#### Problem 0: Homework checklist

✓I didn't talk with any one about this homework.

✓ Source-code are included at the end of this document.

# Problem 1: Prove or disprove

For the proof below, I will assume all the matrices are  $n \times n$  square matrices.

1. The product of two diagonal matrices is diagonal.

 $\boldsymbol{A}$  and  $\boldsymbol{B}$  are diagonal matrices, so

$$\mathbf{A}_{ij} = 0, if \ i \neq j$$

$$\boldsymbol{B}_{ij} = 0, if \ i \neq j$$

$$C = A \times B$$

$$C_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj}$$

 $C_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj}$ Only if i = j = k,  $C_{ij}$  will not be zero, which mean C is also a diagonal matrix.

2. The product of two upper triangular matrices is upper triangular A and Bare two upper triangular matrices, so

$$\mathbf{A}_{ij} = 0, if \ i \geqslant j$$

$$egin{aligned} m{B}_{ij} &= 0, if \ i \geqslant j \ m{C} &= m{A} imes m{B} \end{aligned}$$

$$C = A \times B$$

$$C_{ij} = \sum_{k=1}^{n} A_{ik} B_{ki}$$

 $C_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj}$ When  $i \geqslant j$ ,  $C_{ij} = 0$  because one of  $A_{ik}$  and  $B_{kj}$  must be zero.

3. The product of two symmetric matrices is symmetric. A and B are two symmetric matrices, so

$$oldsymbol{A}_{ij} = oldsymbol{A}_{ji}$$

$$B_{ij} = B_{ji}$$

$$C = A \times B$$

$$C_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj}$$

 $m{C}_{ij} = \sum_{k=1}^n m{A}_{ik} m{B}_{kj}$ Then  $m{C}_{ij}$  is not necessary to be equal to  $m{C}_{ji}$ 

Counter-example:

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$$
 is a symmetric matrix

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$$
 is a symmetric matrix;  
 $B = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  is also a symmetric matrix.

$$C = A \times B = \begin{bmatrix} 2 & 2 \\ 3 & 3 \end{bmatrix}$$
 which is not a symmetric matrices.

4. The product of two orthogonal matrices is orthogonal. A and B are two orthogonal matrices, so

$$A^T A = AA^T = I$$
 $B^T B = BB^T = I$ 

$$\mathbf{B}^T \mathbf{B} = \mathbf{B} \mathbf{B}^T = \mathbf{I}$$

Then 
$$(AB)(AB)^T = ABB^TA^T = AA^T = I$$

So AB is also an orthogonal matrix.

5. The product of two square, full rank matrices is full rank This statement is incorrect.

Counter-example:

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$$
 is a full rank matrix;  
 $B = \begin{bmatrix} 3 & 5 \\ 1 & 1 \end{bmatrix}$  is also a full rank matrix.  
 $C = A \times B = \begin{bmatrix} 4 & 6 \\ 2 & 3 \end{bmatrix}$  which is not a full matrices.

# Problem 2

We know U and V are square orthogonal matrices, then

$$\boldsymbol{U}^T \boldsymbol{U} = \boldsymbol{I}, \boldsymbol{V}^T \boldsymbol{V} = \boldsymbol{I} \tag{1}$$

$$||UAV||_{2}^{2} = V^{T}A^{T}U^{T}UAV$$

$$= V^{T}A^{T}AV$$

$$= V^{T}||A||_{2}^{2}V$$

$$= V^{T}V||A||_{2}^{2}$$

$$= ||A||_{2}^{2}$$

Therefore  $||A||_2$  is orthogonally invariant.

### Problem 3

$$f(\mathbf{A}) = \max_{i,j} |A_{ij}| \tag{2}$$

1. 
$$\bullet$$
  $f(A) > 0, when \mathbf{A} \neq 0,$   
 $f(A) = 0 \ iff \mathbf{A} = 0$ 

• For and scalar 
$$k$$
,  

$$f(k\mathbf{A})$$

$$= \max_{i,j} |kA_{ij}|$$

$$= k \max_{i,j} |A_{ij}|$$

$$= kf(\mathbf{A})$$

• 
$$f(A + B)$$
  
=  $\max_{i,j} |A_{ij} + B_{ij}|$   
 $\leq \max_{i,j} |A_{ij}| + \max_{i,j} |B_{ij}|$   
=  $f(A) + f(B)$ 

Therefore  $f(\mathbf{A})$  is a matrix norm.

