```
In [ ]:
```

16/20

Assignment 3

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Imports

```
In [3]: #from scipy import stats
import numpy as np
import matplotlib.pyplot as plt
import urllib.request as url
from decimal import *
#from statistics import mean
%matplotlib inline
```

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Problem 1

Proofs

Proof of Three point Stencil

To prove :
$$y''(x)_2 = \frac{[y(x+h)-2y(x)+y(x-h)]}{(h*h)}$$

Taking left and right derivatives, i.e when h > 0 or when h < 0, we get -

Eqn-1:
$$f(x_0 + h) = f(x_0) + f'(x_0)h + \frac{f''(x_0)h^2}{2} + \frac{f^{(3)}(x_0)h^3}{6} + \frac{f^{(4)}(\xi_1)h^4}{24}$$

Eqn-2:
$$f(x_0 - h) = f(x_0) - f'(x_0)h + \frac{f''(x_0)h^2}{2} - \frac{f^{(3)}(x_0)h^3}{6} + \frac{f^{(4)}(\xi_2)h^4}{24}$$

Adding Eqn1 and Eqn2

$$f(x_0 + h) + f(x_0 - h) = 2f(x_0) + f''(x_0)h^2 + \frac{f^{(4)}(\xi)h^4}{12}$$

$$f''(x_0) = \frac{f(x_0 + h) + f(x_0 - h) - 2f(x_0)}{h^2} - \frac{f^{(4)}(\xi)h^2}{12} \text{ QED}$$

Proof of Five point Stencil

Eqn-1:
$$f(x_0 + h) = f(x_0) + f'(x_0)h + \frac{f''(x_0)h^2}{2} + \frac{f^{(3)}(x_0)h^3}{6} + \frac{f^{(4)}(x_0)h^4}{24} + \frac{f^{(5)}(x_0)h^5}{120} + \frac{f^{(6)}(\xi_1)h^6}{720}$$

Eqn-2:
$$f(x_0 - h) = f(x_0) - f'(x_0)h + \frac{f''(x_0)h^2}{2} - \frac{f^{(3)}(x_0)h^3}{6} + \frac{f^{(4)}(x_0)h^4}{24} - \frac{f^{(5)}(x_0)h^5}{120} + \frac{f^{(6)}(\xi_2)h^6}{720}$$

Egn-3:
$$f(x_0 + 2h) = f(x_0) + 2f'(x_0)h + \frac{4}{2!}f''(x_0)h^2 + \frac{8}{3!}f^{(3)}(x_0)h^3 + \frac{16f^{(4)}(x_0)h^4}{24} + \frac{32f^{(5)}(x_0)h^5}{120} + \frac{64f^{(6)}(x_0)h^4}{7}$$

Egn- 4:
$$f(x_0 - 2h) = f(x_0) - 2f'(x_0)h + \frac{4}{2!}f''(x_0)h^2 - \frac{8}{3!}f^{(3)}(x_0)h^3 + \frac{16f^{(4)}(x_0)h^4}{24} - \frac{32f^{(5)}(x_0)h^5}{120} + \frac{64f^{(6)}(x_0)h^4}{72}$$

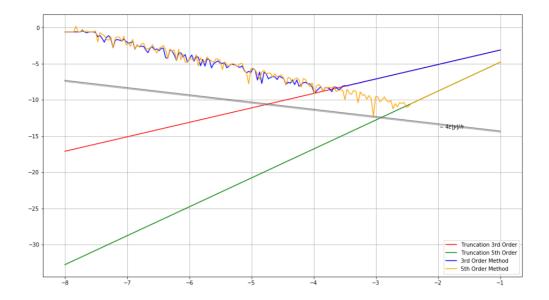
Adding Eqn 1 and Eqn 2 -

$$16f(x_0+h)+16f(x_0-h)=32f(x_0)+16f''(x_0)h^2+\frac{32f^{(4)}(x_0)h^4}{24}+\frac{32f^{(6)}(\xi_5)h^6}{720}$$

