

## Least-Squares Problems

Q: If  $A\underline{x} = \underline{b}$  does not admit a solution what is the best approximate solution?

Definition  $A$  -  $m \times n$  matrix and  $\underline{b}$  in  $\mathbb{R}^m$ , a least-squares solution to  $A\underline{x} = \underline{b}$  is an  $\hat{\underline{x}}$  in  $\mathbb{R}^n$  such that

$$\|\underline{b} - A\hat{\underline{x}}\| \leq \|\underline{b} - A\underline{x}\| \quad \text{for all } \underline{x} \text{ in } \mathbb{R}^n.$$

↑  
called least-squares error

### Remark

If  $A\underline{x} = \underline{b}$  is consistent then we can choose  $\hat{\underline{x}}$  any solution, i.e.  $A\hat{\underline{x}} = \underline{b}$ . If not then  $\hat{\underline{x}}$  is the "best approximate solution" to  $A\underline{x} = \underline{b}$ .

Recall:

1/  $W \subset \mathbb{R}^m$  a subspace,  $\underline{b}$  in  $\mathbb{R}^m$  and  $\hat{\underline{b}}$  in  $W$ , then

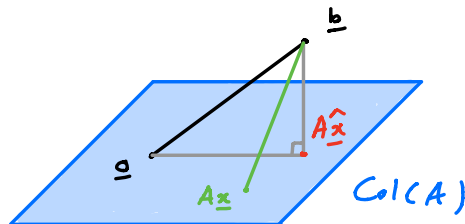
$$\|\underline{b} - \hat{\underline{b}}\| \leq \|\underline{b} - \underline{w}\| \quad \text{for all } \underline{w} \text{ in } W \Leftrightarrow \hat{\underline{b}} = \text{Proj}_W(\underline{b}) \Leftrightarrow \underline{b} - \hat{\underline{b}} \text{ in } W^\perp$$

2/  $\text{Col}(A) = \{A\underline{x} \text{ such that } \underline{x} \text{ in } \mathbb{R}^n\}$

Hence

$$\|\underline{b} - A\hat{\underline{x}}\| \leq \|\underline{b} - A\underline{x}\| \quad \text{for all } \underline{x} \text{ in } \mathbb{R}^n \Leftrightarrow A\hat{\underline{x}} = \text{Proj}_{\text{Col}(A)}(\underline{b})$$
$$\Leftrightarrow \underline{b} - A\hat{\underline{x}} \text{ in } (\text{Col}(A))^\perp$$

Picture in  $\mathbb{R}^3$ :



$$A = (\underline{a}_1, \dots, \underline{a}_n) \Rightarrow \text{Col}(A) = \text{Span}(\underline{a}_1, \dots, \underline{a}_n)$$

$$\underline{b} - A\hat{x} \text{ in } (\text{Col}(A))^\perp \Leftrightarrow \underline{a}_i \cdot (\underline{b} - A\hat{x}) = 0 \text{ for all } i$$

$$(A^T = \begin{pmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_n^T \end{pmatrix})$$

$$\Leftrightarrow \underline{a}_i^T (\underline{b} - A\hat{x}) = 0 \text{ for all } i$$

$$\Leftrightarrow A^T (\underline{b} - A\hat{x}) = \underline{0}$$

$$\Leftrightarrow A^T A \hat{x} = A^T \underline{b}$$

Remark This also follows from the fact that  $(\text{Col}(A))^\perp = \text{Nul}(A^T)$

Conclusion

$$\hat{x} \text{ a least-squares solution to } A\underline{x} = \underline{b} \Leftrightarrow A\hat{x} = \text{Proj}_{\text{Col}(A)}(\underline{b})$$

$$\Leftrightarrow \underline{b} - A\hat{x} \text{ in } (\text{Col}(A))^\perp$$

$$\Leftrightarrow \underline{b} - A\hat{x} \text{ in } \text{Nul}(A^T)$$

$$\Leftrightarrow \hat{x} \text{ a solution to}$$

$$A^T A \underline{x} = A^T \underline{b}.$$

*Called the normal equations for  $A\underline{x} = \underline{b}$*

Example

$$A = \begin{pmatrix} -1 & 2 \\ 2 & -3 \\ -1 & 3 \end{pmatrix}, \quad \underline{b} = \begin{pmatrix} 4 \\ 1 \\ 2 \end{pmatrix}, \quad \hat{x} = ?$$