2. 
$$f(\mathbf{AB})$$
  
 $= \max_{i,j} |A_{ij}B_{ij}|$   
 $\leq \max_{i,j} |A_{ij}| \max_{i,j} |B_{ij}|$   
 $= f(\mathbf{A})f(\mathbf{B})$   
The  $\leq$  is not necessary to

The  $\leq$  is not necessary to be. A counter example:

$$A = \begin{bmatrix} 3 & 3 \\ 3 & 3 \end{bmatrix}$$

$$B = \begin{bmatrix} 3 & 3 \\ 3 & 3 \end{bmatrix}$$

$$AB = \begin{bmatrix} 18 & 18 \\ 18 & 18 \end{bmatrix}$$

$$f(A) = 3$$

$$f(B) = 3$$

$$f(\boldsymbol{A}\boldsymbol{B}) = 18$$
  
Then  $f(\boldsymbol{A}\boldsymbol{B}) > f(\boldsymbol{A})f(\boldsymbol{B})$   
So  $f$  does not satisfy the sub-multiplicative property.

3.

$$g(\mathbf{A}) = \sigma f(\mathbf{A}) \tag{3}$$

$$g(\mathbf{AB}) = \sigma \max_{i,j} |A_{ij}B_{ij}|$$
  

$$g(\mathbf{A})g(\mathbf{B}) = \sigma^2 \max_{i,j} |A_{ij}| \max_{i,j} |B_{ij}|$$

To make

$$g(\mathbf{A}\mathbf{B}) \le g(\mathbf{A})g(\mathbf{B}) \tag{4}$$

$$\sigma \max_{i,j} |A_{ij}B_{ij}| \le \sigma^2 \max_{i,j} |A_{ij}| \max_{i,j} |B_{ij}| \tag{5}$$

$$\sigma \max_{i,j} |A_{ij}B_{ij}| \le \sigma^2 \max_{i,j} |A_{ij}| \max_{i,j} |B_{ij}|$$

$$\sigma \ge \frac{\max_{i,j} |A_{ij}B_{ij}|}{\max_{i,j} |A_{ij}| \max_{i,j} |B_{ij}|}$$
(6)

Therefore there exists  $\sigma > 0$  such that:  $g(\mathbf{A}) = \sigma f(\mathbf{A})$ 

#### Problem 4

$$f(x) = ||x\mathbf{k}^T|| \tag{7}$$

- f(x) > 0, when**A** $\neq 0,$ f(x) = 0 if  $f\mathbf{A} = 0$  because  $||\mathbf{A}||$  is a matrix norm.
  - $f(mx) = ||mx\mathbf{k}^T|| = m||x\mathbf{k}^T|| = mf(x)$
  - $f(x+y) = \|(x+y)\mathbf{k}^T\| = \|x\mathbf{k}^T + y\mathbf{k}^T\| \le \|x\mathbf{k}^T\| + \|y\mathbf{k}^T\| = f(x) + f(y)$

Therefore f is a vector norm.

2. If  $\|A\|$  is a sub-multiplicative matrix norm,

$$f(\mathbf{A}x) = \|\mathbf{A}x\mathbf{k}^T\| \le \|\mathbf{A}\| \|x\mathbf{k}^T\| = \|\mathbf{A}\| f(x)$$
(8)

# Problem 5 (Choice 1)

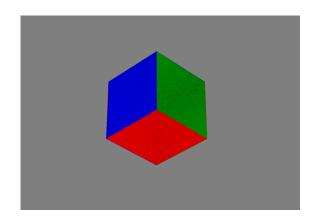
 $f(\mathbf{x}) = \text{sum of two largest entries in } \mathbf{x} \text{ by absolute value is a vector form.}$ Suppose

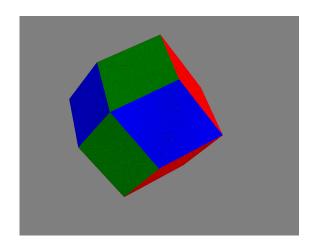
$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

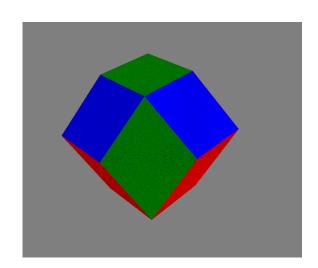
Then the unit-ball should include all points which satisfy conditions:

$$|x_1| + |x_2| = 1, |x_3| \le |x_1|, |x_3| \le |x_2|$$
  
or  $|x_2| + |x_3| = 1, |x_1| \le |x_2|, |x_3| \le |x_3|$   
or  $|x_3| + |x_1| = 1, |x_2| \le |x_3|, |x_3| \le |x_1|$ 

I used Monte Carlo method to generate N = 300000 points which satisfy those conditions, as shown ind Figure 1.







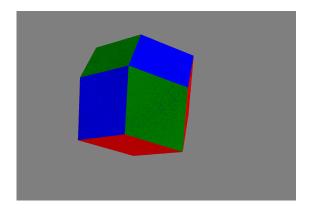


Figure 1: The unit-ball surface from different views. Color images are in electronic version.