

# Maths for Signals and Systems

## Problem Sheet 7

### Singular Value Decomposition (SVD)

Consider a matrix  $A$  of dimension  $m \times n$  with  $m < n$  and rank  $r$ . Recall from the lectures that  $r \leq m$ . The matrix  $AA^T$  is square, symmetric and of dimension  $m \times m$ . The matrix  $A^T A$  is square, symmetric and of dimension  $n \times n$ . The following properties hold:

- Both  $AA^T$  and  $A^T A$  have rank  $r$  (the same rank as the original matrix  $A$ ).
- The  $m$  eigenvalues of  $A^T A$  are identical to the eigenvalues of  $AA^T$  and the rest  $n - m$  eigenvalues are 0.
- The so called **singular values** of  $A$  are the square roots of the non-zero eigenvalues of  $AA^T$  (or  $A^T A$ ).
- Matrix  $A$  has a so called **Singular Value Decomposition (SVD)** of the form  $A = U\Sigma V^T$  where  $U$  is of dimension  $m \times m$ ,  $\Sigma$  is of dimension  $m \times n$  and  $V$  is of dimension  $n \times n$ . Furthermore,  $U$  contains the eigenvectors of  $AA^T$  in its columns,  $V$  contains the eigenvectors of  $A^T A$  in its columns and  $\Sigma_{ij} = \begin{cases} \sigma_i = \sqrt{\lambda_i} & i = j, i \leq r \\ 0 & \text{otherwise} \end{cases}$  with  $\lambda_i, i = 1, \dots, r$  the non-zero eigenvalues of  $AA^T$  (or  $A^T A$ ).
- The above comments imply that  $AA^T = U\Sigma^2 U^T$  and  $A^T A = V\Sigma^2 V^T$ .

The above analysis is straightforward in the case of  $m > n$ .

To understand better the structure of  $\Sigma$ , in the case of a  $3 \times 4$  matrix of rank 2 we have

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 & 0 & 0 \\ 0 & \sigma_2 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \text{ whereas in the case of a } 4 \times 3 \text{ matrix of rank 2 we have } \Sigma = \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

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### Problems

1. Find the singular values of the matrix  $A = \begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$ .

### Solution

We compute  $AA^T$ . (This is the smaller of the two symmetric matrices associated with  $A$ .) We

get  $AA^T = \begin{bmatrix} 3 & 1 & 2 \\ 1 & 1 & 0 \\ 2 & 0 & 2 \end{bmatrix}$ . We next find the eigenvalues of this matrix. The characteristic

polynomial is  $\lambda^3 - 6\lambda^2 + 6\lambda = \lambda(\lambda^2 - 6\lambda + 6)$ . This gives three eigenvalues:  $\lambda_1 = 3 + \sqrt{3}$ ,  $\lambda_2 = 3 - \sqrt{3}$  and  $\lambda_3 = 0$ . Note that all are positive, and that there are two nonzero eigenvalues,

corresponding to the fact that  $A$  has rank 2. For the singular values of  $A$ , we now take the square roots of the eigenvalues of  $AA^T$ , so  $\sigma_1 = \sqrt{3+\sqrt{3}}$  and  $\sigma_2 = \sqrt{3-\sqrt{3}}$ . (We don't have to mention the singular values which are zero.)

2. Find the singular values of the matrix  $B = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ .

**Solution**

We use the same approach:  $BB^T = \begin{bmatrix} 5 & 4 \\ 4 & 5 \end{bmatrix}$ . This has characteristic polynomial  $\lambda^2 - 10\lambda + 9$  so  $\lambda_1 = 9$  and  $\lambda_2 = 1$  are the eigenvalues. Hence, the singular values are 3 and 1.

3. Find the singular values of  $A = \begin{bmatrix} 0 & 1 & 1 \\ \sqrt{2} & 2 & 0 \\ 0 & 1 & 1 \end{bmatrix}$  and find the SVD of  $A$ .

**Solution**

We compute  $AA^T$  and find  $AA^T = \begin{bmatrix} 2 & 2 & 2 \\ 2 & 6 & 2 \\ 2 & 2 & 2 \end{bmatrix}$ . The characteristic polynomial is  $-\lambda^3 + 10\lambda^2 - 16\lambda = -\lambda(\lambda^2 - 10\lambda + 16) = -\lambda(\lambda - 8)(\lambda - 2)$ .

The eigenvalues of  $AA^T$  are  $\lambda_1 = 8$ ,  $\lambda_2 = 2$  and  $\lambda_3 = 0$ . Thus, the singular values are  $\sigma_1 = 2\sqrt{2}$  and  $\sigma_2 = \sqrt{2}$  (and  $\sigma_3 = 0$ ). To give the decomposition, we consider the diagonal matrix of

singular values  $\Sigma = \begin{bmatrix} 2\sqrt{2} & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & 0 \end{bmatrix}$ . Next, we find an orthonormal set of eigenvectors for  $AA^T$ .

