

RISK_Mng_HW5_Duanhong Gao(dg2896)

October 13, 2016

```
In [17]: ##### HW5 Problem 2 differnet lambda values #####
import numpy as np
import scipy.stats as ss
import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt
import plotly.graph_objs as go
import plotly.plotly as py

path = '~/Documents/Semester3/M5320/homework/HW4/AMD-yahoo.csv'
sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
#data1 = sto_AMD.values[1:252*20,:]
data1 = sto_AMD.values
#data1 = data1[:, :-1]
AMD_close = list(data1[:,6])

## exponential weighting paramater lambda
lambda1 = 0.998614
lambda2 = 0.998723
lambda3 = 0.999362
lambda4 = 0.94
lambda5 = 0.97

### List the lambda values

def list_lambdas(lambda_k):
    list_lambda = []
    for i in range(len(data1)):
        lambda_value = (lambda_k**i)
        list_lambda.append(lambda_value)
    return(list_lambda)

list_lambda1 = list_lambdas(lambda1)
list_lambda2 = list_lambdas(lambda2)
list_lambda3 = list_lambdas(lambda3)
list_lambda4 = list_lambdas(lambda4)
list_lambda5 = list_lambdas(lambda5)

def weigthted_vol_and_mu(list_lambda, years):
    wgt_A_vol_lambda = []
    wgt_A_mu_lambda = []
```

```

wgt_log_rtn = []
wgt_log_rtn_sq = []

for i in range( len(data1)-1 ):
    log_return = math.log( AMD_close[i]/AMD_close[i+1] )
    wgt_log_return = log_return * list_lambda[i]
    wgt_log_rtn.append(wgt_log_return)

    log_return_sq = log_return ** 2
    wgt_log_rtn_sq.append( log_return_sq * list_lambda[i] )

for j in range(len(data1)-252*years):
    wgt_mu_lambda = sum(wgt_log_rtn[j:j+252*years])/sum(list_lambda[j:j+252*years])
    wgt_vol_lambda = sum(wgt_log_rtn_sq[j:j+252*years])/sum(list_lambda[j:j+252*years]) - wgt_mu_lambda**2
    wgt_A_vol_lambda.append(wgt_vol_lambda)
    wgt_A_mu_lambda.append(wgt_mu_lambda)

return(wgt_A_vol_lambda, wgt_A_mu_lambda )

years = 5
wgt_A_vol_lambda1 = weighed_vol_and_mu(list_lambda1, years)[0][1: 252*20]
wgt_A_mu_lambda1 = weighed_vol_and_mu(list_lambda1, years)[1][1: 252*20]

wgt_A_vol_lambda2 = weighed_vol_and_mu(list_lambda2, years)[0][1: 252*20]
wgt_A_mu_lambda2 = weighed_vol_and_mu(list_lambda2, years)[1][1: 252*20]

wgt_A_vol_lambda3 = weighed_vol_and_mu(list_lambda3, years)[0][1: 252*20]
wgt_A_mu_lambda3 = weighed_vol_and_mu(list_lambda3, years)[1][1: 252*20]

wgt_A_vol_lambda4 = weighed_vol_and_mu(list_lambda4, years)[0][1: 252*20]
wgt_A_mu_lambda4 = weighed_vol_and_mu(list_lambda4, years)[1][1: 252*20]

wgt_A_vol_lambda5 = weighed_vol_and_mu(list_lambda5, years)[0][1: 252*20]
wgt_A_mu_lambda5 = weighed_vol_and_mu(list_lambda5, years)[1][1: 252*20]

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d,'%Y-%m-%d').date() for d in timeline]

fig, ax = plt.subplots()
ax.plot(timeline, wgt_A_vol_lambda1, 'r-', label='lambda1 = 0.998614')
ax.plot(timeline, wgt_A_vol_lambda2, 'g-', label='lambda2 = 0.998723')
ax.plot(timeline, wgt_A_vol_lambda3, 'y-', label='lambda3 = 0.999362')
ax.plot(timeline, wgt_A_vol_lambda4, 'b-', label='lambda4 = 0.94')
ax.plot(timeline, wgt_A_vol_lambda5, 'm-', label='lambda4 = 0.97')

legend = ax.legend(loc='upper right', shadow=True)
ax.set_title('AMD vols')

