

**Math 218D Problem Session: Week 14**

December 9, 2022

**1. Rules of vector SVD**

Which of the following  $A = \sigma_1 u_1 v_1^T + \cdots + \sigma_r u_r v_r^T$  are valid singular value decompositions? Why/why not?

a)  $A = 1(1, 0)(1, 0)^T + 3(0, 1)(0, 1)^T$

b)  $A = 4(1, 0)(0, 1)^T + 3(0, 1)(1, 0)^T$

c)  $A = 3(1, -1)(1, 0)^T + 2(1, 1)(0, 1)^T$

d)  $A = -3(1/\sqrt{2}, -1/\sqrt{2}, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

e)  $A = 3(-1/\sqrt{2}, 1/\sqrt{2}, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

f)  $A = 5(1, 0, 0)(0, 1)^T + 3(0, 1, 0)(1, 0)^T + 2(0, 0, 1)(0, 1)^T$

- 2. The matrix SVD** Suppose that  $A$  is an  $m \times n$  matrix of rank  $r$ , with SVD  $A = U\Sigma V^T$ .
- a)  $U$  is a  $\text{---} \times \text{---}$  matrix,  $\Sigma$  is a  $\text{---} \times \text{---}$  matrix, and  $V$  is a  $\text{---} \times \text{---}$  matrix. The matrices  $U$  and  $V$  are  $\text{---}$  matrices. The first  $\text{---}$  diagonal entries of  $\Sigma$  are  $> 0$ .
  - b) Expand  $A^T A$  using  $A = U\Sigma V^T$  to see that the matrix  $A^T A$  has symmetric diagonalization  $Q_1 D_1 Q_1^T$ , with  $Q_1 = \text{---}$  and  $D_1 = \text{---}$ . What are the eigenvectors of  $A^T A$ ? What are the eigenvalues?
  - c) Expand  $AA^T$  using  $A = U\Sigma V^T$  to see that the matrix  $AA^T$  has symmetric diagonalization  $Q_2 D_2 Q_2^T$ , with  $Q_2 = \text{---}$  and  $D_2 = \text{---}$ . What are the eigenvectors of  $AA^T$ ? What are the eigenvalues?
  - d) Suppose that  $i \leq r$ . The left singular vector  $u_i$  is the  $i$ th column of  $U$ , the singular value  $\sigma_i$  is the  $i$ th diagonal entry of  $\Sigma$ , and the right singular vector  $v_i$  is the  $i$ th column of  $V$ . Explain why  $Av_i = \sigma_i u_i$  by computing  $V^T v_i$ ,  $\Sigma V^T v_i$ , and  $U\Sigma V^T v_i$ .

### 3. Computing the vector SVD

To

- (1) Find the non-zero eigenvalues  $\lambda_1 \geq \lambda_2 \geq \dots \lambda_r > 0$  of  $A^T A$ .
- (2) Find an orthonormal basis of each of the  $\lambda_i$  eigenspace of  $A^T A$ . Listed in order of decreasing eigenvalue, these are the right singular vectors  $v_1, \dots, v_r$ .
- (3) For  $i = 1, \dots, r$ , set  $\sigma_i = \sqrt{\lambda_i}$  and  $u_i = \frac{Av_i}{\sigma_i}$ . These are the singular values and left singular vectors.
- (4) Write  $A = \sigma_1 u_1 v_1^T + \dots + \sigma_r u_r v_r^T$ .

Compute the vector SVD of each of the following matrices:

a)  $A = \begin{pmatrix} 0 & -1 \\ 3 & 0 \end{pmatrix}$

b)  $A = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -3 & 0 & 0 \end{pmatrix}$

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

#### 4. Computing the matrix SVD

To find the matrix SVD  $A = U\Sigma V^T$  of a matrix  $A$ :

- (1) Find the symmetric diagonalization  $VDV^T$  of  $A^T A$ , where the eigenvalues are listed in decreasing order:  $\lambda_1 \geq \dots \geq \lambda_n$ . The rank  $r$  of  $A$  is the same as the number of non-zero eigenvalues of  $A^T A$  (counted with multiplicity).
- (2) The columns of  $V$  are the right singular vectors  $v_1, \dots, v_r$ , followed by an orthonormal basis  $v_{r+1}, \dots, v_n$  of  $\text{Nul}(A)$ .
- (3) For  $i = 1, \dots, r$ , set  $\sigma_i = \sqrt{\lambda_i}$  and  $u_i = \frac{Av_i}{\sigma_i}$ . These are the singular values and left singular vectors.
- (4) We still need the vectors  $u_{r+1}, \dots, u_m$ : find these by computing an orthonormal basis of  $\text{Nul}(A^T)$  (using RREF to find a basis, Gram–Schmidt to replace it with an orthonormal basis).
- (5) Finally, the matrix  $U$  is the matrix with columns  $u_1, \dots, u_m$ , the matrix  $V$  was found in (1), and  $\Sigma$  has its first  $r$  diagonal entries as  $\sigma_1, \dots, \sigma_r$  and the remaining entries of  $\Sigma$  being zero.

Compute the matrix SVD of each of the following matrices:

a)  $A = \begin{pmatrix} 0 & -1 \\ 3 & 0 \end{pmatrix}$

b)  $A = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -3 & 0 & 0 \end{pmatrix}$

c)  $A = \begin{pmatrix} 1 & 2 \\ 1 & 2 \\ 2 & 4 \end{pmatrix}$

## 5. Sums of rank 1 matrices

This final problem is not about SVDs, but just about sums of rank one matrices.

- a) Without computing  $A$ , explain why

$$A = (1, 2, 1)(1, 1)^T + (1, -1, 1)(-1, 1)^T$$

is a rank 2 matrix.

**Hint:** compute  $A(1, 1)$  and  $A(-1, 1)$ , and use this to show that  $(1, 2, 1)$  and  $(1, -1, 1)$  are in the column space of  $A$ .

- b) If  $A = u_1 v_1^T + \cdots + u_r v_r^T$  for some vectors  $u_i \in \mathbf{R}^m$  and  $v_j \in \mathbf{R}^n$ , explain why the rank of  $A$  is at most  $r$ .

**Hint:** Show that the subspace  $\text{Col}(A)$  is contained in the span  $\text{Span}\{u_1, \dots, u_r\}$ , which is at most  $r$ -dimensional.

- c) If the vectors  $u_1, \dots, u_r \in \mathbf{R}^m$  are a linearly independent set of vectors, and the vectors  $v_1, \dots, v_r \in \mathbf{R}^n$  are also linearly independent, prove that

$$A = u_1 v_1^T + \cdots + u_r v_r^T$$

has rank equal to  $r$ .

**Hint:** Show that there is a vector  $v \in \mathbf{R}^n$  which is orthogonal to  $v_2, \dots, v_r$ , but  $v_1^T v \neq 0$ . Compute  $Av$ , and use this to show that  $u_1 \in \text{Col}(A)$ . The same idea shows that  $u_2, \dots, u_r$  are all in  $\text{Col}(A)$ .

**6. Principal component analysis** Consider the following data points

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 3 \\ 5 \\ 7 \end{pmatrix}, \begin{pmatrix} 4 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix}$$

- a) Calculate the mean value vector  $\bar{\mu}$  then recenter the data points and put them in the matrix form  $A$ .
- b) Calculate the covariance matrix by the formula  $S = \frac{1}{n-1}A \cdot A^T$ . What does the diagonal of this matrix tells you?
- c) Using a calculator, find the eigenvalue of  $S$  and its corresponding eigenvector.
- d) Find the principal component and the variance along that direction. What is the geometric meaning of the principal component?