22S-CSB150-EEB159 Lab2

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TOTAL POINTS

97 / 100

QUESTION 3

```
3 Ex.2b 10 / 10
QUESTION 1

√ - 0 pts Correct

1 Ex.1 20 / 20
                                                           - 5 pts 'eig(af) = -3.0000'

√ - 0 pts Correct

   - 5 pts Your final answer is correct, but I need to
                                                       OUESTION 4
  see your working
                                                       4 Ex.3 47 / 50
   - 5 pts Condition for both eigenvalues to be
                                                           - 0 pts Correct
  negative is: $$b_1^2 > b_2 b_3$$
                                                           - 5 pts Create `5 x 5` matrix and random vector:
   - 5 pts Condition for one positive and one
                                                          ```matlab
 negative eigenvalue is: $$b_1^2 < b_2 b_3$$
 A = rand(5);
 - 2 pts Need to simplify fully
 A = A .* A'; % random symmetric matrix
 vold1=rand(5,1); % a random vector
QUESTION 2
 vold2=rand(5,1); % another random vector
2 Ex.2a 20 / 20
 II1 = []; % place holder for each iteration for first

√ - 0 pts Correct

 random vector
 - 10 pts Code for the 6 iterations should be:
 II2 = []; % place holder for each iteration for
  ```matlab
                                                          second random vector
  v=[1;2;0;3;1];
  u1=af*v/norm(af*v);
                                                           - 10 pts Correctly write `for` loop to iterate and
  u2=af*u1/norm(af*u1);
                                                         find largest eigenvalue:
  u3=af*u2/norm(af*u2);
                                                         ```matlab
 u4=af*u3/norm(af*u3);
 for ii = 1:100
 u5=af*u4/norm(af*u4);
 vnew1=A*vold1/norm(A*vold1);
 u6=af*u5/norm(af*u5);
 II1(ii) = double(vnew1'*A*vnew1);
 u6'*af*u6
 vold1=vnew1;
 vnew2=A*vold2/norm(A*vold2);
 - 10 pts `u6' * af * u6 = -3.0000`
 II2(ii) = double(vnew2'*A*vnew2);
```

vold2=vnew2;

...

- 5 pts Verify results with `eig()`
- 10 pts Repeat process with a \*\*different\*\*

random vector

- **10 pts** Compare results for the two initial vectors
- 3 pts Correct answer: rate of convergence depends on initial vector. It may not look this way if you chose simple matrices and similar vectors or if you look at all 200 iterations in one plot.
- 10 pts Make plot of eigenvalue estimates for

both cases:

```matlab

figure

hold on

plot(ll1,'r--o')

plot(ll2,'g-*')

xlabel('iteration number')

ylabel('estimated leading eigenvalue')

legend(strcat(",num2str(vold1')),strcat(",num2str(v

old2')))

hold off

...

- √ 3 pts Axes labels and title
 - 50 pts not attempted

```
syms b1 b2 b3 b4 b5 b6 b7 b8 b9 b10 b11 b12 b13 b14 b15 b16 eig([b1, b2; b3, b4])
```

ans =

$$\left(\frac{b_1}{2} + \frac{b_4}{2} - \frac{\sqrt{b_1^2 - 2b_1b_4 + b_4^2 + 4b_2b_3}}{2}\right) \\
\frac{b_1}{2} + \frac{b_4}{2} + \frac{\sqrt{b_1^2 - 2b_1b_4 + b_4^2 + 4b_2b_3}}{2}\right)$$

eig([-b1, b2; b3, -b1])

ans =

$$\begin{pmatrix} -b_1 - \sqrt{b_2 b_3} \\ \sqrt{b_2 b_3} - b_1 \end{pmatrix}$$

%exericse 1 solution: -b1 - sqrt(b2b3) <0, so both eigenvalues will be
%negative when sqrt(b2*b3) < b1. One eigenvalue will be negative and the
%other will be positive when sqrt(b2*b3) > b1
af = [-1, 1.5; 1.5, -1];
eig(af)

