

**Math 1080: Spring 2011**  
**Homework #11 (due April 22)**

**Problem 1:**

a) Determine the matrices  $\mathbf{U}$ ,  $\mathbf{\Sigma}$ ,  $\mathbf{V}$  in the singular value decomposition,  $\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$ , of the following matrices

$$\mathbf{A}_1 = \begin{bmatrix} -1 & -1 & -1 & -1 \\ 1 & 1 & -1 & -1 \\ -1 & 1 & 1 & 0 \end{bmatrix}$$

$$\mathbf{A}_2 = \begin{bmatrix} -2 & 2 & 0 & -2 \\ 2 & -1 & 1 & 3 \\ 0 & 1 & 1 & 1 \\ -2 & 3 & 1 & -1 \end{bmatrix}$$

b) Use the results of a) to find, in each case, the rank of  $\mathbf{A}_i$ , the norm  $\|\mathbf{A}_i\|$  and the basis for  $\text{null}(\mathbf{A}_i)$ .

*SOLUTION:*

a)

Case 1:

$$\mathbf{A}_1 \mathbf{A}_1^T = \begin{bmatrix} 4 & 0 & -1 \\ 0 & 4 & -1 \\ -1 & -1 & 3 \end{bmatrix},$$

$$\begin{aligned} \det(\mathbf{A}_1 \mathbf{A}_1^T - \lambda \mathbf{I}) &= \begin{vmatrix} 4-\lambda & 0 & -1 \\ 0 & 4-\lambda & -1 \\ -1 & -1 & 3-\lambda \end{vmatrix} = (4-\lambda)^2 (3-\lambda) - 2(4-\lambda) \\ &= (4-\lambda)(\lambda^2 - 7\lambda + 10) = (4-\lambda)(\lambda-2)(\lambda-5) \end{aligned}$$

The eigenvalues of  $\mathbf{A}_1 \mathbf{A}_1^T$  are  $\lambda_i = 5, 4$ , and  $2$  and hence the singular values of  $\mathbf{A}_1$  are  $\sqrt{5}, 2$ , and  $\sqrt{2}$  and

$$\mathbf{\Sigma} = \begin{bmatrix} \sqrt{5} & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & \sqrt{2} & 0 \end{bmatrix}$$

For  $\lambda_1 = 5$  the eigenvector of  $\mathbf{A}_1 \mathbf{A}_1^T$  obeys  $\begin{bmatrix} -1 & 0 & -1 \\ 0 & -1 & -1 \\ -1 & -1 & -2 \end{bmatrix} \mathbf{x}_1 = \mathbf{0}$ , i.e.,  $\mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$ .

For  $\lambda_2 = 4$  the eigenvector of  $\mathbf{A}_1 \mathbf{A}_1^T$  obeys  $\begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & -1 \\ -1 & -1 & -1 \end{bmatrix} \mathbf{x}_2 = \mathbf{0}$ , i.e.,  $\mathbf{x}_2 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$ .

For  $\lambda_3 = 2$  the eigenvector of  $\mathbf{A}_1 \mathbf{A}_1^T$  obeys  $\begin{bmatrix} 2 & 0 & -1 \\ 0 & 2 & -1 \\ -1 & -1 & 1 \end{bmatrix} \mathbf{x}_3 = \mathbf{0}$ , i.e.,  $\mathbf{x}_3 = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$ .

The columns of  $\mathbf{U}$  are the normalized eigenvectors  $\mathbf{x}_i$ :  $\mathbf{U} = \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{-1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \end{bmatrix}$

$$\mathbf{A}_1 \mathbf{A}_1^T = \begin{bmatrix} 3 & 1 & -1 & 0 \\ 1 & 3 & 1 & 0 \\ -1 & 1 & 3 & 2 \\ 0 & 0 & 2 & 2 \end{bmatrix}$$

The eigenvalues of  $\mathbf{A}_1 \mathbf{A}_1^T$  are  $\lambda_i = 5, 4, 2$ , and 0

For  $\lambda_1 = 5$  the eigenvector obeys  $\begin{bmatrix} -2 & 1 & -1 & 0 \\ 1 & -2 & 1 & 0 \\ -1 & 1 & -2 & 2 \\ 0 & 0 & 2 & -3 \end{bmatrix} \mathbf{x}_1 = \mathbf{0}$ , i.e.,  $\mathbf{x}_1 = \begin{bmatrix} -1 \\ 1 \\ 3 \\ 2 \end{bmatrix}$

For  $\lambda_2 = 4$  the eigenvector obeys  $\begin{bmatrix} -1 & 1 & -1 & 0 \\ 1 & -1 & 1 & 0 \\ -1 & 1 & -1 & 2 \\ 0 & 0 & 2 & -2 \end{bmatrix} \mathbf{x}_2 = \mathbf{0}$ , i.e.,  $\mathbf{x}_2 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$