```

```

fig, ax = plt.subplots()
ax.plot(timeline, wgt_A_mu_lambda1, 'r-', label='lambda1 = 0.998614')
ax.plot(timeline, wgt_A_mu_lambda2, 'g-', label='lambda2 = 0.998723')
ax.plot(timeline, wgt_A_mu_lambda3, 'y-', label='lambda3 = 0.999362')
ax.plot(timeline, wgt_A_mu_lambda3, 'b-', label='lambda4 = 0.94')
ax.plot(timeline, wgt_A_mu_lambda3, 'm-', label='lambda4 = 0.97')

legend = ax.legend(loc='upper right', shadow=True)
ax.set_title('AMD mu')
plt.show()

```

In [1]: ##### HW5 Problem 3 ##### PART 1: unweighted mu and vols for VaR and ES #####

```

import numpy as np
import scipy.stats as ss
import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt

```

```

path = '~/Documents/Semester3/M5320/homework/HW4/INTC-yahoo.csv'
sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
#data1 = sto_AMD.values[1:252*20,:]
data1 = sto_AMD.values
#data1 = data1[:-1]
AMD_close = list(data1[:,6])

```

```

A_log_rtn = []
A_log_rtn_sq = []

```

```

for i in range(1, len(data1)-1):
    log_return = math.log( AMD_close[i]/AMD_close[i+1] )
    A_log_rtn.append(log_return)
    A_log_rtn_sq.append(log_return**2)

```

```

def vol_and_mu(years):
    S0 = 10000
    T = 5/252
    p = 0.99

```

```

    A_vol_years = []
    A_mu_years = []
    A_VaR_years = []
    A_ES_years = []

```

```

    for i in range(len(data1)-252*years):
        vol_years = np.std(A_log_rtn[i:i+252*years]) * np.sqrt(252)
        mu_years = np.mean(A_log_rtn[i:i+252*years])*252 + (vol_years**2)/2
        VaR_years = S0 - S0 * np.exp( vol_years * T**(0.5)* ss.norm.ppf(1-p) + (mu_years - pow(
        A_vol_years.append(vol_years)
        A_mu_years.append(mu_years)

```

```

A_VaR_years.append(VaR_years)

if i<6:
    ES_years = VaR_years
else:
    sum_loss = []
    sum_loss_value = 0
    for k in range(4):
        sum_loss_value = sum_loss_value + A_VaR_years[i-k]
        sum_loss.append(sum_loss_value/(k+1))
    s = pd.Series(sum_loss)
    ES_years = s.quantile(.975)
A_ES_years.append(ES_years)

return(A_vol_years, A_mu_years, A_VaR_years, A_ES_years)

A_VaR_2years = vol_and_mu(2)[2][1:252*20]
A_ES_2years = vol_and_mu(2)[3][1:252*20]

A_VaR_5years = vol_and_mu(5)[2][1:252*20]
A_ES_5years = vol_and_mu(5)[3][1:252*20]

A_VaR_10years = vol_and_mu(10)[2][1:252*20]
A_ES_10years = vol_and_mu(10)[3][1:252*20]

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d,'%Y-%m-%d').date() for d in timeline]

fig, ax = plt.subplots()
ax.plot(timeline, A_VaR_2years, 'y-', label='VaR 2 year')
ax.plot(timeline, A_VaR_5years, 'g-', label='VaR 5 year')
ax.plot(timeline, A_VaR_10years, 'm-', label='VaR 10 year')
ax.plot(timeline, A_ES_2years, 'r-', label='ES 2 year')
ax.plot(timeline, A_ES_5years, 'c-', label='ES 5 year')
ax.plot(timeline, A_ES_10years, 'b-', label='ES 10 year')

legend = ax.legend(loc='upper right', shadow=True)
ax.set_title('INTC VaRs and ESs Windowed')
plt.show()

```

In [3]: ##### HW5 Problem3 ##### PART 2: weighted VaRs and ES

```

import numpy as np
import scipy.stats as ss
import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt

