Assignment 2 Solutions

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Solutions 1

Question 1 1.1

Given the matrix

$$A = \begin{pmatrix} 1 & 2 \\ -1 & 2 \end{pmatrix} \tag{1}$$

1.1.1 Part A

The spectrum of A is found by finding the eigenvalues of A.

$$A - \lambda I = 0 \tag{2}$$

$$A - \lambda I = 0$$

$$\begin{bmatrix} 1 & 2 \\ -1 & 2 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = 0$$

$$\begin{bmatrix} 1 - \lambda & 2 \\ -1 & 2 - \lambda \end{bmatrix} = 0$$
(2)
(3)

$$\begin{bmatrix} 1 - \lambda & 2 \\ -1 & 2 - \lambda \end{bmatrix} = 0 \tag{4}$$

$$((1-\lambda)(2-\lambda)) + 2 = 0$$

$$\lambda^2 - 3\lambda + 4 = 0$$
(5)
(6)

$$\lambda^2 - 3\lambda + 4 = 0 \tag{6}$$

Now we simply factor to find the eigenvalues.

$$\lambda = \frac{3 \pm \sqrt{-7}}{2} \tag{7}$$

$$\lambda = \frac{3 \pm \sqrt{7}i}{2} \tag{8}$$

Therefore the spectrum of A,

$$\sigma(A) = \frac{3 \pm \sqrt{7}i}{2} \tag{9}$$

The spectral radius of A is,

$$p(A) = \max\{\sigma(A)\}$$
 (10)

$$\lambda = \frac{3 - \sqrt{7}i}{2} \tag{12}$$

and converting to the reals (13)

$$= \sqrt{\left(\frac{3 \pm \sqrt{7}i}{2}\right)^2}$$

$$= \sqrt{\frac{9}{4} + \frac{7}{4}}$$

$$= \sqrt{\frac{16}{4}}$$

$$= \sqrt{4}$$
(13)
(14)
(15)
(16)
(17)

