



**HACETTEPE UNIVERSITY
ENGINEERING FACULTY
ELECTRICAL AND ELECTRONICS
ENGINEERING PROGRAM**

2023-2024
SPRING SEMESTER

ELE708
NUMERICAL METHODS IN ELECTRICAL ENGINEERING

HW6

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1) Exercises

1.a) 6.5

6.5-) Determine the critical points of each of the following function and characterize each as a min, max or saddle point. Also determine whether each function has global min or max on \mathbb{R}^2 .

a) $x^2 - 4xy + y^2$

$$\nabla f(x,y) = \begin{bmatrix} 2x-4y \\ -4x+2y \end{bmatrix} \rightarrow \nabla f(x,y) = 0 \text{ at } \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow \text{critical point}$$

$$H_f(x,y) = \begin{bmatrix} 2 & -4 \\ -4 & 2 \end{bmatrix} \rightarrow \text{eigen values } \lambda_1 = 1+3i, \lambda_2 = 1-3i \rightarrow H_f(0,0) \text{ is indefinite}$$

positive and negative

$$\text{so } \begin{bmatrix} 0 \\ 0 \end{bmatrix} \text{ is } \underline{\text{saddle point}}$$

Since $f(x,y)$ has only saddle point, there is no global min or max.

b) $x^4 - 4xy + y^4$

$$\nabla f(x,y) = \begin{bmatrix} 4x^3 - 4y \\ -4x + 4y^3 \end{bmatrix} \rightarrow \nabla f(x,y) = 0 \text{ at } \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \text{ and } \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$H_f(x,y) = \begin{bmatrix} 12x^2 & -4 \\ -4 & 12y^2 \end{bmatrix} \rightarrow H_f(0,0) = \begin{bmatrix} 0 & -4 \\ -4 & 0 \end{bmatrix} \rightarrow \text{eigen values } \lambda_1 = -4, \lambda_2 = 4 \rightarrow H_f(0,0) \text{ indefinite}$$

so $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ is saddle point.

$$\rightarrow H_f(1,1) = \begin{bmatrix} 12 & -4 \\ -4 & 12 \end{bmatrix} \rightarrow \text{eigen values } \lambda_1 = 8, \lambda_2 = 16 \rightarrow H(1,1) \text{ is positive def}$$

so $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is local min

$$\rightarrow H(-1,-1) = \begin{bmatrix} 12 & -4 \\ -4 & 12 \end{bmatrix} \rightarrow \text{eigen values } \lambda_1 = 8, \lambda_2 = 16 \rightarrow H(-1,-1) \text{ is positive def}$$

so $\begin{bmatrix} -1 \\ -1 \end{bmatrix}$ is local min

Since $H(x,y)$ is always positive def,

$\begin{bmatrix} 1 \\ 1 \end{bmatrix}^T$ and $\begin{bmatrix} -1 \\ -1 \end{bmatrix}^T$ are global min

c) $2x^3 - 3x^2 - 6xy(x-y-1)$

$$\nabla f(x,y) = \begin{bmatrix} 6x^2 - 12xy + 6y^2 - 6x + 6y \\ -6x^2 + 12xy + 6x \end{bmatrix} \rightarrow \nabla f(x,y) = 0 \rightarrow \begin{cases} (y-x)=0 \text{ or } (y-x+1)=0 \\ x=0 \text{ or } (2y-x+1)=0 \end{cases}$$

$$\rightarrow \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$H_f(x,y) = \begin{bmatrix} 12x - 12y - 6 & -12x + 12y + 6 \\ -12x + 12y + 6 & 12x \end{bmatrix}$$

$$H_f(0,0) = \begin{bmatrix} -6 & +6 \\ +6 & 0 \end{bmatrix} \rightarrow \text{indefinite so } \begin{bmatrix} 0 \\ 0 \end{bmatrix} \text{ is saddle point}$$

$$H_f(-1,-1) = \begin{bmatrix} -6 & 6 \\ 6 & -12 \end{bmatrix} \rightarrow \text{indefinite so } \begin{bmatrix} -1 \\ -1 \end{bmatrix} \text{ is saddle point}$$

$$H_f(0,-1) = \begin{bmatrix} 6 & -6 \\ -6 & 0 \end{bmatrix} \rightarrow \text{indefinite so } \begin{bmatrix} 0 \\ -1 \end{bmatrix} \text{ is saddle point}$$

$$H_f(1,0) = \begin{bmatrix} 6 & -6 \\ -6 & 6 \end{bmatrix} \rightarrow \text{positive so } \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ is local minimum}$$