ans = 2×1 -2.5000 0.5000

```
% one + one -

af = [-1, 0.5; 0.5, -1];
eig(af)
```

ans = 2×1 -1.5000 -0.5000

```
eig2 = @(b1, b2, b3, b4) b1./2 + b4./2 - (b1.^2 - 2*b1.*b4 + 4*b2.*b3 + b4.^2)^(1/2)/2;
eig([b1,b2,b3; b2,b4,b5; b3,b5,b6])
```

ans =

$$\begin{pmatrix}
\frac{b_1}{3} + \frac{b_4}{3} + \frac{b_6}{3} + \frac{\sigma_3}{\sigma_2} + \sigma_2 \\
\frac{b_1}{3} + \frac{b_4}{3} + \frac{b_6}{3} - \frac{\sigma_3}{2\sigma_2} - \frac{\sigma_2}{2} - \sigma_1 \\
\frac{b_1}{3} + \frac{b_4}{3} + \frac{b_6}{3} - \frac{\sigma_3}{2\sigma_2} - \frac{\sigma_2}{2} + \sigma_1
\end{pmatrix}$$

where

$$\sigma_1 = \frac{\sqrt{3} \left(\frac{\sigma_3}{\sigma_2} - \sigma_2\right) i}{2}$$

$$\sigma_2 = \left(\sqrt{\left(\sigma_4 + \sigma_8 + \sigma_7 + \sigma_6 - \sigma_5 - b_2 \, b_3 \, b_5 - \frac{b_1 \, b_4 \, b_6}{2}\right)^2 - \sigma_3^3} - \sigma_8 - \sigma_7 - \sigma_6 - \sigma_4 + \sigma_5 + b_2 \, b_3 \, b_5 + \frac{b_1 \, b_4 \, b_6}{2}\right)^2$$

$$\sigma_3 = \frac{b_2^2}{3} - \frac{b_1 b_6}{3} - \frac{b_4 b_6}{3} - \frac{b_1 b_4}{3} + \frac{b_3^2}{3} + \frac{b_5^2}{3} + \frac{(b_1 + b_4 + b_6)^2}{9}$$

$$\sigma_4 = \frac{(b_1 + b_4 + b_6) \left(-b_2^2 - b_3^2 - b_5^2 + b_1 b_4 + b_1 b_6 + b_4 b_6\right)}{6}$$

$$\sigma_5 = \frac{(b_1 + b_4 + b_6)^3}{27}$$

$$\sigma_6 = \frac{b_2^2 b_6}{2}$$

$$\sigma_7 = \frac{b_3^2 b_4}{2}$$

$$\sigma_8 = \frac{b_1 b_5^2}{2}$$

% a much more complex answer here

%eig([b1, b2, b3, b4; b2, b5, b6, b7; b3, b6, b8, b9; b4, b7, b9, b10])

ans =
$$b_1 b_4 - b_2 b_3 - b_1 c - b_4 c + c^2$$

```
p = [1 - (b1 + b4) b1*b4-b2*b3];
r = roots(p);
p2 = []; %write the coefficients for your 5th-rder polynomials here.
r = roots(p2);
af=[-1, 0.5, 0.5, 0.5, 0.5;
0.5, -1, 0.5, 0.5, 0.5;
0.5, 0.5, -1, 0.5, 0.5;
0.5, 0.5, 0.5, -1, 0.5;
0.5, 0.5, 0.5, 0.5, -1];
eig(af)
ans = 5 \times 1
  -1.5000
  -1.5000
  -1.5000
  -1.5000
   1.0000
% these 0.5s are coefficients for my 5th order polynomial, matrix vector
% form, theorem does not apply
af=[-1, -0.5, -0.5, -0.5, -0.5;
-0.5, -1, -0.5, -0.5, -0.5;
-0.5, -0.5, -1, -0.5, -0.5;
-0.5, -0.5, -0.5, -1, -0.5;
-0.5, -0.5, -0.5, -0.5, -1];
eig(af)
ans = 5 \times 1
  -3.0000
  -0.5000
  -0.5000
  -0.5000
  -0.5000
m = [1, 10; 10, 1];
V = [1;1];
i1 = m*v / norm(m*v)
i1 = 2 \times 1
   0.7071
   0.7071
i1'*m*i1
ans = 11.0000
eig(m)
```

