

# COL 726 Hw4

K LAXMAN

TOTAL POINTS

59 / 80

## QUESTION 1

1 1 7 / 10

✓ - 0 pts Correct Proof for  $x_n \in \mathcal{K}_n$

- 0 pts Correct value for  $p_n(x)$  and

$\alpha$

- 5 pts Incorrect value for  $p_n(x)$  and

$\alpha$

- 2 pts Incorrect value for  $p_n(x)$

- 5 pts Incorrect/Incomplete Proof for  $x_n \in$

$\mathcal{K}_n$

✓ - 3 pts Incorrect/No Value for  $\alpha$

- 10 pts No Attempt

## QUESTION 2

2 2 10 / 10

✓ - 0 pts Correct

- 2.5 pts Incorrect (a)

- 2.5 pts Incorrect (b)

- 2.5 pts Incorrect (c)

- 2.5 pts Incorrect (d)

- 4 pts Missing Derivatives and Incorrect

Answers

- 2 pts Missing Derivatives

## QUESTION 3

3 3 5 / 10

- 0 pts Correct

- 5 pts Incorrect Code/Plot

✓ - 5 pts Incomplete/Incorrect Explanation of the Convergence

✓ - 2 pts No Plot

+ 2 Point adjustment

## QUESTION 4

4 4 11 / 15

- 0 pts Correct Code and Plots

- 5 pts Incorrect (i) Steepest descent

- 5 pts Incorrect (ii) Newton

- 5 pts Incorrect (iii) Damped Newton

- 6 pts No Plots

✓ - 6 pts No Code

- 3 pts Plot Missing for (iii)

- 15 pts No Attempt

+ 2 Point adjustment

## QUESTION 5

5 5 5 / 10

- 0 pts Correct

✓ - 5 pts Missing/Incorrect Code/Plot for (a)

✓ - 5 pts Missing/Incorrect Code/Plot for (b)

- 3 pts No Plots

- 10 pts No Attempt

- 5 pts Missing Code

+ 5 Point adjustment

## QUESTION 6

6 6 6 / 10

- **0 pts** Correct
- **10 pts** Incorrect Code/Plot

✓ - **5 pts** *No Code*

- **5 pts** No Plot
- **10 pts** No Attempt

+ **1** *Point adjustment*

#### QUESTION 7

7 **7** 15 / 15

✓ - **0 pts** *Correct*

- **7 pts** Missing/Incorrect Proof of the fact that

$f$  is convex

- **8 pts** Missing/Incorrect Proof of the iterates
- **15 pts** No Attempt

Q. 1:- To show that  $x_n$  belongs to the Krylov subspace  $K_n$ .

To show this, we need to show that  $x_n$  can be written as linear combination of first  $n$  powers of  $A$  times  $b$  by induction.

For  $n=1$

$$x_2 = x_1 + \alpha(b - Ax_1) = (I - \alpha A)b + \alpha x_1$$

$\therefore$  This is linear combn. of  $b$  &  $Ax_1$ , so, it belongs to  $K_1$ .

Assume that  $x_n$  belongs to  $K_n$  for some  $n$ , then

$$\begin{aligned} x_{n+1} &= x_n + \alpha(b - Ax_n) \\ &= x_n + \alpha b - \alpha Ax_n \\ &= (I - \alpha A)x_n + \alpha b \end{aligned}$$

by induction hypothesis  $x_n$  written as

$$x_n = c_0 b + c_1 A b + \dots + c_{n-1} A^{n-1} b \quad \text{here } c_0, c_1, \dots, c_{n-1} \text{ are constants}$$

by substituting in above equation.

$$\begin{aligned} x_{n+1} &= (I - \alpha A)(c_0 b + c_1 A b + \dots + c_{n-1} A^{n-1} b) + \alpha b \\ &= c_0 (I - \alpha A)b + c_1 (I - \alpha A)A b + \dots + c_{n-2} (I - \alpha A)A^{n-2} b + \\ &\quad (c_{n-1} (I - \alpha A)A^{n-1} b + \alpha b) \\ &\quad + \alpha b \end{aligned}$$

$\therefore$  This is linear combn. of  $n+1$  powers.

