# MMAT5320 Computational Mathematics

### Assignment 1

Due date: 10th October 2019

Please hand in your assignments to the assignment box on 2/F Lady Shaw Building (opposite the administration office and underneath the notice boards) by 6pm on **Thursday 10th October 2019**. Remember to include your name, ID number and show all of your working!

### Question 1

Calculate the rank, range and nullspace of the following matrices:

$$A_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad A_2 = \begin{pmatrix} 7 & 0 & 1 \\ 3 & 0 & 0 \\ 0 & 0 & 2 \end{pmatrix}, \quad A_3 = \begin{pmatrix} a_1 a_1 & a_1 a_2 & a_1 a_3 \\ a_2 a_1 & a_2 a_2 & a_2 a_3 \\ a_3 a_1 & a_3 a_2 & a_3 a_3 \end{pmatrix}$$

where  $a_1, a_2, a_3$  are non-zero real numbers.

#### 0.1 Solutions

The rank, range and nullspace of  $A_1$  are [1 point each]

$$\operatorname{rank}(A_1) = 1$$
,  $\operatorname{range}(A_1) = \{(x, 0, 0) : x \in \mathbb{R}\}$ ,  $\operatorname{null}(A_1) = \{(0, y, z) : y, z \in \mathbb{R}\}$ .

For  $A_2$  [1 point each]

$$\operatorname{rank}(A_2) = 2$$
,  $\operatorname{range}(A_2) = \{ (7x + z, 3x, 2z) : x, z \in \mathbb{R} \}$ ,  $\operatorname{null}(A_2) = \{ (0, y, 0) : y \in \mathbb{R} \}$ .

For  $A_3$  [1 point for rank, 2 points each for range and nullspace]

$$\operatorname{rank}(A_3) = 1$$
,  $\operatorname{range}(A_3) = \operatorname{span}\{(a_1, a_2, a_3)\}$ ,  $\operatorname{null}(A_3) = \operatorname{span}\{(\frac{1}{a_1}, 0, -\frac{1}{a_3}), (0, \frac{1}{a_2}, -\frac{1}{a_3})\}$ .

[Total 11 points]

## Question 2

(i) Fix any two real numbers p < q, prove that for any  $x \in \mathbb{R}^m$ ,  $m \ge 1$ , the following inequality for vector norms holds

$$||x||_q \le ||x||_p.$$

**Hint:** you can use without proof the following property: if  $(a_i)_{i=1,...,m}$  are nonnegative constants and  $s \ge 1$ , then

$$\sum_{i=1}^{m} a_i^s \le \left(\sum_{i=1}^{m} a_i\right)^s.$$

(ii) Give a counterexample to the claim

$$||x||_2 > ||x||_1$$
.

What modification to the left-hand side is needed to make the inequality true?

- (iii) For a square matrix  $A \in \mathbb{R}^{m \times m}$ , we define a function  $||A||_* := \max_{1 \le i,j \le n} |a_{ij}|$ . Show that this is a norm.
- (iv) Give an example of a matrix  $A \in \mathbb{R}^{2\times 2}$  where  $||A^2||_* > ||A||_*^2$ , and explain why this shows that  $||\cdot||_*$  is not an induced p-matrix norm.
- (v) For the matrix

$$A = \begin{pmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$$

calculate  $||A||_2$  and  $||A||_4$ .

#### 0.2 Solutions

(i) Set  $y = \frac{x}{\|x\|_p}$ , then  $\|y\|_p = 1$ , which means  $\|y\|_p^p = \sum_{i=1}^m |y_i|^p = 1$  [1 point]. Moreover,

$$\sum_{i=1}^{m} |y_i|^q = \sum_{i=1}^{m} |y_i|^{p^{\frac{q}{p}}} \le \left(\sum_{i=1}^{m} |y_i|^p\right)^{\frac{q}{p}} = 1 \quad [3 \text{ points: 1 point each for } =, \le, =]$$

since q/p > 1. This shows  $||y||_q \le 1$ , and so

$$||y||_q = \frac{||x||_q}{||x||_p} \le 1 \implies ||x||_q \le ||x||_p.$$

[2 points: 1 for  $||y||_q \le 1$  and 1 for conclusion].

