

MA2001

LIVE LECTURE 9

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Topics for week 9

5.1 Inner Products in \mathbf{R}^n

5.2 Orthogonal and Orthonormal Bases

5.3 Best Approximation (preview)

Dot product

$\mathbf{u} = (u_1, u_2, \dots, u_n)$, $\mathbf{v} = (v_1, v_2, \dots, v_n)$ vectors in \mathbf{R}^n

The **dot product** of \mathbf{u} and \mathbf{v} is defined as

product of two vectors $\mathbf{u} \cdot \mathbf{v} = u_1v_1 + u_2v_2 + \dots + u_nv_n$ scalar

\mathbf{u} and \mathbf{v} regarded as row matrices $\mathbf{u} \cdot \mathbf{v} = \mathbf{u}\mathbf{v}^T$

\mathbf{u} and \mathbf{v} regarded as column matrices $\mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v}$

$$\mathbf{A}\mathbf{b} = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{a}_n \end{pmatrix} \mathbf{b} = \begin{pmatrix} \mathbf{a}_1 \cdot \mathbf{b} \\ \mathbf{a}_2 \cdot \mathbf{b} \\ \vdots \\ \mathbf{a}_n \cdot \mathbf{b} \end{pmatrix}$$

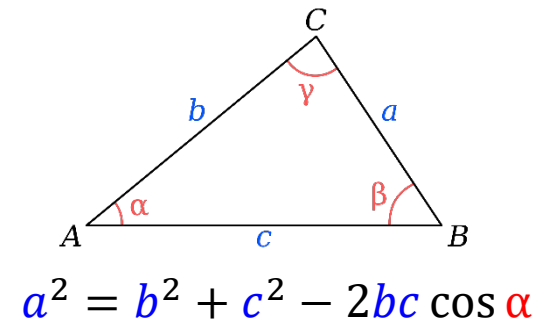
\mathbf{u} regarded as row matrix, \mathbf{v} regarded as column matrix ?

$$\mathbf{u} \cdot \mathbf{v} = \mathbf{u}\mathbf{v}$$

Length, distance and angles in \mathbf{R}^n

$\mathbf{u} = (u_1, u_2, \dots, u_n)$, $\mathbf{v} = (v_1, v_2, \dots, v_n)$ vectors in \mathbf{R}^n

Dot product	$\mathbf{u} \cdot \mathbf{v}$	$u_1v_1 + u_2v_2 + \dots u_nv_n$
Norm (length)	$ \mathbf{u} $	$\sqrt{\mathbf{u} \cdot \mathbf{u}}$ $\sqrt{u_1^2 + u_2^2 + \dots + u_n^2}$
Distance	$ \mathbf{u} - \mathbf{v} $	$\sqrt{(\mathbf{u} - \mathbf{v}) \cdot (\mathbf{u} - \mathbf{v})}$ $\sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + \dots + (u_n - v_n)^2}$
Angle	between \mathbf{u} and \mathbf{v}	$\cos^{-1}\left(\frac{\mathbf{u} \cdot \mathbf{v}}{ \mathbf{u} \mathbf{v} }\right)$ $\cos^{-1}\left(\frac{u_1v_1 + u_2v_2 + \dots + u_nv_n}{ \mathbf{u} \mathbf{v} }\right)$



Comparing vectors

Among the three vectors $\mathbf{u}_1 = (1,1,0,0)$, $\mathbf{u}_2 = (0,1,1,0)$, $\mathbf{u}_3 = (1,0,0,1)$ which is the **best approximation** of $\mathbf{v} = (1,2,3,4)$?

Compare the distance between \mathbf{v} and each of \mathbf{u}_1 , \mathbf{u}_2 , \mathbf{u}_3

$$\|\mathbf{v} - \mathbf{u}_1\| = \sqrt{(1-1)^2 + (2-1)^2 + (3-0)^2 + (4-0)^2} = \sqrt{26}$$

$$\|\mathbf{v} - \mathbf{u}_2\| = \sqrt{(1-0)^2 + (2-1)^2 + (3-1)^2 + (4-0)^2} = \sqrt{22}$$

$$\|\mathbf{v} - \mathbf{u}_3\| = \sqrt{(1-1)^2 + (2-0)^2 + (3-0)^2 + (4-1)^2} = \sqrt{22}$$

The smaller the **distance** between two vectors, the better is the **approximation**

The smaller the **angle** between two vectors, the better is the **approximated ratio** among the coordinates

Properties of dot product

Let c be a scalar and $\mathbf{u}, \mathbf{v}, \mathbf{w}$ vectors in \mathbf{R}^n .