To find degree  $(n+1)$  polynomial corresponds to relation.

$$\begin{aligned} x_{n+1} &= (I - \alpha A)x_n + \alpha b \\ &= (I - \alpha A)(I - \alpha A)x_{n-1} + \alpha (I - \alpha A)b \\ &= (I - \alpha A)^2 x_{n-1} + \alpha (I - \alpha A)b \\ &= (I - \alpha A)^{n-1} x_1 + \alpha (I - \alpha A)^{n-1} b \end{aligned}$$

This is polynomial (degree  $n+1$ ) in  $A$  which can be written as

$$p_{n+1}(x) = x_n + \alpha(xI - A)^{n-1}b \quad \text{here } I \text{ is Identity matrix.}$$

So, finally for value of  $\alpha$  for fast convergence we can use

Formula

$$\alpha = 2/(\lambda_{\max} + \lambda_{\min})$$

$\lambda_{\max}$  &  $\lambda_{\min}$  are eigenvalues of  $A$ .

In this case w.k.t. eigenvalue of  $A$  is spread in circle  $|z-2| \leq \frac{1}{2}$  in complex plane.

Let Eigenvalue of  $A = \lambda_1, \lambda_2, \dots, \lambda_n$

$$|\lambda_i - 2| \leq \frac{1}{2} \quad \text{adding 2 on both sides}$$

$$|\lambda_i| \leq 5/2$$

∴ we have

$$\lambda_{\max} = \max \{ |\lambda_1|, |\lambda_2|, \dots, |\lambda_m| \} \leq 5/2$$

$$\lambda_{\min} = \min \{ |\lambda_1|, |\lambda_2|, \dots, |\lambda_m| \} \geq 3/2$$

Substitute these value in above formula for  $\alpha$

$$\alpha = \frac{2}{(\lambda_{\max} + \lambda_{\min})} \leq \frac{4}{5}$$

for fast convergence.

So, we could recommend  $\alpha = 4/5$

1 1 7 / 10

✓ - 0 pts Correct Proof for  $x_n \in \mathcal{K}_n$

- 0 pts Correct value for  $p_n(x)$  and  $\alpha$

- 5 pts Incorrect value for  $p_n(x)$  and  $\alpha$

- 2 pts Incorrect value for  $p_n(x)$

- 5 pts Incorrect/Incomplete Proof for  $x_n \in \mathcal{K}_n$

✓ - 3 pts Incorrect/No Value for  $\alpha$

- 10 pts No Attempt

(2) given equation  $f(x) = x^2 - 3x + 2 = 0 \rightarrow \textcircled{1}$

given that  $x=2$ , Now we have to find  $|g'_i(2)| \Rightarrow g_{i+1}(x) = \frac{x^2+2}{3}$   
i.e. from eq. (1)

$$\text{Let } 3x = x^2 + 2$$

$$x = \frac{x^2 + 2}{3}$$

i)  $g_1(x) = \frac{x^2+2}{3}$

$$g_1(2) = \frac{2 \times 2}{3} = \frac{4}{3} \quad \text{so, here } g'_1(2) = \frac{4}{3} > 1$$

This condition is not convergent.

ii)  $g_2(x) = \sqrt{3x-2}$

$$\text{Let } x^2 - 3x + 2 = 0 \Rightarrow x^2 = 3x - 2$$
$$\Rightarrow \sqrt{3x-2} = x$$

$$g'_2(x) = \frac{1}{2\sqrt{3x-2}} \cdot 3 = \frac{3}{2\sqrt{3x-2}}$$

$$g'_2(2) = \frac{3}{2\sqrt{6-2}} = \frac{3}{4} < 1 \quad \text{so, } g'_2(x) \text{ is convergent.}$$

iii)  $g_3(x) = \frac{3-2/x}{x}$ ; Let  $x^2 - 3x + 2 = 0$

$$x^2 = 3x - 2$$

$$x = \frac{3-2/x}{x}$$

$$g'_3(x) = \frac{2}{x^2} \quad \text{i.e. } \frac{2}{(2)^2} = \frac{1}{2} < 1$$

$\therefore g'_3(x)$  is convergent

iv)  $g_4(x) = \frac{(x^2-2)}{(2x-3)}$

$$|g'_4(x)| = \frac{(2x-3)2x - (x^2-2)2}{(2x-3)^2}$$

2 2 10 / 10

✓ - 0 pts Correct

- 2.5 pts Incorrect (a)

