This homework is due on Thursday, November 5, 2020, at 10:59PM.

Self-grades are due on Thursday, November 12, 2020, at 10:59PM.

# 1 SVD I

Find the singular value decomposition of the following matrix (leave all work in exact form, not decimal):

$$A = \begin{bmatrix} 2 & 2\\ 3 & -3 \end{bmatrix}$$

- a) Find the eigenvalues of  $AA^{\top}$  and order them from largest to smallest,  $\lambda_1 > \lambda_2$ .
- b) Find orthonormal eigenvectors  $\vec{u}_i$  of  $AA^{\top}$  (all eigenvectors are mutually orthogonal and unit length).
- c) Find the singular values  $\sigma_i = \sqrt{\lambda_i}$ . Find the  $\vec{v}_i$  vectors from:

$$A^{\top}\vec{u}_i = \sigma_i \vec{v}_i$$

d) Write out A as a weighted sum of rank-1 matrices:

$$A = \sigma_1 \vec{u}_1 \vec{v}_1^\top + \sigma_2 \vec{u}_2 \vec{v}_2^\top$$

## 2 Rank 1 Decomposition

In this problem, we will decompose a few images into linear combinations of rank 1 matrices. Remember that outer product of two vectors  $\vec{s}\vec{g}^T$  gives a rank 1 matrix. It has rank 1 because clearly, the column span is one-dimensional — multiples of  $\vec{s}$  only — and the row span is also one dimensional — multiples of  $\vec{g}^T$  only.

For example, if  $\vec{s}$  and  $\vec{g}$  are two vectors of dimension 5, then  $\vec{s}\vec{g}^T$  is given as follows.

$$\vec{s} = \begin{bmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \\ s_5 \end{bmatrix} \qquad \vec{g} = \begin{bmatrix} g_1 \\ g_2 \\ g_3 \\ g_4 \\ g_5 \end{bmatrix}$$
$$\vec{g} = \begin{bmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \\ s_5 \end{bmatrix} \begin{bmatrix} g_1 & g_2 & g_3 & g_4 & g_5 \end{bmatrix} = \begin{bmatrix} s_1g_1 & s_1g_2 & s_1g_3 & s_1g_4 & s_1g_5 \\ s_2g_1 & s_2g_2 & s_2g_3 & s_2g_4 & s_2g_5 \\ s_3g_1 & s_3g_2 & s_3g_3 & s_3g_4 & s_3g_5 \\ s_4g_1 & s_4g_2 & s_4g_3 & s_4g_4 & s_4g_5 \\ s_5g_1 & s_5g_2 & s_5g_3 & s_5g_4 & s_5g_5 \end{bmatrix}$$

a) Consider a standard  $8 \times 8$  chessboard shown in Figure 1. Assume that black colors represent -1 and that white colors represent 1.



Figure 1:  $8 \times 8$  chessboard.

Hence, that the chessboard is given by the following  $8 \times 8$  matrix  $C_1$ :

|                         | -  |    |    |    |    |    |    |    |
|-------------------------|----|----|----|----|----|----|----|----|
| <i>C</i> <sub>1</sub> = | 1  | -1 | 1  | -1 | 1  | -1 | 1  | -1 |
|                         | -1 | 1  | -1 | 1  | -1 | 1  | -1 | 1  |
|                         | 1  | -1 | 1  | -1 | 1  | -1 | 1  | -1 |
|                         | -1 | 1  | -1 | 1  | -1 | 1  | -1 | 1  |
|                         | 1  | -1 | 1  | -1 | 1  | -1 | 1  | -1 |
|                         | -1 | 1  | -1 | 1  | -1 | 1  | -1 | 1  |
|                         | 1  | -1 | 1  | -1 | 1  | -1 | 1  | -1 |
|                         | -1 | 1  | -1 | 1  | -1 | 1  | -1 | 1  |
|                         |    |    |    |    |    |    |    |    |

**Express**  $C_1$  **as a linear combination of outer products.** *Hint: In order to determine how many rank 1 matrices you need to combine to represent the matrix, find the rank of the matrix you are trying to represent.* 

b) For the same chessboard shown in Figure 1, now assume that black colors represent 0 and that white colors represent 1.

