

# PS2-Q1-Ying Sun

January 21, 2019

## 1 Problem Set 2 --Q1

### 1.1 Ying Sun

#### 1. Numerical differentiation exercises

##### 1.1 Problem 1

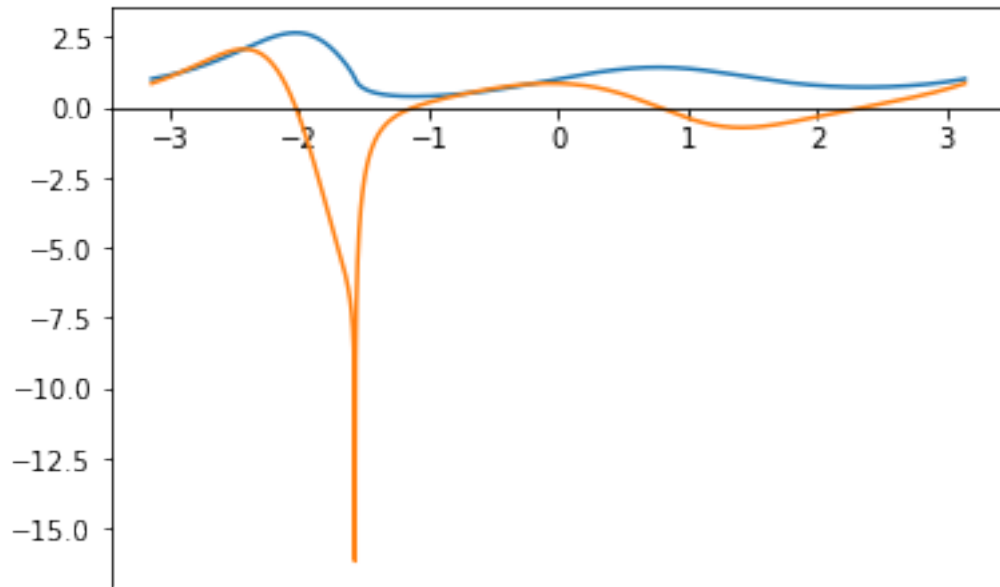
```
In [1]: from matplotlib import pyplot as plt
        from sympy import *
        import numpy as np
        from math import pi
```

```
# Takes the symbolic derivative with respect to x using SymPy
x = Symbol('x')
y = (sin(x) + 1) ** (sin(cos(x)))
diff(y, x)
```

```
Out[1]: (-log(sin(x) + 1)*sin(x)*cos(cos(x)) + sin(cos(x))*cos(x)/(sin(x) + 1))*(sin(x) + 1)**
```

```
In [2]: # Lambdify and plot the resulting function
x_vector = np.linspace(-pi, pi, 10000)
f = lambdify(x, y, 'numpy')
f_prime = lambdify(x, diff(y, x), 'numpy')
ax = plt.gca()
ax.spines["bottom"].set_position("zero")
plt.plot(x_vector, f(x_vector))
plt.plot(x_vector, f_prime(x_vector))
```

```
Out[2]: [<matplotlib.lines.Line2D at 0x11de41ba8>]
```



