#### Simulation Methods in Fin Mat, Lab 3

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#### Black-Scholes market model

$$dS(t) = S(t) \cdot (\mu \cdot dt + \sigma \cdot dB(t))$$

B(t) - the standard Brownian motion  $\sigma$  - the volatility. Can be a function of S(t) and t mu - trend

No-Arbitrage pricing: the price of an European option with exercise time  $\mathcal{T}$  and pay-off p is given by

$$Price(t, S_t) = E[e^{-r(T-t)}p(S(T)) \mid S(t) = S_t]$$

where the expectation is taken uder the (risk-neutral) measure for which the stock price equation is

$$dS(t) = S(t)((r-D) dt + \sigma dB(t)).$$

r - the risk free interest rate, D - dividend rate

# Using MC to find option prices in the case of constant volatility

If  $\sigma$  is constant, then the solution of

$$dS(t) = S(t)((r-D) dt + \sigma dB(t))$$

satisfies

$$S(T) = S(t)e^{(r-D-\frac{\sigma^2}{2})(T-t)+\sigma(B(T)-B(t))}$$

Since  $B(T) - B(t) \sim N(0, T - t)$ , we can write

$$Price(t, S_t) = E[e^{-r(T-t)}p(S(T))]$$

where

$$S(T) = S_t e^{(r-D-\frac{\sigma^2}{2})(T-t)+\sigma\sqrt{T-t}X}$$
],  $X \sim N(0,1)$ 

How to generate S(T) if  $\sigma$  is not constant?

### Euler-Maruyama method

Consider a SDE (Stochastic Differential Equation)

$$dY(t) = \alpha(Y(t), t) dt + \beta(Y(t), t) dB(t), Y(0) = Y_0$$

Euler-Maruyama method: Use the intuitive meaning of SDE: over small interval  $(t_i, t_{i+1})$  the change is Y is approximately the equal to  $\alpha$ , times the change in t plus  $\beta$  times the change in the Brownian motion.

Mathematically:

$$Y(t_{i+1})-Y(t_i)\approx \alpha(Y(t_i),t_i)\cdot(t_{i+1}-t_i)+\beta(Y(t_i),t_i)\cdot(B(t_{i+1})-B(t_i))$$

Use notations  $t_i = i\Delta t$ , where  $\Delta t = \frac{i \cdot T}{m}$ . Define

$$Y_{i+1} = Y_i + \alpha(Y_i, t_i)\Delta t + \beta(Y_i, t_i)\sqrt{\Delta t}X_i, \quad i = 0, 1, \dots, m-1,$$

where  $X_i \sim N(0,1)$ . Similar formulas for systems of SDE.

## Euler-Maruyama method (continued)

$$dY(t) = \alpha(Y(t), t) dt + \beta(Y(t), t) dB(t), Y(0) = Y_0$$

Euler-Maruyama method

$$Y_{i+1} = Y_i + \alpha(Y_i, t_i)\Delta t + \beta(Y_i, t_i)\sqrt{\Delta t}X_i, i = 0, 1, \dots, m-1$$

BS market model

$$dS(t) = S(t) \cdot (\mu \cdot dt + \sigma \cdot dB(t))$$

Euler-Maruyama method

$$S_{i+1} = S_i \cdot (1 + \mu(S_i, t_i)\Delta t + \sigma(S_i, t_i)\sqrt{\Delta t} X_i), i = 0, 1, \dots, m-1$$

Use  $S_m$  instead of S(T)

#### Strong and weak convergence

- ► The trajectory of Brownian motion determines a trajectory of the stock price
- A given trajectory B(t),  $t \in [0, T]$  determines the exact values  $S(t_i)$ , i = 0, ..., m
- ▶ can also compute approximate values  $S_i$ , i = 0, ..., m according to Euler's method with  $\sqrt{\Delta t} X_i = B(t_i) B(t_{i-1})$
- Strong convergence characterizes how close are those values in the average, how quicly the differences in prices goes to 0 as m increases
- ▶ Weak convergence is related to how close are the distributions of S(T) and  $S_m$ . Closeness of distribution is much weaker condition than closeness of values of random variables

### Illustration