$$= \sqrt{\frac{9}{4} + \frac{7}{4}} \tag{15}$$

$$= \sqrt{\frac{16}{4}} \tag{16}$$

$$= \sqrt{4} \tag{17}$$

$$p(A) = 2 \tag{18}$$

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Listing 1: Matlab Commands
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```
A = [1,2;-1,2];
eig(A)
```

ans =

$$1.5000 + 1.3229 i$$

 $1.5000 - 1.3229 i$

abs(ans)

ans =

2.0000

2.0000

diary off

1.1.2 Part B

$$||A||_1 = \max\{2,4\} = \boxed{4} \tag{19}$$

$$||A||_{\inf} = \max\{3,3\} = \boxed{3}$$
 (20)

$$||A||_2 = \sqrt{\lambda_{\text{max}}(A'A)} \tag{21}$$

First we find the spectrum of A'A,

$$A'A = \begin{bmatrix} 1 & -1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -1 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 0 \\ 0 & 8 \end{bmatrix}$$
(22)

The eigenvalues are simply the diagonal entries meaning,

$$\sigma(A'A) = \{2,8\} \tag{24}$$

Therefore,
$$||A||_2 = \sqrt{\max\{2,8\}}$$
 (25)

$$||A||_2 = \boxed{\sqrt{8}} \tag{26}$$

1.1.3 Part C

The left singular vectors of A are eigenvectors of AA'; we will denote this U. The right singular vectors of A are eigenvectors of A'A; we will denote this V. The singular values of A are the square roots of the non-zero eigenvalues of A'A and AA'; we will denote this S.

Note,

$$AA' = \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix}, A'A = \begin{bmatrix} 2 & 0 \\ 0 & 8 \end{bmatrix}$$
 (27)

First, we will find the left singular vectors of A. The eigenvalues of AA' are $\lambda = 2,8$. We will now find eigenvectors for AA'.

Case: $\lambda = 2$,

$$\begin{bmatrix} 3 & 3 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \tag{28}$$

A solution:
$$X = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$
 (29)

Case: $\lambda = 8$,

$$\begin{bmatrix} -3 & 3 \\ 3 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \tag{30}$$

A solution:
$$X = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 (31)

Therefore,

$$U = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \tag{32}$$

Next, we will find the right singular vectors of A. The eigenvalues of A'A are $\lambda = 2.8$ We will now find eigenvectors for A'A.

Case: $\lambda = 2$,

$$\begin{bmatrix} 0 & 0 \\ 0 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \tag{33}$$

A solution:
$$X = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$
 (34)

Case: $\lambda = 8$,

$$\begin{bmatrix} -6 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \tag{35}$$

A solution:
$$X = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
 (36)

Therefore,

$$V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \tag{37}$$

The singular values of A are the square roots of the eigenvalues of AA' and A'A. Therefore,

$$S = \begin{bmatrix} \sqrt{2} & 0\\ 0 & \sqrt{8} \end{bmatrix} \tag{38}$$

The singular value decomposition of A is given by A = USV'.

$$A = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{8} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
 (39)

$$A = \begin{bmatrix} 1.4142 & 2.8284 \\ -1.4142 & 2.8282 \end{bmatrix} \tag{40}$$

The matrix above is in fact not A, but it is only off by a scalar. If we divide by 1.4142 we get the original matrix A. Therefore we will encorporate this scalar division into U. Now,

$$U = \begin{bmatrix} 0.7071 & 0.7071 \\ -0.7071 & 0.7071 \end{bmatrix}$$
 (41)

In full,

$$U = \begin{bmatrix} 0.7071 & 0.7071 \\ -0.7071 & 0.7071 \end{bmatrix}, S = \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{8} \end{bmatrix}, V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
(42)

Listing 3: Matlab Commands

$$[U,S,V] = svd(A)$$

$$\begin{array}{ccc} 0.7071 & -0.7071 \\ 0.7071 & 0.7071 \end{array}$$

$$S =$$

$$\begin{array}{ccc} 2.8284 & & 0 \\ 0 & & 1.4142 \end{array}$$

$$\begin{array}{cc} 0 & -1 \\ 1 & 0 \end{array}$$

1.2 **Question 2**

1.2.1 Part A

The example symmetric matrix is,

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 5 & 7 & 2 \\ 3 & 7 & 2 & 9 \\ 4 & 2 & 9 & 4 \end{bmatrix} \tag{43}$$

The eigenvalues and eigenvectors for A are thus,

$$\lambda = \begin{bmatrix}
17.4770 & 0 & 0 & 0 \\
0 & -7.5267 & 0 & 0 \\
0 & 0 & 3.0000 & 0 \\
0 & 0 & 0 & -0.9502
\end{bmatrix}$$

$$V = \begin{bmatrix}
0.3043 & 0.0760 & 0.2673 & 0.9112 \\
0.4769 & -0.3446 & -0.8018 & 0.1047 \\
0.6022 & 0.7534 & -0.0000 & -0.2640 \\
0.5633 & -0.5549 & 0.5345 & -0.2986
\end{bmatrix}$$
(45)

$$V = \begin{pmatrix} 0.3043 & 0.0760 & 0.2673 & 0.9112 \\ 0.4769 & -0.3446 & -0.8018 & 0.1047 \\ 0.6022 & 0.7534 & -0.0000 & -0.2640 \\ 0.5633 & -0.5549 & 0.5345 & -0.2986 \end{pmatrix}$$
(45)

And the eigenvalues and eigenvectors for A'A are thus,

$$\lambda = \begin{bmatrix} 305.4452 & 0 & 0 & 0 \\ 0 & 56.6519 & 0 & 0 \\ 0 & 0 & 9.0000 & 0 \\ 0 & 0 & 0 & 0.9030 \end{bmatrix}$$

$$V = \begin{bmatrix} 0.3043 & -0.0760 & 0.2673 & -0.9112 \\ 0.4769 & 0.3446 & -0.8018 & -0.1047 \\ 0.6022 & -0.7534 & 0.0000 & 0.2640 \\ 0.5633 & 0.5549 & 0.5345 & 0.2986 \end{bmatrix}$$

$$(46)$$

$$V = \begin{bmatrix} 0.3043 & -0.0760 & 0.2673 & -0.9112 \\ 0.4769 & 0.3446 & -0.8018 & -0.1047 \\ 0.6022 & -0.7534 & 0.0000 & 0.2640 \\ 0.5633 & 0.5549 & 0.5345 & 0.2986 \end{bmatrix}$$
(47)

Therefore, it is clear from this example that the eigenvectors for A and A'A are the same and the eigenvalues of A'A are the eigenvalues of A squared. The conjecture is thus, (λ, ν) is an eigenpair of A'A if and only if $(\sqrt{\lambda}, \nu)$ is an eigenpair of A; (for symmetric A).

