

1. a. Since the space of unit vectors is compact, there exists x^0 such that $\|M\|_2 = \|Mx^0\|_2$. For the same reason, there exist unit vectors x°, y° such that

$$\langle x^\circ, My^\circ \rangle = \max_{\|x\|=\|y\|=1} \langle x, My \rangle$$

Letting $y^0 = Mx^0/\|M\|_2$, we have $\|y\|_2 = 1$. Therefore

$$\|M\|_2 = (y^0)^* Mx^0 \leq \max_{\|y\|=\|x\|=1} \langle y, Mx \rangle$$

Also, by Cauchy-Schwarz we have

$$\max_{\|x\|=\|y\|=1} \langle x, My \rangle = (x^\circ)^* My^\circ \leq \|x^\circ\|_2 \|My^\circ\|_2 \leq \|My^\circ\|_2 \leq \max_{\|y\|=1} \|My\|_2 = \|M\|_2$$

Having shown that

$$\|M\|_2 \leq \max_{\|x\|=\|y\|=1} \langle x, My \rangle \leq \|M\|_2$$

we conclude that the definitions are equivalent

b. Let the SVD of M be written $M = UDV^T$ where $D = \text{diag}(\sigma_1, \dots, \sigma_n)$. Let v_1 be the first column of V , then $\|v_1\|_2 = 1$ and

$$\|Mv_1\|_2 = \|UDV^T v_1\|_2 = \|UDe_1\|_2 = \|De_1\|_2 = \sigma_1$$

Hence

$$\sigma_1 = \|Mv_1\|_2 \leq \max_{\|x\|=1} \|Mx\|_2 = \|M\|_2$$

Meanwhile for any unit vector x , defining $y = V^T x$ we have $\|y\|_2 \leq 1$. Then

$$\max_{\|x\|=1} \|Mx\|_2 = \max_{\|x\|=1} \|UDV^T x\|_2 \leq \max_{\|x\|=1} \|DV^T x\|_2 \leq \max_{\|y\|=1} \|Dy\|_2$$

But defining $a_i = y_i^2$,

$$\max_{\|y\|=1} \|Dy\|_2^2 = \max_{\|y\|=1} \sum_{i=1}^n \sigma_i^2 y_i^2 = \max_{\sum a_i=1, a_i \geq 0} \sum_{i=1}^n \sigma_i^2 a_i$$

is maximized by $a = e_1$, hence $\max_{\|y\|=1} \|Dy\|_2 = \sigma_1$.

Having shown that

$$\sigma_1 \leq \|M\|_2 \leq \sigma_1$$

we conclude that the two definitions are equivalent.

2. a. From 1a we have

$$\|M^*\|_2 = \max_{\|x\|=\|y\|=1} \langle x, M^*y \rangle = \max_{\|x\|=\|y\|=1} \langle y, Mx \rangle = \|M\|_2$$

b. We have

$$\|AB\|_2 = \max_{\|x\|=1} \|ABx\|_2 = \max_{y=Bx \text{ for some } \|x\|=1} \|Ay\|_2$$

Meanwhile, if $y = Bx$, and $\|x\|_2 = 1$, we have $\|y\|_2 \leq \|B\|_2$. Therefore the set $\{y : y = Bx \text{ for some } x \text{ such that } \|x\| = 1\}$ is contained in the set $\{y : \|y\|_2 \leq \|B\|_2\}$. Hence

$$\max_{y=Bx \text{ for some } \|x\|=1} \|Ay\|_2 \leq \max_{\|y\|=\|B\|_2} \|Ay\|_2 = \|A\|_2 \|B\|_2$$

3. i.

$$\|aM\|_2 = \max_{\|x\|=1} \|aMx\|_2 = \max_{\|x\|=1} |a| \|Mx\|_2 = |a| \max_{\|x\|=1} \|Mx\|_2 = a \|M\|_2$$

ii.

$$\|A+B\|_2 = \max_{\|x\|=1} \|Ax+Bx\|_2 \leq \max_{\|x\|=1} \|Ax\|_2 + \|Bx\|_2 \leq \max_{\|x\|=1} \|Ax\|_2 + \max_{\|x\|=1} \|Bx\|_2 = \|A\|_2 + \|B\|_2$$

iii. Proof of contrapositive: If $M \neq 0$, then some column M_i is nonzero. But then $\|Me_i\|_2 = \|M_i\|_2 > 0$, so $\|M\|_2 > 0$.