## 1.2 Problem 2

```
In [3]: def forward_1(x, h=0.001):
        return (f(x+h)-f(x))/h

        def forward_2(x, h=0.001):
            return (4*f(x+h)-f(x+2*h)-3*f(x))/(2*h)

        def backward_1(x, h=0.001):
            return (f(x)-f(x-h))/h

        def backward_2(x, h=0.001):
            return (-4*f(x-h)+f(x-2*h)+3*f(x))/(2*h)

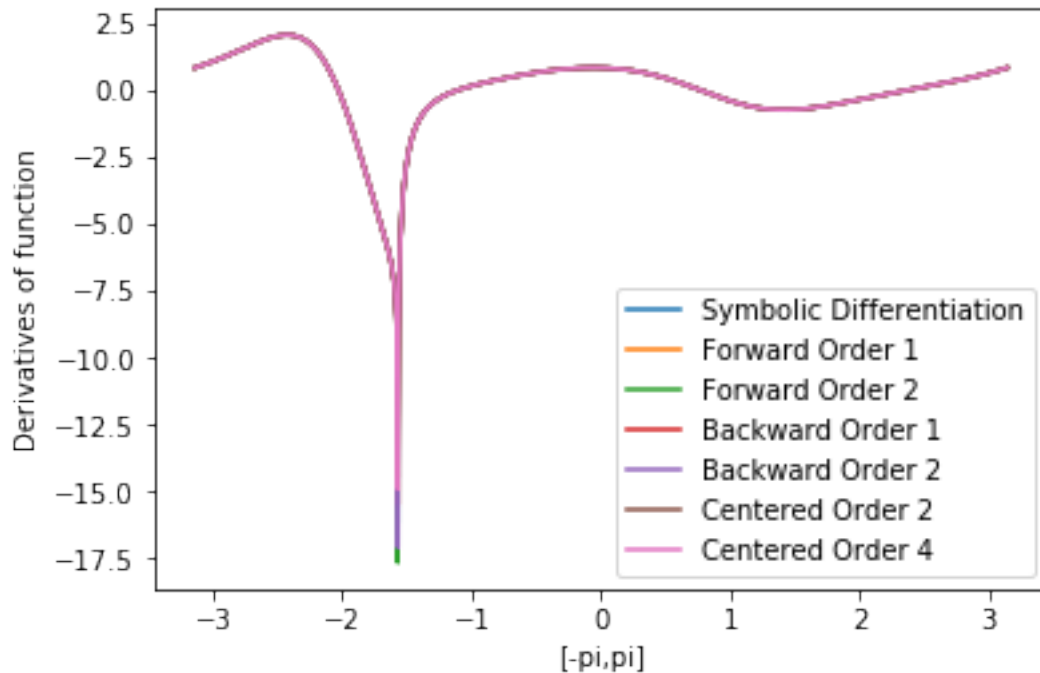
        def center_2(x, h=0.001):
            return (f(x+h)-f(x-h))/(2*h)

        def center_4(x, h=0.001):
            return (f(x-2*h)-8*f(x-h)+8*f(x+h)-f(x+2*h))/(12*h)

In [4]: plt.plot(x_vector, f_prime(x_vector), label = 'Symbolic Differentiation')
        plt.plot(x_vector, forward_1(x_vector), label = 'Forward Order 1')
        plt.plot(x_vector, forward_2(x_vector), label = 'Forward Order 2')
        plt.plot(x_vector, backward_1(x_vector), label = 'Backward Order 1')
        plt.plot(x_vector, backward_2(x_vector), label = 'Backward Order 2')
        plt.plot(x_vector, center_2(x_vector), label = 'Centered Order 2')
        plt.plot(x_vector, center_4(x_vector), label = 'Centered Order 4')
```

```
plt.legend(loc = 'lower right')
plt.xlabel("[-pi,pi]")
plt.ylabel("Derivatives of function")
```

Out[4]: Text(0, 0.5, 'Derivatives of function')