1. $\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$ commutative law

2. $(\mathbf{u} + \mathbf{v}) \cdot \mathbf{w} = \mathbf{u} \cdot \mathbf{w} + \mathbf{v} \cdot \mathbf{w}$
 $\mathbf{w} \cdot (\mathbf{u} + \mathbf{v}) = \mathbf{w} \cdot \mathbf{u} + \mathbf{w} \cdot \mathbf{v}$ distributive law

3. $(c\mathbf{u}) \cdot \mathbf{v} = \mathbf{u} \cdot (c\mathbf{v}) = c(\mathbf{u} \cdot \mathbf{v})$ scalar mult.

4. $\|c\mathbf{u}\| = |c| \|\mathbf{u}\|$ norm (not $c \|\mathbf{u}\|$)

5. (i) $\mathbf{u} \cdot \mathbf{u} \geq 0$ $\mathbf{u} \cdot \mathbf{u} = \|\mathbf{u}\|^2$
(ii) $\mathbf{u} \cdot \mathbf{u} = 0$ if and only if $\mathbf{u} = \mathbf{0}$.

6. $|\mathbf{u} \cdot \mathbf{v}| \leq \|\mathbf{u}\| \cdot \|\mathbf{v}\|$ Cauchy-Schwarz Inequality

“Orthogonal” meaning

- Two vectors \mathbf{u} and \mathbf{v} are orthogonal if $\mathbf{u} \cdot \mathbf{v} = 0$.

(geometrically: \mathbf{u} and \mathbf{v} are perpendicular)

- A set $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$ is an orthogonal set if every pair of distinct vectors in S are orthogonal:

$$\mathbf{u}_1 \cdot \mathbf{u}_2 = 0, \mathbf{u}_1 \cdot \mathbf{u}_3 = 0, \dots, \mathbf{u}_{k-1} \cdot \mathbf{u}_k = 0$$

- A set $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$ is an orthogonal basis of a vector space V if S is an orthogonal set and a basis for V

- A vector \mathbf{u} is orthogonal to a vector space V if \mathbf{u} is orthogonal to every vector in V . i.e. $\mathbf{u} \cdot \mathbf{v} = 0$ for all $\mathbf{v} \in V$

Nullspace vs Row space

$$\mathbf{A}\mathbf{b} = \mathbf{0}$$

\mathbf{b} belongs to the nullspace of \mathbf{A}

$$\begin{pmatrix} \mathbf{a}_1 \cdot \mathbf{b} \\ \mathbf{a}_2 \cdot \mathbf{b} \\ \vdots \\ \mathbf{a}_n \cdot \mathbf{b} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

\mathbf{b} is orthogonal to every row of \mathbf{A}

Row space
of \mathbf{A}

Nullspace
of \mathbf{A}

$$(c_1\mathbf{a}_1 + c_2\mathbf{a}_2 + \dots + c_n\mathbf{a}_n) \cdot \mathbf{b}$$

$$c_1\mathbf{a}_1 \cdot \mathbf{b} + c_2\mathbf{a}_2 \cdot \mathbf{b} + \dots + c_n\mathbf{a}_n \cdot \mathbf{b} = 0$$

\mathbf{b} is orthogonal to the row space of \mathbf{A}

Orthogonal set VS orthonormal set

$$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$$

S is an orthogonal set

$$\mathbf{u}_1 \cdot \mathbf{u}_2 = 0, \mathbf{u}_1 \cdot \mathbf{u}_3 = 0, \dots, \mathbf{u}_{k-1} \cdot \mathbf{u}_k = 0$$

$$\|\mathbf{u}_1\| = \|\mathbf{u}_2\| = \dots = \|\mathbf{u}_k\| = 1$$

$$T = \left\{ \frac{1}{\|\mathbf{u}_1\|} \mathbf{u}_1, \frac{1}{\|\mathbf{u}_2\|} \mathbf{u}_2, \dots, \frac{1}{\|\mathbf{u}_k\|} \mathbf{u}_k \right\}$$