There is no global min or max.

d) $(x-y)^4 + x^2 - y^2 - 2x + 2y + 1$

$$\nabla f(x,y) = \begin{bmatrix} 4(x-y)^3 + 2x - 2 \\ -4(x-y)^3 - 2y + 2 \end{bmatrix} \rightarrow \nabla f(x,y) = 0 \text{ at } \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$H_f(x,y) = \begin{bmatrix} 12(x-y) + 2 & -12(x-y)^2 \\ -12(x-y)^2 & 12(x-y)^2 - 2 \end{bmatrix} \rightarrow H_f(1,1) = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix} \rightarrow \text{indefinite}$$

So $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is saddle point, there is no global min and max

6.6-) Determine the critical points of the Lagrangian function for each of the following problems and determine whether each ~~of~~ the constrained minimum, maximum and neither.

a-) $f(x, y) = x^2 + y^2$ subject to $g(x, y) = x + y - 1 = 0$

$$\nabla f(x, y) = \begin{bmatrix} 2x \\ 2y \end{bmatrix} \quad J_g(x, y) = \nabla g(x, y) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}^T$$

$$\nabla L(x, \lambda) = \begin{bmatrix} 2x + 1 \cdot \lambda \\ 2y + 1 \cdot \lambda \\ x + y - 1 \end{bmatrix} = \begin{bmatrix} \nabla f(x, y) + J_g^T(x, y) \cdot \lambda \\ g(x) \end{bmatrix} \Rightarrow \begin{bmatrix} 0,5 \\ 0,5 \\ 1 \end{bmatrix} \xrightarrow{\text{critical point}}$$

$$B(x, y, \lambda) = \nabla_{xx} L(x, y, \lambda) = H_f(x) + \sum_{i=1}^m \lambda_i \cdot H_{g_i}(x)$$

$$H_f(x, y) = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad H_g(x, y) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$B(x, y, \lambda) = H_f + \lambda \cdot H_g = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad \text{The orthogonal matrix to } J_g(x, y) \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$z^T \cdot B \cdot z = \begin{bmatrix} 1 & -1 \end{bmatrix} \cdot \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 4 \quad \text{so } \begin{bmatrix} 0,5 \\ 0,5 \\ 1 \end{bmatrix} \text{ is constrained minimum.}$$

b-) $f(x,y) = x^3 + y^3$ subject to $g(x,y) = x + y - 1$

① $\nabla f(x,y) = \begin{bmatrix} 3x^2 \\ 3y^2 \end{bmatrix}$ $\nabla g = Jg(x,y) = [1, 1]$

② $\nabla L(x,y,\lambda) = \begin{bmatrix} \nabla f(x,y) + \lambda \cdot J^T(x,y) \\ g(x) \end{bmatrix} = \begin{bmatrix} 3x^2 + \lambda \\ 3y^2 + \lambda \\ x + y - 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0,5 \\ 0,5 \\ -0,75 \end{bmatrix} \xRightarrow{\text{critical point}}$

③ $B(x,y,\lambda) = Hf(x,y) + \lambda \cdot Hg(x,y)$

$$= \begin{bmatrix} 6x & 0 \\ 0 & 6y \end{bmatrix} + \lambda \cdot \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 6x & 0 \\ 0 & 6y \end{bmatrix}$$

④ The orthogonal vector to $Jg(x,y) = [1 \ 1]$ is $z = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

⑤ $z^T \cdot B \cdot z = [1 \ -1] \cdot \begin{bmatrix} 6 \cdot 0,5 & 0 \\ 0 & 6 \cdot 0,5 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 6 > 0$ so $\begin{bmatrix} 0,5 \\ 0,5 \\ -0,75 \end{bmatrix}$ is constrained minimum \Rightarrow

C-) $f(x,y) = 2x+y$ subject to $g(x,y) = x^2+y^2-1=0$

① $\nabla f(x,y) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ $\nabla g(x,y) = \begin{bmatrix} 2x & 2y \end{bmatrix}$

② $\nabla L = \begin{bmatrix} 2 + 2x\lambda \\ 1 + 2y\lambda \\ x^2 + y^2 - 1 \end{bmatrix} \Rightarrow \begin{matrix} x = -1/\lambda \\ y = -1/2\lambda \end{matrix} \Rightarrow \left(\frac{1}{\lambda}\right)^2 + \left(\frac{1}{2\lambda}\right)^2 = 1$

