

MA 522 Solution for Homework Assignment #2
(Based on the solutions of Megan Dailey)

Problem 1:

- (1) The trace of a matrix $A = [a_{ij}]$ is defined as the sum of diagonal entries, i.e. $\text{tr}(A) = \sum_{i=1}^n a_{ii}$. Verify that $\|A\|_F^2 = \text{tr}(A^T A)$.
- (2) If U, V are orthogonal, prove $\|UAV\|_F = \|A\|_F$ and $\|UAF\|_2 = \|A\|_2$.

Proof:

- (1) Recall $(BA)_{ij} = \sum_{k=1}^m b_{ik}a_{kj}$ where $1 \leq i \leq p, 1 \leq j \leq n$ with $B = [b_{ij}] \in \mathbb{R}^{p \times m}, A = [a_{ij}] \in \mathbb{R}^{m \times n}$. Thus,

$$\begin{aligned}\text{tr}(BA) &= \sum_{i=1}^n (BA)_{ii} \\ &= \sum_{i=1}^n \sum_{k=1}^m b_{ik}a_{ki}\end{aligned}$$

Thus, if $B = A^T$, i.e. $b_{ik} = a_{ki}$, then

$$\begin{aligned}\text{tr}(A^T A) &= \sum_{i=1}^n \sum_{k=1}^m a_{ki}a_{ki} \\ &= \sum_{i=1}^n \sum_{k=1}^m a_{ki}^2 \\ &= \|A\|_F^2\end{aligned}$$

- (2) Recall for an orthogonal matrix U ,
- (a) $U^T = U^{-1}$
 - (b) $\|U\mathbf{x}\|_2 = \|\mathbf{x}\|_2$

Also, note

$$\begin{aligned}\text{tr}(AB) &= \sum_{k=1}^m \sum_{i=1}^n a_{ki}b_{ik} \\ &= \sum_{i=1}^n \sum_{k=1}^m b_{ik}a_{ki} \\ &= \text{tr}(BA)\end{aligned}$$

Now, consider

$$\begin{aligned}
\|UAV\|_F^2 &= \text{tr} \left((UAV)^T (UAV) \right) && \text{by (1)} \\
&= \text{tr} (V^T A^T U^T U AV) \\
&= \text{tr} (V^T A^T AV) && \text{by (a)} \\
&= \text{tr} ((A^T AV)V^T) && \text{by } \text{tr}(AB) = \text{tr}(BA) \\
&= \text{tr} (A^T A) && \text{by (a)} \\
&= \|A\|_F^2 && \text{by (1)}
\end{aligned}$$

And,

$$\begin{aligned}
\|UAV\|_2 &= \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|UAV\mathbf{x}\|_2}{\|\mathbf{x}\|_2} && \text{by definition} \\
&= \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|U(AV\mathbf{x})\|_2}{\|\mathbf{x}\|_2} \\
&= \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|AV\mathbf{x}\|_2}{\|\mathbf{x}\|_2} && \text{by (b)} \\
&= \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|AV\mathbf{x}\|_2}{\|V\mathbf{x}\|_2} && \text{by (b)} \\
&= \max_{\mathbf{y} \neq \mathbf{0}} \frac{\|A\mathbf{y}\|_2}{\|\mathbf{y}\|_2} && \text{where } \mathbf{y} = A\mathbf{x} \neq \mathbf{0} \text{ if } \mathbf{x} \neq \mathbf{0} \\
&= \|A\|_2 && \text{by definition}
\end{aligned}$$

□

Problem 2: 1.7

Verify that $\|xy^H\|_F = \|xy^H\|_2 = \|x\|_2 \|y\|_2$ for any $x, y \in \mathbb{C}^n$.

Proof:

Consider:

$$\begin{aligned}
\|xy^H\|_2 &= \max_{\mathbf{z} \neq \mathbf{0}} \frac{\|(\mathbf{x}\mathbf{y}^H)\mathbf{z}\|_2}{\|\mathbf{z}\|_2} \\
&= \max_{\mathbf{z} \neq \mathbf{0}} \frac{\|\mathbf{x}(\mathbf{y}^H\mathbf{z})\|_2}{\|\mathbf{z}\|_2} \\
&= \max_{\mathbf{z} \neq \mathbf{0}} \frac{\|\mathbf{x} \langle \bar{\mathbf{y}}, \mathbf{z} \rangle\|_2}{\|\mathbf{z}\|_2} \\
&= \max_{\mathbf{z} \neq \mathbf{0}} \frac{\langle \bar{\mathbf{y}}, \mathbf{z} \rangle \|\mathbf{x}\|_2}{\|\mathbf{z}\|_2} \\
&\leq \max_{\mathbf{z} \neq \mathbf{0}} \frac{\|\bar{\mathbf{y}}\|_2 \|\mathbf{z}\|_2 \|\mathbf{x}\|_2}{\|\mathbf{z}\|_2} && \text{by Cauchy Schwarz} \\
&= \max_{\mathbf{z} \neq \mathbf{0}} \|\bar{\mathbf{y}}\|_2 \|\mathbf{x}\|_2 \\
&= \|\bar{\mathbf{y}}\|_2 \|\mathbf{x}\|_2 && \text{since } \|\bar{y}_i\| = |y_i|
\end{aligned}$$

