

# ECSE 343: Numerical Methods in Engineering

## Assignment 3

Due Date: 29th nov 2021

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Please type your answers and write your code in this .mlx script. If you choose to provide the handwritten answers, please scan your answers and include those in SINGLE pdf file.

Please submit this .mlx file along with a **PDF** copy of this file.

Note: You can write the function in the appendix section using the provided stencils.

**In this assignment you may use the backslash operator, '\' to solve systems such as  $Ax=b$ .**

**Question 1: Nonlinear Equations for univariate case.**

a) Bisection Method is used for finding the roots of a continuous function,  $f(x)$ , given endpoints;  $a, b$  with  $f(a) \cdot f(b) < 0$ . The interval  $[a, b]$  contains a root,  $x_r$ , because  $f(a)$  and  $f(b)$  have opposite signs.

Bisection method bisects the given interval at the midpoint,  $c = \frac{a+b}{2}$  and then chooses a new interval based such that end of point of interval have opposite signs, i.e., if  $f(a) \cdot f(c) < 0$ , then the new interval is set to  $[a, c]$ , else if  $f(c) \cdot f(b) < 0$  then the interval is set to  $[c, b]$ .

The above procedure is repeated until the following two conditions are simultaneously met,

- The function value at the interval is sufficiently small, i.e.,  $\|f(c)\|_2 < \epsilon_{\text{tolerance}}$
- The new interval is sufficiently small, i.e.,  $\|a - b\|_2 < \epsilon_{\text{tolerance}}$

Implement the Bisection Method in the cell below to find the root of  $f(x) = x^4 - 2x^2 - 4$  in the interval  $[-3, 0]$  using  $\epsilon_{\text{tolerance}} = 10^{-5}$ . Show the number of iterations the bisection method took to converge.

Use the cell below to implement your code.

Note: There is no need to write the function for Bisection Method. However, if you wish to implement the function, use the appendix.

```
%implement your bisection method here
f=@(x) x^4-2*x^2-4;
a=-3;
b=0;
tol = 10^-5;
```

```
[c,counter]= bisectionMethod(f,a,b,tol)
```

```
fc = 1.2621e-06  
c = -1.7989  
counter = 22
```

## b) Newton-Raphson

Bisection Method requires the given function,  $f(x)$  to be continuous, Newton-Raphson requires  $f(x)$  to be continuous and differentiable. While the Newton-Raphson method converges faster than the bisection method, however, Newton-Raphson method does not guarantee convergence. Newton-Raphson works by linearising the  $f(x)$ , at the given initial guess,  $x^{(0)}$  as shown below

$$f(x) = f(x^{(0)}) + f'(x^{(0)})(x - x^{(0)})$$

Setting the expression on the right to zero, provides an approximation to the root, and we obtain

$$x^{(1)} = x^{(0)} - \frac{f(x^{(0)})}{f'(x^{(0)})}$$

Then, the function,  $f(x)$ , is again linearised using approximation,  $x^{(1)}$  as the initial guess, to obtain new approximation of the root. After  $k$  iterations the new approximation for the root becomes,

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})}$$

The above procedure is repeated until the following two convergence conditions are simultaneously met,

- The function value at new guess point is sufficiently small, i.e.,  $\|f(x^{(k+1)})\|_2 < \epsilon_{\text{tolerance}}$
- The difference between the two consecutive solutions is sufficiently small, i.e.,  $\|\Delta x^{(k+1)}\|_2 < \epsilon_{\text{tolerance}}$

where  $\Delta x^{(k+1)} = x^{(k+1)} - x^{(k)}$ .

Implement the **Newton-Raphson** Method in the cell below to find the root of  $f(x) = x^4 - 2x^2 - 4$  use initial guess of  $x^{(0)} = -3$  using  $\epsilon_{\text{tolerance}} = 10^{-5}$ . If the convergence is not reached in 100 iterations, quit the algorithm by displaying an error message indicating that Newton-Raphson failed to converge.

Plot the 2-Norm of difference between consecutive solutions  $\|x^{(k+1)} - x^{(k)}\|_2$  for each iteration.

