

# FIN 514 - PS\_3

Spring 2023

Venu Madhav Menday & Vineeth Kotala

```
In [2]: import numpy as np
import pandas as pd
import scipy.stats as st
import matplotlib.pyplot as plt
```

```
In [3]: # ENTER INPUT FOR: start_step
N = 100
```

```
In [4]: # ENTER INPUT FOR: S0 = Original Stock Price
S0 = 227.29
# ENTER INPUT FOR: K = Exercise Price of Call Option
K = 193.20
# ENTER INPUT FOR: sigma = Annualized (Future) Volatility of Stock Price Returns
sigma = 0.3573
# ENTER INPUT FOR: r = Annualized Continuously Compounded Risk-free Rate
r = 0.048
# ENTER INPUT FOR: T = Time Length of Option in which to Exercise (In Years)
T = 1
# ENTER INPUT FOR: D = proportional dividend
D = 0.0219
# ENTER INPUT FOR: ND = number of dividends and TD = array of dividend dates
ND = 4
TD = np.zeros([ND]) #creates an array with 4 entries
TD = [1/12, 4/12, 7/12, 10/12]
Tac = [3/12, 6/12, 9/12]
```

In [42]: Tac

Out[42]: [0.25, 0.5, 0.75]

## Stock value tree

```
In [1]: def Stock_tree(N, S0, sigma, r, T, D, TD):

    stock_value = np.zeros([N+1, N+1])

    Deltat = T / N
    u = np.exp(r*Deltat + sigma * (Deltat)**0.5)
    d = np.exp(r*Deltat - sigma * (Deltat)**0.5)
    print("u = ", u, "d=", d)

    # FIRST LET'S BUILD A STOCK PRICE TREE WITH DIVIDENDS
    # Let's have dividends at grid points rather than times
    jD1 = [i/Deltat for i in TD]
    jD = [np.ceil(i) for i in jD1]
    print(jD)

    stock_value[0,0] = S0
    for j in range(1,N+1):
        stock_value[j, 0] = stock_value[j-1, 0]*d
        for i in range(1, j+1):
            stock_value[j, i] = stock_value[j-1, i-1]*u

    # This adjusts all stock prices for that j by the size of the dividend
    if j in jD: stock_value[j, :] *= (1-D)
    print(stock_value[3,0], stock_value[50,0], stock_value[50,50], stock_value[100,0], stock_value[100,100])
    return stock_value
```

```
In [5]: stock = Stock_tree(N, S0, sigma, r, T, D, TD)
stock
```

```
u = 1.0368735670785485 d= 0.9653640450761569
[9.0, 34.0, 59.0, 84.0]
204.48134220721911 37.31666485389899 1329.3428558939984 6.126681665793567 7774.879794607821
```

```
Out[5]: array([[2.27290000e+02, 0.00000000e+00, 0.00000000e+00, ...,
               0.00000000e+00, 0.00000000e+00, 0.00000000e+00],
              [2.19417594e+02, 2.35670993e+02, 0.00000000e+00, ...,
               0.00000000e+00, 0.00000000e+00, 0.00000000e+00],
              [2.11817856e+02, 2.27508303e+02, 2.44361023e+02, ...,
               0.00000000e+00, 0.00000000e+00, 0.00000000e+00],
              ...,
              [6.57420249e+00, 7.06118777e+00, 7.58424657e+00, ...,
               7.23172790e+03, 0.00000000e+00, 0.00000000e+00],
              [6.34649871e+00, 6.81661679e+00, 7.32155895e+00, ...,
               6.98125010e+03, 7.49838750e+03, 0.00000000e+00],
              [6.12668167e+00, 6.58051675e+00, 7.06796976e+00, ...,
               6.73944783e+03, 7.23867369e+03, 7.77487979e+03]])
```

```
In [6]: #Coupon dates
Cpn = 0.1185
NC = 12
TC = list(np.linspace(1/NC, 1, NC))
Face = 1000
```



In [9]:

```

def ac_option(S0, Face, K, T, r, sigma, N, stock_value, Cpn, NC, TC, Tac):

    # LIST TO SAVE RESULTS
    convertible_result = []

    # CREATE TWO DIMENSIONAL ARRAY OF SIZE [N+1,N+1] TO STORE ALL STEPS
    # option_value[N+1, N+1]
    option_value = np.zeros([N+1, N+1])

    Deltat = T / N
    u = np.exp(r*Deltat + sigma * (Deltat)**0.5)
    d = np.exp(r*Deltat - sigma * (Deltat)**0.5)
    q = (np.exp(r * Deltat) - d) / (u - d)

    #First, let's calculate the coupon dates and accrued interest

    #jc1 are the exact values of j where the coupons are paid
    #jc are the values of j where we first factor in the coupons -
    #where int will give us the j immediately before or on the coupon/call date
    jc1 = [j/Deltat for j in TC]
    jc = [int(j) for j in jc1]
    #jcall = int(tcalle/Deltat)
    #print("call period starts", jcall)

    jac1 = [j/Deltat for j in Tac]
    jac = [int(j) for j in jac1]

    j = N

    for i in range(0, j+1):
        if stock_value[j,i] >= K:
            option_value[j,i] = Face*(1+Cpn/NC)
        else:
            option_value[j,i] = Face/K*stock_value[j,i] + Face*(Cpn/NC)

    for j in range(N-1, -1, -1):

        if j in jac:
            next_ac_date = Tac[jac.index(j)]
            prev_ac_date = Tac[jac.index(j)-1]
            print("auto call date",j,jac.index(j),next_ac_date,prev_ac_date)

