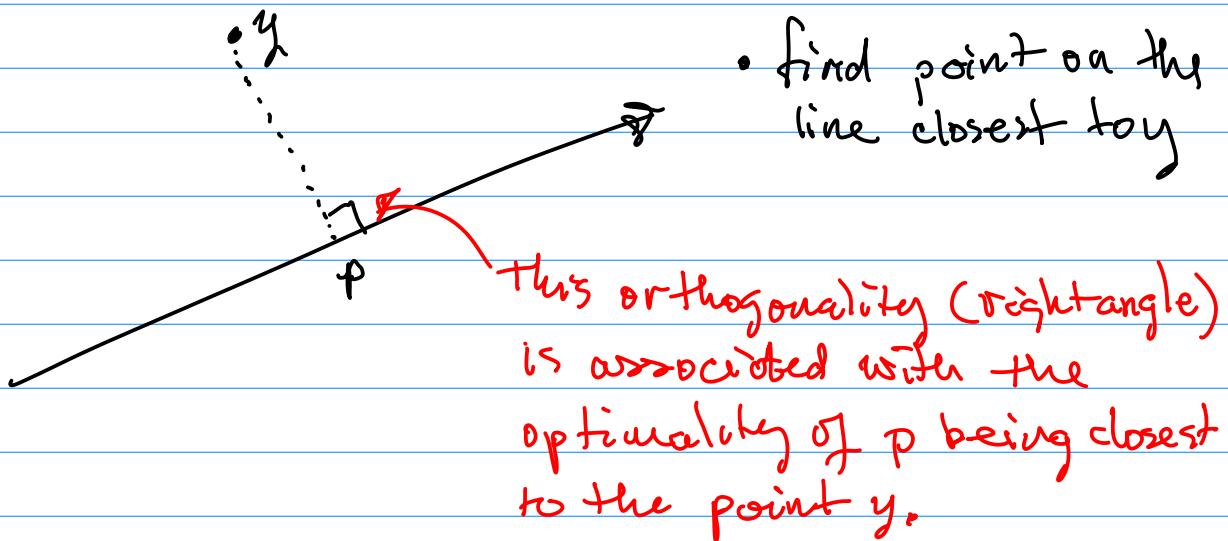


About orthogonal projections...



- find point on the line closest to y

This orthogonality (rightangle)
is associated with the
optimality of p being closest
to the point y .

Orthogonality has something to do with dot products...

Transposes have something to do with dot products...

$$y \cdot Ax = Ax \cdot y = (Ax)^T y = x^T A^T y$$

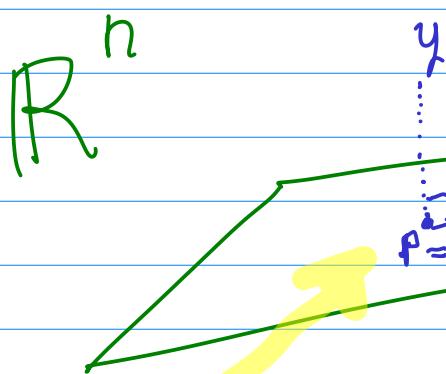
$$= x^T (A^T y) = x \cdot A^T y = A^T y \cdot x$$

This the A jumps over the dot and gets transposed...

$$y \cdot Ax = A^T y \cdot x$$

Goal find the closest point in a plane or hypersurface (actually a subspace) to another point...

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



To find the closest point p in W to y ... need this
this orthogonality.
or This is
 $W = \text{col } A = \text{col } Q$ called the
orthogonal projection
of y onto W .

How:

Let $\{b_1, b_2, \dots, b_q\} \subset \mathbb{R}^n$ be a basis for W .

What does that mean:

- ① The span of the b_i 's is W ...
- ② The b_i 's are linearly independent..

Describe ① and ② using matrices

$$A = \begin{bmatrix} b_1 & b_2 & \dots & b_q \end{bmatrix} \in \mathbb{R}^{n \times q}$$

$$\text{Col } A = \{Ax : x \in \mathbb{R}^q\} = \text{span}\{b_1, b_2, \dots, b_q\} = W$$

Note the b_i 's being linearly independent means

$$\text{Nul } A = \{x : Ax = 0\} = \{0\} \text{ thus } \dim \text{Nul } A = 0$$

Since the columns of A are linearly independent, then

$$A = QR \quad \text{by Gram-Schmidt algorithm.}$$

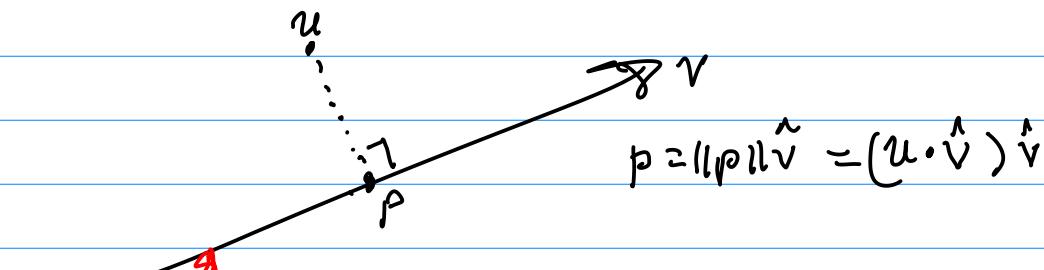
from last week...

