

Lecture 26: Review

- 1 Eigenvalues, eigenvectors, eigenspaces, diagonalization
- 2 Inner product, length, orthogonality, orthogonal complement
- 3 Orthogonal projections, the Gram-Schmidt process
- 4 QR factorization, least-squares problems, orthogonal diagonalization of symmetric matrices

Eigenvalues, -vectors, -spaces, diagonalization

Let A be $n \times n$ matrix

- The eigenvalues of A : solutions of the characteristic equation $\det(A - \lambda I) = 0$
- The eigenspace corresponding to the eigenvalue λ_k : the space $\text{Nul}(A - \lambda_k I)$
- An eigenvector corresponding to the eigenvalue λ_k : an element of $\text{Nul}(A - \lambda_k I)$

Diagonalization: $A = PDP^{-1}$, where D is a diagonal matrix

- A is diagonalizable iff it has n linearly independent eigenvectors
- If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are the eigenvectors, and $\lambda_1, \dots, \lambda_n$ are the corresponding eigenvalues

$$A = PDP^{-1} \quad \text{with} \quad P = [\mathbf{v}_1 \dots \mathbf{v}_n], \quad D = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

- distinct eigenvalues, or symmetric matrix \Rightarrow diagonalizable
- dimension of $\text{Nul}(A - \lambda_k I)$ is less than the multiplicity of $\lambda_k \Rightarrow$ not diagonalizable

Inner product, length, orthogonality

$$\mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v}, \quad \|\mathbf{u}\| = \sqrt{\mathbf{u} \cdot \mathbf{u}}, \quad \text{dist}(\mathbf{u}, \mathbf{v}) = \|\mathbf{u} - \mathbf{v}\|, \quad \mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$

- Normalization: $\mathbf{w} = \frac{1}{\|\mathbf{u}\|} \mathbf{u}$
- Orthogonality: $\mathbf{u} \cdot \mathbf{v} = 0$ ($\mathbf{u} \perp \mathbf{v}$)
- Orthogonal complement: $H^\perp = \{\mathbf{z} \in \mathbb{R}^n : \mathbf{z} \perp H\}$
- For any matrix A ,

$$(\text{Row } A)^\perp = \text{Nul } A \quad \text{and} \quad (\text{Col } A)^\perp = \text{Nul } A^T$$

- Let $H = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$, and $P = [\mathbf{v}_1, \dots, \mathbf{v}_p]$. We have $H = \text{Col } P$.

$$H^\perp = (\text{Col } P)^\perp = \text{Nul } P^T$$

Orthogonal projections, Gram-Schmidt process

Let $\mathcal{U} = \{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ be an orthogonal basis for H .

- For $\mathbf{x} \in \mathbb{R}^n$, let

$$\hat{\mathbf{x}} = \alpha_1 \mathbf{u}_1 + \dots + \alpha_p \mathbf{u}_p \quad \text{with} \quad \alpha_k = \frac{\mathbf{u}_k \cdot \mathbf{x}}{\mathbf{u}_k \cdot \mathbf{u}_k} \quad (k = 1, \dots, p)$$

- $\text{Proj}_H \mathbf{x} = \hat{\mathbf{x}}$ is called the orthogonal projection of \mathbf{x} onto H
- $\mathbf{x} - \hat{\mathbf{x}} \perp H$, in other words, $\mathbf{x} - \hat{\mathbf{x}} \in H^\perp$

Gram-Schmidt: $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} \mapsto \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$

$$\mathbf{u}_1 = \mathbf{v}_1$$

$$\mathbf{u}_2 = \mathbf{v}_2 - \frac{(\mathbf{v}_2 \cdot \mathbf{u}_1)}{(\mathbf{u}_1 \cdot \mathbf{u}_1)} \mathbf{u}_1$$

$$\mathbf{u}_3 = \mathbf{v}_3 - \frac{(\mathbf{v}_3 \cdot \mathbf{u}_1)}{(\mathbf{u}_1 \cdot \mathbf{u}_1)} \mathbf{u}_1 - \frac{(\mathbf{v}_3 \cdot \mathbf{u}_2)}{(\mathbf{u}_2 \cdot \mathbf{u}_2)} \mathbf{u}_2$$

QR, least-squares, orthogonal diagonalization

If A has linearly independent columns

- Use Gram-Schmidt and normalization to orthonormalize the columns of $A \rightarrow Q$
- Calculate $R = Q^T A$, then $A = QR$

Least-squares problems

- $A\hat{\mathbf{x}} = \text{Proj}_{\text{Col}A}\mathbf{b} \Leftrightarrow A^T A\hat{\mathbf{x}} = A^T \mathbf{b}$
- The columns of A are linearly independent $\Leftrightarrow A^T A$ is invertible
- $A^T A$ is invertible and $A = QR$, then $A^T A\hat{\mathbf{x}} = A^T \mathbf{b} \Leftrightarrow \hat{\mathbf{x}} = R^{-1} Q^T \mathbf{b}$

Orthogonal diagonalization of symmetric matrices: Let $A = A^T$

- Find the eigenvalues and bases for the eigenspaces
- Orthonormalize the bases
- Collecting all the vectors from these bases into the columns of U , we have $A = UDU^{-1} = UDU^T$, with the diagonal matrix D consisting of the eigenvalues