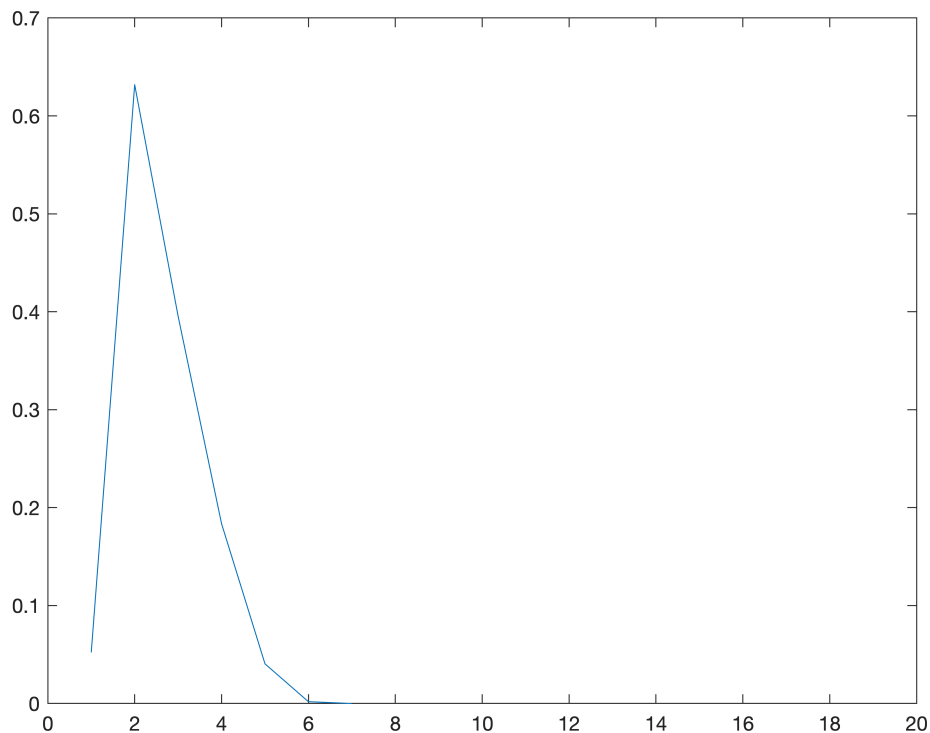
```
% write your code here  
f=@(x)x^4-2*x^2-4;  
fderivative = @(x) 4*x^3-4*x;  
x0=-3;
```

```
tol = 10^-5;
nmax=100;

[x,deltaX]=newtonraphson(f,fderivative,x0,tol,nmax)

x = -1.7989
deltaX = 1×7
    0.0521    0.6319    0.3954    0.1836    0.0405    0.0018    0.0000
```

```
plot(deltaX)
xlim ([0,20]);
```



## Question 2: Nonlinear Equations for N-variables.

Newton-Raphson Method can be used to solve the system of nonlinear equations of N-variables,  $[x_1, x_2, \dots, x_n]$  shown below,

$$f_1(x_1, x_2, \dots, x_N) = 0$$

$$f_2(x_1, x_2, \dots, x_N) = 0$$

⋮

$$f_N(x_1, x_2, \dots, x_N) = 0$$

The above equations can be written as a vector valued function as shown below,

$$\mathbf{f}(\mathbf{X}) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_N) \\ f_2(x_1, x_2, \dots, x_N) \\ \vdots \\ f_N(x_1, x_2, \dots, x_N) \end{bmatrix}$$

where  $\mathbf{X} = [x_1, x_2, \dots, x_N]$  is the vector containing all the variables.

Following the idea of unidimensional Newton-Raphson, we start with an initial guess  $\mathbf{X}^{(0)}$  and we use this to linearise the function  $\mathbf{f}(\mathbf{X})$  around  $\mathbf{X}^{(0)}$

$$\mathbf{f}(\mathbf{X}) = \mathbf{f}(\mathbf{X}^{(0)}) + \left( \frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}^{(0)}) \right) (\mathbf{X} - \mathbf{X}^{(0)})$$

