

**WEEK 3: REVISION** 

**FINAL EXAM** 



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- 1. Four Fundamental Subspaces
- 2. Orthogonal Vectors and Subspaces
- 3. Projections
- 4. Least Squares and Projections onto a Subspace
- 5. Example of Least Squares



## 1. FOUR FUNDAMENTAL SUBSPACES

Suppose **A** is a  $m \times n$  matrix.

$$A = \begin{cases} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{cases}$$
 linearly independent space is  $C(\mathbf{A})$ , a subspace of  $\mathbb{R}^m$ .

Columns for is  $N(\mathbf{A})$ , a subspace of  $\mathbb{R}^n$ .

- The column space is C(A), a subspace of  $\mathbb{R}^m$ .
- The row space is  $C(\underline{\mathbf{A}}^T)$ , a subspace of  $\underline{\mathbb{R}}^n$ .
- The nullspace is N(A), a subspace of  $\mathbb{R}^n$ .
- The left nullspace is  $N(\mathbf{A}^T)$ , a subspace of  $\mathbb{R}^m$ .



C(A)

 $\dim r$ 

column space

all Ax

#### 2. ORTHOGONAL VECTORS AND SUBSPACES

Two real vectors  $\underline{x}$  and yare orthogonal if

$$x^{T} y = 0 \text{ or } y^{T}x = 0$$
  
 $x.y = 0 \text{ or } y.x = 0$ 

 $C(A^{\mathrm{T}})$  $\dim r$ row space all  $A^{\mathrm{T}}y$ 

 $R(A) \bot N(A) \checkmark$  $C(A) \perp N(A^T)$ 

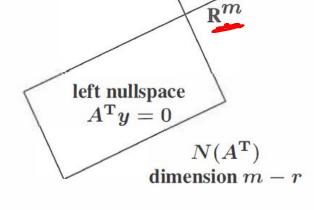
The big picture

orthogonal Complement nullspace Ax = 0N(A)

Rank - number of independent columns - rows

rank(A) + nullity(A) = n

dimension n-r

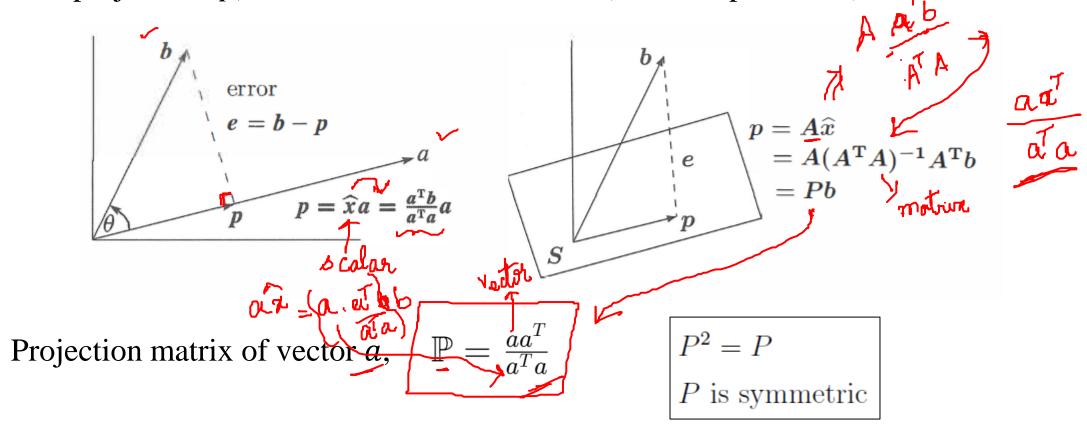


$$\dim(C(A^T)) + \dim(N(A^T)) = m$$



## 3. PROJECTIONS

The projection (*p*) of *b* onto a line and onto S (column space of A).





#### 4. LEAST SQUARES APPROXIMATIONS

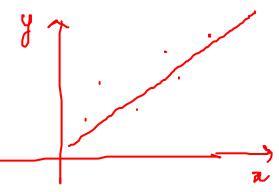
y = mal+C

- $\triangleright$  It often happens that Ax = b has no solution.
- > The usual reason is: *too many equations*.
  - > The matrix A has more rows than columns.
  - $\triangleright$  There are more equations than unknowns (*m* is greater than *n*).
  - $\triangleright$  Then columns span a small part of m-dimensional space.