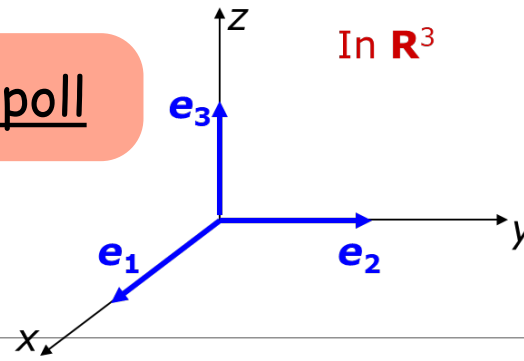
S is an orthonormal set

$$\mathbf{u} \xrightarrow{\text{normalizing}} \frac{1}{\|\mathbf{u}\|} \mathbf{u}$$

any non-zero vector

unit vector

Orthogonal basis



$\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$

- basis for \mathbf{R}^3
- orthogonal basis
- orthonormal basis

$\{(1,0,0), (1,1,0), (1,1,1)\}$

- basis for \mathbf{R}^3
- not orthogonal basis
- not orthonormal basis

$\{(2,0,0), (0,1,1), (0,1,-1)\}$

- basis for \mathbf{R}^3
- orthogonal basis
- not orthonormal basis

$\{(1,0,1), (0,1,0), (1,1,1)\}$

- not basis for \mathbf{R}^3
- not orthogonal basis
- not orthonormal basis

Gram-Schmidt Process Visualization tool

$\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$: a **basis** for a vector space V .

Define $\mathbf{v}_1 = \mathbf{u}_1$

$$\mathbf{v}_2 = \mathbf{u}_2 - \frac{\mathbf{u}_2 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 \quad \text{orthogonal to } \mathbf{v}_1$$

$$\mathbf{v}_3 = \mathbf{u}_3 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 \quad \text{orthogonal to } \mathbf{v}_1 \text{ and } \mathbf{v}_2$$

\vdots

$$\mathbf{v}_k = \mathbf{u}_k - \frac{\mathbf{u}_k \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\mathbf{u}_k \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 - \dots - \frac{\mathbf{u}_k \cdot \mathbf{v}_{k-1}}{\|\mathbf{v}_{k-1}\|^2} \mathbf{v}_{k-1} \quad \text{orthogonal to } \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{k-1}$$

$\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is an **orthogonal basis** for V .

Coordinate vector w.r.t. orthogonal basis

$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$: an orthogonal basis for V

For any vector \mathbf{w} in V ,

$$\mathbf{w} = c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + \dots + c_k \mathbf{u}_k$$

$$\frac{\mathbf{w} \cdot \mathbf{u}_1}{\|\mathbf{u}_1\|^2}$$

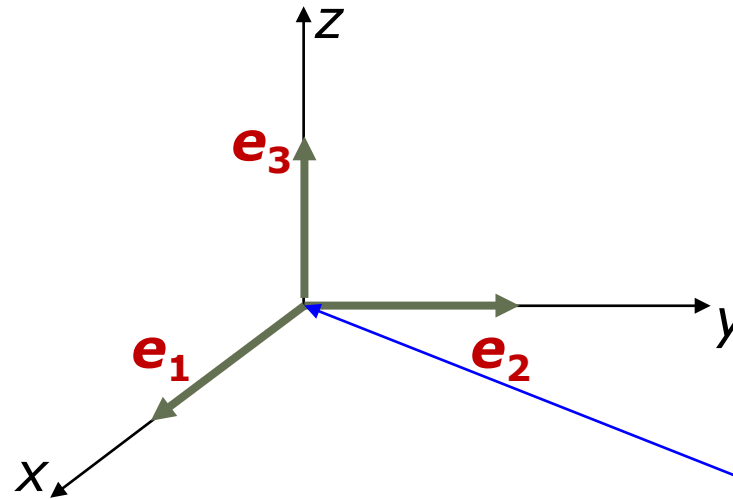
$$\frac{\mathbf{w} \cdot \mathbf{u}_2}{\|\mathbf{u}_2\|^2}$$

$$\frac{\mathbf{w} \cdot \mathbf{u}_k}{\|\mathbf{u}_k\|^2}$$

$$(\mathbf{w})_S = \left(\frac{\mathbf{w} \cdot \mathbf{u}_1}{\|\mathbf{u}_1\|^2}, \frac{\mathbf{w} \cdot \mathbf{u}_2}{\|\mathbf{u}_2\|^2}, \dots, \frac{\mathbf{w} \cdot \mathbf{u}_k}{\|\mathbf{u}_k\|^2} \right)$$

