} \text{rank}(A)$  as expected.

Self-Proof of Corollary

Let  $A \in M_{m \times n}(\mathbb{F})$  and  $B$  be obtained from  $A$  by an elementary row operation. By Theorem 3.1, there exists an elementary matrix  $E$  obtained from  $I_m$  by performing the same elementary row operation, such that  $B = EA$ . Since Theorem 3.2 says  $E$  is invertible, Theorem 3.4 tells us

$\text{rank}(B) = \text{rank}(EA) = \text{rank}(A)$ .

Self-Proof of Theorem 3.5

Ideas

Every  $Ax$  can be generated by  $\{a_j \mid 1 \leq j \leq n\}$

$$Ax = \sum_{j=1}^n x_j a_j$$

$a_j$  be the  $j$ th column of  $A$ .

Proof

Let  $A \in M_{m \times n}(\mathbb{F})$ , and  $x \in \mathbb{F}^n$ ; we know  $Ax = \sum_{j=1}^n x_j a_j$  so  $R(LA) = \text{span}\{a_j \mid 1 \leq j \leq n\}$ . And hence,  $\text{rank}(A) = \dim(\text{span}\{a_j \mid 1 \leq j \leq n\})$

as expected.

Example 1

$$\begin{pmatrix} 1 & 2 & 1 \\ 1 & 0 & 3 \\ 1 & 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 2 & 2 & 4 \\ 1 & 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ 1 & 1 & 2 \end{pmatrix} \text{ ops used column row!} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 1 & 0 & 3 \\ 1 & 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 0 \\ 1 & 0 & 2 \\ 1 & 2 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 2 \\ 1 & 1 & 1 \end{pmatrix}$$

Thus,  $\text{rank}\left(\begin{pmatrix} 1 & 2 & 1 \\ 1 & 0 & 3 \\ 1 & 1 & 2 \end{pmatrix}\right) = \text{rank}\left(\begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 2 \\ 1 & 1 & 1 \end{pmatrix}\right) = \dim(\text{span}\{(1, 1, 0), (1, 1, 2)\}) = 2$ .

### Self-Proof of Theorem 3.6

Similar to the proof for exercise 12 of section 3.1.

### Self-Proof of Corollary 1

From Theorems 3.1, 3.2, and 3.6.

$$\begin{aligned} & ((C^t)^{-1}(B^{-1}D)^t)^t \\ & (C^t)^{-1}(D^t) (B^t)^{-1} \end{aligned}$$

### Self-Proof of Corollary 2

(a) By Corollary 1, there exists a matrix  $D$  of the form  $\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$  and two invertible matrices  $B$  and  $C$ , with  $D = BAC$  (or  $A = B^{-1}DC^{-1}$ ).  
So,  $\text{rank}(A^t) = \text{rank}(B^{-1}DC^{-1})^t = \text{rank}((C^t)^{-1}(D^t)(B^t)^{-1}) = \text{rank}(D^t) = \text{rank}(D) = \text{rank}(A)$  by exercise 5 of section 2.4 and

Theorem 3.4.

(b) Follows from (a) and Theorem 3.5.

(c) By Theorem 3.5 and (b).

$$R(A_n) = I_n$$

$$R^2(A_n) = R(I_n) = A^{-1}$$

### Self-Proof of Corollary 3

Let  $A$  be an invertible  $m \times n$  matrix. So, it has rank  $m$ ; by the proof of Corollary 1, there exists  $p, q$  and corresponding elementary matrices  $E_i$  and  $G_j$  for  $1 \leq i \leq p$  and  $1 \leq j \leq q$ , with  $(D =) I_n = \left( \prod_{i=1}^p E_i \right) A \left( \prod_{j=1}^q G_j \right)$ . Hence,  $A = \left( \prod_{i=1}^p E_i^{-1} \right) \left( \prod_{j=1}^q G_j^{-1} \right)$  where each  $E_i^{-1}$  and  $G_j^{-1}$  are elementary matrices by Theorem 3.2.

### Self-Proof of Theorem 3.7:

(a)  $R(UT) = U[R(T)] \subseteq U[E^W] = R(U)$ . So,  $\text{rank}(UT) \leq \text{rank}(U)$ .

(b) By the Dimension Theorem,  $\text{rank}(U[R(T)]) = \text{rank}(T) - \text{nullity}(U[R(T)]) \leq \text{rank}(T)$ .

(c) By (a):

(a) By (b):

$$\begin{aligned} BI_n &= A^{-1} \\ BBI_n &= A^{-1} \end{aligned}$$

### The Inverse of a Matrix

$$\begin{aligned} I_n &= BA \\ A^{-1} &= B \end{aligned}$$

$$\begin{aligned} R(A) &= I_n \\ R(I_n) &= A^{-1} \end{aligned}$$

$$R^2(A) = A^{-1} \quad \checkmark$$

$$\begin{aligned} I_n &= BAC \\ A &= B^{-1}C^{-1} \\ A^{-1} &= CB \end{aligned}$$

$$\begin{aligned} O(A) &= I_n \\ O(I_n) &= BC + A^{-1} \end{aligned}$$

Exercises

(4) False, it's the number of linearly independent columns of the matrix.  
 (b) False, consider  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$  defined by  $T(e_1) = 0$  and  $T(e_2) = T(e_3) = e_1$ . Then,  $\text{rank}(T^2) = \text{rank}(T_0) = 0 \neq 1 = \text{rank}(T)$ .

- (c) True
- (d) True
- (e) False
- (f) True
- (h) True
- (i) True
- (g) True

2. (a) Notice  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ -1 & 1 & 0 \\ 0 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$  so  $\text{rank} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} = 2$ .

(c) we see that  $\begin{pmatrix} 1 & 0 & 2 \\ 1 & 0 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$  thus  $\text{rank} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} = 2$  as well.

(g) we note that  $\begin{pmatrix} 1 & 1 & 0 & 1 \\ 2 & 2 & 0 & 2 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$  has rank 1.

3. Trivial

4. (a)  $\begin{pmatrix} 1 & 1 & 1 & 2 \\ 1 & 0 & 1 & 2 \\ 1 & 1 & 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$  as required, so it has rank 2.

(b)  $\begin{pmatrix} 2 & 1 \\ -1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 \\ -5 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 \\ 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ , again this has rank 2.

5. (a) We compute  $\begin{pmatrix} 1 & 2 & 1 & 10 \\ 1 & 1 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 & 10 \\ 0 & -1 & 0 & -10 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & -10 \\ 0 & -1 & 0 & -10 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & -10 \\ 0 & -1 & 0 & -10 \end{pmatrix}$ . So,  $\begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix}$  is of rank 2 and has the inverse  $\begin{pmatrix} -1 & 2 \\ 1 & -1 \end{pmatrix}$ .

(c) Again,  $\begin{pmatrix} 1 & 2 & 1 & 10 & 0 \\ 1 & 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 10 & 0 \\ 1 & 1 & 1 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 & 10 & 0 \\ 0 & 1 & 0 & -10 & 0 \\ 0 & -1 & 0 & -10 & 0 \\ 0 & 0 & 0 & -10 & 0 \end{pmatrix}$ . Thus,  $\begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 0 \end{pmatrix}$  is of rank 2 and invertible.

(e) One more,  $\begin{pmatrix} 1 & 2 & 1 & 10 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 2 & 1 & 10 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 & 10 & 0 & 0 \\ 0 & 1 & 0 & -10 & 0 & 0 \\ 0 & -1 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \end{pmatrix}$ . Hence,  $\begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 0 \end{pmatrix}$  has rank 2 and cannot be inverted.

$\begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \\ 0 & 0 & 0 & -10 & 0 & 0 \end{pmatrix}$   
 Hence it has rank 3 and inverse  $\begin{pmatrix} 1 & 3 & 1 \\ 1 & 3 & 1 \\ 1 & 3 & 1 \end{pmatrix}$

6. (a)  $T(1) = -1, T(x) = 2-x, T(x^2) = 2+4x-x^2$

$$\left( \begin{array}{ccc|ccc} -1 & 2 & 2 & 1 & 0 & 0 \\ 0 & -1 & 4 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 2 & -1 & -2 & 0 \\ 0 & 1 & 4 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & -1 & -2 & -10 \\ 0 & 1 & 0 & 0 & -1 & -4 \\ 0 & 0 & 1 & 0 & 0 & -1 \end{array} \right)$$

oops acci did column ops

$T$  is invertible and is given by  $T^{-1}(a+bx+cx^2) = -a-2b-10c - (b+4c)x - cx^2$ .

check:  $f'(x) = -b-4c-2cx, f''(x) = -2c$ , so  $TT^{-1}(a+bx+cx^2) = -2c - 2b - 8c - 4cx + a + 2b + 10c + (b+4c)x + cx^2 = a+bx+cx^2$

(c)  $T(1,0,0) = (1,-1,1), T(0,1,0) = (2,1,0), T(0,0,1) = (1,2,1)$

$$\left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ -1 & 1 & 2 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ 0 & 1 & 3 & 0 & 1 & 1 \\ 0 & -2 & 0 & -1 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 3 & 0 & 1 & 1 \\ 0 & 0 & 6 & -1 & 2 & 3 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & -3 & -2 \\ 0 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & -1 & 2 & 3 \end{array} \right)$$

$T$  is invertible and  $T^{-1}(a,b,c) = (\frac{1}{2}a - \frac{1}{3}b + \frac{1}{2}c, \frac{1}{2}a - \frac{1}{2}c, -\frac{1}{6}a + \frac{1}{3}b + \frac{1}{2}c)$ .

$$\left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ -1 & 1 & 2 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ 0 & -2 & 1 & 1 & 0 & 0 \\ 0 & -2 & 0 & -1 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & -2 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & -2 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right)$$

Therefore,  $\left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ -1 & 1 & 2 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{array} \right)^{-1} = \left[ \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & -2 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & -2 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & -2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & -2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & -2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & -2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & -2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \right]^{-1}$

And so  $\left( \begin{array}{ccc|ccc} 1 & 2 & 1 & 1 & 0 & 0 \\ -1 & 1 & 2 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{array} \right)^{-1} = \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & -1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & -2 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right)$  ✓ checked with GC

8. By Theorem 2.15(c),  $R(L_A) = R(cA) = R(A)$ . Thus  $\text{rank}(cA) = \text{rank}(A)$ .

9. See self-proof.

11. Let  $b_j$  be the  $j$ th column of  $B$ . We see that  $b_1 \notin \text{span}\{b_j \mid 2 \leq j \leq n\}$  so  $\text{rank}(\bar{B}) = r-1$ , where  $\bar{B}$  is the  $m \times (n-1)$  matrix with  $\bar{B}_{ij} = B_{i,j+1}$ .  
Furthermore, since  $\text{rank}(\bar{B})$  is clear isomorphic to  $\text{rank}(LB')$ ,  $\text{rank}(B') = \text{rank}(\bar{B}) = r-1$ .

# IF - Exercise

Prove that every inverse matrix  $A^{-1}$  can be computed exclusively by a finite sequence of elementary row operations on  $A$ .

Idea: Let  $A_n$  be non-invertible matrix.

$n=1$ ,  $A_1$  trivial

Assume  $n$ , let  $a_j^c$  be  $A_n$ 's  $j$ th column,  $a_j^r$  its  $j$ th row

$a_1 \rightarrow e_1$  by  $R_1, R_2, \dots, R_k$   
 $\neq 0$

$\{a_j^c \mid 1 \leq j \leq n\}$  is a basis for  $M_{1 \times n}(\mathbb{F})$   
 $\Rightarrow \sum_{j=1}^n c_j a_j^c = (1, 0, 0, \dots, 0)$

Apply assumption / IH ✓

Proof: We first show that every  $n \times n$  invertible matrix can be transformed into  $I_n$  by a finite sequence of row operations. When  $n=1$ , the result is trivial. So, assume this holds for  $n \in \mathbb{N}$  and let  $A$  be a  $(n+1) \times (n+1)$  invertible matrix.

As usual,  $A$  can be transformed into  $e_1$  by a finite  $k_1$  number of elementary row operations  $R_1$  to  $R_{k_1}$ . Let  $\bar{r}_j$  be the  $j$ th row of  $R_1 R_2 \dots R_{k_1}(A)$ . Since  $R_1 R_2 \dots R_{k_1}(A)$  is of rank  $n+1$ , so by virtue of  $\{\bar{r}_j \mid 1 \leq j \leq n+1\}$  forming a basis for  $M_{1 \times (n+1)}(\mathbb{F})$ , there exists some scalars  $c_j \in \mathbb{F}$  for which  $\sum_{j=1}^{n+1} c_j \bar{r}_j = (1, 0, 0, \dots, 0)$ .

Thus, there are some  $k_2$  such elementary row operations  $\bar{R}_1$  to  $\bar{R}_{k_2}$  with  $\bar{R}_1 \bar{R}_2 \dots \bar{R}_{k_2}(\bar{r}_1) = (1, 0, 0, \dots, 0)$ . Now, the  $n \times n$  matrix  $A'$  given by  $A'_{i,j} := \bar{R}_1 \bar{R}_2 \dots \bar{R}_{k_2} \bar{R}_1 \bar{R}_2 \dots \bar{R}_{k_1}(A)_{i+1,j}$  has rank  $n$ , and can be transformed into  $I_n$  by a finite  $k_3$  number of elementary row operations  $R'_1$  to  $R'_{k_3}$ .

as our assumption / induction hypothesis tells us. Therefore, the finite sequence  $R'_1, R'_2, \dots, R'_{k_3}, \bar{R}_1, \bar{R}_2, \dots, \bar{R}_{k_2}, R_1, R_2, \dots, R_{k_1}$  transforms  $A$  into  $I_{n+1}$ .

consequently, given any invertible  $n \times n$  matrix  $A$ , there exists a finite  $k$  number of elementary row operations  $R_i$  with the composition  $R := R_1 R_2 \dots R_k$  that transforms  $A$  into  $I_n$ . Letting  $E_i := R_i(I_n)$  be the corresponding elementary matrices and  $B := E_1 E_2 \dots E_k$ , we have  $I_n = R(A) = BA$  so that  $A^{-1} = BI_n = B^2 A = R^2(A)$ .

wherefore, the finite sequence of elementary row operations  $R^2$  transforms  $A$  to  $A^{-1}$ . □

13. See self-proof. □

14. (a) Clear from the relevant definitions. □

(b) Trivial from (a) □

(c) Follows immediately from (b). □

15. Let  $x_j$  and  $y_j$  be the  $j$ th columns of  $M$  and  $M(A|B)$ , respectively. Since we know  $y_j = \sum_{i=1}^n (A|B)_{ij} x_i$ , it is clear that  $M(A|B) = (MA|MB)$ . □

16. See self-proof. □

$$B : \text{rank}(BC) \leq \text{rank}(B) \leq 1$$

17. We see that for some suitable scalars,

$$BC = \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{pmatrix} \begin{pmatrix} c_1 & c_2 & c_3 \end{pmatrix} = \begin{pmatrix} b_{11}c_1 & b_{11}c_2 & b_{11}c_3 \\ b_{21}c_1 & b_{21}c_2 & b_{21}c_3 \\ b_{31}c_1 & b_{31}c_2 & b_{31}c_3 \end{pmatrix} \rightarrow \begin{pmatrix} b_{11}c_1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Hence,  $BC$  must have at most rank 1.

→ Ehh nah this kinda bad notation when you look at the following sentence. For some  $3 \times 1$  matrix  $B$ ,  $A = [B \ c_2 B \ c_3 B] = B [c_1 \ c_2 \ c_3]$ . □

Conversely, since  $A$  has rank 1, this suggests that its columns are scalar multiples of each other.

18. Let  $\bar{A}_j$  be the  $n \times n$  matrix whose  $j$ th column is the  $j$ th column of  $A$ , and 0 everywhere else. Then, it is a clear extension of exercise 14 of Section 2-3 to say  $AB = \sum_{j=1}^n \left( \sum_{i=1}^n B_{ji} \bar{A}_j \right)$ . □

~~Section 2-3 to say~~  $AB = \sum_{j=1}^n \left( \sum_{i=1}^n B_{ji} \bar{A}_j \right)$  □

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} \begin{pmatrix} 7 & 8 & 9 \\ 10 & 11 & 12 \end{pmatrix} = \begin{pmatrix} 27 & 30 & 33 \\ 61 & 68 & 75 \\ 95 & 106 & 117 \end{pmatrix}$$

(1)  $(a_1 \ a_2 \ a_3 \ a_4 \ a_5)$

$$\begin{pmatrix} x_1 & \beta_1 \\ x_2 & \beta_2 \\ x_3 & \beta_3 \end{pmatrix} \begin{pmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{pmatrix} = \begin{pmatrix} a_1 x_1 + b_1 \beta_1 & \dots & \dots \\ a_1 x_2 + b_1 \beta_2 & \dots & \dots \\ a_1 x_3 + b_1 \beta_3 & \dots & \dots \end{pmatrix} \text{rank} \leq 2?$$

$$AB = A \begin{pmatrix} I_{r_2} & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} \dots & 0 \\ \dots & 0 \end{pmatrix}$$

$$\begin{matrix} \underbrace{v_1 \ v_2 \ v_3}_{B} = a v_1 + b v_2 \\ \downarrow \\ v_1 - b v_2, (1-b)v_2, v_3 = a v_1 \end{matrix}$$

$$\begin{matrix} u_1 \ u_2 \ (u_1 + u_2) \\ u_1 + u_2 \ 2u_2 \ (u_1 + u_2) \\ v_1 \ v_2 \ v_3 \\ = c(u_1 + u_2) - \frac{(1-c)(2u_2)}{2} \\ = c v_1 - \frac{1-c}{2} v_2 \end{matrix} \begin{pmatrix} 1 & 0 & \frac{1}{61} \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & \frac{62}{61} & \frac{63}{61} \\ 61 & 68 & 75 \\ 95 & 106 & 117 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & \frac{62}{61} & \frac{63}{61} \\ 0 & 6 & 12 \\ 0 & -\frac{3890}{61} & -\frac{5911}{61} \end{pmatrix}$$

$$\leftarrow \begin{pmatrix} 1 & 0 & \frac{1}{61} \\ 0 & 6 & 12 \\ 0 & 0 & -\frac{95}{61} \end{pmatrix}$$

$$b = \frac{c-1}{2} \\ 2b+1 = c$$

$$\begin{pmatrix} A \\ B \end{pmatrix} M = \begin{pmatrix} AM \\ BM \end{pmatrix}$$

$$\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Since  $R_1, R_2, \dots, R_k(A)$  is...  
 The...  
 18. Ideas

$$AB \rightarrow \begin{pmatrix} I_{r_1} & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} I_{r_2} & 0 \\ 0 & 0 \end{pmatrix} \stackrel{\text{claim}}{=} \begin{pmatrix} I_{\min\{r_1, r_2\}} & 0 \\ 0 & 0 \end{pmatrix}$$

consider  $1 \leq i, j \leq r$   
 $(AB)_{ij} = \sum_{k=1}^r A_{ik} B_{kj} = \sum_{k=1}^r \delta_{ik} \delta_{kj} = \delta_{ij}$   
 $(AB)_{ij} = \sum_{k=1}^r 0 \cdot B_{kj} = 0$  as  $A_{ik} = 0$  if  $k < i$   
 $(AB)_{ij} = 0$  if  $r_2 < i < r_1$   
 clearly  $\text{rank} \leq 1$

$$\begin{pmatrix} | & & | \\ \hline & & \\ \hline | & & | \end{pmatrix} B = \begin{pmatrix} b_1^A & b_2^A & \dots & b_p^A \end{pmatrix}$$