Setting the expression on the left side to zero, we find the approximation of the solution vector. After  $k$  iteration the approximation for the solution vector can be written as

$$\mathbf{X}^{(k+1)} = \mathbf{X}^{(k)} - \left( \frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}^{(k)}) \right)^{-1} \mathbf{f}(\mathbf{X}^{(k)})$$

where  $\mathbf{f}(\mathbf{X}^{(k)})$  is the vector valued function evaluated at  $\mathbf{X}^{(k)}$  and  $\left( \frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}^{(k)}) \right)$  is the jacobian evaluated at  $\mathbf{X}^{(k)}$ . The structure of the Jacobian is given by,

$$\frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

The above procedure is repeated until the following two convergence conditions are simultaneously met,

- The function value at new guess point is sufficiently small, i.e.,  $\|\mathbf{f}(\mathbf{X}^{(k+1)})\|_2 < \epsilon_{\text{tolerance}}$
- The difference between the two consecutive solutions is sufficiently small, i.e.,  $\|\mathbf{X}^{(k+1)} - \mathbf{X}^{(k)}\|_2 < \epsilon_{\text{tolerance}}$

Find a solution of the following system of three nonlinear equations using Newton-Raphson method.

$$x_1^2 + x_2^2 + x_3^2 = 3$$

$$x_1^2 + x_2^2 - x_3 = 1$$

$$x_1 + x_2 + x_3 = 3$$

Note that this can be expressed as

$$f(X) = \begin{bmatrix} x_1^2 + x_2^2 + x_3^2 - 3 \\ x_1^2 + x_2^2 - x_3 - 1 \\ x_1 + x_2 + x_3 - 3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \quad \text{where } X = [x_1 \ x_2 \ x_3]^T$$

a) Write a MATLAB function named *evaluateEquations(x)*, that evaluates the above equation at a given input vector. Use the provided *X* to check your implementation.

Use the framework of the function provided in the Appendix.

```
% X in this case is an example
```

```
X = [2;3;4];
```

```
f = evaluateEquations(X)
```

```
f = 3x1
     26
      8
      6
```

b) Write a MATLAB function named *evaluateJacobian()*, that evaluates the Jacobian of the above equations. Use the provided *X* to check your implementation.

Use the framework of the function provided in the Appendix.

```
X = [2;3;4];
```

```
J = evaluateJacobian(X)
```

```
J = 3x3
     4     6     8
     4     6    -1
     1     1     1
```

## Newton-Raphson

c) Implement the Newton-Raphson method, use the function named *NewtonRaphson()* to write the code for this.

Start with initial guess,  $X^{(0)} = [0 \ 0 \ 0]^T$  and then find a suitable initial guess. Plot the 2-Norm of difference between consecutive solutions  $\|X^{(k+1)} - X^{(k)}\|_2$  for each iteration.

Use the framework of the function provided in the Appendix. In the *NewtonRaphson()* function use the methods implemented in part (a) and (b).

```
% write your code here
func = @evaluateEquations;
Jac = @evaluateJacobian;
Xguess=[1;2;3]
```

```
Xguess = 3x1
```

```
1
```

```
2
```

```
3
```

```
[X,deltaX] = NewtonRaphson(func,Jac,Xguess,1e-5)
```

```
X = 3x1
```

```
10-10 x
```

```
0.9025
```

```
0.9025
```

```
0
```

```
deltaX = 1x19
```

```
2.4265
```

```
1.2418
```

```
0.6226
```

```
0.3113
```

```
0.1556
```

```
0.0778
```

```
0.0389
```

```
0.0195 ...
```

