

## ISYE 6669 – HOMEWORK 4

### 1 – CONVEX OPTIMIZATION

#### ANS 1.1 -

$$f(x) = e^x - x$$

$$f'(x) = e^x - 1$$

$$f''(x) = e^x$$

$$x^{k+1} = x^k + f'(x^k) / f''(x^k)$$

$$x^0 = -1$$

Let's convert this to a simple program to do this for us –

```
e = 2.71828183
x0 = -1
der1 = (e**x0) - 1
der2 = (e**x0)
count = 1

while abs(der1) >= 1/(10**5):

    print("\nIteration ",count,': ')
    print('X value: ',x0)
    print('First Derivative: ',der1)
    print('Second Derivative: ',der2)
    x0 = x0 - (der1/der2)
    der1 = (e**x0) - 1
    der2 = (e**x0)
    count += 1
```

Upon running a python program with a while loop conditioned as mentioned in the problem – the absolute value of first derivative must be lesser than  $10^{-5}$ , we get –

Iteration 1 :

X value: -1

First Derivative: -0.6321205590371033

Second Derivative: 0.36787944096289676

Iteration 2 :

X value: 0.7182818300000002

First Derivative: 1.0509063766879514

Second Derivative: 2.0509063766879514

Iteration 3 :

X value: 0.20587112776936967

First Derivative: 0.22859486217556269

Second Derivative: 1.2285948621755627

Iteration 4 :

X value: 0.019809091199582074

First Derivative: 0.02000659321423348

Second Derivative: 1.0200065932142335

Iteration 5 :