(ii) Counterexample (any would suffices), e.g. x = (1, 1, 0). [2 full points awarded to calculations showing  $||x||_2 \le ||x||_1$ , otherwise just 1 point for stating the counter example]

Modifications would be  $m||x||_2 > ||x||_1$  for  $x \in \mathbb{R}^m$ . [1 point]

- (iii) [1 point each for the three requirements of a norm, total 3 points]
- (iv) Example

$$A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

then calculations show [2 points: 1 point each for computing matrix norms]  $||A^2||_* = 2 > ||A||_*^2 = 1$ .

If  $\|\cdot\|_*$  is a induced *p*-matrix norm, then from Lemma in Lecture notes slide (Inequalities I) we must have

$$||A^2||_* \le ||A||_*^2$$

but the counterexample shows this cannot be true. [2 points: 1 for applying Lemma, 1 for deducing contradiction]

2

(v) For  $x \in \mathbb{R}^2$ ,  $Ax = (2x_2, 0, 0)$ . Then,

$$||Ax||_2 = 2|x_2|$$
,  $||Ax||_4 = 2|x_2|$ . [2 points: 1 for each]

So

$$\|A\|_2 = \sup_{x_1^2 + x_2^2 = 1} 2|x_2| = 2, \quad \|A\|_4 = \sup_{x_1^4 + x_2^4 = 1} 2|x_2| = 2 \quad [\text{2 points: 1 for each}]$$

[Total 20 points]

### Question 3

(i) Compute a full SVD for the following matrices:

$$A_1 = \begin{pmatrix} 2 & 1 \end{pmatrix}, \quad A_2 = \begin{pmatrix} 1 & i \\ 0 & 0 \end{pmatrix}$$

where  $i = \sqrt{-1}$  is the imaginary unit. Also write down a reduced SVD for each matrix.

- (ii) Let  $A \in \mathbb{C}^{m \times m}$  be a hermitian and unitary matrix, with  $A = U \Sigma V^*$  be a SVD of A. Show that
  - all eigenvalues of A are 1 or -1;
  - all singular values of A are 1.

#### 0.3 Solutions

(i) Compute [1 point]

$$A_1^{\mathsf{T}} A_1 = \begin{pmatrix} 4 & 2 \\ 2 & 1 \end{pmatrix}$$

Eigenvalues are roots of characteristic polynomial  $x^2 - 5x = 0$  implies  $\sigma_1^2 = 5$  and  $\sigma_2^2 = 0$  [2 points]. Eigenvectors for  $\sigma_1^2$  and  $\sigma_2^2$  with unit length are [2 points]

$$v_1 = \begin{pmatrix} 2/\sqrt{5} \\ 1/\sqrt{5} \end{pmatrix} \quad v_2 = \begin{pmatrix} 1/\sqrt{5} \\ -2/\sqrt{5} \end{pmatrix}$$

Then,  $u_1 = \frac{1}{\sigma_1} A_1 v_1 = 1$  [1 point], and the full SVD is [1 point]

$$A_1 = \underbrace{1}_{U} \underbrace{\left(\sqrt{5} \quad 0\right)}_{\Sigma} \underbrace{\frac{1}{\sqrt{5}} \begin{pmatrix} 2 & 1\\ 1 & -2 \end{pmatrix}}_{V^{\mathsf{T}}}.$$

A reduced SVD is [1 point]

$$A_1 = \underbrace{\begin{pmatrix} 1 & 0 \end{pmatrix}}_{\hat{U}} \underbrace{\begin{pmatrix} \sqrt{5} & 0 \\ 0 & 0 \end{pmatrix}}_{\hat{\Sigma}} \underbrace{\frac{1}{\sqrt{5}} \begin{pmatrix} 2 & 1 \\ 1 & -2 \end{pmatrix}}_{V^{\mathsf{T}}}.$$

For  $A_2$ , compute [1 point]

$$A_2^* A_2 = \begin{pmatrix} 1 & i \\ -i & 1 \end{pmatrix}$$

Eigenvalues are roots of characteristic polynomial  $x^2 - 2x = 0$  implies  $\sigma_1^2 = 2$  and  $\sigma_2^2 = 0$  [2 points]. Eigenvectors with unit length are [2 points]

$$v_1 = \begin{pmatrix} 1/\sqrt{2} \\ -i/\sqrt{2} \end{pmatrix}, \quad v_2 = \begin{pmatrix} 1/\sqrt{2} \\ i/\sqrt{2} \end{pmatrix}$$

Then,

$$u_1 = \frac{1}{\sigma_1} A v_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

and we take  $u_2 = (0,1)^{\mathsf{T}}$  [2 points]. Then, the full SVD is [1 point]

$$A_2 = \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}}_{I_I} \underbrace{\begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix}}_{\Sigma} \underbrace{\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & i \\ 1 & -i \end{pmatrix}}_{V^*}.$$