- 2.5 pts Incorrect (b)

- 2.5 pts Incorrect (c)

- 2.5 pts Incorrect (d)

- 4 pts Missing Derivatives and Incorrect Answers

- 2 pts Missing Derivatives

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Q3

```
% Nonlinear system using Newton and Broyden's methods
x_initial = [-0.5; 1.40]; % Initial guess for x
x_true = [0; 1]; % True value of x
tolerance = eps; % Tolerance for convergence
max_iterations = 20; % Maximum number of iterations
fprintf('Newton method:\n');
fprintf('k  x(1)      x(2)      error\n');
k = 0;
x = x_initial;
step_size = ones(size(x));
error = norm(x - x_true);
fprintf('%3d %17.10e %17.10e %17.10e\n', k, x(1), x(2), error)

while norm(step_size) > tolerance && k < max_iterations
    k = k + 1;
    step_size = -(Df(x)\f(x));
    x = x + step_size ;
    error = norm(x - true_x);
    fprintf('%3d %17.10e %17.10e %17.10e\n', k, x(1), x(2), error);
end

fprintf('\nBroyden method:\n');
fprintf('k  x(1)      x(2)      error\n');
k = 0;
x = x_initial;
fx = f(x);
B = Df(x);
step_size = ones(size(x));
error = norm(x - x_true);
fprintf('%3d %17.10e %17.10e %17.10e\n', k, x(1), x(2), error)

while norm(step_size) > tolerance && k < max_iterations
    k = k + 1;
    step_size = -(B\fx);
    x = x + step_size;
    y = fx;
    fx = f(x);
    y = fx - y;
    B = B + ((y - B*step_size)*(step_size')) / (step_size'*step_size);
```



