

# Positive Definite Matrices

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# Recap and agenda for today's class

- Discussed eigenvalues and eigenvectors for a symmetric matrix  $A$
- Discuss the concept of positive definite matrices
  - Chapter 6.5 of the book

# Positive Definite Matrices

- Positive definite matrices - **symmetric matrices with positive eigenvalues  $\lambda$**
- How to recognize positive definite matrices
  - You may say, just find eigenvalues and test whether they are  $> 0$
  - This is what we want to avoid - calculating eigenvalues is a work
- When eigenvalues are needed, we can calculate them
- But if we just want to know whether they are +ve, there are faster ways
- Here are two goals of today's class
  - To find quick tests on a symmetric matrix that guarantee positive eigenvalues
  - To explain important applications of positive definiteness
- Every eigenvalue is real because the matrix is symmetric

# Positive Definite Matrices – Test-1

- When is  $2 \times 2$  matrix  $S = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$  have  $\lambda_1 > 0$  and  $\lambda_2 > 0$
- Test1: Eigenvalues of  $S$  are positive if and only if

$$a > 0 \text{ and } ac - b^2 > 0$$

- Proof: Their product  $\lambda_1 \lambda_2$  is equal to the determinant i.e.,  $ac - b^2$ 
  - If  $\lambda_1 \lambda_2 > 0$  then  $ac - b^2 > 0$
- Their sum  $\lambda_1 + \lambda_2$  is equal to the trace i.e.,  $a + c$ 
  - If  $\lambda_1 + \lambda_2 > 0$  then  $a + c > 0$
  - Then  $a$  and  $c$  are both positive, (if  $a$  or  $c$  is not positive,  $ac - b^2 > 0$  will fail)
  - If  $a > 0$  and  $ac - b^2 > 0$  then it automatically implies  $c > 0$
- This test uses the 1 by 1 determinant  $a$  and the 2 by 2 determinant  $ac - b^2$
- When  $S$  is 3 by 3, we need to also check  $\det S > 0$

# Positive Definite Matrices – Test-2

- When is  $2 \times 2$  matrix  $S = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$  have  $\lambda_1 > 0$  and  $\lambda_2 > 0$
- Test2: Eigenvalues of  $S$  are positive if and only if **the pivots are positive**

$$a > 0 \text{ and } \frac{ac - b^2}{a} > 0$$

- $a > 0$  is required in both tests
  - So  $ac > b^2$  is also required, for the determinant test and now the pivot test
- Point is to recognize that ratio as the second pivot of  $S$

$$\begin{array}{ccc} \begin{bmatrix} a & b \\ b & c \end{bmatrix} & \begin{array}{c} \text{The first pivot is } a \\ \xrightarrow{\quad\quad\quad} \\ \text{The multiplier is } b/a \end{array} & \begin{bmatrix} a & b \\ 0 & c - \frac{b}{a}b \end{bmatrix} \end{array} \quad \begin{array}{c} \text{The second pivot is} \\ c - \frac{b^2}{a} = \frac{ac - b^2}{a} \end{array}$$

# Positive Definite Matrices – Test-2

- This connects two big parts of linear algebra
  - Positive eigenvalues mean positive pivots and vice versa
- Each pivot is a ratio of upper left determinants
  - Pivots give a quick test for  $A > 0$
  - Pivots are a lot faster to compute than the eigenvalues
- Satisfying to see pivots and determinants and eigenvalues come together

# Positive Definite Matrices – Test 3

- We have  $S\mathbf{x} = \lambda\mathbf{x}$ . We now perform inner product with  $\mathbf{x}$

$$\mathbf{x}^T S \mathbf{x} = \lambda \mathbf{x}^T \mathbf{x} \Rightarrow \mathbf{x}^T S \mathbf{x} = \lambda \|\mathbf{x}\|^2$$

- Since  $\lambda > 0$  for a positive definite matrix. This implies that  $\mathbf{x}^T S \mathbf{x} > 0$
- In many applications this number is the energy of the system
- Idea is that  $\mathbf{x}^T S \mathbf{x}$  is positive for all nonzero vectors  $\mathbf{x}$ , **not just eigenvectors**
- If  $S$  and  $T$  are positive definite, so is  $S + T$ 
  - Reason:  $\mathbf{x}^T (S + T) \mathbf{x} = \mathbf{x}^T S \mathbf{x} + \mathbf{x}^T T \mathbf{x}$
  - Since, both  $\mathbf{x}^T S \mathbf{x} > 0$  and  $\mathbf{x}^T T \mathbf{x} > 0$ , their sum is  $> 0$
- $\mathbf{x}^T S \mathbf{x}$  also connects with our final way to recognize a positive definite matrix

# Positive Definite Matrices – Test 4

- Consider matrix  $A$ , possibly rectangular
  - We know that  $S = A^T A$  is square and symmetric
- More than that,  $S$  will be positive definite when  $A$  has independent columns:
- Test: If the columns of  $A$  are independent, then  $S = A^T A$  is positive definite

$$\mathbf{x}^T A^T A \mathbf{x} = (A\mathbf{x})^T A\mathbf{x} = \|A\mathbf{x}\|^2$$

- If  $A$  has independent columns, vector  $A\mathbf{x}$  is not zero when  $\mathbf{x} \neq 0$
- Then  $\mathbf{x}^T S \mathbf{x}$  is the positive number  $\|A\mathbf{x}\|^2$  and the matrix  $S$  is positive definite



# Positive Definite Matrices – Summary

- When a symmetric matrix  $S$  has one of these five properties, it has them all:
  - All  $n$  eigenvalues of  $S$  are positive
  - All  $n$  upper left determinants are positive
  - All  $n$  pivots of  $S$  are positive
  - $\mathbf{x}^T S \mathbf{x}$  is positive except at  $\mathbf{x} = 0$ . This is the energy-based definition
  - $S$  equals  $A^T A$  for a matrix  $A$  with independent columns
- Cholesky factorization of positive definite matrix

$$S = Q \Lambda Q^T = Q \sqrt{\Lambda} \sqrt{\Lambda}^T Q^T = Q \sqrt{\Lambda} (Q \sqrt{\Lambda})^T = A_1 A_1^T$$

- Square-root of a positive definite matrix

$$\begin{aligned}\sqrt{S} &= Q \sqrt{\Lambda} Q^T \\ \sqrt{S} \sqrt{S} &= Q \sqrt{\Lambda} Q^T Q \sqrt{\Lambda} Q^T = Q \Lambda Q^T = S\end{aligned}$$

# Positive Semidefinite Matrices

- All eigenvalues of  $S$  are  $\geq 0$
- $\mathbf{x}^T S \mathbf{x} \geq 0$