Homework 2 MA 574 - Python

Numerical Differentiation

How to find the derivative of a function numerically?

. Forward Difference Formula

$$f'(a)pprox rac{f(a+h)-f(a)}{h}$$

. Backward Difference Formula

$$f'(a)pprox rac{f(a)-f(a-h)}{h}$$

. Central Difference Formula

$$f'(a)pprox rac{f(a+h)-f(a-h)}{2h}$$

See this book for more details and Python implementation.

```
In [4]: # Forward Difference Formula
    def fdiff(f,a,h):
        derivative_forward_diff = (f(a+h)-f(a))/h
        return derivative_forward_diff

In [5]: def bdiff(f,a,h):
        derivative_backward_diff = (f(a)-f(a-h))/h
        return derivative_backward_diff

In [6]: def cdiff(f,a,h):
        derivative_central_diff = (f(a+h)-f(a-h))/(2*h)
```

Numerical Integration

return derivative_central_diff

General Riemann Sum

$$\int_a^b f(x) \; dx pprox \sum_{i=1}^n f(x_i^*) \; \Delta x,$$

where x_i^* is the leftmost point of the i-th subinterval for left Riemann sum, rightmost point for right Riemann sum and middle-point for the mid-point sum.

```
In [ ]:
In [7]:
         Following function could be used to find all kinds of Riemann Sums by adjuting the
         argument. Set it = 0 for left sum, =1 for right sum and 0.5 for mid point sum.
         def RiemannSum(f,a,b,n,shift=0):
             if shift < 0 or shift > 1:
                  print("Please provide appropriate value for the shift from 0 to 1.0.")
                  return
             deltax = (b-a)/n
             sum=0.0
             a = a+shift*deltax
             for i in range(n):
                  sum = sum + f(a+i*deltax)
             return sum*deltax
In [10]: ## Example
         f = lambda x: 3*x**2
         L40 = RiemannSum(f,0,2,600, shift=1.0)
         print(L40)
```

8.02001111111119

Trapezoidal Rule

The trapezoidal rule for numerical approximation of a definite integral could be thought as the average fo the leftand right Riemann sums.

$$\int_a^b f(x) dx pprox \sum_{i=0}^{n-1} rac{f(x_i) + f(x_{i+1})}{2} \Delta x.$$

Or more efficiently as

$$\int_a^b f(x) dx pprox rac{\Delta x}{2} \Biggl(f(x_0) + 2 \left(\sum_{i=1}^{n-1} f(x_i)
ight) + f(x_n) \Biggr) \, .$$

```
In [11]: def Trapezoidal(f,a,b,n):
    deltax = (b-a)/n
    sum=0.0
    for i in range(1,n):
        sum = sum + f(a+i*deltax)
    sum = f(a)+2*sum+f(b)
    return sum*deltax/2
In [12]: ## Examples
f = lambda x: 3*x**2
T40 = Trapezoidal(f,0,2,40)
    print(T40)
```

8.0025000000000003

Simpson's 1/3-rule

Number of subintervals n of [a,b] must be even. Let n=2m, then

$$\int_a^b f(x)dx pprox rac{\Delta x}{3} \Bigg[f(x_0) + 4\left(\sum_{i=1}^m f(x_{2i-1})
ight) + 2\left(\sum_{i=1}^{m-1} f(x_{2i})
ight) + f(x_{2m}) \Bigg]\,.$$

If you are interested to find more about Simpson's method and its implimentation in Python, check this online book.

