

A PROOF OF HADAMARD INEQUALITY FOR
NONSINGULAR OR POSITIVE DEFINITE MATRICES

Sumitra Purkayastha

Bayesian and Interdisciplinary Research Unit
Indian Statistical Institute
203, B.T. Road, Kolkata, 700 108, INDIA
e-mail: sumitra@isical.ac.in

Abstract: Hadamard inequality for nonsingular matrices states the following: if $\mathbf{A} = ((a_{ij}))$ is an $n \times n$ nonsingular matrix, then $[\det(\mathbf{A})]^2 \leq \prod_{i=1}^n (\sum_{j=1}^n a_{ij}^2)$. This result is equivalent to the following fact about positive definite matrices: if $\mathbf{B} = ((b_{ij}))$ is an $n \times n$ positive definite matrix, then $\det(\mathbf{B}) \leq \prod_{i=1}^n b_{ii}$. We provide a simple proof of this later fact that uses in a straightforward manner spectral decomposition of a positive definite matrix and the inequality between arithmetic mean and geometric mean of a set of positive numbers.

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1. Introduction

We begin by stating Hadamard inequality for nonsingular matrices.

Theorem 1. *Let $\mathbf{A} = ((a_{ij}))$ be an $n \times n$ nonsingular matrix. Then*

$$[\det(\mathbf{A})]^2 \leq \prod_{i=1}^n \left(\sum_{j=1}^n a_{ij}^2 \right). \quad (1)$$

Remark 1. Notice that any singular matrix of order $n \times n$ can be written as the limit of a sequence of $n \times n$ non-singular matrices, where limit of a sequence

of matrices $\mathbf{M}_n = ((m_{ij}^{(n)}))$ is defined as the matrix $\mathbf{M} \stackrel{\text{def}}{=} ((\lim_{n \rightarrow \infty} m_{ij}^{(n)}))$, assuming $\lim_{n \rightarrow \infty} m_{ij}^{(n)}$ to exist for every (i, j) . Notice, moreover, that both sides of (1) are continuous functions of the entries of \mathbf{A} . Theorem 1 is, therefore, easily seen to hold *also* for any singular matrix of order $n \times n$.

Remark 2. Notice that a positive definite matrix $\mathbf{B} = ((b_{ij}))$ of order $n \times n$ can be written as $\mathbf{B} = \mathbf{A}\mathbf{A}^T$, for some $n \times n$ non-singular matrix $\mathbf{A} = ((a_{ij}))$. With this choice of \mathbf{A} , the quantity appearing on the left-hand side of (1) equals $\det(\mathbf{B})$. Also, $b_{ii} = \sum_{j=1}^n a_{ij}^2$ for every $i = 1, \dots, n$. Thus, Theorem 1 above implies the following.

Theorem 2. Let $\mathbf{B} = ((b_{ij}))$ be an $n \times n$ positive definite matrix. Then

$$\det(\mathbf{B}) \leq \prod_{i=1}^n b_{ii}. \quad (2)$$

Remark 3. A close scrutiny of arguments leading to validity of Theorem 2 assuming Theorem 1 to hold (cf. Remark 2) will show that Theorem 2 implies Theorem 1. In other words, Theorems 1 and 2 are equivalent.

Remark 4. Since any non-negative definite matrix can be written as the limit of a sequence of $n \times n$ positive definite matrices, and also since both sides of (2) are continuous functions of the entries of \mathbf{B} , Theorem 2 is easily seen to hold *also* for any non-negative definite matrix.

Remark 5. Many proofs of Theorem 1 or of Theorem 2 are known to exist in literature. We describe two of them below, and mention briefly two more.