Assume  $k$

$$\begin{pmatrix} | & & | \\ \hline & & \\ \hline | & & | \end{pmatrix} P$$

$S_1$  to  $S_n$   
 $S_k$  := the map matrix  
 with  $(S_k)_{ij} := A_{ij}$  if  $i=j=k$ , 0 otherwise.  
 $\sum_{k=1}^n S_k = \begin{pmatrix} I_{\text{rank}(AB)} & 0 \\ 0 & 0 \end{pmatrix} = P(AB)Q$  for invertible  $P, Q$   
 $AB = \sum_{k=1}^n P^{-1} S_k Q^{-1}$

$\text{rank}(AET) \stackrel{\text{claim}}{=} \text{rank}(AT)$  if  $E$  is invertible  
 $R(AET) = [L_A L_E [R(AT)]]$   
 $= L_A^{-1} X$  since  $L_E$  is invertible; its  $\det$

$$\mathbb{F}^p \xrightarrow{L_B} \mathbb{F}^n \xrightarrow{L_A} \mathbb{F}^m$$

$\text{rank}(L_B) \leq n$   $R(L_A L_B) = L_A [R(L_B)]$   
 $\text{rank}(L_A L_B) = \text{rank}(L_A R(L_B))$   
 $\leq n - \text{nullity}(L_A R(L_B))$   
 $\leq n$

Theorem 3.7  
 $\text{rank}(AB) \leq \text{rank}(B) \leq n$

Proof

First, notice that by Theorem 3.7,  $\text{rank}(AB) \leq \text{rank}(B) \leq n$ . Hence, for some invertible matrices  $P$  and  $Q$ ,  $\begin{pmatrix} I_{\text{rank}(AB)} & 0 \\ 0 & 0 \end{pmatrix} = P(AB)Q$  according to Corollary 1 to Theorem 3.6. Define the map matrices  $S_k$  by  $(S_k)_{ij} = A_{ij}$  when  $i=j=k$ , and 0 otherwise. It is thus clear that  $\text{rank}(S_k) \leq 1$  and  $\sum_{k=1}^n S_k = \begin{pmatrix} I_{\text{rank}(AB)} & 0 \\ 0 & 0 \end{pmatrix}$ . Combining these results, we have that  $AB = \sum_{k=1}^n P^{-1} S_k Q^{-1}$ , where  $\text{rank}(P^{-1} S_k Q^{-1}) = \text{rank}(S_k) \leq 1$  by Theorem 3.4(1). □



19. Ideas  $\mathbb{F}^p \xrightarrow{L_B} \mathbb{F}^n \xrightarrow{L_A} \mathbb{F}^m$

This suggests  $p \geq n \geq m$ ,  $L_B$  is surjective so  $\text{rank}(L_A L_B) = \text{rank}(L_A) = m$ .

Proof

We see that  $p \geq n \geq m$  for  $\text{rank}(B) = n$  and  $\text{rank}(A) = m$ , where  $L_B$  must be surjective, so  $\text{rank}(AB) = \text{rank}(L_A L_B) = \text{rank}(L_A) = m$ .

20. (a) We want to find 2 possible columns of  $M$  which are nonzero but will evaluate to zero in  $AM$ :

$$a_1 \begin{pmatrix} 1 \\ -1 \\ 3 \end{pmatrix} + a_2 \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} + a_3 \begin{pmatrix} -1 \\ 3 \\ -5 \end{pmatrix} + a_4 \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} + a_5 \begin{pmatrix} 1 \\ 0 \\ -6 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

By GC:

$$\begin{aligned} a_1 &= a_3 + 3a_5 \\ a_2 &= -2a_3 + a_5 \\ a_3 &= a_5 \\ a_4 &= -2a_5 \\ a_5 &= a_5 \end{aligned}$$

So, two possible columns are  $\begin{pmatrix} 1 \\ -2 \\ 1 \\ 0 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 4 \\ -1 \\ -1 \\ 1 \\ 1 \end{pmatrix}$ . Correspondingly, we have the matrix  $M =$

$$M = \begin{pmatrix} 1 & 4 & 0 & 0 & 0 \\ -2 & -1 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{pmatrix}$$

Indeed we can verify that  $AM = 0$ .

(b) Ideas

$$\begin{pmatrix} 1 & 0 & -1 & 2 & 1 \\ -1 & 1 & 3 & -1 & 0 \\ -2 & 1 & 3 & -1 & 0 \\ 3 & -1 & 5 & 1 & 6 \\ -5 & 0 & 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -1 & 2 & 1 \\ 0 & 1 & 3 & -1 & 0 \\ 0 & 1 & 3 & 1 & 4 \\ 0 & -1 & 8 & -5 & 3 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -1 & 2 & 1 \\ 0 & 1 & 3 & -1 & 0 \\ 0 & 0 & 0 & 2 & 4 \\ 0 & 0 & 11 & -6 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 2 & 1 \\ 0 & 1 & 5/2 & -1 & 0 \\ 0 & 0 & 1 & 2 & 4 \\ 0 & 0 & 8 & -6 & 3 \end{pmatrix}$$

$$\mathbb{F}^5 \xrightarrow{L_B} \mathbb{F}^5 \xrightarrow{L_A} \mathbb{F}^4$$

$$\text{rank}(B) = 2, \text{rank}(L_A) = 4$$

$$\text{nullity}(L_B) = 3, \text{nullity}(L_A) = 1$$

$$\begin{aligned} \text{rank}(AB) + \text{nullity}(AB) &= 5 \\ 0 + \text{nullity}(AB) &= 5 \end{aligned}$$

$$\begin{aligned} \text{nullity}(B) + \text{nullity}(A_{R(B)}) &= 5 \\ \text{nullity}(A_{R(B)}) &= 2 \end{aligned}$$

$$L_{A_{R(B)}} : \underbrace{R(B)}_{\dim 2} \rightarrow \mathbb{F}^4$$

$$\begin{aligned} (a_1 \ a_2) \begin{pmatrix} b_1 & b_2 \\ b_3 & b_4 \end{pmatrix} \\ \mathbb{R}^2 \rightarrow \mathbb{R}^4 \\ (1, 1) \begin{pmatrix} 1 \\ -1 \end{pmatrix} \end{aligned}$$

20. (b) Notice  $A$  is of rank 3 so nullity  $(A) = 2$  by the Dimension Theorem. As such, for  $AB = 0$ ,  $\text{rank}(B) \leq 2$ .
21. Since  $\text{rank}(A) = m$ , the columns  $a_i$  of  $A$  generate  $\mathbb{F}^m$ . So, for each  $1 \leq j \leq m$ , there exists scalars  $B_{ij}$  for  $1 \leq i \leq n$ , such that  $e_j = \sum_{i=1}^n B_{ij} a_i$ .  
 Defining  $B$  to be the  $n \times m$  matrix with entries  $B_{ij}$ , it is clear that  $AB = I_m$ .
22. Similarly, by virtue of  $\text{rank}(B) = m$ , the rows  $b_j$  of  $B$  generate  $M_{1 \times m}(\mathbb{F})$ . As such, for any  $1 \leq i \leq n$ , there exists scalars  $A_{ij}$  for  $1 \leq j \leq m$  with  $\sum_{j=1}^m A_{ij} b_j$  being the  $i$ th row of  $I_m$ . Hence, it is again straightforward to notice  $AB = I_m$ .

### Self-Proof of Theorem 3.8

Trivial

### Self-Proof of Corollary

If  $m < n$ ,  $\dim(K) = n - \text{rank}(A) \geq n - m > 0$ .

### Self-Proof of Theorem 3.9

Let  $s'$  be a solution to  $Ax = b$ . Then  $A(s' - s) = 0$  so  $s' = s + (s' - s) \in s + K_H$ . Conversely, for  $k \in K_H$ ,  $A(s+k) = As + Ak = b$ .

Hence,  $K = s + K_H$ .

### Self-Proof of Theorem 3.10

When  $A$  is invertible, for any solution  $x$  to  $Ax = b$  we indeed have  $x = A^{-1}b$ . Which means  $A^{-1}b$  is the only solution. So, consider  $Ax = b$  having only 1 solution, say some  $s$ . Then, the solution space to  $Ax = 0$  must be  $\{0\}$ , lest there exists some solution  $s+k \neq s$  to  $Ax = b$  by Theorem 3.9.

Accordingly, nullity  $(A) = 0$  so  $A$  is invertible.

### Self-Proof of Theorem 3.11

Ideas  
 If  $\exists$  soln  $x$ , i.e.  $Ax = b$ ,  $(A|b) \begin{pmatrix} x \\ -1 \end{pmatrix} = 0$   $b$  is a lin comb of columns of  $A$ .

If  $b \neq 0$  & all soln  $y$  to  $(A|b)y = 0$  are s.t.  $y = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$ ,  $\exists$  no soln to  $Ax = b$

$\text{rank}(A) \leq \text{rank}(A|b)$

If  $\text{rank}(A|b) = \text{rank}(A) + 1$ ,  $\mathbb{P}^n \setminus \mathcal{U}(b)$  is a lin subsp (normal) &  $\sum_{i=1}^n c_i a_i \neq b \forall c_i$   
maximal set of lin columns

$m \times (n+1)$

When  $\text{rank}(A) = \text{rank}(A|b)$ ,  $b$  lin comb of cols of  $A$

$$b = \sum_{j=1}^n c_j a_j$$

$$\text{Soln: } \begin{pmatrix} c_1 \\ \vdots \\ c_n \\ -1 \end{pmatrix}$$

$$\text{to } (A|b)y = 0$$

# 1 - Proof of Theorem 3.11

For a consistent system of linear equations  $Ax = b$ , with some solution  $x$ . Then,  $b$  can be expressed as a linear combination of columns in  $A$ . Therefore,  $\text{rank}(A) = \text{rank}(A|b)$  is clear by considering the maximal number of linearly independent columns. Conversely, when  $\text{rank}(A) = \text{rank}(A|b)$ ,  $b$  is a linear combination of columns of  $A$ . In other words, there are scalars  $c_j \in \mathbb{F}$  with  $b = \sum_{j=1}^n c_j a_j$ . Hence, it is clear that  $\begin{pmatrix} x \\ c_j \end{pmatrix}$  is a solution  $Ax = b$ . Which implies the consistency of the system  $Ax = b$ . □

cases

True ✓  
False ✓

True, considering any system  $Ax = 0$ ,  $x = 0$  is a solution.

False ✓

1) False. Consider  $\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$ , then all  $x \in \mathbb{F}^2$  is a solution.

2) False ✓

3) True. The only solution would be  $x = A^{-1}0 = 0$ .

4) False. Consider  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ . Then  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  is the sole solution, thus the solution set does not contain  $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$  and can't be a subspace of  $\mathbb{F}^n$ . Nice

5)  $\begin{pmatrix} 1 & 3 & 6 \\ 2 & 2 & 6 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 \\ 2 & 2 \end{pmatrix}$  has rank 1. Thus, the solution space has dimension 1 and is spanned by  $\left\{ \begin{pmatrix} 3 \\ -1 \end{pmatrix} \right\}$ . the basis

6)  $\begin{pmatrix} 1 & 2 & -1 \\ 2 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 3 & 3 & -1 \\ 3 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 \\ 1 & 0 & -1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 2 \\ 1 & 0 & 0 \end{pmatrix}$  clearly has rank 2. So its solution space has dimension 1 and the basis  $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$ .

7)  $(1 \ 2 \ -3 \ 1)$  clearly is of rank 1, so its solution space is of dimension 3 and has the basis  $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \right\}$ . Its linear independence is seen from the fact that

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & -2 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

has rank 3.

2. (g)  $\begin{pmatrix} 1 & 2 & 1 & 1 \\ 0 & 1 & -1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & -1 & 1 \end{pmatrix}$  has rank 2. Therefore, its solution space of dimension 2 has the basis  $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \right\}$ .

3. (a)  $x_1 = -3x_2$  NOT correct btw forms! 3. are nonhomogeneous! Oops

(c)  $-x_1 = x_2 = x_3$

(e) Any solution can be represented as  $a \begin{pmatrix} 1 \\ 0 \\ 0 \\ -1 \end{pmatrix} + b \begin{pmatrix} 0 \\ 1 \\ -2 \\ 0 \end{pmatrix} + c \begin{pmatrix} 0 \\ 0 \\ 1 \\ 3 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \\ -a-2b+3c \end{pmatrix}$ , which is equivalent to  $x_4 = -x_1 - 2x_2 + 3x_3$ .

(g) Similarly for (g), any solution can be written as  $a \begin{pmatrix} -2 \\ 1 \\ 1 \\ 0 \end{pmatrix} + b \begin{pmatrix} -3 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -2a-3b \\ a+b \\ a+b \\ b \end{pmatrix}$ . Hence, any solution is given by the defining equations  $x_1 = -2x_4 - 3x_2$  and  $x_3 = x_4 + x_2$ .

(a) (1) We first compute the inverse:  $\left( \begin{array}{cc|cc} 1 & 3 & 1 & 0 \\ 2 & 5 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{cc|cc} 1 & 3 & 1 & 0 \\ 0 & -1 & -2 & 1 \end{array} \right) \rightarrow \left( \begin{array}{cc|cc} 1 & 0 & -5 & 3 \\ 0 & -1 & -2 & 1 \end{array} \right)$ .

(2) Then, the unique solution is given by  $\begin{pmatrix} -5 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 4 \\ 3 \end{pmatrix} = \begin{pmatrix} -11 \\ 5 \end{pmatrix}$ , i.e.  $-11x_1 = 5x_2$ .

(b) (1) Again, the inverse is  $\begin{pmatrix} 1 & 2 & -1 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 \\ 2 & -2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 \\ 0 & -6 & 3 & -2 & 0 \\ 0 & 0 & -9 & 4 & -6 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & \frac{1}{3} & 0 \\ 0 & 1 & -2 & 1 & -1 \\ 0 & 0 & -9 & 4 & -6 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & \frac{1}{3} & 0 \\ 0 & 1 & 0 & \frac{1}{3} & -1 \\ 0 & 0 & 1 & \frac{1}{3} & -2 \end{pmatrix}$ .

(2) Now, the unique solution is again  $\begin{pmatrix} \frac{1}{3} & 0 & 0 & \frac{1}{3} & 0 \\ 0 & 1 & 0 & \frac{1}{3} & -1 \\ 0 & 0 & 1 & \frac{1}{3} & -2 \end{pmatrix} \begin{pmatrix} 5 \\ 1 \\ 4 \end{pmatrix} = \begin{pmatrix} 3 \\ 0 \\ -2 \end{pmatrix}$ . In other words,  $3x_1 = -2x_3$ .

One example is:

$$\begin{aligned} x_1 + x_2 + 0x_3 + 0x_4 + \dots + 0x_n &= 1 \\ x_1 + x_2 + 0x_3 + 0x_4 + \dots + 0x_n &= 1 \\ &\vdots \\ x_1 + x_2 + 0x_3 + 0x_4 + \dots + 0x_n &= 1 \end{aligned}$$

It is represented by  $\begin{pmatrix} 1 & 1 & 0 & 0 & \dots & 0 \\ 1 & 1 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 0 & 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}$ .

There are  $\infty$  solutions at least, because for each  $x_1 \in \mathbb{R}$ ,  $(x_1, 1-x_1, 0, 0, \dots, 0)^t$  is a solution. As such, there indeed are infinitely many solutions.

$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 1 \\ 11 \end{pmatrix}$  which would be true for all  $x \in \mathbb{R}^2$ .  
 We want to find  $(a, b, c) \in \mathbb{R}^3$  such that  $T(a, b, c) = (a+b, 2a-c) = (1, 11)$ . We have the system of linear equations:

$$a+b=1 \quad \text{and} \quad 2a-c=11,$$

which translates to

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 1 \\ 11 \end{pmatrix}.$$

To solve this, we first compute that  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & -2 & -1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$  has rank 2. The solution set of the corresponding homogeneous system hence has dimension 1 and has  $\left\{ \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \right\}$ . Furthermore,  $\begin{pmatrix} 1 \\ 0 \\ 11 \end{pmatrix}$  is a solution (to the non-homogeneous system). This means that the general solution is  $\left\{ \lambda \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 \\ 0 \\ 11 \end{pmatrix} \mid \lambda \in \mathbb{R} \right\}$ .

7. (a) We compute the ranks:

i.  $\begin{pmatrix} 1 & 1 & -1 & 2 \\ 1 & 1 & 2 & 0 \\ 2 & 2 & 1 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & 0 & 3 & 4 \\ 2 & 0 & 3 & 4 \end{pmatrix}$  has rank 2, while  
 ii.  $\begin{pmatrix} 1 & 1 & -1 & 2 & 2 \\ 1 & 1 & 2 & 0 & 1 \\ 2 & 2 & 1 & 2 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 2 & 1 & 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \end{pmatrix}$  has rank 3.

As such, no solutions exist.

(c) We again conduct rank computation:

$$\begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & -1 \\ 1 & 2 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

tells us the coefficient matrix has identical rank as the augmented matrix.

Consequently, solutions certainly exist.

(e) Once more, compute ranks:

$$\begin{pmatrix} 1 & 2 & -1 \\ 2 & 1 & 2 \\ 1 & -4 & 7 \end{pmatrix} \begin{pmatrix} 1 \\ 3 \\ 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 2 & -3 & 4 \\ 1 & -6 & 8 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \rightarrow \begin{pmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.$$

The rank of the coefficient matrix is unequal to that of the augmented matrix. Thus, no solutions exist.

no. (0,0) or

8. (a)

The condition  $(a+b, b-2c, a+2c) = (1, 3, -2)$  translates to the system of linear equations  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \\ 1 & 0 & 2 \end{pmatrix} x = \begin{pmatrix} 1 \\ 3 \\ -2 \end{pmatrix}$ . We compute the ranks needed

$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \\ 1 & 0 & 2 \end{pmatrix} \begin{matrix} | \\ | \\ | \end{matrix} \begin{matrix} 1 \\ 3 \\ -2 \end{matrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \\ 0 & -1 & 2 \end{pmatrix} \begin{matrix} | \\ | \\ | \end{matrix} \begin{matrix} 1 \\ 3 \\ -3 \end{matrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} \begin{matrix} | \\ | \\ | \end{matrix} \begin{matrix} 1 \\ 3 \\ -1 \end{matrix}$ . By the rank equality, a solution exists.