Adding Eqn 3 and Eqn 4 -

$$f(x_0 + 2h) + f(x_0 - 2h) = 2f(x_0) + 4f''(x_0)h^2 + \frac{32f^{(4)}(x_0)h^4}{24} + \frac{128f^{(6)}(\xi_6)h^6}{720}$$

```
In [23]: # Below are the mathematical functions for f(x), f'(x), df^{*}
         4/d^4x(f(x)), df^6/d^6x(f(x))
         def f(x):
              return np.sqrt(x)
         def f 2(x):
             return -1/4*x**(-3/2)
         def f 4(x):
              return -15/(16*x**(7/2))
         def f 6(x):
              return -945/(64*x**(11/2))
          # main program
          # differrent differentiation schemes
         logh = -np.linspace(8,1,200) # -8...0
         h = 10.**logh
         eps = 2.22e-16
         x = 1
         analyitical value = -1/4
         # numerical differentiation
         \# p = 2 \text{ diff. formula}
         d1f 2 = (f(x+h)+f(x-h)-2*f(x))/h**2
          \# p = 4 \text{ diff. formula}
         d1f 4 = (16*f(x+h)+16*f(x-h)-f(x-2*h)-f(x+2*h) -30*f(x))/(1
         2*h*h)
          residual 2 = np.absolute(d1f 2 - f 2(x))
          residual 4 = np.absolute(analyitical value - d1f 4 )
         trucation 2 = (abs(f 4(x))*h**2)/12
         trucation 4 = \text{np.absolute}((96*f 6(x)*h**4)/8640)
         plt.figure(figsize=(16,9)) # Setting Larger Graph Size
         plt.plot(logh,np.log10(trucation 2),"red")
         plt.plot(logh,np.log10(trucation_4),"green")
         plt.plot(logh,np.log10(residual_2), "blue")
         plt.plot(logh,np.log10(residual 4),"orange")
          for p in [2.4]:
             Not sure how this code segment inside for loop works, d
          irectly copied from the example text file
             E round = np.array((5*p)**0.5/2*eps*abs(f(x))/h)
             logE = np.log10(E round)
             c = 0.8 - p/8
             plt.plot(logh,logE,color=(c,c,c),alpha=0.7)
         plt.grid()
         plt.text(-2.,-13.9,'$\sim 4\epsilon |y|/h$')
         axes = plt.gca()
         axes.legend(["Truncation 3rd Order","Truncation 5th Orde
          r", "3rd Order Method", "5th Order Method"])
         plt.show()
```



Q1 5/6 Need to comment on the result/whether it is what you expect.

Problem 2

Planet	Semi-major Axis
Mercury	0.387098 AU
Venus	0.723332 AU
Earth	1.000001018 AU
Mars	1.523679 AU
Ceres	2.7691651545 AU
Jupiter	5.2044 AU
Saturn	9.5826 AU
Uranus	19.2184 AU
Neptune	30.11 AU

References

"Ceres (Dwarf Planet)". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Ceres_(dwarf_planet)).

"Earth's Orbit". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Earth%27s_orbit). (https://en.wikipedia.org/wiki/Earth%27s_orbit).

"Jupiter". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Jupiter (https://en.wiki/Jupiter (<a href="https://en.wik

"Mars". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Mars (https://en.wiki/Mars (<a href="https://en.wiki/Mar

"Mercury (Planet)". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Mercury_(planet)).

"Neptune". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Neptune (https://en.wiki/Neptune (<a href="https://en.wik

"Saturn". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Saturn (https://en.wiki/Saturn (https://en.wiki/Saturn (https://en.wiki/Saturn (https://en.wiki/Saturn (https://en.wiki/Saturn (<a href="https:/

"Uranus". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Uranus (https://en.wiki/Uranus (https://en.wiki/Uranus (https://en.wiki/Uranus (https://en.wiki/Uranus (https://en.wiki/Uranus (https://en.wiki/Uranus (https://en.