In perhaps the shortest proof of Theorem 2 (see Mirsky [3, p. 418]), one defines $\mathbf{C} = ((c_{ij}))$ by $c_{ij} = b_{ij}/\sqrt{b_{ii}b_{jj}}$. Then one notices that (2) is equivalent to the following: $\det(\mathbf{C}) \leq 1$. To see this, one notices that $\det(\mathbf{C}) = \alpha_1 \cdots \alpha_n$ and $1 = [(c_{11} + \cdots + c_{nn})/n]^n = [(\alpha_1 + \cdots + \alpha_n)/n]^n$, where $\alpha_1, \dots, \alpha_n$ are the eigenvalues of \mathbf{C} . Since \mathbf{C} is positive definite, each $\alpha_i > 0$. Hence, $[(\alpha_1 + \cdots + \alpha_n)/n]^n \geq \alpha_1 \cdots \alpha_n$, which, in turn, implies $\det(\mathbf{C}) \leq 1$.

In another proof of Theorem 2 (see Mirsky [3, pp. 416-417]), one defines $\mathbf{B}_k \stackrel{\text{def}}{=} ((b_{ij}))_{i,j=k,\dots,n}$, $k = 1, \dots, n$, and also $\mathbf{B}' \stackrel{\text{def}}{=} \mathbf{B}$ except the $(1, 1)$ -th entry of \mathbf{B}' which is given by $b_{11} - \det(\mathbf{B})/\det(\mathbf{B}_2)$, and shows that \mathbf{B}' is positive semi-definite. Consequently, $\det(\mathbf{B}) \leq b_{11} \det(\mathbf{B}_2)$, and the proof follows by induction.

Yet another proof of Theorem 2 (see Bellman [2, p. 126]) is as follows. Notice that $\det(\mathbf{B}) = b_{11}\mathbf{B}_{11} + \det(\mathbf{B}_{0,11})$, where \mathbf{B}_{11} is the co-factor of b_{11} and $\mathbf{B}_{0,11}$ is the matrix obtained from \mathbf{B} by making its $(1, 1)$ -th entry 0 and leaving

all other entries unchanged. It is easy to verify, by noticing that the matrix obtained by deleting the first row and first column of $\mathbf{B}_{0,11}$ is positive definite, that $\det(\mathbf{B}_{0,11}) \leq 0$. The rest of the proof follows by induction.

The last proof we wish to mention (see Bellman [2, p. 127], Beckenbach and Bellman [1, pp. 63-64]) uses the relation between $\det(\mathbf{B})$ and the integral of $\exp(-\mathbf{x}^T \mathbf{B} \mathbf{x} / 2)$ over \mathbb{R}^n , and some clever arguments.

We present in this paper a proof of Theorem 2 that uses in a straightforward manner spectral decomposition of a positive definite matrix and also the inequality between arithmetic mean and geometric mean of a set of positive numbers. The proof is provided in Section 2.

2. A Proof of Theorem 2

Suppose

$$\mathbf{B} = \mathbf{P} \mathbf{L} \mathbf{P}^T \tag{3}$$

is a spectral decomposition of \mathbf{B} , where $\mathbf{P} = ((p_{ij}))$ is an $n \times n$ orthogonal matrix and \mathbf{L} is an $n \times n$ diagonal matrix with $\mathbf{L} = \text{diag}(\lambda_1, \dots, \lambda_n)$. Each λ_i is positive.

Notice that

$$\det(\mathbf{B}) = \prod_{i=1}^n \lambda_i. \tag{4}$$

Also, (3) implies the following: $b_{ii} = \sum_{k=1}^n \lambda_k p_{ik}^2$ for $i = 1, \dots, n$. Therefore,

$$\prod_{i=1}^n b_{ii} = \sum_{k_1=1}^n \sum_{k_2=1}^n \cdots \sum_{k_n=1}^n \lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2. \tag{5}$$

Notice now that

$$\sum_{k_1=1}^n \sum_{k_2=1}^n \cdots \sum_{k_n=1}^n p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 = \prod_{j=1}^n \left(\sum_{k_j=1}^n p_{jk_j}^2 \right) = 1, \tag{6}$$

since $\sum_{k_1=1}^n p_{1k_1}^2 = \sum_{k_2=1}^n p_{2k_2}^2 = \cdots = \sum_{k_n=1}^n p_{nk_n}^2 = 1$, which is implied by orthogonality of \mathbf{P} . Moreover, $p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 \geq 0 \forall k_1, k_2, \dots, k_n$. Hence,

$$\sum_{k_1=1}^n \sum_{k_2=1}^n \cdots \sum_{k_n=1}^n \lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2$$

is a weighted average of the following set of positive quantities:

$$\{\lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n} : k_j = 1, \dots, n; j = 1, \dots, n\}.$$

The weight associated with $\lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n}$ is $p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2$.

