# Solving Linear Systems in C

Various Matrix Decompositions for Ax = b

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#### A Note on Notation

- Hermitian Conjugate:  $A^*$ ,  $A^{\dagger}$ , or  $A^H$ 
  - $A^*$  is complex case of the transpose:  $A^T$
  - Hermitian Matrix:  $A = A^*$
- Unitary Matrix: Q
  - $-Q^*$  is complex case of an orthogonal matrix  $(A^{-1} = A^T)$
  - I.e.,  $QQ^* = Q^*Q = I$  and  $AA^T = A^TA = I$
- Lower Triangular Matrix: L
- Upper Triangular Matrix: U
- Diagonal Matrix: D

## Matrix Decompositions

Solving systems of linear equations: Ax = b

- QR decomposition: A = QR (need not be square)
- *LU* decompositions:
  - No pivoting: A = LU
  - Partial pivoting: PA = LU
  - Full pivoting: PAQ = LU
  - Unit triangular and diagonal: A = LDU
  - LU reduction (runs LU in parallel)
- Cholesky decomposition:  $A = L^*L$

### Matrix Decompositions

Eigenvalues and related concepts:  $Ax = \lambda x$ 

- Eigendecomposition:  $A = PDP^{-1}$
- Jordan:  $A = PJP^{-1}$
- Schur: *A* = *QUQ*\*
- Singular Value:  $A = USV^*$  (need not be square)

$$A = QR$$

$$Ax = b$$

$$QRx = b$$

$$Q^*QRx = Q^*b$$

$$Rx = Q^*b$$

$$Rx = y$$

#### LU

- Twice as fast as QR
- Pivoting adds stability

$$A = LU$$

$$LUx = b$$

$$Ly = b$$

$$Ux = y$$

### Cholesky

- Must be Hermitian and positive-definite
- I.e., positive eigenvalues
- Twice as fast as LU (where appropriate)

$$A = L^*L$$

$$L^*Lx = b$$

$$L^*y = b$$

$$Lx = y$$