True or false

Can we find an orthogonal set of **four** vectors in \mathbf{R}^3 ?



$$\mathbf{e}_1 = (1, 0, 0) \quad \mathbf{e}_2 = (0, 1, 0) \quad \mathbf{e}_3 = (0, 0, 1) \quad \mathbf{0} = (0, 0, 0)$$

Orthogonal \Rightarrow Linearly independent

Let S be an **orthogonal** set of **nonzero** vectors in a vector space.

Then S is **linearly independent**.

To check whether S is an **orthogonal basis** for V :



Only need to check:

- (i) S is **orthogonal** and
- (ii) $\text{span}(S) = V$

A vector orthogonal to a subspace

To **show** a vector \mathbf{v} is orthogonal to $V = \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$

Show: $\mathbf{v} \cdot \mathbf{u}_1 = 0, \mathbf{v} \cdot \mathbf{u}_2 = 0, \dots, \mathbf{v} \cdot \mathbf{u}_k = 0$

To **find** a vector \mathbf{v} that is orthogonal to $V = \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$

Let $\mathbf{v} = (x_1, x_2, \dots, x_n)$ (unknowns)

Set up: $\mathbf{v} \cdot \mathbf{u}_1 = 0, \mathbf{v} \cdot \mathbf{u}_2 = 0, \dots, \mathbf{v} \cdot \mathbf{u}_k = 0$

Convert into a homogeneous system.

Solve the system.

W^\perp

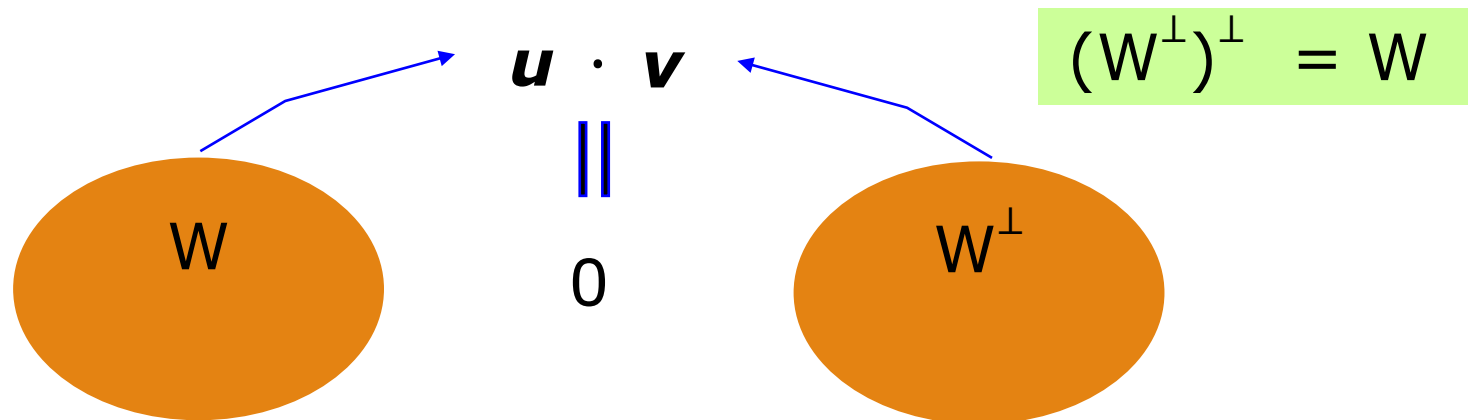
Pronounced
as **W-perp**

Let W be a subspace of \mathbf{R}^n .