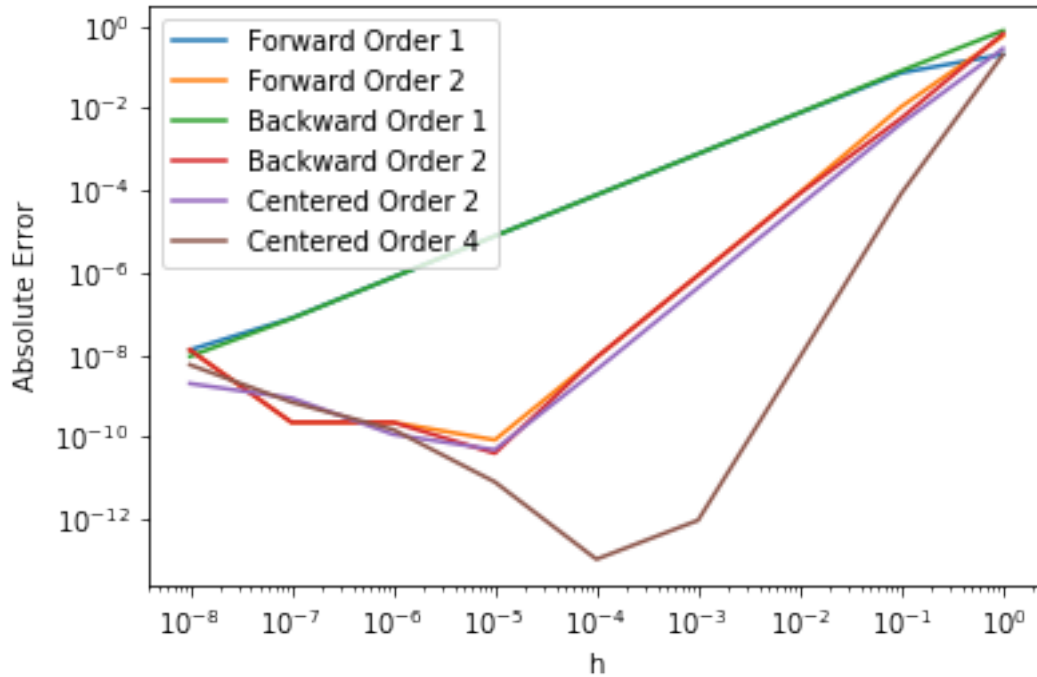
### 1.3 Problem 3

In [5]: *# Compute absolute error*

```
def Compare(x):
    x0 = f_prime(x)
    hvec = np.logspace(-8,0,9)
    plt.plot(hvec, abs([forward_1(x, h=i) for i in hvec]-x0), label = 'Forward Order 1')
    plt.plot(hvec, abs([forward_2(x, h=i) for i in hvec]-x0), label = 'Forward Order 2')
    plt.plot(hvec, abs([backward_1(x, h=i) for i in hvec]-x0), label = 'Backward Order 1')
    plt.plot(hvec, abs([backward_2(x, h=i) for i in hvec]-x0), label = 'Backward Order 2')
    plt.plot(hvec, abs([center_2(x, h=i) for i in hvec]-x0), label = 'Centered Order 2')
    plt.plot(hvec, abs([center_4(x, h=i) for i in hvec]-x0), label = 'Centered Order 4')
    plt.legend(loc='upper left')
    plt.xlabel("h")
    plt.ylabel("Absolute Error")
    plt.loglog()
```

In [6]: *# Graph at fixed point (x = 1)*

```
Compare(1)
```



#### 1.4 Problem 4

```
In [7]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")

df = np.load("plane.npy")
plane_df = pd.DataFrame(df, columns = ['t', 'alpha', 'beta'])
plane_df['alpha'] = np.deg2rad(plane_df['alpha'])
plane_df['beta'] = np.deg2rad(plane_df['beta'])
a = 500

# Calculate the location of the plane
plane_df['x'] = (a*np.tan(plane_df['beta']))/(np.tan(plane_df['beta'])-\
                                             np.tan(plane_df['alpha']))
plane_df['y'] = (a*np.tan(plane_df['beta'])*np.tan(plane_df['alpha']))\
/(np.tan(plane_df['beta'])-np.tan(plane_df['alpha']))

# Approximate x_prime and y_prime
plane_df['x_prime'] = 0
plane_df['y_prime'] = 0
plane_df['x_prime'][0] = plane_df['x'][1] - plane_df['x'][0]
plane_df['x_prime'][7] = plane_df['x'][7] - plane_df['x'][6]
plane_df['y_prime'][0] = plane_df['y'][1] - plane_df['y'][0]
plane_df['y_prime'][7] = plane_df['y'][7] - plane_df['y'][6]
```

```

for i in range(1, 7):
    plane_df['x_prime'][i] = 0.5 * (plane_df['x'][i+1] - plane_df['x'][i-1])
    plane_df['y_prime'][i] = 0.5 * (plane_df['y'][i+1] - plane_df['y'][i-1])
plane_df['speed'] = np.sqrt(plane_df['x_prime'] **2 + plane_df['y_prime']**2)
plane_df

