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# Proof semi-positive definition of the product of a positive diagonal matrix and a symmetric semi-positive definite matrix

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**Problem statement:** We have  $\Lambda$  a real symmetric semi-positive definite matrix and D a real positive diagonal matrix (thus also symmetric). The product  $\Lambda D$  is a square matrix but non-symmetric. We want to prove that  $\Lambda D$  quadratic form,  $x^T \Lambda D x$ , is semi-positive definite, i.e.  $x^T \Lambda D x \geq 0 \,\forall x \in \mathbb{R}^n \setminus \{0\}$ .

The proof is the following:

*Proof.* Let us recall some useful matrix definitions and properties:

**Definition 1** (Real symmetric semi-positive definite matrix). A is a real symmetric matrix (its eigenvalues are thus real): A is semi-positive definite  $\iff$  all its eigenvalues are non-negative.

**Definition 2** (Quadratic form semi-positive definition). A matrix A is semi-positive definite  $\iff x^T A x \ge 0, \forall x \in \mathbb{R}^n \setminus \{0\}$ 

**Property 1.** If A is a real semi-positive definite matrix, then  $B^TAB$  is semi-positive definite for any matrix B.

**Property 2** (Matrix congruent to a symmetric matrix). Any matrix congruent to a symmetric matrix is again symmetric: If A is a symmetric matrix then so is  $B^TAB$  for any matrix B.

Let us prove that  $\Lambda D$  is semi-positive definite with respect to the Definition 2.

#### First, we prove that the eigenvalues of $\Lambda D$ are reals and non-negatives:

Because D is diagonal and positive we can write  $D = D^{\frac{1}{2}}D^{\frac{1}{2}}$ ,  $D^{\frac{1}{2}}$  is also real positive and symmetric thus invertible. Let us reformulate the matrix  $\Lambda D$ :

$$\Lambda D = \Lambda D^{\frac{1}{2}} D^{\frac{1}{2}} 
= D^{-\frac{1}{2}} (D^{\frac{1}{2}} \Lambda D^{\frac{1}{2}}) D^{\frac{1}{2}}$$
(1)

It corresponds to a change of basis of  $D^{\frac{1}{2}}$ . Because the eigenvalues (denoted  $\lambda$ ) are invariant to change of basis we have:

$$\lambda(\Lambda D) = \lambda(D^{\frac{1}{2}}\Lambda D^{\frac{1}{2}}) \tag{2}$$

Using the Properties.1 and 2, because  $\Lambda$  is a real symmetric semi-positive definite matrix,  $(D^{\frac{1}{2}})^T \Lambda D^{\frac{1}{2}} = D^{\frac{1}{2}} \Lambda D^{\frac{1}{2}}$  is a real symmetric semi-positive definite matrix. Thus its eigenvalues

are real and non-negatives (Definition.1) and so are the ones of  $\Lambda D$  because of Eq.2. Then we have proven that the eigenvalues of  $\Lambda D$  are are real and non-negative.

#### Now, let us prove the semi-positive definition of $\Lambda D$ :

**Property 3** (Types of Quadratic Forms). Quadratic forms can be classified according to the nature of the eigenvalues of the matrix of the quadratic form:

- If all  $\lambda_i$  are positive, the form is said to be positive definite.
- If all  $\lambda_i$  are negative, the form is said to be negative definite.
- If all  $\lambda_i$  are non-negative (positive or zero), the form is said to be positive semi-definite.
- If all  $\lambda_i$  are non-positive (zero or negative), the form is said to be negative semi-definite.
- If the  $\lambda_i$  represent a mixture of positive, zero, and negative values, the form is said to be indefinite

Using the Property 3 of [1], because the eigenvalues of  $\Lambda D$  are are real and non-negative we have directly the semi-positive definition of the quadratic form.

This property can be proven as follows:

**Definition 3** (Toeplitz decomposition). Every square matrix A can be decomposed uniquely as the sum of two matrices U and V, where U is symmetric and V is skew-symmetric.

$$A = U + V = \frac{1}{2}(A + A^{T}) + \frac{1}{2}(A - A^{T})$$
(3)

In our case  $A = \Lambda D$  and  $U = \frac{1}{2}(\Lambda D + (\Lambda D)^T)$  is symmetric.

We recall the fact that the quadratic form of a skew-symmetric matrix equals to zero. Indeed, by definition  $V^T = -V$  and thus  $x^T V x = (x^T V^T x)^T = -x^T V x$  which holds only if it equals to zero. Thus, the quadratic form of  $x^T \Lambda D x$  is the same one of  $x^T U x$ , i.e.

$$x^{T}\Lambda Dx = x^{T} \left(\frac{\Lambda D + (\Lambda D)^{T}}{2}\right) x \tag{4}$$

One can thus prove the semi-positive definition of the symmetric matrix U to prove the semi-positive definition of  $\Lambda D$ .

We use the following theorem on matrices [2] (Chapter 10, Theorem 10.28):

**Theorem 1.** Let A be an  $n \times n$  matrix with eigenvalues  $\lambda_1(A), ..., \lambda_n(A)$  and  $\mathcal{R}_e \lambda_i(A)$  their real part. Let  $\lambda(A) = \sum_{i=1}^n \lambda_i(A)$  and  $\mathcal{R}_e \lambda(A) = \sum_{i=1}^n \mathcal{R}_e \lambda_i(A)$ . Then:

$$\mathcal{R}_e \lambda(A) \le \lambda \left(\frac{A + A^T}{2}\right) \tag{5}$$

Then, because the eigenvalues of  $\Lambda D$  are real and non-negative (as proven in the previous part), we have using the Theorem 1:

$$0 \le \lambda(\Lambda D) \le \lambda(U) \tag{6}$$

Thus, the eigenvalues of U are non-negative, delivering the proof of the semi-positive definition of U, because U is symmetric (see Definition 1). Then,  $x^TUx \geq 0$ ,  $\forall x \in \mathbb{R}^n \setminus \{0\}$  and using Eq.4, we obtain  $x^T \Lambda Dx \geq 0$ ,  $\forall x \in \mathbb{R}^n \setminus \{0\}$ : proving of the semi-positive definition of  $\Lambda D$ .

## References

- [1] J. D. Carroll and P. E. Green. Decomposition of Matrix Transformations: Eigen structures and Quadratic Forms. Mathematical Tools for Applied Multivariate Analysis. Elsevier Inc., 1997.
- [2] Fuzhen Zhang. Matrix theory: basic results and techniques. Springer, 2011.