

Lab3_sol

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1 Lab 3 sample solutions

First we should import libraries what we need Always we import scipy

```
In [20]: import numpy as np
```

In this lab we also need to draw graphs, so let us import plot and show from pylab

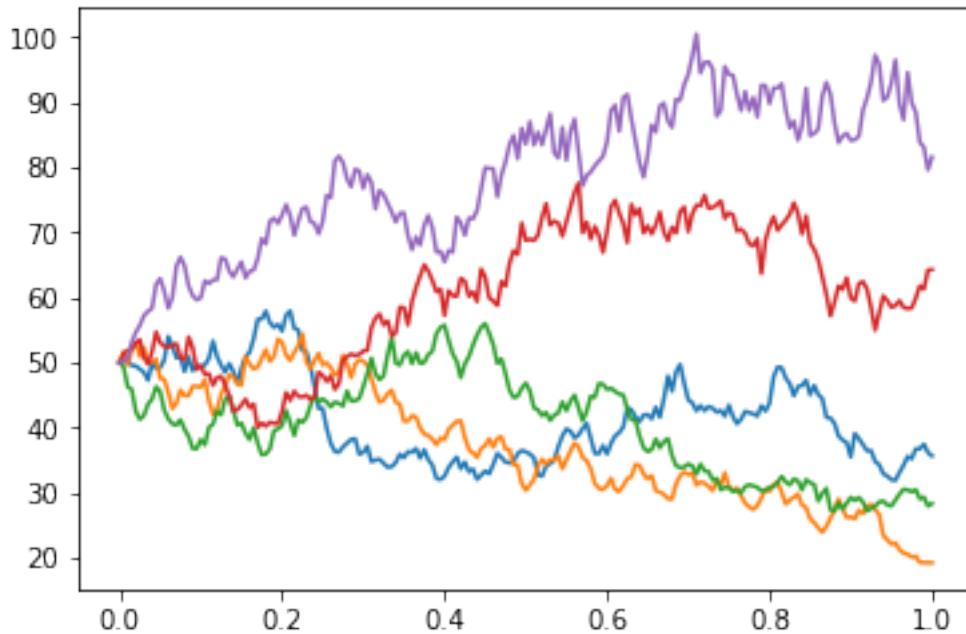
```
In [21]: import pylab as pl
```

1.1 Exercise 1

Write a function `BSgraph(S0,n,m,mu,sigma,T)` that plots the graph of n trajectories of the stock price on the interval $[0, T]$, corresponding to the Black-Scholes market model with constant parameters μ and σ . For computing the values of the stock prices divide the interval $[0, T]$ into m equal subintervals (ie. use the time points $t_i = \frac{iT}{m}$, $i = 0, 1, \dots, m$) and use the Euler's method.

Solution

```
In [22]: def BSgraph(S0,n,m,mu,sigma,T):
    hi=T/m
    S=np.zeros(shape=(m+1,n))
    S[0,:]=S0
    for i in np.arange(1,m+1):
        Xi=np.random.randn(n)
        S[i,:]=S[i-1,:]*(1+mu*hi+sigma*np.sqrt(hi)*Xi)
    t=np.linspace(0,T,m+1)
    pl.plot(t,S)
    pl.show()
BSgraph(S0=50,n=5,m=200,mu=0.1,sigma=0.5,T=1)
```



1.2 Exercise 2

In the case of pricing European options by Monte-Carlo methods, we do not want to look at the trajectories of the stock prices but need only to generate values of $S(T)$. Define a function $ST(S_0, n, m, \mu, \sigma, T)$ that returns a vector of n randomly generated values of $S(T)$ according to BS market model with constant μ and non-constant volatility given by a function $\sigma(s, t)$. Compute the mean value and standard deviation of 100000 generated stock prices at time $t = T$ in the case $\mu = 0.05$, $m = 100$, $S_0 = 10$, $T = 1$ and $\sigma(s, t) = \frac{e^{-0.1 \cdot t}}{1 + 0.005 s^2}$. (For checking answers: mean and standard deviation should be approximately 10.5 and 6.1)

Solution

```
In [23]: def ST(S0,n,m,mu,sigma,T):
        """
        generator of final stock prices S(T) in the case of BS market model
        sigma should be a function of two variables
        """
        hi=T/m
        S=np.zeros(shape=(m+1,n))
        S[0,:]=S0
        t=np.linspace(0,T,m+1)
        for i in range(1,m+1):
            Xi=np.random.randn(n)
            S[i,:]=S[i-1,:]*(1+mu*hi+sigma(S[i-1,:],t[i-1])*np.sqrt(hi)*Xi)
        return S[m,:]
```

```

def sigma_ex2(s,t):
    return np.exp(-0.1*t)/(1+0.005*s**2)
result=ST(n=100000,S0=10,mu=0.05,m=100,T=1,sigma=sigma_ex2)
print("mean:",np.mean(result))
print("standard deviation",np.std(result))

```

mean: 10.505523167

standard deviation 6.07615963614

1.3 Exercise 3

Often we have several stochastic processes in a market model. Let us consider a model with stochastic interest rate:

$$\begin{aligned}
 dS(t) &= S(t)(r(t) dt + 0.5dB_1(t)), \\
 dr(t) &= (0.05 - r(t)) dt + 0.02 dB_2(t),
 \end{aligned}$$

where B_1 and B_2 are independent Brownian motions. Use Euler-Maruyama method for defining a function that for given n and m outputs n generated values of $S(0.5)$ in the case $S(0) = 100$, $r(0) = 0.04$.

Solution

```

In [24]: def Ex3(m,n):
    #parameters fixed in the problem
    T=0.5
    S0=100
    r0=0.04
    delta_t=T/m
    S=np.zeros(shape=(m+1,n))
    r=np.zeros(shape=(m+1,n))
    S[0,:]=S0
    r[0,:]=r0
    for i in range(1,m+1):
        dB1=np.sqrt(delta_t)*np.random.randn(n)
        dB2=np.sqrt(delta_t)*np.random.randn(n) #each time we use randn, new (independen
        S[i,:]=S[i-1,:]*(1+r[i-1,:]*delta_t+0.5*dB1)
        r[i,:]=(0.05-r[i-1,:])*delta_t+0.02*dB2+r[i-1,:]
    #need to return only values for stock prices at T=0.5
    return S[m,:]
#try if the function works (each time we use the generator, we get different numbers)
print(Ex3(10,5))
print(Ex3(10,5))
#the average of a large number of generated values is an approximation
#of the expected value of S(T) for reasonably large values of m
ST_values=np.mean(Ex3(100,100000))
print("average of 100000 values for m=100:",ST_values)

```

```
[ 91.28955645  89.699855   91.10688581 105.73033266 125.26055018]
[ 134.87974335  98.14246629 127.20016126 115.88339273 144.073344 ]
average of 100000 values for m=100: 101.989826068
```