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- 0 pts Correct

- 5 pts Incorrect Code/Plot

✓ - 5 pts *Incomplete/Incorrect Explanation of the Convergence*

✓ - 2 pts *No Plot*

+ 2 *Point adjustment*

```

error = norm(x - x_true);
fprintf(' %3d %17.10e %17.10e %17.10e\n', k, x(1), x(2), error);
end

% Function to calculate f(x)
function [y] = f(x)
    y = [(x(1) + 3)*(x(2)^3 - 7) + 18;
        sin(x(2)*exp(x(1)) - 1)];
end

% Function to calculate the Jacobian of f(x)
function [j] = Df(x)
    j = [x(2)^3 - 7, 3*x(2)^2*(x(1) + 3);
        x(2)*exp(x(1))*cos(x(2)*exp(x(1)) - 1), exp(x(1))*cos(x(2)*exp(x(1)) - 1)];
end

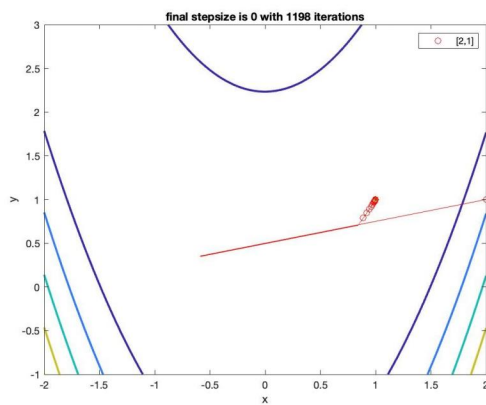
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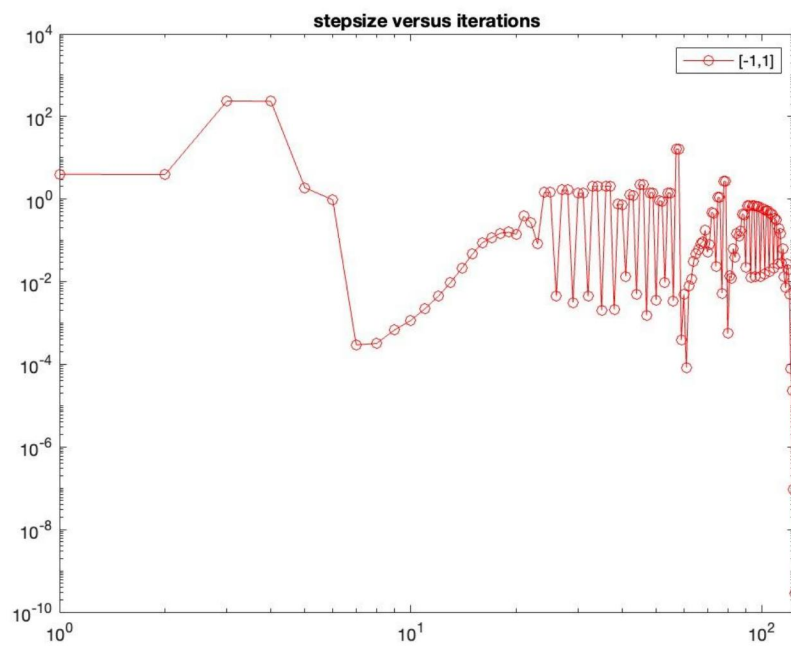
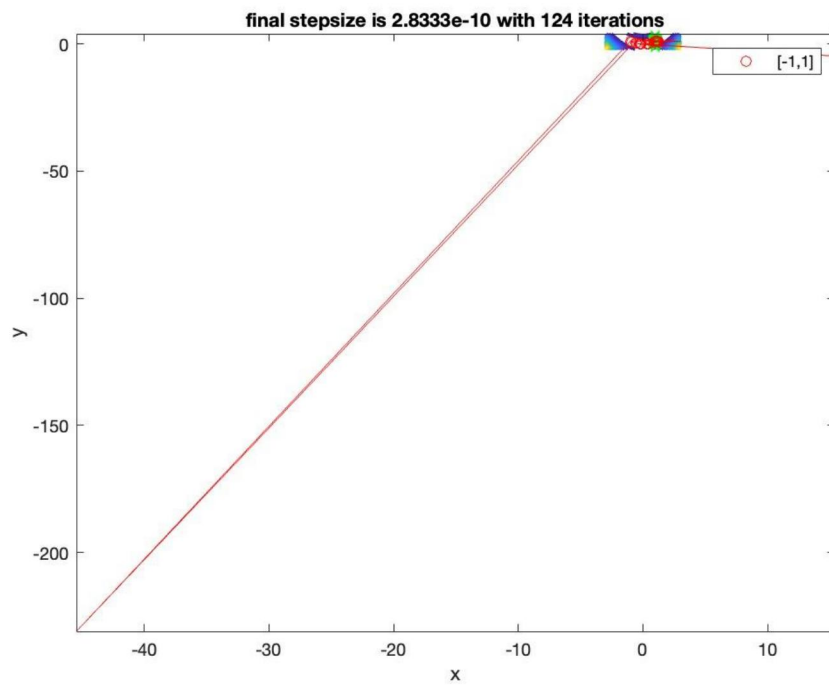
```

