# EM316 - Numerical Methods for EEE

Problem Sheet 3

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### Question 1

Given that,

$$A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \quad X_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

 $\mathbf{a}$ 

Iteration 1,

$$AX_0 = \lambda X_1$$

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ -0.5 \end{bmatrix}$$
(1)

Iteration 2,

$$AX_1 = \lambda X_2$$

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -0.5 \end{bmatrix} = \begin{bmatrix} 2.5 \\ -2 \end{bmatrix} = 2.5 \begin{bmatrix} 1 \\ -0.8 \end{bmatrix}$$
(2)

Iteration 3,

$$AX_2 = \lambda X_3$$

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -0.8 \end{bmatrix} = \begin{bmatrix} 2.8 \\ -2.6 \end{bmatrix} = 2.8 \begin{bmatrix} 1 \\ -0.929 \end{bmatrix}$$
(3)

For an approximate eigenvector of A,

$$X = \begin{bmatrix} 1 \\ -0.929 \end{bmatrix}$$

Thus, the approximate dominant eigenvalue,

$$\lambda_{max} = 2.8$$

#### b

Considering the final result,

$$X_3 = \begin{bmatrix} 1 \\ -0.929 \end{bmatrix}$$

Approximate Dominant eigenvalue using Rayleigh Quotient,

$$\lambda = \frac{X^T A X}{X^T X}$$

$$= \frac{\begin{bmatrix} 1 & -0.929 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -0.929 \end{bmatrix}}{\begin{bmatrix} 1 & -0.929 \end{bmatrix} \begin{bmatrix} 1 \\ -0.929 \end{bmatrix}}$$

$$= \frac{\begin{bmatrix} 1 & -0.929 \end{bmatrix} \begin{bmatrix} 2.929 \\ -2.858 \end{bmatrix}}{1.863041}$$

$$= \frac{5.584082}{1.863041}$$

$$= 2.9973$$

 $\mathbf{c}$ 

For exact values of Eigenvectors and Eigenvalues

$$det(A - \lambda I) = 0$$
$$(2 - \lambda)^{2} - 1 = 0$$
$$\lambda_{1} = 3, \quad \lambda_{2} = 1$$

For the Dominant Eigenvector,

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 3 \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

This gives,

$$2x_1 - x_2 = 3x_1$$
$$-1x_1 + 2x_2 = 3x_2$$
$$x_1 = -x_2$$

Thus the eigenvector,

$$X = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

 $\mathbf{d}$ 

For the error  $\epsilon$ ,

$$\epsilon = \frac{|3 - 2.8|}{3} \times 100\%$$
$$= 6.667\%$$

 $\mathbf{e}$ 

For the inverse matrix of A,

$$A^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$
$$= \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

Let's take  $X_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ 

Iteration 1,

$$A^{-1}X_0 = \mu X_1$$

$$\frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \frac{2}{3} \begin{bmatrix} 1 \\ 0.5 \end{bmatrix}$$
(1)

Iteration 2,

$$A^{-1}X_1 = \mu X_2$$

$$\frac{1}{3} \begin{bmatrix} 2 & 1\\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1\\ 0.5 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2.5\\ 2 \end{bmatrix} = \frac{2.5}{3} \begin{bmatrix} 1\\ 0.8 \end{bmatrix}$$
(2)

Iteration 3,

$$A^{-1}X_2 = \mu X_3$$

$$\frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0.8 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2.8 \\ 2.6 \end{bmatrix} = \frac{2.8}{3} \begin{bmatrix} 1 \\ 0.929 \end{bmatrix}$$
(3)