Then the distance of the projection from the origin is in the \hat{v} direction

$$\|p\| = u \cdot \hat{v} \approx (2, 3) \cdot \frac{(4, 1)}{\sqrt{17}} = \frac{8+3}{\sqrt{17}} = \frac{11}{\sqrt{17}}$$

Thus

$$p = \|p\| \hat{v} = \frac{11}{\sqrt{17}} \cdot \frac{(4, 1)}{\sqrt{17}} = \frac{11}{17} (4, 1) = \left(\frac{44}{17}, \frac{11}{17} \right)$$



Unit vector means $\|\hat{v}\|=1$ or $\|\hat{v}\|^2=1$

$$\hat{v} \cdot \hat{v} = 1 \quad \text{or} \quad \hat{v}^T \hat{v} = 1$$

Compare this with $Q^T Q = I$

Use this to find p in a higher dimensional space W .

Note that $\text{col } A = \text{col } Q$. Why?

Correct
but there
is a
more general
relationship

The Gram-Schmidt process uses column operations on the columns of A to obtain the columns of Q . Since the column operations are invertible, the columns of A can also be obtained from the columns of Q . Therefore $\text{col } A = \text{col } Q$.

$$A = QR$$

$n \times q$ $n \times q$ $q \times q$

$$\text{col } A = \{ Ax : x \in \mathbb{R}^q \} = \{ QRx : x \in \mathbb{R}^q \}$$

$$y = Rx \in \mathbb{R}^q$$

$$\{Rx : x \in \mathbb{R}^q\} \subseteq \mathbb{R}^q$$

$$\text{Col } A \subseteq \{Qy : y \in \mathbb{R}^q\} = \text{Col } Q$$

more general)

$$\text{No matter what factorization } A = BC$$

$$\text{Col } A \subseteq \text{Col } B$$

$$\text{Know } R \text{ is invertible so } A = QR \text{ then}$$

$$Q = AR^{-1}$$

this is a factorization of Q

$$\text{Col } Q \subseteq \text{Col } A$$

Since $\text{Col } A \subseteq \text{Col } Q$ and $\text{Col } Q \subseteq \text{Col } A$ then it must be that $\text{Col } A = \text{Col } Q \dots$

What is p ?

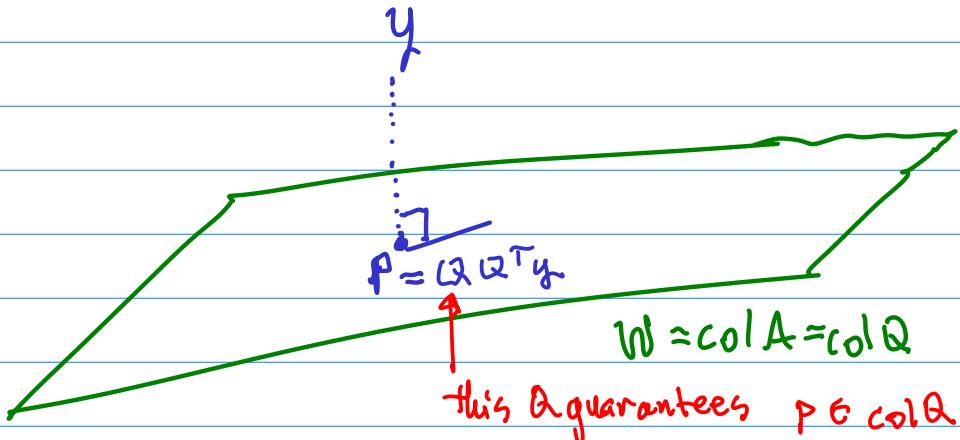
$$p = \|p\| \hat{v} = (\hat{u} \cdot \hat{v}) \hat{v} = \hat{v} (\hat{v}^T \hat{u})$$

1-dimensional case ...

m -dimensional case ...

$$p = Q(Q^T y)$$

check this ...



- To check $p = QQ^T y$ I need to show the vector from y to p is perpendicular to any vector $w \in W$.

$w \in \text{col } Q$ so $w = Qz$ for some $z \in \mathbb{R}^q$

To be perpendicular the dot product must be zero

$$(p-y) \cdot w = p \cdot w - y \cdot w = QQ^T y \cdot w - y \cdot w$$

$$= \cancel{QQ^T y} \cdot \cancel{Qz} - \cancel{y} \cdot \cancel{Qz}$$

try to get rid of some of the Q 's here
using the fact that $Q^T Q = I$

$$= \cancel{Q^T Q} \cancel{Q^T y} \cdot z - Q^T y \cdot z$$

$$= Q^T y \cdot z - Q^T y \cdot z = 0$$

recall fact about transposes...

$$y \cdot Ax = A^T y \cdot x$$

In summary to find the orthogonal projection of y onto W or the closest point in W to y ...

$$\checkmark \text{Let } p = QQ^T y$$

where $A = QR$ is the QR factorization of A and
 $A = [b_1 | b_2 | \dots | b_q]$ with $\{b_1, b_2, \dots, b_q\}$ a basis of W .

Back to solving $Ax = b$, ... but suppose I can't because it doesn't have a solution, ... Next best thing minimize $\|Ax - b\|$

↑
closest point in the column space of A

to the point b ...

$$Ax = p = QQ^T b \quad \begin{matrix} \text{since } p \in \text{col } A \text{ then} \\ \text{Ax} = p \text{ has a solution} \\ \text{what we just did...} \end{matrix}$$

Thus

$$QRx = QQ^T b$$

$$Q^T QRx = Q^T QQ^T b$$

$$Rx = Q^T b$$

the solution to this
is the x that
minimizes $\|Ax - b\|$

One more thing, orthogonal complement ...

$$W^\perp = \{y : y \cdot w = 0 \text{ for all } w \in W\}.$$

Theorem: b.1

EM 3

Let A be an $m \times n$ matrix. The orthogonal complement of the row space of A is the null space of A , and the orthogonal complement of the column space of A is the null space of A^T :

$$(\text{Row } A)^\perp = \text{Nul } A \quad \text{and} \quad (\text{Col } A)^\perp = \text{Nul } A^T$$

please read about this for next time...