>> thirdqs
Newton method:
k  x(1)          x(2)          error
0 -5.0000000000e-01 1.4000000000e+00 6.4031242374e-01
1 2 3 4 5
Broyden method:
k  x(1)          x(2)          error
0 -5.0000000000e-01 1.4000000000e+00 6.4031242374e-01
1 -5.5315135718e-02 1.0280665838e+00 6.2028198166e-02
2 5.0995307015e-04 1.0001236435e+00 5.2472835528e-04
3 -2.3384786364e-04 1.0000765609e+00 2.4606176062e-04
4 -4.0826221037e-05 1.0000135979e+00 4.3031185873e-05
5 -1.3275089604e-07 1.0000000453e+00 1.4027960438e-07
6 -5.3912058040e-10 1.0000000002e+00 5.6858615398e-10
7 1.6678561809e-12 1.0000000000e+00 1.7584764462e-12
8 -7.4352765976e-16 1.0000000000e+00 8.6605346667e-16
9 3.5985405887e-17 1.0000000000e+00 1.1670861614e-16
10 1.4149822529e-16 1.0000000000e+00 1.7985466189e-16
>>

```

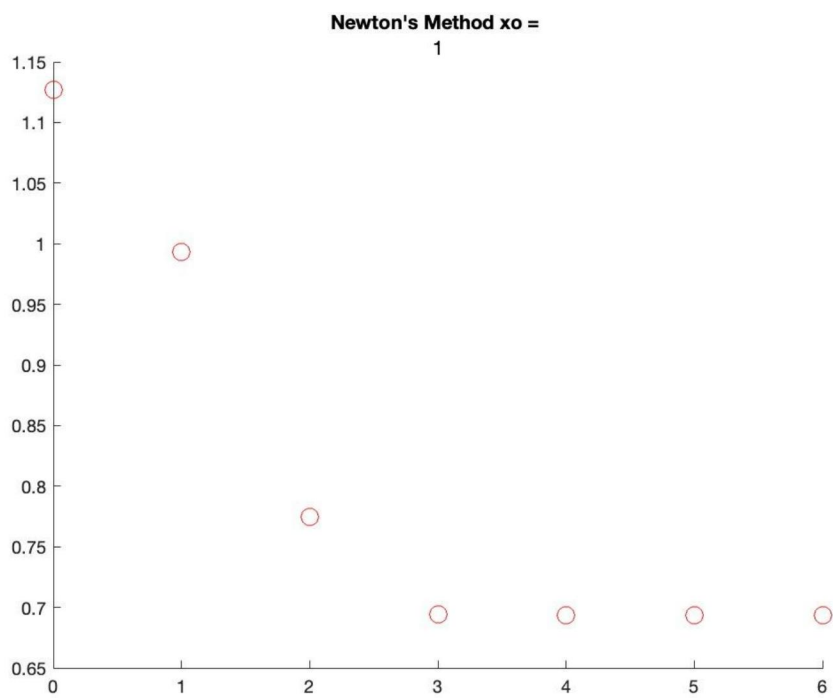
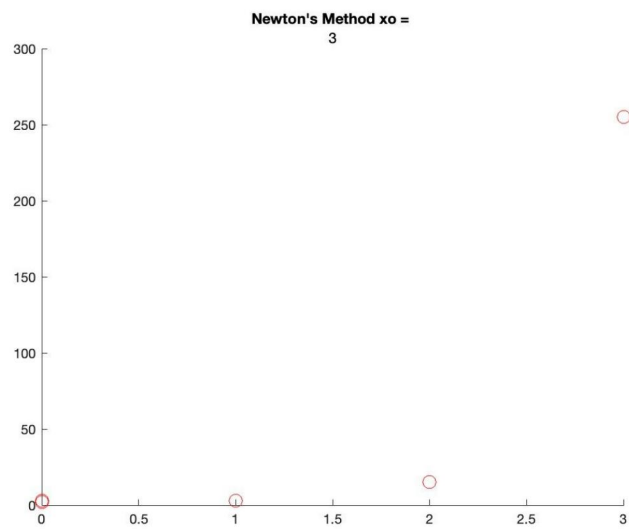
Q4





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- 0 pts Correct Code and Plots
- 5 pts Incorrect (i) Steepest descent
- 5 pts Incorrect (ii) Newton
- 5 pts Incorrect (iii) Damped Newton
- 6 pts No Plots
- ✓ - 6 pts *No Code*
- 3 pts Plot Missing for (iii)
- 15 pts No Attempt
- + 2 *Point adjustment*



