

# 1 Lecture Review

## 1.1 Positive Definite Matrices

If a symmetric matrix  $S$  has one of these properties, it has them all:

1. All eigenvalues are  $> 0$  ( $S$  is positive definite).
2.  $S = LU$  where  $L$  is unit lower triangular and  $U$  is upper triangular with positive diagonal entries.
3.  $S = LDL^T$  where  $L$  is unit lower triangular and  $D$  is a diagonal with positive diagonal entries.
4. All  $n$  upper left determinants are positive.
5.  $x^T S x > 0$  unless  $x = 0$ .
6.  $S = A^T A$  for  $A$  with independent columns.
7. The compact SVD = the full SVD = an eigenfactorization.

## 2 Problems

1. Show that if  $S$  is positive semidefinite, then  $x^T S x \geq 0$  for any vector  $x$ . Then show the other direction: if  $x^T S x \geq 0$  for any vector  $x$ , then  $S$  is positive semidefinite.

*Solution.* Assume first that  $S$  is positive semidefinite. Since  $S$  is symmetric, we have a diagonalization  $S = Q\Lambda Q^{-1}$  with an orthogonal  $Q$ , so in fact  $S = Q\Lambda Q^T$ . Since  $S$  is positive semidefinite, the diagonal entries  $\lambda_i$  of  $\Lambda$  are nonnegative. Let us denote by  $q_i$  the  $i$ th column of  $Q$  – the  $i$ th eigenvector of  $S$ , so  $Sq_i = \lambda_i q_i$ .

Now take any vector  $x$ . Since  $S$  is diagonalizable, its eigenvectors span the whole vector space, so  $x = \sum_i c_i q_i$ . Then compute:

$$x^T S x = x^T S \sum_i c_i q_i = \sum_i c_i x^T \lambda_i q_i = \sum_i c_i \lambda_i \left( \sum_j c_j q_j^T q_i \right).$$

Since  $Q$  is orthogonal, the system of vectors  $q_1, \dots, q_n$  is orthonormal, so we can continue the sequence of equalities:

$$x^T S x = \sum_i c_i \lambda_i (c_i q_i^T q_i) = \sum_i c_i^2 \lambda_i.$$

Since we know that  $\lambda_i \geq 0$  and  $c_i^2$  is always nonnegative, we can conclude that  $x^T S x \geq 0$  for every vector  $x$ .

Now prove the reverse direction. Assume that for any vector  $x$ , we have  $x^T S x \geq 0$ . Now, if  $x$  is an eigenvector with eigenvalue  $\lambda$ , then  $0 \leq x^T S x = x^T \lambda x = \lambda \|x\|^2$ . Therefore, the eigenvalue  $\lambda$  must be nonnegative.  $\square$

2. Show that if  $S = A^T A$  for some matrix  $A$ , then  $x^T S x \geq 0$  for any vector  $x$ .

*Solution.*  $x^T S x = x^T A^T A x = (Ax)^T (Ax) = \|Ax\|^2 \geq 0$ .

□

3. Let  $P$  be the  $n \times n$  matrix

$$\begin{pmatrix} \frac{1}{n} & \cdots & \frac{1}{n} \\ \vdots & \ddots & \vdots \\ \frac{1}{n} & \cdots & \frac{1}{n} \end{pmatrix}.$$

Is this matrix: (i) symmetric, (ii) positive semi-definite, (iii) positive-definite, (iv) Markov, (v) positive Markov, (vi) a projection matrix, (vii) rank 1? Explain, then determine the eigenvalues of  $P$ .

*Solution.* (i) Yes.

- (ii) Observe that  $v = (1, \dots, 1)^T$  is an eigenvector with eigenvalue 1. Moreover,  $P$  is of rank one, so the nullspace is of dimension  $n - 1$ , so 0 is an eigenvalue with multiplicity  $n - 1$ . So the eigenvalues are 0 and 1, both nonnegative, so  $P$  is positive semidefinite.
- (iii) No, because 0 is an eigenvalue. Unless  $n = 1$ , then  $P = (1)$  is positive definite.
- (iv) Yes.
- (v) Yes.
- (vi) Yes, because  $P^2 = P$  and  $P^T = P$ .
- (vii) Yes, because all the columns are equal.

□

4. Let  $P$  be the matrix from the previous problem. Let  $M = I - P$  and answer (i)-(vii) from the previous problem for this matrix. What is the rank of  $M$  and what are its eigenvalues?

*Solution.* (i) Yes, because it is a difference between two symmetric matrices and taking transpose is a linear operation.

- (ii) If  $x$  is an eigenvector of  $P$  with eigenvalue  $\lambda$ , then it is also an eigenvector of  $M$  with a different eigenvalue  $1 - \lambda$ , because  $Mx = (I - P)x = x - Px = x - \lambda x = (1 - \lambda)x$ . So  $M$  has one eigenvalue 0 and the eigenvalue of 0 has multiplicity  $n - 1$ . So it is also positive semidefinite.
- (iii) No, because 0 is an eigenvalue.
- (iv) No, because it has negative entries  $-\frac{1}{n}$  (if  $n > 1$ ) and moreover the sum of entries in each column is equal to 0, not to 1.
- (v) No, because it is not even Markov.
- (vi) Yes, because  $(I - P)^2 = I - P - P + P^2 = I - P$ , using  $P^2 = P$ , and  $(I - P)^T = I - P$ .
- (vii) No, unless  $n = 2$ . The rank of the matrix  $M$  is  $\text{rk } M = \dim \text{Col } M = n - \dim \text{Nul } M = n - 1$ .

