

MA2001

LIVE LECTURE 8

Q&A: log in to Pollev.com/vtpoll

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Topics for week 8

4.1 Row spaces and Column spaces

4.2 Ranks

4.3 Nullspaces and Nullities

Row Space and Column Space

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 2 & 1 \\ 3 & 6 & 6 & 3 \\ 4 & 9 & 9 & 5 \\ -2 & -1 & -1 & 1 \\ 5 & 8 & 9 & 4 \\ 4 & 2 & 7 & 3 \end{pmatrix} \begin{matrix} \mathbf{r}_1 \\ \mathbf{r}_2 \\ \mathbf{r}_3 \\ \mathbf{r}_4 \\ \mathbf{r}_5 \\ \mathbf{r}_6 \end{matrix} \xrightarrow[\text{Elimination}]{\text{Gaussian}} \begin{pmatrix} 1 & 2 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{matrix} \mathbf{v}_1 \\ \mathbf{v}_2 \\ \mathbf{v}_3 \\ \\ \\ \end{matrix}$$

pivot columns

REF

row space of $\mathbf{A} = \text{span}\{\mathbf{r}_1 \mathbf{r}_2 \mathbf{r}_3 \mathbf{r}_4 \mathbf{r}_5 \mathbf{r}_6\}$

column space of $\mathbf{A} = \text{span}\{\mathbf{c}_1 \mathbf{c}_2 \mathbf{c}_3 \mathbf{c}_4\}$

Basis for row space of $\mathbf{A} = \{\mathbf{v}_1 \mathbf{v}_2 \mathbf{v}_3\}$

Basis for column space of $\mathbf{A} = \{\mathbf{c}_1 \mathbf{c}_2 \mathbf{c}_3\}$

Bases are
not unique

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

Row space, column space, nullspace

m x n matrix \mathbf{A}	Subspace	Basis	Dimension
Row space	Subspace of \mathbf{R}^n	Non-zero rows in REF	Rank (# non-zero rows in REF)
Column space	Subspace of \mathbf{R}^m	Corresponding “pivot” columns in \mathbf{A}	Rank (# pivot columns in REF)
Nullspace	Subspace of \mathbf{R}^n Same as solution space of $\mathbf{Ax} = \mathbf{0}$	From the spanning vectors in the general solution	Nullity (# parameters in general solution)

Rank

If \mathbf{R} is a row-echelon form of \mathbf{A} ,

$\text{rank}(\mathbf{A})$ = the number of nonzero rows of \mathbf{R}

= the number of leading entries in \mathbf{R}

= the number of pivot columns in \mathbf{R}

= largest number of linearly independent rows in \mathbf{A} = \dim (row space of \mathbf{A})

= largest number of linearly independent columns in \mathbf{A} = \dim (column space of \mathbf{A})

For an $m \times n$ matrix \mathbf{A} , $\text{rank}(\mathbf{A}) \leq \min\{m, n\}$.

An $m \times n$ matrix \mathbf{A} with $\text{rank}(\mathbf{A}) = \min\{m, n\}$ is said to be of full rank.

Rank(**A**) = **largest** number of **linearly independent rows** in **A**
= **largest** number of **linearly independent columns** in **A**

Rank by inspection

What is the rank of each of the following matrices?

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 2 \\ 3 & 3 & 3 \end{pmatrix} \quad \text{rank}(\mathbf{A}) = 1$$

All rows are scalar multiples of each other

$$\mathbf{B} = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 3 & 4 \\ 0 & 0 & 0 \end{pmatrix} \quad \text{rank}(\mathbf{B}) = 2$$