Now let $\mathbf{z} = \frac{\bar{\mathbf{y}}}{\|\bar{\mathbf{y}}\|_2}$. Then, $\|\mathbf{z}\|_2 = 1$ and

$$\begin{aligned} |\langle \bar{\mathbf{y}}, \mathbf{z} \rangle| &= \left| \left\langle \bar{\mathbf{y}}, \frac{\bar{\mathbf{y}}}{\|\bar{\mathbf{y}}\|_2} \right\rangle \right| \\ &= \frac{|\langle \bar{\mathbf{y}}, \bar{\mathbf{y}} \rangle|}{\|\bar{\mathbf{y}}\|_2} \\ &= \frac{\|\bar{\mathbf{y}}\|_2^2}{\|\bar{\mathbf{y}}\|_2} \\ &= \|\bar{\mathbf{y}}\|_2 \end{aligned}$$

Thus,

$$\begin{aligned} \|xy^H\|_2 &= \max_{\mathbf{z} \neq \mathbf{0}} \frac{|\langle \bar{\mathbf{y}}, \mathbf{z} \rangle| \|\mathbf{x}\|_2}{\|\mathbf{z}\|_2} \\ &\geq \frac{\|\bar{\mathbf{y}}\|_2 \|\mathbf{x}\|_2}{\|\mathbf{z}\|_2} && \text{from above} \\ &= \|\bar{\mathbf{y}}\|_2 \|\mathbf{x}\|_2 && \text{since } \|\mathbf{z}\|_2 = 1 \\ &= \|\mathbf{y}\|_2 \|\mathbf{x}\|_2 \end{aligned}$$

Hence, we have

$$\|\mathbf{x}\|_2 \|\mathbf{y}\|_2 \leq \|\mathbf{xy}^H\|_2 \leq \|\mathbf{x}\|_2 \|\mathbf{y}\|_2$$

Which implies,

$$\|\mathbf{xy}^H\|_2 = \|\mathbf{x}\|_2 \|\mathbf{y}\|_2$$

Consider now

$$\begin{aligned} \|\mathbf{xy}^H\|_F^2 &= \sum_{j=1}^n \sum_{i=1}^n (x_i \bar{y}_j)^2 \\ &= \sum_{j=1}^n \sum_{i=1}^n x_i^2 \bar{y}_j^2 \\ &= \sum_{i=1}^n x_i^2 \cdot \sum_{j=1}^n \bar{y}_j^2 \\ &= \left(\sum_{i=1}^n x_i^2 \right) \left(\sum_{j=1}^n \bar{y}_j^2 \right) \\ &= \|\mathbf{x}\|_2^2 \|\bar{\mathbf{y}}\|_2^2 \end{aligned}$$

Hence,

$$\begin{aligned} \|\mathbf{xy}^H\|_F &= \|\mathbf{x}\|_2 \|\bar{\mathbf{y}}\|_2 \\ &= \|\mathbf{x}\|_2 \|\mathbf{y}\|_2 \end{aligned}$$

Thus,

$$\|\mathbf{xy}^H\|_2 = \|\mathbf{x}\|_2 \|\mathbf{y}\|_2 = \|\mathbf{xy}^H\|_F$$

□

Problem 3: Let k be a positive integer and $X \in \mathbb{R}^{n \times n}$ be such that $X^k = \mathbf{0}$. Prove that $I - X$ is invertible and

$$(I - X)^{-1} = \sum_{i=0}^{k-1} X^i$$

Proof:
Since

$$\begin{aligned} (I - X) \cdot \sum_{i=0}^{k-1} X^i &= \sum_{i=0}^{k-1} X^i - X \cdot \sum_{i=0}^{k-1} X^i \\ &= \sum_{i=0}^{k-1} X^i - \sum_{i=0}^{k-1} X^{i+1} \\ &= X^0 - X^k \\ &= I - \mathbf{0} && \text{by definition and assumption} \\ &= I \end{aligned}$$

Hence, $(I - X)$ is invertible and moreover $(I - X)^{-1} = \sum_{i=0}^{k-1} X^i$.

□

Problem 4: Show that if $\|E\| < 1/2$, then

$$\|(I - E)^{-1} - (I + E)\| \leq 2\|E\|^2$$

where $\|\cdot\|$ is any matrix operator norm.

Proof:
Remarks:

(a) For $|r| < 1$

$$\sum_{i=0}^{\infty} r^i = \frac{1}{1-r}$$

So,

$$\begin{aligned} \sum_{i=2}^{\infty} r^i &= \sum_{i=0}^{\infty} r^i - 1 - r \\ &= \frac{1}{1-r} - (1+r) \\ &= \frac{1}{1-r} - \frac{(1+r)(1-r)}{1-r} \\ &= \frac{r^2}{1-r} \end{aligned}$$

(b) Consider

$$\begin{aligned}\|E\| &< \frac{1}{2} \\ -\|E\| &> -\frac{1}{2} \\ 1 - \|E\| &> \frac{1}{2} \\ \frac{1}{1 - \|E\|} &< 2\end{aligned}$$

Recall from class:

$$(I - E)^{-1} = \sum_{i=0}^{\infty} E^i$$

Thus, we have

$$(I - E)^{-1} - (I - E) = \sum_{i=2}^{\infty} E^i$$

Hence,

$$\begin{aligned}\|(I - E)^{-1} - (I - E)\| &= \left\| \sum_{i=2}^{\infty} E^i \right\| \\ &\leq \sum_{i=2}^{\infty} \|E^i\| \\ &= \frac{\|E\|^2}{1 - \|E\|} \\ &\leq 2\|E\|^2\end{aligned}$$

by the triangle inequality

by Remark (a)

by Remark (b)

□