5 5 / 10

- 0 pts Correct

✓ - 5 pts *Missing/Incorrect Code/Plot for (a)*

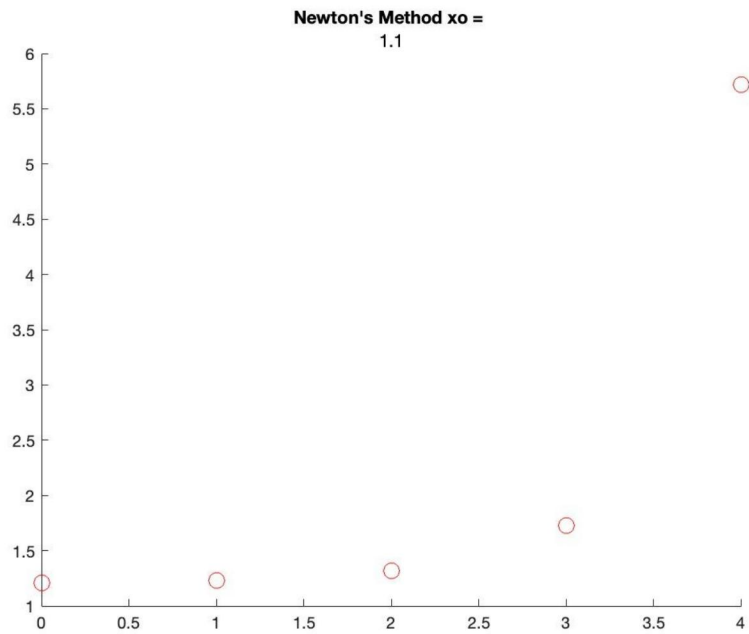
✓ - 5 pts *Missing/Incorrect Code/Plot for (b)*

- 3 pts No Plots

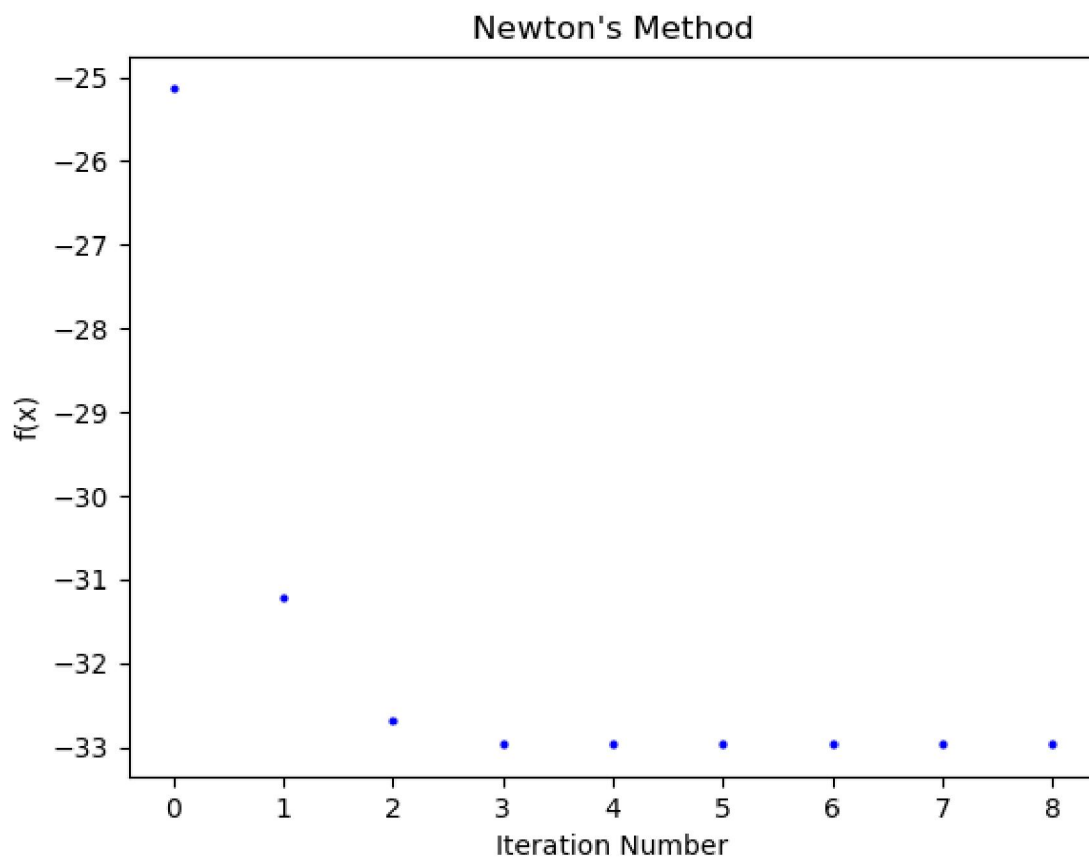
- 10 pts No Attempt

- 5 pts Missing Code

+ 5 *Point adjustment*



Q6



6 6 / 10

- 0 pts Correct

- 10 pts Incorrect Code/Plot

✓ - 5 pts *No Code*

- 5 pts No Plot

- 10 pts No Attempt

+ 1 *Point adjustment*



$$= \frac{2x^2 - 6x + 4}{(2x-3)^2}$$

$$g'_4(2) = \frac{2(2)^2 - 6(2) + 4}{(2 \times 2 - 3)^2} \Rightarrow 0 < 1$$

$\therefore g'_4(x)$  is convergent

Q. 7) :- By using the hint given in Q5 and examining the optimisation problem

$$\text{minimize } x_1 y_1 + \sqrt{\gamma} x_2 y_2$$

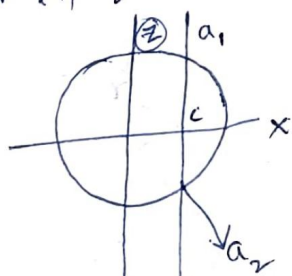
$$\text{Subject to } y_1^2 + y_2^2 \leq 1$$

$$y_1 \geq 1/\sqrt{1+\gamma} \quad \text{here } y_1, y_2 \text{ are variables}$$

It is represented as

i.e. vertical line through

$$y_1 = (1+\gamma)^{-1/2}$$



to maximize the inner product of 'y' with coeff vector  $(x_1, \sqrt{\gamma} x_2)$ . There are 3 cases depending on orientation of coeff vector.