"Venus". En.Wikipedia.Org, 2019, https://en.wikipedia.org/wiki/Venus (https://en.wiki/Venus (ht

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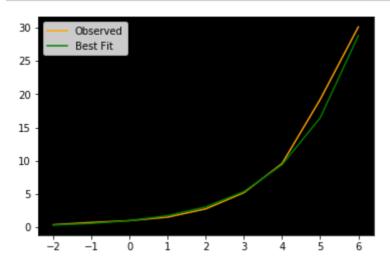
```
obs axis = [0.387098, 0.723332, 1.000001018, 1.523679, 2.769165]
In [93]:
          1545,5.2044,9.5826,19.2184,30.11]
          d obs = np.array(obs axis)
          def d unvec(x):
               Calculate all 9 distance values for given X
               n = np.array([-2, -1, 0, 1, 2, 3, 4, 5, 6])
               return x**n
          d = np.vectorize(d unvec)
          def e(distance):
               Calculate square of the arithmetic average of relative
           deivations
               return np.sum( ((distance - d obs)/d obs)**2 )/9
          e vec = np.vectorize(e)
          \overline{\text{def}} \overline{\text{dE}} \overline{\text{dx}}(x):
               11 II II
               First Derivative of the function e(d(x)) using first pr
           inciples, for arbitarily small number
               1e-10 as h, the smaller this number the more accurate t
           he derivative is
               return (e(d(x+1e-10)) - e(d(x)))/1e-10
```

```
In [96]:
           a n,b n = 1.5,2.0
           m = (a n + b n)/2
           f m = \overline{d}E dx(\overline{m})
           \overline{\text{list m}}, \overline{\text{list f m}} = [],[]
           n = \overline{0}
           while not(np.isclose([f m],[0])):
                 Bisection Method Implementation, but instead of restric
            ting to N iterations, this method will continue until
                 the midoint isclose to 0
                m = (a n + b n)/2
                f m = \overline{dE} dx(\overline{m})
                f_a = dE_dx(a_n)
                f b = dE dx(b n)
                list m.append(m)
                list_f_m.append(f_m)
                if f^{-}m^*f a < 0:
                     a n, b n = a n, m
                elif \overline{f} m*\overline{f} b < \overline{0}:
                     a_n, b_n = m, b n
                n += 1
           print("X Value for Minimun E((d(x))) - ",m,"Value of E(d(x)))
            (x)) at the X Value - ",f m)
           print("Number of Iterations -", n)
           X Value for Minimun E((d(x))) - 1.7509384118020535 Value o
```

f E(d(x)) at the X Value - 0.0

Number of Iterations - 27

```
In [99]: n = np.array([-2,-1,0,1,2,3,4,5,6])
    plt.plot(n,d_obs,'orange')
    plt.plot(n,d(m),'green')
    axes = plt.gca()
    axes.set_facecolor('xkcd:black') # Setting Background colou
    r as black
    axes.legend(['Observed','Best Fit'])
    plt.show()
    residuals = np.array(d(m) - d_obs)
    for i in range(9):
        print("Planet Number(N) : ",n[i],"Residual : ",residual
    s[i])
```



```
Planet Number(N) :
                   -2 Residual : -0.06091730076811058
                   -1 Residual :
Planet Number(N) :
                                 -0.15220968452643224
Planet Number(N) :
                   O Residual :
                                 -1.0180000000747924e-06
Planet Number(N) :
                   1 Residual :
                                 0.22725941180205345
Planet Number(N) :
                   2 Residual :
                                 0.2966201674238973
Planet Number(N) :
                   3 Residual :
                                 0.16360128249547667
Planet Number(N) : 4 Residual :
                                 -0.18356035987598496
Planet Number(N) : 5 Residual :
                                 -2.761260460056711
Planet Number(N) :
                   6 Residual :
                                 -1.2945622311269247
```

Q2 3.5/4 Requested a semilog plot to show power law.