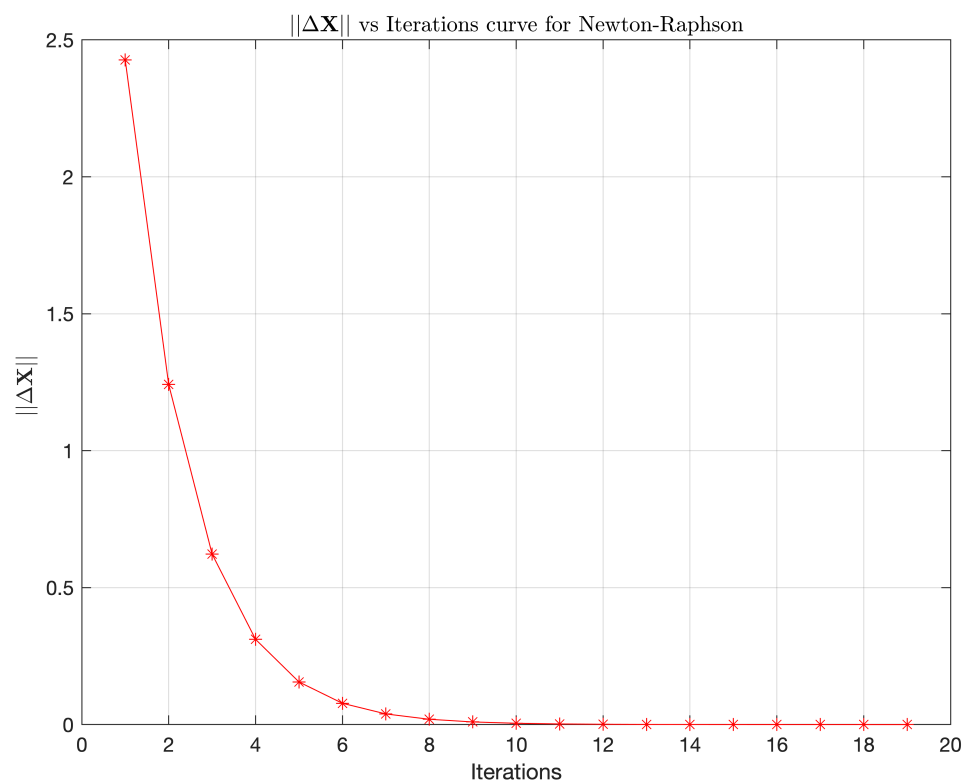
```
plot(deltaX,'r-*')
```

```
xlabel('Iterations')
```

```
ylabel('  $\|\Delta \mathbf{X}\|$  ', 'Interpreter', 'latex')
```

```
grid on
```

```
title('  $\|\Delta \mathbf{X}\|$  vs Iterations curve for Newton-Raphson', 'Interpreter', 'Interpreter')
```



**Newton-Raphson with Broyden's method to approximate Jacobian**

d) Suppose you do not have access to the Jacobian of the nonlinear equations, use the Broyden's method to compute/ update the Jacobian.

At the  $(k + 1)^{th}$  iteration of Newton Raphson, the solution vector  $\mathbf{X}$  is given by

$$\mathbf{X}^{(k+1)} = \mathbf{X}^{(k)} - \left( \frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}^{(k)}) \right)^{-1} \mathbf{f}(\mathbf{X}^{(k)})$$

let,  $J_k = \left( \frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}^{(k)}) \right)$ . Using the Broyden's update,  $J_k$  can be computed as

$$J_k = J_{k-1} + \frac{(\mathbf{f}(\mathbf{X}^{(k)}) - \mathbf{f}(\mathbf{X}^{(k-1)})) - J_{k-1} (\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)})}{\| (\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)}) \|_2^2} (\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)})^T$$

Write a MATLAB function named `NR_Broyden_Jacobian()`, that evaluates the Jacobian of the above equations.

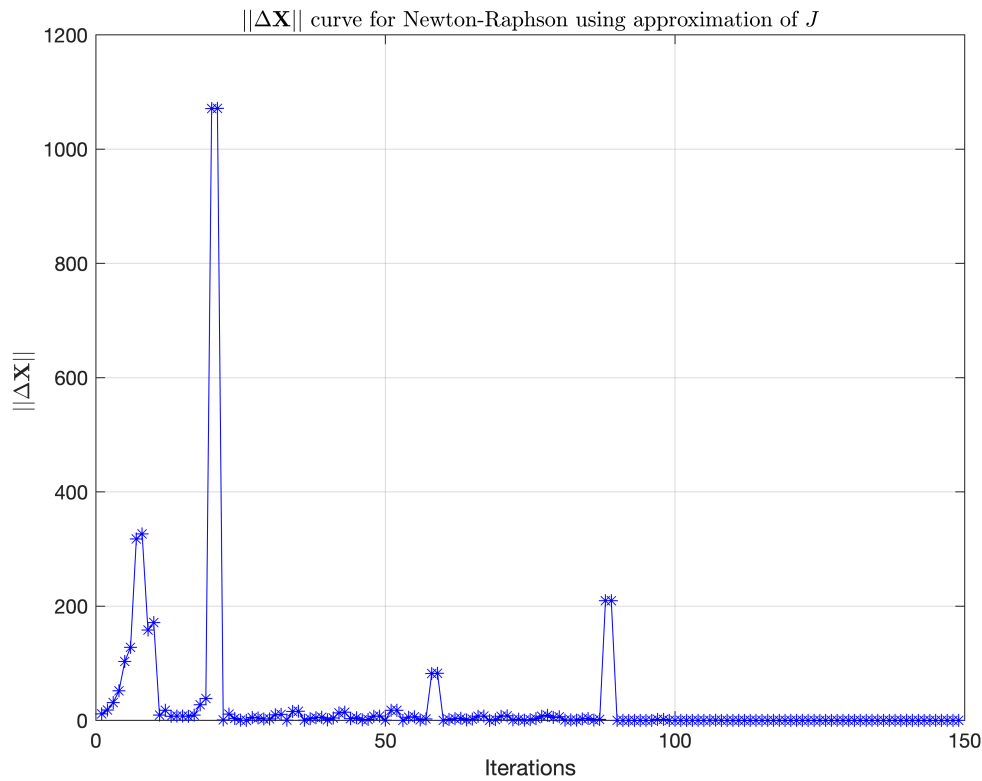
Use the initial guess,  $\mathbf{X}^{(0)}$  you used in part c). For the initial guess for Jacobian  $J_0 = I_{n \times n}$ , where  $I_{n \times n}$  is the identity matrix and  $n$  is the number of equations.