(b)

Again, we have a system of linear equations  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \end{pmatrix} x = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ . Computing the necessary ranks,  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & -2 \end{pmatrix} \begin{matrix} | \\ | \end{matrix} \begin{matrix} 2 \\ 1 \end{matrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & -2 \end{pmatrix} \begin{matrix} | \\ | \end{matrix} \begin{matrix} 2 \\ 1 \end{matrix}$ , we see that the rank is again equal. Hence, a solution exists.

9. Trivial.

10. When any  $m \times n$  coefficient matrix  $A$  has rank  $m$ , the  $m \times (n+1)$  augmented matrix  $(A|b)$  must also be of equal rank.

Hence, a solution definitely exists.

11. To solve  $(I-A)x = 0$ , we notice that Theorem 3.12 says it has a one-dimensional solution set spanned by a nonnegative vector, which can be  $\begin{pmatrix} 1 \\ 4 \end{pmatrix}$  for example. Thus,  $\frac{1}{1+\frac{1}{4}+1} \begin{pmatrix} 1 \\ 4 \end{pmatrix} = \begin{pmatrix} \frac{1}{6} \\ \frac{2}{3} \end{pmatrix}$  is the required solution which tells us the ratio needed is 4:3.

12. Simply solve  $\begin{pmatrix} 0.6 & 0.3 \\ 0.4 & 0.6 \end{pmatrix} x = 0$  into the required form.

13. We compute the inverse  $(I-A)^{-1}$ , if it exists:

$\begin{pmatrix} 1 & -\frac{1}{5} \\ -\frac{1}{3} & \frac{4}{5} \end{pmatrix} \begin{matrix} | \\ | \end{matrix} \begin{matrix} 1 & 0 \\ 0 & 1 \end{matrix} \rightarrow \begin{pmatrix} 1 & -\frac{1}{5} \\ 0 & \frac{2}{3} \end{pmatrix} \begin{matrix} | \\ | \end{matrix} \begin{matrix} 1 & 0 \\ \frac{2}{3} & 1 \end{matrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{matrix} | \\ | \end{matrix} \begin{pmatrix} \frac{14}{5} & \frac{3}{5} \\ 1 & \frac{3}{2} \end{pmatrix}$

The inverse is thus  $\begin{pmatrix} \frac{14}{5} & \frac{3}{5} \\ 1 & \frac{3}{2} \end{pmatrix}$ . Accordingly, the solution is represented by  $\begin{pmatrix} \frac{14}{5} & \frac{3}{5} \\ 1 & \frac{3}{2} \end{pmatrix} \begin{pmatrix} 2 \\ 5 \end{pmatrix} = \begin{pmatrix} \frac{29}{5} \\ \frac{19}{2} \end{pmatrix}$ .

This means that  $\frac{29}{5}$  units of commodity 1 and  $\frac{19}{2}$  units of commodity 2 should be produced.

f. In the notation of the open model of Leontief, suppose that

$$A = \begin{pmatrix} \frac{1}{2} & \frac{1}{5} \\ \frac{3}{10} & \frac{3}{5} \end{pmatrix} \quad \text{and} \quad d = \begin{pmatrix} 90 \\ 20 \end{pmatrix}.$$

We note that  $(I-A)^{-1} = \begin{pmatrix} \frac{20}{7} & \frac{10}{7} \\ \frac{15}{7} & \frac{25}{7} \end{pmatrix}$ . So,  $(I-A)^{-1}d = \begin{pmatrix} \frac{2000}{7} \\ \frac{1850}{7} \end{pmatrix}$ . which means <sup>the total output of the economic system must be</sup> \$550 billion to support this defense system.





2.(g)

$$\left( \begin{array}{cccc|c} 2 & -2 & -1 & 6 & -2 \\ 1 & -1 & 1 & 2 & -1 \\ 4 & -4 & 5 & 7 & -1 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & -1 & -\frac{1}{2} & 3 & -1 \\ 0 & 0 & \frac{3}{2} & -1 & 0 \\ 0 & 0 & 7 & -5 & 3 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & -1 & 0 & \frac{5}{2} & -1 \\ 0 & 0 & 1 & -\frac{2}{3} & 0 \\ 0 & 0 & 0 & -\frac{1}{3} & 3 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & -1 & 0 & 0 & -\frac{8}{3} \\ 0 & 0 & 1 & 0 & -6 \\ 0 & 0 & 0 & 1 & -9 \end{array} \right)$$

The solution set is, as a result,

$$\left\{ \begin{pmatrix} 8 \\ 0 \\ \frac{23}{9} \\ 0 \\ 0 \end{pmatrix} + s \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} -8 \\ 0 \\ \frac{6}{29} \\ -\frac{9}{29} \\ 1 \end{pmatrix} \mid s, t \in \mathbb{R} \right\}$$

$$\begin{pmatrix} -23 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$t \begin{pmatrix} -23 \\ 0 \\ 6 \\ 9 \\ 1 \end{pmatrix}$$

(f)

$$\left( \begin{array}{cccc|c} 1 & 2 & -1 & 3 & 2 \\ 2 & 4 & -1 & 6 & 5 \\ 0 & 1 & 0 & 2 & 3 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & 2 & -1 & 3 & 2 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 2 & 3 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & 2 & -1 & 3 & 2 \\ 0 & 1 & 0 & 2 & 3 \\ 0 & 0 & 1 & 0 & 1 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & 0 & 0 & -1 & -3 \\ 0 & 1 & 0 & 2 & 3 \\ 0 & 0 & 1 & 0 & 1 \end{array} \right)$$

The solution set is hence

$$\left\{ \begin{pmatrix} 1 \\ 3 \\ 5 \\ 0 \end{pmatrix} + r \begin{pmatrix} -1 \\ -2 \\ 0 \\ 0 \end{pmatrix} \mid r \in \mathbb{R} \right\}$$

$$\left\{ \begin{pmatrix} -3 \\ 3 \\ -1 \\ 0 \end{pmatrix} + r \begin{pmatrix} -1 \\ -3 \\ 0 \\ 1 \end{pmatrix} \mid r \in \mathbb{R} \right\}$$

~~Handwritten scribbles and corrections.~~

near that  $\begin{pmatrix} 0 & -1 & 3 \\ 1 & 1 & 3 \end{pmatrix}$

2. (i) 
$$\begin{pmatrix} 3 & -1 & 2 & 4 & 1 & | & 2 \\ 1 & -1 & 2 & 3 & 1 & | & -1 \\ 2 & -3 & 6 & 9 & 4 & | & -5 \\ 7 & -2 & 4 & 8 & 1 & | & 6 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 2 & 3 & 1 & | & -1 \\ 0 & 2 & -4 & -5 & -2 & | & 5 \\ 0 & -1 & 2 & 3 & 2 & | & -3 \\ 0 & 5 & -10 & -13 & -6 & | & 13 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 & -1 & | & 2 \\ 0 & 1 & -2 & -3 & -2 & | & 3 \\ 0 & 0 & 0 & 1 & 2 & | & -1 \\ 0 & 0 & 0 & 2 & 4 & | & -2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 & -1 & | & 2 \\ 0 & 1 & -2 & 0 & 4 & | & 0 \\ 0 & 0 & 0 & 1 & 2 & | & -1 \\ 0 & 0 & 0 & 0 & 0 & | & 0 \end{pmatrix}$$

solution set: 
$$\left\{ \begin{pmatrix} 2 \\ 0 \\ 0 \\ -1 \\ 0 \end{pmatrix} + r \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + s \begin{pmatrix} 1 \\ 0 \\ 0 \\ -2 \\ 0 \end{pmatrix} \mid r, s \in \mathbb{R} \right\}$$



3. (a) Let  $\text{rank}(A) = r$ . When  $\text{rank}(A') \neq \text{rank}(A'b')$ , the  $(r+1)$ th row of  $A'$  is zero but the same row of  $(A'b')$  must be nonzero by the difference in rank. Hence, this nonzero entry must lie in the last column. Conversely, consider the existence of a row  $k$  in which the only nonzero entry lies in the last column. So, the column  $b$  must be linearly independent of all other columns in  $A'$ . (consequently,  $\text{rank}(A'b') > \text{rank}(A')$ )

by Theorem 3.16 (a) and condition (a)

(b) We see that  $Ax=b$  is consistent iff  $\text{rank}(A) = \text{rank}(A'b)$  iff  $\text{rank}(A') = \text{rank}(A'b')$  iff  $\text{rank}(A') = \text{rank}(A'b')$  iff no row in which the only nonzero entry lies in the last column by the corollary to Theorem 3.13 and part (a) above.

4. (a) We compute the reduced row echelon form:

$$\begin{pmatrix} 1 & 2 & -1 & 1 & | & 2 \\ 2 & 1 & 1 & -1 & | & 3 \\ 1 & 2 & -3 & 2 & | & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & -1 & 1 & | & 2 \\ 0 & -3 & 3 & -3 & | & -1 \\ 0 & 0 & -2 & 1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 1 & -1 & | & \frac{5}{3} \\ 0 & 1 & -1 & 1 & | & \frac{2}{3} \\ 0 & 0 & -2 & 1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & -\frac{1}{2} & | & \frac{11}{6} \\ 0 & 1 & 0 & \frac{1}{2} & | & \frac{2}{3} \\ 0 & 0 & 1 & -\frac{1}{2} & | & 0 \end{pmatrix}$$

So, the solution set of the homogeneous system is  $\left\{ r \begin{pmatrix} -1 \\ -1 \\ -2 \end{pmatrix} \mid r \in \mathbb{R} \right\}$  which has the basis  $\left\{ \begin{pmatrix} -1 \\ -1 \\ -2 \end{pmatrix} \right\}$ .

(c) We see that the reduced row echelon form has a row in which the only nonzero entry lies in the last column:

$$\begin{pmatrix} 1 & 1 & -3 & 1 & | & 1 \\ 1 & 1 & 1 & -1 & | & 2 \\ 1 & 1 & -1 & 0 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & -3 & 1 & | & 1 \\ 0 & 0 & 4 & -2 & | & 1 \\ 0 & 0 & 2 & -1 & | & -1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & -\frac{1}{2} & | & \frac{3}{2} \\ 0 & 0 & 1 & -\frac{1}{2} & | & \frac{1}{4} \\ 0 & 0 & 0 & 0 & | & -\frac{3}{2} \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & -\frac{1}{2} & | & 0 \\ 0 & 0 & 1 & -\frac{1}{2} & | & 0 \\ 0 & 0 & 0 & 0 & | & 1 \end{pmatrix}$$

Therefore, the system is inconsistent.



5. Let the matrix  $B$  be the reduced row echelon form of  $A$ . Then, using the notation of Theorem 3.16,  $j_1 = 1, j_2 = 2, j_3 = 4$ . Consequently, since  $b_1 = 1b_{j_1}, b_2 = 1b_{j_2}$ , and  $b_4 = 1b_{j_3}$ ,  $\{a_1, a_2, a_4\}$  should be linearly independent. To check for this, we row reduce:

$$\left( \begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ -1 & -1 & -2 & 0 \\ 3 & 1 & 0 & 0 \end{array} \right) \rightarrow \left( \begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & -1 & -3 & 0 \\ 0 & 1 & 3 & 0 \end{array} \right) \rightarrow \left( \begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

Indeed, we see that the only solution is  $\begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$ . In other words,  $\left\{ \begin{pmatrix} 1 \\ -1 \\ 3 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix} \right\}$  is linearly independent and can be the first, second, and fourth columns of  $A$ , respectively.

7. As per normal, we row reduce:

$$\left( \begin{array}{ccccc} 2 & 1 & -8 & 1 & -3 \\ -3 & 4 & 12 & 37 & -5 \\ 1 & -2 & -4 & -17 & 8 \end{array} \right) \rightarrow \left( \begin{array}{ccccc} 1 & 3 & -4 & 18 & -11 \\ 0 & -2 & 0 & 88 & -29 \\ 0 & -5 & 0 & -35 & 19 \end{array} \right) \rightarrow \left( \begin{array}{ccccc} 1 & 0 & 8 & -300 & 104 \\ 0 & 1 & -4 & 106 & -40 \\ 0 & 0 & -20 & 405 & -181 \end{array} \right) \rightarrow \left( \begin{array}{ccccc} 1 & 0 & 0 & -462 & \frac{104}{20} \\ 0 & 1 & 0 & 25 & \frac{181}{20} \\ 0 & 0 & 1 & -20 & \frac{181}{20} \end{array} \right)$$

Gr. C. Check

So,  $\{u_1, u_2, u_3\}$  is a basis for  $\mathbb{R}^3$ .

$$\left( \begin{array}{ccccc} 2 & 1 & -8 & 1 & -3 \\ -3 & 4 & 12 & 37 & -5 \\ 1 & -2 & -4 & -17 & 8 \end{array} \right) \rightarrow \left( \begin{array}{ccccc} 1 & 3 & -4 & 18 & -11 \\ 0 & 13 & 0 & 41 & -38 \\ 0 & -5 & 0 & -35 & 19 \end{array} \right) \rightarrow \left( \begin{array}{ccccc} 1 & 0 & -4 & -3 & 0 \\ 0 & 1 & 0 & 7 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right)$$

Try 2

So,  $\{u_1, u_2, u_5\}$  is a basis for  $\mathbb{R}^3$ .

11. (a) It suffices to notice that  $1 - 2(2) + 3(1) - 0 + 2(0) = 1 - 4 + 3 = 0$ , so  $(1, 2, 1, 0, 0) \in V$ . Hence  $\{(1, 2, 1, 0, 0)\}$  is a linearly independent subset of  $V$ .

(b) Vectors in  $V$  have the form  $(x_1, x_2, x_3, x_4, x_5) = (2t_1 - 3t_2 + t_3 - 2t_4, t_1, t_2, t_3, t_4) = t_1(2, 1, 0, 0, 0) + t_2(-3, 0, 1, 0, 0) + t_3(1, 0, 0, 0, 0) + t_4(-2, 0, 0, 0, 1)$ . It is clear that  $\{(2, 1, 0, 0, 0), (-3, 0, 1, 0, 0), (1, 0, 0, 0, 0), (-2, 0, 0, 0, 1)\}$  spans  $V$  by having  $t_i = x_i$ , and linear independence is trivial by noticing each vector in the above set has a nonzero entry that is zero in all other vectors. In other words, this set is a basis for  $V$ .

By row-reducing  $\begin{pmatrix} 2 & 1 & -3 & -2 & 0 \\ -3 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 1 \end{pmatrix}$ , we got  $\begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & -2 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$ . Which means  $\{(1, 2, 1, 0, 0), (2, 1, 0, 0, 0), (1, 0, 0, 0, 0), (-2, 0, 0, 0, 1)\}$  is a basis for  $V$ .

2. (a) First, notice that  $S \subseteq V$  as expected since  $0 - (-1) + 2 - 3 + 0 = 0$  and  $2(0) - (-1) - 0 + 3 - 4 + 4(0) = 0$ ;  $1 - 0 + 2 - 3 + 0 = 0$  and  $2(1) - 0 - 1 + 3 - 4 + 4(0) = 0$ .  
 Furthermore, linear independence is ensured as  $a(0, -1, 0, 1, 1, 0) + b(1, 0, 1, 1, 1, 0) = (b, -a, b, a+b, a+b, 0) = 0$  means  $-a = b = 0$ .

(b) As usual, we now reduce the coefficient matrix to find a basis for the solution space:  

$$\begin{pmatrix} 1 & -1 & 0 & 2 & -3 & 1 \\ 2 & -1 & -1 & 3 & -4 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 0 & 2 & -3 & 1 \\ 0 & 1 & -1 & -1 & 2 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -1 & 1 & -1 & 3 \\ 0 & 1 & -1 & -1 & 2 & 2 \end{pmatrix}$$

An equivalent system is hence

$$\begin{aligned} x_1 - x_3 + x_4 - x_5 + 3x_6 &= 0 \\ x_2 - x_3 - x_4 + 2x_5 + 2x_6 &= 0 \end{aligned}$$

Let  $t_i = x_{i+2}$ , then any solution  $(x_1, x_2, x_3, x_4, x_5, x_6) = (t_1 - t_2 + t_3 - 3t_4, t_1 + t_2 - 2t_3 - 2t_4, t_1, t_2, t_3, t_4) =$   
 $t_1(1, 1, 1, 0, 0, 0) + t_2(-1, 1, 0, 1, 0, 0) + t_3(1, -2, 0, 0, 1, 0) + t_4(-3, -2, 0, 0, 0, 1)$ . By Theorem 3.15, the set  
 $\{(1, 1, 1, 0, 0, 0), (-1, 1, 0, 1, 0, 0), (1, -2, 0, 0, 1, 0), (-3, -2, 0, 0, 0, 1)\}$  is a basis for  $V$ . Again, we now reduce to  
 find that the reduced row echelon form of  $B$ : the new matrix with the first columns the vectors from  $S$  and the rest of the columns the basis vectors above.

$$\begin{pmatrix} 0 & 1 & 1 & -1 & -1 & -3 \\ -1 & 0 & 1 & 1 & -2 & -2 \\ 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \text{ is } \begin{pmatrix} 1 & 0 & -1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$B$ : by looking at the pivot columns

As such, it is clear from Theorem 3.16 (d) that an extension of  $S$  to a basis for  $V$  is the set  
 $\{(0, -1, 0, 1, 1, 0), (1, 0, 1, 1, 1, 0), (-1, 1, 0, 1, 0, 0), (-3, -2, 0, 0, 0, 1)\}$ .

Ideas

Let  $B$  and  $C$  be the reduced row echelon form of  $A$ .

$$\begin{aligned} MA &= NA \\ M &= N \end{aligned}$$

exactly  $r$  nonzero rows  $b_{ji}$  of  $B$  and  $c_{ki}$  of  $C$

Proof

Let  $A$  be a  $1 \times n$  matrix, then the row reduced echelon form of  $A$  is either  $0$  or  $e_1$ , and these are non overlapping cases. Hence, uniqueness holds for  $n=1$ .  
 Assume that this is true of any  $m \times n$  matrix and let  $A$  now be a  $(m+1) \times n$  matrix instead. Furthermore, let  $A_m$  be the  $m \times n$  matrix with  $(A_m)_{ij} = A_{ij}$  and  $a_{m+1}$  the  $(m+1)$ th row of  $A$ . If  $M$  and  $N$  are invertible matrices so that  $MA$  and  $NA$  are the reduced row echelon forms of  $A$ , then  $MA_m$  and  $NA_m$  are the reduced row echelon forms of  $A_m$ , by exercise 15 of section 3.2 and exercise 14 of this section. By our assumption / induction hypothesis,  $MA_m = NA_m$ . Thus,  $MA = NA$ .