Hence, the chessboard is given by the following  $8 \times 8$  matrix  $C_2$ :

$$C_2 = \begin{vmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{vmatrix}$$

#### Express C<sub>2</sub> as a linear combination of outer products.

c) Now consider the Swiss flag shown in Figure 2. Assume that red colors represent 0 and that white colors represent 1.



Figure 2: Swiss flag.

Assume that the Swiss flag is given by the following  $5 \times 5$  matrix *S*:

$$S = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Furthermore, we know that the Swiss flag can be viewed as a superposition of the following pairs of images:



Figure 3: Pairs of images - Option 1



Figure 4: Pairs of images - Option 2

Express the *S* in two different ways: i) as a linear combination of the outer products inspired by the Option 1 images and ii) as a linear combination of outer products inspired by the Option 2 images.

## **3** SVD properties

In this question, we look at some properties of SVD. Below we consider a m by n matrix whose SVD writes  $A = \sum_{i=1}^{r} \sigma_i \vec{u}_i \vec{v}_i^T$  where r is the rank of the matrix.

a) We know that *A* can also be represented in matrix form as  $A = \sum_{i=1}^{r} \sigma_{i} \vec{u}_{i} \vec{v}_{i}^{T} = U_{1} S V_{1}^{T}$ where  $U_{1} = [\vec{u}_{1}, \vec{u}_{2}, ..., \vec{u}_{r}]$ ,  $S = \text{diag}(\sigma_{1}, \sigma_{2}, ..., \sigma_{r})$ , and  $V_{1} = [\vec{v}_{1}, \vec{v}_{2}, ..., \vec{v}_{r}]$ . Show that  $\sum_{i=1}^{r} \sigma_{i} \vec{u}_{i} \vec{v}_{i}^{T} = U_{1} S V_{1}^{T}$ . Note that the math does not assume any further property for  $U_{1}$  and  $V_{1}$  than that they are of compatible shape as long as *S* is diagonal.

The full SVD of *A* writes  $A = U\Sigma V^T$  where  $U = [\vec{u}_1, \vec{u}_2, \dots, \vec{u}_r, \vec{u}_{r+1}, \dots, \vec{u}_m]$  and  $V = [\vec{v}_1, \vec{v}_2, \dots, \vec{v}_r, \vec{v}_{r+1}, \dots, \vec{v}_n]$  are orthornormal matrices with the first r columns being the same as those of *U*' and *V*'.

$$\Sigma = \begin{bmatrix} S_{r \times r} & 0_{r \times (n-r)} \\ 0_{(m-r) \times r} & 0_{(m-r) \times (n-r)} \end{bmatrix}.$$

- b) Suppose  $r < \min(m, n)$ , what is a basis of the null space of *A*? Prove your answer.
- c) What is the basis of the range of *A*? Prove your answer.
- d) Suppose n = m = r, that is, A is square and full rank. Find the inverse of A in terms of  $U, \Sigma$ , and V.

## 4 Induced Matrix Norms

Often, the general effect of matrices on their inputs is really hard to predict. To overcome this, we usually try to "bound" the effect a matrix has on input vectors. We will work through a simple case of bounding the output of an  $n \times n$  matrix T given that T is diagonalizable and symmetric. We will consider the system

$$\vec{y} = T\vec{x}$$
,

where  $\vec{x}$  is the input vector and  $\vec{y}$  is the output vector.

- a) Let  $U = [\vec{u}_1, \vec{u}_2, ..., \vec{u}_n]$  be a set of orthonormal eigenvectors of the matrix *T*. Decompose the generic input vector  $\vec{x}$  into a linear combination of these eigenvectors.
- b) Let the eigenvalues of *T* be  $\{\lambda_1, \lambda_2, ..., \lambda_n\}$  with  $|\lambda_1| \ge |\lambda_2| \ge ... |\lambda_n|$ . If we represent  $\vec{x}$  as a linear combination of the eigenvectors of *T*, what is the Euclidean norm of the output vector  $\vec{y}$ ,  $\|\vec{y}\|$ ?

Hint: You can use the fact that Euclidean norms are preserved with orthonormal transforms.

c) Say you do not know all the eigenvalues of *T*, but you know the largest eigenvalue  $\lambda_1$ . If the norm  $||\vec{x}|| = \alpha$ , how big could  $||\vec{y}||$  be?