Use the framework of the function provided in the Appendix

```
func = @evaluateEquations;
[Xout,deltaXBroyden] = NR_Broyden_Jacobian(func, Xguess,1e-5)
```

```
Xout = 3x1
10-10 x
    0.2448
    0.3375
   -0.0000
deltaXBroyden = 1x149
103 x
    0.0114    0.0182    0.0315    0.0517    0.1031    0.1281    0.3179    0.3267 ...
```

```
plot(deltaXBroyden,'b-*)
xlabel('Iterations')
ylabel('$ || \Delta \textbf{X} || $ ', 'Interpreter', 'latex')
grid on
title('$ || \Delta \textbf{X} || $ curve for Newton-Raphson using approximation of $
```



### Newton-Raphson with Broyden's method to approximate Inverse of Jacobian

e) In part d) , you implemented the Broyden's method for updating the Jacobian, in this part use the Broyden's update for computing inverse of Jacobian. The inverse of Jacobian can be computed as

$$J_k^{-1} = J_{k-1}^{-1} + \frac{(\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)}) - J_{k-1}^{-1} (f(\mathbf{X}^{(k)}) - f(\mathbf{X}^{(k-1)}))}{(\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)})^T J_{k-1}^{-1} (f(\mathbf{X}^{(k)}) - f(\mathbf{X}^{(k-1)}))} \left( (\mathbf{X}^{(k)} - \mathbf{X}^{(k-1)})^T J_{k-1}^{-1} \right)$$

Write a MATLAB function named `NR_Broyden_InvJacobian()`, that evaluates the Jacobian of the above equations.

Use the initial guess,  $\mathbf{X}^{(0)}$  you used in part c). For the initial guess for inverse of Jacobian  $J_0^{-1} = I_{n \times n}$ , where  $I_{n \times n}$  is the identity matrix and  $n$  is the number of equations.