For  $\lambda_3 = 2$  the eigenvector obeys  $\begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 1 & 0 \\ -1 & 1 & 1 & 2 \\ 0 & 0 & 2 & 0 \end{bmatrix} \mathbf{x}_3 = \mathbf{0}$ , i.e.,  $\mathbf{x}_3 = \begin{bmatrix} -1 \\ 1 \\ 0 \\ -1 \end{bmatrix}$

For  $\lambda_4 = 0$  the eigenvector obeys  $\begin{bmatrix} 3 & 1 & -1 & 0 \\ 1 & 3 & 1 & 0 \\ -1 & 1 & 3 & 2 \\ 0 & 0 & 2 & 2 \end{bmatrix} \mathbf{x}_4 = \mathbf{0}$ , i.e.,  $\mathbf{x}_4 = \begin{bmatrix} 1 \\ -1 \\ 2 \\ -2 \end{bmatrix}$

The columns of  $\mathbf{V}$  are the normalized eigenvectors  $\mathbf{x}_i$  multiplied by  $\pm 1$  so that

$$\mathbf{A}_1 \mathbf{v}_i = \sigma_i \mathbf{u}_i:$$

$$\mathbf{V} = \begin{bmatrix} \frac{1}{\sqrt{15}} & \frac{-1}{\sqrt{2}} & \frac{-1}{\sqrt{3}} & \frac{1}{\sqrt{10}} \\ \frac{-1}{\sqrt{15}} & \frac{-1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{10}} \\ \frac{-3}{\sqrt{15}} & 0 & 0 & \frac{2}{\sqrt{10}} \\ \frac{-2}{\sqrt{15}} & 0 & \frac{-1}{\sqrt{3}} & \frac{-2}{\sqrt{10}} \end{bmatrix}$$

Case 2:

The matrix  $\mathbf{A}_2$  is symmetric and hence it has the orthogonal decomposition  $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T$

which yields immediately a singular value decomposition with  $\mathbf{U} = \mathbf{Q}$ ,  $\mathbf{\Sigma} = |\mathbf{\Lambda}|$ , and

$$\mathbf{V} = \mathbf{Q} \text{sign}(\mathbf{\Lambda}).$$

$$\begin{aligned} \det(\mathbf{A}_2 \mathbf{A}_2^T - \lambda \mathbf{I}) &= \begin{vmatrix} -2-\lambda & 2 & 0 & -2 \\ 2 & -1-\lambda & 1 & 3 \\ 0 & 1 & 1-\lambda & 1 \\ -2 & 3 & 1 & -1-\lambda \end{vmatrix} \\ &= -(2+\lambda) \begin{vmatrix} -1-\lambda & 1 & 3 \\ 1 & 1-\lambda & 1 \\ 3 & 1 & -1-\lambda \end{vmatrix} - 2 \begin{vmatrix} 2 & 1 & 3 \\ 0 & 1-\lambda & 1 \\ -2 & 1 & -1-\lambda \end{vmatrix} \\ &\quad + 2 \begin{vmatrix} 2 & -1-\lambda & 1 \\ 0 & 1 & 1-\lambda \\ -2 & 3 & 1 \end{vmatrix} \\ &= \lambda^2 (\lambda^2 + 3\lambda - 18) \\ &= \lambda^2 (\lambda - 3)(\lambda + 6) \end{aligned}$$

The eigenvalues are -6, 3, 0 and 0, and hence  $\mathbf{\Sigma} = \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ .

For  $\lambda_1 = -6$  the eigenvector obeys  $\begin{bmatrix} 4 & 2 & 0 & -2 \\ 2 & 5 & 1 & 3 \\ 0 & 1 & 7 & 1 \\ -2 & 3 & 1 & 5 \end{bmatrix} \mathbf{x}_1 = \mathbf{0}$ , i.e.,  $\mathbf{x}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \\ -1 \end{bmatrix}$

For  $\lambda_2 = 3$  the eigenvector obeys  $\begin{bmatrix} -5 & 2 & 0 & -2 \\ 2 & -4 & 1 & 3 \\ 0 & 1 & -2 & 1 \\ -2 & 3 & 1 & -4 \end{bmatrix} \mathbf{x}_2 = \mathbf{0}$ , i.e.,  $\mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}$

For  $\lambda_3 = 0$  the eigenvector obeys  $\begin{bmatrix} -2 & 2 & 0 & -2 \\ 2 & -1 & 1 & 3 \\ 0 & 1 & 1 & 1 \\ -2 & 3 & 1 & -1 \end{bmatrix} \mathbf{x}_3 = \mathbf{0}$ , i.e.,  $\mathbf{x}_3 = \begin{bmatrix} -s+t \\ t \\ -s-t \\ s \end{bmatrix}$ .

