

Class 15: 10/25

Outline: I. Symmetric Matrices
II. Positive Definite Matrices

Questions from last time? (Systems of linear differential equations).

I. Symmetric matrices ($A^T = A$).

Key facts about symmetric matrices:

1. A symmetric matrix has all real eigenvalues
2. Eigenvectors can be taken to be orthonormal
3. Symmetric matrices can always be diagonalized.

Spectral Theorem: Every symmetric matrix A can be written as

$$A = Q \Lambda Q^{-1} = Q \Lambda Q^T \quad (Q^{-1} = Q^T)$$

Example 1: Last time we had P = projection onto $y=x$ in $\mathbb{R}^2$.
Projection matrices are symmetric.

- We had orthogonal eigenvectors (can be normalized).
- Eigenvalues were 0 and 1 (real numbers).

Example 2: $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$

$$\det(A - \lambda I) = \det \begin{bmatrix} 1-\lambda & 2 \\ 2 & 4-\lambda \end{bmatrix} = (1-\lambda)(4-\lambda) - 4 = -5\lambda + \lambda^2$$

Eigenvalues are $\lambda_1 = 0$, $\lambda_2 = 5$

Eigenvectors: For λ_1 : Solve $\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \vec{x} = \vec{0} \leadsto \vec{x}_1 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$

For λ_2 : Solve $\begin{bmatrix} -4 & 2 \\ 2 & -1 \end{bmatrix} \vec{x} = \vec{0} \leadsto \vec{x}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

(Note: $\begin{bmatrix} -2 \\ 1 \end{bmatrix}$ is a basis for $N(A)$.)

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ is a basis for the row space / column space.

These are orthogonal!).

Eigenvectors aren't orthonormal yet.

How do we make them orthonormal?

Normalize: In this case, divide by $\sqrt{5}$.

So $Q = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix}$ and $\Lambda = \begin{bmatrix} 0 & 0 \\ 0 & 5 \end{bmatrix}$

$$A = Q \Lambda Q^{-1} = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 5 \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$= \frac{1}{5} \begin{bmatrix} 0 & 5 \\ 0 & 10 \end{bmatrix} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix} = \frac{1}{5} \begin{bmatrix} 5 & 10 \\ 10 & 20 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \checkmark$$

Why are eigenvalues real?

$$A \vec{x} = \lambda \vec{x} \implies A \bar{\vec{x}} = \bar{\lambda} \bar{\vec{x}}, \text{ and } \bar{\vec{x}}^T A = \bar{\vec{x}}^T \bar{\lambda}$$

$$\text{So } \bar{\vec{x}}^T A \vec{x} = \bar{\vec{x}}^T \lambda \vec{x} = \bar{\vec{x}}^T \bar{\lambda} \bar{\vec{x}} \quad \bar{\vec{x}}^T \vec{x} = |x_1|^2 + \dots + |x_n|^2 \neq 0$$

$$\text{So } \lambda = \bar{\lambda} \text{ and } \lambda \text{ is real.}$$

Why are eigenvectors orthogonal when they have different eigenvalues?

$$\text{Say } A \vec{x}_1 = \lambda_1 \vec{x}_1, \text{ and } A \vec{x}_2 = \lambda_2 \vec{x}_2.$$

$$\begin{aligned} \vec{x}_1^T A \vec{x}_2 &= \vec{x}_1^T \lambda_2 \vec{x}_2 = \lambda_2 \vec{x}_1^T \vec{x}_2 \\ \vec{x}_1^T A^T \vec{x}_2 &= \vec{x}_1^T \lambda_1 \vec{x}_1 = \lambda_1 \vec{x}_1^T \vec{x}_2 \\ (A \vec{x}_1)^T \vec{x}_2 &= \lambda_1 \vec{x}_1^T \vec{x}_2 \end{aligned}$$

$$\text{Since } \lambda_1 \neq \lambda_2, \vec{x}_1^T \vec{x}_2 = 0 \text{ and } \vec{x}_1 \text{ and } \vec{x}_2 \text{ are orthogonal}$$

Fact: $A = Q \Lambda Q^T$ implies that

$$A = \lambda_1 P_1 + \dots + \lambda_n P_n \quad \text{where } P_i \text{ is projection onto the eigenspace with eigenvalue } \lambda_i$$

$$(\text{i.e. } A = \lambda_1 \vec{x}_1 \vec{x}_1^T + \lambda_2 \vec{x}_2 \vec{x}_2^T + \dots)$$

Fact: If $A \rightarrow U$ requires no permutations, then the signs of the pivots are the same as the signs of the eigenvalues.