Problem 3

Proofs:

Proving the term in iterations is

$$x_{n+1} = x_n \times (2 - c \times x_n)$$

General form for Newton's Method:

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}, \text{ where } f(x) \text{ and } f'(x) \text{ are as follows}$$

$$f(x): 1/x - c, f'(x): -1/x^2$$

$$\Rightarrow x_{n+1} = x_n - \frac{(1/x_n - c)}{(-1/x_n^2)}$$

$$\Rightarrow x_{n+1} = \frac{\frac{-1}{x_n} - \frac{1}{x_n} + c}{\frac{-1}{x_n^2}}$$

$$\Rightarrow x_{n+1} = \frac{2 - c \times x_n}{x_n} \times x_n^2$$

$$\Rightarrow x_{n+1} = x_n \times (2 - c \times x_n) \text{ QED}$$

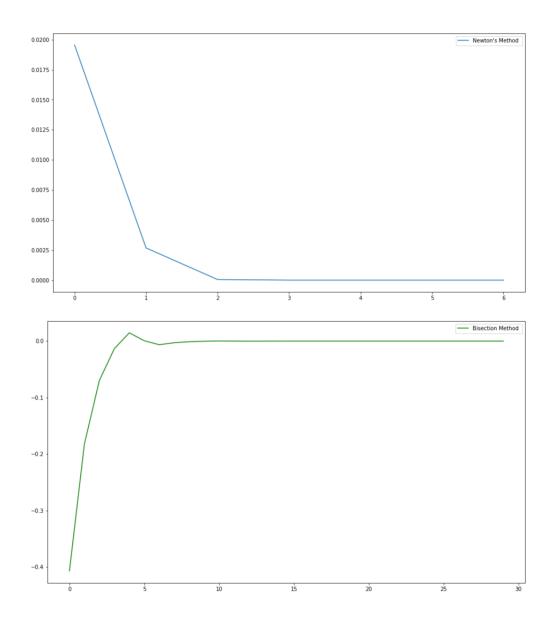
Solving for C = 7, $x_1 = 0.09 < 1/7$

Need to prove the inequality in general.

```
In [131]:
              n = 0
              lastVal, currentVal = 0, 0.09
              approximation = []
              while not lastVal == currentVal:
                  Newton's Method Implementation, using while loop to sto
              p when the previous value is the same as the current value
                  instead of using vectorized function for a given n valu
                   11 11 11
                  lastVal = currentVal
                  currentVal = newton(f,df,currentVal)
                  approximation.append(currentVal)
                  n += 1
              print("Newton's Method : Value, Iterations-",approximation
              [-1],n)
              iterations = np.arange(n)
              count = 0
              a.b = 0.1.1
              m = (a+b)/2
              f m = f(m)
              fa = f(a)
              f b = f(b)
              m values = []
              while not np.isclose([f_m],[0]):
                  Bisection Algorithm, exactly same as the one used in Pr