Vector  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  refers to basis  $\begin{pmatrix} 0 & 13 & 4 \\ -5 & 0 & 18 \end{pmatrix}$ .  $\frac{15}{-38}$   $\frac{4}{-22}$   $\begin{pmatrix} 0 & -20 & 106 & 109 \\ & 405 & -40 & \end{pmatrix}$   $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$

15. Let  $A$  be any  $n \times n$  matrix, and use the notation of Theorem 3.16. So the  $k$ th column of  $A$  is  $\sum_{i=1}^n d_{ki} a_{ji}$ . Then, for any two (possibly distinct) reduced row echelon forms  $B$  and  $C$  of  $A$ , the  $k$ th columns of  $B$  and  $C$  are both just  $\sum_{i=1}^n d_{ki} a_{ji}$ . Since this holds for all  $1 \leq k \leq n$  columns of  $B$  and  $C$ ,  $B=C$  must consequently hold true. So, the reduced row echelon form of any matrix is unique. □

Dude what was I thinking bruh.  
 The indexing  $j_i$  may be different for  $B$  and  $C$ .

X



# Self-Proof of Lemma

Idea

$$\det(B) = \sum_{j=1}^n (-1)^{1+j} B_{1j} \det(\tilde{B}_{2j})$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} \quad \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^{e_1} (-1)^{1+1} \cdot 1 = 1$$

$$(-1)^{1+1} \det \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + (-1)^{1+2} \det \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} + (-1)^{1+3} \det \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 1 & 0 & 0 \\ 3 & 2 & 1 \end{pmatrix} \quad (-1)^{2+1} \det \begin{pmatrix} 2 & 3 \\ 2 & 1 \end{pmatrix} = (-1)(-4) = 4$$

$$1 \cdot (-1)^{1+1} \det \begin{pmatrix} 0 & 0 \\ 2 & 1 \end{pmatrix} + 2 \cdot (-1)^{1+2} \det \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} + 3 \cdot (-1)^{1+3} \det \begin{pmatrix} 1 & 0 \\ 3 & 2 \end{pmatrix} = -2 + 6 = 4$$

$i=1$ : there's only 1 nonzero summand

$n=2$  Simple case analysis

Assume 1, when  $i \geq 2$ :

$$\sum_{j=1}^n (-1)^{i+j} B_{ij} \det(\tilde{B}_{2j}) = (-1)^{i+k} \sum_{j=1}^n (-1)^j B_{ij} \det(\tilde{B}_{2j})_{i-1,k}$$

$$\det(\tilde{B}_{ik}) = \sum_{j=1}^n (-1)^j \det(\tilde{B}_{ik})_{2j}$$

$$= \sum_{j=1}^n (-1)^{2+j} (b'_{ik})_{2j} \det(\tilde{B}_{ik})_{2j}$$

## Proof

When  $n=2$ , a simple case analysis suffices:

- $i=1 \neq k=1$ :  $(-1)^{1+1} \det((d)) = d = \det \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ ,
- $i=1 \neq k=2$ :  $(-1)^{1+2} \det((c)) = -c = \det \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ ,

$$3. i=2 \neq k=1: (-1)^{2+1} \det((b)) = -b = \det \begin{pmatrix} a & b \\ c & d \end{pmatrix},$$

$$4. i=2 \neq k=2: (-1)^{2+2} \det((a)) = a = \det \begin{pmatrix} a & b \\ c & d \end{pmatrix},$$

for any scalars  $a, b, c, d \in \mathbb{F}$ . Clearly, the result holds for  $n=2$ . So assume it holds for  $n \geq 2$ . Tailoring the trivial case of  $i=1$  first, we see that  $\det(B) = \sum_{j=1}^n (-1)^{1+j} B_{1j} \det(\tilde{B}_{2j}) = (-1)^{1+k} \det(\tilde{B}_{2j})_{i-1,k}$ . If  $i \geq 2$ , then for any row  $b_j \in \mathbb{F}^n$  of  $B$ , let the row vectors  $\tilde{b}_j \in \mathbb{F}^n$ ,  $\tilde{b}_j \in \mathbb{F}^n$ ,  $\tilde{b}_j \in \mathbb{F}^n$  be  $b_j$  but with the  $j$ th entry, the  $k$ th entry, and the  $j$ th and  $k$ th entries, respectively removed. As such, using our assumption / induction hypothesis (I.H.) and the Corollary to Theorem 4.3 which tells us  $\det \begin{pmatrix} \tilde{b}_1 \\ \vdots \\ \tilde{b}_k \\ \vdots \\ \tilde{b}_n \end{pmatrix} = 0$  for  $j=k$ ,  $\det(B) = \sum_{j=1}^n (-1)^{1+j} B_{1j} \det \begin{pmatrix} \tilde{b}_1 \\ \vdots \\ \tilde{b}_j \\ \vdots \\ \tilde{b}_n \end{pmatrix} = (-1)^{1+k} \sum_{j=0}^n (-1)^j B_{1j} \det \begin{pmatrix} \tilde{b}_1 \\ \vdots \\ \tilde{b}_{j-1} \\ \vdots \\ \tilde{b}_{j+1} \\ \vdots \\ \tilde{b}_n \end{pmatrix} = (-1)^{1+k} \det(\tilde{B}_{ik})$ , as expected. As a result, the claim is true for  $n+1$ . Consequently, it is true for all natural numbers  $n \geq 2$ .

Oops had a really dumb brain leading me astray. Write some really dumb stuff egh.

# Self-proof of Theorem 4.3

Idea  
 When  $n=1$ , trivial  
 Assume  $n$

$$\det \begin{pmatrix} a_{11} & \dots & a_{1j} & \dots & a_{1n} \\ \vdots & & \vdots & & \vdots \\ a_{i1} & \dots & a_{ij} & \dots & a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{n1} & \dots & a_{nj} & \dots & a_{nn} \end{pmatrix} = \sum_{i=1}^{n+1} (-1)^{i+j} a_{ij} \det(\tilde{A}_{2j})$$

$$\tilde{A}_{2j} = \begin{pmatrix} a_{11} & \dots & a_{1j} & \dots & a_{1n} \\ \vdots & & \vdots & & \vdots \\ a_{i1} & \dots & a_{ij} & \dots & a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{n1} & \dots & a_{nj} & \dots & a_{nn} \end{pmatrix}$$

Proof  
 When  $n=1$ , the result is trivial as  $\det(a+kb) = a+kb = \det(a) + k \det(b)$  for any  $a, b, k \in \mathbb{F}$ . Assume this holds for  $n \in \mathbb{N}$ . To show the same is true of

$n+1$ , first for any row vector  $v \in \mathbb{F}^{n+1}$ , let  $\tilde{v} \in \mathbb{F}^n$  be  $v$  but with the  $j$ th entry completely removed, so that  $\begin{pmatrix} a_1 \\ \vdots \\ a_{j-1} \\ v \\ a_{j+1} \\ \vdots \\ a_{n+1} \end{pmatrix} = \begin{pmatrix} a_1 \\ \vdots \\ a_{j-1} \\ \tilde{v} \\ a_{j+1} \\ \vdots \\ a_{n+1} \end{pmatrix}$  by noticing  $(v)_j = a_j + kv_j$ .  
 Hence,  $\det \begin{pmatrix} a_1 \\ \vdots \\ a_{j-1} \\ v \\ a_{j+1} \\ \vdots \\ a_{n+1} \end{pmatrix} = \sum_{i=1}^{n+1} (-1)^{i+j} (a_i)_{2j} \det \begin{pmatrix} \tilde{a}_1 \\ \vdots \\ \tilde{a}_{i-1} \\ \tilde{a}_{i+1} \\ \vdots \\ \tilde{a}_{n+1} \end{pmatrix} = \sum_{i=1}^{n+1} (-1)^{i+j} (a_i)_{2j} \left[ \det \begin{pmatrix} \tilde{a}_1 \\ \vdots \\ \tilde{a}_{i-1} \\ \tilde{a}_{i+1} \\ \vdots \\ \tilde{a}_{n+1} \end{pmatrix} + k \det \begin{pmatrix} \tilde{a}_1 \\ \vdots \\ \tilde{a}_{i-1} \\ \tilde{a}_{i+1} \\ \vdots \\ \tilde{a}_{n+1} \end{pmatrix} \right] = \sum_{i=1}^{n+1} (-1)^{i+j} (a_i)_{2j} \det \begin{pmatrix} \tilde{a}_1 \\ \vdots \\ \tilde{a}_{i-1} \\ \tilde{a}_{i+1} \\ \vdots \\ \tilde{a}_{n+1} \end{pmatrix} + k \sum_{i=1}^{n+1} (-1)^{i+j} (a_i)_{2j} \det \begin{pmatrix} \tilde{a}_1 \\ \vdots \\ \tilde{a}_{i-1} \\ \tilde{a}_{i+1} \\ \vdots \\ \tilde{a}_{n+1} \end{pmatrix} = \det \begin{pmatrix} a_1 \\ \vdots \\ a_{j-1} \\ v \\ a_{j+1} \\ \vdots \\ a_{n+1} \end{pmatrix} + k \det \begin{pmatrix} a_1 \\ \vdots \\ a_{j-1} \\ kv_j \\ a_{j+1} \\ \vdots \\ a_{n+1} \end{pmatrix}$

as expected. In other words, the claim holds for  $n+1$ . By induction, this is true of all natural  $n$ .

In the case that  $r=1$ , the result is immediate as  $(a_i)_{2j} = (u+kv)_{2j} = u_{2j} + kv_{2j}$  and  $\begin{pmatrix} u+kv \\ a_1 \\ \vdots \\ a_{n+1} \end{pmatrix}_{2j} = \begin{pmatrix} u \\ a_1 \\ \vdots \\ a_{n+1} \end{pmatrix}_{2j} = \begin{pmatrix} v \\ a_1 \\ \vdots \\ a_{n+1} \end{pmatrix}_{2j}$

## Self-Proof of Corollary / Exercise 24

As in Theorem 4.3, let  $a_j$  denote the  $j$ th row of  $A$  and suppose  $a_r = 0$  for some  $1 \leq r \leq n$ . Now,  $\det(A) = \det \begin{pmatrix} a_1 \\ \vdots \\ a_{r-1} \\ 0 \\ a_{r+1} \\ \vdots \\ a_n \end{pmatrix} = \det \begin{pmatrix} a_1 \\ \vdots \\ a_{r-1} \\ 0 \\ a_{r+1} \\ \vdots \\ a_n \end{pmatrix} - \det \begin{pmatrix} a_1 \\ \vdots \\ a_{r-1} \\ 0 \\ a_{r+1} \\ \vdots \\ a_n \end{pmatrix} = 0$  as expected.  $\square$



4. (a) Area =  $|\det \begin{pmatrix} 3 & -2 \\ 2 & 5 \end{pmatrix}| = |15+4| = 19 \text{ units}^2$  ✓

(b) Area =  $|\det \begin{pmatrix} 1 & 3 \\ -3 & 1 \end{pmatrix}| = |1+9| = 10 \text{ units}^2$  ✓

(c) Area =  $|\det \begin{pmatrix} 4 & -1 \\ -6 & -2 \end{pmatrix}| = |-8-6| = 14 \text{ units}^2$  ✗

(d) Area =  $|\det \begin{pmatrix} 3 & 4 \\ 2 & -6 \end{pmatrix}| = |-18-8| = 26 \text{ units}^2$  ✓

5. Let  $A := \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , so  $B = \begin{pmatrix} c & d \\ a & b \end{pmatrix}$ . Hence,  $\det(B) = cb - da = -(ad - bc) = -\det(A)$ .

6. In that case,  $A$  is of the form  $\begin{pmatrix} a & a \\ b & b \end{pmatrix}$  so  $\det(A) = ab - ab = 0$  as expected.

7. Once more, let  $A := \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , so  $A^t = \begin{pmatrix} a & c \\ b & d \end{pmatrix}$ . Hence,  $\det(A^t) = ad - cb = ad - bc = \det(A)$ .

8. Given  $A$  is diagonal,  $A = \begin{pmatrix} a & 0 \\ 0 & d \end{pmatrix}$  for some scalars  $a, d$  in  $\mathbb{F}$ . Accordingly,  $\det(A) = ad - 0 \cdot 0 = ad$ , as we wanted.

9. Method 1

Let  $A := \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix}$  and  $B = \begin{pmatrix} b_1 & b_2 \\ b_3 & b_4 \end{pmatrix}$ . Thus,  $AB = \begin{pmatrix} a_1 b_1 + a_2 b_3 & a_1 b_2 + a_2 b_4 \\ a_3 b_1 + a_4 b_3 & a_3 b_2 + a_4 b_4 \end{pmatrix}$ . Hence,  $\det(AB) = (a_1 b_1 + a_2 b_3)(a_3 b_2 + a_4 b_4) - (a_1 b_2 + a_2 b_4)(a_3 b_1 + a_4 b_3) = a_1 b_1 a_3 b_2 + a_1 b_1 a_4 b_4 + a_2 b_3 a_3 b_2 + a_2 b_3 a_4 b_4 - (a_1 b_2 a_3 b_1 + a_1 b_2 a_4 b_3 + a_2 b_4 a_3 b_1 + a_2 b_4 a_4 b_3) = a_1 b_1 a_3 b_2 + a_1 b_1 a_4 b_4 + a_2 b_3 a_3 b_2 + a_2 b_3 a_4 b_4 - a_1 b_2 a_3 b_1 - a_1 b_2 a_4 b_3 - a_2 b_4 a_3 b_1 - a_2 b_4 a_4 b_3 = a_1 a_4 (b_1 b_4 - b_2 b_3) - a_2 a_3 (b_1 b_4 - b_2 b_3) = (a_1 a_4 - a_2 a_3)(b_1 b_4 - b_2 b_3) = \det(A) \det(B)$ .

Method 2 (Probably more easily generalizable + elegance)

Idea:  
 $p_i = \sum_{j=1}^2 A_{ij} b_j$   
 $\det \begin{pmatrix} p_1 \\ p_2 \end{pmatrix} = \sum_{j=1}^2 A_{2j} \det \begin{pmatrix} b_j \\ p_2 \end{pmatrix} = \sum_{j=1}^2 A_{2j} \sum_{k=1}^2 A_{1k} \det \begin{pmatrix} b_j \\ b_k \end{pmatrix}$   
 $= A_{11} A_{22} \det \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} + A_{12} A_{21} \det \begin{pmatrix} b_2 \\ b_1 \end{pmatrix}$  as  $\det \begin{pmatrix} b_1 \\ b_1 \end{pmatrix} = \det \begin{pmatrix} b_2 \\ b_2 \end{pmatrix} = 0$   
 $= (A_{11} A_{22} - A_{12} A_{21}) \det \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$  by exercise 5

Proof

Let  $p_i$  and  $b_i$  be the  $i$ th rows of  $AB$  and  $B$  respectively. We know  $p_i = \sum_{j=1}^2 A_{ij} b_j$ , therefore

$\det(AB) = \det \begin{pmatrix} p_1 \\ p_2 \end{pmatrix} = \sum_{j=1}^2 A_{2j} \sum_{k=1}^2 A_{1k} \det \begin{pmatrix} b_j \\ b_k \end{pmatrix}$  by Theorem 4.1,  
 $= A_{11} A_{22} \det \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} + A_{12} A_{21} \det \begin{pmatrix} b_2 \\ b_1 \end{pmatrix}$  as  $\det \begin{pmatrix} b_1 \\ b_1 \end{pmatrix} = \det \begin{pmatrix} b_2 \\ b_2 \end{pmatrix} = 0$  by exercise 6 above,  
 $= (A_{11} A_{22} - A_{12} A_{21}) \det \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$  since  $\det \begin{pmatrix} b_2 \\ b_1 \end{pmatrix} = -\det \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$  by exercise 5 above,  
 $= \det(A) \det(B)$ .

Exercises

$0 < \alpha < \pi \Rightarrow 0 < \sin(\alpha) \leq 1$

13. Ideas

Example 2 of 52.1:  $T_\alpha(u) = v \Rightarrow (u_1 \cos(\alpha) - u_2 \sin(\alpha), u_1 \sin(\alpha) + u_2 \cos(\alpha)) = (v_1, v_2)$

$$\det \begin{pmatrix} u_1 \cos(\alpha) - u_2 \sin(\alpha) & u_1 \sin(\alpha) + u_2 \cos(\alpha) \\ u_1 \sin(\alpha) + u_2 \cos(\alpha) & u_1 \cos(\alpha) - u_2 \sin(\alpha) \end{pmatrix} = u_1^2 \sin(\alpha) + u_1 u_2 \cos(\alpha) - u_1 u_2 \cos(\alpha) + u_2^2 \sin(\alpha) = (u_1^2 + u_2^2) \sin(\alpha)$$

sign 1  $u^2 + v^2$

Proof

Let  $\alpha$  be the angle between  $u$  and  $v$ , not necessarily satisfying  $0 < \alpha < \pi$ . By example 2 of section 2.1, we see that

$$\det \begin{pmatrix} u \\ v \end{pmatrix} = \det \begin{pmatrix} u_1 \cos(\alpha) - u_2 \sin(\alpha) & u_1 \sin(\alpha) + u_2 \cos(\alpha) \\ u_1 \sin(\alpha) + u_2 \cos(\alpha) & u_1 \cos(\alpha) - u_2 \sin(\alpha) \end{pmatrix} = u_1^2 \cos(\alpha) + u_1 u_2 \cos(\alpha) - u_1 u_2 \cos(\alpha) + u_2^2 \sin(\alpha) = (u_1^2 + u_2^2) \sin(\alpha)$$

That is,  $\mathcal{O} \begin{pmatrix} u \\ v \end{pmatrix} = 1$  iff  $0 < \alpha < \pi$ , to ensure  $\sin(\alpha) > 0$ .

1. (a) False ✓

(b) True ✓

(c) False ✓

(d) False ✓

(e) True ✓

2. (a)  $\det \begin{pmatrix} 6 & -3 \\ 2 & 4 \end{pmatrix} = 6 \cdot 4 + 3 \cdot 2 = 24 + 6 = 18 \cdot 30$

(b)  $\det \begin{pmatrix} -5 & 2 \\ 6 & 1 \end{pmatrix} = -5 - 12 = -17$  ✓

(c)  $\det \begin{pmatrix} 8 & 0 \\ 3 & -1 \end{pmatrix} = -8 - 0 = -8$  ✓

3. (a)  $\det \begin{pmatrix} -1+i & 1-4i \\ 3+2i & 2-3i \end{pmatrix} = (-1+i)(2-3i) - (1-4i)(3+2i) = -2+3+3i+2i - (3+8-12i+2i) = -10+15i$  ✓

(b)  $\det \begin{pmatrix} 5-2i & 6+4i \\ -3+i & 7i \end{pmatrix} = 35i+14 + (3-i)(6+4i) = 14+35i+18+4-6i+12i = 36+41i$  ✓

(c)  $\det \begin{pmatrix} 2i & 3 \\ 4 & 6i \end{pmatrix} = -12-12 = -24$  ✓

1. (a) Let  $A := \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  once again, then  $C = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$ . Consequently,  $AC = \begin{pmatrix} ad-bc & -ab+ba \\ cd-dc & -cb+da \end{pmatrix} = \begin{pmatrix} ad-bc & 0 \\ 0 & ad-bc \end{pmatrix} = \begin{pmatrix} ad-bc & 0 & 0 \\ 0 & ad-bc & 0 \\ 0 & 0 & ad-bc \end{pmatrix} = (A)$ , which is clearly just  $\det(A) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = [\det(A)]I$ .

