

## Eigenvalues and Positivity of a Matrix

**Definition.** A matrix  $A \in \mathbb{R}^{n \times n}$  is called positive definite (semi-definite) if  $x^T A x > 0$  ( $\geq 0$ ) for all  $x \in \mathbb{R}^n \setminus \{0\}$ .

Let  $A \in \mathbb{R}^{n \times n}$  be any matrix. Please show:

1. If  $A$  is positive definite (semi-definite), then  $\lambda > 0$  ( $\geq 0$ ) for all eigenvalues  $\lambda \in \sigma(A)$ .  
(Hint: Rayleigh quotient.)
2. The reverse is not true: Find an example of a matrix, which has positive eigenvalues, but is *not* positive definite.  
(Hint: Consider, e.g., a triangular  $2 \times 2$  matrix.)
3. The reverse is true for symmetric matrices: Let  $A$  be *symmetric* and  $\lambda > 0$  ( $\geq 0$ ) for all eigenvalues  $\lambda \in \sigma(A)$ , then  $A$  is positive definite (semi-definite).  
(Hint: Eigendecomposition.)
4. Let  $A$  be *symmetric*, then  $A$  is invertible if and only if  $\lambda \neq 0$  for all eigenvalues  $\lambda \in \sigma(A)$ .  
(Hint: Eigendecomposition.)

### Solution:

1. Let  $A$  be positive definite (semi-definite) and let  $(\lambda, v)$  be an eigenpair of  $A$ . Then by using the Rayleigh quotient we find

$$\lambda = \frac{v^T A v}{v^T v} = \frac{v^T A v}{\|v\|_2^2} \stackrel{\substack{>0 \text{ } (\geq 0) \\ >0 \text{ since } v \neq 0}}{>0 \text{ } (\geq 0)} > 0 \text{ } (\geq 0).$$

2. Take for example

$$A = \begin{pmatrix} 1 & -4 \\ 0 & 2 \end{pmatrix}$$

which has positive eigenvalues  $\sigma(A) = \{1, 2\}$ , but for  $x = (1, 1)^T$  we have

$$x^T A x = -3 + 2 = -1 < 0.$$

3. For symmetric  $A \in \mathbb{R}^{n \times n}$  we find an orthogonal matrix  $Q \in \mathbb{R}^{n \times n}$  and diagonal matrix  $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$  with  $\lambda_i \in \sigma(A)$ , so that

$$A = Q \Lambda Q^T \quad (\text{eigendecomposition}).$$

Since  $Q$  is orthogonal, any vector  $x \in \mathbb{R}^n$  can be written as  $x = Ix = Q Q^T x = Q \mu$ , with coordinates  $\mu := Q^T x$  where  $\mu \neq 0$  for  $x \neq 0$ . Therefore

$$x^T A x = \mu^T Q^T Q \Lambda Q^T Q \mu = \mu^T \Lambda \mu = \sum_{i=1}^n \mu_i^2 \lambda_i \geq 0.$$

We have “ $>$ ” if  $x \neq 0$ .

4. For symmetric  $A \in \mathbb{R}^{n \times n}$  we find an orthogonal matrix  $Q \in \mathbb{R}^{n \times n}$  and diagonal matrix  $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$  with  $\lambda_i \in \sigma(A)$ , so that

$$A = Q \Lambda Q^T \quad (\text{eigendecomposition}).$$

Therefore:

$$A \text{ invertible} \stackrel{(*)}{\Leftrightarrow} \Lambda \text{ invertible} \stackrel{(**)}{\Leftrightarrow} \lambda \neq 0 \quad \forall \lambda \in \sigma(A).$$

(\*) product of matrices is invertible if all factors are invertible and orthogonal matrices are invertible with inverse  $Q^T = Q^{-1}$

(\*\*) diagonal matrices are invertible, if and only if diagonal elements are nonzero

Example:  $A$  symmetric and positive definite (spd)  $\Rightarrow A$  invertible.