

"""

Homework 4 Problem 13.13

Develop a program using a programming or macro language to implement the golden-section search algorithm. Design the program so that it is expressly designed to locate a maximum or minimum based on user preference. The subroutine should have the following features:

- iterate until the relative error falls below a stopping criterion or exceeds a maximum number of iterations.
- return both the optimal x and $f(x)$
- minimize the number of function evaluations

@author: Jacob Needham

"""

#getting user decision on max or min

```
def wantMaxOrMin():
    count = 0
    answer = False
    while answer != True and count < 5:
        print("Type \"MAX\" if you would like to find the max or \"MIN\" \
              to find the minimum of the function")
        user = input()
        count += 1
        if user == "MAX":
            return True
            break
        if user == "MIN":
            return False
            break
        elif user != "MAX" or user != "MIN":
            print("\nInvalid input, try again")
```

#function equation, in a real problem the function would come from the user

```
def function(x):
    y = (x**3 - 30*x**2 - 661*x - 1791)/1791
    return y
```

#Golden search subroutine

```
def goldenSearchMax(es, iMax, L, U):
    iCount = 0
    ea = es
    R = ((5**.5) - 1)/2
    xL = L
    xU = U
    d = R * (xU - xL)
    x1 = xL + d
    x2 = xU - d
    f1 = function(x1)
    f2 = function(x2)
    if f1 > f2 :
        xopt = x1
        fx = f1
    else:
        xopt = x2
        fx = f2
```

```

for iCount in range(iMax):
    d = R*d
    xinit = xU - xL
    if f1 > f2:
        xL = x2
        x2 = x1
        x1 = xL + d
        f2 = f1
        f1 = function(x1)
    else:
        xU = x1
        x1 = x2
        x2 = xU - d
        f1 = f2
        f2 = function(x2)
    iCount += 1
    if f1 > f2:
        xopt = x1
        fx = f1
    else:
        xopt = x2
        fx = f2
    if xopt != 0:
        ea = (1-R)* abs(xinit/xopt)*100
    if ea < es:
        break
print("\nNumber of iterations to convergence:", iCount)
print("The optimal value is:", round(fx,4))
print("The x value for the optimal value is:", round(xopt,4))

```

```

def goldenSearchMin(es,iMax,L,U):
    iCount = 0
    ea = es
    R = ((5**0.5)-1)/2
    xL = L
    xU = U
    d = R *(xU - xL)
    x1 = xL + d
    x2 = xU - d
    f1 = function(x1)
    f2 = function(x2)
    if f1<f2 :
        xopt = x1
        fx = f1
    else:
        xopt = x2
        fx = f2
    for iCount in range(iMax):
        d = R*d
        xinit = xU - xL
        if f1 < f2:
            xL = x2
            x2 = x1
            x1 = xL + d
            f2 = f1
            f1 = function(x1)

```

```

else:
    xU = x1
    x1 = x2
    x2 = xU - d
    f1 = f2
    f2 = function(x2)
    iCount += 1
    if f1 < f2:
        xopt = x1
        fx = f1
    else:
        xopt = x2
        fx = f2
    if xopt != 0:
        ea = (1-R)* abs(xinit/xopt)*100
    if ea < es:
        break
print("\nNumber of iterations to convergence:", iCount)
print("The minimal value is:", round(fx,4))
print("The x value for the minimal value is:", round(xopt,4))

def main():
    answer = wantMaxOrMin()

    if answer == True:
        goldenSearchMax(.001,200,-20,0)
    else:
        goldenSearchMin(.001,200,-20,0)
main()

```

OUTPUT

Type "MAX" if you would like to find the max or "MIN"
to find the minimum of the function

MAX

Number of iterations to convergence: 25
The optimal value is: 0.595
The x value for the optimal value is: -7.8978

```
"""
```

Homework 4 Problem 13.15

Develop a program using a programming or macro language to implement Newton's method. The subroutine should have the following features:

- Iterate until the relative error falls below a stopping criterion or exceeds a maximum number of iterations.
- Returns both the optimal x and f(x).

Test your program with the same problem as Example 13.3.

@author: Jacob Needham

```
"""
```

```
#function equation
```

```
def function(x):
```

```
    y = -1.5*x**6 - 2*x**4 + 12*x
```

```
    return y
```

```
#first derivative
```

```
def functionDer(x):
```

```
    y = -9*x**5 - 8*x**3 + 12
```

```
    return y
```

```
#second derivative
```

```
def functionDer2(x):
```

```
    y = -45*x**4 - 24*x**2
```

```
    return y
```

```
#Newton Raphson technique
```

```
def newtonMethod(x0,es,iMax):
```

```
    ea=es
```

```
    xR = x0
```

```
    iCount = xOld = ea = None
```

```
    for iCount in range(0,iMax):
```

```
        iCount +=1
```

```
        xOld = xR
```

```
        xR = xOld - (functionDer(xOld)/functionDer2(xOld))
```

```
        if xR != 0:
```

```
            ea = abs((xR-xOld)/xR)*100
```

```
        if ea<es or iCount > iMax:
```

```
            break
```

```
    print("The max was found after", iCount,"iterations")
```

```
    print("The x value of the max of the function is: ", round(xR,4))
```

```
    print("The function evaluated at the max is:", round(function(xR),4))
```

```
    print("The error is: ", round(ea*100,4) , "%")
```

```
newtonMethod(2, .001, 30)
```