~~(b) From Theorem 4.1 and exercise 9 of this section,  $\det([\det(A)]I) = \det(A) \begin{pmatrix} 1 & 0 \\ 0 & \det(A) \end{pmatrix} = [\det(A)]^2 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \det(AC) = \det(A)\det(C)$ . Hence,  $\det(C) = \det(A)$  when  $\det(A) \neq 0$ .~~

~~Even if  $\det(A) = 0$ ,  $\det(C) = ad+bc$~~

(b)  $\det(A) = ad-bc = da - (-b)(-c) = \det(C)$ .

(c) The classical adjoint of  $A^t = \begin{pmatrix} a & c \\ b & d \end{pmatrix}$  is  $\begin{pmatrix} d & -c \\ -b & a \end{pmatrix}$ , i.e.  $C^t$ .

(d) If  $A$  is invertible, Theorem 4.2 immediately says  $A^{-1} = [\det(A)]^{-1}C$ .

(a) If  $E$  is a type 2 elementary matrix, then we can suppose without loss of generality that  $E = \begin{pmatrix} a & 0 \\ 0 & 1 \end{pmatrix}$  for some scalar  $a \in \mathbb{F}$ . Hence,  $\delta(E) \stackrel{(ii)}{=} a \delta \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \stackrel{(iii)}{=} a_1 = \det(E)$ .

When  $E$  is a type 3 elementary matrix, then the existence of some scalar  $b \in \mathbb{F}$  with  $E = \begin{pmatrix} 1 & b \\ 0 & 1 \end{pmatrix}$  is something we can assume without loss of generality.

Therefore,  $\delta(E) \stackrel{(ii)}{=} \delta \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + b \delta \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} \stackrel{(iii)}{=} 1 = \det(E)$ .

Finally, given  $E$  is a type 1 elementary matrix, it is certain that  $E = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ . Therefore,  $\delta(E) \stackrel{(i)}{=} \delta \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} - \delta \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \stackrel{(ii)}{=} \delta \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} - \delta \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \stackrel{(iii)}{=} -1 = \det(E)$  since  $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$  is a type 3 elementary matrix.

So, as long as  $E$  is an elementary matrix,  $\delta(E) = \det(E)$ .

(b) Let us follow the notation and work done in exercise 9 of this section: we know that

$$\delta(EB) = E_{11}E_{22} \delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} + E_{22}E_{11} \delta \begin{pmatrix} b_2 \\ b_1 \end{pmatrix}.$$

Now, notice that  $\delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \stackrel{(ii)}{=} \delta \begin{pmatrix} b_2-b_1 \\ b_1 \end{pmatrix} + \delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \stackrel{(iii)}{=} \delta \begin{pmatrix} b_2-b_1 \\ b_2 \end{pmatrix} + \delta \begin{pmatrix} b_2-b_1 \\ b_1-b_2 \end{pmatrix} \stackrel{(i)}{=} \delta \begin{pmatrix} b_2 \\ b_2 \end{pmatrix} - \delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} + \delta \begin{pmatrix} b_1-b_2 \\ b_1-b_2 \end{pmatrix} \stackrel{(ii)}{=} -\delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$ . Hence,  $\delta(EB) = (E_{11}E_{22} - E_{22}E_{11}) \delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \det(E) \delta \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \stackrel{(ca)}{=} \delta(E) \delta(B)$ .

First, if  $A$  is invertible, then  $A = \prod_{i=1}^k E_i$  for some natural  $k$  and corresponding elementary matrices  $E_i$  by Corollary 3 of Theorem 3.6. As such,

$$\delta(A) \stackrel{(1)(b)}{=} \prod_{i=1}^k \delta(E_i) \stackrel{(1)(a)}{=} \prod_{i=1}^k \det(E_i);$$

which is just  $\det(A)$  according to exercise 9 of this section. When  $A$  is not invertible, it reduces to the form

$\begin{pmatrix} a & 0 \\ 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} a & 0 \\ a & 0 \end{pmatrix}$  by elementary row operations. Thus,  $\delta(A) \stackrel{(iii)}{=} 0 = \det(A)$ .

First = det ...  $\det(E)$  ...  $\det \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 0 & \dots & 1 \end{pmatrix}$  ...

Self-proof of Theorem 4.4

Idea: Let  $B_j$  be  $A$  but with its  $i$ th row being  $e_j$ .  $\rightarrow a_i = \sum_{j=1}^n A_{ij} e_j$   
 $\det(A) = \sum_{j=1}^n A_{ij} \det(B_j) = \sum_{j=1}^n (-1)^{i+j} \det(\tilde{A}_{ij})$  as  $i$ th column is deleted so  $(B_j)_{ij} = \tilde{A}_{ij}$   
 $\det(A) = \sum_{j=1}^n A_{ij} \det(B_j) = \sum_{j=1}^n (-1)^{i+j} \det(\tilde{A}_{ij})$  as  $\det(A) = \sum_{j=1}^n A_{ij} \det(B_j) = \sum_{j=1}^n (-1)^{i+j} \det(\tilde{A}_{ij})$

Proof: Let  $B_j$  be  $A$  but with its  $i$ th row being  $e_j$  instead, so by virtue of the  $i$ th row of  $A$  being  $\sum_{j=1}^n A_{ij} e_j$ , we know  $\det(A) = \sum_{j=1}^n A_{ij} \det(B_j) = \sum_{j=1}^n (-1)^{i+j} \det(\tilde{A}_{ij}) = \sum_{j=1}^n (-1)^{i+j} \det(\tilde{A}_{ij})$  because  $B_j$  and  $A$  only differ in the  $i$ th row, and since that is removed, we have  $(B_j)_{ij} = \tilde{A}_{ij}$ . □

Self-proof of Corollary

Idea:  $\begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix} = (-1)^{1+1} (1) \det \begin{pmatrix} 1 & 1 \\ 2 & 3 \end{pmatrix} + (-1)^{1+2} (2) \det \begin{pmatrix} 1 & 1 \\ 1 & 3 \end{pmatrix}$

Proof: When  $n=2$ , the result is clear from noticing that  $\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc = 0$  for any  $a, b, c, d \in \mathbb{F}$ . And  $n=1$  is vacuous. So, assume the statement is true for  $n-1 \geq 2$ , and  $A \in M_{nn}(\mathbb{F})$  are identical through  $i \neq j$ . Let  $1 \leq k \leq n$  so  $k \neq i$  and  $k \neq j$ . Then, by Theorem 4.4,  $\det(A)$  is just the cofactor expansion of  $A$  along the  $i$ th row:  $\det(A) = \sum_{k=1}^n (-1)^{i+k} A_{ik} \det(\tilde{A}_{ik})$ . Since the  $i$ th and  $j$ th rows of  $\tilde{A}_{ik}$  must be identical,  $\det(\tilde{A}_{ik}) = 0$  by assumption / the induction hypothesis, for each  $1 \leq k \leq n$ . Consequently,  $\det(A)$  is a sum of zeros and must be 0. In other words, the claim holds for  $n$  too. Therefore, induction says that this is true for all  $n \in \mathbb{N}$ . □

### Self-Proof of Theorem 4.5 (Continuation)

As such, to represent  $\tilde{B}_{kj}$  in a form suitable to apply our induction hypothesis, let  $R_{pq}(A)$  be the matrix obtained from  $A$  by interchanging rows  $p$  and  $q$ . Note that  $\tilde{A}_{ij} = R_{k-2, k-1} R_{k-3, k-2} \dots R_{i, i+1}(\tilde{B}_{kj})$ , a sequence of  $k-1-i$  elementary row operations of type 1. As such,  $\det(\tilde{A}_{ij}) = (-1)^{k-1-i} \det(\tilde{B}_{kj})$  by our assumption / induction hypothesis. Thus,  $\det(A) := \sum_{j=1}^n (-1)^{i+j} A_{ij} = \det(\tilde{A}_{ij}) = - \sum_{j=1}^n (-1)^{k+j} B_{ij} = \det(B)$ , as expected. So, the determinants also holds for  $n$ . Consequently, it must hold true for all  $n \in \mathbb{N}$ . □

### Self-Proof of Theorem 4.6

Let  $a_j$  be the  $j$ th row of  $A$ . Say  $B$  is a matrix obtained by adding  $c$  times of the  $i$ th row to the  $j$ th row of  $A$ . Now, by Theorem 4.5 and the Corollary to Theorem 4.4

$$\det(B) = \det \begin{pmatrix} a_1 \\ \vdots \\ a_i + ca_j \\ \vdots \\ a_j \\ \vdots \\ a_n \end{pmatrix} \stackrel{\text{Th 4.5}}{=} c \det \begin{pmatrix} a_1 \\ \vdots \\ a_i \\ \vdots \\ a_j \\ \vdots \\ a_n \end{pmatrix} + \det \begin{pmatrix} a_1 \\ \vdots \\ a_i \\ \vdots \\ 0 \\ \vdots \\ a_n \end{pmatrix} \stackrel{\text{Corollary}}{=} 0 + \det \begin{pmatrix} a_1 \\ \vdots \\ a_i \\ \vdots \\ a_j \\ \vdots \\ a_n \end{pmatrix} = \det(A) \text{ as we thought.}$$

(Here  $i > j$  is shown but the display of  $a_i$  right above  $a_n$  is just for the sake of clarity. It is easy to notice that the above argument holds when  $j > i$ .)

### Self-Proof of Corollary

As before, let  $a_j$  be the  $j$ th row of  $A$ . Given  $A$  has rank less than  $n$ ,  $a_j = \sum_{i=1}^n c_i a_i$  for some scalars  $c_i \in \mathbb{F}$  and  $1 \leq j \leq n$ . Hence, we notice that by the Corollary to Theorem 4.4

$$\det(A) \stackrel{\text{Th 4.6}}{=} \det \begin{pmatrix} a_1 \\ \vdots \\ a_j - \sum_{i=1}^n c_i a_i \\ \vdots \\ a_n \end{pmatrix} = \det \begin{pmatrix} a_1 \\ \vdots \\ 0 \\ \vdots \\ a_n \end{pmatrix} \stackrel{\text{Corollary}}{=} 0. \quad \checkmark$$

# Exercises

- 1. (a) False ✓
- (b) True ✓
- (c) True ✓
- (d) True ✓
- (e) False ✓
- (f) False ✓
- (g) False ✓
- (h) True ✓

Note: To save time, I'll avoid writing (computations) out in the same way I do with proofs, and instead just, well, compute.

2. We notice that for row vectors  $A, B, C \in \mathbb{R}^3$

$$\det \begin{pmatrix} 3A \\ 3B \\ 3C \end{pmatrix} = 3 \det \begin{pmatrix} A \\ B \\ C \end{pmatrix} = 3^3 \det \begin{pmatrix} A \\ B \\ C \end{pmatrix}.$$

Hence, the desired  $k$  is 27.

3. By using the information about how row operations affect determinants mentioned in page 217,

$$\det \begin{pmatrix} 2a_1 & 2a_2 & 2a_3 \\ 3b_1+5c_1 & 3b_2+5c_2 & 3b_3+5c_3 \\ 7c_1 & 7c_2 & 7c_3 \end{pmatrix} = (2)(3)(7) \det \begin{pmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{pmatrix}.$$

So,  $k = 42$ .

4. Again, repeating a similar procedure,

$$\det \begin{pmatrix} b_1+c_1 & b_2+c_2 & b_3+c_3 \\ c_1+c_1 & c_2+c_2 & c_3+c_3 \\ a_1+b_1 & a_2+b_2 & a_3+b_3 \end{pmatrix} = \det \begin{pmatrix} b_1+c_1 & b_2+c_2 & b_3+c_3 \\ a_1+c_1 & a_2+c_2 & a_3+c_3 \\ b_1-c_1 & b_2-c_2 & b_3-c_3 \end{pmatrix} = 2 \det \begin{pmatrix} b_1 & b_2 & b_3 \\ a_1+c_1 & a_2+c_2 & a_3+c_3 \\ c_1 & c_2 & c_3 \end{pmatrix} = 2 \det \begin{pmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{pmatrix}.$$

Thus,  $k = 2$ .

$$5. \det \begin{pmatrix} 0 & 1 & 2 \\ -1 & 0 & -3 \\ 2 & 3 & 0 \end{pmatrix} = 0 + 1 \cdot \det \begin{pmatrix} -1 & -3 \\ 2 & 0 \end{pmatrix} + 2 \cdot \det \begin{pmatrix} -1 & 0 \\ 2 & 3 \end{pmatrix} = +(-6) + 2(-3) = -6 - 6 = -12$$

$$7. \det \begin{pmatrix} 0 & 1 & 2 \\ -1 & 0 & -3 \\ 2 & 3 & 0 \end{pmatrix} = -1 \cdot \det \begin{pmatrix} 1 & 2 \\ 3 & 0 \end{pmatrix} + 0 - 3 \cdot \det \begin{pmatrix} 0 & 1 \\ 2 & 3 \end{pmatrix} = -6 - 3(-2) = -6 + 6 = 0 \quad \chi$$

9.  $\det \begin{pmatrix} 0 & 1+i & 2 \\ -2i & 0 & 1-i \\ 3 & 4i & 0 \end{pmatrix} = 0 + (1+i) \cdot \det \begin{pmatrix} -2i & 1-i \\ 3 & 0 \end{pmatrix} + 2 \cdot \det \begin{pmatrix} -2i & 1-i \\ 3 & 4i \end{pmatrix} = (1+i)(1-i) + 2(-4i) = 1 - i^2 - 8i = 2 - 8i$

$\det \begin{pmatrix} 0 & 1+i & 2 \\ -2i & 0 & 1-i \\ 3 & 4i & 0 \end{pmatrix} = (-1)^{3+1} 3 \cdot \det \begin{pmatrix} 1+i & 2 \\ 0 & 1-i \end{pmatrix} + 4i \det \begin{pmatrix} 0 & 2 \\ -2i & 1-i \end{pmatrix} + 0 = 3(1+i)(1-i) + 4i(-4i) = 6 - 16 = -10$

11.  $\det \begin{pmatrix} 0 & 2 & 1 & 3 \\ 1 & 0 & -2 & 2 \\ 3 & -1 & 0 & 1 \\ -1 & 1 & 2 & 0 \end{pmatrix} = (-1)^{4+1}(-1) \det \begin{pmatrix} 2 & 1 & 3 \\ 0 & -2 & 2 \\ -1 & 0 & 1 \end{pmatrix} + (-1)^{4+2}(1) \det \begin{pmatrix} 0 & 1 & 3 \\ 3 & -1 & 1 \\ -1 & 2 & 0 \end{pmatrix} + (-1)^{4+3}(2) \det \begin{pmatrix} 0 & 2 & 3 \\ 3 & -1 & 1 \\ -1 & 1 & 0 \end{pmatrix} + 0$   
 $= \left[ 0 - 2(-1)^{2+2} \det \begin{pmatrix} 2 & 3 \\ -1 & 1 \end{pmatrix} + 2(-1)^{2+3} \det \begin{pmatrix} 2 & 1 \\ -1 & 0 \end{pmatrix} \right] + \left[ (-1)^{1+2} \det \begin{pmatrix} 1 & 3 \\ 3 & 1 \end{pmatrix} + 3(-1)^{1+3} \det \begin{pmatrix} 1 & -2 \\ 3 & 0 \end{pmatrix} \right]$   
 $= -2(2+3) - 2(1) - (1-6) + 3(6) + 2[-2(1-6) + 3(-1)]$   
 $= -10 - 2 + 5 + 18 + 2(10 - 3) = -10 - 2 + 5 + 18 + 14 = 25$

13. We transform the matrix into an upper triangular one via a type I row operation:  
 $\begin{pmatrix} 4 & 5 & 6 \\ 0 & 2 & 3 \\ 0 & 0 & 1 \end{pmatrix} \xrightarrow{T1} \begin{pmatrix} 4 & 5 & 6 \\ 0 & 2 & 3 \\ 0 & 0 & 1 \end{pmatrix}$ . Hence, the required determinant is just  $4 \cdot 2 \cdot 1 = 8$ .

15. As before, we transform the matrix with elementary row operations so it becomes upper triangular:  
 $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{2 \times T3} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & -6 & -12 \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & 0 & 0 \end{pmatrix}$ . Thus, the determinant required is  $1 \cdot (-3) \cdot 0 = 0$ .

17. Again transforming the matrix to an upper triangular one,  
 $\begin{pmatrix} 0 & 1 & 1 \\ 1 & 2 & -5 \\ 6 & -4 & 3 \end{pmatrix} \xrightarrow{T1} \begin{pmatrix} 1 & 2 & -5 \\ 0 & 1 & 1 \\ 6 & -4 & 3 \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} 1 & 2 & -5 \\ 0 & 1 & 1 \\ 0 & -16 & 33 \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} 1 & 2 & -5 \\ 0 & 1 & 1 \\ 0 & 0 & -16 \end{pmatrix}$   
 So the determinant we need has value  $1 \cdot 1 \cdot (-16) = -16$ .  
 $33 + 16 = 49$   
 $33 + 1 \cdot 1 \cdot (-16) = 16$   
 $-1 \cdot 1 \cdot (-49) = -49$

19. Transforming the matrix so it is upper triangular,

$$\begin{pmatrix} i & 2 & -1 \\ 3 & 1+i & 2 \\ -2i & 1 & 4-i \end{pmatrix} \xrightarrow{2 \times T1} \begin{pmatrix} i & 2 & -1 \\ -2i & 1 & 4-i \\ 3 & 1+i & 2 \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} i & 2 & -1 \\ -2i & 1 & 4-i \\ 0 & 5 & 2-i \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} i & 2 & -1 \\ 0 & 5 & 2-i \\ 0 & 5 & 2-i \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} i & 2 & -1 \\ 0 & 5 & 2-i \\ 0 & 0 & -\frac{2i}{5} - \frac{1}{5}i \end{pmatrix}$$

As such, the determinant  $\det \begin{pmatrix} i & 2 & -1 \\ -2i & 1 & 4-i \\ 3 & 1+i & 2 \end{pmatrix}$  is just  $3(5)(-\frac{2i}{5} - \frac{1}{5}i) = -28 - 2i$ .

