ENGG 5781: Matrix Analysis and Computations	2018-19 First Term
Assignment 1	
Instructor: Wing-Kin Ma	Due: October 5, 2018

Answer Problems 1–3, and either Problem 4 or Problem 5.

Note:

- 1. By submitting this assignment, we are assumed to have read the homework guideline http://www.ee.cuhk.edu.hk/~wkma/engg5781/hw/hw\_guidelines.pdf thereby understanding and respecting the guideline mentioned there.
- 2. You are allowed to use properties and theorems up to Lecture 2; unless specified, that includes all the results and proofs in the additional notes.

**Problem 1** (30%) Are the following sets subspaces? Provide your answer with a proof.

- (a)  $S = \{ \mathbf{X} \in \mathbb{R}^{m \times n} \mid \mathbf{X}^T \mathbf{Y} + \mathbf{Y}^T \mathbf{X} = \mathbf{0} \}$ , where  $\mathbf{Y} \in \mathbb{R}^{m \times n}$  is given.
- (b)  $S = {\mathbf{X} \in \mathbb{R}^{m \times n} \mid \mathbf{X} = \mathbf{A}\mathbf{B}^T, \mathbf{A} \in \mathbb{R}^{m \times r}, \mathbf{B} \in \mathbb{R}^{n \times r}}, \text{ where } r < \min\{m, n\}.$
- (c)  $S = \{ \mathcal{X} \in \mathbb{R}^{m \times n \times p} \mid x_{ijk} = \sum_{\ell=1}^{r} a_{i\ell} b_{j\ell} c_{k\ell}, \ \forall i, j, k, \ c_{k\ell} \in \mathbb{R} \ \forall k, \ell \}, \text{ where } a_{i\ell}, b_{j\ell} \in \mathbb{R} \text{ are given.}$
- (d)  $S = \{ \mathbf{y} \in \mathbb{C}^m \mid |\sum_{i=1}^m y_i(\alpha_i)^i| = 0, \ j = 1, \dots, n \}, \text{ where } \alpha_1, \dots, \alpha_n \in \mathbb{C} \text{ are given.}$

**Problem 2** (20%) A non-empty subset S of  $\mathbb{R}^m$  is said to be an affine set if

$$\alpha \in \mathbb{R}, \mathbf{x}, \mathbf{y} \in \mathcal{S} \implies \alpha \mathbf{x} + (1 - \alpha) \mathbf{y} \in \mathcal{S}.$$

(a) Show that if S is affine, then any affine combination of  $\mathbf{a}_1, \ldots, \mathbf{a}_n \in S$ , i.e.,

$$\mathbf{y} = \sum_{i=1}^{n} \alpha_i \mathbf{a}_i, \quad \alpha_1, \dots, \alpha_n \in \mathbb{R}, \quad \sum_{i=1}^{n} \alpha_i = 1,$$

lies in S.

(b) Show that an affine set S can always be represented by  $S = V + \mathbf{b}$ , where  $\mathbf{b} \in S$  and V is a subspace<sup>1</sup>.

Problem 3 (20%) Let  $S = \{ \mathbf{y} \in \mathbb{R}^m \mid \mathbf{y} = \mathbf{A}\mathbf{x} - \mathbf{b}, \mathbf{x} \in \mathbb{R}^n, \|\mathbf{x}\|_2 \le 1 \}$ . Suppose  $\mathbf{b} \notin \mathcal{R}(\mathbf{A})$ .

- (a) Is  $S \cap S^{\perp} = \{0\}$ , or  $S \cap S^{\perp} = \emptyset$ ?
- (b) Show that  $S^{\perp} = \mathcal{N}([\mathbf{A} \mathbf{b}]^T)$ .

 $<sup>^{1} \</sup>text{The notation } \overline{\mathbf{b} + \mathcal{V} \text{ means that } \mathbf{b} + \mathcal{V} = \{ \mathbf{y} = \mathbf{b} + \mathbf{v} \mid \mathbf{v} \in \mathcal{V} \}.$ 

**Problem 4 (30%)** Let  $\{\mathbf{a}_1, \ldots, \mathbf{a}_n\} \subset \mathbb{R}^m$  be a given set of linearly independent vectors. Let  $\mathbf{q}_1, \ldots, \mathbf{q}_n \in \mathbb{R}^m$  whose construction will be specified, and let  $S_i = \text{span}\{\mathbf{q}_1, \ldots, \mathbf{q}_i\}$  for the sake of notational convenience. Consider the following procedure.

