hw3_Q5.x

March 16, 2024

```
[]: # imports
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm
from IPython.display import display, Markdown

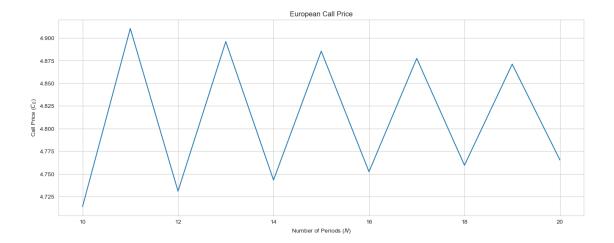
# set up the environment
%matplotlib inline
plt.rc("figure", figsize=(16, 6)) # set default size of plots
sns.set_style("whitegrid") # set default seaborn style
rng = np.random.default_rng() # random generator
```

0.1 Question 5.1

```
[]: def stock_tree(SO, sigma, T, N):
         """Stock price tree
        Args:
            SO (float): initial stock price
             sigma (float): volatility
             T (float): time to maturity
             N (int): number of steps
        Returns:
             numpy.ndarray: stock price tree
        dt = T / N \# time step
        u = np.exp(sigma * np.sqrt(dt)) # up factor (Cox-Ross-Rubinstein model)
        d = 1 / u \# down factor
        S = np.zeros((N + 1, N + 1)) # stock price tree
        S[0, 0] = S0 # initial stock price
        for t in range(1, N + 1):
             for i in range(t + 1):
                 S[i, t] = S0 * u**(t - i) * d**i # i down, t - i up
        return S
```

```
"""European call option price tree
         Args:
             SO (float): initial stock price
             sigma (float): volatility
             T (float): time to maturity
             N (int): number of steps
             K (float): strike price
             r (float): risk-free rate
             q (float, optional): dividend yields. Defaults to O.
         Returns:
             numpy.ndarray: European call option price tree
         S = stock_tree(S0, sigma, T, N) # stock price tree
         dt = T / N \# time step
         u = np.exp(sigma * np.sqrt(dt)) # up factor
         d = 1 / u \# down factor
         p = (np.exp((r - q) * dt) - d) / (u - d) # risk-neutral probability
         euro_call = np.zeros((N + 1, N + 1)) # European call option price tree
         for i in range(N + 1):
             euro_call[i, N] = max(S[i, N] - K, 0) # payoff at maturity
         for t in range(N - 1, -1, -1):
             for i in range(t + 1):
                 euro_call[
                     i,
                     # backward induction
                     t] = np.exp(-r * dt) * (p * euro_call[i, t + 1] + (1 - p) *_{\sqcup}
      \rightarroweuro_call[i + 1, t + 1])
         return euro call
[]: calls = np.zeros(11)
     for n in range(10, 21):
         calls[n - 10] = euro_call_tree(S0=50, sigma=0.3,
                                         T=0.5, K=50, r=0.05, N=n) [0, 0]
     # plot
     fig, ax = plt.subplots()
     ax.plot(range(10, 21), calls)
     plt.xlabel(r"Number of Periods ($N$)")
     plt.ylabel(r"Call Price ($C_0$)")
     plt.title("European Call Price")
    plt.show()
```

def euro_call_tree(S0, sigma, T, N, K, r, q=0):



0.2 Question 5.2

```
[]: def Black_Scholes_Call(S, K, r, T, sigma, q=0):
           """Black-Scholes call option price
           Args:
                S (float): initial stock price
                K (float): strike price
                r (float): risk-free rate
                T (float): time to maturity
                sigma (float): volatility
                q (float, optional): dividend yields. Defaults to O.
           Returns:
                float: Black-Scholes (European) call option price
           11 11 11
           if sigma == 0:
                return max(S * np.exp(-q * T) - K * np.exp(-r * T), 0)
           else:
                d1 = (np.log(S / K) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
                d2 = d1 - sigma * np.sqrt(T)
                \texttt{return} \ \texttt{S} \ * \ \texttt{np.exp}(-\texttt{q} \ * \ \texttt{T}) \ * \ \texttt{norm.cdf(d1)} \ - \ \texttt{K} \ * \ \texttt{np.exp}(-\texttt{r} \ * \ \texttt{T}) \ * \ \texttt{norm}.
        \hookrightarrowcdf(d2)
```

Black-Scholes Call Price: $C_0 = 4.8174$

```
[]: n = 10
while True:
    call_price_tree = euro_call_tree(
        S0=50, sigma=0.3, T=0.5, K=50, r=0.05, N=n)[0, 0]
    if abs(call_price_bs - call_price_tree) < 0.01: # penny accuracy (0.01)
        display(Markdown("Number of periods to get penny accuracy: $N = {}$".
        oformat(n)))
        break
        n += 1</pre>
```