21. Using elementary row operations to transform the given matrix into an upper triangular matrix,

$$\begin{pmatrix} 1 & 0 & -2 & 3 \\ -3 & 1 & 1 & 2 \\ 0 & 4 & -1 & 1 \\ 2 & 3 & 0 & 1 \end{pmatrix} \xrightarrow{3 \times T3} \begin{pmatrix} 1 & 0 & -2 & 3 \\ 0 & 1 & -5 & 11 \\ 0 & 4 & -1 & 1 \\ 0 & 3 & 4 & -5 \end{pmatrix} \xrightarrow{2 \times T3} \begin{pmatrix} 1 & 0 & -2 & 3 \\ 0 & 1 & -5 & 11 \\ 0 & 0 & 19 & -43 \\ 0 & 0 & 19 & -38 \end{pmatrix} \xrightarrow{T3} \begin{pmatrix} 1 & 0 & -2 & 3 \\ 0 & 1 & -5 & 11 \\ 0 & 0 & 19 & -43 \\ 0 & 0 & 0 & 5 \end{pmatrix}$$

As a result, the necessary determinant is  $1 \cdot 1 \cdot 19 \cdot 5 = 95$ .

23. This is immediate for  $1 \times 1$  matrices as  $\det(a) = a$  by definition (for any  $a \in \mathbb{F}$ ). Assume this is true for  $n \times n$  matrices and let  $A$  be a  $(n+1) \times (n+1)$  upper triangular matrix. We shall do cofactor expansion along the  $(n+1)$ th row of  $A$ . But first, notice that  $A_{n+1,j} = \begin{cases} 1 & \text{if } j=n+1 \\ 0 & \text{otherwise} \end{cases}$  and  $\tilde{A}_{n+1,j}$  is an  $n \times n$  upper triangular matrix since its  $(i_1, i_2)$ th entry is just  $A_{i_1, i_2}$  for  $1 \leq i_1, i_2 \leq n$ . So, our assumption / induction hypothesis says  $\det(\tilde{A}_{n+1,j}) = \prod_{i=1}^n (\tilde{A}_{n+1,j})_{i,i} = \prod_{i=1}^n A_{i,i}$ . Hence,  
 $\det(A) = (-1)^{n+1+n+1} A_{n+1,n+1} \cdot \prod_{i=1}^n A_{i,i} = \prod_{i=1}^{n+1} A_{i,i}$ . In other words, for  $(n+1) \times (n+1)$  matrices, the statement holds. So by induction, it must be true of any square matrix.

24. See the corresponding self-proof.

25. Let  $a_j$  represent the  $j$ th row of  $A$ , then

$$\det(kA) = \det \begin{pmatrix} ka_1 \\ ka_2 \\ \vdots \\ ka_n \end{pmatrix} = k \det \begin{pmatrix} a_1 \\ ka_2 \\ \vdots \\ ka_n \end{pmatrix} = k^n \det \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} = k^n \det(A)$$

by using Theorem 4.3  $n$  times.

26. By exercise 25,  $\det(-A) = (-1)^n \det(A)$ . So, the desired equality  $\det(-A) = \det(A)$  holds iff  $n$  is even.



27. Suppose  $A \in \text{Mat}_n(\mathbb{F})$  has two identical columns. ~~We notice that elementary row operations keep identical columns identical.~~  
 Hence,  $\det(A) = 0$  by the analogy to Theorem 4.6.

28. By example 4 of this section, we know  $\det(I) = 1$ . So, by using the information on how elementary row operations affect determinants, as provided on page 217, we see that  $\det(E_1) = -1$ ,  $\det(E_2) = c$  where  $E_2$  is obtained from  $I$  by multiplying one row by the scalar  $c \in \mathbb{F}$ , and  $\det(E_3) = 1$ .

29. Ideas

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$T3: E_{ij} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}^t = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ c & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \rightarrow I_+$$

$$E_{ij} = \begin{cases} c & \text{if } i=r \text{ \& } j=s, \\ I_{ij} & \text{otherwise.} \end{cases}$$



T2: Equality holds by symmetry.

$$T1: E_{ij} = \begin{cases} 1 & \text{if } i=j \text{ is not } r \text{ or } s, \\ 1 & \text{if } i=r \text{ \& } j=s \text{ or } i=s \text{ \& } j=r, \\ 0 & \text{otherwise.} \end{cases} = \begin{cases} 1 & \text{if } j=i \text{ is not } r \text{ or } s, \\ 1 & \text{if } j=r \text{ \& } i=s \text{ or } j=s \text{ \& } i=r, \\ 0 & \text{otherwise.} \end{cases} = E_{ji}$$

Proof

We first see that for  $n \times n$  elementary matrices  $E$  of types 1 and 2, they are symmetric. If  $E$  is of type 1, that is, obtained by interchanging rows  $r$  and  $s$  of  $I$  for some  $1 \leq r, s \leq n$ , then  $E_{ij} = \begin{cases} 1 & \text{if } i=r \text{ \& } j=s \text{ or } i=s \text{ \& } j=r, \\ I_{ij} & \text{otherwise.} \end{cases} = \begin{cases} 1 & \text{if } j=r \text{ \& } i=s \text{ or } j=s \text{ \& } i=r, \\ I_{ji} & \text{otherwise.} \end{cases} = E_{ji}$ . Similarly, when  $E$  is of type 2, then it is also symmetric.

obtained by multiplying some row  $r$  of  $I$  by a scalar  $c \in \mathbb{F}$ , so  $E$  is only nonzero on its diagonal entries; it is symmetric. Hence,  $E = E^t$  in both cases. As we saw,  $\det(E) = \det(E^t)$  holds. Now consider  $E$  being of type 3, thus it is obtained by adding  $c \in \mathbb{F}$  times of row  $r$  to row  $s$  of  $I$ . Then,  $E^t$  is obtained by adding  $c \in \mathbb{F}$  times of row  $s$  to row  $r$  because  $(E^t)_{ij} = E_{ji} = \begin{cases} c & \text{if } j=s \text{ \& } i=r, \\ I_{ji} & \text{otherwise.} \end{cases} = \begin{cases} c & \text{if } j=r \text{ \& } i=s, \\ I_{ij} & \text{otherwise.} \end{cases}$  In other words,  $E^t$  is also a type 3 elementary matrix.

By exercise 28 of this section,  $\det(E) = \det(E^t) = 1$ .

or, for T3, we could use a cofactor expansion along

30. If  $n = 2m + 1$  for some  $m \in \mathbb{N}$ , then we see that:

$$B = \begin{pmatrix} a_n \\ a_{n-1} \\ \vdots \\ a_{m+1} \\ a_m \\ \vdots \\ a_2 \\ a_1 \end{pmatrix} \xrightarrow{\text{1st}} \begin{pmatrix} a_1 \\ a_{n-1} \\ \vdots \\ a_{m+1} \\ a_m \\ \vdots \\ a_2 \\ a_n \end{pmatrix} \xrightarrow{\text{2nd}} \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_{m+1} \\ a_m \\ \vdots \\ a_2 \\ a_{n-1} \\ a_n \end{pmatrix} \xrightarrow{\text{3rd}} \dots \xrightarrow{\text{mth}} \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_{n-1} \\ a_n \end{pmatrix} = A$$

Even if  $n = 2m$  for some  $m \in \mathbb{N}$  instead, simply remove the ' $a_m$ ' entry from above and the same argument still applies.  
 In any case,  $m$  type 1 row operations suffice to transform  $B$  to  $A$ . So,  $\det(B) = (-1)^m \det(A) = \begin{cases} (-1)^{\frac{n}{2}} \det(A) & \text{if } n \text{ odd,} \\ (-1)^{\frac{n}{2}} \det(A) & \text{if } n \text{ even.} \end{cases}$

# Self-Proof of Theorem 4.9 (Cramer's Rule)

Idea:

If  $\det(A) \neq 0$ ,  $\text{rank}(A) = n$  and since  $\text{rank}(A) \leq \text{rank}(A|b) \leq n$ ,  $\text{rank}(A|b) = n$  also. So,  $\det(A|b) = \det(A^{-1}A|A^{-1}b) = \det(I_n|b')$  for some column vector  $b' \in \mathbb{F}^n$ .

Indeed, system has exactly one solution, namely  $b' = (b'_1, b'_2, \dots, b'_n)$ .

$$b'_k = \frac{\det(N_k)}{\det(I_n)} \text{ because } \det(N_k) = \det \begin{pmatrix} 1 & 0 & 0 & \dots & 0 & b'_1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 1 & 0 & \dots & 0 & b'_2 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 1 & \dots & 0 & b'_3 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & \dots & \cdot & \cdot & \cdot & \cdot & \dots & 0 & 0 \\ 0 & 0 & 0 & \dots & \cdot & \cdot & \cdot & \cdot & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & b'_n & 0 & 0 & \dots & 0 & 1 & 0 \end{pmatrix} = \det \begin{pmatrix} 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ b'_{1k} & \dots & b'_{2k} & \dots & b'_{nk} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \dots & 0 & \dots & 0 \end{pmatrix} = b'_{1k} \det(I_n)$$

$\det(e_1, e_2, \dots, b', \dots, e_n) = \det(e_1, e_2, \dots, b'_{1k}e_k, \dots, e_n)$  where  $\{e_i | 1 \leq i \leq n\}$  is the standard ordered basis for  $\mathbb{F}^n$  in column form.

$$= \frac{\det(N_k)}{\det(A)\det(A^{-1})}$$

$$\stackrel{?}{=} \frac{\det(A^{-1}M_k)}{\det(A)\det(A^{-1})}$$

$$N_k \stackrel{?}{=} A^{-1}M_k$$

$$M_k = (a_1, a_2, \dots, b, \dots, a_n)$$

$$A^{-1}M_k = (A^{-1}a_1, A^{-1}a_2, \dots, A^{-1}b, \dots, A^{-1}a_n) = (e_1, e_2, \dots, b', \dots, e_n)$$

(Exercise 15 of section 3-2)  
 $\rightarrow M(A|B) = (MA|MB)$

## Proof

Given  $\det(A) \neq 0$ ,  $\text{rank}(A) = n$  by the corollary to Theorem 4.6, and since  $\text{rank}(A) \leq \text{rank}(A|b) \leq n$ ,  $\text{rank}(A|b) = n$  too. Accordingly, the solution space of the homogeneous system corresponding to  $Ax = b$  has dimension 0. Hence, the nonhomogeneous system  $Ax = b$  has exactly one solution, namely  $b' := (b'_1, b'_2, \dots, b'_n) := A^{-1}b$  because  $(A^{-1}A|A^{-1}b) = (I_n|b')$  is the reduced row echelon form of  $(A|b)$ . Let  $N_k$  be the  $n \times n$  matrix obtained from  $I_n$  by replacing column  $k$  of  $I_n$  by  $b'$ . First, notice that  $\det(N_k) = \det(e_1, e_2, \dots, e_{k-1}, b', e_{k+1}, \dots, e_{n-1}, e_n) \stackrel{\text{Thm 4.3}}{=} \det(e_1, e_2, \dots, e_{k-1}, b'_{1k}e_k, e_{k+1}, \dots, e_{n-1}, e_n) \stackrel{\text{Thm 4.3}}{=} b'_{1k} \det(I_n) = b'_{1k}$  from Theorems 4.6 and 4.3, where  $\{e_i | 1 \leq i \leq n\}$  is the standard ordered basis of  $\mathbb{F}^n$  in column form. Secondly, by having  $a_j$  be the  $j$ th column of  $A$ , we see that  $A^{-1}M_k = A^{-1}(a_1, a_2, \dots, a_{k-1}, b, a_{k+1}, \dots, a_{n-1}, a_n) \stackrel{\text{Ex 15}}{=} (A^{-1}a_1, A^{-1}a_2, \dots, A^{-1}a_{k-1}, A^{-1}b, A^{-1}a_{k+1}, \dots, A^{-1}a_{n-1}, A^{-1}a_n) = (e_1, e_2, \dots, e_{k-1}, b', e_{k+1}, \dots, e_{n-1}, e_n) = N_k$  from exercise 15 of section 3-2.

As such,  $b'_k = \frac{\det(N_k)}{\det(I_n)} = \frac{\det(A^{-1}M_k)}{\det(A^{-1}A)} \stackrel{\text{Thm 4.7}}{=} \frac{\det(A^{-1}) \cdot \det(M_k)}{\det(A^{-1}) \cdot \det(A)} = \frac{\det(M_k)}{\det(A)}$ .

# Self-Proof of Corollary

Idea

If  $A$  invertible,  $\mathbb{1} = \det(I_n) = \det(AA^{-1}) = \det(A) \det(A^{-1})$  by Theorem 4.7 so  $\det(A^{-1}) = \frac{1}{\det(A)}$ .

When  $\det(A) \neq 0$ ,  $\text{rank}(A) = n$  by the Corollary of Theorem 4.6.

Proof

When  $A$  is invertible, by Theorem 4.7 we see that  $\det(A) \det(A^{-1}) = \det(AA^{-1}) = \det(I_n) = \mathbb{1}$ . So,  $\det(A) \neq 0$  and  $\det(A^{-1}) = \frac{1}{\det(A)}$  as expected.  $\square$

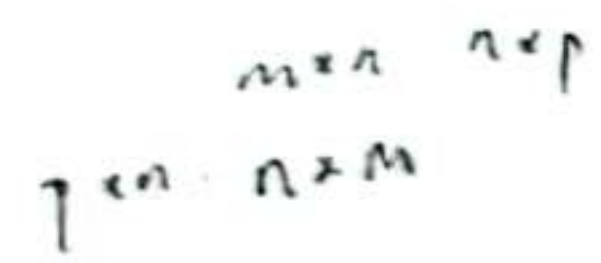
Conversely, given  $\det(A) \neq 0$ , the Corollary to Theorem 4.6 says  $\text{rank}(A) = n$ . Consequently,  $A$  is invertible. Therefore, the biconditional holds.  $\square$

# Self-Proof of Theorem 4.8

Idea

$A$  not invertible  $\Rightarrow A^t$  not invertible  $\Rightarrow$  trivial

$A$  invertible  $\Rightarrow$



Proof

If  $A$  is not invertible, then  $A^t$  — being of equal rank — must not be invertible. Thus,  $\det(A^t) = \det(A) = 0$ . Now consider  $A$  being invertible.

Then for some nonnegative integer  $n$  and elementary matrices  $E_i$ ,  $A = E_n E_{n-1} \dots E_1$  so  $\det(A^t) = \det(E_1^t E_2^t \dots E_n^t) \stackrel{\text{Thm 4.7}}{=} \det(E_1) \cdot \det(E_2) \cdot \dots \cdot \det(E_n)$

$\stackrel{\text{Ex 2.8}}{=} \det(E_1) \cdot \det(E_2) \cdot \dots \cdot \det(E_n) = \det(E_n) \cdot \det(E_{n-1}) \cdot \dots \cdot \det(E_1) \stackrel{\text{Thm 4.7}}{=} \det(E_n E_{n-1} \dots E_1) = \det(A)$  by Theorem 4.7 and exercise 29 of section 4.2.  $\square$

4.2.

1. (a) False  $\det \begin{pmatrix} 2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} = 2$

(b) True ✓

(c) False, it should be  $\det(M) \neq 0$  instead

(d) True ✓

(e) False,  $\det(A^T) = \det(A)$

(f) True ✓

(g) False ✓

(h) False ✓ Nice! 😊

3. Since  $\det \begin{pmatrix} 2 & 1 & -3 \\ 1 & -2 & 1 \\ 3 & 4 & -2 \end{pmatrix} = -\det \begin{pmatrix} 1 & -2 & 1 \\ 0 & 5 & -5 \\ 0 & 10 & -5 \end{pmatrix} = -\det \begin{pmatrix} 1 & -2 & 1 \\ 0 & 5 & -5 \\ 0 & 0 & 5 \end{pmatrix} = -1 \cdot 5 \cdot 5 = -25$ , Cramer's Rule indeed applies. Using the notation

of Theorem 4.9, we have  $x_1 = \frac{\det \begin{pmatrix} 5 & 1 & -3 \\ 0 & 4 & -2 \end{pmatrix}}{-25} = \frac{\det \begin{pmatrix} 5 & 1 & -3 \\ 0 & 4 & -2 \end{pmatrix}}{-25} = \frac{\det \begin{pmatrix} 5 & 1 & -3 \\ 0 & 4 & -2 \end{pmatrix}}{-25} = \frac{(5)(-4)(5)}{-25} = 4$ ,

$x_2 = \frac{\det \begin{pmatrix} 2 & 5 & -3 \\ 3 & 0 & -2 \end{pmatrix}}{-25} = \frac{-\det \begin{pmatrix} 1 & 10 & 1 \\ 0 & -15 & -5 \end{pmatrix}}{-25} = \frac{\det \begin{pmatrix} 1 & 10 & 1 \\ 0 & 15 & 5 \end{pmatrix}}{25} = \frac{1 \cdot 15 \cdot 5}{25} = -3$ , and  $x_3 = \frac{\det \begin{pmatrix} 2 & 1 & 5 \\ 1 & 4 & 0 \end{pmatrix}}{-25} = \frac{-\det \begin{pmatrix} 1 & 2 & 10 \\ 0 & 5 & -30 \end{pmatrix}}{-25} = \frac{\det \begin{pmatrix} 1 & 2 & 10 \\ 0 & 5 & -30 \end{pmatrix}}{25} = 0$

Therefore, the unique solution to this system of linear equations is  $(x_1, x_2, x_3) = (4, -3, 0)$ .

To verify:  
 $2(4) + (-3) - 3(0) = 8 - 3 = 5$  ✓  
 $4 - 2(-3) + 0 = 4 + 6 = 10$  ✓  
 $3(4) + 4(-3) - 2(0) = 12 - 12 = 0$  ✓

5. Since  $\det \begin{pmatrix} -8 & -1 & 4 \\ 2 & -1 & 1 \\ 0 & 1 & -7 \end{pmatrix} = \det \begin{pmatrix} 1 & -1 & 4 \\ 0 & -5 & 33 \\ 0 & 1 & -7 \end{pmatrix} = -\det \begin{pmatrix} 1 & -1 & 4 \\ 0 & 1 & -7 \\ 0 & 0 & -2 \end{pmatrix} = -1(1)(-2) = 2$ , using the notation of Theorem 4.9; Cramer's Rule

$x_1 = \frac{\det \begin{pmatrix} -4 & -1 & 4 \\ 8 & 3 & 1 \\ 0 & -1 & 1 \end{pmatrix}}{2} = \frac{\det \begin{pmatrix} -4 & -1 & 4 \\ 0 & 1 & 9 \\ 0 & 0 & 10 \end{pmatrix}}{2} = \frac{-4(1)(10)}{2} = -20$ ,  $x_2 = \frac{\det \begin{pmatrix} -8 & 4 & 4 \\ 2 & 0 & 1 \\ 0 & 8 & -7 \end{pmatrix}}{2} = \frac{\det \begin{pmatrix} 1 & -4 & 4 \\ 0 & -24 & 33 \\ 0 & 8 & -7 \end{pmatrix}}{2} = \frac{-\det \begin{pmatrix} 1 & -4 & 4 \\ 0 & 8 & -7 \\ 0 & 0 & 12 \end{pmatrix}}{2} = \frac{-1(8)(12)}{2} = -48$ , and

$x_3 = \frac{\det \begin{pmatrix} -8 & -1 & -4 \\ 2 & -1 & 0 \\ 0 & 1 & 1 \end{pmatrix}}{2} = \frac{\det \begin{pmatrix} 1 & -1 & -4 \\ 0 & -5 & -24 \\ 0 & 0 & 16 \end{pmatrix}}{2} = \frac{-1(1)(16)}{2} = -8$ . Thus, the unique solution to this system is  $(x_1, x_2, x_3) = (-20, -48, -8)$ .