4. a. Let m_i denote the columns of M^* .

$$\|M\|_2 = \|M^*\|_2 = \max_{\|x\|=1} \|M^*x\|_2 = \left\| \sum_{i=1}^m x_i m_i \right\|_2 \leq \sum_{i=1}^m |x_i| \|m_i\|_2$$

Now let μ be the vector defined by $\mu_i = \|m_i\|_2$. By Hölder's inequality, we have

$$\|M\|_2 \leq \max_{\|x\|_2=1} \langle x, \mu \rangle \leq \max_{\|x\|_2=1} \|x\|_1 \|\mu\|_\infty$$

But $\max_{\|x\|_2=1} \|x\|_1 = \sqrt{m}$ and $\|\mu\|_\infty = \max_i \|m_i\|_2$. Hence

$$\|M\|_2 \leq \sqrt{m} \max_i \|m_i\|_2$$

as needed.

We see that the upper bound is tight by taking $M = 1_m 1_n^T$, in which case $m_1 = \dots = m_m = 1_n$

$$\|M\|_2 = \|M \frac{1}{\sqrt{n}} 1_n\|_2 = \|\sqrt{n} 1_m\|_2 = \sqrt{m} \|1_n\|_2 = \sqrt{m} \max_i \|m_i\|_2$$

b. Using the fact that $\|x\|_2 \geq \frac{1}{\sqrt{m}} \|x\|_1$ for any vector $x \in \mathbb{R}^m$, we have

$$\|M\|_2 \geq \|M \frac{1}{\sqrt{n}} 1_n\|_2 = \|\mu\|_2 \geq \frac{1}{\sqrt{m}} \|\mu\|_1 = \frac{1}{\sqrt{mn}} \sum_{i=1}^m |m_i^* 1_n|$$

where μ is the m -vector with $m_i = \frac{1}{\sqrt{n}} m_i^* 1_n$. Hence the upper bound is proved.

To show that the upper bound is tight, take $M = 1_m 1_n^T$, in which case

$$\|M\|_2 = \sqrt{mn} = \frac{1}{\sqrt{mn}} mn = \frac{1}{\sqrt{mn}} \sum_{i=1}^m |m_i^* 1_n|$$

5. a. We have

$$\|M\|_F = \sqrt{\sum_i \sum_j M_{ij}^2} = \sqrt{\text{tr} M^T M} = \sqrt{\text{tr} M^T M}$$

Without loss of generality assume $n \leq m$. Since the eigenvalues of $M^T M$ are $\sigma_1^2, \dots, \sigma_n^2$, we have

$$\text{tr} M^T M = \sum_{i=1}^n \sigma_i^2$$

hence dropping the assumption that $n \leq m$, we have

$$\|M\|_F = \sqrt{\sum_{i=1}^{\min(n,m)} \sigma_i^2}$$

b. Only $\sigma_1, \dots, \sigma_r$ are nonzero, where $r = \text{rank}(M)$. Letting $d = (\sigma_1, \dots, \sigma_r)$, we have $\|M\|_F = \|d\|_2$. But we know that for all vectors $x \in \mathbb{R}^r$,

$$\|x\|_\infty \leq \|x\|_2 = \sqrt{\sum_{i=1}^r x_i^2} \leq \sqrt{\sum_{i=1}^r \|x\|_\infty^2} = \sqrt{r} \|x\|_\infty$$

Hence

$$\|M\|_2 = \sigma_1 = \|d\|_\infty \leq \|d\|_2 = \|M\|_F = \|d\|_2 \leq \sqrt{r} \|d\|_\infty = \sqrt{r} \|M\|_2$$

as needed.