Two rows that are not scalar multiples of each other

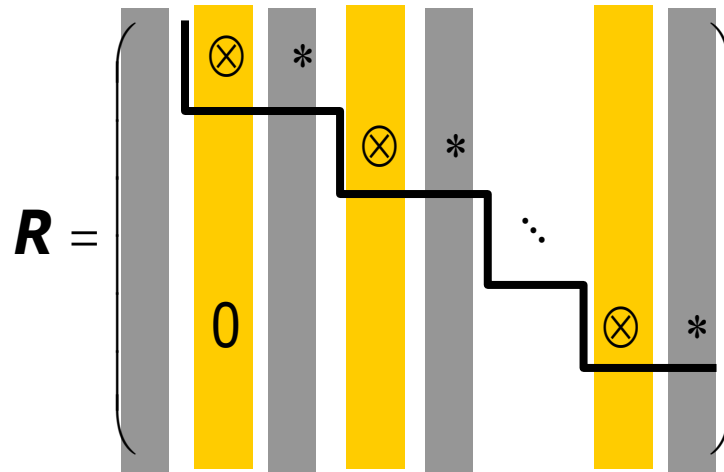
$$\mathbf{C} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \quad \text{rank}(\mathbf{C}) = 2$$

Third row is the sum of first two rows
Third column is the same as first column

Dimension Theorem of Matrix

$$\text{rank}(\mathbf{A}) + \text{nullity}(\mathbf{A}) = \# \text{ columns of } \mathbf{A}$$

row-echelon
form



 pivot columns

(correspond to basis for column space of \mathbf{A}) $\text{rank}(\mathbf{A})$

 non-pivot columns

(correspond to parameters in general solutions) $\text{nullity}(\mathbf{A})$

Row equivalence

$$\mathbf{A} \rightarrow \rightarrow \rightarrow \mathbf{B}$$

Preserve **row space** and **rank**

Preserve **nullspace** and **nullity**

Does not preserve column space, but preserve **linearity relations among columns**

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 2 & 1 \\ 3 & 6 & 6 & 3 \\ 4 & 9 & 9 & 5 \\ -2 & -1 & -1 & 1 \\ 5 & 8 & 9 & 4 \\ 4 & 2 & 7 & 3 \end{pmatrix} \xrightarrow[\text{Elimination}]{\text{Gaussian}} \mathbf{R} = \begin{pmatrix} 1 & 2 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

For both \mathbf{A} and \mathbf{R} :

- column#1 + column#4 = column#3
- column#1, #3, #4 are linearly dependent
- column#2, #3, #4 are linearly independent

True or False

If \mathbf{A} is an $n \times n$ invertible matrix,
then the row space of \mathbf{A} is \mathbf{R}^n



The rows of \mathbf{A} form
a basis for \mathbf{R}^n



$\text{span}\{\text{rows of } \mathbf{A}\} = \mathbf{R}^n$



If \mathbf{A} is an $n \times n$ invertible matrix,
then the column space of \mathbf{A} is \mathbf{R}^n



The columns of \mathbf{A}
form a basis for \mathbf{R}^n



$\text{span}\{\text{columns of } \mathbf{A}\} = \mathbf{R}^n$



Basis for Linear Span

Let $V = \text{span}\{ \mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k \}$ Basis

We need to remove the redundant vectors among $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ to get a basis for V .

Column space method:

- Form a matrix \mathbf{A} using $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ as its columns.
- Reduce \mathbf{A} to row echelon form \mathbf{R}
- Identify the pivot columns in \mathbf{R}
- The corresponding columns in \mathbf{A} form a basis for V .

Example

$\text{span}\{(1,2,2,1), (3,6,6,3), (4,9,9,5), (-2,-1,-1,1), (5,8,9,4)\}$

$$\begin{pmatrix} 1 & 3 & 4 & -2 & 5 \\ 2 & 6 & 9 & -1 & 8 \\ 2 & 6 & 9 & -1 & 9 \\ 1 & 3 & 5 & 1 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 3 & 4 & -2 & 5 \\ 0 & 0 & 1 & 3 & -2 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

Basis for the linear span:

$\{(1,2,2,1), \textcolor{red}{(3,6,6,3)}, (4,9,9,5), \textcolor{red}{(-2,-1,-1,1)}, (5,8,9,4)\}$

Deriving basis for \mathbb{R}^n

$$S = \{(2, -1, 0), (1, -1, 3), (-5, 1, 0), (1, 0, 1)\}$$

How to get a basis from S for \mathbb{R}^3 ?

