

Least squares.

3.8

1/3

consider  $Ax = b$  where  $A$  is  $(m \times n)$ .

If  $x \in \mathbb{R}^n$  then  $r = Ax - b$  is the residual vector &  $x^*$  that yields the smallest possible  $r$  is the least squares solution to  $Ax = b$ . That is

$$\|Ax^* - b\| \leq \|Ax - b\| \text{ for all } x \in \mathbb{R}^n$$

Tomorrow we will derive the formula ... today we will use it... The least squares solution to  $Ax = b$  is  $A^T A x = A^T b$ .

ex1:  $A = \begin{bmatrix} 1 & 2 & 4 \\ -2 & -3 & -7 \\ 1 & 3 & 5 \end{bmatrix}$  &  $b = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$

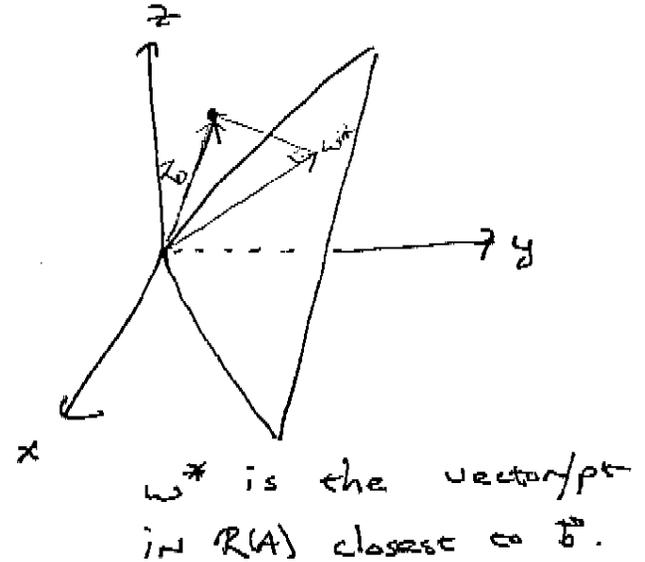
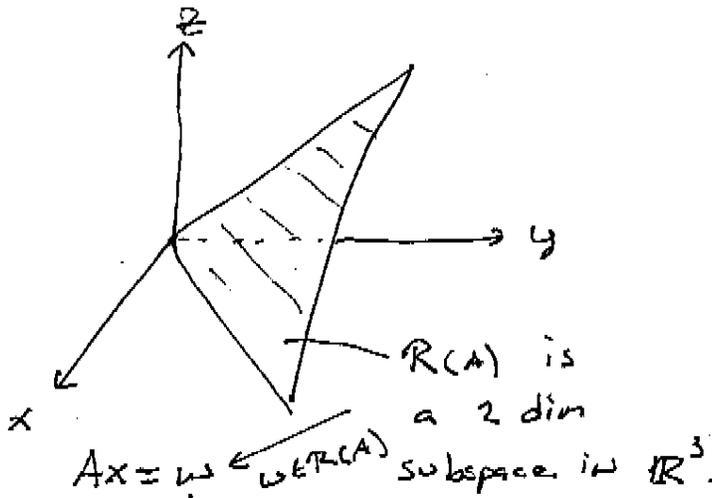
ex2: linear fit to:

$x$	-2	0	1	2
$y$	2	1	0	0

ex3: quadratic fit to:

$t$	0	1	2	3
$y$	0	0	1	2

To understand the origin of the formula  $A^T A x = A^T b$  for finding least squares sol. consider an inconsistent sys  $(3 \times 2)$  sys.



$$\Rightarrow \|w^* - b\| \leq \|w - b\| \quad \forall w \in R(A).$$

We can see that  $(w^* - b) \perp$  to every vector in  $R(A)$ . In particular, it is  $\perp$  to the cols of  $A$  which are a spanning set for  $R(A)$ .

$$\Rightarrow A_1^T (w^* - b) = 0$$

$$\Rightarrow A_2^T (w^* - b) = 0$$

$$\Rightarrow A^T (w^* - b) = 0$$

$$\text{But } w^* \in R(A) \Rightarrow \exists x^* \in \mathbb{R}^2 \text{ s.t. } Ax^* = w^*.$$

$$\Rightarrow A^T (Ax^* - b) = 0$$

$$\Rightarrow A^T A x^* = A^T b.$$

3.8
3/3

Thm: Consider the  $(m \times n)$  sys.  $Ax = b$ .

- (1) The associated sys.  $A^T A x = A^T b$  is always consistent.
- (2) The least squares sol. of  $Ax = b$  are precisely the sol. of  $A^T A x = A^T b$ .
- (3) The least-squares sol. is unique iff  $A$  has rank  $n$ .

Defn. If the rank of  $A$  is less than  $n$ , then we say  $A$  is rank deficient.

Concern w/ round-off.

ex 4:  $A = \begin{bmatrix} 120 & 148 & 175 & 204 & 232 & 260 & 288 & 316 & 345 & 375 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$

$$b = [3, 4, 5, 5.5, 6, 7.5, 8.8, 10, 11.1, 12]^T$$

check out  $A^T A$  &  $A^T b$ ....