## HW1\_Q7\_Python

February 24, 2024

```
[]: # imports
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
plt.rcParams["figure.figsize"] = (12, 9) #set default figure size
rng = np.random.default_rng() # random number generator
```

Two ways to simulate a GBM process:

$$dZ(t) = \mu Z(t)dt + \sigma Z(t)dB(t)$$

1. Exact solution to the SDE

$$Z(t) = Z(0) \cdot \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)t + \sigma B(t)\right)$$

2. Approximate the GBM by a discretized version of the SDE (Euler-Maruyama method)

$$\begin{split} \Delta \tilde{Z}(t_i) &= \tilde{Z}(t_{i-1}) \cdot [\mu \Delta t + \sigma \Delta B] \\ \tilde{Z}(t_i) &= \tilde{Z}(t_{i-1}) \cdot [1 + \mu \Delta t + \sigma \Delta B] \end{split}$$

```
[]:  # Parameters

ZO = 100  # initial value

mu = 0.1  # drift

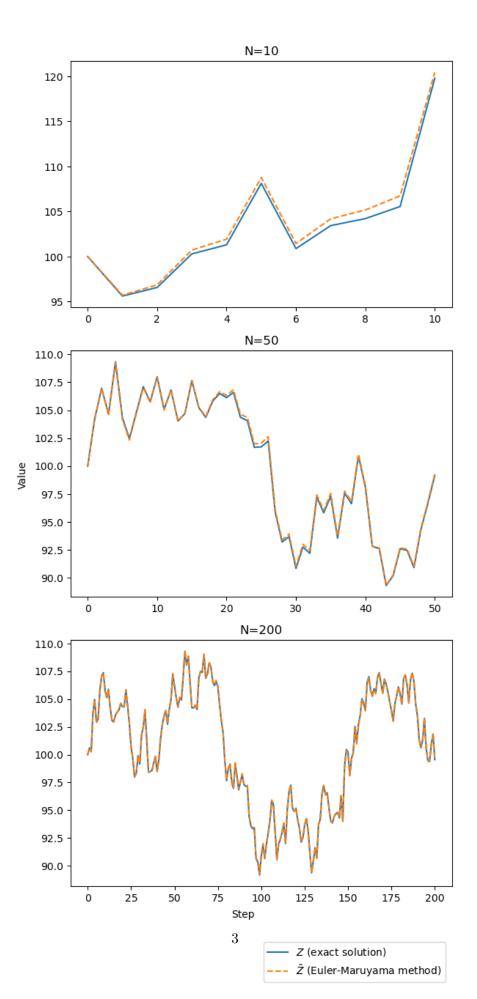
sigma = 0.2  # volatility

T = 1  # time horizon
```

```
[]: # Simulation for different N
fig, ax = plt.subplots(3, 1, figsize=(6, 12))
a = 0
for N in [10, 50, 200]:
    # Time step
    t = np.linspace(0, T, N + 1)
    dt = T / N

# Brownian motion
B = np.zeros(N + 1)
B[0] = 0
B[1:] = np.cumsum(np.sqrt(dt) * rng.standard_normal(N))
```

```
# GBM (exact solution)
   Z = Z0 * np.exp((mu - 0.5 * sigma**2) * t + sigma * B)
   # GBM (Euler-Maruyama method)
   Z_EM = np.zeros(N + 1)
   Z_EM[0] = Z0
   Z_{EM[1:]} = Z0 * np.cumprod(1 + mu * dt + sigma * np.diff(B))
   # plot
   ax[a].plot(Z, linestyle='-', color="CO")
   ax[a].plot(Z_EM, linestyle='--', color="C1")
   ax[a].set_title(f'N={N}')
   a += 1
# labels
fig.text(0.5, 0, 'Step', ha='center', va='center')
fig.text(0, 0.5, 'Value', ha='center', va='center', rotation='vertical')
fig.legend([r"$Z$ (exact solution)", r"$\tilde{Z}$ (Euler-Maruyama method)"],
           bbox_to_anchor=(0.5, -0.1),
           loc=3,
           borderaxespad=2)
fig.tight_layout()
plt.show()
```



As N becomes large, becomes negligible.	the difference	between the $\epsilon$	exact solution a	and the Euler-Ma	aruyama method