```

```

path = '~/Documents/Semester3/M5320/homework/HW4/INTC-yahoo.csv'
sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
data1 = sto_AMD.values
AMD_close = list(data1[:,6])

## exponential weighting paramater lambda
lambda1 = 0.998614

### List the lambda values

def list_lambdas(lambda_k):
    list_lambda = []
    for i in range(len(data1)):
        lambda_value = (lambda_k**i)
        list_lambda.append(lambda_value)
    return(list_lambda)

list_lambda1 = list_lambdas(lambda1)

def weigthed_VaR_and_ES(list_lambda, years):
    wgt_A_vol_years = []
    wgt_A_mu_years = []

    wgt_log_rtn = []
    wgt_log_rtn_sq = []

    for i in range( len(data1)-1 ):
        log_return = math.log( AMD_close[i]/AMD_close[i+1] )
        wgt_log_return = log_return * list_lambda[i]
        wgt_log_rtn.append(wgt_log_return)

        log_return_sq = log_return ** 2
        wgt_log_rtn_sq.append( log_return_sq * list_lambda[i] )

    S0 = 10000
    T = 5/252
    p = .99

    wgt_A_VaR_years = []
    wgt_A_ES_years = []

    for j in range(len(data1)-252*years):
        wgt_mu_lambda = sum(wgt_log_rtn[j:j+252*years])/sum(list_lambda[j:j+252*years])
        wgt_vol_lambda = sum(wgt_log_rtn_sq[j:j+252*years])/sum(list_lambda[j:j+252*years])- wgt_mu_lambda**2
        wgt_VaR_years = S0 - S0 * np.exp( wgt_vol_lambda * T**(0.5)* ss.norm.ppf(1-p) + (wgt_mu_lambda*T) )
        wgt_A_vol_years.append(wgt_vol_lambda)
        wgt_A_mu_years.append(wgt_mu_lambda)

```

```

wgt_A_VaR_years.append(wgt_VaR_years)

if j<6:
    wgt_ES_years = wgt_VaR_years
else:
    wgt_sum_loss = []
    wgt_sum_loss_value = 0
    for k in range(4):
        wgt_sum_loss_value = wgt_sum_loss_value + wgt_A_VaR_years[j-k]
        wgt_sum_loss.append(wgt_sum_loss_value/(k+1))
    s = pd.Series(wgt_sum_loss)
    wgt_ES_years = s.quantile(.975)
wgt_A_ES_years.append(wgt_ES_years)

return(wgt_A_VaR_years, wgt_A_ES_years)

years_2 = 2
wgt_A_VaR_2years = weighed_VaR_and_ES(list_lambda1, years_2)[0][1: 252*20]
wgt_A_ES_2years = weighed_VaR_and_ES(list_lambda1, years_2)[1][1: 252*20]

years_5 = 5
wgt_A_VaR_5years = weighed_VaR_and_ES(list_lambda1, years_5)[0][1: 252*20]
wgt_A_ES_5years = weighed_VaR_and_ES(list_lambda1, years_5)[1][1: 252*20]

years_10 = 10
wgt_A_VaR_10years = weighed_VaR_and_ES(list_lambda1, years_10)[0][1: 252*20]
wgt_A_ES_10years = weighed_VaR_and_ES(list_lambda1, years_10)[1][1: 252*20]

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d, '%Y-%m-%d').date() for d in timeline]

fig, ax = plt.subplots()
ax.plot(timeline, wgt_A_VaR_2years, 'r-', label='VaR 2 year')
ax.plot(timeline, wgt_A_ES_2years, 'y-', label='ES 2 year')

ax.plot(timeline, wgt_A_VaR_5years, 'b-', label='VaR 5 year')
ax.plot(timeline, wgt_A_ES_5years, 'g-', label='ES 5 year')

ax.plot(timeline, wgt_A_VaR_10years, 'm-', label='VaR 10 year')
ax.plot(timeline, wgt_A_ES_10years, 'c-', label='ES 10 year')

legend = ax.legend(loc='upper right', shadow=True)
ax.set_title('INTC VaRs and ESs Exponential Windowed Equivalent')

plt.show()

```

In [5]: ##### HW5 Problem3 ##### PART 3: VaRs and ES exponential lamda =0.94 /0.97

```

import numpy as np
import scipy.stats as ss