□

5. True or False:

- (a) Every positive definite matrix is invertible.
- (b) The only positive definite projection matrix is  $P = I$ .
- (c) Every projection matrix is positive semidefinite.
- (d) A diagonal matrix with positive diagonal entries is positive definite.
- (e) A symmetric matrix with positive determinant is positive definite.

*Solution.* (a) True. Because if  $S$  is positive definite, then  $S = Q\Lambda Q^{-1}$  – a product of three invertible matrices, and here  $\Lambda$  is invertible, because it is diagonal with nonzero diagonal entries.

- (b) True. If  $P$  is a projection matrix, then from  $P^2 = P$  we conclude that its eigenvalues are necessarily 0 or 1. If  $P$  is in addition positive definite, then they should all be 1. So  $P$  is similar to the identity matrix  $I$ , hence must be  $I$  itself.
- (c) True, because the possible eigenvalues are 0 and 1.
- (d) True, because it is symmetric and the diagonal entries are its eigenvalues.
- (e) False, because we can take the  $2 \times 2$  matrix  $-I$  which is negative definite and whose determinant is 1.

□

6. Without multiplying

$$S = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 5 \end{pmatrix} \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix},$$

find

- |                            |                                            |
|----------------------------|--------------------------------------------|
| (a) the determinant of $S$ | (c) the eigenvectors of $S$                |
| (b) the eigenvalues of $S$ | (d) a reason why $S$ is positive definite. |

*Solution.* Note that we can introduce the notation  $S = Q\Lambda Q^T$ . Here  $Q$  is an orthogonal matrix, so we also have  $S = Q\Lambda Q^{-1}$ . Therefore, we are given a diagonalization of  $S$ .

- (a)  $\det S = \det \Lambda = 2 \cdot 5 = 10$ , and here we used the fact that  $\det S$  is equal to the determinant of its diagonalized form.
- (b) Since  $S = Q\Lambda Q^{-1}$  is a diagonalization, the eigenvalues are 2 and 5 – the diagonal entries of  $\Lambda$ .
- (c) The eigenvectors of  $S$  are the two columns of  $Q$ .
- (d) The matrix  $S$  is symmetric, because  $S^T = (QQ^T)^T = Q\Lambda Q^T = S$ , and has positive eigenvalues 2 and 5, and therefore is positive definite.

□

7. Suppose  $A = \begin{pmatrix} a & b \\ b & d \end{pmatrix}$ . Explain why  $ad - b^2 > 0$  and  $a + d > 0$  imply  $A$  is positive definite. How should we change these conditions to ensure that the eigenvalues have opposite sign.

*Solution.* Let  $\lambda$  and  $\mu$  be the two eigenvalues of the symmetric matrix  $A$ . Then  $\lambda\mu = \det A = ad - b^2 > 0$  and hence  $\lambda$  and  $\mu$  must have the same sign. If in addition  $\lambda + \mu = \operatorname{tr} A = a + d > 0$ , then these values must be positive. So  $A$  is necessarily positive definite.

If we want opposite signs of the eigenvalues, we must require  $\det A = ad - b^2 < 0$ . □

8. Find an example of a  $3 \times 3$  matrix with positive determinant and positive trace which is not positive definite.

*Solution.* Empty spaces denote zeroes:

$$\begin{pmatrix} 3 & & \\ & -1 & \\ & & -1 \end{pmatrix}.$$

□

9. True or False (assume  $A$  is  $n \times n$ ):

- (a) If  $A$  is a matrix whose columns sum to 0, then  $A + I$  is a Markov matrix.
- (b) If  $A$  is a diagonal matrix and is a Markov matrix, then  $A = I$ .
- (c) If  $A$  is Markov then  $I - A$  is positive semidefinite.
- (d) If  $A$  is positive Markov then  $I - A$  has rank  $n - 1$ .

*Solution.* (a) False, for example take  $A = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$ .

- (b) True, because then the only nonzero entry in each column is one on the diagonal, hence it must be 1 to ensure that the sum is equal to 1.
- (c) True. If  $A$  is Markov, then any of its eigenvalues is at most one. So if  $Ax = \lambda x$ , then  $(I - A)x = (1 - \lambda)x$ , and  $1 - \lambda > 0$ .
- (d) True. If  $A$  is positive Markov, then there is only one eigenvalue equal to 1. If  $\lambda$  is an eigenvalue of  $A$ , then  $1 - \lambda$  is an eigenvalue of  $I - A$ , and all eigenvalues of  $I - A$  can be obtained this way. So the only way to get a zero eigenvalue for  $I - A$  is to take  $\lambda = 1$  in  $A$ , but there is exactly one eigenvalue of  $A$  equal to 1. So the nullspace of  $I - A$  is one-dimensional, therefore its rank is  $n - 1$ .

□