

# Analytical Geometry and Linear Algebra II, Lab 10

Symmetric matrices

Positive definite matrices and minima



## How I spent last weekend







- 1 A symmetric matrix S has n real eigenvalues  $\lambda_i$  and n orthonormal eigenvectors  $q_1, \dots, q_n$ .
- **2** Every real symmetric S can be diagonalized:  $S = Q\Lambda Q^{-1} = Q\Lambda Q^{T}$
- $\bf 3$  The number of positive eigenvalues of S equals the number of positive pivots.

# Symmetric Matrices (2)

Symmetric matrices S have orthogonal eigenvector matrices Q. Look at this again:

**Symmetry** 
$$S = X\Lambda X^{-1}$$
 becomes  $S = Q\Lambda Q^{T}$  with  $Q^{T}Q = I$ .

This says that every 2 by 2 symmetric matrix is (rotation)(stretch)(rotate back)

$$S = Q\Lambda Q^{\mathrm{T}} = \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 \end{bmatrix} \begin{bmatrix} \lambda_1 & \\ & \lambda_2 \end{bmatrix} \begin{bmatrix} \mathbf{q}_1^{\mathrm{T}} \\ \mathbf{q}_2^{\mathrm{T}} \end{bmatrix}. \tag{5}$$

Columns  $q_1$  and  $q_2$  multiply rows  $\lambda_1 q_1^T$  and  $\lambda_2 q_2^T$  to produce  $S = \lambda_1 q_1 q_1^T + \lambda_2 q_2 q_2^T$ .

Write A as S + N, symmetric matrix S plus skew-symmetric matrix N:

$$A = egin{bmatrix} 1 & 2 & 4 \\ 4 & 3 & 0 \\ 8 & 6 & 5 \end{bmatrix} = S + N \qquad \quad (S^{\mathrm{T}} = S \ \ \mathrm{and} \ \ N^{\mathrm{T}} = -N).$$

For any square matrix,  $S = \frac{1}{2}(A + A^{T})$  and N = \_\_\_\_ add up to A.

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#### **Answer**

$$A = \begin{bmatrix} 1 & 3 & 6 \\ 3 & 3 & 3 \\ 6 & 3 & 5 \end{bmatrix} + \begin{bmatrix} 0 & -1 & -2 \\ 1 & 0 & -3 \\ 2 & 3 & 0 \end{bmatrix} = \frac{1}{2}(A + A^{T}) + \frac{1}{2}(A - A^{T})$$
= symmetric + skew-symmetric.

Find an orthogonal matrix Q that diagonalizes  $S = \begin{bmatrix} -2 & 6 \\ 6 & 7 \end{bmatrix}$ . What is  $\Lambda$ ?

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#### **Answer**

$$\lambda=10 \text{ and } -5 \text{ in } \Lambda=\begin{bmatrix}10&0\\0&-5\end{bmatrix}, \ x=\begin{bmatrix}1\\2\end{bmatrix} \text{ and } \begin{bmatrix}2\\-1\end{bmatrix} \text{ have to be normalized to}$$
 unit vectors in  $Q=\frac{1}{\sqrt{5}}\begin{bmatrix}1&2\\2&-1\end{bmatrix}$ .

*True* (with reason) *or false* (with example).

- (a) A matrix with real eigenvalues and n real eigenvectors is symmetric.
- (b) A matrix with real eigenvalues and n orthonormal eigenvectors is symmetric.
- (c) The inverse of an invertible symmetric matrix is symmetric.
- (d) The eigenvector matrix Q of a symmetric matrix is symmetric.

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- (a) A matrix with real eigenvalues and n real eigenvectors is symmetric.
- (b) A matrix with real eigenvalues and n orthonormal eigenvectors is symmetric.
- The inverse of an invertible symmetric matrix is symmetric.
- The eigenvector matrix Q of a symmetric matrix is symmetric.

#### **Answer**

(a) False. 
$$A=\begin{bmatrix}1&2\\0&1\end{bmatrix}$$
 (b) True from  $A^{\mathrm{T}}=Q\Lambda Q^{\mathrm{T}}=A$  (c) True from  $S^{-1}=Q\Lambda^{-1}Q^{\mathrm{T}}$ 

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$$A^{\mathrm{T}} = Q\Lambda Q^{\mathrm{T}} = A$$

(c) True from 
$$S^{-1} = Q\Lambda^{-1}Q^{-1}$$

## **Positive Definite Matrices**

Applications from ML

- Cholesky decomposition  $A = LL^H$  (A special case of A = LU)
- Least squares computation reduction
- Support Vector Machine (SVM), kernel Positive-definite kernel
- Representer Theorem

Five tests

**Positive definite matrices** are the best. How to test S for  $\lambda_i > 0$ ?

- Test 1 Compute the **eigenvalues** of S: All eigenvalues positive
- Test 2 The energy  $x^{\mathrm{T}}Sx$  is positive for every vector  $x \neq 0$
- Test 3 The **pivots** in elimination on S are all positive
- Test 4 The upper left **determinants** of S are all positive
- Test 5  $S = A^T A$  for some matrix A with independent columns

### **Positive Definite Matrices**

*Applications from Robotics* 

Matrix form of Lagrange

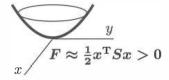
## **Positive Definite Matrices**

Important Application: Test for a Minimum

Does F(x,y) have a minimum if  $\partial F/\partial x=0$  and  $\partial F/\partial y=0$  at the point (x,y)=(0,0)?

For f(x), the test for a minimum comes from calculus: df/dx is zero and  $d^2f/dx^2 > 0$ . Two variables in F(x,y) produce a symmetric matrix S. It contains four second derivatives. Positive  $d^2f/dx^2$  changes to positive definite S:

Second derivatives 
$$S = \left[ \begin{array}{ccc} \partial^2 F/\partial x^2 & \partial^2 F/\partial x \partial y \\ \partial^2 F/\partial y \partial x & \partial^2 F/\partial y^2 \end{array} \right]$$



F(x,y) has a minimum if  $\partial F/\partial x = \partial F/\partial y = 0$  and S is positive definite.

Reason: S reveals the all-important terms  $ax^2 + 2bxy + cy^2$  near (x, y) = (0, 0). The second derivatives of F are 2a, 2b, 2b, 2c. For F(x, y, z) the matrix S will be 3 by 3.

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### Reference material

- Lecture 28, Positive Definite Matrices and Minima
- "Introduction to Linear Algebra", pdf pages 349–374
   6.4 Symmetric, 6.5 Positive Definite matrices
- "Linear Algebra and Applications", pdf pages 355–376
   Positive Definite Matrices 6.1, 6.2

