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## proof of Hadamard's inequality

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Let's first prove the second inequality. If  $A$  is singular, the thesis is trivially verified, since for a Hermitian positive semidefinite matrix the right-hand side is always nonnegative, all the diagonal entries being nonnegative. Let's thus assume  $\det(A) \neq 0$ , which means,  $A$  being Hermitian positive semidefinite,  $\det(A) > 0$ . Then no diagonal entry of  $A$  can be 0 (otherwise, since for a Hermitian positive semidefinite matrix,  $0 \leq \lambda_{\min} \leq a_{ii} \leq \lambda_{\max}$ ,  $\lambda_{\min}$  and  $\lambda_{\max}$  being respectively the minimal and the maximal eigenvalue, this would imply  $\lambda_{\min} = 0$ , that is  $A$  is singular); for this reason we can define  $D = \text{diag}(d_{11}, d_{22}, \dots, d_{nn})$ , with  $d_{ii} = a_{ii}^{-\frac{1}{2}} \in \mathbb{R}$ , since all  $a_{ii} \in \mathbb{R}^+$ . Let's furthermore define  $B = DAD$ . It's easy to check that  $B$  too is Hermitian positive semidefinite, so its eigenvalues  $\lambda_B$  are all non-negative (actually, since  $A = A^H$  and since  $D$  is real and diagonal,  $B^H = (DAD)^H = D^H(DA)^H = D^H A^H D^H = DAD = B$ ; on the other hand, for any  $\mathbf{x} \neq \mathbf{0}$ ,  $\mathbf{x}^H B \mathbf{x} = \mathbf{x}^H DAD \mathbf{x} = (\mathbf{x}^H D)A(D\mathbf{x}) = (D^H \mathbf{x})^H A(D\mathbf{x}) = (D\mathbf{x})^H A(D\mathbf{x}) = \mathbf{y}^H A \mathbf{y} \geq 0$ ). Moreover, we have obviously  $b_{ii} = d_{ii} a_{ii} d_{ii} = 1$  so that  $\text{tr}(B) = n$  and, recalling the <http://planetmath.org/ArithmeticGeometricMeansInequality> geometric-arithmetic mean inequality, which holds in this case because the eigenvalues of  $B$  are all non-negative,

$$\det(B) = \prod_{i=1}^n \lambda_B^{(i)} \leq \left( \frac{1}{n} \sum_{i=1}^n \lambda_B^{(i)} \right)^n = \left( \frac{1}{n} \text{tr}(B) \right)^n = 1,$$

and since  $\det(B) = \det(D)^2 \det(A) = \left( \prod_{i=1}^n a_{ii} \right)^{-1} \det(A)$ , we have the thesis. Since  $\det(A) = \prod_{i=1}^n a_{ii}$  if and only if  $\det(B) = 1$  and since in the geometric-arithmetic inequality equality holds if and only if all terms are equal, we must have  $\lambda_B^{(i)} = \lambda_B$ , so that  $\prod_{i=1}^n \lambda_B^{(i)} = \lambda_B^n = 1$ , whence  $\lambda_B = 1$  ( $\lambda_B$  having to be non-negative), and since  $B$  is Hermitian and hence is diagonalizable, we obtain  $B = I$ , and so  $A = D^{-1} B D^{-1} = D^{-2} = \text{diag}(a_{11}, a_{22}, \dots, a_{nn})$ . So we can conclude that equality holds if and only if  $A$  is diagonal.

Let's now derive the more general first inequality. Let  $A$  be a complex-valued  $n \times n$  matrix. If  $A$  is singular, the thesis is trivially verified. Let's thus assume  $\det(A) \neq 0$ ; then  $B = AA^H$  is a Hermitian positive semidefinite matrix (actually,  $B^H = (AA^H)^H = (A^H)^H A^H = AA^H = B$  and, for any  $\mathbf{x} \neq \mathbf{0}$ ,  $\mathbf{x}^H B \mathbf{x} = \mathbf{x}^H AA^H \mathbf{x} = (\mathbf{x}^H A)(A^H \mathbf{x}) = (A^H \mathbf{x})^H (A^H \mathbf{x}) = \mathbf{y}^H \mathbf{y} = \|\mathbf{y}\|_2^2 \geq 0$ ). Therefore, the second inequality can be applied to  $B$ , yielding:

$$\begin{aligned} |\det(A)|^2 &= \det(A) \det^*(A) = \det(A) \det(A^H) = \det(AA^H) = \det(B) \\ &\leq \prod_{i=1}^n b_{ii} = \prod_{i=1}^n \sum_{j=1}^n a_{ij} (a^H)_{ji} = \prod_{i=1}^n \sum_{j=1}^n a_{ij} a_{ij}^* = \prod_{i=1}^n \sum_{j=1}^n |a_{ij}|^2. \end{aligned}$$

As we proved above, for  $\det(B)$  to be equal to  $\prod_{i=1}^n b_{ii}$ ,  $B$  must be diagonal, which means that  $\sum_{k=1}^n a_{ik} a_{jk}^* = |a_{ii}|^2 \delta_{ij}$ . So we can conclude that

equality holds if and only if the rows of  $A$  are orthogonal.  $\square$

## References

- [1] R. A. Horn, C. R. Johnson, *Matrix Analysis*, Cambridge University Press, 1985