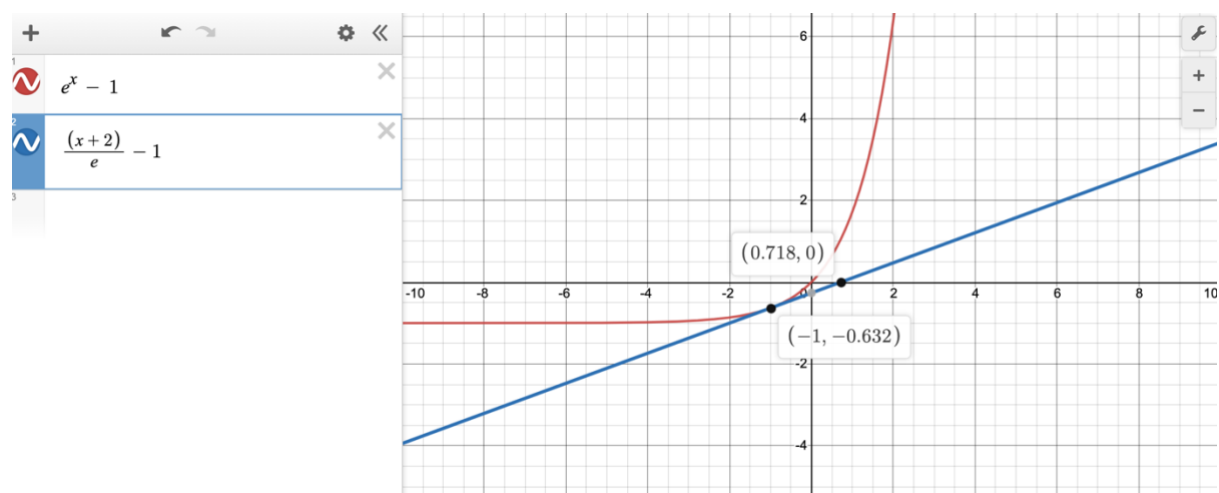
X value: 0.0001949109116008707

First Derivative: 0.00019492990807723487

Second Derivative: 1.0001949299080772

## ANS 1.2 -

Tangent 1 at  $x^0$  –



In the form  $k(x-a) + b = 0$ , the tangent would be –

$$(x - (e - 2)) = 0$$

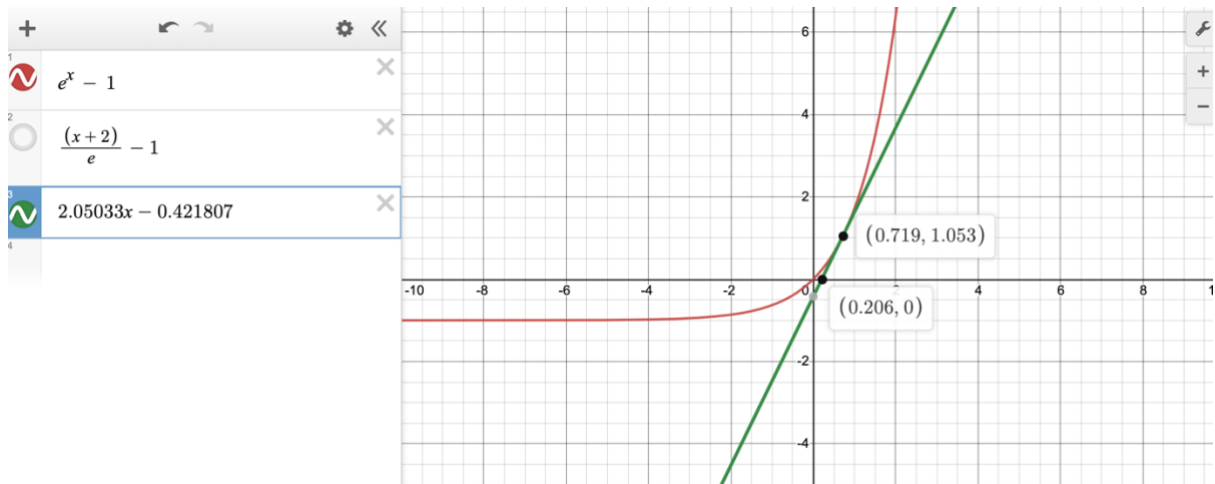
$$k = 1$$

$$a = e - 2$$

$$b = 0$$

$$x^1 = 0.718 \text{ (value from Newton's method = 0.2718)}$$

Tangent 1 at  $x^1$  –



In the form  $k(x-a) + b = 0$ , the tangent would be –

$$2.05033(x) + (-0.421807) = 0$$

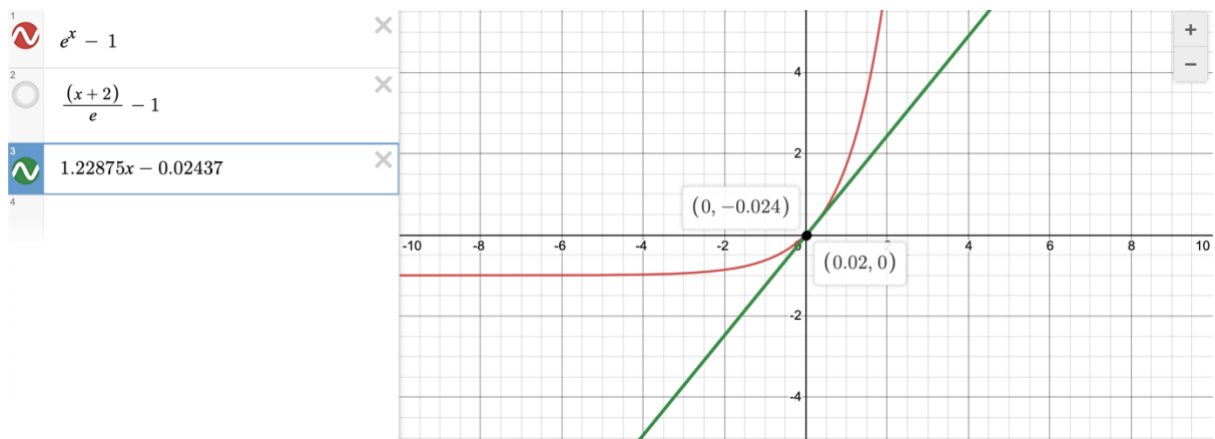
$$k = 2.05033$$

$$a = 0$$

$$b = -0.421807$$

$$x^2 = 0.206 \text{ (value from Newton's method} = 0.2058)$$

Tangent 1 at  $x^2$  –



In the form  $k(x-a) + b = 0$ , the tangent would be –

$$1.22875(x) + (-0.02437) = 0$$

$$k = 1.22875$$

$$a = 0$$

$$b = -0.02437$$

$$x^3 = 0.02 \text{ (value from Newton's method} = 0.019)$$

The new values from these observations match the new values from part 1!

## 2 – NONCONVEX OPTIMIZATION

### ANS 2.1 -

Minimum  $(x_1, x_2) = (1.47568569, -0.32234972)$

Objective Value = 1.6129287968151138

Code for finding global minimum using the BFGS update method –

```
import scipy.optimize as optimize

def f(x):

    x1 = x[0]
    x2 = x[1]
    y1 = (1-x1+(x1*x2))**2
    y2 = (2-x1+((x1**2)*x2))**2
    y3 = (3-x1+((x1**2)*x2))**2
    return y1 + y2 + y3

bnds = ((-5, 5), (-5, 5))

res = optimize.minimize(fun=f, x0=[0,0], method='L-BFGS-B', bounds=bnds)
res.x
```

