# WEEK 3: REVISION



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### 1. Four Fundamental Subspaces

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

$$A = egin{bmatrix} a_{11} & a_{12} & ... & a_{1n} \ a_{21} & a_{22} & ... & a_{2n} \ dots & dots & \ddots & dots \ a_{m1} & a_{m2} & ... & a_{mn} \end{bmatrix}$$

- The column space is  $C(\mathbf{A})$ , a subspace of  $\mathbb{R}^m$ .
- The row space is  $C(A^T)$ , a subspace of  $\mathbb{R}^n$ . 2.
- The nullspace is  $N(\mathbf{A})$ , a subspace of  $\mathbb{R}^n$ . A  $\mathbb{Z} = \mathbb{O}$ The left nullspace is  $N(\mathbf{A}^T)$ , a subspace of  $\mathbb{R}^m$ . 3.
- 4.

$$A = \begin{bmatrix} 1 & 4 & 2 \\ 2 & 8 & 4 \\ -1 & -4 & -2 \end{bmatrix} R_2 + R_3 + R_4 = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 4 & 2 \\ 2 & 8 & 4 \\ -1 & -4 & -2 \end{bmatrix} R_2 + R_3 + R_4 = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

Row space is 
$$C(A^T)$$
 span  $\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\}$ 

Nullspace is 
$$N(A)$$
 span  $\left\{ \begin{bmatrix} -4 \\ 0 \end{bmatrix} \begin{bmatrix} -2 \\ 0 \end{bmatrix} \right\}$ 

Left nullspace is 
$$N(A^T)$$

$$A^{T} = \begin{cases} 1 & 2 & -1 \\ 4 & 8 & -4 \\ 2 & 4 & -2 \end{cases} R_{2} - 2R_{1} \qquad \begin{array}{c} Pivet \\ 2 & -1 \\ 0 & 0 \end{array} \qquad \begin{array}{c} Pivet \\ 2 & -1 \\ 0 & 0 \end{array}$$

$$\begin{bmatrix} 1 & 4 & 2 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Left nullspace is 
$$N(A^T)$$

$$\begin{bmatrix}
1 & 2 & -1 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{bmatrix}
\begin{bmatrix}
y_1 \\
y_2 \\
y_3
\end{bmatrix} = \begin{bmatrix}
0 \\
0 \\
0
\end{bmatrix}$$

$$y_1 + 2y_2 - y_3 = 0$$

$$y_1 + 2 = 0$$

$$y_1 - 1 = 0$$

$$y_1 = 1$$

### Solution to Ax = b

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

[Ab] 
$$A = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ -2 & -3 \end{bmatrix}$$
 and  $b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$ 

$$\begin{bmatrix} 1 & 2 & b_2 \\ -2 & -3 & b_3 \end{bmatrix} R_2 - R_1$$

$$\begin{bmatrix} 1 & 2 & b_2 \\ -2 & -3 & b_3 \end{bmatrix} R_3 + 2R_1$$

$$\begin{bmatrix} 1 & 1 & b_1 \\ 0 & 1 & b_2 - b_1 \\ 0 & -1 & b_3 + b_1 \end{bmatrix} R_3 + R_2$$