# SOME SOLVED PROBLEMS



$$span \left[ egin{array}{c} 1 \ 2 \ -1 \end{array} 
ight]$$

Nullspace is N(A)

$$span \begin{bmatrix} -4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

Find the fundamental spaces of 
$$A = \begin{bmatrix} 1 & 4 & 2 \\ 2 & 8 & 4 \\ -1 & -4 & -2 \end{bmatrix}$$
  $\Rightarrow \begin{bmatrix} 4 & 2 \\ 2 & 8 & 4 \\ 0 & 0 & 0 \end{bmatrix}$ 

Row space is 
$$\mathfrak{C}(A^T)$$
,  $\mathfrak{F}_{-4}$   $\mathfrak{$ 

Left nullspace is  $N(\mathbf{A}^T)$ 

$$span \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$



## SOLUTION TO Ax = b

$$A = \begin{bmatrix} 1 & 3 & 0 & 2 \\ 0 & 0 & 1 & 4 \\ 1 & 3 & 1 & 6 \end{bmatrix}$$

Find the condition on  $(b_1, b_2, b_3)$  for Ax = b to be solvable.

$$\begin{bmatrix} A & b \end{bmatrix} = \begin{bmatrix} 1 & 3 & 0 & 2 & b_1 \\ 0 & 0 & 1 & 4 & b_2 \\ 1 & 3 & 1 & 6 & b_3 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 3 & 0 & 2 & b_1 \\ 0 & 0 & 1 & 4 & b_2 \\ 0 & 0 & 0 & 0 & b_3 - b_1 - b_2 \end{bmatrix}$$

Now we get 0 = 0 in the third equation only if  $b_3 - b_1 - b_2 = 0$ .

$$b_1 + b_2 = b_3$$

This condition puts b in the column space of A.



How to find the projection matrix for 
$$a = \begin{bmatrix} -1 \\ 3 \\ -2 \end{bmatrix}$$
?

$$\mathbb{P} = rac{aa^T}{a^Ta}$$

$$aa^{T} = \begin{bmatrix} -1 & 3 & -2 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 3 \\ -2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & -3 & 2 & -1 \\ -3 & 9 & -6 & 3 \\ 2 & -6 & 4 & -2 \\ -1 & 3 & -2 & 1 \end{bmatrix}$$
 Projection of  $b = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix}$  onto  $a$ :

$$\underline{a^Ta} = egin{bmatrix} -1 \ 3 \ -2 \ 1 \end{bmatrix} egin{bmatrix} -1 \ 3 \ -2 \ 1 \end{bmatrix} = 15$$

$$\mathbb{P} = rac{1}{15} egin{bmatrix} 1 & -3 & 2 & -1 \ -3 & 9 & -6 & 3 \ 2 & -6 & 4 & -2 \ -1 & 3 & -2 & 1 \end{bmatrix}$$

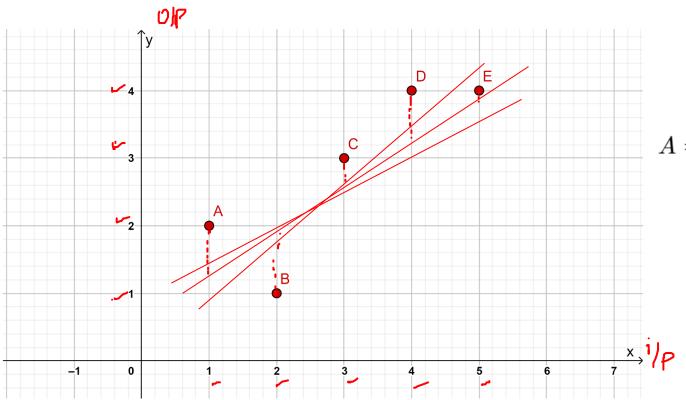
Projection of 
$$b = \begin{bmatrix} -1\\0\\1\\1 \end{bmatrix}$$
 onto  $a$ :

$$p=\mathbb{P}*b$$

$$p = \frac{1}{15} \begin{bmatrix} 1 & -3 & 2 & -1 \\ -3 & 9 & -6 & 3 \\ 2 & -6 & 4 & -2 \\ -1 & 3 & -2 & 1 \end{bmatrix} * \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{15} \begin{bmatrix} 2 \\ -6 \\ 4 \\ -2 \end{bmatrix}$$



#### LEAST SQUARES METHOD



Solve: 
$$\mathbf{A}^T A \widehat{\mathbf{x}} = \mathbf{A}^T \mathbf{b}$$
  $\hat{x} = \begin{bmatrix} \theta' \\ \hat{\theta}'' \end{bmatrix}$ 

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 1 \\ 3 & 1 \\ 4 & 1 \\ 5 & 1 \end{bmatrix} A^{T} = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} b = \begin{bmatrix} 2 \\ 1 \\ 3 \\ 4 \\ 4 \end{bmatrix}$$

$$\begin{bmatrix} 55 & 15 \\ 15 & 5 \end{bmatrix} \begin{bmatrix} \hat{\theta}' \\ \hat{\theta}'' \end{bmatrix} = \begin{bmatrix} 49 \\ 14 \end{bmatrix}$$

Solving this we get, 
$$\hat{x} = \begin{bmatrix} 0.7 \\ 0.7 \end{bmatrix}$$

Best fit line: 
$$y = 0.7x + 0.7$$

Distance of points from line = residuals Least squares regression line: Minimizes sum of square residuals



Best fit line: y = 0.7x + 0.7