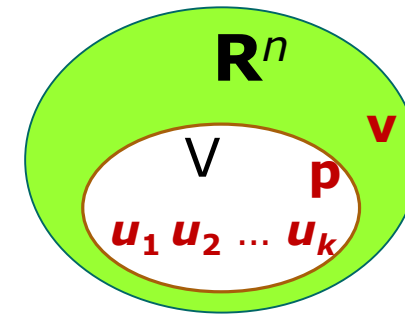
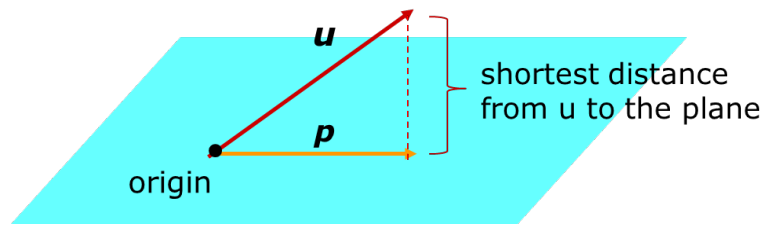
Define $W^\perp = \{ \mathbf{u} \in \mathbf{R}^n \mid \mathbf{u} \text{ is orthogonal to } W \}$

Exercise 5 Q7 W^\perp is also a subspace of \mathbf{R}^n

Every vector in W^\perp is orthogonal to every vector in W .



Projection



The projection p of v onto a subspace V :

p is the vector in V “nearest” to the given vector v

p is the **best approximation** of v in the subspace V

$v - p$ is orthogonal to V

good for checking,
but not finding projection

$S = \{u_1, u_2, \dots, u_k\}$: an orthogonal basis for V

The projection p of v onto V is given by:

$$p = \frac{v \cdot u_1}{||u_1||^2} u_1 + \frac{v \cdot u_2}{||u_2||^2} u_2 + \dots + \frac{v \cdot u_k}{||u_k||^2} u_k$$

Alternative method: use **least squares solution**

Extending orthogonal basis

$$\mathbf{u}_1 = (1, 2, 2, -1), \mathbf{u}_2 = (1, 1, -1, 1), \mathbf{u}_3 = (-1, 1, -1, -1)$$

$S = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ orthogonal set

Extend S to an orthogonal basis for \mathbf{R}^4

Two methods:

$$\mathbf{p} = \frac{\mathbf{v} \cdot \mathbf{u}_1}{\|\mathbf{u}_1\|^2} \mathbf{u}_1 + \frac{\mathbf{v} \cdot \mathbf{u}_2}{\|\mathbf{u}_2\|^2} \mathbf{u}_2 + \cdots + \frac{\mathbf{v} \cdot \mathbf{u}_k}{\|\mathbf{u}_k\|^2} \mathbf{u}_k$$

1. Use row space method:

- find the “missing” row \mathbf{r} in the row echelon form;
- find the projection \mathbf{p} of \mathbf{r} onto $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$
- take the vector $\mathbf{r} - \mathbf{p}$

2. Find a non-zero vector \mathbf{v} orthogonal to $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$

Solve: $\mathbf{u}_1 \cdot \mathbf{v} = 0, \mathbf{u}_2 \cdot \mathbf{v} = 0, \mathbf{u}_3 \cdot \mathbf{v} = 0$

$$\begin{pmatrix} \mathbf{u}_1 \cdot \mathbf{v} \\ \mathbf{u}_2 \cdot \mathbf{v} \\ \mathbf{u}_3 \cdot \mathbf{v} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \mathbf{u}_3 \end{pmatrix} \mathbf{v} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

The projection \mathbf{p} of \mathbf{v} onto a subspace V : $\mathbf{v} - \mathbf{p}$ is orthogonal to V

Exercise 5 Q19 (Hint)

$$\mathbf{A}^2 = \mathbf{A} = \mathbf{A}^T$$

(a) show that $(\mathbf{A}\mathbf{u}) \cdot \mathbf{v} = \mathbf{u} \cdot (\mathbf{A}\mathbf{v})$

- Regard dot product as matrix multiplication $\mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v}$

(b) show that $\mathbf{A}\mathbf{w}$ is the projection of \mathbf{w} onto the subspace

$$V = \{\mathbf{u} \in \mathbf{R}^n \mid \mathbf{A}\mathbf{u} = \mathbf{u}\} \text{ of } \mathbf{R}^n$$