Use the framework of the function provided in the Appendix.

```
func = @evaluateEquations;
[Xout,deltaXBroyden] = NR_Broyden_InvJacobian(func, Xguess,1e-5)
```

```
Xout = 3x1
10^-7 x
    0.9949
    0.6433
```



```

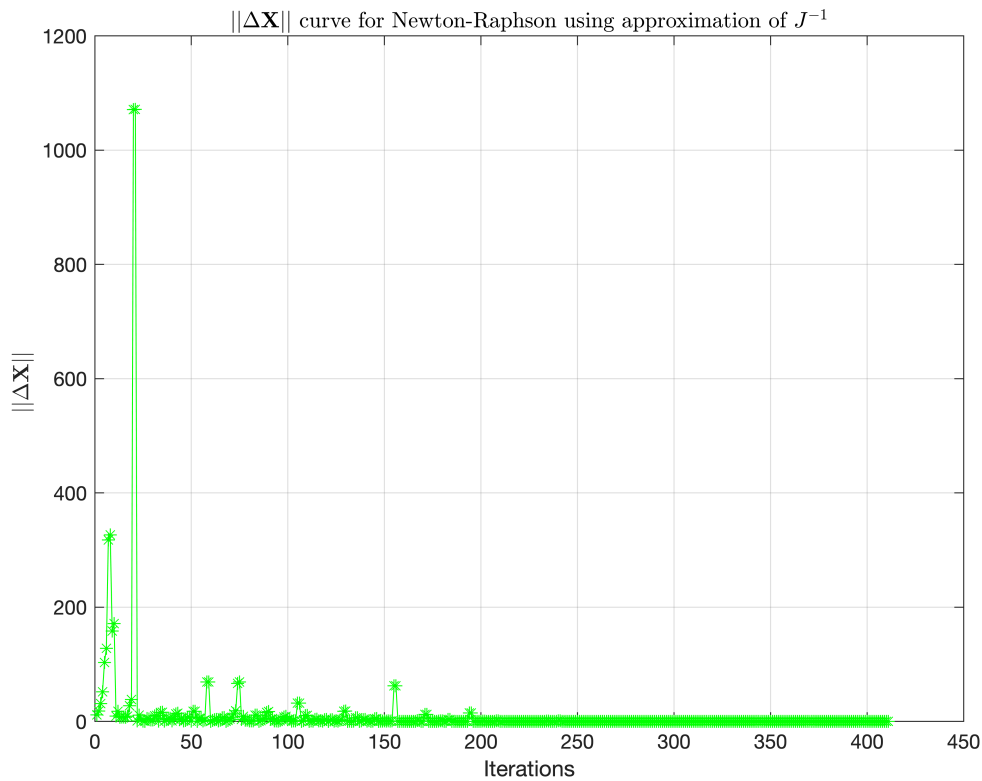
0.0000
deltaXBroyden = 1x411
103 ×
0.0114    0.0182    0.0315    0.0517    0.1031    0.1281    0.3179    0.3267 ...

```

```

plot(deltaXBroyden, 'g-*)
xlabel('Iterations')
ylabel(' $ || \Delta \textbf{X} || $ ', 'Interpreter', 'latex')
grid on
title(' $ || \Delta \textbf{X} || $ curve for Newton-Raphson using approximation of $

```



f) Given a nonlinear system with  $n$  number of nonlinear equations, shown below,

$$f_1(x_1, x_2, \dots, x_n) = 3x_1 - 2x_1^2 - 2x_2 + 1 = 0$$

$$f_i(x_1, x_2, \dots, x_n) = 3x_i - 2x_i^2 - x_{i-1} - 2x_{i+1} + 1 = 0, \quad \text{for } 1 < i < n$$

$$f_n(x_1, x_2, \dots, x_n) = 3x_n - 2x_n^2 - x_{n-1} + 1 = 0$$

where,  $x_1, x_2, \dots, x_n$  are the variables.

The nonlinear vector of equations can written as,

$$\mathbf{f}(\mathbf{X}) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_i(x_1, x_2, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, \dots, x_n) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

The Matlab function to evaluate the above function is named *nlfunction()* and it is provided to you in appendix.

Suppose you do not have access to the jacobian, so you chose to solve the above nonlinear function using Broyden's methods you developed in part (d) and (e). The code cell below uses the Broyden's algorithms you implemented in part d) and e) to solve the above nonlinear system for different size of the system ranging from  $10 \leq n \leq 200$ .

