

Matrix Factorizations

$$A = LU = \begin{pmatrix} \text{lower triangular } L \\ \text{1's on the diagonal} \end{pmatrix} \begin{pmatrix} \text{upper triangular } U \\ \text{pivots on the diagonal} \end{pmatrix}$$

Requirements: No row exchanges as Gaussian elimination reduces square A to U

$$A = LDU = \begin{pmatrix} \text{lower triangular } L \\ \text{1's on the diagonal} \end{pmatrix} \begin{pmatrix} \text{pivot matrix } D \\ \text{is diagonal} \end{pmatrix} \begin{pmatrix} \text{upper triangular } U \\ \text{1's on the diagonal} \end{pmatrix}$$

Requirements: No row exchanges. The pivots in D are divided out to leaves 1's on the diagonal of U . If A is symmetric then U is L^T and $A = LDL^T$.

$$PA = LU = (\text{permutation matrix } P \text{ to avoid zeros in the pivot positions.})$$

Requirements: A is invertible. Then P, L, U are invertible.

$$EA = R \text{ (} m \text{ by } m \text{ invertible } E \text{) (any } m \text{ by } n \text{ matrix } A \text{) = (reduced row echelon form } R \text{)}$$

Requirements: None!

$$S = C^T C = (\text{lower triangular}) (\text{upper triangular}) \text{ with } \sqrt{D} \text{ on both diagonals}$$

Requirements: S is symmetric and positive definite.

$$A = QR = (\text{orthonormal columns in } Q) (\text{upper triangular } R)$$

Requirements: A has independent columns. Those are orthogonalized in Q by the Gram-Schmidt process.

$$A = X \Lambda X^{-1} = (\text{eigenvectors in } X) (\text{eigenvalues in } \Lambda) (\text{left eigenvectors in } X^{-1})$$

Requirements: A must have n linearly independent eigenvectors.

$$S = Q \Lambda Q^T = (\text{orthogonal matrix } Q) (\text{real eigenvalue matrix } \Lambda) (Q^T \text{ is } Q^{-1})$$

Requirements: S is real and symmetric: $S^T = S$. This is the **Spectral Theorem**.

$$A = U \Sigma V^T = \begin{pmatrix} \text{orthogonal} \\ U \text{ is } m \times m \end{pmatrix} \begin{pmatrix} m \times n \text{ singular value matrix} \\ \sigma_1, \sigma_2, \dots, \sigma_r \text{ on its diagonal} \end{pmatrix} \begin{pmatrix} \text{orthogonal} \\ V \text{ is } n \times n \end{pmatrix}$$

Requirements: None.

Eigenvalues and Eigenvectors of Special Matrices

Matrix	Eigenvalues	Eigenvectors
Symmetric: $S^T = S = Q \Lambda Q^T$	real eigenvalues	orthogonal $\bar{x}_i^T x_j = 0$
Orthogonal: $Q^T = Q^{-1}$	all $ \lambda = 1$	orthogonal $\bar{x}_i^T x_j = 0$
Skew-symmetric: $A^T = -A$	imaginary λ 's	orthogonal $\bar{x}_i^T x_j = 0$
Positive Definite: $x^T S x > 0$	all $\lambda > 0$	orthogonal $\bar{x}_i^T x_j = 0$
Similar: $A = B C B^{-1}$	$\lambda(A) = \lambda(C)$	B times eigenvector of C
Projection: $P = P^2 = P^T$	$\lambda = 1; 0$	column space; null space
Rank One: uv^T	$\lambda = v^T u; 0, \dots, 0$	u ; whole plane v^\perp
Inverse: A^{-1}	$1/\lambda(A)$	keep eigenvectors of A
Shift: $A + cI$	$\lambda(A) + c$	keep eigenvectors of A
Stable Powers: $A^n \rightarrow 0$	all $ \lambda < 1$	any eigenvectors
Diagonalizable: $A = X \Lambda X^{-1}$	diagonal of Λ	columns of X