Show: $\mathbf{w} - \mathbf{A}\mathbf{w}$ is orthogonal to V

Let $\mathbf{u} \in V$.



$$\mathbf{A}\mathbf{u} = \mathbf{u} \quad (*)$$

Show $\mathbf{u} \cdot (\mathbf{w} - \mathbf{A}\mathbf{w}) = 0$



Use part (a) and (*)



Least squares solutions

When a linear system $\mathbf{Ax} = \mathbf{b}$ is inconsistent

A least squares solution \mathbf{x}_0 of $\mathbf{Ax} = \mathbf{b}$:

- \mathbf{x}_0 is the best approximation to a solution of $\mathbf{Ax} = \mathbf{b}$

$$\mathbf{Ax}_0 \underset{\text{closest}}{\approx} \mathbf{b} \quad ||\mathbf{b} - \mathbf{Ax}_0|| \underset{\text{smallest}}{}$$


i.e. $||\mathbf{b} - \mathbf{Ax}_0|| \leq ||\mathbf{b} - \mathbf{Av}||$ for all \mathbf{v} in \mathbf{R}^n

To find the least squares solution \mathbf{x}_0 of $\mathbf{Ax} = \mathbf{b}$:

solve the new system $\mathbf{A}^T\mathbf{Ax} = \mathbf{A}^T\mathbf{b}$

Example (Least Squares)

$$\begin{array}{rcl} 2x + y & = & 20 \\ 6x + y & = & 18 \\ 20x + y & = & 10 \\ 30x + y & = & 6 \\ 40x + y & = & 2 \end{array} \quad \begin{pmatrix} 2 & 1 \\ 6 & 1 \\ 20 & 1 \\ 30 & 1 \\ 40 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 20 \\ 18 \\ 10 \\ 6 \\ 2 \end{pmatrix}$$

A  **b**

$Ax = b$ is inconsistent

Find least squares solution for **$Ax = b$**

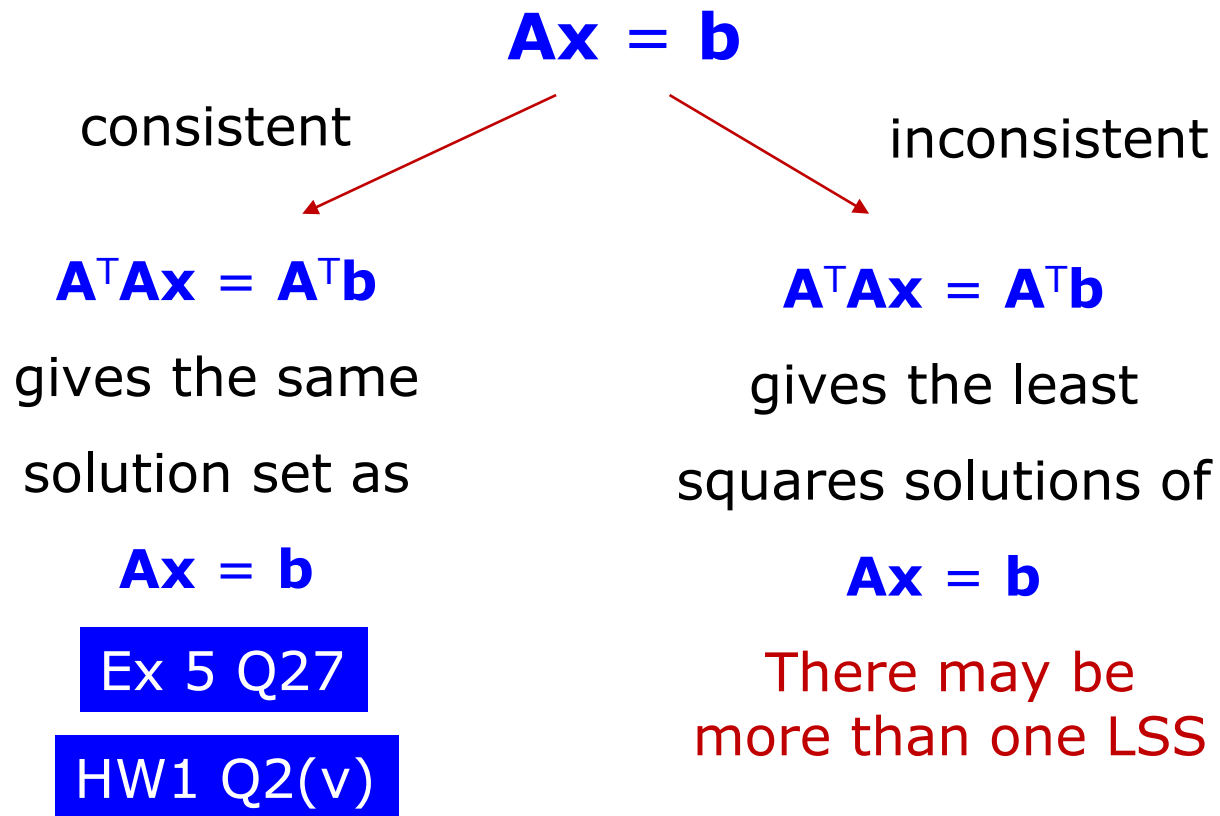
$$\begin{pmatrix} 2940 & 98 \\ 98 & 5 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 608 \\ 56 \end{pmatrix}$$

