

# Monte Carlo Simulation

Simulate Stochastic Differential Equations (SDE)

# Simulation of SDE

- Certain SDE's are solvable in closed form, e.g., geometric Brownian Motion:

$$dS_t = rS_t dt + \sigma S_t dW_t$$

Which has an explicit solution

$$S_t = S_0 \exp \left( \left( r - \frac{1}{2}\sigma^2 \right) t + \sigma W_t \right)$$

- However, most of SDE's are NOT solvable, which requires us to resort to numerical solutions.

# Convergence of SDE solutions

- Suppose we have a general SDE:

$$dX(t) = b(X_t, t)dt + \sigma(X_t, t)dW_t$$

- Suppose we have a solution  $\hat{X}_t^\Delta$  (via numerical solution through discretization of time step  $\Delta$ )

- $\hat{X}_t$  has a strong order of convergence of  $\alpha$  if

$$E[|\hat{X}_t^\Delta - X_t|] \leq C\Delta^\alpha \text{ for some fixed constant } C$$

- $\hat{X}_t$  has a weak order of convergence of  $\alpha$  if

$$E[|f(\hat{X}_t^\Delta) - f(X_t)|] \leq C\Delta^\alpha \text{ for all sufficiently smooth function } f \text{ and some fixed constant } C$$

# Convergence of SDE Solutions

- The larger the order  $\alpha$ , the better quality the solution has.
- Generally, for any given solution scheme, its strong order of convergence is typically less than its weaker order of convergence.