```

```

import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt

path = '~/Documents/Semester3/M5320/homework/HW4/INTC-yahoo.csv'
sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
data1 = sto_AMD.values
AMD_close = list(data1[:,6])

## exponential weighting paramater lambda
lambda1 = 0.94
lambda2 = 0.97

### List the lambda values

def list_lambdas(lambda_k):
    list_lambda = []
    for i in range(len(data1)):
        lambda_value = (lambda_k**i)
        list_lambda.append(lambda_value)
    return(list_lambda)

list_lambda1 = list_lambdas(lambda1)
list_lambda2 = list_lambdas(lambda2)

def weigthed_VaR_and_ES(list_lambda, years):
    wgt_A_vol_years = []
    wgt_A_mu_years = []

    wgt_log_rtn = []
    wgt_log_rtn_sq = []

    for i in range( len(data1)-1 ):
        log_return = math.log( AMD_close[i]/AMD_close[i+1] )
        wgt_log_return = log_return * list_lambda[i]
        wgt_log_rtn.append(wgt_log_return)

        log_return_sq = log_return ** 2
        wgt_log_rtn_sq.append( log_return_sq * list_lambda[i] )

    S0 = 10000
    T = 5/252
    p = .99

    wgt_A_VaR_years = []
    wgt_A_ES_years = []

    for j in range(len(data1)-252*years):
        wgt_mu_lambda = sum(wgt_log_rtn[j:j+252*years])/sum(list_lambda[j:j+252*years])
        wgt_vol_lambda = sum(wgt_log_rtn_sq[j:j+252*years])/sum(list_lambda[j:j+252*years])- wgt_mu_lambda**2
        wgt_VaR_years = S0 - S0 * np.exp( wgt_vol_lambda * T**(0.5)* ss.norm.ppf(1-p) + (wgt_mu_lambda * T) )

```

```

wgt_A_vol_years.append(wgt_vol_lambda)
wgt_A_mu_years.append(wgt_mu_lambda)
wgt_A_VaR_years.append(wgt_VaR_years)

if j<6:
    wgt_ES_years = wgt_VaR_years
else:
    wgt_sum_loss = []
    wgt_sum_loss_value = 0
    for k in range(4):
        wgt_sum_loss_value = wgt_sum_loss_value + wgt_A_VaR_years[j-k]
        wgt_sum_loss.append(wgt_sum_loss_value/(k+1))
    s = pd.Series(wgt_sum_loss)
    wgt_ES_years = s.quantile(.975)
wgt_A_ES_years.append(wgt_ES_years)

return(wgt_A_VaR_years, wgt_A_ES_years)

years_2 = 2
wgt_A_VaR_lambda1 = weighed_VaR_and_ES(list_lambda1, years_2)[0][1: 252*20]
wgt_A_ES_lambda1 = weighed_VaR_and_ES(list_lambda1, years_2)[1][1: 252*20]

wgt_A_VaR_lambda2 = weighed_VaR_and_ES(list_lambda2, years_2)[0][1: 252*20]
wgt_A_ES_lambda2 = weighed_VaR_and_ES(list_lambda2, years_2)[1][1: 252*20]

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d, '%Y-%m-%d').date() for d in timeline]

fig, ax = plt.subplots()
ax.plot(timeline, wgt_A_VaR_lambda1, 'r-', label='VaR lambda=0.94')
ax.plot(timeline, wgt_A_ES_lambda1, 'y-', label='ES lambda=0.94')

ax.plot(timeline, wgt_A_VaR_lambda2, 'b-', label='VaR lambda=0.97')
ax.plot(timeline, wgt_A_ES_lambda2, 'g-', label='ES lambda=0.97')

legend = ax.legend(loc='upper right', shadow=True)
ax.set_title('INTC VaRs and ESs Exponential')

plt.show()

```

```

In [6]: ##### HW5 Problem3 ##### PART 5: VaR 2yr Window vs Equivalent Exponential
import numpy as np
import scipy.stats as ss
import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt

path = '~/Documents/Semester3/M5320/homework/HW4/INTC-yahoo.csv'

```



```

sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
data1 = sto_AMD.values
AMD_close = list(data1[:,6])

A_log_rtn = []
A_log_rtn_sq = []

for i in range(1 , len(data1)-1):
    log_return = math.log( AMD_close[i]/AMD_close[i+1] )
    A_log_rtn.append(log_return)
    A_log_rtn_sq.append(log_return**2)

def vol_and_mu(years):
    S0 = 10000
    T = 5/252
    p = 0.99

    A_vol_years = []
    A_mu_years = []
    A_VaR_years = []
    A_ES_years = []

    for i in range(len(data1)-252*years):
        vol_years = np.std(A_log_rtn[i:i+252*years]) * np.sqrt(252)
        mu_years = np.mean(A_log_rtn[i:i+252*years])*252 + (vol_years**2)/2
        VaR_years = S0 - S0 * np.exp( vol_years * T**(0.5)* ss.norm.ppf(1-p) + (mu_years - pow(
            vol_years, 2)/2)*T)
        A_vol_years.append(vol_years)
        A_mu_years.append(mu_years)
        A_VaR_years.append(VaR_years)

        if i<6:
            ES_years = VaR_years
        else:
            sum_loss = []
            sum_loss_value = 0
            for k in range(4):
                sum_loss_value = sum_loss_value + A_VaR_years[i-k]
                sum_loss.append(sum_loss_value/(k+1))
            s = pd.Series(sum_loss)
            ES_years = s.quantile(.975)
            A_ES_years.append(ES_years)

    return(A_VaR_years, A_ES_years, A_mu_years, A_vol_years)

A_VaR_2years = vol_and_mu(10) [0] [1:252*20]

#####

## exponential weighting paramater lambda