$A^T A$ **$A^T b$**

Find the actual solutions of **$A^T Ax = A^T b$**

$$\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$$

Always consistent !



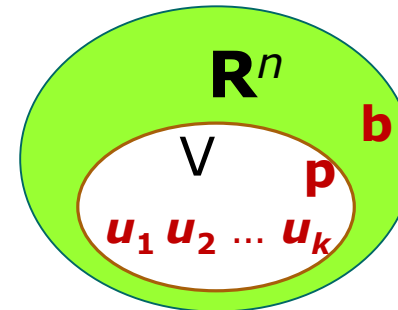
Finding projection using least squares

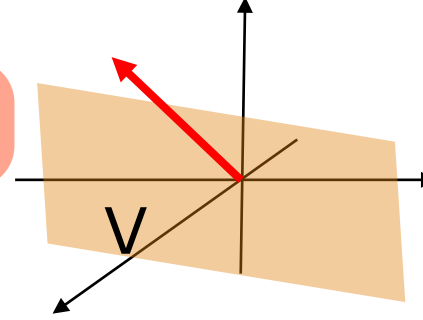
Let \mathbf{x}_0 be a **least squares solution** of $\mathbf{Ax} = \mathbf{b}$

Then $\mathbf{Ax}_0 =$ **projection** of \mathbf{b} onto column space of \mathbf{A}

To find projection \mathbf{p} of \mathbf{b} onto a subspace V :

- Find **any** basis $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$ for V = column space of \mathbf{A}
- Form matrix \mathbf{A} using $\mathbf{u}_1, \dots, \mathbf{u}_k$ as column vectors
- Solve the system $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$ to get \mathbf{x}_0
- $\mathbf{Ax}_0 =$ **projection** \mathbf{p}





Example (Projection)

Find the **projection** of $(1, -1, 1)$ onto the plane $V: x + y + z = 0$ in \mathbf{R}^3

$$V = \text{span}\{(1, -1, 0), (1, 0, -1)\}$$

Form matrix $\mathbf{A} = \begin{pmatrix} 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{pmatrix}$ and column vector $\mathbf{b} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$

Find the least squares solution of $\mathbf{Ax} = \mathbf{b}$

Solve $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$ $\begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix} \longrightarrow \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 4/3 \\ -2/3 \end{pmatrix}$

Projection of $(1, -1, 1)$ on V : $\begin{pmatrix} 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 4/3 \\ -2/3 \end{pmatrix} = \begin{pmatrix} 2/3 \\ -4/3 \\ 2/3 \end{pmatrix}$

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Announcement

❖ Tutorial

- Tutorial set 8 (include Q23, 27)
- Exercise set 5 solution (to be uploaded this weekend)

❖ Homework

- HW2 results published
- HW2 solutions available
- HW3 due 22 October (next Friday)

❖ MATLAB WS5

- OTOT starting this week

❖ Online quiz 9

- Due next Thursday