```
Listing 4: Matlab Commands
A = [1,2,3,4;2,5,7,2;3,7,2,9;4,2,9,4];
[P1,D1] = eigs(A)
P1 =
    0.3043
               0.0760
                           0.2673
                                      0.9112
    0.4769
              -0.3446
                          -0.8018
                                      0.1047
    0.6022
                          -0.0000
               0.7534
                                     -0.2640
    0.5633
              -0.5549
                           0.5345
                                     -0.2986
D1 =
   17.4770
                                0
                                           0
                     0
              -7.5267
                                           0
          0
                                0
          0
                           3.0000
                                           0
          0
                     0
                                     -0.9502
                                0
[P2,D2] = eigs(A'*A)
P2 =
    0.3043
             -0.0760
                           0.2673
                                     -0.9112
    0.4769
               0.3446
                          -0.8018
                                     -0.1047
    0.6022
                           0.0000
                                      0.2640
              -0.7534
    0.5633
               0.5549
                           0.5345
                                      0.2986
D2 =
  305.4452
                                0
                                           0
                     0
          0
              56.6519
                                0
                                           0
          0
                     0
                           9.0000
                                           0
          0
                     0
                                      0.9030
                                0
diary off
```

1.2.2 Part B

Since *A* is a symmetric matrix we know that *A* can be decomposed to $A = PDP^{-1}$. Where the columns of P are the eigenvectors of A and D is a diagonal matrix where the diaganol entries are the eigenvalues of A. The proof is as follows,

$$A = PDP^{-1} (48)$$

$$A^{2} = PDP^{-1}PDP^{-1}$$

$$= PD(P^{-1}P)P^{-1}$$

$$= PDDP^{-1}$$

$$= PDDP^{-1}$$

$$= PD^{2}P^{-1}$$
(50)
(51)
(52)
(53)

$$= PD(P^{-1}P)P^{-1} (51)$$

$$= PDDP^{-1} (52)$$

$$= PD^2P^{-1} (53)$$

We know that $A^2 = A'A$. The columns of P are the eigenvectors of A'A and D^2 is a diagonal matrix where the diagonal entries are the eigenvalues of A'A. Therefore, the eigenvalues of A'A are the eigenvalues of A squared. This concludes the proof.

1.2.3 Part C

Theorem: The 2-norm of a symmetric matrix A is as follows,

$$||A||_2 = p(A) \tag{54}$$

Where,
$$p(A) := \max\{|\lambda| : \lambda \in \sigma(A)\}$$
 (55)

We must check the theorm against,

$$B = \begin{bmatrix} -92 & 144 \\ 144 & -8 \end{bmatrix} \tag{56}$$

The eigenvalues of B are -200 and 100. Therefore, $\sigma(B) = \{-200, 100\}$ and p(B) = 200. Using Matlab we find $||B||_2 = 200$. Therefore, the theorem holds for

Listing 5: Matlab Commands

$$B = [-92,144;144,-8];$$

```
eigs (B)

ans =

-200.0000
100.0000

abs (ans)

ans =

200.0000
100.0000
norm(B,2)

ans =

200
diary off
```

1.2.4 Part D

This theorem does not hold for the non-symmetric matrix,

$$C = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \tag{57}$$

We find $\sigma(C) = \{5.3723, -0.3723\}$ and p(C) = 5.3723. Using Matlab we find $||C||_2 = 5.4650$. Therefore, the theorem does not hold for C.

```
Listing 6: Matlab Commands

C = [1,2;3,4];
eigs(C)

ans =
```

```
5.3723

-0.3723

abs(ans)

ans =

5.3723

0.3723

norm(C,2)