\Rightarrow critical points

$\begin{bmatrix} -0,894 \\ -0,447 \\ 1,12 \end{bmatrix}$ and $\begin{bmatrix} -0,894 \\ -0,447 \\ 1,12 \end{bmatrix}$

③ $B(x,y,\lambda) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + \lambda \cdot \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 2\lambda & 0 \\ 0 & 2\lambda \end{bmatrix}$

④ The orthogonal vector to ∇g is $\mathbf{z} = \begin{bmatrix} y \\ -x \end{bmatrix}$

⑤ $\mathbf{z}^T \cdot B \cdot \mathbf{z} = \begin{bmatrix} y & -x \end{bmatrix} \cdot \begin{bmatrix} 2\lambda & 0 \\ 0 & 2\lambda \end{bmatrix} \cdot \begin{bmatrix} y \\ -x \end{bmatrix} = y^2 \cdot 2\lambda + x^2 \cdot 2\lambda$

for $\begin{bmatrix} -0,894 \\ -0,447 \\ 1,12 \end{bmatrix} \Rightarrow \mathbf{z}^T \cdot B \cdot \mathbf{z} = 2,24 > 0$ so constrained min

for $\begin{bmatrix} -0,894 \\ -0,447 \\ -1,12 \end{bmatrix} \Rightarrow \mathbf{z}^T \cdot B \cdot \mathbf{z} = -2,24 < 0$ so constrained max

d-) $f(x,y) = x^2 + y^2$ subject to $g(x,y) = x \cdot y^2 - 1 = 0$

① $\nabla f(x,y) = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$, $Jg(x,y) = \begin{bmatrix} y^2 & 2xy \end{bmatrix}$

② $\nabla L(x,y,\lambda) = \begin{bmatrix} 2x + \lambda \cdot y^2 \\ 2y + \lambda \cdot 2xy \\ xy^2 - 1 \end{bmatrix} \Rightarrow \begin{cases} x = -\lambda \cdot y^2 / 2 \\ x = -1/\lambda \end{cases} \Rightarrow \begin{cases} \lambda^3 = -2 \\ \lambda = \underline{\underline{-1,25992 \dots}} \end{cases}$

$\begin{bmatrix} 0,794 \\ 1,12 \\ -0,126 \end{bmatrix}$

③ $B(x,y,\lambda) = \begin{bmatrix} 2 & 2\lambda y \\ 2\lambda y & 2 + 2\lambda x \end{bmatrix}$

④ $z = \begin{bmatrix} 2x \\ -y \end{bmatrix}$

⑤ $z^T \cdot B \cdot z = 15,1 > 0$ so $\begin{bmatrix} 0,794 \\ 1,12 \\ -1,26 \end{bmatrix}$ is constrained minimum.

2) Computer Problems

2.a) 6.3

Use a library routine, or one of your own design, to find a minimum of each of the following functions on the interval $[0, 3]$. Draw a plot of each function to confirm that it is unimodal.

(a) $f(x) = x^4 - 14x^3 + 60x^2 - 70x$.