The reduced SVD is equal to the full SVD since the matrix is a square matrix [1 point]

(ii) If A is hermitian and unitary, then  $A^*=A$  and  $A^*A=A^2=I$  [2 points]. If  $A=U\Sigma V^*$  is a SVD of A, then

$$I = A^*A = V\Sigma^*U^*U\Sigma V^* = V\Sigma^*\Sigma V^*$$
. [2 points]

Since  $\Sigma$  is a real diagonal matrix,  $\Sigma^*\Sigma = \Sigma^2$  which contains the singular values of A. Therefore,

$$I = V^*V = V^*(V\Sigma^2V^*)V = \Sigma^2$$
,

and so all singular values of A are 1. [1 point]

If  $\lambda$  is an eigenvalue of A with corresponding eigenvector x, then

$$x = A^*Ax = \lambda A^*x = \lambda Ax = \lambda^2x$$
 [2 points]

and so  $\lambda^2 = 1$ , which implies  $\lambda = \pm 1$ . [1 point]

[Total 25 points]

## Question 4

(i) For the matrix

$$A = \begin{pmatrix} 2 & 1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix}$$

find the orthogonal projector P onto range(A). What is the projection of a point  $(1,2,3)^{\mathsf{T}}$  to range(A)?

(ii) Consider the matrix

$$A = \begin{pmatrix} 1 & 2 & -1 \\ 4 & 8 & -4 \end{pmatrix}.$$

- Find the orthogonal projector  $P_C$  onto range(A).
- Find the orthogonal projector  $P_R$  onto range  $(A^{\mathsf{T}})$ .
- Prove that the following relation holds for general matrices  $A \in \mathbb{R}^{m \times n}$

$$P_CAP_R = A$$
.

#### 0.4 Solutions

(i) Orthogonal projector to range(A) is [3 points]

$$P = A(A^{\mathsf{T}}A)^{-1}A^{\mathsf{T}} = \begin{pmatrix} 2 & 1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} 1/6 & -1/6 \\ -1/6 & 2/3 \end{pmatrix} \begin{pmatrix} 2 & 0 & 2 \\ 1 & 1 & 0 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 2 & 1 & 1 \\ 1 & 2 & -1 \\ 1 & -1 & 2 \end{pmatrix}$$

Projection of point  $(1,2,3)^{\mathsf{T}}$  is [1 point]

$$P\begin{pmatrix} 1\\2\\3 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 7\\2\\5 \end{pmatrix}.$$

(ii) Note that the columns 2 and 3 are linear combinations of column 1, and so range(A) = span{ $(1,4)^{\mathsf{T}}$ } [1 point]. The orthogonal projector onto range(A) is therefore [2 points]

$$P_C = \frac{1}{\|c_1\|_2^2} (c_1 c_1^{\mathsf{T}}) = \frac{1}{17} \begin{pmatrix} 1 & 4 \\ 4 & 16 \end{pmatrix}.$$

Meanwhile, row 2 is a linear combination of row 1, and so range( $A^{\top}$ ) = span{(1, 2, -1)} [1 point]. The orthogonal projector onto range( $A^{\top}$ ) is therefore [2 points]

$$P_R = \frac{1}{\|r_1\|_2^2} r_1 r_1^{\mathsf{T}} = \frac{1}{6} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \\ -1 & -2 & 1 \end{pmatrix}.$$

For general matrices  $A \in \mathbb{R}^{m \times n}$  and any vector  $v \in \mathbb{R}^n$ , we see that  $Av \in \text{range}(A)$  and so  $P_C Av = Av$ , i.e.,  $P_C A = A$  [1 point]. Since  $P_R$  is the orthogonal projector to range( $A^{\mathsf{T}}$ ), for any pair of vectors w and v, it holds

$$(AP_Rv)^{\mathsf{T}}w = (AP_R^{\mathsf{T}}v)^{\mathsf{T}}w$$
 [hermitian/symmetric of projector]  
=  $v^{\mathsf{T}}P_RA^{\mathsf{T}}w = v^{\mathsf{T}}A^{\mathsf{T}}w = (Av)^{\mathsf{T}}w$ .

This implies  $AP_R = A$  [2 point], and so  $P_CAP_R = P_CA = A$  [1 point].

[Total 14 points]