$$\Rightarrow A^T A = \begin{pmatrix} -1 & 2 & -1 \\ 2 & -3 & 3 \end{pmatrix} \begin{pmatrix} -1 & 2 \\ 2 & -3 \\ -1 & 3 \end{pmatrix} = \begin{pmatrix} 6 & -11 \\ -11 & 22 \end{pmatrix}$$

$$A^T \underline{b} = \begin{pmatrix} -1 & 2 & -1 \\ 2 & -3 & 3 \end{pmatrix} \begin{pmatrix} 4 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} -4 \\ 11 \end{pmatrix}$$

$$\left( \begin{array}{cc|c} 6 & -11 & -4 \\ -11 & 22 & 11 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 6 & -11 & -4 \\ 1 & -2 & -1 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & -2 & -1 \\ 6 & -11 & -4 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & -2 & -1 \\ 0 & 1 & 2 \end{array} \right)$$

$$\hat{x} = \begin{pmatrix} 3 \\ 2 \end{pmatrix} \Leftrightarrow \left( \begin{array}{cc|c} 1 & 0 & 3 \\ 0 & 1 & 2 \end{array} \right)$$

least squares solution  
to  $A\underline{x} = \underline{b}$

## Remark

$A\underline{x} = \underline{b}$  has a unique least squares solution

$\Leftrightarrow A\underline{x} = \hat{\underline{b}}$  has a unique solution

$\Leftrightarrow$  Columns of  $A$  are L.I.

$\Leftrightarrow A^T A$  invertible

In this case  $\hat{\underline{x}} = (A^T A)^{-1} A^T \underline{b}$

Recall:  $A - m \times n$  with L.I. columns  $\Rightarrow A = QR$  ↓  $m \times n$  with orthonormal columns ←  $n \times n$  upper triangular

$$\Rightarrow A^T = R^T Q^T \Rightarrow A^T A = R^T Q^T Q R = R^T R$$

$$\Rightarrow \hat{\underline{x}} = \overset{A^T A}{\parallel} (R^T R)^{-1} \overset{A^T}{\parallel} (R^T Q^T) \underline{b} = R^{-1} (R^T)^{-1} R^T Q^T \underline{b} = R^{-1} Q^T \underline{b}$$

Conclusion  $A - m \times n$  matrix with L.I. columns,  $\underline{b}$  in  $\mathbb{R}^m$

↓  $m \times n$  with orthonormal columns  
 $A = QR$  ←  $n \times n$  upper triangular  $\Rightarrow$  Unique Least-squares solution is  $\hat{\underline{x}} = R^{-1} Q^T \underline{b}$   
( $\Leftrightarrow R \hat{\underline{x}} = Q^T \underline{b}$ )

## Example

$$A = \begin{pmatrix} 1 & 3 & 5 \\ 1 & 1 & 0 \\ 1 & 1 & 2 \\ 1 & 3 & 3 \end{pmatrix}, \quad \underline{b} = \begin{pmatrix} 3 \\ 5 \\ 7 \\ -3 \end{pmatrix}$$

Gram-Schmidt  $\Rightarrow A = \begin{pmatrix} 1/2 & 1/2 & 1/2 \\ 1/2 & -1/2 & -1/2 \\ 1/2 & -1/2 & 1/2 \\ 1/2 & 1/2 & -1/2 \end{pmatrix} \begin{pmatrix} 2 & 4 & 5 \\ 0 & 2 & 3 \\ 0 & 0 & 2 \end{pmatrix}$

$$\Rightarrow Q^T \underline{b} = \begin{pmatrix} 1/2 & 1/2 & 1/2 & 1/2 \\ 1/2 & -1/2 & -1/2 & 1/2 \\ 1/2 & -1/2 & 1/2 & -1/2 \end{pmatrix} \begin{pmatrix} 3 \\ 5 \\ 7 \\ -3 \end{pmatrix} = \begin{pmatrix} 6 \\ -6 \\ 4 \end{pmatrix}$$

$\hat{x}$  is the unique solution to the linear system

$$\left( \begin{array}{ccc|c} 2 & 4 & 5 & 6 \\ 0 & 2 & 3 & -6 \\ 0 & 0 & 2 & 4 \end{array} \right) \longrightarrow \left( \begin{array}{ccc|c} 1 & 0 & 0 & 10 \\ 0 & 1 & 0 & -6 \\ 0 & 0 & 1 & 2 \end{array} \right) \Rightarrow \hat{x} = \begin{pmatrix} 10 \\ -6 \\ 2 \end{pmatrix}$$