This gives approximate smallest eigenvalue,

$$\lambda_{min} = \frac{1}{\mu}$$

$$= \frac{3}{2.8}$$

$$= 1.071$$

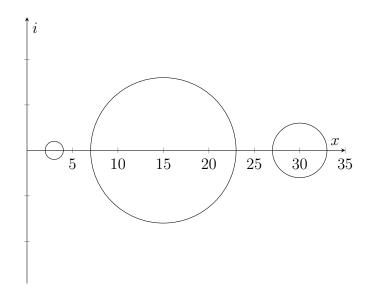
# Question 2

Given that,

$$A = \begin{bmatrix} 30 & 1 & 2 \\ 4 & 15 & -4 \\ -1 & 0 & 3 \end{bmatrix} \quad X_0 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

 $\mathbf{a}$ 

$$D_i = \left(a_{i,i}, \sum_{j \neq i} |a_{i,j}|\right) \quad D_1 = (30,3) \quad D_2 = (15,8) \quad D_3 = (3,1)$$



b

Iteration 1,

$$AX_{0} = \lambda X_{1}$$

$$\begin{bmatrix} 30 & 1 & 2 \\ 4 & 15 & -4 \\ -1 & 0 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -28 \\ -8 \\ 4 \end{bmatrix} = 28 \begin{bmatrix} -1 \\ -0.286 \\ 0.143 \end{bmatrix}$$
(1)

Iteration 2,

$$AX_{1} = \lambda X_{2}$$

$$\begin{bmatrix} 30 & 1 & 2 \\ 4 & 15 & -4 \\ -1 & 0 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ -0.286 \\ 0.143 \end{bmatrix} = \begin{bmatrix} -30.00 \\ -8.862 \\ 1.429 \end{bmatrix} = 30 \begin{bmatrix} -1 \\ -0.2954 \\ 0.048 \end{bmatrix}$$
(2)

Iteration 3,

$$AX_{2} = \lambda X_{3}$$

$$\begin{bmatrix} 30 & 1 & 2 \\ 4 & 15 & -4 \\ -1 & 0 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ -0.295 \\ 0.048 \end{bmatrix} = \begin{bmatrix} -30.199 \\ -8.617 \\ 1.144 \end{bmatrix} = 30.199 \begin{bmatrix} -1 \\ -0.285 \\ 0.037 \end{bmatrix}$$
(3)

 $\mathbf{c}$ 

For the dominant eigenvalue, choosing the scalar as 30,

$$A - 30I = \begin{bmatrix} 0 & 1 & 2 \\ 4 & -15 & -4 \\ -1 & 0 & -27 \end{bmatrix}$$
$$(A - 30I)^{-1} = \frac{1}{82} \begin{bmatrix} 405 & 27 & 26 \\ 112 & 2 & 8 \\ -15 & -1 & -4 \end{bmatrix}$$

Selecting 
$$X_0 = \begin{bmatrix} -1\\0\\1 \end{bmatrix}$$
,

Iteration 1,

Iteration 2,

Iteration 3,

Thus, the eigenvalue,

$$\lambda_{max} = \frac{1}{\mu} + \sigma$$

$$= \frac{1}{4.936} + 30$$

$$= 30.203$$

It is very clear that the shifted inverse power method converges to the actual eigen value faster than the power method.

#### $\mathbf{d}$

For the inverse matrix of A,

$$A^{-1} = \frac{1}{1372} \begin{bmatrix} 45 & -3 & -34 \\ -8 & 92 & 128 \\ 15 & -1 & 446 \end{bmatrix}$$

Iteration 1,

$$A^{-1}X_0 = \mu X_1$$

$$\frac{1}{1372} \begin{bmatrix} 45 & -3 & -34 \\ -8 & 92 & 128 \\ 15 & -1 & 446 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = \frac{1}{1372} \begin{bmatrix} -79 \\ 136 \\ 431 \end{bmatrix} = 0.314 \begin{bmatrix} -0.183 \\ 0.315 \\ 1 \end{bmatrix}$$
 (1)