OUTPUT

The max was found after 7 iterations

The x value of the max of the function is: 0.9169

The function evaluated at the max is: 8.6979

The error is: 0.0197 %

```

# -*- coding: utf-8 -*-
"""
Homework 4 Problem 13.20
The normal distribution is a bell-shaped curve defined by  $y=e^{-x^2}$ .
Use the golden-section search to determine the location of the inflection
point of this curve for positive x.

@author: Jacob Needham
"""
import math

#function equation
def function(x):
    y = math.exp(-x**2)
    return y

def functionDer(x):
    y = -2*x*math.exp(-x**2)
    return y

def functionDer2(x):
    y = (4*x**2-2)*math.exp(-x**2)
    return y

def goldenSearchMin(es,iMax,L,U):
    iCount = 0
    ea = es
    R = ((5**.5)-1)/2
    xL = L
    xU = U
    d = R *(xU - xL)
    x1 = xL + d
    x2 = xU - d
    f1 = functionDer(x1)
    f2 = functionDer(x2)
    if f1<f2 :
        xopt = x1
        fx = f1
    else:
        xopt = x2
        fx = f2
    for iCount in range(iMax):
        d = R*d
        xinit = xU - xL
        if f1 < f2:
            xL = x2
            x2 = x1
            x1 = xL + d
            f2 = f1
            f1 = functionDer(x1)
        else:
            xU = x1
            x1 = x2
            x2 = xU - d
            f1 = f2
            f2 = functionDer(x2)

```

```

iCount += 1
if f1 < f2:
    xopt = x1
    fx = f1
else:
    xopt = x2
    fx = f2
if xopt != 0:
    ea = (1-R)* abs(xinit/xopt)*100
if ea < es:
    break
print("The x value for the inflection point is:",round(function(fx),4))
print("Number of iterations to convergence:", iCount)

```

```
goldenSearchMin(.001,100,-20,20)
```

OUTPUT

```

The x value for the inflection point is: 0.4791
Number of iterations to convergence: 32

```

```

"""
Homework 4 Problem 14.9
Develop a program using a programming or macro language to implement the
random search method. Design the subprogram so that it is expressly designed
to locate a maximum. Test the program with f(x, y) from Prob. 14.7. Use a
range of -2 to 2 for both x and y.

@author: Jacob Needham
"""

import random

#function equation
def function(x,y):
    z = 4*x + 2*y + x**2 - 2*x**4 + 2*x*y - 3*y**2
    return z

#Random Search takes in a number of points and finds the max value of a function
def randomSearch(maxNumPoints):
    listOfPoints = []
    for i in range(maxNumPoints):
        randPointX = random.uniform(-2,2)
        randPointY = random.uniform(-2,2)
        listOfPoints.append ([function(randPointX,randPointY),randPointX,randPointY])
    index = listOfPoints.index(max(listOfPoints))
    domainMax = listOfPoints[index][0]
    xValue = listOfPoints[index][1]
    yValue = listOfPoints[index][2]

    print("The max of the set")
    print("4*x + 2*y + x^2 - 2*x^4 + 2*x*y - 3*y^2")
    print("bounded by -2<=x<=2 and -2<=y<=2 and with", maxNumPoints, "searches is:")
    print(round(domainMax,3),
          " at x=", round(xValue,3),
          ", y=", round(yValue,3),sep="")

randomSearch(1000)

```

OUTPUT

```

The max of the set
4*x + 2*y + x^2 - 2*x^4 + 2*x*y - 3*y^2
bounded by -2<=x<=2 and -2<=y<=2 and with 1000
searches is:
4.271 at x=0.899, y=0.534

```

```
"""
```

Homework 4 Problem 14.10

The grid search is another brute force approach to optimization. The two-dimensional version is depicted in Fig. P14.10. The x and y dimensions are divided into increments to create a grid. The function is then evaluated at each node of the grid. The denser the grid, the more likely it would be to locate the optimum.

Develop a program using a programming or macro language to implement the grid search method. Design the program so that it is expressly designed to locate a maximum. Test it with the same problem as Example 14.1.

@author: Needh

```
"""
```

```
import numpy as np
```

```
def function(x,y):
```

```
    z = y - x - 2*x**2 - 2*x*y - y**2
```

```
    return z
```

```
#Golden search subroutine
```

```
def randomSearch(gridIncrement,xLower,xUpper,yLower,yUpper):
```

```
    listOfPoints = []
```

```
    xStep = np.arange(xLower,xUpper,gridIncrement)
```

```
    yStep = np.arange(yLower,yUpper,gridIncrement)
```

```
    for x in xStep:
```

```
        for y in yStep:
```

```
            listOfPoints.append([function(x,y),x,y])
```

```
    index = listOfPoints.index(max(listOfPoints))
```

```
    domainMax = listOfPoints[index][0]
```

```
    xValue = listOfPoints[index][1]
```

```
    yValue = listOfPoints[index][2]
```

```
    print("The max of the set")
```

```
    print("y - x - 2*x^2 - 2*x*y - y^2")
```

```
    print("bounded by ",xLower,"<=x<=",xUpper," and ",yLower, "<=y<=",yUpper,\n          " with a grid increment of ", gridIncrement, " is:",sep="")
```

```
    print(round(domainMax,3)," located at x=",xValue," y=",yValue,sep="")
```

```
randomSearch(.005,-2,2,1,3)
```

