Matrix Calculation: Homework #2

Due on Sept 26, 2022 at 3:10pm

Professor Jun Lai Monday

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Problem 1

(Page 80, Problem 2.4.2)

Proof. By SVD decomposition, there exists orthogonal matrix U, V such that

$$U^{t}AV = \begin{bmatrix} D & O \\ O & O \end{bmatrix}, D = \operatorname{diag}(\sigma_{1}, \cdots, \sigma_{r}).$$
(1)

WLOG, we set $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_r$, and assume that $U = [y_1, \cdots, y_m], V = [x_1, \cdots, x_n]$, it means that:

$$y_i^t A x_j = \begin{cases} \sigma_i \ (i = j, i, j \le r) \\ 0 \ (otherwise). \end{cases}$$
 (2)

where $||y_i|| = ||x_i|| = 1$. As $\{y_i\}$, $\{x_j\}$ are orthonormal basis, respectively, we can write:

$$x = \sum_{i=1}^{n} \beta_i x_i$$

$$y = \sum_{i=1}^{m} \alpha_i y_i$$
(3)

Then:

$$y^{t}Ax = \sum \alpha_{i}\beta_{i}\sigma_{i}$$

$$\leq \sigma_{\max} \sum |\alpha_{i}\beta_{i}|$$

$$\leq \sigma_{\max} \sqrt{\sum \alpha_{i}^{2} \sum \beta_{i}^{2}}$$

$$= \sigma_{\max} ||x||_{2} ||y||_{2}.$$

$$(4)$$

It means that

$$\sigma_{\max} \ge \max \frac{y^t A x}{\|x\|_2 \|y\|_2}.\tag{5}$$

On the other hand, set $x = x_1$, $y = y_1$, we can see $\frac{y^t A x}{\|x\|_2 \|y\|_2} = \sigma_1 = \sigma_{\text{max}}$. Q.E.D.

Problem 2

(Page 80, Problem 2.4.6)

By corollary 2.4.7, $A = \sigma_1 u_1 v_1^t + \sigma_2 u_2 v_2^t$. In one hand, set $B = \sigma_1 u_1 v_1^t$, we can get $||A - B||_F = \sigma_2$. In the other hand, assume r(B) = 1, then dim ker B = 1, assume that Bx = 0. Expand the vector x to an orthonormal basis for \mathbb{R}^2 as $\{x, y\}$, then:

$$||A - B||_F^2 \ge ||(A - B)x||_2^2$$

$$= ||Ax||_2^2$$

$$= \sigma_1^2 (v_1^t x)^2 + \sigma_2^2 (v_2^t x)^2$$

$$\ge \sigma_2^2.$$
(6)

So the nearest rank-1 matrix $B = \sigma_1 u_1 v_1^t$.

Problem 3

(Page 80, Problem 2.4.7)

Proof. By corollary 2.4.3, we can see:

$$||A||_F = \sqrt{\sum_{i=1}^r \sigma_i^2} \tag{7}$$

$$||A||_2 = \sigma_i.$$

As
$$\sum_{i=1}^r \sigma_i^2 \le r\sigma_1^2$$
, it means that $||A||_F \le \sqrt{r}\sigma_1 = \sqrt{r}||A||_2$.