```

```

lambda1 = 0.998614

### List the lambda values

def list_lambdas(lambda_k):
    list_lambda = []
    for i in range(len(data1)):
        lambda_value = (lambda_k**i)
        list_lambda.append(lambda_value)
    return(list_lambda)

list_lambda1 = list_lambdas(lambda1)

def weigthed_VaR_and_ES(list_lambda, years):
    wgt_A_vol_years = []
    wgt_A_mu_years = []

    wgt_log_rtn = []
    wgt_log_rtn_sq = []

    for i in range( len(data1)-1 ):
        log_return = math.log( AMD_close[i]/AMD_close[i+1] )
        wgt_log_return = log_return * list_lambda[i]
        wgt_log_rtn.append(wgt_log_return)

        log_return_sq = log_return ** 2
        wgt_log_rtn_sq.append( log_return_sq * list_lambda[i] )

    S0 = 10000
    T = 5/252
    p = .99

    wgt_A_VaR_years = []
    wgt_A_ES_years = []

    for j in range(len(data1)-252*years):
        wgt_mu_lambda = 100* sum(wgt_log_rtn[j:j+252*years])/sum(list_lambda[j:j+ 252*years])
        wgt_vol_lambda =((sum(wgt_log_rtn_sq[j:j+252*years])/sum(list_lambda[j:j+ 252*years])))-
        wgt_VaR_years = S0 - S0 * np.exp(wgt_vol_lambda * T**(0.5)* ss.norm.ppf(1-p) + (wgt_mu_
        wgt_A_vol_years.append(wgt_vol_lambda)
        wgt_A_mu_years.append(wgt_mu_lambda)
        wgt_A_VaR_years.append(wgt_VaR_years)

    if j<6:
        wgt_ES_years = wgt_VaR_years
    else:
        wgt_sum_loss = []
        wgt_sum_loss_value = 0
        for k in range(4):
            wgt_sum_loss_value = wgt_sum_loss_value + wgt_A_VaR_years[j-k]

```

```

        wgt_sum_loss.append(wgt_sum_loss_value/(k+1))
        s = pd.Series(wgt_sum_loss)
        wgt_ES_years = s.quantile(.975)
        wgt_A_ES_years.append(wgt_ES_years)

    return(wgt_A_VaR_years, wgt_A_ES_years)

years_2 = 10
wgt_A_VaR_2years = weighed_VaR_and_ES(list_lambda1, years_2)[0][1: 252*20]

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d,'%Y-%m-%d').date() for d in timeline]

fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
ax2.plot(timeline, wgt_A_VaR_2years, 'r-', label='VaR 10yr Exp Equiv')
ax1.plot(timeline, A_VaR_2years, 'b-', label='VaR 10yr windowed')

legend1 = ax1.legend(loc='upper right', shadow=True)
legend2 = ax2.legend(loc='upper right', shadow=True)
ax1.set_title('INTC VaR 10yr Window vs Equivalent Exponential')
plt.show()

```

In [9]: ##### HW5 Problem 4 ##### PART 1: Φ AND D_Φ

```

from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
import pylab
def norm_func(x,h):
    pdf = norm.pdf(x)
    cdf = norm.cdf(x)
    D_Phi = (norm.cdf(x+h)-norm.cdf(x-h))/(2*h)
    DErr = norm.pdf(x) - D_Phi
    first_derivative = np.gradient(cdf)
    second_derivative = np.gradient(first_derivative)
    return(pdf, D_Phi, DErr, second_derivative)

x = np.linspace(-3,3, 100)
h1 = 1e-04
fun1 = norm_func(x, h1)
h2 = 1e-05
fun2 = norm_func(x, h2)
h3 = 1e-06
fun3 = norm_func(x, h3)

fig, ax1 = plt.subplots()
ax1.plot(x, fun1[0], 'r-', lw=1, alpha=0.6, label='phi')

ax1.plot(x, fun1[1], 'b-', lw=1, label='d(Phi,h=1e^-04)')

ax1.plot(x, fun2[1], 'g-', lw=1, label='d(Phi,h=1e^-05)')