```

```

Out[7]:
      t      alpha      beta      x      y  x_prime  y_prime  \
0  7.0  0.981748  1.178795  1311.271337  1962.456239      44      12
1  8.0  0.969181  1.161866  1355.936476  1975.114505      45      12
2  9.0  0.956440  1.144761  1401.918398  1987.346016      47      12
3 10.0  0.943525  1.127308  1450.497006  2000.840713      48      13
4 11.0  0.930959  1.110378  1498.640350  2013.512411      46      12
5 12.0  0.919614  1.095020  1543.798955  2025.792234      49      13
6 13.0  0.906524  1.077217  1598.041382  2040.990583      51      14
7 14.0  0.895005  1.061509  1647.596093  2055.065571      49      14

      speed
0  45.607017
1  46.572524
2  48.507731
3  49.729267
4  47.539457
5  50.695167
6  52.886671
7  50.960769

```

```

In [8]: plane_df[['t', 'speed']]

```

```

Out[8]:
      t      speed
0  7.0  45.607017
1  8.0  46.572524
2  9.0  48.507731
3 10.0  49.729267
4 11.0  47.539457
5 12.0  50.695167
6 13.0  52.886671
7 14.0  50.960769

```

## 1.5 Problem 5

```

In [9]: def Jacobian(f, x0, h):
        num_f = len(f)
        dim = len(x0)
        J = zeros(num_f, dim)
        for i, sf in enumerate(f):
            for j in range(dim):
                func= lambdify((x,y), sf, 'numpy')
                plus = x0 + h*(np.identity(dim))[:, j]
                minus = x0 - h*(np.identity(dim))[:, j]

```

```

        # Use the second order centered difference quotient
        J[i, j] = (func(plus[0], plus[1]) - func(minus[0], minus[1])) / (2 * h)
    return J

```

In [10]: *#Test the Jacobian calculation function*

```

x = Symbol('x')
y = Symbol('y')
x0 = (1, 1)
h = 0.001
Jacobian([x**2, x**3 - y], x0, h)

```

Out[10]: Matrix([  
 [1.999999999999984, 0.0],  
 [3.000000099999986, -0.999999999999945]])

## 1.6 Problem 7

```

In [11]: import time
import math
import sympy as sy
from autograd import numpy as anp
from autograd import grad
import numpy as np

def Time(N):
    t1 = np.zeros(N, dtype='float')
    t2 = np.zeros(N, dtype='float')
    t3 = np.zeros(N, dtype='float')
    abs_error1 = np.array([1e-18] * N)
    abs_error2 = np.zeros(N, dtype='float')
    abs_error3 = np.zeros(N, dtype='float')
    y = lambda x: (anp.sin(x) + 1) ** (anp.sin(anp.cos(x)))
    gy = grad(y)
    for i in range(N):
        x = np.random.uniform(-math.pi, math.pi)
        time1 = time.clock()
        z = sy.symbols('z')
        y_prime = sy.diff((sy.sin(z) + 1) ** sy.sin(sy.cos(z)), z)
        f_prime = sy.lambdify(z, y_prime, "numpy")
        prime = f_prime(x)
        time2 = time.clock()
        t1[i] = time2 - time1
        time3 = time.clock()
        approx_prime = center_4(x, h = 0.01)
        time4 = time.clock()
        t2[i] = time4 - time3
        abs_error2[i] = abs(approx_prime - prime)
        time5 = time.clock()
        auto_approx_prime = gy(x)

```

```

        time6 = time.clock()
        t3[i] = time6 - time5
        abs_error3[i] = abs(auto_approx_prime- prime)

    return t1, t2, t3, abs_error1, abs_error2, abs_error3

t1, t2, t3, abs_error1, abs_error2, abs_error3 = Time(200)

plt.scatter(t1, abs_error1, label='Sympy')
plt.scatter(t2, abs_error2, label='Difference Quotients')
plt.scatter(t3, abs_error3, label='Autograd')
plt.loglog()
plt.xlim(10**-5, 10**-2)
plt.ylim(10**-19, 10**-5)
plt.xlabel("Time")
plt.ylabel("absolute error")
plt.legend()

```

Out[11]: <matplotlib.legend.Legend at 0x120c0f320>