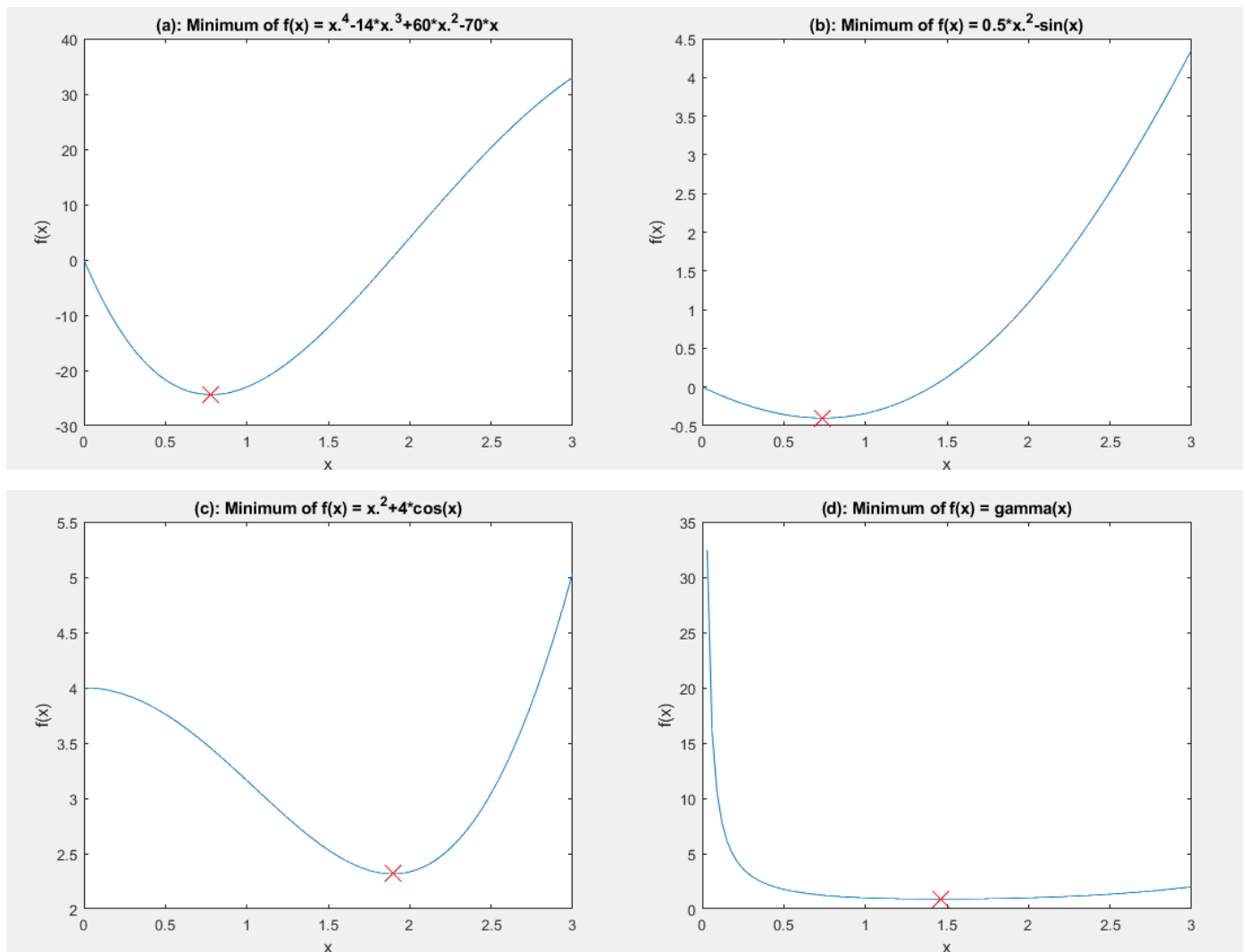
(b) $f(x) = 0.5x^2 - \sin(x)$.

(c) $f(x) = x^2 + 4 \cos(x)$.

(d) $f(x) = \Gamma(x)$.

$$\Gamma(x) = \int_0^\infty t^{x-1} e^{-t} dt, \quad x > 0,$$

1. The equations are defined.
2. The minimum of equation in bounded interval find by using `fminbnd(equation, 0, 3)`.
3. The equation is plotted and then the minimum point is marked.