Iteration 2,

$$A^{-1}X_1 = \mu X_2$$

$$\frac{1}{1372} \begin{bmatrix} 45 & -3 & -34 \\ -8 & 92 & 128 \\ 15 & -1 & 446 \end{bmatrix} \begin{bmatrix} -0.183 \\ 0.315 \\ 1 \end{bmatrix} = \frac{1}{1372} \begin{bmatrix} -43.18 \\ 158.444 \\ 442.94 \end{bmatrix} = 0.32 \begin{bmatrix} -0.097 \\ 0.358 \\ 1 \end{bmatrix}$$
 (2)

Iteration 3,

$$A^{-1}X_2 = \mu X_3$$

$$\frac{1}{1372} \begin{bmatrix} 45 & -3 & -34 \\ -8 & 92 & 128 \\ 15 & -1 & 446 \end{bmatrix} \begin{bmatrix} -0.097 \\ 0.358 \\ 1 \end{bmatrix} = \frac{1}{1372} \begin{bmatrix} -39.44 \\ 161.71 \\ 444.19 \end{bmatrix} = 0.323 \begin{bmatrix} -0.088 \\ 0.364 \\ 1 \end{bmatrix}$$
 (3)

Thus the smallest Eigenvalue approximation,

$$\lambda_{min} = \frac{1}{\mu}$$

$$= \frac{1}{0.323}$$

$$= 3.0959$$

#### Question 3

Given that,

$$A = \begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix}$$

 $\mathbf{a}$ 

Eigenvalues,

$$\lambda_1 = 3$$
  $\lambda_2 = -6$   $\lambda_3 = 0$ 

Corresponding eigen vectors,

$$v_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$
  $v_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$   $v_3 = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ 

b

Taking 
$$X_0 = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}$$

Iteration 1,

$$AX_{0} = \lambda X_{1}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} = \begin{bmatrix} -4 \\ 2 \\ 8 \end{bmatrix} = -8 \begin{bmatrix} 0.5 \\ -0.25 \\ -1 \end{bmatrix}$$
(1)

Iteration 2,

$$AX_{1} = \lambda X_{2}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 0.5 \\ -0.25 \\ -1 \end{bmatrix} = \begin{bmatrix} -5.25 \\ -0.75 \\ 3.75 \end{bmatrix} = -5.25 \begin{bmatrix} 1 \\ 0.143 \\ -0.714 \end{bmatrix}$$
(2)

Iteration 3,

$$AX_{2} = \lambda X_{3}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 0.143 \\ -0.714 \end{bmatrix} = \begin{bmatrix} -4.713 \\ 0.429 \\ 5.571 \end{bmatrix} = -5.571 \begin{bmatrix} 0.846 \\ 0.076 \\ -1 \end{bmatrix}$$
(3)

Iteration 4,

$$AX_{3} = \lambda X_{4}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 0.846 \\ 0.076 \\ -1 \end{bmatrix} = \begin{bmatrix} -5.616 \\ -0.078 \\ 5.46 \end{bmatrix} = -5.616 \begin{bmatrix} 1 \\ 0.014 \\ -0.972 \end{bmatrix}$$
(4)

Iteration 5,

$$AX_4 = \lambda X_5$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 0.014 \\ -0.972 \end{bmatrix} = \begin{bmatrix} -5.874 \\ -0.042 \\ 5.958 \end{bmatrix} = -5.958 \begin{bmatrix} 0.986 \\ 0.007 \\ -1 \end{bmatrix}$$
(5)

Taking 
$$X_0 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

Iteration 1,

$$AX_{0} = \lambda X_{1}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \\ 4 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$
(1)

Iteration 2,

$$AX_{1} = \lambda X_{2}$$

$$\begin{bmatrix} -2 & 1 & 4 \\ 1 & 1 & 1 \\ 4 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$
(2)

Since  $X_2 = X_1$ , over the next iterations it will produce the same answer with eigen value equals to 3, and the eigenvector in this scenario converges to  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ 

### Question 4

Given that,

$$f(x) = \begin{bmatrix} x_1^2 - x_1 + x_2 - 1 = 0 \\ x_1^2 - 2x_2^2 - x_2 = 0 \end{bmatrix}$$