              oblem 2
                  m = (a+b)/2
                  if count > 52:
                      break
                  f m = f(m)
                  fa = f(a)
                  f^{-}b = f(b)
                  m values.append(m)
                  if f m*f a < 0:
                      a,b = a,m
                  elif f m*f b < 0:
                      a,b = m,b
                  count += 1
              print("Bisection Method : Value, Iterations-", m, count,'\n
              ')
              bisection = np.array(m values)
              approximation = np.array(approximation)
              residuals = 1/7 - approximation
              residual bisection = 1/7 - bisection
              counts = np.arange(count)
              print('Residual Plots','\n')
              plt.figure(figsize=(16,9)) # Setting Larger Graph Size
              plt.plot(iterations, residuals)
              axes = plt.qca()
              axes.legend(["Newton's Method "])
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```

Newton's Method : Value, Iterations- 0.14285714285714285 7 Bisection Method : Value, Iterations- 0.14285714281722903 3 $_{\odot}$

Residual Plots

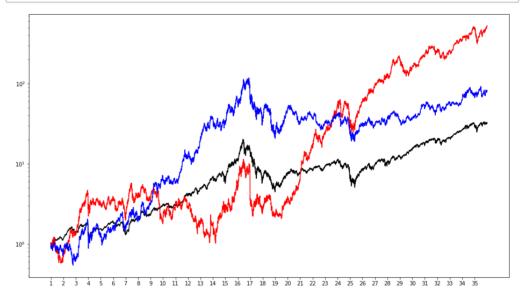


Q3 3/4 Proofs missing last part + comment on result.

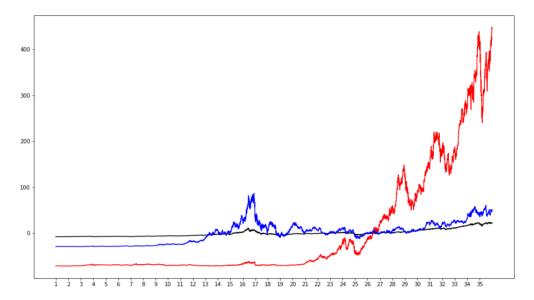
Problem 4

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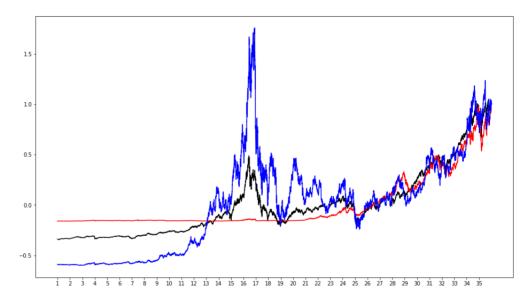
```
In [61]: url.urlretrieve("http://planets.utsc.utoronto.ca/~pawel/pyt
h/IXIC.dat", "IXIC.dat")
url.urlretrieve("http://planets.utsc.utoronto.ca/~pawel/pyt
h/AAPL.dat", "AAPL.dat")
url.urlretrieve("http://planets.utsc.utoronto.ca/~pawel/pyt
h/INTC.dat", "INTC.dat")
pass
#Retriving Files
```



