## Some Linear Algebra Notes

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**Definition 1** A matrix A is positive semi-definite (PSD) iff for all vectors  $\vec{x}$ ,  $\vec{x}^T A \vec{x} \ge 0$ .

**Definition 2** The conjugate of a complex number z = a + bi is  $\overline{z} = a - bi$ .

**Definition 3** A square matrix A is Hermitian if  $A = A^H$ , that is  $A_{ij} = \overline{A}_{ji}$ . For real matrices, Hermitian and symmetric are equivalent (and  $A^H \equiv A^T$ ).

**Theorem 1** A matrix A is PSD iff  $\exists B$  such that  $A = B^H B$ .

**Theorem 2** A matrix A is PSD iff  $\exists$  a Hermitian matrix C such that  $A = C^2$ .

**Theorem 3** If A is PSD then  $A^{-1}$  exists and is PSD.

**Theorem 4** If A is PSD, then  $\forall$  integer  $k > 0 \exists$  a unique PSD matrix B with  $A = B^k$ . B also satisfies (1) AB = BA, (2) B = p(A) for some polynomial p, (3) rank(B) = rank(A), and (4) if A is real then so is B.

**Theorem 5** The matrix  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  is PSD iff  $b^2 < 4ac$ .

**Theorem 6** A real matrix A is PSD iff its symmetric part  $B = (A + A^T)/2$  is PSD. Indeed  $\vec{x}^T A \vec{x} = \vec{x}^T B \vec{x} \ \forall \vec{x}$ .

**Definition 4**  $\lambda$  is an eigenvalue of a square matrix A iff for some non-zero  $\vec{x}$ ,  $A\vec{x} = \lambda \vec{x}$ .  $\vec{x}$  is called an eigenvector with corresponding eigenvalue  $\lambda$ . We define  $\lambda_i(A)$  to be the *i*<sup>th</sup> largest eigenvalue of the matrix A.

**Definition 5** The trace of a square matrix is the sum of its diagonal elements,  $tr(A) = \sum_{i} A_{ii}$ .

**Theorem 7** Let A be square. Then  $tr(A) = \sum_i \lambda_i(A)$ .

**Theorem 8** The eigenvalues of a Hermitian matrix are all real.

**Definition 6** The spectral norm of a matrix A is the square root of the largest eigenvalue of  $A^H A$ ,  $||A||_2 = \sqrt{\lambda_1(A^H A)} = \max_{\|\vec{x}\| \neq 0} \frac{||A\vec{x}||_2}{\|\vec{x}\|_2}$ .

**Definition 7** Let  $||\vec{x}||$  be a vector norm. The induced matrix norm is  $||A|| = \max_{||\vec{x}||=1} ||A\vec{x}||$ . The spectral norm is the  $L_2$  induced matrix norm.