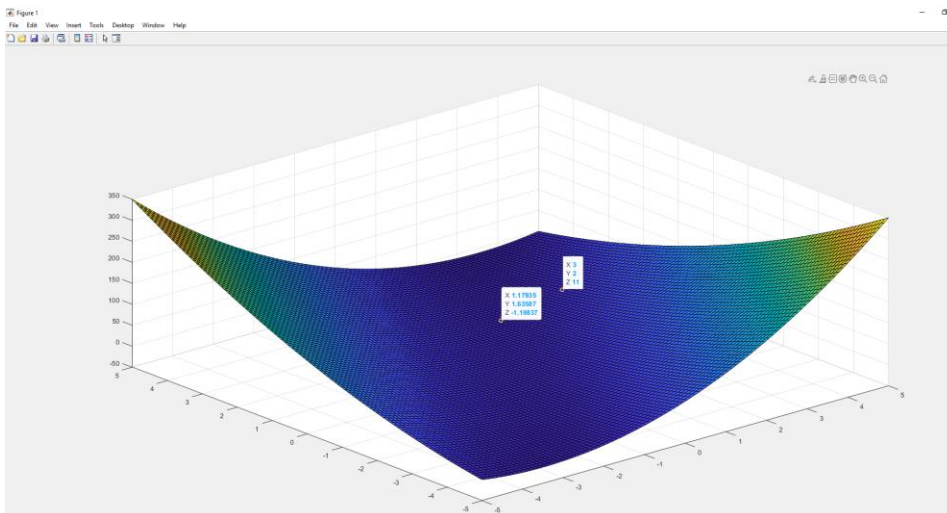


2.b) 6.6

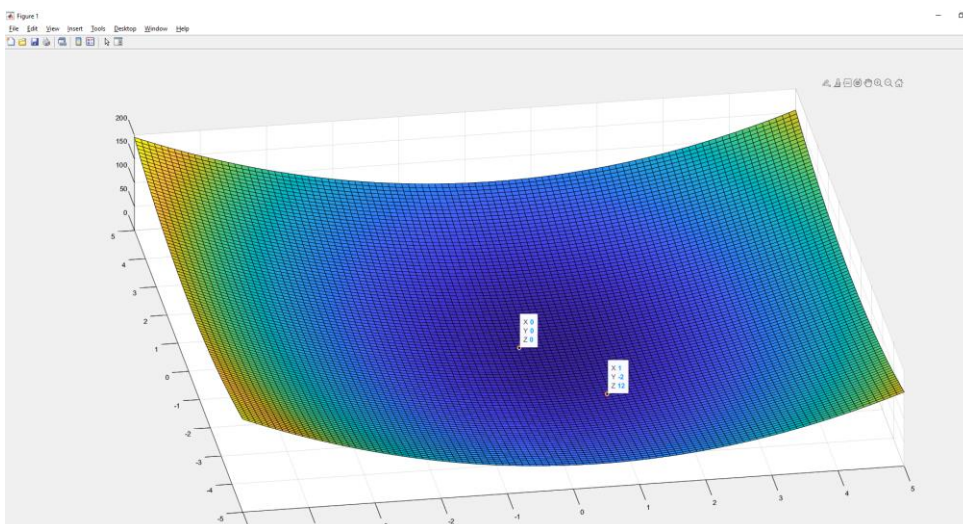
Write a general-purpose line search routine. Your routine should take as input a vector defining the starting point, a second vector defining the search direction, the name of a routine defining the objective function, and a convergence tolerance. For the resulting one-dimensional optimization problem, you may call a library routine or one of your own design. In any case, you will need to determine a bracket for the minimum along the search direction using some heuristic procedure. Test your routine for a variety of objective functions and search directions. This routine will be useful in some of the other computer exercises in this section.

The LineSearch algorithm created by using library routine function 'fminbnd' and iteration for each step. By doing these optimal minimum point can be found.

1) $x_0 = [3; 2];$
 $s = [-5; -1];$
 $y = 5 * x(1)^2 + 2 * x(2)^2 - 7 * x(1) * x(2);$



2) $x_0 = [1; -2];$
 $s = [-1; 2];$
 $Y = 5 * X1.^2 + 2 * X2.^2 - X1.*4;$



2.c) 6.9

Write a program to find a minimum of Rosenbrock's function

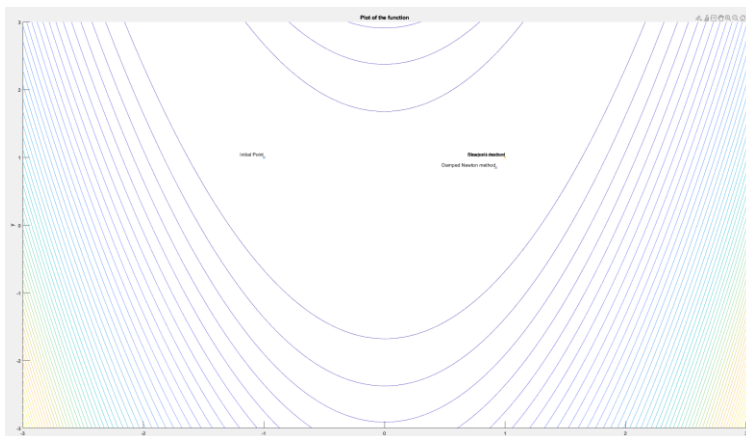
$$f(x, y) = 100(y - x^2)^2 + (1 - x)^2$$

using each of the following methods:

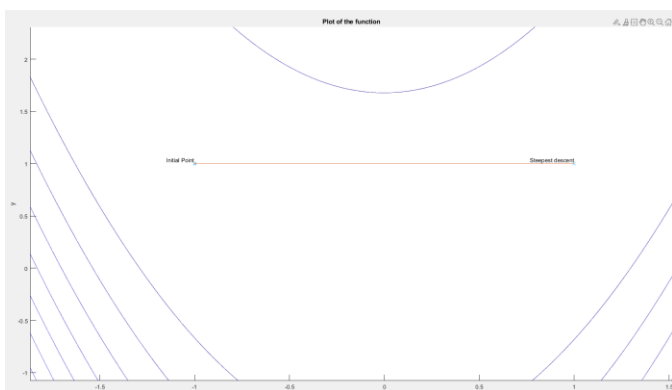
- (a) Steepest descent
- (b) Newton
- (c) Damped Newton (Newton's method with a line search)

You should try each of the methods from each of the three starting points $[-1 \ 1]^T$, $[0 \ 1]^T$, and $[2 \ 1]^T$. For any line searches and linear system solutions required, you may use either library routines or routines of your own design. Plot the path taken in the plane by the approximate solutions for each method from each starting point.