Case 1)  $x_1 > 0$  and  $|x_2| \leq x_1$ , coeff vector lies b/w  $(1, -\sqrt{\gamma})$  &  $(1, \sqrt{\gamma})$

$$\text{optimum is } \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \frac{1}{x_1^2 + \gamma x_2^2} \begin{bmatrix} x_1 \\ \sqrt{\gamma} x_2 \end{bmatrix}$$

$$\text{optimal value is } (x_1^2 + \gamma x_2^2)^{1/2}$$

Case 2) If  $x_2 \leq 0$  and  $x_2 < -x_1$ , then  $\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \frac{1}{\sqrt{1+\gamma}} \begin{bmatrix} 1 \\ -\sqrt{\gamma} \end{bmatrix}$

$$\text{optimal value is } \frac{(x_1 - \gamma x_2)}{(1+\gamma)^{1/2}}$$

Case 3) If  $x_2 \geq 0$  and  $x_1 < x_2$  then  $\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \frac{1}{\sqrt{1+\gamma}} \begin{bmatrix} 1 \\ \sqrt{\gamma} \end{bmatrix}$

$$\text{optimal value is } \frac{x_1 + \gamma x_2}{(1+\gamma)^{1/2}}$$

by considering the gradient descent algorithm.

Note that iterates satisfies  $|x_2^{(k)}| < x_1^{(k)}$  so they are in the interior of the region where  $f(x_1, x_2) = (x_1^2 + \gamma x_2^2)^{1/2}$

$$\nabla f(x) = \frac{1}{\sqrt{x_1^2 + \gamma x_2^2}} \begin{bmatrix} x_1 \\ \gamma x_2 \end{bmatrix}$$

Verify the expressions

