2-Norm

Proposition 2.7.5. If $A \in \mathbb{R}^{m \times n}$, then

$$\|A\|_2 = \sqrt{\max_{\mu \in \lambda(A^T A)} \mu},$$

i.e., $||A||_2$ is the square root of the largest eigenvalue of A^TA . *Proof.* To calculate $||A||_2$, we find first $||A||_2^2$. Thus,

$$\|A\|_{2} = \max_{\|\mathbf{x}\|_{2}=1} \|A\mathbf{x}\|_{2} \Rightarrow \|A\|_{2}^{2} = \max_{\|\mathbf{x}\|_{2}=1} \|A\mathbf{x}\|_{2}^{2} = \max_{\mathbf{x}^{T}\mathbf{x}=1} \mathbf{x}^{T} A^{T} A \mathbf{x}.$$

Let $A^T A = B \in \mathbf{R}^{n \times n}$. The matrix B is a symmetric matrix because

$$B^T = (A^T A)^T = A^T A$$

and

$$\mathbf{x}^{T} A^{T} A \mathbf{x} = \mathbf{x}^{T} B \mathbf{x} = \begin{bmatrix} \xi_{1} & \cdots & \xi_{n} \end{bmatrix} \begin{bmatrix} b_{1;1} & \cdots & b_{1;n} \\ \vdots & \cdots & \vdots \\ b_{n;1} & \cdots & b_{n;n} \end{bmatrix} \begin{bmatrix} \xi_{1} \\ \vdots \\ \xi_{n} \end{bmatrix} =$$
$$= \begin{bmatrix} \xi_{1} & \cdots & \xi_{n} \end{bmatrix} \begin{bmatrix} b_{1;1}\xi_{1} + \dots + b_{1;n}\xi_{n} \\ \cdots & \cdots \\ b_{n;1}\xi_{1} + \dots + b_{n;n}\xi_{n} \end{bmatrix} =$$
$$= \begin{bmatrix} \xi_{1} \sum_{j=1}^{n} b_{1;j}\xi_{j} + \dots + \xi_{n} \sum_{j=1}^{n} b_{n;j}\xi_{j} \end{bmatrix},$$
$$\mathbf{x}^{T} \mathbf{x} = \sum_{j=1}^{n} \xi_{j}^{2},$$

then $\mathbf{x}^T A^T A \mathbf{x}$ is a function of *n* variables ξ_1, \ldots, ξ_n and

$$\frac{\partial \left(\mathbf{x}^{T} A^{T} A \mathbf{x}\right)}{\partial \xi_{i}} = \sum_{j=1}^{n} b_{i; j} \xi_{j} + \sum_{j=1}^{n} b_{j; i} \xi_{j} \stackrel{b_{i; j}=b_{j; i}}{=} 2 \sum_{j=1}^{n} b_{i; j} \xi_{j} = 2 \left[A^{T} A \mathbf{x}\right]_{i},$$
$$\frac{\partial \left(\mathbf{x}^{T} \mathbf{x}\right)}{\partial \xi_{i}} = 2\xi_{i} = 2 \left[\mathbf{x}\right]_{i}.$$

The problem of finding $\max_{\mathbf{x}^T \mathbf{x} = 1} \mathbf{x}^T A^T A \mathbf{x}$ is a problem of finding the relative extremum. To solve our problem we form an auxiliary function

$$\Phi(\xi_1;\ldots,\xi_n;\rho) = \mathbf{x}^T A^T A \mathbf{x} + \mu \left(1 - \mathbf{x}^T \mathbf{x}\right) \,.$$

To find the stationary points of Φ , we form the system of equations:

$$\frac{\partial \Phi}{\partial \xi_i} = 0$$
 (i = 1 : n) $\wedge \frac{\partial \Phi}{\partial \rho} = 0$,

i.e.,

$$\begin{cases} 2 \left[A^T A \mathbf{x} \right]_i - 2\mu \left[\mathbf{x} \right]_i = 0 \ (i = 1 : n) \\ 1 - \mathbf{x}^T \mathbf{x} = \mathbf{0} \end{cases}$$

or

$$\begin{cases} A^T A \mathbf{x} = \mu \mathbf{x} ,\\ \|\mathbf{x}\|_2 = 1 . \end{cases}$$

Thus, any stationary point for relative extremum is the normed vector \mathbf{x} corresponding to an eigenvalue of $A^T A$. Let us express from the relation $A^T A \mathbf{x} = \mu \mathbf{x}$ the eigenvalue μ . We obtain that $\mu = \mathbf{x}^T A^T A \mathbf{x}$, where $\|\mathbf{x}\|_2 = 1$. Comparing this result with the original formula for finding $\|A\|_2^2$, we notice that $\|A\|_2^2 = \max_{\mu \in \lambda(A^T A)} \mu$. Thus,

$$||A||_2 = \sqrt{\max_{\mu \in \lambda(A^T A)} \mu},$$

i.e., $\|\,A\|_2$ is the square root of the largest eigenvalue of A^TA .

Corollary 2.7.1. If matrix $A \in \mathbb{R}^{m \times n}$ is symmetric, then

$$\|A\|_2 = \max_{\lambda \in \lambda(A)} \lambda.$$