# Two SDE solution schemes

- Euler scheme

$$X_{i+1} = X_i + b(X_i, t_i)(t_{i+1} - t_i) + \sigma(X_i, t_i)(W_{i+1} - W_i)$$

- Milstein scheme

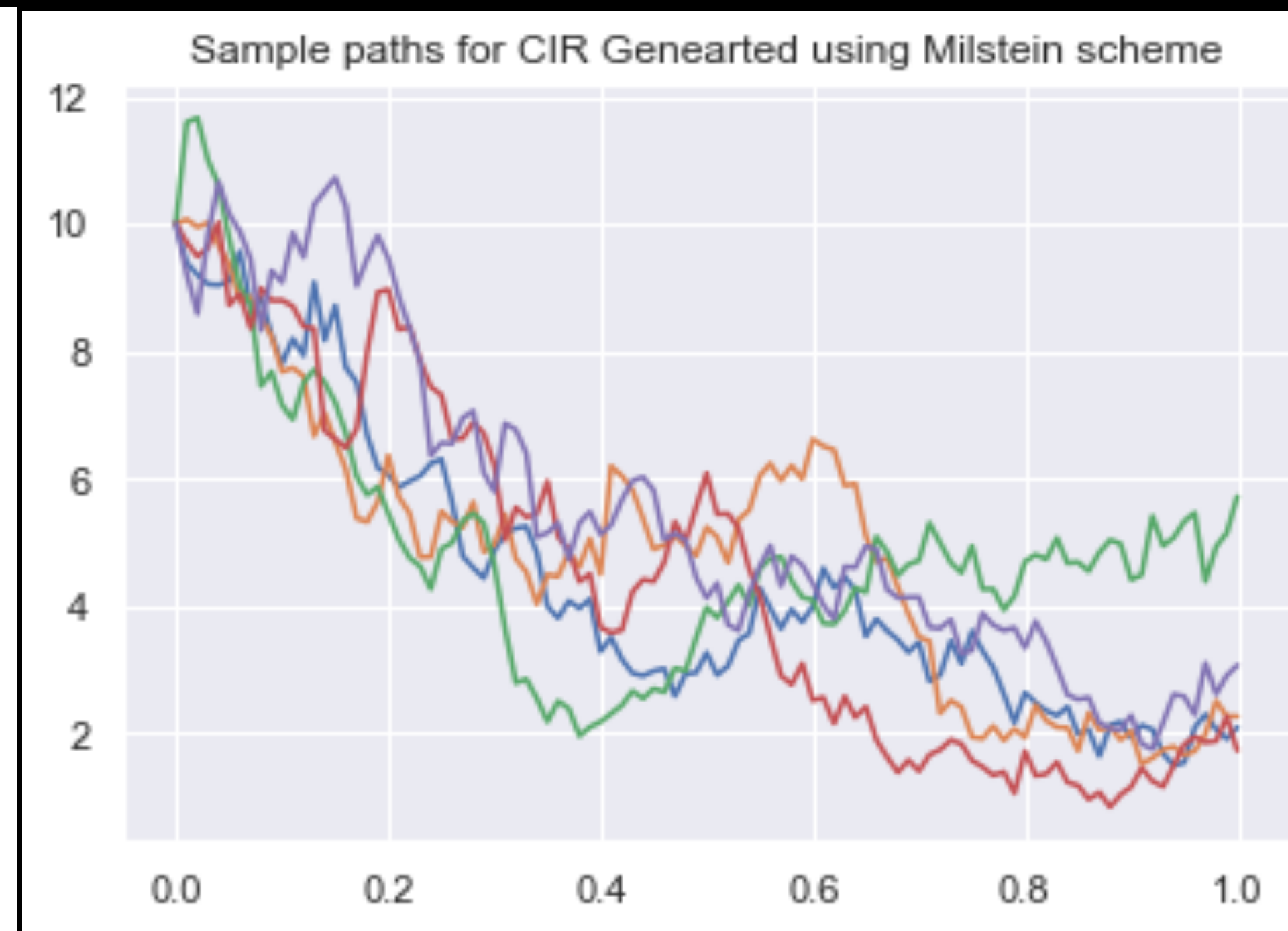
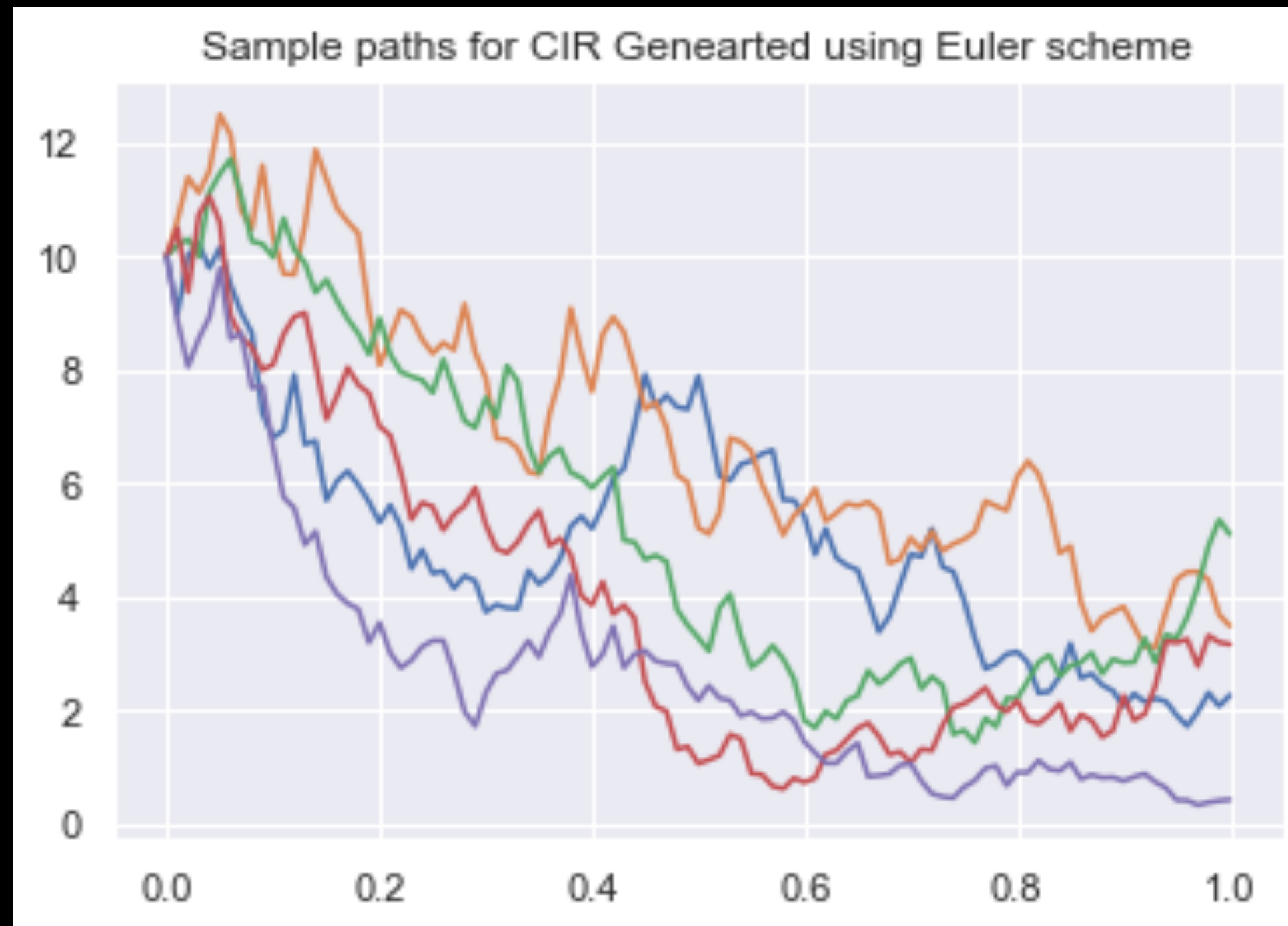
$$X_{i+1} = X_i + b(X_i, t_i)(t_{i+1} - t_i) + \sigma(X_i, t_i)(W_{i+1} - W_i) + \frac{1}{2}\sigma(X_i, t_i)\sigma_x(X_i, t_i)\left((W_{i+1} - W_i)^2 - (t_{i+1} - t_i)\right)$$

- Euler scheme is more straightforward, but Milstein scheme is more accurate (note that for Milstein scheme derivative of diffusion function with respect to x is required)

# Two SDE solution schemes

- For Euler scheme, strong and weak orders of convergences are  $1/2$  and  $1$  respectively
- For Milsten scheme, strong and weak orders of convergences are  $1$  in both cases.

# SDE simulation trajectories



# Application - Option Pricing with Heston Model

- Heston Stochastic Volatility Model

$$dS_t = rS_t dt + \sqrt{V_t} S_t dW_t$$

$$dV_t = \kappa(\theta - V_t)dt + \sigma\sqrt{V_t}dB_t$$

$$d[W_t, B_t] = \rho dt$$

- Using Milstein scheme we get pretty accurate option price (see notebook)