$$x_1^{(k)} = \gamma \left( \frac{\gamma-1}{\gamma+1} \right)^k, \quad x_2^{(k)} = \left( -\frac{\gamma-1}{\gamma+1} \right)^k.$$

for  $k=0$ , starting point  $x^{(0)} = (\gamma, 1)$  gradient at  $x^{(k)} \propto (x_1^{(k)}, \gamma x_2^{(k)})$

$\therefore$  exact line search minimizes  $f$  along the line

$$\begin{bmatrix} (1-t)x_1^{(k)} \\ (1-\gamma t)x_2^{(k)} \end{bmatrix} = \left( \frac{\gamma-1}{\gamma+1} \right)^k \begin{bmatrix} (1-t)\gamma \\ (1-\gamma t)(-1)^k \end{bmatrix}$$

$f$  can be written as

$$f((1-t)x_1^{(k)}, (1-\gamma t)x_2^{(k)}) = (\gamma^2(1-t)^2 + \gamma(1-\gamma t)^2)^{1/2} \left( \frac{\gamma-1}{\gamma+1} \right)^k$$

This is minimised by  $t = 2/(1+\gamma)$  so,

$$x^{(k+1)} = \left( \frac{\gamma-1}{\gamma+1} \right)^k \begin{bmatrix} (1-t)\gamma \\ (1-\gamma t)(-1)^k \end{bmatrix}$$

$$= \left( \frac{\gamma-1}{\gamma+1} \right)^{k+1} \begin{bmatrix} \gamma \\ (-1)^{k+1} \end{bmatrix}$$

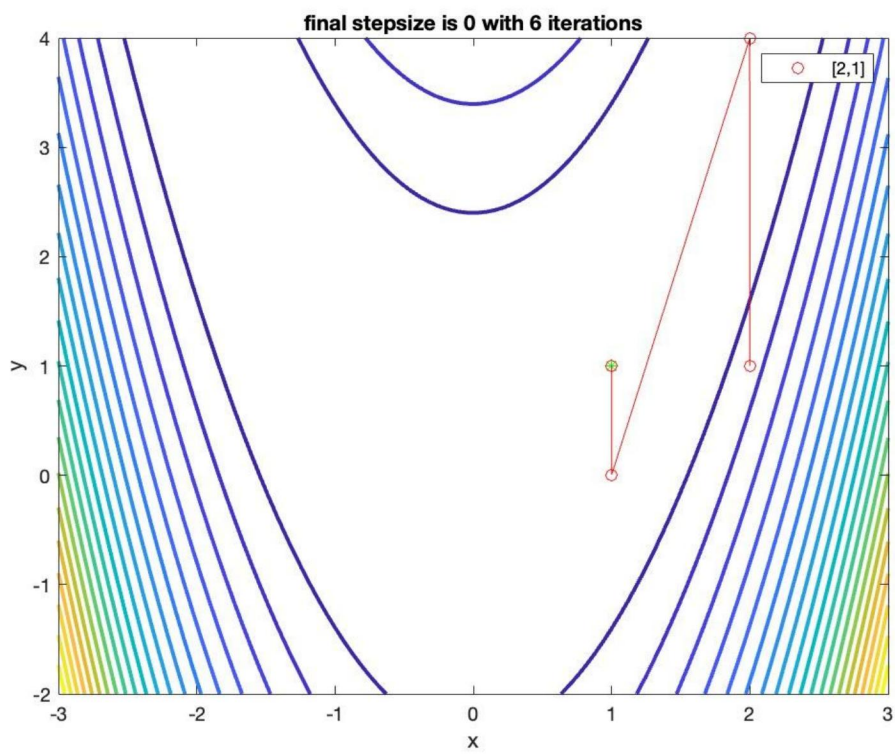
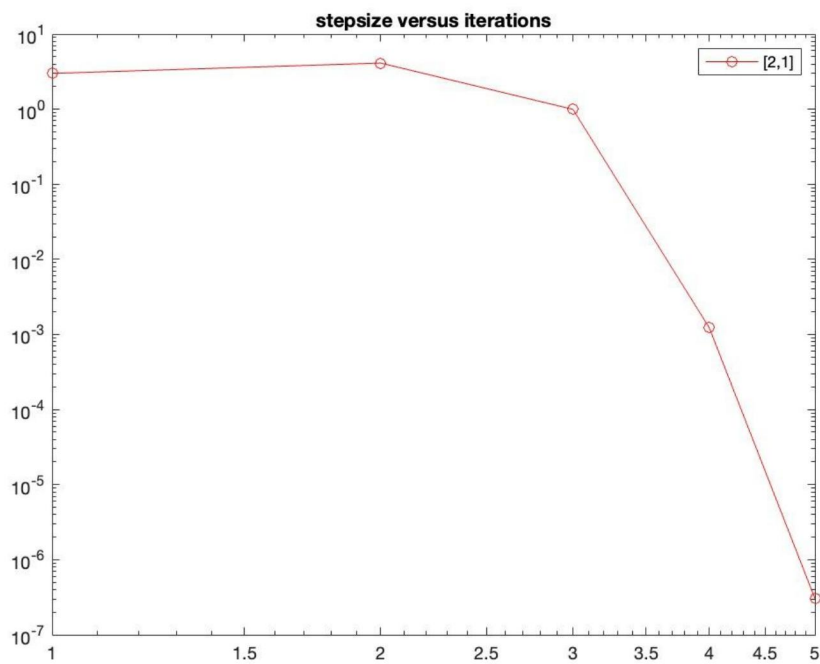
7 15 / 15

✓ - 0 pts Correct

- 7 pts Missing/Incorrect Proof of the fact that  $f$  is convex

- 8 pts Missing/Incorrect Proof of the iterates

- 15 pts No Attempt



**Q5**