Number of periods to get penny accuracy: N = 103

0.3 Question 5.3

```
[]: def euro_put_tree(S0, sigma, T, N, K, r, q=0):
         """European put option price tree
         Args:
             SO (float): initial stock price
             sigma (float): volatility
             T (float): time to maturity
             N (int): number of steps
             K (float): strike price
             r (float): risk-free rate
             q (float, optional): dividend yields. Defaults to O.
         Returns:
             numpy.ndarray: European put option price tree
         S = stock_tree(S0, sigma, T, N)
         dt = T / N
         u = np.exp(sigma * np.sqrt(dt))
         d = 1 / u
         p = (np.exp((r - q) * dt) - d) / (u - d) # risk-neutral probability
         euro_put = np.zeros((N + 1, N + 1))
         for i in range(N + 1):
             euro_put[i, N] = max(K - S[i, N], 0)
         for t in range(N - 1, -1, -1):
             for i in range(t + 1):
                 euro_put[i, t] = np.exp(-r * dt) * (p * euro_put[i, t])
                                                                    t + 1] + (1 - p) *_{11}
      \rightarroweuro_put[i + 1, t + 1])
         return euro_put
     def american_put_tree(S0, sigma, T, N, K, r, q=0):
         """American put option price tree
```

```
Arqs:
             SO (float): initial stock price
             sigma (float): volatility
             T (float): time to maturity
             N (int): number of steps
             K (float): strike price
             r (float): risk-free rate
             q (float, optional): dividend yields. Defaults to O.
         Returns:
             numpy.ndarray: American put option price tree
         S = stock_tree(S0, sigma, T, N)
         dt = T / N
         u = np.exp(sigma * np.sqrt(dt))
         d = 1 / u
         p = (np.exp((r - q) * dt) - d) / (u - d) # risk-neutral probability
         american_put = np.zeros((N + 1, N + 1))
         for i in range(N + 1):
             american_put[i, N] = max(K - S[i, N], 0) # payoff at maturity
         for t in range(N - 1, -1, -1):
             for i in range(t + 1):
                 american put[i, t] = max(
                     K - S[i, t],
                     np.exp(-r * dt) * (p *
                                         american_put[i, t + 1] + (1 - p) *_{\sqcup}
      \rightarrowamerican_put[i + 1, t + 1]),
                 ) # allowing early exercise
         return american_put
[]: early_exercise_premium = (american_put_tree(S0=50, sigma=0.3, T=1, N=500, K=50,
      \Rightarrowr=0.05, q=0.02)[0, 0] -
```

Early Exercise Premium: 0.1752

0.4 - 5.5

```
[]: def MC_Heston(SO, K, T, r, vO, kappa, theta, gamma, rho, N, M, q=0):
    """Monte Carlo simulation for Heston model

Args:
    SO (float): initial stock price
```

```
T (float): time to maturity
             r (float): risk-free rate
             v0 (float): initial variance
             kappa (float): mean reversion speed
             theta (float): mean reversion level
             gamma (float): volatility of variance
             rho (float): correlation between stock price and variance
             N (int): number of steps
             M (int): number of paths
             q (float, optional): dividend yields. Defaults to 0.
         Returns:
             float: Heston call option price
         dt = T / N \# time step
         logS = np.zeros((M, N + 1)) # log stock price paths
         v = np.zeros((M, N + 1)) # variance paths
         logS[:, 0] = np.log(S0) # initial stock price
         v[:, 0] = v0 # initial variance
         for i in range(1, N + 1):
             Z1 = rng.standard_normal(size=M)
             Z2 = rho * Z1 + np.sqrt(1 - rho**2) * rng.standard_normal(size=M)
             logS[:, i] = logS[:, i - 1] + 
                 (r - q - 0.5 * v[:, i - 1]) * dt + np.sqrt(v[:, i - 1] * dt) * Z1
             v[:, i] = np.maximum(v[:, i - 1] + kappa * (theta - v[:, i - 1])
                                  * dt + gamma * np.sqrt(v[:, i - 1] * dt) * Z2, 0)
         call_mean = np.mean(
             np.exp(-r * T) * np.maximum(np.exp(logS[:, -1]) - K, 0))
         call_std = np.std(
             np.exp(-r * T) * np.maximum(np.exp(logS[:, -1]) - K, 0)) / np.sqrt(M)
         return call_mean, call_std
[]: call_mean, call_std = MC_Heston(
         S0=100.
         K=100,
         T=1,
         r=0.01,
         v0 = 0.09,
         kappa=0.2,
         theta=0.5,
         gamma=0.1,
         rho = -0.6,
         N=100,
         M=1000,
     display(Markdown(f"Heston Call Price: ${call_mean:.4f} \pm {call_std:.4f}$"))
```

K (float): strike price

Heston Call Price: 14.1308 ± 0.7552