Throw out redundant vectors from S

Arrange the vectors as **columns** of a matrix

Look for **pivot columns** of the REF

$$T = \left\{ \begin{pmatrix} 2 \\ 1 \\ -5 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 3 \\ 0 \\ 1 \end{pmatrix} \right\}$$

How to extend T to a basis for \mathbb{R}^4 ?

Add on non-redundant vectors to T

Arrange the vectors as **rows** of a matrix

Look for '**missing**' leading entries of the REF

Extending basis

$$S = \{(1, 4, -2, 5, 1), (2, 9, -1, 8, 2), (2, 9, -1, 9, 3)\}$$

complete R to a 5×5 matrix

$$\mathbf{A} = \begin{pmatrix} 1 & 4 & -2 & 5 & 1 \\ 2 & 9 & -1 & 8 & 2 \\ 2 & 9 & -1 & 9 & 3 \end{pmatrix}$$

Gaussian
Elimination

$$\mathbf{R} = \begin{pmatrix} 1 & 4 & -2 & 5 & 1 \\ 0 & 1 & 3 & -2 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & x & * & * \\ 0 & 0 & 0 & 0 & y \end{pmatrix}$$

form a basis for \mathbf{R}^5

are not redundant
in row space of A

E.g. $(0 \ 0 \ 1 \ 0 \ 0)$

E.g. $(0 \ 0 \ 0 \ 0 \ 1)$

$$\{ (1, 4, -2, 5, 1), (2, 9, -1, 8, 2), (2, 9, -1, 9, 3), \\ (0, 0, 1, 0, 0), (0, 0, 0, 0, 1) \}$$

True or False

We can also use column space method to extend a set in \mathbf{R}^n to a basis for \mathbf{R}^n

$$\mathbf{A} = (\mathbf{c}_1 \ \mathbf{c}_2 \ \dots \ \mathbf{c}_k) \xrightarrow{\text{row echelon form}} \mathbf{R} = (\underbrace{\mathbf{c}_1^* \ \mathbf{c}_2^* \ \dots \ \mathbf{c}_k^*}_{\text{add the "missing" pivot columns}})$$

$$\mathbf{A} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{pmatrix} \xrightarrow{\text{G.E.}} \mathbf{R} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \quad \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$\mathbf{Av} \in \text{Column space of } \mathbf{A}$ Visualization

$$\mathbf{A}: m \times n$$

$$\mathbf{A} = (\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n) \quad \mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$

$$\mathbf{Av} = (\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n) \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$