The maximum factor by which a square matrix can grow the norm of a vector is called the induced norm for that matrix. Although we had you do the derivation above for a symmetric matrix *T*, the fact that  $||A\vec{x}|| = \sqrt{\vec{x}^T A^T A \vec{x}}$  can be used to show how to generalize this concept of induced norm for general matrices.

### 5 Closed-Loop Control of SIXT33N

To make our control more robust, we introduce feedback, turning our open-loop controller into a closed-loop controller. In this problem, we derive the closed-loop control scheme you will use to make SIXT33N reliably drive straight.

We introduce  $\delta(k) = d_L(k) - d_R(k)$  as the difference in positions between the two wheels. We will consider a proportional control scheme, which introduces a feedback term into our input equation in which we apply gains  $k_L$  and  $k_R$  to  $\delta(k)$  to modify our input at each timestep in an effort to prevent  $|\delta(k)|$  from growing without bound. To do this, we will modify our inputs  $u_L(k)$  and  $u_R(k)$  to be:

$$u_L(k) = \frac{v^* + \beta_L}{\theta_L} - k_L \frac{\delta(k)}{\theta_L}$$
$$u_R(k) = \frac{v^* + \beta_R}{\theta_R} + k_R \frac{\delta(k)}{\theta_R}$$

Substituting into the open-loop equations

$$d_L(k+1) - d_L(k) = \theta_L u_L(k) - \beta_L$$
(1)  
$$d_R(k+1) - d_R(k) = \theta_R u_R(k) - \beta_R$$

we obtain:

$$d_L(k+1) - d_L(k) = v^* - k_L \delta(k)$$
(2)  
$$d_R(k+1) - d_R(k) = v^* + k_R \delta(k)$$

- a) Let's look a bit more closely at picking  $k_L$  and  $k_R$ . First, we need to figure out what happens to  $\delta(k)$  over time. Find  $\delta(k + 1)$  in terms of  $\delta(k)$ .
- b) Given your work above, what is the eigenvalue of the system defined by  $\delta(k)$ ? For discrete-time systems like our system,  $\lambda \in (-1, 1)$  is considered stable. Are  $\lambda \in [0, 1)$  and  $\lambda \in (-1, 0]$  identical in function for our system? Which one is "better"? (*Hint:* Preventing oscillation is a desired benefit.)

Based on your choice for the range of  $\lambda$  above, how should we set  $k_L$  and  $k_R$  in the end?

c) Let's re-introduce the model mismatch in order to model environmental discrepancies, disturbances, etc. How does closed-loop control fare under model mismatch? Find  $\delta_{ss} = \delta[k \to \infty]$ , assuming that  $\delta[0] = \delta_0$ . What is  $\delta_{ss}$ ? (To make this easier, you may leave your answer in terms of appropriately defined *c* and  $\lambda$  obtained from an equation in the form of  $\delta(k+1) = \delta(k)\lambda + c$ .)

Check your work by verifying that you reproduce the equation in part (c) if all model mismatch terms are zero. Is it better than the open-loop model mismatch?

$$d_L(k+1) - d_L(k) = (\theta_L + \Delta \theta_L)u_L(k) - (\beta_L + \Delta \beta_L)$$
$$d_R(k+1) - d_R(k) = (\theta_R + \Delta \theta_R)u_R(k) - (\beta_R + \Delta \beta_R)$$

$$u_L(k) = \frac{v^* + \beta_L}{\theta_L} - k_L \frac{\delta(k)}{\theta_L}$$
$$u_R(k) = \frac{v^* + \beta_R}{\theta_R} + k_R \frac{\delta(k)}{\theta_R}$$

# 6 Homework Process and Study Group

Citing sources and collaborators are an important part of life, including being a student! We also want to understand what resources you find helpful and how much time homework is taking, so we can change things in the future if possible.

- a) What sources (if any) did you use as you worked through the homework?
- b) **If you worked with someone on this homework, who did you work with?** List names and student ID's. (In case of homework party, you can also just describe the group.)
- c) Roughly how many total hours did you work on this homework?
- d) Do you have any feedback on this homework assignment?