For  $\lambda_1 = 8$ , we find an eigenvector  $[1 \ 2 \ 1]^T$  - normalizing gives  $u_1 = [1/\sqrt{6} \ 2/\sqrt{6} \ 1/\sqrt{6}]^T$ . For  $\lambda_2 = 2$  we find  $u_2 = [-1/\sqrt{3} \ 1/\sqrt{3} \ -1/\sqrt{3}]^T$ , and finally for  $\lambda_3 = 0$  we get  $u_3 = [1/\sqrt{2} \ 0 \ -1/\sqrt{2}]^T$ . This gives the matrix:

$$U = \begin{bmatrix} 1/\sqrt{6} & -1/\sqrt{3} & 1/\sqrt{2} \\ 2/\sqrt{6} & 1/\sqrt{3} & 0 \\ 1/\sqrt{6} & -1/\sqrt{3} & -1/\sqrt{2} \end{bmatrix}.$$

Finally, we have to find an orthogonal set of eigenvectors for  $A^T A = \begin{bmatrix} 2 & 2\sqrt{2} & 0 \\ 2\sqrt{2} & 6 & 2 \\ 0 & 2 & 2 \end{bmatrix}$ .

This can be done in two ways. We show both ways, starting with orthogonal diagonalization. We already know that the eigenvalues will be  $\lambda_1 = 8$ ,  $\lambda_2 = 2$  and  $\lambda_3 = 0$ . This gives eigenvectors  $v_1 = [1/\sqrt{6} \ 3/\sqrt{12} \ 1/\sqrt{12}]^T$ ,  $v_2 = [1/\sqrt{3} \ 0 \ -2/\sqrt{6}]^T$  and  $v_3 = [1/\sqrt{2} \ -1/2 \ 1/2]^T$ . Put these together to get:

$$V = \begin{bmatrix} 1/\sqrt{6} & 1/\sqrt{3} & 1/\sqrt{2} \\ 3/\sqrt{12} & 0 & -1/2 \\ 1/\sqrt{12} & -2/\sqrt{6} & 1/2 \end{bmatrix}$$

For a quicker method, we calculate the columns of  $V$  using those of  $U$  using the formula:

$$v_i = \frac{1}{\sigma_i} A^T u_i$$

$$v_1 = \frac{1}{2\sqrt{2}} \begin{bmatrix} 0 & \sqrt{2} & 0 \\ 1 & 2 & 1 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix} = \frac{1}{2\sqrt{2}} \begin{bmatrix} 2\sqrt{2}/\sqrt{6} \\ 6/\sqrt{6} \\ 2/\sqrt{6} \end{bmatrix} = \begin{bmatrix} 1/\sqrt{6} \\ 3/\sqrt{12} \\ 1/\sqrt{12} \end{bmatrix}.$$

We can similarly calculate the other two columns.

Either way we can now verify the formula  $A = U\Sigma V^T$ .

4. Find the SVD of the matrix  $A = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix}$ .

**Solution**

We first compute  $AA^T = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$  and  $A^T A = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix}$ .

We see immediately that the eigenvalues of  $AA^T$  are  $\lambda_1 = \lambda_2 = 2$  (and hence, the eigenvalues of  $A^T A$  are 2 and 0, both with multiplicity 2). Thus, the matrix  $A$  has singular value  $\sigma_1 = \sigma_2 = \sqrt{2}$ .

Next, an orthonormal basis of eigenvectors of  $AA^T$  is  $u_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $u_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ . (You can choose any orthonormal basis of  $R^2$  here because  $AA^T$  is a multiple of the identity, but the one chosen makes computation easiest.) Thus, we set  $U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ .

For  $A^T A$ , the eigenvectors which correspond to the value of 2 are obtained from the formula below:

$$\begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = 2 \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} \Rightarrow \begin{cases} x + z = 2x \Rightarrow z = x \\ y + w = 2y \Rightarrow y = w \\ x + z = 2z \Rightarrow x = z \\ y + w = 2w \Rightarrow y = w \end{cases}$$

Therefore the eigenvectors which correspond to the eigenvalue of 2 are of the form

$$\begin{bmatrix} x \\ y \\ x \\ y \end{bmatrix} = x \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} + y \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}. \text{ Therefore two orthonormal eigenvectors are } \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} \text{ and } \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}.$$

The eigenvectors which correspond to the eigenvalue of 0 are obtained from the formula below:

$$\begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{matrix} x+z=0 \Rightarrow z=-x \\ y+w=0 \Rightarrow y=-w \end{matrix}$$

Therefore the eigenvectors which correspond to the eigenvalue of 0 are of the form

$$\begin{bmatrix} x \\ y \\ -x \\ -y \end{bmatrix} = x \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \end{bmatrix} + y \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \end{bmatrix}. \text{ Therefore two orthonormal eigenvectors are } \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \end{bmatrix} \text{ and } \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \end{bmatrix}.$$

$$\text{Therefore, } V = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 & 0 & 0 \\ 0 & \sqrt{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$$