```

```

ax1.plot(x, fun3[1], 'm-',lw=1, label='d(Phi,h=1e^-06)')

legend1 = ax1.legend(loc='upper right', shadow=True)
ax1.set_title('phi vs dPhi(h)')
plt.show()

In [3]: ##### HW5 Problem 4 ##### PART 2: phi - dPhi(h)
from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
import pylab
def norm_func(x,h):
    pdf = norm.pdf(x)
    cdf = norm.cdf(x)
    D_Phi = (norm.cdf(x+h)-norm.cdf(x-h))/(2*h)
    DErr = norm.pdf(x) - D_Phi
    first_derivative = np.gradient(cdf)
    second_derivative = np.gradient(first_derivative)
    return(pdf, D_Phi, DErr, second_derivative)

x = np.linspace(-3,3, 100)
h1 = 1e-04
fun1 = norm_func(x, h1)
h2 = 1e-05
fun2 = norm_func(x, h2)
h3 = 1e-06
fun3 = norm_func(x, h3)

fig, ax1 = plt.subplots()
ax1.plot(x, fun1[2], 'r-', lw=10, label='DErr(h=1e^-04)')

#ax1.plot(x, fun1[1], 'b-', lw=1, label='DErr(h=1e^-04)')

ax1.plot(x, fun2[2], 'g-',lw=2, label='DErr(h=1e^-05)')
ax1.plot(x, fun3[2], 'm-',lw=2, label='DErr(h=1e^-06)')

legend1 = ax1.legend(loc='upper right', shadow=True)
ax1.set_title('DErr = phi - dPhi(h)')
plt.show()

In [4]: ##### HW5 Problem 4 ##### PART 3: Second Derivatives
from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
import pylab
def norm_func(x,h):
    pdf = norm.pdf(x)
    cdf = norm.cdf(x)
    D_Phi = (norm.cdf(x+h)-norm.cdf(x-h))/(2*h)
    DErr = norm.pdf(x) - D_Phi
    first_derivative = np.gradient(cdf)
    second_derivative = np.gradient(first_derivative)
    return(pdf, D_Phi, DErr, second_derivative)

x = np.linspace(-3,3, 100)

```

```

h1 = 1e-05
fun1 = norm_func(x, h1)
h2 = 1e-06
fun2 = norm_func(x, h2)
h3 = 1e-07
fun3 = norm_func(x, h3)

fig, ax1 = plt.subplots()
ax1.plot(x, fun1[3], 'm-', lw=1, label='ddPhi(x,1e-05)')
ax1.plot(x, fun2[3], 'g-', lw=2, label='ddPhi(x,1e-06)')
ax1.plot(x, fun3[3], 'r-', label='ddPhi(x,1e-07)')
legend = ax1.legend(loc='upper right', shadow=True)
ax1.set_title('Second derivatives ')
plt.show()

```

In [5]: ##### HW5 Problem 5 ##### PART I : BS CALL And Delta

```

from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
import pylab
import numpy as np
import scipy.stats as ss
import time

class Solution():
    def __init__(self, type1, S0, K, r, sigma, T):
        self.type1= type1
        self.S0=S0
        self.K=K
        self.r=r
        self.sigma=sigma
        self.T=T

    def d1(self):
        return (np.log(self.S0/self.K) + (self.r + self.sigma**2 / 2) * T)/(self.sigma * np.sqrt(T))

    def d2(self):
        return (np.log(self.S0 / self.K) + (self.r - self.sigma**2 / 2) * self.T) / (self.sigma * np.sqrt(T))

    def BlackScholes(self):
        if self.type1=="C":
            return self.S0 * ss.norm.cdf(self.d1()) - self.K * np.exp(-self.r * self.T) * ss.norm.cdf(self.d2())
        else:
            return self.K * np.exp(-self.r * self.T) * ss.norm.cdf(-d2()) - self.S0 * ss.norm.cdf(-d1())

    def Delta(self):
        return np.exp(-self.r * self.T) * ss.norm.cdf(self.d1())

S0 = np.linspace(50, 150, 1000)
K = 100
r=0.05
sigma = 0.25