## Algorithm 1:

```
1 \mathbf{q}_1 = \mathbf{a}_1/\|\mathbf{a}_1\|_2;

2 for i=2,\ldots,n do

3 \tilde{\mathbf{q}}_i = \Pi_{\mathcal{S}_{i-1}^{\perp}}(\mathbf{a}_i);

4 \mathbf{q}_i = \tilde{\mathbf{q}}_i/\|\tilde{\mathbf{q}}_i\|_2;

5 end
```

(a) Show that, for any  $i \in \{1, ..., n\}$ ,

$$\operatorname{span}\{\mathbf{a}_1,\ldots,\mathbf{a}_i\}=\mathcal{S}_i.$$

DO NOT use the proof in Lecture 1, page 47–50. Consider the problem as if you did not know what is Gram-Schmidt, which you can easily find in textbooks or in the world-wide web. Use ONLY the basic notions of subspace, with projection onto subspaces included, to rediscover the result.

- (b) Use the basic notions of subspaces, as well as those of orthogonality and LS, to show that  $\tilde{\mathbf{q}}_i = \mathbf{a}_i \mathbf{Q}_{i-1}\mathbf{Q}_{i-1}^T\mathbf{a}_i$ , where  $\mathbf{Q}_i = [\mathbf{q}_1, \dots, \mathbf{q}_i]$ .
- (c) Suggest how we may modify the algorithm when  $\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$  is linearly dependent. Note that the ultimate goal is to find an orthogonal basis for span $\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$ .

Note: Since this problem requires some explanation, we consider it as a semi-essay problem. English fluency, clarity of presentation, originality of the presentation (relative to others), etc., will be heavily taken into account.

**Problem 5** (30%) This is a MATLAB problem. The problem we deal with is a face recognition problem. Your main reference is [1], Chapter 9. I also recommend [2] as an additional reference for you to get better understanding of the context. Download from the course website the following three files: X.mat, y\_part\_a.mat, and y\_part\_b.mat. Loading X.mat on MATLAB, you will see five matrices, namely, X\_1, X\_2, X\_3, X\_4, X\_5. I will sometimes use  $X_1, X_2, X_3, X_4, X_5$ , our standard matrix notations, to describe them. Each  $X_i$  is a collection of 63 images, with size  $192 \times 168$ , taken from the same person. Specifically, each column of  $X_i$  is an image stored in the vectorized form. You can see them, say, for  $X_1$ , by calling

```
>> for i=1:63, subplot(8,8,i); imshow(reshape(X_1(:,i),192,168)); end;
```

I should mention that the data come from "Yale Face Database B."

(a) Load y\_part\_a.mat. You will find a vector y, or y. You can see it by calling

```
>> imshow(reshape(y,192,168));
```

You will see that it is a noisy image. Form a multi-person data matrix  $\mathbf{X} = [\mathbf{X}_1, \dots, \mathbf{X}_5]$ , apply LS

$$\min_{\mathbf{w} \in \mathbb{R}^n} \|\mathbf{y} - \mathbf{X}\mathbf{w}\|_2^2,$$

and use the LS solution to construct an estimated, and hopefully noise-cleaned, image. Show us the image you recover. Then, from the LS solution, identify the person **y** is associated with. Note: You can certainly watch the images and *subjectively* identify who that person is, but I would like to see a quantitative way.

(b) Load y\_part\_b.mat, which contains a vector y. Using imshow, you will see that a small part of the image is severely corrupted. Try LS and show the recovered image. Then, try Algorithm 11 in [1], AltMin for Robust Regression (AM-RR), with parameter k = 4775. You should also write a short description concerning what is the rationale of AM-RR, and how it works. Show the recovered image of AM-RR.

Note: You will need to submit your MATLAB code online via Blackboard. You also need to provide a description on your assignment. We do not do reverse engineering tasks such as guessing how a MATLAB code works in the absence of any description. If you are in doubt, talk to us to understand more.

## References

- [1] P. Jain and P. Kar. Non-convex optimization for machine learning. Foundations and Trends<sup>(R)</sup> in Machine Learning, 10(3–4):142–336, 2017.
- [2] J. Wright, A. Y. Yang, A. Ganesh, S. S. Sastry, and Y. Ma. Robust face recognition via sparse representation. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 31(2):210–227, 2009.