# SI231 - Matrix Computations, Fall 2020-21

## Homework Set #1

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## **Acknowledgements:**

- 1) Deadline: 2020-09-27 23:59:59
- 2) No handwritten is accepted. You need to use LATEX. (If you have difficulty in using LATEX, you are allowed to use **Word** for the first and the second homework to accommodate yourself.)
- 3) Do use the given template.

#### I. UNDERSTANDING RANK, RANGE SPACE AND NULL SPACE

# **Problem 1.** (4 points $\times$ 5)

- 1) For matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , prove that  $\mathbb{R}^n = \mathcal{N}(\mathbf{A}) \oplus \mathcal{R}(\mathbf{A}^T)^{-1}$ . **Hint:**  $\dim(\mathcal{N}(\mathbf{A})) + \dim(\mathcal{R}(\mathbf{A}^T)) = n$ .
- 2) For matrices  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{B} \in \mathbb{R}^{m \times n}$ , prove that  $\operatorname{rank}(\mathbf{A} + \mathbf{B}) \leq \operatorname{rank}(\mathbf{A}) + \operatorname{rank}(\mathbf{B})$ .
- 3) For matrices  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{B} \in \mathbb{R}^{n \times p}$ , prove that  $\operatorname{rank}(\mathbf{AB}) \leq \min\{\operatorname{rank}(\mathbf{A}), \operatorname{rank}(\mathbf{B})\}$  and  $\operatorname{rank}(\mathbf{AB}) = n$  only when  $\mathbf{A}$  has full-column rank and  $\mathbf{B}$  has full-row rank.
- 4) For matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  and  $\mathbf{B} \in \mathbb{R}^{m \times p}$ , prove that  $\mathcal{R}(\mathbf{A}|\mathbf{B}) = \mathcal{R}(\mathbf{A}) + \mathcal{R}(\mathbf{B})^2$
- 5) For matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  and  $\mathbf{B} \in \mathbb{R}^{m \times p}$ , prove that

$$\mathsf{rank}(\mathbf{A}|\mathbf{B}) = \mathsf{rank}(\mathbf{A}) + \mathsf{rank}(\mathbf{B}) - \dim(\mathcal{R}(\mathbf{A}) \cap \mathcal{R}(\mathbf{B})).$$

**Hint:** Recall the result in 4).

#### II. UNDERSTANDING SPAN, SUBSPACE

**Problem 1.** (10 points) For a set of vectors  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ , prove that  $\mathrm{span}(S)$  is the intersection of all subspaces that contain S, i.e., prove that  $\mathrm{span}(S) = \mathcal{M}$  where  $\mathcal{M} := \bigcap_{s \subseteq \mathcal{V}} \mathcal{V}$  is the intersection of all subspaces that contain S and V denotes the subspace containing S.

**Hint:** Prove that  $span(S) \subseteq M$  and  $M \subseteq span(S)$ .

#### III. BASIS, DIMENSION AND PROJECTION

**Problem 1.** (2 points  $\times$  2) Determine the dimension of each of the following vector spaces:

1) The space of polynomials having degree n or less;

<sup>1</sup>Let  $S_1$  and  $S_2$  be two subspaces of  $\mathbb{R}^n$ , if  $S_1 \cap S_2 = \{0\}$  and  $S_1 + S_2 = \mathbb{R}^n$ , we define the **direct sum**  $\mathbb{R}^n = S_1 \oplus S_2$ .

<sup>2</sup>Here 
$$\mathbf{A}|\mathbf{B}$$
 denotes a new matrix combined by  $\mathbf{A}$  and  $\mathbf{B}$ . For example,  $\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ ,  $\mathbf{B} = \begin{bmatrix} b_{11} \\ b_{21} \end{bmatrix}$ , then  $\mathbf{A}|\mathbf{B} = \begin{bmatrix} a_{11} & a_{12} & b_{11} \\ a_{21} & a_{22} & b_{21} \end{bmatrix}$ .

2) The space of  $n \times n$  symmetric matrices.

## Problem 2. Some Important linear transformations

- 1) **Rotations.** (6 points) A rotation matrix  $\mathbf{R} \in \mathbb{R}^{n \times n}$  is an orthogonal matrix  $(\mathbf{R}\mathbf{R}^T = \mathbf{I})$  such that  $\det(\mathbf{R}) = 1$ .
  - According to above definition, find all rotation matrix in  $\mathbb{R}^{2\times 2}$ .
  - Geometrically, if  $\mathbf{R} \in \mathbb{R}^{2 \times 2}$ , then  $\mathbf{R} \mathbf{x}$  means we rotate the vector  $\mathbf{x} \in \mathbb{R}^2$  from some angle  $\theta \in [0, 2\pi]$  in anti-clockwise direction. For  $\mathbf{x} = [\cos(\pi/4), \sin(\pi/4)]^T$ , compute  $\mathbf{R} \mathbf{x}$ , where  $\mathbf{R}$  represents the matrix that rotating  $\mathbf{x}$  by  $7/12\pi$  in anti-clockwise direction.

