

Short Introduction

These example below is taken from the Class: GRA6535 Derivatives, at BI Norwegian Business school. The task was to program different scenarios for the options. This was an excellent exercise to combine knowledge with programming to understand how to build pricing functions for options.

Exercise 1

Task description:

Find the price of an American put with $S_0=K= 30$, $r= 0.04$, $\sigma= 0.2$, and maturity in 7 months using the binomial method. the option numerically (without using any analytical formulas of the greeks).

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In [1]: # First we need to import necessary modules in this case we only need the math
import math

# Building our Binomial Tree (this is our method for pricing the American put
def put_binomial(s0, K, r, sigma, T, n):
    # Calculating the time interval
    dt = T / n

    # Calculating the discount factor (Our risk-free)
    rf_discount = math.exp(-r * dt)

    # Calculate the up and down factors
    u = math.exp(sigma * math.sqrt(dt))
    d = 1 / u

    # Calculating the risk-neutral probability p
    p = (math.exp(r * dt) - d) / (u - d)

    # Creating our binomial price tree with a double for loop
    tree = [[0 for j in range(i + 1)] for i in range(n + 1)]
    tree[0][0] = s0

    # Forward calculating the stock price in our tree based on u and d
    for i in range(1, n + 1):
        tree[i][0] = tree[i - 1][0] * u # Upward prices
        for j in range(1, i + 1):
            tree[i][j] = tree[i - 1][j - 1] * d # Downward prices

    # Calculating the option values at the maturity, the end of our trees
    options_in_tree = [[0 for j in range(i + 1)] for i in range(n + 1)]
    for j in range(n + 1):
        options_in_tree[n][j] = max(0, K - tree[n][j])

    # Then we need to work our way backwards, using the backward induction met
    for i in range(n - 1, -1, -1):
        for j in range(i + 1):
            # Since this is an American option we need to take the early exerc
            exercise_value = max(0, K - tree[i][j])
            # Finding the PV by discounting the price
            continuation_value = rf_discount * (p * options_in_tree[i + 1][j + 1] + (1 - p) * options_in_tree[i + 1][j])
            # Choose either the early exercise price or the continuation value
            options_in_tree[i][j] = max(exercise_value, continuation_value)
    return options_in_tree[0][0] # Returning the option value of the stock

# Inserting our values into the function for finding the price for the America
option_price = put_binomial(s0=30, K=30, r=0.04, T=7/12, sigma=0.2, n=4)
print(f"The American put option price is: ${round(option_price,4)}")

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The American put option price is: \$1.1386

Exercise 2

Task description:

Find the price of a European call with $S_0=K= 50$, $r= 0.03$, $\sigma= 0.25$, and maturity in 5 months using the Monte Carlo method. Compute the delta, gamma, and vega of the option numerically (without using any analytical formulas of the greeks)

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In [2]: # The only module we need to import is the numpy
import numpy as np

# Defining our global variables with values
s0 = 50
K = 50
r = 0.03
T = 5/12
sigma = 0.25
num_simulations = 10000

# Finding the price for European Call option
def call_pricing(s0, K, r, T, sigma, num_simulations):
    # Our Wiener process
    z = np.random.standard_normal(num_simulations)
    st = s0 * np.exp((r-0.5 * sigma**2)*T + np.sqrt(T)*z)
    payoffs = np.maximum(st-K, 0)
    price = np.exp(-r*T)*np.mean(payoffs)
    return price

# Computing the price
call_price = call_pricing(s0=s0, K=K, r=r, T=T, sigma=sigma, num_simulations=num_simulations)
print(f"The European Call option price is: ${round(call_price, 3)}")

# Finding the greeks for our European call
def calculate_greeks():
    # We define the small changes as delta
    delta_stock = 0.01
    delta_sigma = 0.01

    # Calculating call price for Delta up and down
    call_price_delta_up = call_pricing(s0+delta_stock, K=K, r=r, T=T, sigma=sigma)
    call_price_delta_down = call_pricing(s0 - delta_stock, K=K, r=r, T=T, sigma=sigma)

    # Calculating the call price for Vega
    call_price_vega = call_pricing(s0, K=K, r=r, T=T, sigma=sigma+delta_sigma, num_simulations=num_simulations)