For the Jacobian Matrix J,

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}$$
$$= \begin{bmatrix} 2x_1 - 1 & 1 \\ 2x_1 & -4x_2 - 1 \end{bmatrix}$$

Iteration 1,

$$X_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \tag{1}$$

$$J(X_0) = \begin{bmatrix} -1 & 1\\ 0 & -1 \end{bmatrix} \tag{2}$$

$$f(X_0) = \begin{bmatrix} -1\\0 \end{bmatrix} \tag{3}$$

Using Newtons Method,

$$J(X_0)\delta X_{1,0} = -f(X_0)$$

$$\begin{bmatrix} -1 & 1\\ 0 & -1 \end{bmatrix} \delta X_{1,0} = -\begin{bmatrix} -1\\ 0 \end{bmatrix} = \begin{bmatrix} 1\\ 0 \end{bmatrix}$$

This gives,

$$X_1 = \begin{bmatrix} -1\\0 \end{bmatrix} \tag{4}$$

$$J(X_1) = \begin{bmatrix} -3 & 1\\ -2 & -1 \end{bmatrix} \tag{5}$$

$$f(X_1) = \begin{bmatrix} 1\\1 \end{bmatrix} \tag{6}$$

Using Newton's Method,

$$J(X_1)\delta X_{2,1} = -f(X_1)$$

$$\begin{bmatrix} -3 & 1 \\ -2 & -1 \end{bmatrix} \delta X_{2,1} = -\begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

Using Gaussian Elimination,

$$\begin{bmatrix} -3 & 1 & | & -1 \\ -2 & -1 & | & -1 \end{bmatrix}_{R2-R3 \times \frac{2}{3}}$$

$$\begin{bmatrix} -3 & 1 & | & -1 \\ 0 & -\frac{5}{3} & | & -\frac{1}{3} \end{bmatrix}_{R2 \times \frac{-3}{5}}$$

$$\begin{bmatrix} -3 & 1 & | & -1 \\ 0 & 1 & | & \frac{1}{5} \end{bmatrix}_{R_1-R_2}$$

$$\begin{bmatrix} -3 & 0 & | & \frac{-6}{5} \\ 0 & 1 & | & \frac{1}{5} \end{bmatrix}_{R_1 \times \frac{1}{3}}$$

$$\begin{bmatrix} 1 & 0 & | & \frac{2}{5} \\ 0 & 1 & | & \frac{1}{5} \end{bmatrix}$$

This gives,

$$X_2 = \begin{bmatrix} -\frac{3}{5} \\ \frac{1}{5} \end{bmatrix} \tag{7}$$

$$J(X_2) = \begin{bmatrix} -\frac{11}{5} & 1\\ -\frac{12}{5} & -\frac{9}{5} \end{bmatrix}$$
 (8)

$$f(X_2) = \begin{bmatrix} \frac{4}{25} \\ \frac{2}{25} \end{bmatrix} \tag{9}$$

Using Newton's Method,

$$J(X_1)\delta X_{3,2} = -f(X_2)$$

$$\begin{bmatrix} -\frac{11}{5} & 1\\ -\frac{12}{5} & -\frac{9}{5} \end{bmatrix} \delta X_{3,2} = \begin{bmatrix} -\frac{4}{25}\\ -\frac{2}{25} \end{bmatrix}$$

Using Gaussian Elimination,

$$\begin{bmatrix} -\frac{11}{5} & 1 & | -\frac{4}{25} \\ -\frac{12}{5} & -\frac{9}{5} & | -\frac{2}{25} \end{bmatrix}$$
$$\begin{bmatrix} 1 & 0 & | \frac{46}{795} \\ 0 & 1 & | -\frac{26}{795} \end{bmatrix}$$

This gives,

$$X_3 = \begin{bmatrix} -0.658\\ 0.232 \end{bmatrix}$$