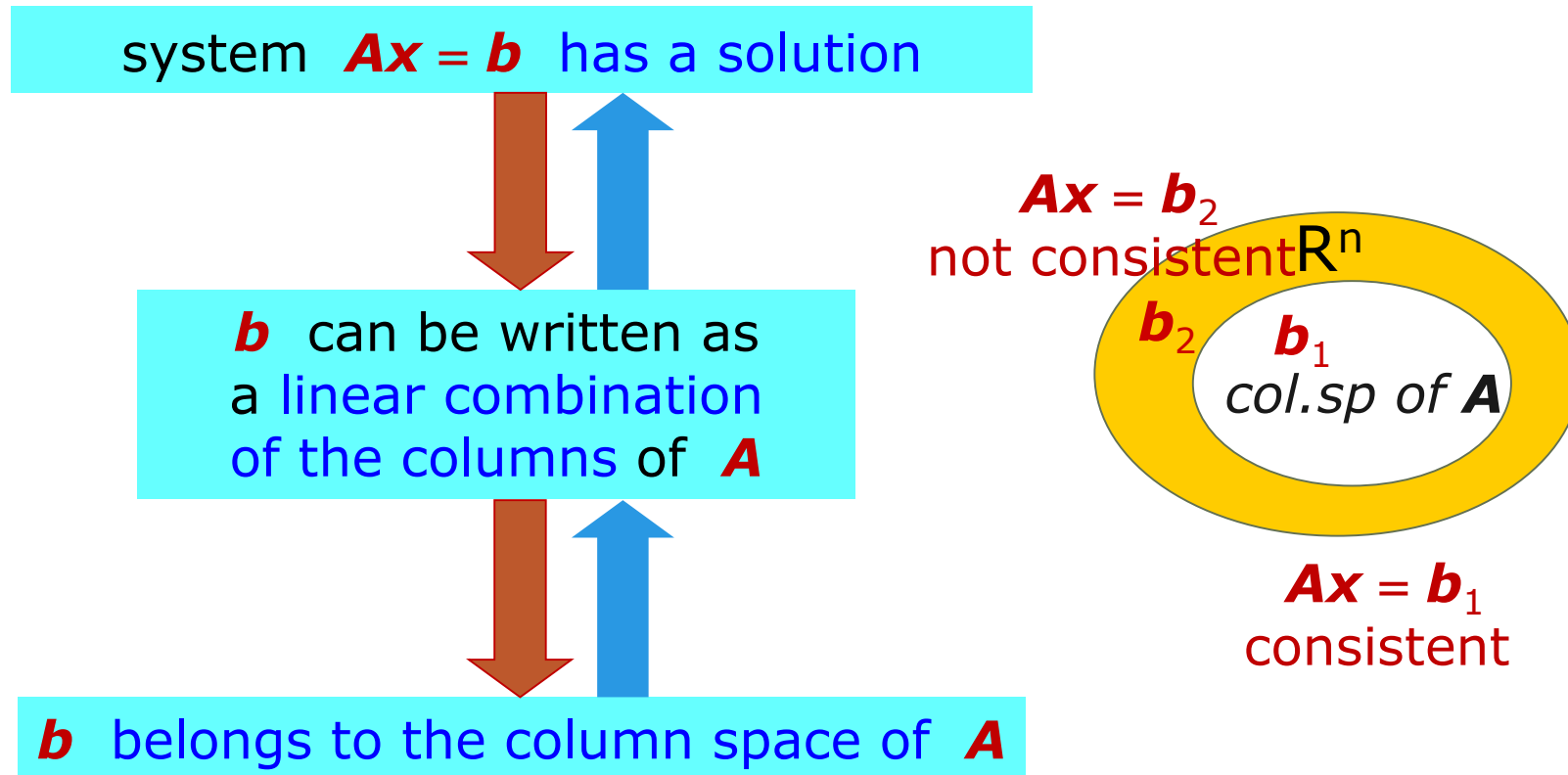
$$= v_1 \mathbf{a}_1 + v_2 \mathbf{a}_2 + \dots + v_n \mathbf{a}_n$$

a linear combination of the columns of \mathbf{A}
belong to column space of \mathbf{A}

For any $\mathbf{v} \in \mathbf{R}^n$, \mathbf{Av} belongs to the column space of \mathbf{A}

Every vector in the column space of \mathbf{A}
has the form \mathbf{Av} for some $\mathbf{v} \in \mathbf{R}^n$

Consistency of LS and column space



Column space of AB

$A: m \times n$

$B: n \times k \quad B = (b_1 \ b_2 \ \dots \ b_k)$

$$AB = (Ab_1 \ Ab_2 \ \dots \ Ab_k)$$

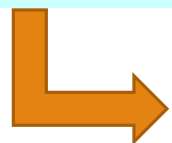
columns of AB in terms of columns of B

each column of $AB \in$ column space of A

column space of $AB \subseteq$ column space of A

Rank(AB) VS Rank(A)

column space of $AB \subseteq$ column space of A



$$\text{rank}(AB) \leq \text{rank}(A) \quad \checkmark$$

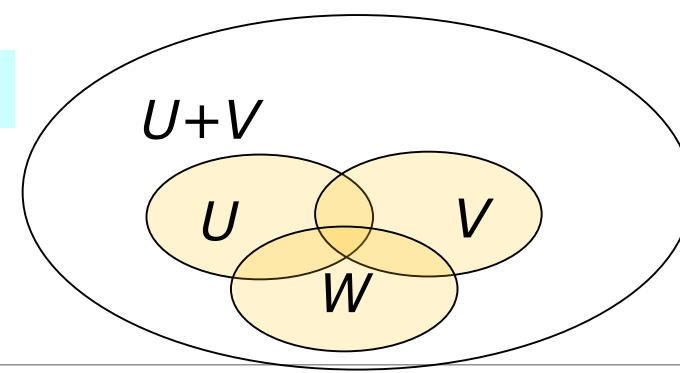
True or false:

column space of $AB \subseteq$ column space of $B \quad \times$

$$\text{rank}(AB) \leq \text{rank}(B) \quad \checkmark$$

row space of $AB \subseteq$ row space of $B \quad \checkmark$

$$\dim(U + V) = \dim(U) + \dim(V) - \dim(U \cap V)$$



Exercise 4 Q23

$$\text{rank}(\mathbf{A} + \mathbf{B}) \leq \text{rank}(\mathbf{A}) + \text{rank}(\mathbf{B})$$

$$\mathbf{A} = (\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n)$$

$$U = \text{Column space of } \mathbf{A} = \text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$$

$$\mathbf{B} = (\mathbf{b}_1 \ \mathbf{b}_2 \ \dots \ \mathbf{b}_n)$$

$$V = \text{Column space of } \mathbf{B} = \text{span}\{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n\}$$

$$\mathbf{A} + \mathbf{B} = (\mathbf{a}_1 + \mathbf{b}_1 \ \mathbf{a}_2 + \mathbf{b}_2 \ \dots \ \mathbf{a}_n + \mathbf{b}_n)$$

$$W = \text{Column space of } \mathbf{A} + \mathbf{B} = \text{span}\{\mathbf{a}_1 + \mathbf{b}_1, \mathbf{a}_2 + \mathbf{b}_2, \dots, \mathbf{a}_n + \mathbf{b}_n\}$$

$$U + V = \text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n, \mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n\}$$

$$W \subseteq U + V \Rightarrow \dim W \leq \dim(U + V) \leq \dim(U) + \dim(V)$$

Exercise 4 Q20

Suppose $\mathbf{AB} = \mathbf{0}$.

Show that column space of $\mathbf{B} \subseteq$ nullspace of \mathbf{A}

$$\mathbf{B} = (\mathbf{b}_1 \ \mathbf{b}_2 \ \dots \ \mathbf{b}_k) \quad \mathbf{AB} = (\mathbf{Ab}_1 \ \mathbf{Ab}_2 \ \dots \ \mathbf{Ab}_k)$$

$$\mathbf{AB} = \mathbf{0}$$

$$\Rightarrow (\mathbf{Ab}_1 \ \mathbf{Ab}_2 \ \dots \ \mathbf{Ab}_k) = \mathbf{0}$$

$$\Rightarrow \mathbf{Ab}_1 = \mathbf{0}, \ \mathbf{Ab}_2 = \mathbf{0}, \ \dots, \ \mathbf{Ab}_k = \mathbf{0}$$

$$\Rightarrow \mathbf{b}_1, \ \mathbf{b}_2, \ \dots, \ \mathbf{b}_k \in \text{nullspace of } \mathbf{A}$$

$$\Rightarrow \text{column space of } \mathbf{B} \subseteq \text{nullspace of } \mathbf{A}$$

Solution set of non-homogeneous system

If we know the general solution of $\mathbf{Ax} = \mathbf{0}$ and one particular solution of $\mathbf{Ax} = \mathbf{b}$, then we have the general solution of $\mathbf{Ax} = \mathbf{b}$.

general solution
of $\mathbf{Ax} = \mathbf{b}$



general solution
of $\mathbf{Ax} = \mathbf{0}$

nullspace of \mathbf{A}

+ \mathbf{v} ← Fixed solution of $\mathbf{Ax} = \mathbf{b}$

The solution set of $\mathbf{Ax} = \mathbf{b} = \{ \mathbf{u} + \mathbf{v} \mid \mathbf{u} \text{ belongs to nullspace of } \mathbf{A} \}$

Solving Linear system with MATLAB

$$\mathbf{Ax} = \mathbf{0}$$

>> rref([A 0])

>> null(A) (give basis for nullspace of A)

$$\mathbf{Ax} = \mathbf{b} \text{ (consistent)}$$

>> rref([A b])

>> A \ b

>> linsolve(A, b)

} (give a particular solution)

general solution
of $\mathbf{Ax} = \mathbf{b}$

The diagram illustrates the construction of the general solution for the linear system $\mathbf{Ax} = \mathbf{b}$. It shows two paths leading to the final result. The first path starts with the command `>> null(A)`, which provides a basis for the nullspace of A . The second path starts with the command `>> A \ b` (or `>> linsolve(A, b)`), which provides a particular solution. Both of these components are then combined to form the general solution of the system.

Exercise 4 Q25(a) [Tutorial 7]

nullspace of \mathbf{A} equals nullspace of $\mathbf{A}^T\mathbf{A}$

Strategy: Show $S \subseteq T$ and $T \subseteq S$

$$\mathbf{v} \in \text{nullspace of } \mathbf{A} \Rightarrow \mathbf{A}\mathbf{v} = \mathbf{0} \Rightarrow \mathbf{A}^T\mathbf{A}\mathbf{v} = \mathbf{0}$$

$$\Rightarrow \mathbf{v} \in \text{nullspace of } \mathbf{A}^T\mathbf{A}$$

So nullspace of $\mathbf{A} \subseteq \text{nullspace of } \mathbf{A}^T\mathbf{A}$

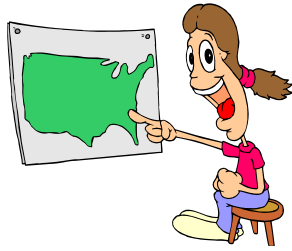
$$\mathbf{v} \in \text{nullspace of } \mathbf{A}^T\mathbf{A} \Rightarrow \mathbf{A}^T\mathbf{A}\mathbf{v} = \mathbf{0} \Rightarrow \mathbf{v}^T\mathbf{A}^T\mathbf{A}\mathbf{v} = 0 \Rightarrow (\mathbf{A}\mathbf{v})^T\mathbf{A}\mathbf{v} = 0 \Rightarrow \mathbf{A}\mathbf{v} = \mathbf{0}$$

$$\Rightarrow \mathbf{v} \in \text{nullspace of } \mathbf{A}$$

hint: sum of squares

So nullspace of $\mathbf{A}^T\mathbf{A} \subseteq \text{nullspace of } \mathbf{A}$

Map of LA



A is an $n \times n$ matrix

A is invertible	chapter 2	A is not invertible
$\det A \neq 0$	chapter 2	$\det A = 0$
rref of A is identity matrix	chapter 1	rref of A has a zero row
$AX = 0$ has only the trivial solution	chapter 1	$AX = 0$ has non-trivial solutions
$AX = B$ has a unique solution	chapter 1	$AX = B$ has no solution or infinitely many solutions
rows (columns) of A are linearly independent	chapter 3	rows (columns) of A are linearly dependent
row (column) space of $A = \mathbb{R}^n$	chapter 4	row (column) space of $A \neq \mathbb{R}^n$
i.e. A is full rank $\text{rank}(A) = n$	chapter 4	$\text{rank}(A) < n$
i.e. nullspace of A is the zero space $\text{nullity}(A) = 0$	chapter 4	$\text{nullity}(A) > 0$

to be continued