Run the cell below and explain the results in Figure Question 2 f.

```
clear all
func1 = @nlfunction;
c = 1
```

```
c = 1
```

```
N = 200
```

```
N = 200
```

```
time_BroydenIJ = zeros(N-10+1,1);
time_BroydenJ = zeros(N-10+1,1);
for I=10:N
    Xguess = zeros(I,1);
    brIJstart =tic ;
    [XBrIJ,deltaXBroyden] = NR_Broyden_InvJacobian(func1,Xguess,1e-6);
    time_BroydenIJ(c) = toc(brIJstart) ;

    brJstart =tic ;
    [XBrJ,deltaXBroyden] = NR_Broyden_Jacobian(func1,Xguess,1e-6);
    time_BroydenJ(c) = toc(brJstart) ;

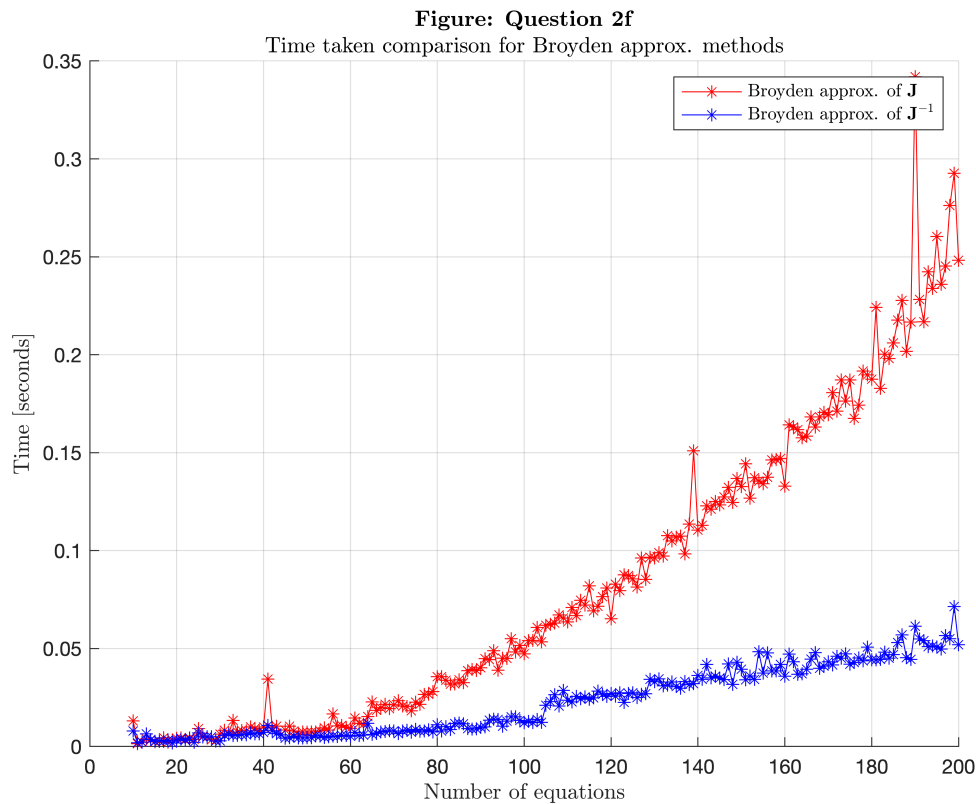
    c =c +1;
end

figure()
clf
hold on
plot(10:N,time_BroydenJ, 'r-*)
plot(10:N,time_BroydenIJ, 'b-*)
```

```

legend('Broyden approx. of  $\textbf{J}$ ', 'Broyden approx. of  $\textbf{J}^{-1}$ ', 'Inter
title({'\textbf{Figure: Question 2f}', 'Time taken comparison for Broyden approx. method
ylabel('Time [seconds]', 'Interpreter', 'latex')
xlabel('Number of equations', 'Interpreter', 'latex')
grid on

```



% Write your explanation here

Broyden's approximation of  $\textbf{J}^{-1}$  in e) is faster than Broyden's approximation of  $\textbf{J}$  done in d).

## Appendix

Function provided for Question 2 part e)

```

function [F] = nlfunction(x)
% x is a columns vector.
% Evaluate the vector function

n = length(x);
F = zeros(n,1);
i = 2:(n-1);
F(i) = (3-2*x(i)).*x(i)-x(i-1)-2*x(i+1) + 1;

```

```

F(n) = (3-2*x(n)).*x(n)-x(n-1) + 1;
F(1) = (3-2*x(1)).*x(1)-2*x(2) + 1;
% % % % % Evaluate the Jacobian if nargout > 1
% % % % if nargout > 1
% % % %     d = -4*x + 3*ones(n,1); D = sparse(1:n,1:n,d,n,n);
% % % %     c = -2*ones(n-1,1); C = sparse(1:n-1,2:n,c,n,n);
% % % %     e = -ones(n-1,1); E = sparse(2:n,1:n-1,e,n,n);
% % % %     J = C + D + E;
% % % % end
end