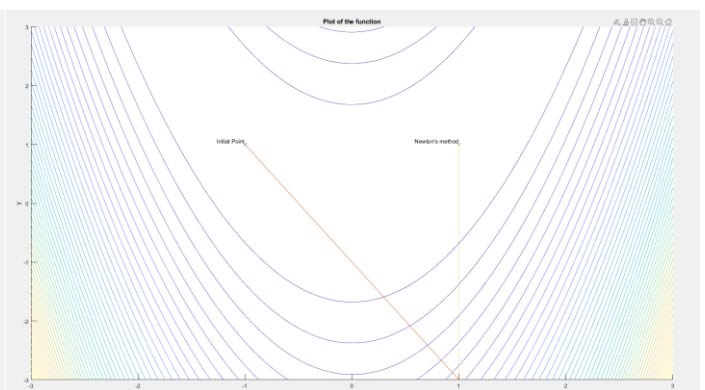
Plot 1, starting point = $[-1 \ 1]^T$



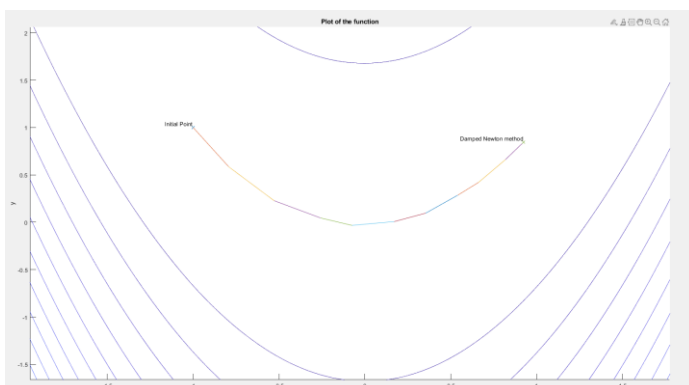
Steepest Descent



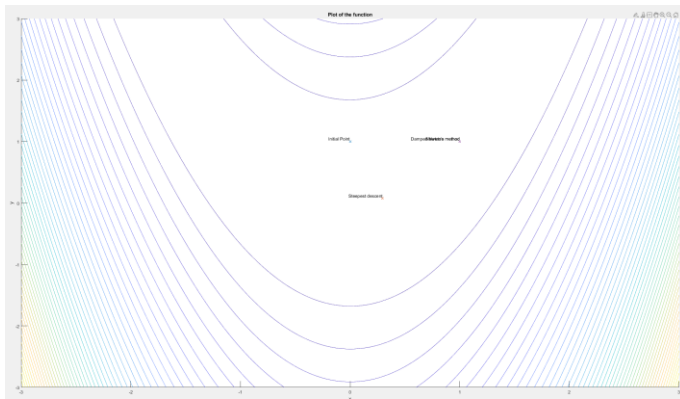
Newton's Method



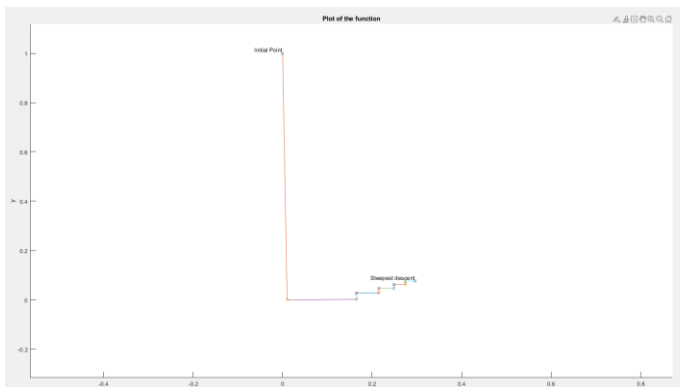
Damped Newton method:



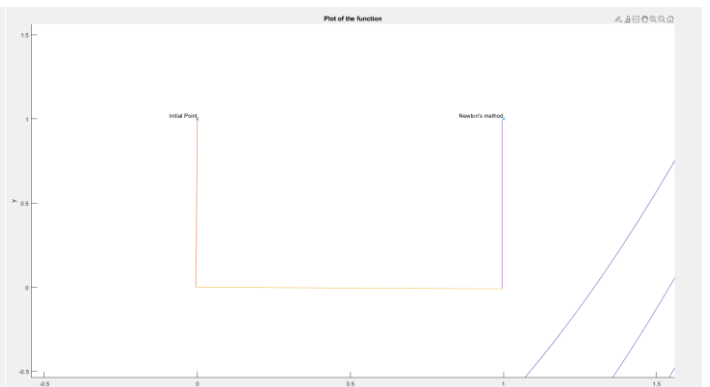
Plot 2, starting point = [0 1]T



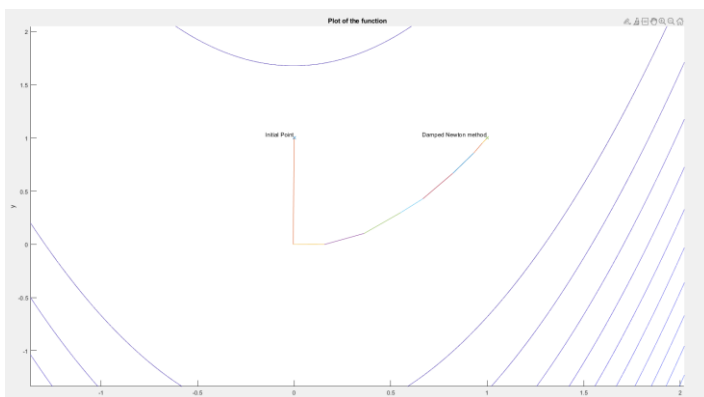
Steepest Descent



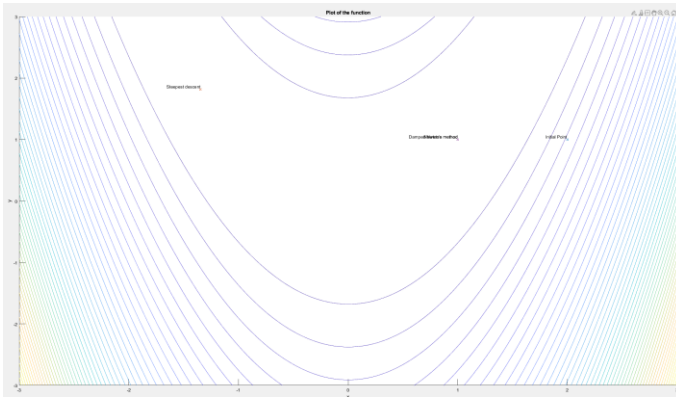
Newton's Method



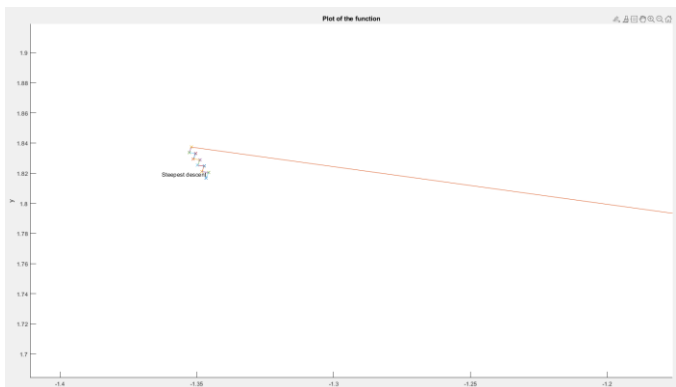
Damped Newton method:



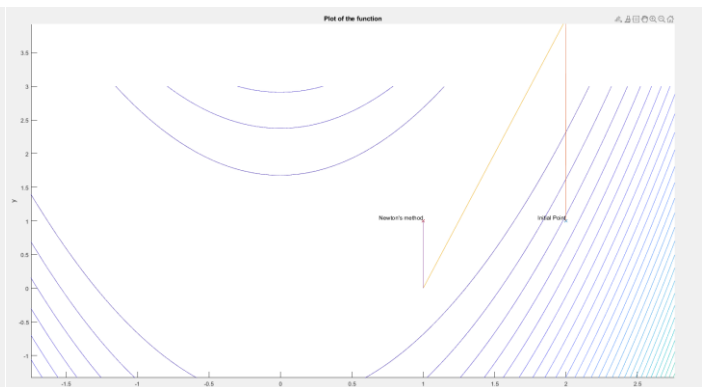
Plot 3, starting point = $[2 \ 1]^T$



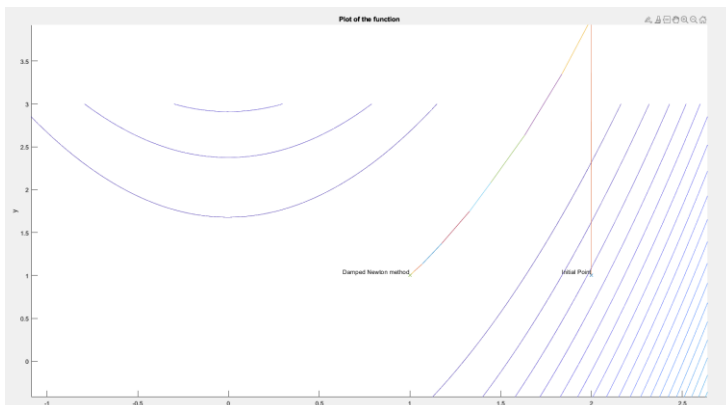
Steepest Descent



Newton's Method



Damped Newton method:



2.d) 6.14 The concentration of a drug in the bloodstream is expected to diminish exponentially with time. We will fit the model function

$$y = f(t, x) = x_1 e^{x_2 t}$$

to the following data:

t 0.5 1.0 1.5 2.0

y 6.80 3.00 1.50 0.75

t 2.5 3.0 3.5 4.0

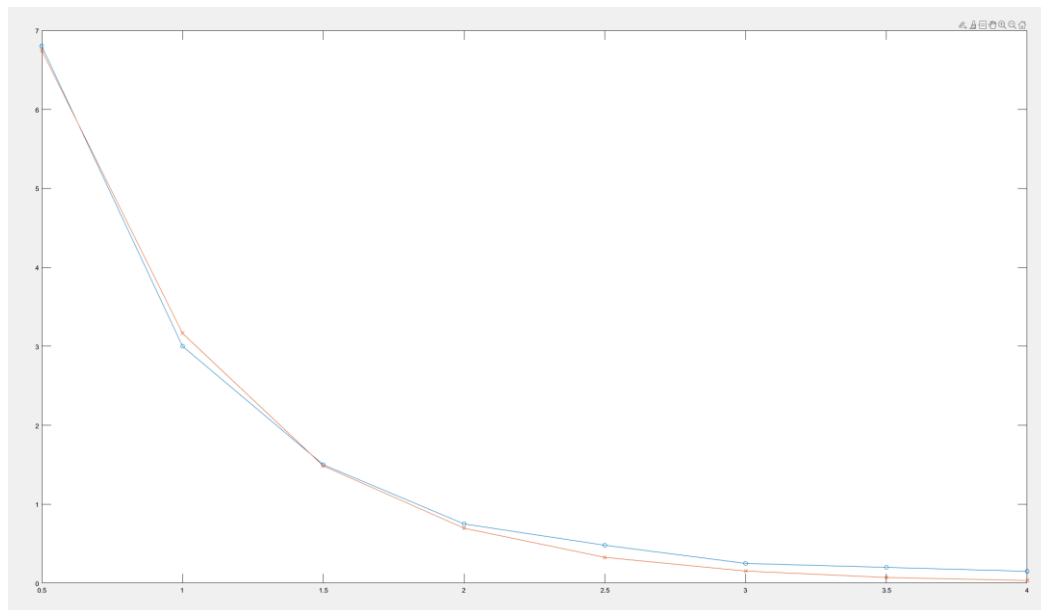
y 0.48 0.25 0.20 0.15

(a) Perform the exponential fit using nonlinear least squares. You may use a library routine or one of your own design, perhaps using the GaussNewton method.

(b) Taking the logarithm of the model function gives $\log(x_1) + x_2 t$, which is now linear in x_2 . Thus, an exponential fit can also be done using linear least squares, assuming that we also take logarithms of the data points y_i . Use linear least squares to compute x_1 and x_2 in this manner. Do the values obtained agree with those determined in part a? Why?

The blue line is nonlinear least square with curve fitting.

The red line is linear least square.



There is differences between linear and non-linear fit. It is because first one adjust its variables for best fit to the values and the other is adjust for exact fit for the equation.

```
>> CP_6_14
(a) Nonlinear fit:
    14.3766
    -1.5139

(b) Linear fit:

x =

    8.6350
   -1.0967
```