

## Lecture 7 Least Squares Problems

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Introduction to Numerical Methods

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## The Linear Least Squares Problem

- In general,  $Ax = b$  with  $m > n$  has no solution
- Instead, try to minimize the *residual*  $r = b - Ax$
- With the 2-norm we obtain the linear *least squares problem* (LSP):

Given  $A \in \mathbb{C}^{m \times n}$ ,  $m \geq n$ ,  $b \in \mathbb{C}^m$ ,  
find  $x \in \mathbb{C}^n$  such that  $\|b - Ax\|_2$  is minimized

- The minimizer  $x$  is the solution to the *normal equations*

$$A^*Ax = A^*b$$

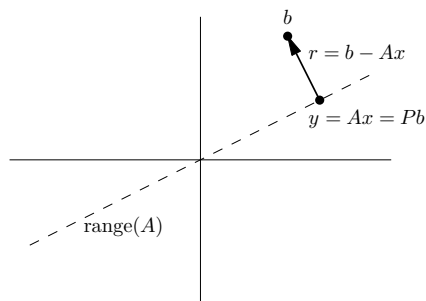
or, in terms of the *pseudoinverse*  $A^+$ :

$$x = A^+b, \quad \text{where } A^+ = (A^*A)^{-1}A^* \in \mathbb{C}^{n,m}$$

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## Geometric Interpretation

- Find the point  $Ax$  in  $\text{range}(A)$  closest to  $b$
- This  $x$  will minimize the 2-norm of  $r = b - Ax$
- $Ax = Pb$  where  $P$  is an orthogonal projector onto  $\text{range}(A)$ , so the residual must be orthogonal to  $\text{range}(A)$



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## Solving the LSP – 1. Normal Equations

- If  $A$  has full rank,  $A^*A$  is square, hermitian positive definite system
- Solve by *Cholesky factorization* (Gaussian elimination)

### Algorithm: Least Squares via Normal Equations

1. Form the matrix  $A^*A$  and the vector  $A^*b$
2. Compute the Cholesky factorization  $A^*A = R^*R$
3. Solve the lower-triangular system  $R^*w = A^*b$  for  $w$
4. Solve the upper-triangular system  $Rx = w$  for  $x$

- Work  $\sim$  Forming  $A^*A$  + Cholesky  $\sim mn^2 + n^3/3$  flops
- Fast, but sensitive to rounding errors

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## Solving the LSP – 2. QR Factorization

- Using  $A = \hat{Q}\hat{R}$ ,  $b$  can be projected onto  $\text{range}(A)$  by  $P = \hat{Q}\hat{Q}^*$
- Insert into  $Ax = b$  to get  $\hat{Q}\hat{R}x = \hat{Q}\hat{Q}^*b$ , or  $\hat{R}x = \hat{Q}^*b$

### Algorithm: Least Squares via QR Factorization

1. Compute the reduced QR factorization  $A = \hat{Q}\hat{R}$
2. Compute the vector  $\hat{Q}^*b$
3. Solve the upper-triangular system  $\hat{R}x = \hat{Q}^*b$  for  $x$

- Work  $\sim$  QR Factorization  $\sim 2mn^2 - 2n^3/3$  flops
- Good stability, relatively fast, used in MATLAB's "backslash" \

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## Solving the LSP – 3. SVD

- Using  $A = \hat{U}\hat{\Sigma}V^*$ ,  $b$  can be projected onto  $\text{range}(A)$  by  $P = \hat{U}\hat{U}^*$
- Insert into  $Ax = b$  to get  $\hat{U}\hat{\Sigma}V^*x = \hat{U}\hat{U}^*b$ , or  $\hat{\Sigma}V^*x = \hat{U}^*b$

### Algorithm: Least Squares via SVD

1. Compute the reduced SVD  $A = \hat{U}\hat{\Sigma}V^*$
2. Compute the vector  $\hat{U}^*b$
3. Solve the diagonal system  $\hat{\Sigma}w = \hat{U}^*b$  for  $w$
4. Set  $x = Vw$

- Work  $\sim$  SVD  $\sim 2mn^2 + 11n^3$  flops
- Very good stability properties, use if  $A$  is close to rank-deficient

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