```

```

T = 1
type1='C'
call = Solution(type1, S0, K, r, sigma, T)
call.BlackScholes()
call.Delta()

fig, ax1 = plt.subplots()
#ax.plot(x, fun[0], 'r-', lw=1, alpha=0.6, label='norm pdf')
ax2 = ax1.twinx()
ax1.plot(S0, call.BlackScholes(), 'b-', lw=3, label='BSCall')
ax2.plot(S0, call.Delta(), 'r-', lw=3, label='Delta')
legend1 = ax1.legend(loc='upper right', shadow=True)
ax1.set_title('BS Call(blue) and delta(red)')
plt.show()

```

In [41]: ##### HW5 Problem 5 ##### PART 2: Difference derivative deltas

```

from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
import pylab
import numpy as np
import scipy.stats as ss
import time

class Solution():
    def __init__(self, type1, S0, K, r, sigma, T):
        self.type1= type1
        self.S0=S0
        self.K=K
        self.r=r
        self.sigma=sigma
        self.T=T

    def d1(self):
        return (np.log(self.S0/self.K) + (self.r + self.sigma**2 / 2) * T)/(self.sigma * np.sqrt(T))

    def d2(self):
        return (np.log(self.S0 / self.K) + (self.r - self.sigma**2 / 2) * self.T) / (self.sigma * np.sqrt(T))

    def BlackScholes(self):
        if self.type1=="C":
            return self.S0 * ss.norm.cdf(self.d1()) - self.K * np.exp(-self.r * self.T) * ss.norm.cdf(self.d2())
        else:
            return self.K * np.exp(-self.r * self.T) * ss.norm.cdf(-d2()) - self.S0 * ss.norm.cdf(-d1())

    def Delta(self):
        return np.exp(-self.r * self.T) * ss.norm.cdf(self.d1())

    def Diff_BS(h):
        h =1
        S1 = np.linspace(50, 150, 1000)+h
        S2 = np.linspace(50, 150, 1000)-h

```

```

K = 100
r=0.05
sigma = 0.25
T = 1
type1='C'

call1 = Solution(type1, S1, K, r, sigma, T)
call2 = Solution(type1, S2, K, r, sigma, T)

diff_BS = (call1.BlackScholes()-call2.BlackScholes())/(2*h)
return diff_BS

S0 = np.linspace(50, 150, 1000)
K = 100
r=0.05
sigma = 0.25
T = 1
type1='C'
call = Solution(type1, S0, K, r, sigma, T)
call.BlackScholes()
call.Delta()

h1=5*(1e-04)
diff_BS1 = Diff_BS(h1)

h2=3
diff_BS2 = Diff_BS(h2)

fig, ax1 = plt.subplots()
ax1.plot(S0, diff_BS1, 'b-', lw=1, label='dBS(x,h=1e-04)')
ax1.plot(S0, diff_BS2, 'r-', lw=1, label='dBS(x,h=1e-05)')
ax1.plot(S0, call.Delta(), 'm-', lw=1, label='delta')

legend1 = ax1.legend(loc='upper right', shadow=True)
ax1.set_title('Difference derivative deltas')
plt.show()

```

In [65]: ##### HW5 Problem3 ##### PART 2: weighted VaRs and ES

```

import numpy as np
import scipy.stats as ss
import pandas as pd
import math
import matplotlib.pyplot as plt
import datetime as dt

path = '~/Documents/Semester3/M5320/homework/HW4/AMD-yahoo.csv'
sto_AMD= pd.read_csv(path, header = 0)
sto_AMD.shape
data1 = sto_AMD.values
AMD_close = list(data1[:,6])

A_log_rtn = []