Run in jupyter –

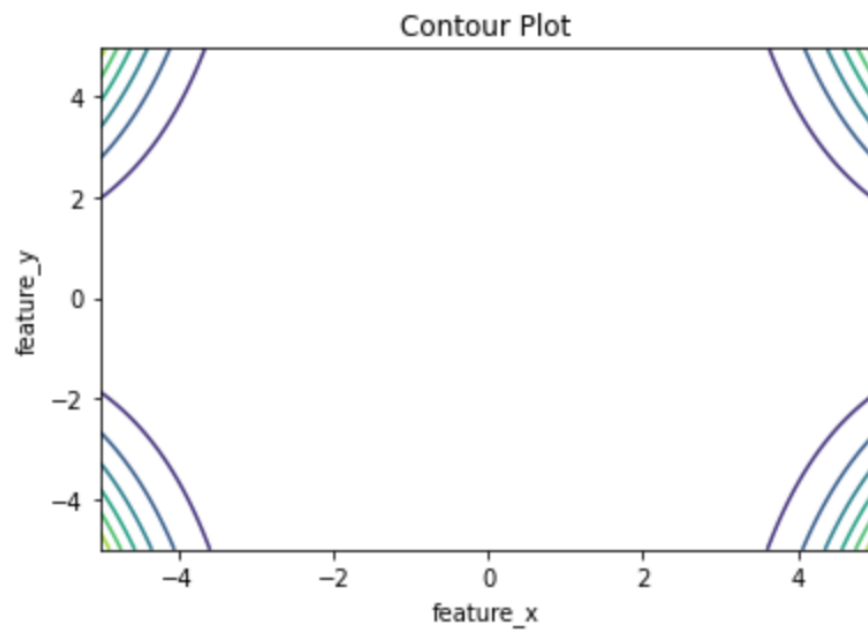
```
1 import scipy.optimize as optimize

1 def f(x):
2
3     x1 = x[0]
4     x2 = x[1]
5     y1 = (1-x1+(x1*x2))**2
6     y2 = (2-x1+((x1**2)*x2))**2
7     y3 = (3-x1+((x1**2)*x2))**2
8     return y1 + y2 + y3
9
10 bnds = ((-5, 5), (-5, 5))

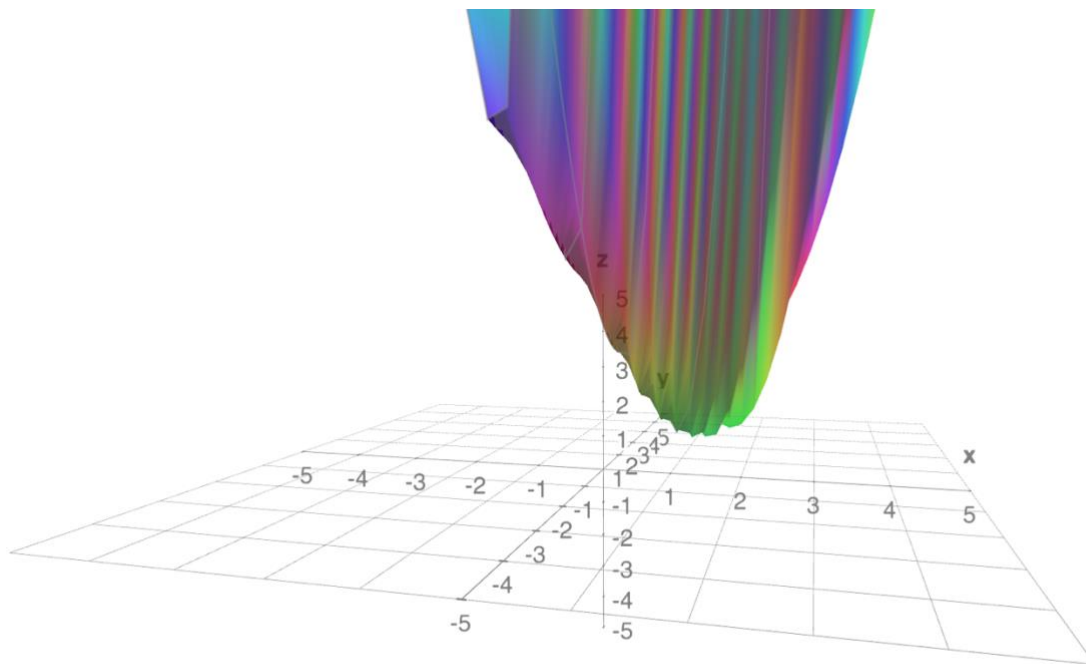
1 res = optimize.minimize(fun=f, x0=[0,0], method='L-BFGS-B', bounds=bnds)
2 res

fun: 1.6129287968151138
hess_inv: <2x2 LbfgsInvHessProduct with dtype=float64>
jac: array([-6.86117833e-06,  6.15063556e-06])
message: b'CONVERGENCE: NORM_OF_PROJECTED_GRADIENT_ <= _PGTOL'
nfev: 42
nit: 11
njev: 14
status: 0
success: True
x: array([ 1.47568569, -0.32234972])
```

2D Contour Plot –



3D Plot –



**ANS 2.2 -**

For global minima –

$$F1 = -2.062605516810367$$

```
X1, X2 = [-1.349385, -1.349385]
X1, X2 = [ 1.34938658, -1.3493853 ]
X1, X2 = [-1.3493853, 1.34938658]
X1, X2 = [-1.3493853, 1.34938658]
```

For local minima –

```
F = -1.554458311873118
X1, X2 = [-7.63259085 7.63259226]
```

**Note** – Both global and local minima were derived from looking at contour plots within the interval of [-10, 10] and taking starting points within likely regions.

This is the code for a local minimum –

```
import scipy.optimize as optimize

def f(x):

    x1 = x[0]
    x2 = x[1]
    pwr = abs(100 - (((x1**2 + x2**2)**0.5)/3.14159))
    inner = abs(np.sin(x1)*np.sin(x2)*(2.718281**pwr))
    outer = -0.0001*(inner-1)**0.1
    return outer

bnds = ((-10, 10), (-10, 10))

res = optimize.minimize(fun=f,x0=[-7.5,7.5],method='L-BFGS-B',bounds=bnds)
res
```

Run in jupyter for global minima as well –

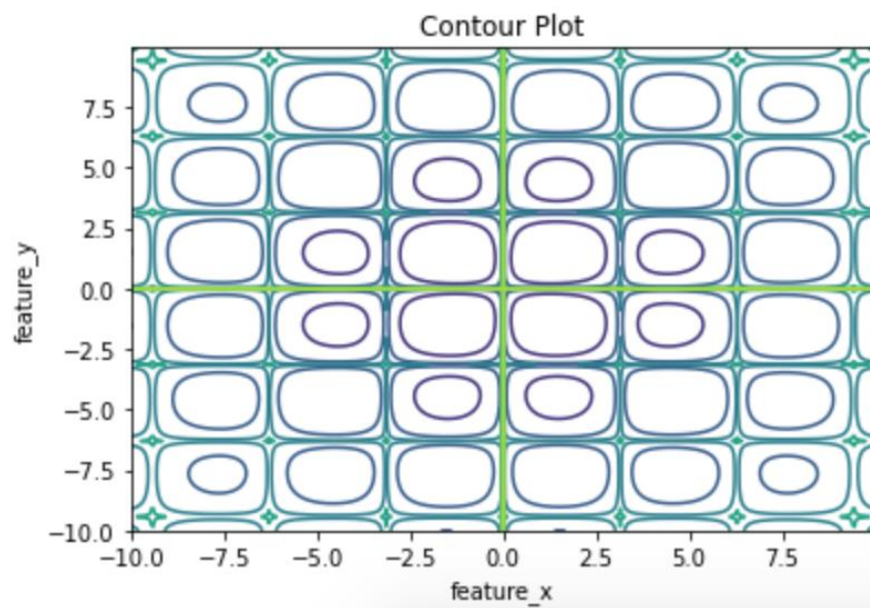
```
1 import scipy.optimize as optimize

1 def f(x):
2
3     x1 = x[0]
4     x2 = x[1]
5     pwr = abs(100 - (((x1**2 + x2**2)**0.5)/3.14159))
6     inner = abs(np.sin(x1)*np.sin(x2)*(2.718281**pwr))
7     outer = -0.0001*(inner-1)**0.1
8     return outer
9
10 bnds = ((-10, 10), (-10, 10))

1 #res = optimize.minimize(fun=f,x0=[-7.5,7.5],method='L-BFGS-B',bounds=bnds)
2 res1 = optimize.minimize(fun=f,x0=[-1.0,-1.0],method='L-BFGS-B',bounds=bnds)
3 res2 = optimize.minimize(fun=f,x0=[1.0,-1.0],method='L-BFGS-B',bounds=bnds)
4 res3 = optimize.minimize(fun=f,x0=[-1.0,1.0],method='L-BFGS-B',bounds=bnds)
5 res4 = optimize.minimize(fun=f,x0=[1.0,1.0],method='L-BFGS-B',bounds=bnds)
6 print(res1.fun)
7 print(res2.fun)
8 print(res3.fun)
9 print(res4.fun)

-2.0626055168018405
-2.062605516810367
-2.062605516810367
-2.062605516816397
```

2D Contour Plot –



3D Plot –