**Hint:** draw a plot of x and Rx.

2) **Reflections.** (8 points) Let  $\mathbf{u} \in \mathbb{R}^n$  be a unit vector,  $\|\mathbf{u}\|_2 = 1$ . For a given vector  $\mathbf{x} \in \mathbb{R}^n$  and a hyperplane  $\mathcal{H}_u = \{\mathbf{x} \in \mathbb{R}^n | \mathbf{u}^T \mathbf{x} = 0\}$ . Let  $\mathbf{Q} = \mathbf{I} - \mathbf{u}\mathbf{u}^T$ . Then a vector  $\mathbf{y} \in \mathbb{R}^n$  is said to be a *reflection* of  $\mathbf{x}$  with respect to  $\mathcal{H}$  if their projections onto the hyperplane  $\mathcal{H}$  (denoted as  $\mathbf{Q}\mathbf{x}$  and  $\mathbf{Q}\mathbf{y}$  respectively) satisfy

$$\mathbf{Q}\mathbf{x} = \mathbf{Q}\mathbf{y}$$
,  $\|\mathbf{x} - \mathbf{Q}\mathbf{x}\|_2 = \|\mathbf{y} - \mathbf{Q}\mathbf{y}\|_2$ .

See Figure.1 for visualization.

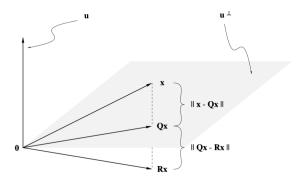


Figure 1. Reflection of x

A Householder matrix has the form  $\mathbf{H} = \mathbf{I} - 2\mathbf{u}\mathbf{u}^T$ . Prove that  $\mathbf{H}\mathbf{x}$  is a reflection of  $\mathbf{x}$  with respect to  $\mathcal{H}_u$ .

#### IV. DIRECT SUM

**Problem 1.** (10 points) Let  $\mathcal{V}$  be a vector space, and  $\mathcal{B}$  be a basis for  $\mathcal{V}$ . Suppose that there exist subsets  $\mathcal{B}_1, \mathcal{B}_2$  of  $\mathcal{B}$ , such that  $\mathcal{B} = \mathcal{B}_1 \cup \mathcal{B}_2$  and  $\mathcal{B}_1 \cap \mathcal{B}_2 = \emptyset$ . Then show that  $\mathcal{V} = \text{span}(\mathcal{B}_1) \oplus \text{span}(\mathcal{B}_2)$ .

**Problem 2.** (10 points) Let  $\mathcal{V}$  be a real vector space of dimension n. Let  $\mathcal{S}$  be a subspace of  $\mathcal{V}$  of dimension d < n. Prove that there exists a subspace  $\mathcal{T}$  of  $\mathcal{V}$  such that  $\mathcal{V} = \mathcal{S} \oplus \mathcal{T}$ .

# V. UNDERSTANDING THE MATRIX NORM

**Problem 1.** (7 points  $\times$  2) Matrix norm is induced by vector norm,

$$\|\mathbf{A}\|_p = \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|\mathbf{A}\mathbf{x}\|_p}{\|\mathbf{x}\|_p} = \max_{\|\mathbf{x}\|_p = 1} \|\mathbf{A}\mathbf{x}\|_p, \quad \mathbf{A} \in \mathbb{R}^{m \times n}, \mathbf{x} \in \mathbb{R}^{n \times 1},$$

prove that

1) the matrix 1-norm

$$\|\mathbf{A}\|_1 = \max_{\|\mathbf{x}\|_1 = 1} \|\mathbf{A}\mathbf{x}\|_1 = \max_j \sum_i^m |a_{ij}|$$

= the largest absolute column sum.

2) the matrix  $\infty$ -norm

$$\|\mathbf{A}\|_{\infty} = \max_{\|\mathbf{x}\|_{\infty} = 1} \|\mathbf{A}\mathbf{x}\|_{\infty} = \max_{i} \sum_{j}^{n} |a_{ij}|$$

= the largest absolute row sum.

#### VI. UNDERSTANDING THE HÖLDER INEQUALITY

**Problem 1.** (6 points  $\times$  3) Hölder inequality:

$$|\mathbf{x}^T \mathbf{y}| \leq ||\mathbf{x}||_p ||\mathbf{y}||_q$$

for any p,q such that 1/p + 1/q = 1,  $p \ge 1$ . Derive this inequality by exexcuting the following steps:

1) Consider the function  $f(t) = (1 - \lambda) + \lambda t - t^{\lambda}$  for  $0 < \lambda < 1$ , establish the inequality

$$\alpha^{\lambda} \beta^{1-\lambda} \le \lambda \alpha + (1-\lambda)\beta$$
,

for nonnegative real numbers  $\alpha$  and  $\beta$ .

2) Let  $\hat{\mathbf{x}} = \mathbf{x}/\|\mathbf{x}\|_p$  and  $\hat{\mathbf{y}} = \mathbf{y}/\|\mathbf{y}\|_q$ , and apply the inequality of part (a) to obtain

$$\sum_{i=1}^{n} |\hat{x}_i \hat{y}_i| \le \frac{1}{p} \sum_{i=1}^{n} |\hat{x}_i|^p + \frac{1}{q} \sum_{i=1}^{n} |\hat{y}_i|^q = 1.$$

- 3) Deduce the Hölder inequality with the above results.
- 4) (Bouns question: 10 points) Prove the general form of triangle inequality

$$\|\mathbf{x} + \mathbf{y}\|_p \le \|\mathbf{x}\|_p + \|\mathbf{y}\|_p.$$

**Hint:** For p > 1, let q be the number such that 1/q = 1 - 1/p. Verify that for scalars  $\alpha$  and  $\beta$ ,

$$|\alpha + \beta|^p = |\alpha + \beta||\alpha + \beta|^{p/q} \le |\alpha||\alpha + \beta|^{p+q} + |\beta||\alpha + \beta|^{p/q}$$

and make use of Hölder's inequality.