ans =

5.4650

diary off
```

1.2.5 Part E

The singular values of a symmetric matrix A are the absolute values of the non-zero eigenvalues of A.

1.3 Question 3

1.3.1 Part A

We want to prove that (λ, ν) is an eigenpair of A if and only if $(\frac{1}{\lambda}, \nu)$ is an eigenpair of A^{-1} . The proof is as follows,

$$Av = \lambda v \tag{58}$$

$$A^{-1}Av = A^{-1}\lambda v (59)$$

$$(A^{-1}A)v = A^{-1}\lambda v (60)$$

$$Iv = A^{-1}\lambda v \tag{61}$$

$$v = A^{-1}\lambda v \tag{62}$$

$$\frac{1}{\lambda}v = A^{-1}v \tag{63}$$

The 'only if' portion follows by symmetry.

1.3.2 Part B

The 2-norm of A^{-1} should be 1 / (the smallest singular value of A). We will denote the singular values of A as s(A). Stated formally,

$$||A^{-1}||_2 = \max_{s_A \in s(A)} \left\{ \frac{1}{s_a} \right\} \tag{64}$$

This is true because we know from (a) that (λ, ν) is an eigenpair for A if and only if $(\frac{1}{\lambda}, \nu)$ is an eigenpair of A^{-1} . Therefore, it must be the case that (λ, ν) is an eigenpair for A'A if and only if $(\frac{1}{\lambda}, \nu)$ is an eigenpair of $(A^{-1})'A^{-1}$ because the 2-norm is preserved. Since the λ 's of A'A are the singular values of A it follows that $\frac{1}{\lambda}$'s of A'A are the singular values of A^{-1} . Therefore, since $||A||_2 = \max\{s(A)\}$,

$$||A^{-1}||_2 = \max_{s_A \in s(A)} \left\{ \frac{1}{s_a} \right\}$$
 (65)

1.4 Question 4

QR factor the matrix Z where matrix Z is,

$$Z = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 7 \\ 4 & 2 & 3 \\ 4 & 2 & 2 \end{bmatrix}$$
 (66)

This will require three iterations. Iteration 1:

$$x = \begin{bmatrix} 1 \\ 4 \\ 7 \\ 4 \\ 4 \end{bmatrix}, y = \begin{bmatrix} 9.8995 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, w = \begin{bmatrix} -0.6704 \\ 0.3013 \\ 0.5273 \\ 0.3013 \\ 0.3013 \end{bmatrix}$$

$$\begin{bmatrix} 0.1010 & 0.4041 & 0.7071 & 0.4041 & 0.4041 \end{bmatrix}$$

$$\begin{bmatrix} 0.7071 & 0.4041 & 0.4041 & 0.4041 \end{bmatrix}$$

$$H = \begin{bmatrix} 0.1010 & 0.4041 & 0.7071 & 0.4041 & 0.4041 \\ 0.4041 & 0.8184 & -0.3178 & -0.1816 & -0.1816 \\ 0.7071 & -0.3178 & 0.4438 & -0.3178 & -0.3178 \\ 0.4041 & -0.1816 & -0.3178 & 0.8184 & -0.1816 \\ 0.4041 & -0.1816 & -0.3178 & -0.1816 & 0.8184 \end{bmatrix}$$
(68)

$$Q = \begin{bmatrix} 0.1010 & 0.4041 & 0.7071 & 0.4041 & 0.4041 \\ 0.4041 & 0.8184 & -0.3178 & -0.1816 & -0.1816 \\ 0.7071 & -0.3178 & 0.4438 & -0.3178 & -0.3178 \\ 0.4041 & -0.1816 & -0.3178 & 0.8184 & -0.1816 \\ 0.4041 & -0.1816 & -0.3178 & -0.1816 & 0.8184 \end{bmatrix}$$
(69)

$$R = \begin{bmatrix} 9.8995 & 9.4954 & 9.6975 \\ -0.0000 & 1.6311 & 2.9897 \\ -0.0000 & 2.1044 & 1.7320 \\ -0.0000 & -1.3689 & -0.0103 \\ -0.0000 & -1.3689 & -1.0103 \end{bmatrix}$$

$$(70)$$

Iteration 2:

$$x = \begin{bmatrix} 9.4954 \\ 1.6311 \\ 2.1044 \\ -1.3689 \\ -1.3689 \end{bmatrix}, y = \begin{bmatrix} 9.4954 \\ 3.2919 \\ 0 \\ 0 \\ 0 \end{bmatrix}, w = \begin{bmatrix} 0 \\ -0.5023 \\ 0.6364 \\ -0.4140 \\ -0.4140 \end{bmatrix}$$

$$H = \begin{bmatrix} 1.0000 & 0 & 0 & 0 & 0 \\ 0 & 0.4955 & 0.6393 & -0.4158 & -0.4158 \\ 0 & 0.6393 & 0.1900 & 0.5269 & 0.5269 \\ 0 & -0.4158 & 0.5269 & 0.6572 & -0.3428 \\ 0 & -0.4158 & 0.5269 & -0.3428 & 0.6572 \end{bmatrix}$$

$$Q = \begin{bmatrix} 0.1010 & 0.3162 & 0.8185 & 0.3316 & 0.3316 \\ 0.4041 & 0.3534 & 0.2714 & -0.5649 & -0.5649 \\ 0.7071 & 0.3906 & -0.4537 & 0.2661 & 0.2661 \\ 0.4041 & -0.5580 & 0.1590 & 0.5082 & -0.4918 \\ 0.4041 & -0.5580 & 0.1590 & -0.4918 & 0.5082 \end{bmatrix}$$

$$R = \begin{bmatrix} 9.8995 & 9.4954 & 9.6975 \\ -0.0000 & 3.2919 & 3.0129 \\ -0.0000 & -0.0000 & 1.7026 \\ -0.0000 & 0.0000 & 0.0089 \end{bmatrix}$$

$$(74)$$

-0.9911

Iteration 3:

-0.0000

0.0000

$$x = \begin{bmatrix} 9.6975 \\ 3.0129 \\ 1.7026 \\ 0.0089 \\ -0.9911 \end{bmatrix}, y = \begin{bmatrix} 9.6975 \\ 3.0129 \\ 1.9701 \\ 0 \\ 0 \end{bmatrix}, w = \begin{bmatrix} 0 \\ 0 \\ -0.2606 \\ 0.0086 \\ -0.9654 \end{bmatrix}$$
(75)

$$H = \begin{bmatrix} 1.0000 & 0 & 0 & 0 & 0 \\ 0 & 1.0000 & 0 & 0 & 0 \\ 0 & 0 & 0.8642 & 0.0045 & -0.5031 \\ 0 & 0 & 0.0045 & 0.9999 & 0.0167 \\ 0 & 0 & -0.5031 & 0.0167 & -0.8641 \end{bmatrix}$$
(76)

$$Q = \begin{bmatrix} 0.1010 & 0.3162 & 0.5420 & 0.3408 & -0.6928 \\ 0.4041 & 0.3534 & 0.5162 & -0.5730 & 0.3422 \\ 0.7071 & 0.3906 & -0.5248 & 0.2684 & 0.0028 \\ 0.4041 & -0.5580 & 0.3871 & 0.5006 & 0.3534 \\ 0.4041 & -0.5580 & -0.1204 & -0.4825 & -0.5273 \end{bmatrix}$$
(77)

$$R = \begin{bmatrix} 9.8995 & 9.4954 & 9.6975 \\ -0.0000 & 3.2919 & 3.0129 \\ -0.0000 & -0.0000 & 1.9701 \\ -0.0000 & 0.0000 & -0.0000 \\ 0.0000 & -0.0000 & 0.0000 \end{bmatrix}$$
(78)

Therefore the QR factorization of Z is

$$Q = \begin{bmatrix} 0.1010 & 0.3162 & 0.5420 & 0.3408 & -0.6928 \\ 0.4041 & 0.3534 & 0.5162 & -0.5730 & 0.3422 \\ 0.7071 & 0.3906 & -0.5248 & 0.2684 & 0.0028 \\ 0.4041 & -0.5580 & 0.3871 & 0.5006 & 0.3534 \\ 0.4041 & -0.5580 & -0.1204 & -0.4825 & -0.5273 \end{bmatrix}$$
(79)

$$R = \begin{bmatrix} 9.8995 & 9.4954 & 9.6975 \\ -0.0000 & 3.2919 & 3.0129 \\ -0.0000 & -0.0000 & 1.9701 \\ -0.0000 & 0.0000 & -0.0000 \\ 0.0000 & -0.0000 & 0.0000 \end{bmatrix}$$
(80)

```
Listing 7: Matlab Commands
Z=[1,2,3;4,5,6;7,8,7;4,2,3;4,2,2];
R=Z;
% Iteration 1
x=R(:,1);
a=norm(x(1:5),2);
y=[x(1:0)' a zeros(1,4)]';
w=(x-y)/norm(x-y, 2);
H = eye(5) - 2*w*w';
Q=eye(5)*H;
R=H*R;
% Iteration 2
x=R(:,2);
a = norm(x(2:5), 2);
y=[x(1:1)' a zeros(1,3)]';
w=(x-y)/norm(x-y, 2);
H = eye(5) - 2*w*w';
O=O*H;
R=H*R;
% Iteration 3
x=R(:,3);
a = norm(x(3:5), 2);
y=[x(1:2)' a zeros(1,2)]';
w=(x-y)/norm(x-y, 2);
H = eye(5) - 2*w*w';
Q=Q*H;
R=H*R;
```

1.5 Question 5