ans = 2×1

```
-9.0000
11.0000
```

```
m = [2, 4;7,3];
v = [1;1];
i1 = m*v / norm(m*v);
c(1) = double(i1'*m*i1)
c =
129
17
i2=m*i1/norm(m*i1);
c(2)=double(i2'*m*i2)
c =
\left(\frac{129}{17} \quad \frac{3884}{493}\right)
i3=m*i2/norm(m*i2);
c(3)=double(i3'*m*i3)
c =
\left(\frac{129}{17} \quad \frac{3884}{493} \quad \frac{1096313283462169}{140737488355328}\right)
i4=m*i3/norm(m*i3);
c(4)=double(i4'*m*i4)
c =
 (129 3884 1096313283462169 275277734408835)
\overline{17} \overline{493}
                                  35184372088832
             140737488355328
i5=m*i4/norm(m*i4);
c(5)=double(i5'*m*i5)
c =
\left(\frac{129}{17}\right)
      3884 1096313283462169 275277734408835 8795377287796131
             <u>140737488355328</u> <u>35184372088832</u> <u>1125899906842624</u>
       493
i6=m*i5/norm(m*i5);
c(6)=double(i6'*m*i6)
c =
 129 3884 1096313283462169 275277734408835 8795377287796131
                                                                           2200071254027361
       493
             140737488355328 35184372088832 1125899906842624 281474976710656
i7=m*i6/norm(m*i6);
c(7) = double(i7'*m*i7)
```

c =

```
\left(\frac{129}{17} \ \frac{3884}{493} \ \frac{1096313283462169}{140737488355328} \ \frac{275277734408835}{35184372088832} \ \frac{8795377287796131}{1125899906842624} \ \frac{2200071254027361}{281474976710656} \ \frac{21996}{2814} \right)
```

i8=m*i7/norm(m*i7);
c(8)=double(i8'*m*i8)

c =

i9=m*i8/norm(m*i8);
c(9)=double(i9'*m*i9)

c =

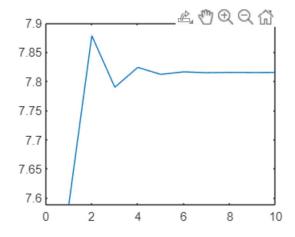
 $\left(\frac{129}{17} \ \frac{3884}{493} \ \frac{1096313283462169}{140737488355328} \ \frac{275277734408835}{35184372088832} \ \frac{8795377287796131}{1125899906842624} \ \frac{2200071254027361}{281474976710656} \ \frac{21996}{2814} \right)$

i10=m*i9/norm(m*i9); c(10)=double(i10'*m*i10)

c =

$$\left(\frac{129}{17} \ \frac{3884}{493} \ \frac{1096313283462169}{140737488355328} \ \frac{275277734408835}{35184372088832} \ \frac{8795377287796131}{1125899906842624} \ \frac{2200071254027361}{281474976710656} \ \frac{21996}{2814} \right)$$

% the outputs eigenvalues are infinitely closing into the value 7.78977 or % approximately, this is called convergence plot(c)