```
In [5]: # Detrending the inflation trend line of 3%
        # Line in the form v = mx + c
        def detrend(data):
            Returns detrended list, substracting the inflation rate
            m = 0.03/252
            # This is the inflation rate per market day, as 252 mar
        ket days a year and 0.03 inflation a year
            \# c = mean(y) - m*mean(x)
            c = np.mean(data) - m*np.mean(t)
            line = m*t + c
            return data-line
        def normalize(data):
            Returns normalized list, where the data ends at value 1
            lastVal = data[-1]
            return data/lastVal
        # Creating Detrended Time Series
        ixic detrended = detrend(ixic)
        aapl detrended = detrend(aapl)
        intc detrended = detrend(intc)
        plt.figure(figsize=(16,9)) # Setting Larger Graph Size
        plt.plot(t,ixic detrended,color=(0,0,0))
        plt.plot(t,aapl detrended,color=(1,0,0))
        plt.plot(t,intc detrended,color=(0,0,1))
        plt.xticks(range(1,len(ixic),252),range(1,36))
        plt.show()
        print("The above graph is the detrended data graph and belo
        w graph is normalized data graph ")
        #Creating Renormalized Time Series
        ixic normal = normalize(ixic detrended)
        aapl normal = normalize(aapl detrended)
        intc normal = normalize(intc detrended)
        plt.figure(figsize=(16,9)) # Setting Larger Graph Size
        plt.plot(t,ixic normal,color=(0,0,0))
        plt.plot(t,aapl normal,color=(1,0,0))
        plt.plot(t,intc normal,color=(0,0,1))
        plt.xticks(range(1,len(ixic),252),range(1,36))
        plt.show()
```



The above graph is the detrended data graph and below graph is normalized data graph

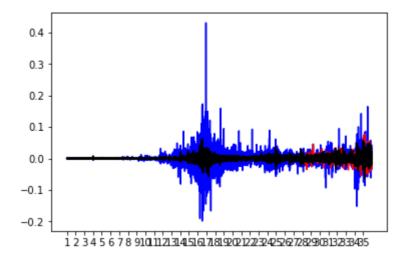


This is not what we meant by detrending. Just subtract a line with slope 3% per year.

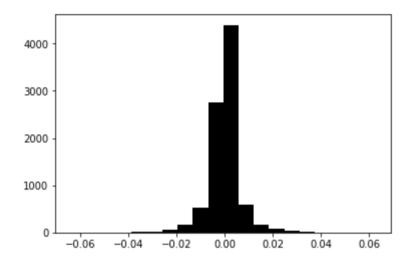
Looking at the normalized data graph, it is clear that the general growth of the market or the IXIC data growth is mostly due to historic inflation graph. While the blue graph, intel stock price, INTC is very slightly above the general historic inflation growth, however about 20 years ago, the INTC stock was growing significantly faster than the inflation rate. Finally, the apple stock or AAPL, was growing negatively relative to the inflation rate for about the first 20 years of this data but in recent years has been growing significantly faster than inflation rate.

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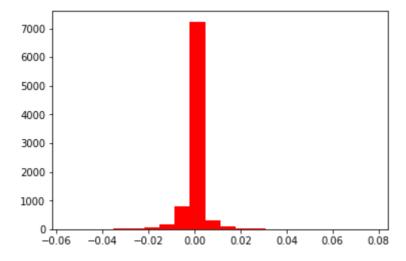
```
In [6]:
        def derrivativeP(data):
            Returns rate of change of the time series by comparing
        the previous and next value
            firstValue = data[:-1]
            nextValue = data[1:]
            #print(len(firstValue),len(nextValue))
            return firstValue - nextValue
        #Creating Rate of change
        dIxic = derrivativeP(ixic normal)
        dAapl = derrivativeP(aapl normal)
        dIntc = derrivativeP(intc normal)
        # Plotting rate of change
        plt.plot(t[:-1],dIntc,color=(0,0,1))
        plt.plot(t[:-1],dAapl,color=(1,0,0))
        plt.plot(t[:-1],dIxic,color=(0,0,0))
        plt.xticks(range(1,len(ixic),252),range(1,36))
        plt.show()
        #Plot rate of change as histogram
        plt.hist(dIxic,bins=20,color=(0,0,0))
        print("IXIC- Mean, Std",dIxic.mean(),dIxic.std())
        plt.show()
        plt.hist(dAapl,bins=20,color=(1,0,0))
        print("AAPL- Mean, Std",dAapl.mean(),dAapl.std())
        plt.show()
        plt.hist(dIntc,bins=20,color=(0,0,1))
        print("INTC- Mean, Std",dIntc.mean(),dIntc.std())
        plt.show()
```



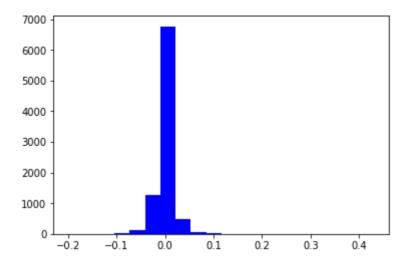
IXIC- Mean, Std -0.00015185066751938043 0.00681536123856707 3



AAPL- Mean, Std -0.00013140212236247294 0.00503116722039568



INTC- Mean, Std -0.00018019146294228 0.01947618869723907



Proof:

Given: $Y = c^*X + b$

To prove: r = +-1, r = sign(c)

Symbols µ for mean

$$Cov(X, Y) = E(X - \mu_X)(cX - c\mu_X) \implies c \times E(X - \mu_X)^2$$

$$\implies \sigma^2 \times c$$

$$Variance(Y) \implies Var(cX + b) \implies Var(cX) \implies \sigma_X^2 \times c^2$$

$$r = \frac{Cov(X, Y)}{\sigma_X \times \sigma_y} \implies \frac{\sigma^2 \times c}{\sigma_X^2 \times |c|}$$

$$\implies r = \frac{c}{|c|} \text{ QED}$$