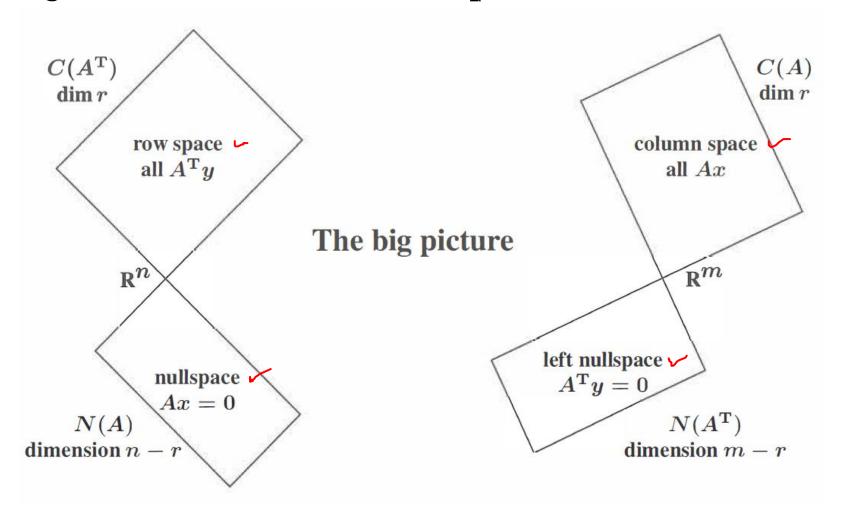
$$\begin{bmatrix} 1 & 1 & b_1 \\ 0 & 1 & b_2 - b_1 \\ 0 & 0 & b_3 + b_2 + b_1 \end{bmatrix} \Rightarrow D$$

$$All yero news \rightarrow b \in C(A)$$

7 E C(A)

### 2. Orthogonal Vectors and Subspaces



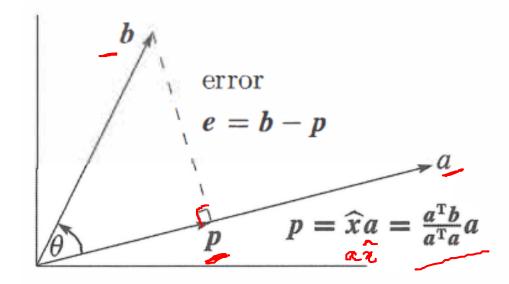


$$rank(A) + nullity(A) = n$$

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

#### 3. Projections

• The projection *p* of *b* onto a line:



Projection matrix of vector a,  $\mathbb{P} = \frac{aa^T}{a^Ta}$ 

$$P^2 = P$$
 $P$  is symmetric

• Projection matrix of 
$$a = \begin{bmatrix} -1 \\ 3 \\ -2 \\ 1 \end{bmatrix}$$
  $\mathbb{P} = \frac{aa^T}{a^Ta}$ 

$$aa^{T} = \begin{bmatrix} -1 \\ 3 \\ -2 \\ 1 \end{bmatrix} \begin{bmatrix} -1 & 3 & -2 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -3 & 2 & -1 \\ -3 & 9 & -6 & 3 \\ 2 & -6 & 4 & -2 \\ -1 & 3 & -2 & 1 \end{bmatrix}$$

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

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

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

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

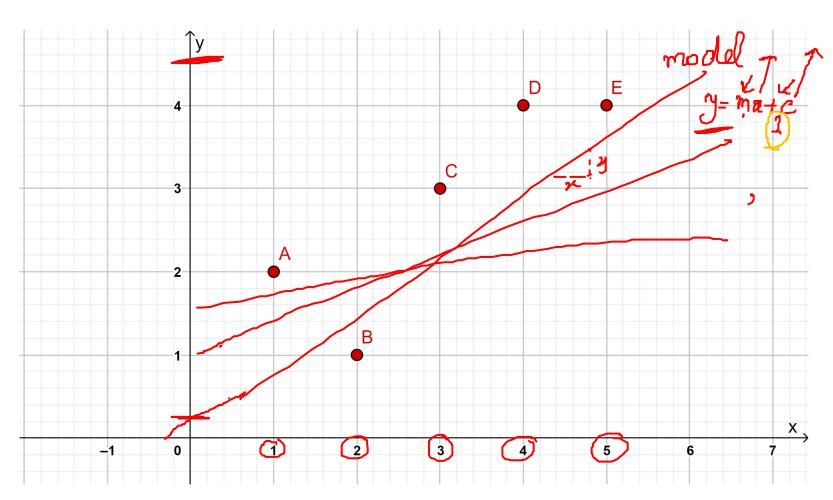
$$p = rac{1}{15} egin{bmatrix} 1 & -3 & 2 & -1 \ -3 & 9 & -6 & 3 \ 2 & -6 & 4 & -2 \ -1 & 3 & -2 & 1 \end{bmatrix} * egin{bmatrix} 1 \ 0 \ 1 \ 1 \end{bmatrix} = rac{1}{15} egin{bmatrix} 2 \ -6 \ 4 \ -2 \end{bmatrix}$$