Example: $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$ Pivots? $U = \begin{bmatrix} 1 & 3 \\ 0 & -8 \end{bmatrix}$ 1, -8

$$\text{Eigenvalues? } A - \lambda I = \begin{bmatrix} 1-\lambda & 3 \\ 3 & 1-\lambda \end{bmatrix}$$

$$\text{characteristic poly} = (1-\lambda)^2 - 9 = \lambda^2 - 2\lambda - 8$$

$$\lambda_1 = 4, \lambda_2 = -2$$

If A is not symmetric, the signs might not match.

$$A = \begin{bmatrix} 1 & 6 \\ -1 & 4 \end{bmatrix} \quad \begin{array}{l} \text{Eigenvalues } -1, -2 \\ \text{Pivots: } 1, 2. \end{array}$$

II. Positive Definite Matrices:

Definition: A symmetric matrix with all positive eigenvalues is called positive definite.

For 2×2 : When is $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ positive definite?

Need λ_1 and $\lambda_2 > 0$, so $\lambda_1, \lambda_2 = \text{determinant} > 0$
and $\lambda_1 + \lambda_2 = \text{trace} > 0$.

So a symmetric 2×2 matrix is positive definite when
 $ac - b^2 > 0$
 $a + c > 0$.

But if a or $c < 0$, $ac - b^2 < 0$, so really need
 $a > 0$ and $c > 0$.

2×2 test: $A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$ is positive definite when
 $ac - b^2 > 0$
 $a > 0$.

Alternate definition: A symmetric matrix A is positive definite if
 $\vec{x}^T A \vec{x} > 0$ for every $\vec{x} \neq \vec{0}$.

$$\begin{aligned} \text{Need } [x, y] \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} &= [x, y] \begin{bmatrix} ax + by \\ bx + cy \end{bmatrix} \\ &= ax^2 + 2bxy + cy^2 > 0. \end{aligned}$$

Fact: 1) If A and B are positive definite, so is $A+B$.
Use alternate def.

$$\begin{aligned} \vec{x}^T (A+B) \vec{x} &= \vec{x}^T A \vec{x} + \vec{x}^T B \vec{x} > 0 \\ \text{and } (A+B)^T &= A^T + B^T = A+B \text{ so symmetric.} \end{aligned}$$

2) If $A = R^T R$, A is symmetric. If R has independent columns, $A = R^T R$ is positive definite.

$$\begin{aligned} \vec{x}^T A \vec{x} &= \vec{x}^T R^T R \vec{x} = (R\vec{x})^T R\vec{x} = \|R\vec{x}\|^2 > 0 \quad \text{if} \\ &\quad R \text{ has indep. columns} \end{aligned}$$

Equivalent properties for symmetric matrices

1. Pivots are all positive and no permutations were necessary to get to U .
2. All n upper left determinants are positive
3. All n eigenvalues are positive
4. $\vec{x}^T A \vec{x} > 0$ except for $\vec{x} = 0$
5. $A = R^T R$ for a matrix R with independent columns.

(If a symmetric matrix has any one of these properties, it is positive definite).

Example: $A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$

Test 2: determinants are 2, 3, 4.

Test 3: eigenvalues? $2, 2+\sqrt{2}, 2-\sqrt{2}$

Test 1: pivots: $2, \frac{3}{2}, \frac{4}{3}$.

$$A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & -1 \end{bmatrix} \quad (\text{Test 5})$$

$$\begin{aligned} \text{Test 4: } \vec{x}^T A \vec{x} &= 2(x_1^2 - x_1 x_2 + x_2^2 - x_2 x_3 + x_3^2) \\ &= 2(x_1 - \frac{1}{2}x_2)^2 + \frac{3}{2}(x_2 - \frac{2}{3}x_3)^2 + \frac{4}{3}x_3^2 > 0 \end{aligned}$$