    # Calculating Delta, Vega and Gamma
    delta = (call_price_delta_up - call_price)/delta_stock
    gamma = (call_price_delta_up - 2*call_price + call_price_delta_down) / (delta_stock**2)
    vega = (call_price_vega - call_price) / delta_sigma

    print(f"The Delta is: {round(delta, 4)}")
    print(f"The Gamma is: {round(gamma, 4)}")
    print(f"The Vega is: {round(vega, 4)}")

if __name__ == "__main__":
    calculate_greeks()

```

The European Call option price is: \$20.511

The Delta is: -89.3

The Gamma is: -8018.4551

The Vega is: -73.4494

Exercise 3

Task description:

Find the price of a European call with $S_0=K= 40$, $r= 0.05$, and maturity in 8 months. The risk-neutral dynamics of the stock price is $dS_t=rS_tdt+\sigma S_t\sigma dz_t$. Here the volatility of the stock price is stochastic and: $d\sigma_t=u\sigma_tdt+v\sigma_tdw_t$, $\sigma_0= 0.3$, where $u= 0.3$, $v= 0.2$, and w_t is a Wiener process independent to z_t .

Choose the appropriate numerical method. Compute the delta, gamma, and vega of the option numerically (without using any analytical formulas of the greeks)


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In [3]: # Import our module numpy
import numpy as np
from numba import jit
import matplotlib.pyplot as plt

# Defining our necessary global variables with values
s0 = 40
K = 40
r = 0.05
sigma_0 = 0.3
v = 0.2
u = 0.3
T = 8/12
num_simulations = 100
dt = T/1000 # Choosing the delta time to be small timestamps

# Simulating a number of different price paths for the stock and its volatility
# Adding the NUMBA Jit function
@jit(nopython=True)
def price_path(s0=s0, sigma_0=sigma_0, r=r, u=u, v=v, T=T, dt=dt):
    N = int(T/dt)
    S = np.zeros(N)
    sigma = np.zeros(N)
    S[0] = s0
    sigma[0] = sigma_0

    # Defining our paths
    for t in range(1, N):
        # Our standard normal distributed stochastic variables
        z, w = np.random.standard_normal(2)

        # Calculating the stochastic volatility
        sigma[t] = sigma[t-1] + u*sigma[t-1]*dt + v*sigma[t-1]*w*np.sqrt(dt)

        # The risk-neutral price dynamics of the stock
        S[t] = S[t-1] + r*S[t-1]*dt + sigma[t-1] * S[t-1] * z * np.sqrt(dt)
    return S[-1], sigma[-1]

# Since we have the opportunity to choose which method we want to utilize, I c
def mc_method(s0=s0, K=K, sigma_0=sigma_0, r=r, u=u, v=v, T=T, num_simulations=n
    payoffs = [max(price_path(s0=s0, sigma_0=sigma_0, r=r, u=u, v=v, T=T, dt=dt
        for _ in range(num_simulations))]
    price = np.exp(-r*T)*np.mean(payoffs)
    return price

# Calculating the necessary greeks
def calculate_greeks():
    # We define the small changes as delta
    delta_stock = 0.01
    delta_sigma = 0.01

    # Calculating the normal call price
    call_price = mc_method(s0=s0, K=K, sigma_0=sigma_0, r=r, u=u, v=v, T=T, num
    # Calculating call price for Delta up and down

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call_price_delta_up = mc_method(s0 + delta_stock,K=K, r=r, T=T,sigma_0=sig
call_price_delta_down = mc_method(s0 - delta_stock, K=K, r=r, T=T, sigma_0

# Calculating the call price for Vega
call_price_vega = mc_method(s0, K=K, r=r, T=T, sigma_0=sigma_0+delta_sigma

# Calculating Delta, Vega and Gamma
delta = (call_price_delta_up - call_price)/delta_stock
gamma = (call_price_delta_up - 2*call_price + call_price_delta_down) / del
vega = (call_price_vega - call_price) / delta_sigma
print(f"The price for the Call Option is: ${round(call_price, 4)}")
print(f"The Delta is: {round(delta, 4)}")
print(f"The Gamma is: {round(gamma, 4)}")
print(f"The Vega is: {round(vega, 4)}")

# Since we want to visualize the price path, we needed to return the whole arr
@jit(nopython=True)
def price_path_2(s0=s0, sigma_0=sigma_0, r=r,u=u, v=v, T=T, dt=dt):
    N = int(T/dt)
    S = np.zeros(N)
    sigma = np.zeros(N)
    S[0] = s0
    sigma[0] = sigma_0

    # Defining our paths
    for t in range(1, N):
        # Our standard normal distributed stochastic variables
        z, w = np.random.standard_normal(2)

        # Calculating the stochastic volatility
        sigma[t] = sigma[t-1] + u*sigma[t-1]*dt + v*sigma[t-1]*w*np.sqrt(dt)

        # The risk-neutral price dynamics of the stock
        S[t] = S[t-1] + r*S[t-1]*dt + sigma[t-1] * S[t-1] * z * np.sqrt(dt)
    return S, sigma

# Want to visualize the computed price path of the stock
def plot_price_path():
    N = int(T / dt)
    S, _ = price_path_2()
    time_stamps = np.linspace(0,T, N)

    plt.plot(time_stamps, S)
    plt.title("Stock Price Path")
    plt.xlabel("Time Stamps")
    plt.ylabel("Stock Price in $")
    plt.grid(True)
    plt.show()

if __name__ == "__main__":
    calculate_greeks()
    plot_price_path()

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The price for the Call Option is: \$5.3462

The Delta is: -133.7115

The Gamma is: -16816.7819

The Vega is: 117.7271