Checking:  
 $-20 - (-48) + 4(-8) = -20 + 48 - 32 = -4$  ✓  
 $-8(-20) + 3(-48) + (-8) = 160 - 144 - 8 = 8$  ✓  
 $2(-20) - (-48) + (-8) = -40 + 48 - 8 = 0$  ✓

7. As  $\det \begin{pmatrix} -\frac{3}{2} & -\frac{1}{2} & 0 \\ 1 & 2 & 1 \\ 0 & 0 & 0 \end{pmatrix} = -\det \begin{pmatrix} 1 & 2 & 1 \\ 0 & 3 & 2 \\ 0 & -5 & -2 \end{pmatrix} = -\det \begin{pmatrix} 1 & 2 & 1 \\ 0 & 3 & 2 \\ 0 & 0 & \frac{1}{3} \end{pmatrix} = -1(3)(\frac{1}{3}) = -4$ , (Cramer's Rule can be used. Hence, using Theorem 4.9)

$x_1 = \frac{\det \begin{pmatrix} 4 & 1 & 0 \\ 12 & -1 & 0 \\ -8 & 2 & 1 \end{pmatrix}}{-4} = \frac{\det \begin{pmatrix} 4 & 1 & 1 \\ 0 & -4 & -3 \\ 0 & 4 & 3 \end{pmatrix}}{-4} = \frac{\det \begin{pmatrix} 4 & 1 & 1 \\ 0 & -4 & -3 \\ 0 & 0 & 0 \end{pmatrix}}{-4} = 0$ ,  $x_2 = \frac{\det \begin{pmatrix} -\frac{3}{2} & 4 & 0 \\ 1 & 12 & 0 \\ 0 & -8 & 1 \end{pmatrix}}{-4} = \frac{-\det \begin{pmatrix} 1 & -8 & 1 \\ 0 & -4 & 2 \\ 0 & 0 & 12 \end{pmatrix}}{-4} = \frac{-(-4)(12)}{-4} = -12$ ,

and lastly,  $x_3 = \frac{\det \begin{pmatrix} -\frac{1}{2} & -\frac{1}{2} & 4 \\ 1 & 2 & 1 \\ 0 & 0 & 0 \end{pmatrix}}{-4} = \frac{-\det \begin{pmatrix} 1 & 2 & -8 \\ 0 & 3 & 2 \\ 0 & -5 & -2 \end{pmatrix}}{-4} = \frac{-\det \begin{pmatrix} 1 & 2 & -8 \\ 0 & 3 & 2 \\ 0 & 0 & \frac{6}{3} \end{pmatrix}}{-4} = \frac{-1(3)(\frac{6}{3})}{-4} = 16$ . The unique solution to the system is hence  $(x_1, x_2, x_3) = (0, -12, 16)$ .

To check:

- $3(0) - 12 + 16 = 4$  ✓
- $1 - 2(0) - (-12) + 0 = 12$  ✓
- $1(0) + 2(-12) + 16 = -8$  ✓

8. Let  $u, v, a_i$  be column vectors in  $\mathbb{F}^n$  for  $1 \leq i \leq n$  and  $k$  be a scalar in  $\mathbb{F}$ . Then by Theorems 4.3 and 4.8, given any  $1 \leq r \leq n$ ,

$$\det(a_1, \dots, a_{r-1}, u+kv, a_{r+1}, \dots, a_n) \stackrel{\text{Thm 4.3}}{=} \det(a_1, \dots, a_{r-1}, u, a_{r+1}, \dots, a_n) + k \det(a_1, \dots, a_{r-1}, v, a_{r+1}, \dots, a_n)$$

So,  $n$ -linearity holds column-wise. □

An upper triangular matrix is invertible iff its determinant is nonzero, by the corollary to Theorem 4.7, iff the product of its diagonal entries is nonzero, by Exercise 23 of Section 4.2, iff all of its diagonal entries are nonzero, since fields have no zero divisors.

When  $\det(M) \neq 0$ , the corollary to Theorem 4.7 says  $M$  is invertible. So,  $M$  can be expressed as a product of some  $n$  elementary matrices  $E_i$  ( $1 \leq i \leq n$ ). That is,  $M = E_n E_{n-1} \dots E_1$ . Since multiplication with elementary matrices is rank preserving,  $\text{rank}(M^k) = \text{rank}(E_n^k E_{n-1}^k \dots E_1^k) = \text{rank}(E_n E_{n-1} \dots E_1) = \text{rank}(M) = n$  for any positive integer  $k$ . Since  $\text{rank}(0) = 0$ ,  $M^k \neq 0$  for each positive integer  $k$ . □

Given that  $M$  is skew symmetric and  $n$  is odd, Theorem 4.8 together with Exercise 25 of Section 4.2 says  $\det(M) \stackrel{\text{Thm 4.8}}{=} \det(M^t) = \det(-M) \stackrel{\text{Ex 25}}{=} (-1)^n \det(M) = -\det(M)$ . So,  $2 \det(M) = 0$ . Thus,  $\det(M) = 0$  means  $M$  is not invertible by the corollary to Theorem 4.7. When  $n$  is even, let  $M$  be the matrix with its  $j$ th row,  $m_j := \begin{cases} e_{j+1} & \text{if } j \text{ odd,} \\ -e_{j-1} & \text{if } j \text{ even.} \end{cases}$  so  $\text{rank}(M) = n$  is certain as  $j+1 \neq j-1$  for any integer  $j$ . Moreover,  $M_{ij} = \begin{cases} 1 & \text{if } j \text{ odd \& } i=j+1, \\ -1 & \text{if } j \text{ even \& } i=j-1, \\ 0 & \text{otherwise.} \end{cases}$   $\begin{cases} -1 & \text{if } i \text{ even \& } j=i-1, \\ 1 & \text{if } i \text{ odd \& } j=i+1, \\ 0 & \text{otherwise.} \end{cases} = -M_{ji}$  shows that  $M$  is skew-symmetric. Therefore, for any even  $n$ , a  $n \times n$  skew symmetric matrix can possibly have nonzero determinant, for example the skew-symmetric matrix above has  $\det(M) \neq 0$  as  $\text{rank}(M) = n$ .

determinant, for example the skew-symmetric matrix above has  $\det(M) \neq 0$  as  $\text{rank}(M) = n$ .

11. Idea

$$M_{ij} := \begin{cases} 0 & \text{if } i=j \\ -1 & \text{if } i < j \\ 1 & \text{if } i > j \end{cases}$$

$M_j$  being row  $j$  of  $M$

$$M_j = [-1 \ -1 \ -1 \ \dots \ -1 \ 0 \ 1 \ 1 \ \dots \ 1]$$

nth

$$M_{ij} := \begin{cases} 1 & \text{if } j=i+1 \\ -1 & \text{if } i=j+1 \\ 0 & \text{otherwise} \end{cases}$$

let  $M_j$  be row  $j$ ,  
 $M_1 = e_2$   
 $M_n = -e_{n-1}$

$$M_3 := M_1 + M_3 \quad M_{n-3} := M_n + M_{n-3}$$

$$\begin{pmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & -1 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 \end{pmatrix}$$

$$\left( \begin{array}{ccc|ccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{array} \right)$$

$$\left( \begin{array}{ccc|ccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 & -1 & 0 \end{array} \right)$$

$$\left( \begin{array}{ccc|ccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & 0 & 0 & -1 & 0 \end{array} \right)$$

$$\left( \begin{array}{ccc|ccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & -1 & 0 \end{array} \right)$$

$$\begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 0 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

Assume  $Q$  is orthogonal.

12. From Theorems 4.7 and 4.8,  $\det(Q)^2 \stackrel{\text{Thm 4.7}}{=} \det(Q) \det(Q^t) \stackrel{\text{Thm 4.8}}{=} \det(QQ^t) = \det(I) = 1$ . Hence,  $\det(Q) = \pm 1$ . □

13. (a) When  $n=1$ ,  $\det(\overline{a+bi}) = \det(a-bi) = a-bi = \overline{a+bi} = \overline{\det(a+bi)}$  for any  $a, b \in \mathbb{R}$ . Hence,  $\det(\overline{M}) = \overline{\det(M)}$  for  $M \in M_{1 \times 1}(\mathbb{C})$ . Suppose this is true of all matrices  $M \in M_{n \times n}(\mathbb{C})$ . Then, letting  $M \in M_{(n+1) \times (n+1)}(\mathbb{C})$  instead,  $\det(\overline{M}) = \sum_{j=1}^{n+1} (-1)^{2+j} \overline{M_{2j}} \det(\overline{M}_{2j}) = \sum_{j=1}^{n+1} (-1)^{2+j} \overline{M_{2j}} \det(\overline{M}_{2j}) \stackrel{\text{I.H.}}{=} \sum_{j=1}^{n+1} (-1)^{2+j} \overline{M_{2j}} \overline{\det(M_{2j})} = \overline{\sum_{j=1}^{n+1} (-1)^{2+j} M_{2j} \det(M_{2j})} = \overline{\det(M)}$ . So, the claim is still true for every  $(n+1) \times (n+1)$  matrix. By induction, the claim holds for all  $n \times n$  matrices and any natural number  $n$ . □

(b) Assume  $Q$  is a unitary matrix. Then by Theorems 4.7 and 4.8, as well as exercises 12 and 13(a) above, we have that  $1 = \det(I) = \det(QQ^*) \stackrel{\text{Thm 4.7}}{=} \det(Q) \det(Q^*) \stackrel{\text{I.H.}}{=} \det(Q) \overline{\det(Q)} = |\det(Q)|^2$ . Hence,  $|\det(Q)| = 1$ . □

14. Clearly,  $\det(B) \neq 0$  iff  $\text{rank}(B) = n$ , by the corollary to Theorem 4.6, iff  $\dim(\text{span}(\beta)) = n$  iff  $\beta$  is a basis for  $\mathbb{F}^n$ , since  $|\beta| = n$ . □

15. Recall that matrices  $A, B \in M_{n \times n}(\mathbb{F})$  are similar iff there exists another <sup>(invertible)</sup> matrix  $Q \in M_{n \times n}(\mathbb{F})$  so  $B = Q^{-1}AQ$ . Consequently, using the definitions above and supposing  $A, B \in M_{n \times n}(\mathbb{F})$  are indeed similar, we have that  $\det(B) \stackrel{\text{Thm 4.7}}{=} \det(Q^{-1}) \det(A) \det(Q) = \det(A) \det(Q^{-1}) \det(Q) \stackrel{\text{Thm 4.7}}{=} \det(A) \det(Q^{-1}Q) = \det(A) \det(I) = \det(A)$  by Theorem 4.7. □

16. By Theorem 4.7 we have  $\det(A) \cdot \det(B) \stackrel{\text{Thm 4.7}}{=} \det(AB) = \det(I) = 1$ . Hence,  $\det(A)$  and  $\det(B)$  must be nonzero, implying that  $A$  and  $B$  are invertible. That is,  $B = A^{-1}(AB) = A^{-1}I = A^{-1}$ . □

17. Suppose  $n$  is odd and  $\mathbb{F}$  is not a field of characteristic two. Then by Theorems 4.3 and 4.7,  $\det(A) \cdot \det(B) = \det(AB) = \det(-BA) \stackrel{\text{Thm 4.7}}{=} \det(-B) \cdot \det(A) \stackrel{\text{Thm 4.3}}{=} (-1)^n \cdot \det(B) \cdot \det(A) = -\det(A) \cdot \det(B)$ . So,  $\det(A) \cdot \det(B) = 0$ . Since fields have no zero divisors, either  $\det(A)$  or  $\det(B)$  must be 0. This corresponds to  $A$  or  $B$  being uninvertible by the corollary to Theorem 4.7. □

19. Since Theorem 4.8 informs us that  $\det(A) = \det(A^t)$ ,  $\det(A)$  is the product of the diagonal entries of itself; as  $(A^t)_{ii} = A_{ii}$ . Let  $A \in M_{n \times n}(\mathbb{F})$ , for some  $n \in \mathbb{N}$ . □

~~20. There are some elementary matrices  $E_i$  of types 1 and 3, of which  $k$  are type 1, so we have  $E_1 \dots E_k A$  being upper triangular. Hence, letting  $G := E_1 \dots E_k$ , we see that high pivoting can be so arranged to achieve~~



20. There exists some  $m_1$  and  $m_2$  elementary row operations of types 1 and 3 respectively, so  $A$  is transformed into an upper triangular matrix  $A'$ . By applying the same sequence of elementary row operations on  $M$ , it becomes some  $\begin{pmatrix} A' & B' \\ 0 & I \end{pmatrix}$ . Thus,  $\det(M) = (-1)^{m_1+m_2} \det \begin{pmatrix} A' & B' \\ 0 & I \end{pmatrix} = (-1)^{m_1+m_2} \prod_{i=1}^n A_{ii} = \det(A)$ . □

(finally a satisfactory phrasing!) ☺

21. Almost the same as 20 except that we need to apply 2 sequence of eros  $A \xrightarrow{m_1} A'$  &  $C \xrightarrow{m_2} C'$  to  $M$ . □

22. (a) Notice  $T(x^j) = [c_0^j, c_1^j, \dots, c_n^j] = \sum_{i=1}^{n+1} c_i^j e_i$ . So,  $M_{ij} = c_i^j$  as we thought.

(b) Since  $T$  is an isomorphism according to exercise 22 of section 2.4,  $LM$  is also an isomorphism (as Figure 2.2 confirms). As such,  $\text{rank}(M) = n+1$  so the Corollary to Theorem 4.7 says  $\det(M) \neq 0$ .

(c) Ideas

$$\exists P: P \begin{pmatrix} c_0 \\ \vdots \\ c_n \end{pmatrix} = \begin{pmatrix} c_0^k \\ \vdots \\ c_n^k \end{pmatrix} ?$$

$$\begin{pmatrix} 1 & c_0 & c_0^2 & \dots & c_0^n \\ 0 & c_1 - c_0 & c_1^2 - c_0^2 & \dots & c_1^n - c_0^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & c_n - c_0 & c_n^2 - c_0^2 & \dots & c_n^n - c_0^n \end{pmatrix} \rightarrow \begin{pmatrix} 1 & c_0 & c_0^2 & \dots & c_0^n \\ 0 & c_1 - c_0 & (c_1 - c_0)(c_1 + c_0) & \dots & (c_1 - c_0)(c_1^{n-1} + \dots + c_0^{n-1}) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & c_n - c_0 & (c_n - c_0)(c_n + c_0) & \dots & (c_n - c_0)(c_n^{n-1} + \dots + c_0^{n-1}) \end{pmatrix}$$

$n=1$ :  $\det(M) = 1$ , vacuously true  
Assume true for  $n=k$ . For  $M \in M_{n+1}(IF)$ ,

$[i, k+1]$ th entry at the  $(k+1)$ th iteration:  $v_{k+1}^i = v_k^i +$

OM  
1st

$$c_i^j - c_0^j$$

$$c_i^j - c_0^j + \frac{c_0 - c_i}{c_0} \cdot c_0^j = c_i^j - c_i c_0^{j-1}$$

$$c_i^j - c_i^2 c_0^{j-1} + \frac{c_i c_0 - c_i^2}{c_0^2} \cdot c_0^j = c_i^j - c_i^2 c_0^{j-2}$$

From 0:  $c_i^j - c_i^k c_0^{j-k}$

$(\mathcal{L}_{ij})_{k+1}$

$$(\mathcal{L}_{ij})_k = \frac{\mathcal{L}_{ij}}{c_0}$$

$$\frac{c_i^j - c_0^j}{\mathcal{L}_{ij}} = \frac{c_i^k}{c_0^k}$$

$$(\mathcal{L}_{ij})_k = (\mathcal{L}_{i, k+1})_k c_0^{j-k-1} \quad (\mathcal{L}_{ij})_k = (\mathcal{L}_{ij})_k - \frac{(\mathcal{L}_{i, k+1})_k}{c_0^{k+1}}$$

$$(\mathcal{L}_{ij})_{k+1} = (\mathcal{L}_{ij})_k - \frac{(\mathcal{L}_{i, k+1})_k}{c_0^{k+1}} \mathcal{L}_{ij}$$

$$c_i^j - c_i^2 c_0^{j-2} + c_i c_0^{j-1} - c_i^2 c_0^{j-2}$$

$$c_i^j \rightarrow c_i^j - c_i^{i-1} c_0^{j+1-i} \quad \star c_i^j \rightarrow c_i^j - c_i^{i-1} c_0 \quad \star$$

From 1:  $c_i^j - c_i^{k-1} c_0^{j-k-1}$

22

Idea

$$m_j - c_0 m_{j-1}$$

$$c_i^j - c_0 c_i^{j-1} = c_i^{j-1} (c_i - c_0)$$

Proof

Clearly, if  $M \in M_{n+1}(\mathbb{F})$ ,  $\det(M) = 1 = \prod_{0 \leq i < j \leq n} (c_j - c_i)$ , the empty product. So assume this is true of all  $(n+1) \times (n+1)$  matrices, and let  $M \in M_{(n+2) \times (n+2)}(\mathbb{F})$  instead, with  $m_j$  being its  $j$ th column. Starting from  $j = n+2$ , we subtract  $c_0$  times of  $m_{j-1}$  from  $m_j$  till we reach  $j=1$ , hence transforming each  $c_i^j$  into  $c_i^{j-1} (c_i - c_0)$ . Thus, we have that by assumption / our induction hypothesis (I.H.):

$$\det(M) = \left( \prod_{i=1}^{n+1} (c_i - c_0) \right) \cdot \det \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 \\ (c_1 - c_0)^{-1} & 1 & c_1 & \cdots & c_1^n \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ (c_{n+1} - c_0)^{-1} & 1 & c_{n+1} & \cdots & c_{n+1}^n \end{pmatrix} = (-1)^{2+2} \cdot 1 \cdot \left( \prod_{i=1}^{n+1} (c_i - c_0) \right) \cdot \det \begin{pmatrix} 1 & c_1 & \cdots & c_1^n \\ \vdots & \vdots & \cdots & \vdots \\ 1 & c_{n+1} & \cdots & c_{n+1}^n \end{pmatrix}$$

$$\stackrel{\text{I.H.}}{=} \left( \prod_{i=1}^{n+1} (c_i - c_0) \right) \cdot \left( \prod_{1 \leq i < j \leq n+1} (c_j - c_i) \right)$$

$$= \prod_{0 \leq i < j \leq n+1} (c_j - c_i)$$

Therefore, the statement is true of all  $M \in M_{(n+2) \times (n+2)}(\mathbb{F})$ . Consequently, it is true by induction for all matrices in  $\mathbb{F}$ . □