% from the plot on the right, we can see that the eigenvalue vector

% undulates then converges around 7.7-7.8 value range

%exercise 2

```
af = [-1, -0.5, -0.5, -0.5, -0.5]
    -0.5, -1, -0.5, -0.5, -0.5;
    -0.5, -0.5, -1, -0.5, -0.5;
    -0.5, -0.5, -0.5, -1, -0.5;
    -0.5, -0.5, -0.5, -0.5, -1];
v = [1;2;0;3;1];
u1 = af*v/norm(af*v);
u2 = af*u1/norm(af*u1);
u3 = af*u2/norm(af*u2);
u4 = af*u3/norm(af*u3);
u5 = af*u4/norm(af*u4);
u6 = af*u5/norm(af*u5);
u1'*af*u1
ans = -2.9637
u2'*af*u2
ans = -2.9990
u3'*af*u3
ans = -3.0000
u4'*af*u4
ans = -3.0000
u5'*af*u5
ans = -3.0000
u6'*af*u6
ans = -3.0000
eig(af)
ans = 5 \times 1
  -3.0000
  -0.5000
  -0.5000
  -0.5000
  -0.5000
eigtrue = eig(af);
difference = abs(u6'*af*u6) - max(abs(eigtrue));
%answering the question in 2b: how close is the largest eigenvalue from the
%calculated approximation in magnitude?
%
```

```
% The substraction shows that the result is a miniscule number, although % not zero, so we can conclude that the approximation is very close but not % the same.
```

```
% exercise 3
af = rand(5,5);

v= rand(5,1);

u1 = af*v/norm(af*v);
%u = size(100);
v2 = rand(5,1);
array = [];

for i = 1:100
    % u(i+1) = af*u(i) / norm(af*u(i));
    v2 = af*v/norm(af*v);
    array(i) = v2'* af *v2;
    % u(i)' * af*u(i);
    v = v2;
end

eig(af)
```

```
ans = 5×1 complex

2.0607 + 0.0000i

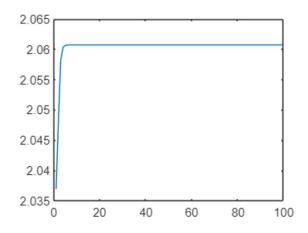
-0.1945 + 0.0000i

0.2540 + 0.0000i

0.1031 + 0.0542i

0.1031 - 0.0542i
```

plot(array);



```
%next, construct a new random vector
af = rand(5,5);
v = rand(5,1);
```

```
u1 = af*v/norm(af*v);
%u = size(100);
v2 = rand(5,1);
array = [];

for i = 1:100
    % u(i+1) = af*u(i) / norm(af*u(i));
    v2 = af*v/norm(af*v);
    array(i) = v2'* af *v2;
    % u(i)' * af*u(i);
    v = v2;
end

eig(af)
```

```
ans = 5×1 complex

2.2663 + 0.0000i

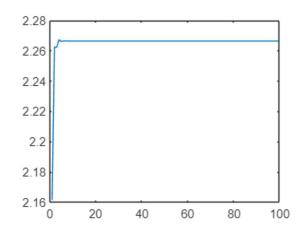
-0.7731 + 0.0000i

0.1610 + 0.1571i

0.1610 - 0.1571i

-0.3282 + 0.0000i
```

plot(array);



 $\mbox{\ensuremath{\mbox{\%}}\xspace}$ plot of eigenvalue estimates for all iterations to look at convergence $\mbox{\ensuremath{\mbox{\%}}\xspace}$ in both cases

%Answer to the question in 3: in both cases, because I was told that af %should be a random 5x5 matrix, the convergence goes to different values, %so it is appr. 2. 06 in the first and 2.26 in the second. The initial guess %does not affect the convergence because eventually over many iterations, %the true eigenvalue will be aligned to the estimation. The convergence %value should be the same, but the rate of convergence may differ depending %on the initial choice of the vector v.

%If you choose a non-symmetric matrix, and your eigenvalue with largest real part is a complex %happens as you iterate? Note: upper bound of oscillations is magnitude of largest eigenvalue. %Answer: After iteration, the complex conjugate pair does not take on the

%imaginary part (c1*i) where c1 denotes the coefficient of the imaginary %part.