We pick, for example  $\mathbf{x}_3 = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}$ ,  $\mathbf{x}_4 = \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$ .

Thus,  $\mathbf{U}_2 = \mathbf{Q} = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 & 0 & 1 & -1 \\ 1 & 1 & 1 & 0 \\ 0 & 1 & -1 & -1 \\ -1 & 1 & 0 & 1 \end{bmatrix}$ ,  $\mathbf{V}_2 = \mathbf{Q} \text{sign}(\mathbf{\Lambda}) = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 0 & 1 & -1 \\ -1 & 1 & 1 & 0 \\ 0 & 1 & -1 & -1 \\ 1 & 1 & 0 & 1 \end{bmatrix}$

b)

$\text{rank}(\mathbf{A}_1) = \# \text{ of nonzero singular values} = 3$ ,  $\|\mathbf{A}_1\| = \sigma_1 = \sqrt{5}$ ,

$\text{null}(\mathbf{A}_1) = \left\langle \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ -1 \\ 2 \\ -2 \end{bmatrix} \right\rangle$

$\text{rank}(\mathbf{A}_2) = 2$ ,  $\|\mathbf{A}_1\| = \sigma_1 = 6$ ,  $\text{null}(\mathbf{A}_1) = \left\langle \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}, \frac{1}{\sqrt{3}} \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix} \right\rangle$

### Computer Assignment 7:

- a) Write a MATLAB function `[U,S,V]=svdsimp(A)` that computes the singular value decomposition,  $\mathbf{A} = \mathbf{U}\mathbf{S}\mathbf{V}^T$ , of an  $m \times n$  matrix  $\mathbf{A}$ . The output variables are the  $m \times m$  orthogonal matrix  $\mathbf{U}$ , the  $m \times n$  diagonal matrix  $\mathbf{S}$ , and the  $n \times n$  orthogonal matrix  $\mathbf{V}$ .
- b) Use the function `svdsimp` to calculate the singular values of the following matrix

$$\mathbf{A} = \begin{bmatrix} -3 & 2 & 6 & -8 & 0 & 5 & 3 \\ 2 & -9 & 5 & 3 & 4 & 4 & 1 \\ 0 & 0 & 4 & 2 & -1 & -1 & -1 \\ 3 & 2 & 0 & -3 & -3 & 1 & 1 \\ 5 & 4 & 4 & 4 & 5 & 8 & 9 \end{bmatrix}$$

### PROGRAM:

```
function [U,S,V] = svdsimp(A)
[m,n] = size(A);
[V,D] = eig(A'*A);
[d,ind] = sort(diag(D), 'descend');
V = V(:,ind);
[U,S] = qr(A*V);
end
```

### OUTPUT:

```
>> A = [3 -2 6 -8 0 5 3; 2 -9 5 3 4 4 1; 0 0 4 2 -1 -1 -1; 3 2 0 -3 -3
1 1; 5 4 4 4 5 8 9];
>> [U,S,V] = svdsimp(A)
```

U =

```
0.4430    -0.1948    -0.8077     0.0022     0.3367
0.4446    -0.7730     0.2944    -0.1111    -0.3252
0.0299    -0.1243     0.0783     0.9862     0.0701
0.0484     0.2300    -0.3952     0.1215    -0.8797
0.7764     0.5442     0.3139     0.0168     0.0463
```

S =

```
17.8860    -0.0000     0.0000    -0.0000     0.0000    -0.0000     0.0000
0    11.3687     0.0000     0.0000     0.0000     0.0000     0.0000
0         0    10.9608    -0.0000     0.0000    -0.0000    -0.0000
0         0         0     4.5280     0.0000     0.0000    -0.0000
0         0         0         0     2.8638     0.0000    -0.0000
```

V =

0.3492	0.1126	-0.1323	0.0515	-0.7151	-0.5075	-0.2761
-0.0942	0.8782	-0.0519	0.2884	0.2373	-0.0098	-0.2788
0.4532	-0.2950	-0.1647	0.7664	0.3002	-0.0550	-0.0017
0.0453	0.0420	0.9072	0.2924	-0.2461	0.1610	0.0356
0.3067	-0.0824	0.3517	-0.3778	0.5235	-0.4808	-0.3571
0.5716	0.0565	-0.0751	-0.2569	-0.0688	0.6866	-0.3496
0.4909	0.3426	0.0203	-0.1806	0.0529	-0.1041	0.7714

>> norm(U\*S\*V'-A)

ans =

1.5274e-014