```

```

A_log_rtn_sq = []

for i in range(1 , len(data1)-1):
    log_return = math.log( AMD_close[i]/AMD_close[i+1] )
    A_log_rtn.append(log_return)
    A_log_rtn_sq.append(log_return**2)

def vol_and_mu(years):
    S0 = 10000
    T = 5/252
    p = 0.99

    A_vol_years = []
    A_mu_years = []
    A_VaR_years = []
    A_ES_years =[]

    for i in range(len(data1)-252*years):
        vol_years = np.std(A_log_rtn[i:i+252*years]) * np.sqrt(252)
        mu_years = np.mean(A_log_rtn[i:i+252*years])*252 + (vol_years**2)/2
        VaR_years = S0 - S0 * np.exp( vol_years * T**(0.5)* ss.norm.ppf(1-p) + (mu_years - pow

        if i<6:
            ES_years = VaR_years
        else:
            sum_loss = []
            sum_loss_value = 0
            for k in range(4):
                sum_loss_value = sum_loss_value + A_VaR_years[i-k]
                sum_loss.append(sum_loss_value/(k+1))
            s = pd.Series(sum_loss)
            ES_years = s.quantile(.975)
            A_ES_years.append(ES_years)

    return(A_VaR_years, A_ES_years, A_mu_years, A_vol_years)

A_VaR_2years = vol_and_mu(2) [0] [1:252*20]
A_ES_2years = vol_and_mu(2) [1] [1:252*20]
#####

## exponential weighting paramater lambda
lambda1 = 0.998614

### List the lambda values

def list_lambdas(lambda_k):

```



```

list_lambda = []
for i in range(len(data1)):
    lambda_value = (lambda_k**i)
    list_lambda.append(lambda_value)
return(list_lambda)

list_lambda1 = list_lambdas(lambda1)

def weighed_VaR_and_ES(list_lambda, years):
    wgt_A_vol_years = []
    wgt_A_mu_years = []

    wgt_log_rtn = []
    wgt_log_rtn_sq = []

    for i in range( len(data1)-1 ):
        log_return = math.log( AMD_close[i]/AMD_close[i+1] )
        wgt_log_return = log_return * list_lambda[i]
        wgt_log_rtn.append(wgt_log_return)

        log_return_sq = log_return ** 2
        wgt_log_rtn_sq.append( log_return_sq * list_lambda[i] )

    S0 = 10000
    T = 5/252
    p = .99

    wgt_A_VaR_years = []
    wgt_A_ES_years = []

    for j in range(len(data1)-252*years):
        wgt_mu_lambda = 100* sum(wgt_log_rtn[j:j+252*years])/sum(list_lambda[j:j+ 252*years])
        wgt_vol_lambda =((sum(wgt_log_rtn_sq[j:j+252*years])/sum(list_lambda[j:j+ 252*years]))
        wgt_VaR_years = S0 - S0 * np.exp(wgt_vol_lambda * T**(0.5)* ss.norm.ppf(1-p) + (wgt_mu
        wgt_A_vol_years.append(wgt_vol_lambda)
        wgt_A_mu_years.append(wgt_mu_lambda)
        wgt_A_VaR_years.append(wgt_VaR_years)

        if j<6:
            wgt_ES_years = wgt_VaR_years
        else:
            wgt_sum_loss = []
            wgt_sum_loss_value = 0
            for k in range(4):
                wgt_sum_loss_value = wgt_sum_loss_value + wgt_A_VaR_years[j-k]
                wgt_sum_loss.append(wgt_sum_loss_value/(k+1))
            s = pd.Series(wgt_sum_loss)
            wgt_ES_years = s.quantile(.975)
        wgt_A_ES_years.append(wgt_ES_years)

```

```

return(wgt_A_VaR_years, wgt_A_ES_years, wgt_A_mu_years, wgt_A_vol_years)

years_2 = 2
wgt_A_VaR_2years = weighed_VaR_and_ES(list_lambda1, years_2)[0][1: 252*20]
#wgt_A_VaR_2years = [x*50 for x in wgt_A_VaR_2years ]

wgt_A_vol_2years = weighed_VaR_and_ES(list_lambda1, years_2)[3][1: 252*20]
A_vol_2years = vol_and_mu(years_2)[3][1: 252*20]

wgt_A_mu_2years = weighed_VaR_and_ES(list_lambda1, years_2)[2][1: 252*20]
A_mu_years = vol_and_mu(years_2)[2][1: 252*20]

#print(wgt_A_vol_2years[1:10],A_vol_2years[1:10],wgt_A_mu_2years[1:10], A_mu_years[1:10])

timeline = data1[1:252*20,0]
timeline = [dt.datetime.strptime(d,'%Y-%m-%d').date() for d in timeline]

fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
ax2.plot(timeline, wgt_A_VaR_2years, 'r-', label='Wgted VaR 2 year')
ax1.plot(timeline, A_VaR_2years, 'b-', label='VaR 2 year')
plt.show()

```

In []: