

## A PROOF OF THE GENERALIZED HADAMARD INEQUALITY VIA INFORMATION THEORY

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The generalized Hadamard inequality is considered in this paper. The new proof presented here is motivated by the idea of mutual information of random variables. For a strict algebraic proof see e.g. [3].

Let  $T$  be an  $m \times n$  matrix of real numbers and denote its column vectors with  $t_1, t_2, \dots, t_n$ . Let the set of column indices  $J = \{1, 2, \dots, n\}$  be partitioned into sets  $J_1, J_2, \dots, J_\rho$ , that is  $J_i \cap J_j = \emptyset$  for  $i \neq j$ ,  $\bigcup_{k=1}^{\rho} J_k = J$ . Let  $T_k = \{t_j : j \in J_k\}$ . Assume that the vectors  $t_j \in T_k$  are linearly independent for  $k = 1, 2, \dots, \rho$ . Now the generalized Hadamard inequality can be stated as follows.

**Theorem.**

$$\det(T^*T) \leq \det(T_1^*T_1)\det(T_2^*T_2) \dots \det(T_\rho^*T_\rho)$$

and equality holds iff  $T_i^*T_j = 0$  for all  $i \neq j$ . ( $A^*$  stands for the transpose of the matrix  $A$ ).

Note that if  $\rho = n$  and  $T_1 = \{t_1\}, T_2 = \{t_2\}, \dots, T_n = \{t_n\}$  then the assumption says that  $t_j \neq 0$  for all  $j$  and the theorem states the well-known version of the Hadamard inequality, that is

$$\det(T^*T) \leq t_1^2 t_2^2 \dots t_n^2$$

and equality holds iff  $t_i^*t_j = 0$  for all  $i \neq j$ .

Our proof goes through a couple of remarks.

**Remark 1.** Consider a real-valued function  $f(x), x = (x_1, x_2, \dots, x_n)$ . Let the set of indices  $J = \{1, 2, \dots, n\}$  be partitioned into sets  $J_1, \dots, J_\rho$ . After a proper rearranging of the components  $x_1, x_2, \dots, x_n$  we can split the vector  $x$  into subvectors  $x_1, x_2, \dots, x_\rho$  such that  $x_k$  contains the components  $x_j, j \in J_k$ . With this notation  $f(x) = f(x_1, x_2, \dots, x_\rho)$ . Assume that

- (i)  $f(x_1, x_2, \dots, x_\rho) \geq 0 \quad (x_1, x_2, \dots, x_\rho) \in \mathbb{R}^n$ ,
- (ii)  $\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_\rho) dx_1 dx_2 \dots dx_\rho = 1$ .

Assumptions (i), (ii) obviously hold for a density function of an  $n$ -dimensional random variable. Introducing the functions

$$f_k(\mathbf{x}_k) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_\rho) d\mathbf{x}_1 \dots d\mathbf{x}_{k-1} d\mathbf{x}_{k+1} \dots d\mathbf{x}_\rho, \quad k = 1, 2, \dots, \rho$$

we have

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f_k(\mathbf{x}_k) d\mathbf{x}_k = 1 \quad k = 1, 2, \dots, \rho$$

and

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_\rho) \log \frac{f(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_\rho)}{f_1(\mathbf{x}_1) f_2(\mathbf{x}_2) \dots f_\rho(\mathbf{x}_\rho)} d\mathbf{x}_1 d\mathbf{x}_2 \dots d\mathbf{x}_\rho \geq 0.$$

Here equality holds iff  $f(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_\rho) = f_1(\mathbf{x}_1) f_2(\mathbf{x}_2) \dots f_\rho(\mathbf{x}_\rho)$ . The latter inequality plays an important role in our discussion and is well-known in information theory (Shannon lemma). It can be proved from the fact that

$$\log t \geq 1 - \frac{1}{t} \quad \text{for } t \geq 0$$

and equality holds iff  $t = 1$ .