Question 1 Consider the sigmoid function given by $f(x) = \tanh(x)$. Create a tabular comparison of its exact derivative at x = 0.5 with the approximations by the three numerical differentiation formulas provided above for $h = 0.1, \ 0.01, \ 0.001, \ 0.0001, \ 0.00001$.

```
In [88]:
                                                   import math
                                                   from tabulate import tabulate
                                                   import pandas as pd
                                                   hList = [0.1, 0.01, 0.001, 0.0001, 0.00001]
                                                   dictDiff = {"forwardDifference":[],"backwardDifference":[],"centralDifference":[]}
                                                   def fdiff(f,a,h):
                                                                         derivative_forward_diff = (math.tanh(a+h)-math.tanh(a))/h
                                                                         return derivative forward diff
                                                   def bdiff(f,a,h):
                                                                         derivative_backward_diff = (math.tanh(a)-math.tanh(a-h))/h
                                                                         return derivative_backward_diff
                                                   def cdiff(f,a,h):
                                                                         derivative_central_diff = (math.tanh(a+h)-math.tanh(a-h))/(2*h)
                                                                         return derivative_central_diff
                                                   for h in hList:
                                                                          print(fdiff(f,0.5,h))
                                                                                 print(bdiff(f,0.5,h))
                                                                                  print(cdiff(f,0.5,h))
                                                                         dictDiff["forwardDifference"].append(fdiff(f,0.5,h))
                                                                         dictDiff["backwardDifference"].append(fdiff(f,0.5,h))
                                                                         dictDiff["centralDifference"].append(fdiff(f,0.5,h))
                                                   # print(dictDiff)
                                                   # df = pd.DataFrame(dictDiff)
                                                   # df.index = hList
                                                   # df.index.name = "h"
                                                   # print(df)
                                                   mydata = [
                                                                          ["0.1", dictDiff["forwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictDiff["backwardDiff["backwardDifference"][0],dictDiff["backwardDifference"][0],dictD
                                                                         ["0.01", dictDiff["forwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dictDiff["backwardDiff["backwardDifference"][1],dictDiff["backwardDifference"][1],dict
                                                                         ["0.001", dictDiff["forwardDifference"][2],dictDiff["backwardDifference"][2],d
                                                                         ["0.0001", dictDiff["forwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDiff["backwardDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDifference"][3],dictDiffer
                                                                         ["0.00001", dictDiff["forwardDifference"][4],dictDiff["backwardDifference"][4]
                                                   # create header
                                                   head = ["H Value", "Forward Difference", "Backward Difference", "Central Difference"
```

```
# display table
print(tabulate(mydata, headers=head, tablefmt="grid"))
 H Value | Forward Difference | Backward Difference | Central Difference
          0.749324
                            0.749324
  0.1
                 0.749324
          0.782804
  0.01
                     0.782804
                                0.782804
 ------
  0.001
           0.786084
                      0.786084
  0.0001
          0.786411
                     0.786411
 ------
          0.786444
                     0.786444
  1e-05
                                0.786444
 ------
```

Question 2 Understand the example of implementation of Trapezoidal rule. Modify this code to find the definite integral by Simpson's rule.

Evaluate the integral of $\int_{-1}^{1} \frac{1}{1+x^2}$ exactly and compare this with the approximations by

- 1. Mid-point Riemann sum.
- 2. Trapezoidal Rule
- 3. Simpson's Rule

Take n=20 above.

```
In [86]: #Evaluate the integral using scipy (Reference - https://scipy.org/)
from tabulate import tabulate
import math
from scipy import integrate
limit = [-1,1]

def indegrand(x):
    return 1/(1+x**2)

ans = integrate.quad(indegrand,limit[0],limit[1])
print("Integration of 1/(1+x**2) from -1 to 1 is ----->",ans[0])

#Evaluate using Mid-point Riemann sum approximation

def RiemannSum(f,a,b,n,shift):
    if shift < 0 or shift > 1:
        print("Please provide appropriate value for the shift from 0 to 1.0.")
        return
```

```
deltax = (b-a)/n
    sum=0.0
    a = a+shift*deltax
    for i in range(n):
        sum = sum + f(a+i*deltax)
    return sum*deltax
f = lambda x: 1/(1+x**2)
RS = RiemannSum(f,limit[0],limit[1],20, shift=0.5)
#Evaluate using Trapezoidal Rule approximation
def Trapezoidal(f,a,b,n):
    deltax = (b-a)/n
    sum=0.0
    for i in range(1,n):
        sum = sum + f(a+i*deltax)
    sum = f(a)+2*sum+f(b)
    return sum*deltax/2
## Examples
f = lambda x: 1/(1+x**2)
T20 = Trapezoidal(f,limit[0],limit[1],20)
#Evaluate using Simpson Rule approximation
def Simpson(f,a,b,n):
    deltax = (b-a)/n
    m = n/2
    sum1=0.0; sum2 = 0.0; sum = 0.0
    for i in range(1,int(m)):
        sum1 = sum1 + f(a+((2*i)-1)*deltax)
        sum2 = sum2 + f(a+(2*i)*deltax)
    sum1 = sum1 + f(a+((2*m)-1)*deltax)
    sum = f(a)+4*sum1+2*sum2+f(b)
    return sum*(deltax/3)
f = lambda x: 1/(1+x**2)
S20 = Simpson(f, limit[0], limit[1], 20)
# print(RS)
# print(T20)
# print(520)
mydata = [
    ["Riemann sum approximation", RS],
    ["Trapezoidal Rule approximation", T20],
    ["Simpson's 1/3-rule", S20]
# create header
head = ["Name", "Approximation"]
# display table
print(tabulate(mydata, headers=head, tablefmt="grid"))
```