```

Question 1) a)

```

function [x, iterations] = bisectionMethod(f,a,b,tol)

if f(a)*f(b) > 0
    exit(1)
end
c = (a + b)/2.0;
fc = f(c);
iteration_counter = 1;
while abs(fc) > tol
    fa = f(a);
    fb = f(b);
    if fa*fc > 0
        a = c;
    else
        b = c;
    end
    c = (a + b)/2;
    fc = f(c);
    iteration_counter = iteration_counter + 1;
end
x = c;
iterations = iteration_counter;
fc
end

```

Question 1 b)

```

function [X_new,normdeltaX]=newtonraphson(f,fderiv,x0,tol,itermax)

i = 1;
x(i) = x0;
f_value = f(i);

while abs(f_value) > tol && i < itermax

    x(i+1) = x(i) - (f_value)/fderiv(x(i));

```

```

        deltaX=x(i+1)-x(i);

        normdeltaX(i)=norm(deltaX);

        x(i)=x(i+1);
        X_new = x(i+1);
        f_value = f(x(i+1));
        i = i + 1;

    end
    % Here, either a solution is found, or too many iterations
    if abs(f_value) > tol
        i = -1;
    end

end

```

Question 2) a)

```

function f = evaluateEquations(x)

    f = zeros(3,1);
    % write your code here

    f = zeros(3,1);
    % write your code here
    x1=x(1,1);
    x2=x(2,1);
    x3=x(3,1);

    f1=x1*x1+x2*x2+x3*x3-3;
    f2=x1*x1+x2*x2-x3-1;
    f3=x1+x2+x3-3;

    f= [f1 ;f2 ;f3];

end

```

Question 2) b)

```

function J = evaluateJacobian(x)
    J = zeros(3,3);

    % write your code here

    x1=x(1,1);
    x2=x(2,1);

```

```

x3=x(3,1);

J(1,1) = 2*x1;
J(1,2) = 2*x2;
J(1,3) = 2*x3;

J(2,1) = 2*x1;
J(2,2) = 2*x2;
J(2,3) = -1;

J(3,1) = 1;
J(3,2) = 1;
J(3,3) = 1;

```

```
end
```

### Question 2) c)

```

function [Xout,normDeltaX] = NewtonRaphson(func,Jac,Xguess,tol)

x_prev=Xguess;
iteration=1;

for iteration=1:1000

    x_current = x_prev-(Jac(x_prev))^(-1)*func(x_prev);

    Xout = func(x_current);

    normDeltaX(iteration)=norm(x_current - x_prev);

    if norm(Xout)<tol && (normDeltaX(iteration))<tol
        break
    end

    x_prev=x_current;

    iteration=iteration+1;

end
end

```

### Appendix: Question 2 d)

```

function [Xout,deltaX] = NR_Broyden_Jacobian(func, Xguess,tol)

```

```

J = eye(size(Xguess,1)); %keep this line

iteration=0;

x_prev = Xguess;

while (1)

    iteration=iteration+1;
    j_c=j;

    if iteration ~= 1

        x_current= x_prev-(J_c)^-1 * func(x_prev);

    else

        x_current=x_prev-(J)^-1 * func(x_prev);
    end

    J_c=J+((func(x_current)-func(x_prev))-J*(x_current-x_prev))/norm(x_current-x_prev)';

    Xout=func(x_current);

    deltaX(iteration)=norm(x_current-x_prev);

    if (deltaX(iteration))<tol && norm(Xout)<tol
        break
    end
    x_prev = x_current;

    J=J_c;

end

iteration ;

end

```

#### Appendix: Question 2 e)

```

function [Xout,deltaX] = NR_Broyden_InvJacobian(func, Xguess,tol)

J = eye(size(Xguess,1)); %keep this line

x_prev=Xguess;

for iterations=1:1000

    j_c =j;

```

```

if iterations ~= 1
    x_current=x_prev-(J_c)*func(x_prev);
else
    x_current=x_prev-(J)*func(x_prev);
end

J_c=J+((x_current-x_prev)-J*(func(x_current)-func(x_prev)))/((x_current-x_prev).'*J_c);

Xout=func(x_current);

deltaX(iterations)=norm(x_current-x_prev);

if norm(Xout)<tol && (deltaX(iterations))<tol
    break
end

x_prev=x_current;

J=J_c;

iterations=iterations+1;

end

end

```