**Remark 2.** Consider a matrix  $T$  for which all assumptions incurred in the theorem are valid. Then the  $n \times n$  matrix  $T^*T$  is positive definite, symmetric and can be partitioned as follows.

$$T^*T = \begin{array}{|c|c|c|c|} \hline T_1^*T_1 & T_1^*T_2 & \dots & T_1^*T_\rho \\ \hline T_2^*T_1 & T_2^*T_2 & \dots & T_2^*T_\rho \\ \hline \vdots & \vdots & & \vdots \\ \hline T_\rho^*T_1 & T_\rho^*T_2 & \dots & T_\rho^*T_\rho \\ \hline \end{array}$$

With this  $T^*T$  let us consider the function

$$g(\mathbf{x}) = \frac{1}{\pi^{n/2}} \frac{1}{(\det(T^*T))^{1/2}} \exp\left(-\mathbf{x}^*(T^*T)^{-1}\mathbf{x}\right).$$

$g(\mathbf{x})$  is the density function of an  $n$ -dimensional random variable with Gauss distribution. Obviously

$$g(\mathbf{x}) \geq 0 \quad \mathbf{x} \in \mathbb{R}^n,$$

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g(x) dx = 1.$$

Let us partition the components of  $x$  into subvectors  $x_1, x_2, \dots, x_\rho$  according to the partition  $J_1, J_2, \dots, J_\rho$  of the column indices  $\{1, 2, \dots, n\}$ . One can show by evaluating the integral (see e.g.[1]) that

$$\begin{aligned} g_k(x_k) &= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g(x_1, x_2, \dots, x_\rho) dx_1 \dots dx_{k-1} dx_{k+1} \dots dx_\rho = \\ &= \frac{1}{\pi^{|J_k|/2}} \frac{1}{(\det(T_k^* T_k))^{1/2}} \exp\left(-x_k^*(T_k^* T_k)^{-1} x_k\right) \end{aligned}$$

where  $|J_k|$  denotes the cardinality of the set  $J_k$ .

**Remark 3.** Evaluating the integrals one can easily check that

$$\begin{aligned} &\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g(x_1, x_2, \dots, x_\rho) \log \frac{g(x_1, x_2, \dots, x_\rho)}{g_1(x_1)g_2(x_2)\dots g_\rho(x_\rho)} dx_1 dx_2 \dots dx_\rho = \\ &= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g(x_1, x_2, \dots, x_\rho) \log g(x_1, x_2, \dots, x_\rho) dx_1 dx_2 \dots dx_\rho - \\ &\quad - \sum_{k=1}^{\rho} \left( \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g_k(x_k) \log g_k(x_k) dx_k \right) = \\ &= \frac{1}{2} \log \frac{\det(T_1^* T_1) \det(T_2^* T_2) \dots \det(T_\rho^* T_\rho)}{\det(T^* T)}. \end{aligned}$$

Now applying the inequality of Remark 1 to  $g(x)$  states that the integral above is non-negative, and equals zero iff

$$\begin{array}{cccc|c|cccc} T_1^* T_1 & T_1^* T_2 & \dots & T_1^* T_\rho & & T_1^* T_1 & 0 & \dots & 0 \\ T_2^* T_1 & T_2^* T_2 & \dots & T_2^* T_\rho & & 0 & T_2^* T_2 & \dots & 0 \\ \vdots & \vdots & & \vdots & & \vdots & \vdots & & \vdots \\ T_\rho^* T_1 & T_\rho^* T_2 & \dots & T_\rho^* T_\rho & = & 0 & 0 & \dots & T_\rho^* T_\rho \end{array}$$

Hence the theorem is proved.

**Corollary.** Let us consider a positive definite symmetric matrix  $C$  of size  $n \times n$ . Take a partition of indices  $J = \{1, 2, \dots, n\}$  into the sets  $J_1, J_2, \dots, J_\rho$  and denote the submatrix of elements  $c_{ij}$   $i \in J_k, j \in J_l$  with  $C_{kl}$ . Then

$$\det(C) \leq \det(C_{11})\det(C_{22}) \dots \det(C_{\rho\rho})$$

and equality holds iff  $C_{kl} = 0$  for all  $k \neq l$ . To prove the corollary it should be noted that according to Cholesky's factorization theorem for every positive definite, symmetric matrix  $C = T^*T$  with a suitable non-singular matrix  $T$ . Then the theorem directly applies.

### References

- [ 1 ] **Anderson T.W.**, *An Introduction to Multivariate Statistical Analysis*, John Wiley & Sons Inc., New York, 1957.
- [ 2 ] **Csiszár I. and Körner J.**, *Information Theory*, Academic Press, 1981.
- [ 3 ] **Gantmacher F.R.**, *The theory of matrices*, Chelsea Publishing Company, New York, N.Y., 1959.
- [ 4 ] **Klafszky E. and Kas P.**, Megjegyzés egy többdimenziós Cauchy eloszlásról, *Alkalmazott Matematikai Lapok*, **13**(1987-88), 145-161.

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