```
In [221:
                         def covariance(x,y):
                                     11 11 11
                                     Returns covariance of X and Y
                                    return (x- np.mean(x))*(y- np.mean(y))
                         def r(x,y,cov):
                                    Returns coeffient of covarience for each year between X
                          and Y
                                     11 11 11
                                    relations = []
                                    for i in range(0,35):
                                               # Ap
                                               relations.append(np.sum(cov[i*252:(i+1)*252])/(np.s
                         um((x[i*252:(i+1)*252] - np.mean(x))**2)*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25])*np.sum((y[i*25
                         2:(i+1)*252] - np.mean(y))**2))**(1/2))
                                     return relations
                         def r full(x,y,cov):
                                    Returns coeffient of covarience for entire data set of
                                     return np.mean(cov)/(np.std(x)*np.std(y))
                          r NA = r(ixic normal, aapl normal, covariance(ixic normal, aap
                          l normal)) # NASDAQ-Apple
                          r NI = r(ixic normal,intc normal,covariance(ixic normal,int
                          c normal)) # NASDAQ-INTEL
                          r IA = r(intc normal, aapl normal, covariance(intc normal, aap
                          l normal)) # INTEL-Apple
                          print("Year\t","NA\t\t","NI\t\t","IA\n")
                          for i in range(0,35):
                                     print(i+1,r_NA[i],r_NI[i],r_IA[i])
                          r NA full = r full(ixic normal, aapl normal, covariance(ixic
                          normal, aapl normal)) # NASDAQ-Apple
                          r NI full = r full(ixic normal,intc normal,covariance(ixic
                          normal,intc normal)) # NASDAQ-INTEL
                          r IA full = r full(intc normal, aapl normal, covariance(intc
                          normal,aapl normal)) # INTEL-Apple
                          print("\nAverage",r NA full,r NI full,r IA full)
```

Year

NA

ΙA

1 0.9999351771413449 0.9999465993019827 0.9999981126661226 2 0.999905082123299 0.999832324625482 0.9999743862719062 3 0.9999640653187579 0.9999697748261174 0.9999966381482311 4 0.9999333553223355 0.9999544424140836 0.9999895403868692 5 0.9998767189501863 0.9999145322612504 0.9999915361546918 6 0.999873166105377 0.999852491199096 0.9999591724440073 7 0.9995380892317527 0.9995960505042796 0.9999623630729927 8 0.9999094019725308 0.9999164853115835 0.9998848374936237 9 0.9993745516999291 0.999806148419211 0.9988489016369941 10 0.9998407338287316 0.9999046862522329 0.9999024457134069 11 0.9972993456154381 0.9979161400380376 0.9921729590626722 12 0.9978673821728263 0.9949734796339822 0.991194867370334 13 0.9867094989328324 0.33886599981572085 0.199325364772887
6 14 0.9755458845785038 -0.19774440496976772 -0.2567191774776
1186 15 -0.3037161306581286 0.5920037079923467 -0.90686082929086
83 16 -0.9467384144718061 0.9571242322447178 -0.94440016824970
29 17 -0.018217752344679903 0.4233090812473575 -0.900015258976
9097 18 0.9286455694058353 -0.3704651369611163 -0.68198904650309
51 19 0.9737346179467342 0.039671345184796425 -0.1829107125044
2079 20 0.9747266179213485 -0.8059953678745773 -0.89399664362782
48
21 0.9625098903195777 -0.848420420124127 -0.930942683250846 2
22 0.8060843740663133 -0.15195473403650306 -0.3800132929554 6543
23 -0.7525218929615805 0.9442229310774368 -0.67490513092973
95 24 -0.006608566709371569 0.8067718185106827 -0.517293455083
3561 25 0.9953318664277514 0.9718861722522337 0.9622325903156156
26 0.297160900116172 0.11139065321876443 0.1611310327196377
2 27 0.8907166668965475 0.8854630585480904 0.8456105374808474
28 0.9914566931019279 0.9379864568239813 0.9324813342672365
29 0.9127242448045143 0.9331158570248247 0.795646449864182
30 0.9925445130900916 0.9344619097169077 0.9621439858279555
31 0.9986738038501282 0.970155689879934 0.97093377546489 32 0.9964136613554837 0.9951914948050716 0.9912287153921686
33 0.9982476651741266 0.9923433354528138 0.986104619113764
34 0.9974302750081178 0.9920422881954374 0.9851166905770773
35 0.996329246941892 0.9937047615236932 0.9877805484957529

NI

Average 0.9177502025836916 0.8578807347233194 0.61818290509 85176

Q4 4.5/6 Comment on the last part results.

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