It follows from the preceding facts and the standard inequality between arithmetic mean and geometric mean of a set of positive numbers that

$$\sum_{k_1=1}^n \sum_{k_2=1}^n \cdots \sum_{k_n=1}^n \lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 \geq \prod_{k_1=1}^n \prod_{k_2=1}^n \cdots \prod_{k_n=1}^n (\lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n}) p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2. \tag{7}$$

Letting $q_j \stackrel{\text{def}}{=} \prod_{k_1=1}^n \prod_{k_2=1}^n \cdots \prod_{k_n=1}^n \lambda_{k_j} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2$, $j = 1, \dots, n$, and $Q \stackrel{\text{def}}{=} \prod_{j=1}^n q_j$, we re-write (7) as

$$\sum_{k_1=1}^n \sum_{k_2=1}^n \cdots \sum_{k_n=1}^n \lambda_{k_1} \lambda_{k_2} \cdots \lambda_{k_n} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 \geq Q. \tag{8}$$

Consider now the term $q_1 = \prod_{k_1=1}^n \prod_{k_2=1}^n \cdots \prod_{k_n=1}^n \lambda_{k_1} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2$ in Q . Notice that (cf. (6))

$$\sum_{k_2=1}^n \cdots \sum_{k_n=1}^n p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 = \prod_{j=2}^n \left(\sum_{k_j=1}^n p_{jk_j}^2 \right) = 1 \tag{9}$$

and so

$$\prod_{k_2=1}^n \cdots \prod_{k_n=1}^n \lambda_{k_1} p_{1k_1}^2 p_{2k_2}^2 \cdots p_{nk_n}^2 = (\lambda_{k_1} p_{1k_1}^2)^{\sum_{k_2=1}^n \cdots \sum_{k_n=1}^n p_{2k_2}^2 \cdots p_{nk_n}^2} = \lambda_{k_1} p_{1k_1}^2. \tag{10}$$

It follows, therefore, from (9) and (10) that $q_1 = \prod_{k_1=1}^n \lambda_{k_1} p_{1k_1}^2$. It can be proved in this manner that $q_j = \prod_{k_j=1}^n \lambda_{k_j} p_{jk_j}^2$ for $j = 1, \dots, n$. This implies

$$Q = \prod_{i=1}^n \left(\prod_{k_i=1}^n \lambda_{k_i} p_{ik_i}^2 \right). \tag{11}$$

Notice now from (11) that $Q = \prod_{j=1}^n \left(\lambda_j^{\sum_{i=1}^n p_{ij}^2} \right)$. Moreover, orthogonality of \mathbf{P} implies $\sum_{i=1}^n p_{ij}^2 = 1$ for every $j = 1, \dots, n$. Hence,

$$Q = \lambda_1 \lambda_2 \cdots \lambda_n. \tag{12}$$

The theorem follows now from (4), (5), (8), and (12).

Remark 6. The proof presented in this paper relies on spectral decomposition of a positive definite matrix and also on the standard inequality between

arithmetic mean and geometric mean of a set of positive numbers, but does not depend on any trick. Even though our proof may look complicated at a first sight, we believe that on close inspection it will be found to be simple.

References

- [1] E.F. Beckenbach, R. Bellman, *Inequalities*, Springer-Verlag, Berlin (1961).
- [2] R. Bellman, *Introduction to Matrix Analysis*, McGraw-Hill, New York (1960).
- [3] L. Mirsky, *An Introduction to Linear Algebra*, Clarendon, Oxford (1955).