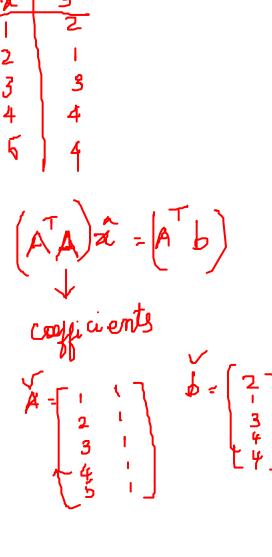
### Least Squares approximations

- It often happens that  $Ax \equiv b$  has no solution.
- The usual reason is: too many equations.
- The matrix A has more rows than columns.
- There are more equations than unknowns ( $\underline{m}$  is greater than  $\underline{n}$ ).
- Then columns span a small part of *m*-dimensional space.
- We cannot always get the error e = b Ax down to zero. When e is zero, x is an exact solution to Ax = b.
- When the length of e is as small as possible,  $\hat{x}$  is a least squares solution.

• Least Squares method: Solving 
$$A^T A \hat{x} = A^T b$$
 we get  $\hat{x} = \int_{-x}^{x} dx$  intercept

## Least Squares approximations





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

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

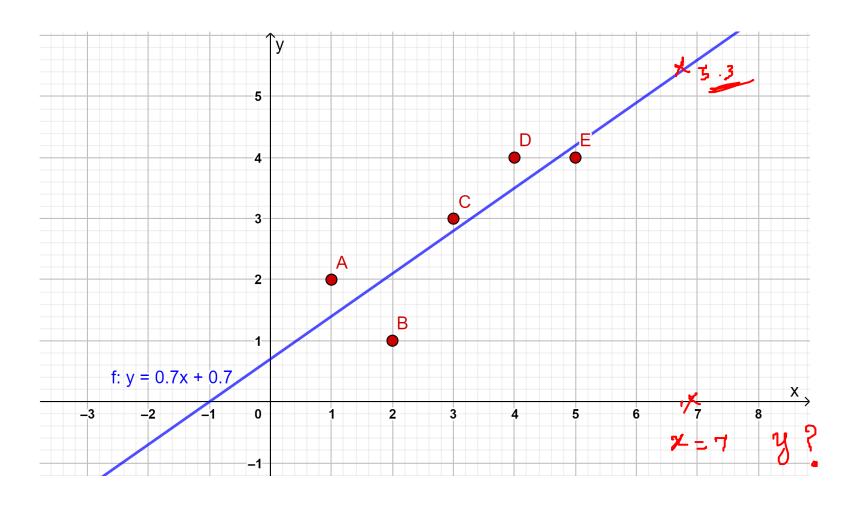
$$A^{T} = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$A = \begin{bmatrix} 55 & 515 \\ 15 & 5 \end{bmatrix} \begin{bmatrix} \hat{\theta}' \\ \hat{\theta}'' \end{bmatrix} = \begin{bmatrix} 49 \\ 14 \end{bmatrix}$$

$$b = \begin{bmatrix} 2 \\ 1 \\ 3 \\ 4 \\ 4 \end{bmatrix}$$
 
$$\hat{x} = \begin{bmatrix} \hat{\theta}' \\ \hat{\theta}'' \end{bmatrix}$$
 Solving this we get, 
$$\hat{x} = \begin{bmatrix} 0.7 \\ 0.7 \end{bmatrix}$$
 into the continuous points of  $\hat{x}$  into the continuous points of  $\hat{x}$  in the continuou

Best fit line: y = 0.7x + 0.7

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



# THANK YOU