22. (a) Ideas

Goal:  $\underbrace{[(c_1 - c_0)(c_2 - c_0) \cdots (c_n - c_0)]}_n \underbrace{[(c_2 - c_1)(c_3 - c_1) \cdots (c_n - c_1)]}_{n-1} \cdots \underbrace{[(c_n - c_{n-1})]}_1$

$$c_3^2 - c_3 c_0 \quad c_3^3 - c_3 c_0^2 + \frac{c_3 c_0 - c_3^2}{c_0^2} = c_3^3 - c_3 c_0^2 + c_3 c_0^2 - c_3^2 c_0 = c_3^3 - c_3^2 c_0$$

$$1(c_1 - c_0) (c_2^2 - c_2 c_0) (c_3^3 - c_3^2 c_0) \cdots (c_n^n - c_n^{n-1} c_0)$$

$$c_2 c_3^2 c_4^3 \cdots c_n^{n-1} (c_1 - c_0)(c_2 - c_0) \cdots (c_n - c_0)$$

$$n=2: (c_2 - c_1)(c_2 - c_0)(c_1 - c_0) = c_2(c_1 - c_0)(c_2 - c_0)$$

When  $n=0: 1=1$

$n=1: c_1 - c_0 = c_1 - c_0$

$$1(c_1 - c_0)(c_2^2 - c_2 c_0)$$

$$c_2(c_1 - c_0)(c_2 - c_0)$$

Assume true for  $n-1$ ,  $\Pi_n = (c_n - c_0)(c_n - c_1) \cdots (c_n - c_{n-1}) \Pi_{n-1}$

Show  $(c_n - c_0)(c_n - c_1) \cdots (c_n - c_{n-1}) = c_n^{n-1} (c_n - c_0)$

$$\det \begin{pmatrix} 1 & c_0 & c_0^2 \\ 1 & c_1 & c_1^2 \\ 1 & c_2 & c_2^2 \end{pmatrix} = \det \begin{pmatrix} 1 & c_0 & c_0^2 \\ 0 & c_1 - c_0 & c_1^2 - c_0^2 \\ 0 & c_2 - c_0 & c_2^2 - c_0^2 \end{pmatrix} = \det \begin{pmatrix} 1 & c_0 & c_0^2 \\ 0 & c_1 - c_0 & c_1^2 - c_0^2 \\ 0 & 0 & c_2^2 - c_2 c_0 \end{pmatrix} = (c_1 - c_0)(c_2^2 - c_2 c_0)$$

$\frac{c_0 - c_2}{c_1 - c_0} (c_2^2 - c_0^2 - (c_2 - c_0)(c_1 + c_0))$   
 $= (c_2 - c_0)(c_2 + c_0 - c_1 - c_0)$   
 $= (c_2 - c_0)(c_2 - c_1)$

$\text{since } (c_2^2 - c_0^2) + \frac{c_0 - c_2}{c_1 - c_0} \cdot c_0^2$   
 $= c_2^2 - c_0^2 + c_0^2 - c_2 c_0$   
 $= c_2^2 - c_2 c_0$

Qn suggests this is  $(c_2 - c_1)(c_2 - c_0)(c_1 - c_0) = c_2(c_2 - c_0)(c_1 - c_0)$

$$\Rightarrow c_2 - c_1 = c_2$$

$$c_1 = 0 ???$$

25. Ideas

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 5 & 0 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} 5 \\ 0 \\ -15 \end{pmatrix} = \begin{pmatrix} -40 \\ 0 \\ 0 \end{pmatrix} = -40 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$$c_{11} = -5 \quad c_{12} = 0 \quad c_{13} = -15$$

$$\det = -40$$

$$c_{21} = 4 \quad c_{22} = -8 \quad c_{23} = 4 \quad \begin{pmatrix} 1 & 2 & 3 \\ 0 & 5 & 0 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} 4 \\ -8 \\ 4 \end{pmatrix} = \begin{pmatrix} -40 \\ 0 \\ 0 \end{pmatrix} = -40 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

When  $\det(A) = 0$ ,  $(A|0) \rightarrow$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} -3 \\ 6 \\ -3 \end{pmatrix} = 0$$

$$c_{11} = -3 \quad c_{12} = 6 \quad c_{13} = -3$$

$$A_{jk} = \det(A) \cdot e_j$$

Given  $\det(A) \neq 0$ ,

$$x_{jk} = \frac{\det(A_{jk})}{\det(A)} = \det(A_{jk}) = \det(A)$$

$$= (-1)^{j+k} \cdot \det(\bar{A}_{jk})$$

$$= c_{jk}$$

← (a)

b) Proof

23. (a) Clearly, for any  $m \times m$  submatrix  $A_m$  of  $A$  with nonzero determinant,  $\text{rank}(A_m) = m \leq \text{rank}(A) =: r$ . Let  $\{a_{k_1}, a_{k_2}, \dots, a_{k_r}\}$  be an extension of  $a_i$  to a basis for  $\mathbb{F}^n$ , where  $a_j$  is the  $j$ th row of  $A$  and  $1 = k_1 < k_2 < \dots < k_r \leq n$ . Then for each  $j \notin \{k_s \mid 1 \leq s \leq r\}$ ,  $a_j = \sum_{i=1}^k c_{ij} a_{k_i}$  for some scalars  $c_{ij}$ . By type 3 row operations, all these  $n-r$  rows  $a_j$  are reduced to 0 in the transformed matrix  $A'$ . Repeating this process on the columns of  $A'$  now, we get a matrix  $A''$  with  $n-r$  zero rows and  $n-r$  zero columns. Removing them gives a submatrix  $A_r$  of  $A$  that has rank  $r$ , so  $\det(A_r) \neq 0$ . As such,  $r = k$ .

(b)  
Proven in (a).

*Ops, A contains entries  $\pm 1$ , not  $\pm i$ !*

24. Let  $R_{i,j}(A+tI)$  denote the matrix  $A+tI$  after  $t^{-1}$  times of its  $i$ th row is added to its  $j$ th row. We see that

$$R_{n-1,n} R_{n-2,n-1} \dots R_{1,2}(A+tI) = \begin{pmatrix} t & 0 & 0 & \dots & 0 & a_0 \\ 0 & t & 0 & \dots & 0 & a_1 + t^{-1}a_0 \\ 0 & 0 & t & \dots & 0 & a_2 + t^{-1}a_1 + t^{-2}a_0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 0 & a_{n-1} + \sum_{i=0}^{n-1} t^{-i} a_i \\ & & & & & + t \end{pmatrix}$$

*ops forget the  $tI$  in the entry!*

$$t + t^{n-1} a_{n-1} + \sum_{i=0}^{n-1} t^i a_i = \left( \sum_{i=0}^{n-1} t^i a_i \right) + t^{n-1}$$

25. (a) Ideas

$$\det(B) \stackrel{+8}{=} \det(B^t) = (-1)^{k+j} \det((\tilde{B}^t)_{kj}) = (-1)^{k+j} \det(\tilde{B}_{jlk})$$

$$\text{Proof} \quad \text{By Theorem 4.8, } \det(B) \stackrel{+8}{=} \det(B^t) = (-1)^{k+j} \det((\tilde{B}^t)_{kj}) = (-1)^{k+j} \det(\tilde{B}_{jlk}) = (-1)^{k+j} \det(\tilde{A}_{jlk}) = c_{jk}, \text{ by doing}$$

cofactor expansion along the  $k$ th row of  $B^t$ .

(b) Ideas

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\begin{matrix} c_{11} = 1 & c_{21} = 0 \\ c_{12} = 0 & c_{22} = 1 \end{matrix}$$

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 4 \\ -3 \end{pmatrix} = 4 \begin{pmatrix} 1 \\ 3 \end{pmatrix} - 3 \begin{pmatrix} 2 \\ 4 \end{pmatrix} = \begin{pmatrix} 4 \\ 12 \end{pmatrix} - \begin{pmatrix} 6 \\ 12 \end{pmatrix} = \begin{pmatrix} -2 \\ 0 \end{pmatrix} = -2 \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \det \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \cdot e_1$$

$$\begin{matrix} c_{11} = 4 & c_{21} = -2 \\ c_{12} = -3 & c_{22} = 1 \\ \det = -2 & \det = -2 \end{matrix}$$

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ -2 \end{pmatrix}$$

$\sum_{j=1}^n c_j - c_i$   
05/05/2011  
17/1

$(c_{ij})_{j=1}^n$

$c_{11}$

$(-1)^{1+2}$

or by all

at all

25. Idea

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 5 & 0 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} 5 \\ 0 \\ -15 \end{pmatrix} = \begin{pmatrix} -40 \\ 0 \\ 0 \end{pmatrix} = -40 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$c_{11} = 5 \quad c_{12} = 0 \quad c_{13} = -15$

When  $\det(A) = 0$ ,  $(A|0) \rightarrow$

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 5 & 0 \\ -7 & 8 & 4 \end{pmatrix} \begin{pmatrix} -3 \\ 6 \\ -3 \end{pmatrix} = 0$$

$c_{11} = -3 \quad c_{12} = 6 \quad c_{13} = -3$

$A_{1k} = \det(A) \cdot e_j$   
Given  $\det(A) \neq 0$ ,

$$x_{1k} = \frac{\det(M_{1k})}{\det(A)} = \det(N_k) = \det(U_k)^k$$

$$= (-1)^{j+k} \cdot \mathbb{1} \cdot \det(\bar{A}_{jk})$$

← (a)

$$= c_{jk}$$

$\det = -40$

$c_{21} = 4 \quad c_{22} = -8 \quad c_{23} = 4$

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 5 & 0 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} 4 \\ -8 \\ 4 \end{pmatrix} = \begin{pmatrix} -40 \\ 0 \\ 0 \end{pmatrix} = -40 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

(b) Proof

R. 1. matrix A

26. (a)  $\begin{pmatrix} A_{22} & -A_{21} \\ -A_{21} & A_{11} \end{pmatrix}^t \begin{pmatrix} A_{22} & -A_{21} \\ -A_{21} & A_{11} \end{pmatrix}$

(e)  $\begin{pmatrix} -3i & 4 & 4+10i \\ 0 & -1+i & -5-3i \\ 0 & 0 & 3+3i \end{pmatrix}^t \begin{pmatrix} -3i & 0 & 0 \\ 4 & -1+i & 0 \\ 10+16i & -5-3i & 3+3i \end{pmatrix}$

One has thinking of 25.

26. (e)  $\det = (-3i)(-1+i)(3+3i) = (3-3i)(-3+3i) = -(9-9-18i) = 18i$   $A \det = -3-3i \quad (-3-3i)^2 = 18i$

27. (a) Ideas

$\det(C) = \sum_{j=1}^{n-1} (-1)^{2+j} c_{2j} \det(\tilde{C}_{2j})$

$[\det(A)]^{-1} = \left[ \sum_{j=1}^{n-1} (-1)^{2+j} A_{2j} \det(\tilde{A}_{2j}) \right]^{-1} = \left[ \sum_{j=1}^{n-1} A_{2j} C_{2j} \right]^{-1}$

Find a matrix B so the cofactor of row i, column k of B is  $(\tilde{C}_{2j})_{ik}$   $1 \leq i \leq n, 1 \leq k \leq n$

$(-1)^{i+k} \det(\tilde{B}_{ik}) = (-1)^{i+k} \det(\tilde{A}_{i+1, k})$  if  $k < j$   
 $= (-1)^{i+k} \det(\tilde{A}_{i+1, k+1})$  if  $k \geq j$

(a) Proof 1

From exercise 25. (c), it is straightforward to notice that  $\det(A) \det(C) = [\det(A)]^{-1}$  so  $\det(C) = [\det(A)]^{-n-1}$ .

(b) we see that  $(C^t)_{ij} = c_{ji} = (-1)^{j+i} \det(\tilde{A}_{ij}) = (-1)^{j+i} \det(\tilde{A}_{ij})^t = (-1)^{i+j} \det((\tilde{A}^t)_{ji})$ , the  $ji$ -cofactor of  $A^t$ , since removing row i and column j then transposing is the same as removing row j and column i after transposing.

(c) Ideas

$A = \begin{pmatrix} a & b & c \\ 0 & d & e \\ 0 & 0 & f \end{pmatrix} \begin{matrix} adf \neq 0 \\ a, d, f \neq 0 \end{matrix} \Rightarrow \begin{matrix} c_{22} = -bf \\ c_{32} = be - cd \\ c_{33} = -ae \end{matrix} \Rightarrow C = \begin{pmatrix} df & 0 & 0 \\ -bf & ad & 0 \\ becd - ae & ad & ad \end{pmatrix}^t$

$\det(\tilde{A}_{ij})$

$\left( N \mid \begin{matrix} I_n \\ e_{n+1} \end{matrix} \right) \rightarrow \left( \begin{matrix} I_n \\ ae_{n+1} \end{matrix} \mid \begin{matrix} N^{-1} \\ e_{n+1} \end{matrix} \right)$

Proof 1

When  $n=1$ , there is nothing to prove since all  $1 \times 1$  matrices are upper triangular. So assume  $B^{-1}$  is upper triangular if  $B \in M_{n-1}(\mathbb{F})$  is, and let  $A \in M_{n \times n}(\mathbb{F})$ . Thus, by some elementary row operations,  $(A \mid I_n) \rightarrow \left( \begin{matrix} I_{n-1} & c_1 & \dots & c_{n-1} \\ 0 & \dots & 0 & \tilde{A}_{nn} \end{matrix} \mid \begin{matrix} \tilde{A}_{nn}^{-1} & 0 \\ \dots & \dots \\ 0 & \dots & 0 & 1 \end{matrix} \right) \rightarrow \left( \begin{matrix} I_n & \tilde{A}_{nn}^{-1} & \dots & -c_1 \tilde{A}_{nn}^{-1} \\ \dots & \dots & \dots & \dots \\ 0 & \dots & 0 & \tilde{A}_{nn} \end{matrix} \right)$   
 Since  $\tilde{A}_{nn}^{-1}$  is upper triangular by assumption / our induction hypothesis,  $A^{-1} = \begin{pmatrix} \tilde{A}_{nn}^{-1} & -c_1 \tilde{A}_{nn}^{-1} \\ \dots & \dots \\ 0 & \dots & 0 & \tilde{A}_{nn} \end{pmatrix}$  is clearly also upper triangular, as expected.  
 Hence, by induction, for any upper triangular matrix A,  $A^{-1}$  is also upper triangular. By exercise 25(d),  $C = \det(A) \cdot A^{-1}$  is also upper triangular.

$b_{ij}(A+tI)$  denote the  $n \times n$  matrix  $R_{n-2, n-1, \dots, R}$

$\det(A) \neq 0$ .  $A^{-1}$  with  $n \times n$  rank  $(A)$

Think about alt. proof of 27. (4)

Alternate Proofs

27. (4) Ideas

$$\det(C) = [\det(A)]^{-1} = [\det(A)]^{-1} \det(\det(A)I) \Rightarrow \det(C) = \det(A)I$$

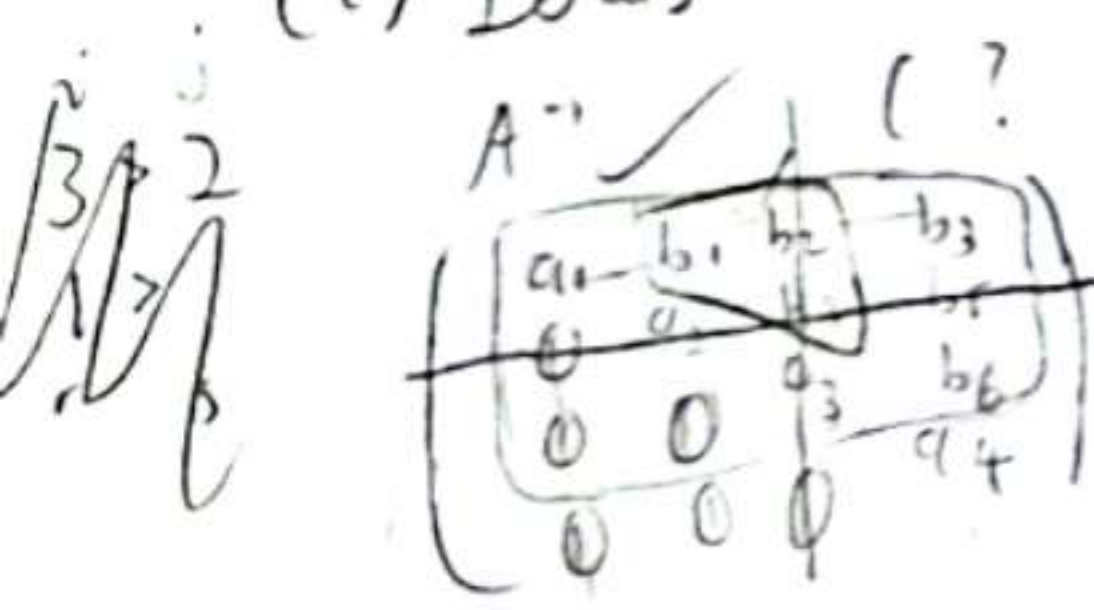
$$\det(C) = \det(A^{n-1})$$

$$=$$

$$\begin{cases} A_{rs} & \text{if } r < j, s < i \\ A_{rs} & \text{if } r \geq j, s < i \\ A_{r, s+1} & \text{if } r < j, s \geq i \\ A_{r, s+1} & \text{if } r \geq j, s \geq i \end{cases}$$

$$r < j < i \leq s$$

(c) Ideas



Proof 2

In Proof 1, we showed  $A^{-1}$  is upper triangular without using exercise 25. So, we shall now show a proof for  $C$  being upper triangular, without using exercise 25. We notice that

$$(\tilde{A}_{ji})_{rs} = \begin{cases} A_{rs} & \text{if } r < j \neq s < i, \\ A_{rs} & \text{if } r \geq j \neq s < i, \\ A_{r, s+1} & \text{if } r < j \neq s \geq i, \\ A_{r, s+1} & \text{if } r \geq j \neq s \geq i. \end{cases}$$

Thus, when  $n \geq i > j \geq 1$  and  $r > s$ , the only possibility for  $(\tilde{A}_{ji})_{rs} \neq 0$  is if  $r < j \neq s \geq i$ , such that  $(\tilde{A}_{ji})_{rs} = A_{r, s+1}$ . But then  $r < j < i \leq s$ , a contradiction. Therefore,  $\tilde{A}_{ji}$  must be upper triangular.

Furthermore,  $(\tilde{A}_{ji})_{jj} = A_{j+1, j} = 0$  means  $\det(\tilde{A}_{ji})_{jj} = 0$ . Hence,  $C_{ij} = 0$ . (consequently,

$C$  is certainly upper triangular.

$$\begin{pmatrix} a_1 & b_1 & b_2 \\ 0 & 0 & b_3 \\ 0 & 0 & a_2 \end{pmatrix}$$