OUTPUT

The max of the set

y - x - 2*x^2 - 2*x*y - y^2

bounded by -2<=x<=2 and 1<=y<=3 with a grid increment
of 0.005 is:

1.25 located at x=-1.0, y=1.5

Mineral methods

Homework #4 Problem #6

Find the gradient & Hessian for the following functions.

1) $f(x,y) = \ln(x^2 + 3xy + 2y^2)$

$$\nabla f(x,y) = \begin{bmatrix} \frac{2x + 3y}{x^2 + 3xy + 2y^2} \\ \frac{3x + 4y}{x^2 + 3xy + 2y^2} \end{bmatrix}$$

$$\text{Hessian } f(x,y) = \begin{vmatrix} \frac{-2x^2 + 6xy + 5y^2}{(x^2 + 3xy + 2y^2)^2} & \frac{-2x + 3y}{(x^2 + 3xy + 2y^2)^2} \\ \frac{-2x + 3y}{(x^2 + 3xy + 2y^2)^2} & \frac{-5x^2 + 6xy + 8y^2}{(x^2 + 3xy + 2y^2)^2} \end{vmatrix}$$

$$= \frac{(-2x^2 + 6xy + 5y^2)(-5x^2 + 6xy + 8y^2)}{(x^2 + 3xy + 2y^2)^4} - \frac{(2x + 3y)^2}{(x^2 + 3xy + 2y^2)^4}$$

3) $f(x,y,z) = x^2 + y^2 + 3z^2$

$$\nabla f(x,y,z) = \begin{bmatrix} 2x \\ 2y \\ 6z \end{bmatrix}$$

$$\text{Hessian} = \begin{vmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 6 \end{vmatrix} = \underline{\underline{24}}$$

Homework #4 Problem #7

A) Start w/ an initial guess of $(x_0, y_0) = (1, 1)$ and apply two iterations of the steepest ascent method to maximize $f(x, y)$.

$$f(x, y) = 2xy + 2y - 1.5x^2 - 2y^2$$

$$\nabla f(x, y) = \begin{bmatrix} 2y - 3x \\ 2x + 2 - 4y \end{bmatrix}$$

$$\nabla f(x, y) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

$$\begin{aligned} g(h) &= f(1-h, 1) \\ &= 2(1-h) + 2 - 1.5(1-h)^2 - 2(1)^2 \\ &= .5h - 1.5h^2 \end{aligned}$$

$$g'(h) = -3h + 1 \quad h^* = 1/3$$

$$\begin{aligned} (x_1, y_1) &= (x_0, y_0) + h \nabla f(x_0, y_0) \\ &= (1, 1) + \frac{1}{3}(-1, 0) \end{aligned}$$

Iteration #1 $\rightarrow \underline{\underline{(2/3, 1)}}$

$$\begin{aligned} g(h) &= f\left(\frac{2}{3}, 1\right) + h(0, 2/3) \\ &= f\left(\frac{2}{3}, 1 - \frac{2}{3}h\right) \\ &= 2\left(\frac{2}{3}\right)\left(1 - \frac{2}{3}h\right) + 2\left(1 - \frac{2}{3}h\right) - 1.5\left(\frac{2}{3}\right)^2 - 2\left(1 - \frac{2}{3}h\right)^2 \\ &= -4(h + .5)(2h - 3) \end{aligned}$$

$$g'(h) = .8 - 1.7h \quad h^* = \frac{4}{17}$$

$$\begin{aligned} (x_2, y_2) &= (x_1, y_1) + h \nabla f(x_1, y_1) \\ &= (2/3, 1) + \frac{1}{2}(0, 2/3) \end{aligned}$$

Iteration #2 $\rightarrow \underline{\underline{(2/3, 5/3)}}$

B)

$$f(x, y) = 2xy + 2y - 1.5x^2 - 2y^2$$

$$\nabla f(x, y) = \begin{bmatrix} 2y - 3x \\ 2x - 4y + 2 \end{bmatrix}$$

$$\frac{\partial^2 f}{\partial x^2} = -3$$

$$\frac{\partial^2 f}{\partial x \partial y} = 2$$

$$\frac{\partial^2 f}{\partial y^2} = -4$$

$$\text{Hessian } f(x, y) = \begin{vmatrix} -3 & 2 \\ 2 & -4 \end{vmatrix} = 8$$

Because $|H|$ is > 0 & $\frac{\partial^2 f}{\partial x^2} < 0$, the value is a max

Solving for max point:

$$2y - 3x = 0 \quad x = 1/2$$

$$2x - 4y + 2 = 0 \quad \underline{\underline{y = 3/4}}$$