```

```

ndt = Tac[jac.index(j)]-Deltat*j

for i in range(0, j+1):
    cont = np.exp(-r * Deltat) * (q * option_value[j + 1, i + 1] + (1-q) * option_value[j + 1, i])
    if j in jc:
        cont = cont + Face*Cpn/NC*np.exp(-r*(TC[jc.index(j)]-Deltat*j))

    option_value[j, i] = cont

    if j in jac:
        if stock_value[j,i]>= S0:
            option_value[j, i] = Face*(1+Cpn/NC)

        elif stock_value[j,i] * np.exp(r*ndt + sigma * (ndt)**0.5) >= S0:
            option_value[j, i] = Face*(1+Cpn/NC) * np.exp(-r*ndt)
            print(option_value[j,i])

output = {'num_steps': N, 'Value': option_value[0,0]}
convertible_result.append(output)

return convertible_result

```

```

In [10]: value = ac_option(S0, Face, K, T, r, sigma, N, stock, Cpn, NC, TC, Tac)
value

```

```

auto call date 75 2 0.75 0.5
auto call date 50 1 0.5 0.25
auto call date 25 0 0.25 0.75

```

```

Out[10]: [{'num_steps': 100, 'Value': 972.9871806870709}]

```

## Problem 2

```
In [10]: # ENTER INPUT FOR: S0 = Original Stock Price
S0 = 100
# ENTER INPUT FOR: K = Excercise Price of Call Option
K = 105
# ENTER INPUT FOR: sigma = Annualized (Future) Volatility of Stock Price Returns
sigma = 0.3
# ENTER INPUT FOR: r = Annualized Continously Compounded Risk-free Rate
r = 0.04
# ENTER INPUT FOR: T = Time Length of Option in which to Exercise (In Years)
T = 0.5
# ENTER INPUT FOR: D = proportional dividend
D = 0.0
# ENTER INPUT FOR: ND = number of dividends and TD = array of dividend dates
ND = 0
TD = np.zeros([ND]) #creates an array with 4 entries
```

- **Cox, Ross and Rubinstein**





```

In [11]: def crr_European(S0, K, T, r, sigma, N, D, TD):

    # LIST TO SAVE RESULTS
    Euro_result = []

    # CREATE TWO DIMENSIONAL ARRAY OF SIZE [N+1,N+1] TO STORE ALL STEPS
    # option_value[N+1, N+1]
    stock_value = np.zeros ([N+1, N+1])
    option_value = np.zeros([N+1, N+1])

    # FOR LOOP STATEMENT: For a Binomial Tree from start_step to N

    Deltat = T / N
    u = np.exp(sigma*(Deltat)**0.5)
    d = np.exp(-sigma*(Deltat)**0.5)
    q = (np.exp(r * Deltat) - d) / (u - d)
    #print("u: ", u, "d: ", d)

    #####

    #STOCK TREE

    jD1 = [i/Deltat for i in TD]
    jD = [np.ceil(i) for i in jD1]

    stock_value[0,0] = S0
    for j in range (1,N+1):
        stock_value[j, 0] = stock_value[j-1, 0]*d
        for i in range(1, j+1):
            stock_value[j, i] = stock_value[j-1, i-1]*u

    # This adjusts all stock prices for that j by the size of the dividend
    if j in jD: stock_value[j, :] *= (1-D)
    #####

    # Start at the last step number because we are going to be moving backwards from step number n to step number 0
    j = N
    for i in range(0, j+1):
        # Then, calculate the value of the option at that exact position within the binomial tree

        option_value[j, i] = np.maximum(K - stock_value[j, i], 0)

    # Now, Lets calculate the option value at each position (i) within the binomial tree at each previous step number (j) until t

```

```
for j in range(N-1, -1, -1):  
  
    # Then, create a FOR iteration on the position number (i), from the top position all the way down to the bottom position of 0  
    for i in range(j, -1, -1):  
  
        # Now, calculation the continuation value of the option at that specific position and step number  
        cv = np.exp(-r * Deltat) * (q*option_value[j+1,i+1] + (1-q)* option_value[j+1,i])  
        option_value[j, i] = cv  
    # RELAY OUTPUTS TO DICTIONARY  
    output = option_value[0,0]  
    Euro_result.append(output)  
  
return Euro_result, option_value
```

## • Rendleman and Bartter

In [ ]:



In [12]: `def rb_European(S0, K, T, r, sigma, N, D, TD):`

```

    # LIST TO SAVE RESULTS
    Euro_result = []

    # CREATE TWO DIMENSIONAL ARRAY OF SIZE [N+1,N+1] TO STORE ALL STEPS
    # option_value[N+1, N+1]
    stock_value = np.zeros ([N+1, N+1])
    option_value = np.zeros([N+1, N+1])

    # FOR LOOP STATEMENT: For a Binomial Tree from start_step to N

    Deltat = T / N
    u = np.exp((r-D-0.5*sigma**2)*Deltat + sigma*(Deltat)**0.5)
    d = np.exp((r-D-0.5*sigma**2)*Deltat - sigma*(Deltat)**0.5)
    q = (np.exp(r * Deltat) - d) / (u - d)
    #print("u: ", u, "d: ", d)

    #####

    #STOCK TREE

    jD1 = [i/Deltat for i in TD]
    jD = [np.ceil(i) for i in jD1]

    stock_value[0,0] = S0
    for j in range (1,N+1):
        stock_value[j, 0] = stock_value[j-1, 0]*d
        for i in range(1, j+1):
            stock_value[j, i] = stock_value[j-1, i-1]*u

    # This adjusts all stock prices for that j by the size of the dividend
    if j in jD: stock_value[j, :] *= (1-D)
    #####

    # Start at the last step number because we are going to be moving backwards from step number n to step number 0
    j = N
    for i in range(0, j+1):
        # Then, calculate the value of the option at that exact position within the binomial tree

        option_value[j, i] = np.maximum(K - stock_value[j, i], 0)

    # Now, Lets calculate the option value at each position (i) within the binomial tree at each previous step number (j) until t

```

```
for j in range(N-1, -1, -1):  
  
    # Then, create a FOR iteration on the position number (i), from the top position all the way down to the bottom position of 0  
    for i in range(j, -1, -1):  
  
        # Now, calculation the continuation value of the option at that specific position and step number  
        cv = np.exp(-r * Deltat) * (q*option_value[j+1,i+1] + (1-q)* option_value[j+1,i])  
        option_value[j, i] = cv  
    # RELAY OUTPUTS TO DICTIONARY  
    output = option_value[0,0]  
    Euro_result.append(output)  
  
return Euro_result, option_value
```

## • Leisen and Reimer

In [ ]:



```

In [13]: def lr_European(S0, K, T, r, sigma, N, D, TD):

    # LIST TO SAVE RESULTS
    Euro_result = []

    # CREATE TWO DIMENSIONAL ARRAY OF SIZE [N+1,N+1] TO STORE ALL STEPS
    # option_value[N+1, N+1]
    stock_value = np.zeros ([N+1, N+1])
    option_value = np.zeros([N+1, N+1])

    # FOR LOOP STATEMENT: For a Binomial Tree from start_step to N

    Deltat = T / N

    def hf(x,N):
        return 0.5 + np.sign(x) * np.sqrt(0.25 - (0.25 * np.exp(-(x/(N+1/3))**2 * (N + 1/6))))

    d1 = (np.log(S0/K) + (r-D+0.5*sigma**2)*Deltat)/(sigma*Deltat**0.5)
    d2 = (np.log(S0/K) + (r-D-0.5*sigma**2)*Deltat)/(sigma*Deltat**0.5)

    u = np.exp((r-D)*Deltat) * hf(d1,N)/hf(d2,N)

    d = (np.exp((r-D)*Deltat) - hf(d2,N)*u)/(1-hf(d2,N))

    q = (np.exp(r * Deltat) - d) / (u - d)
    #print("u: ", u, "d: ", d)

    #####

    #STOCK TREE

    jD1 = [i/Deltat for i in TD]
    jD = [np.ceil(i) for i in jD1]
    #print("Dividend points", jD)

    stock_value[0,0] = S0
    for j in range (1,N+1):
        stock_value[j, 0] = stock_value[j-1, 0]*d
        for i in range(1, j+1):
            stock_value[j, i] = stock_value[j-1, i-1]*u

```

```

# This adjusts all stock prices for that j by the size of the dividend
    if j in jD: stock_value[j, :] *= (1-D)
#####

# Start at the last step number because we are going to be moving backwards from step number n to step number 0
j = N
for i in range(0, j+1):
    # Then, calculate the value of the option at that exact position within the binomial tree

        option_value[j, i] = np.maximum(K - stock_value[j, i], 0)

# Now, Lets calculate the option value at each position (i) within the binomial tree at each previous step number (j) until t
for j in range(N-1, -1, -1):

    # Then, create a FOR iteration on the position number (i), from the top position all the way down to the bottom position of 0
        for i in range(j, -1, -1):

            # Now, calculation the continuation value of the option at that specific position and step number
                cv = np.exp(-r * Deltat) * (q*option_value[j+1,i+1] + (1-q)* option_value[j+1,i])
                option_value[j, i] = cv
# RELAY OUTPUTS TO DICTIONARY
output = option_value[0,0]
Euro_result.append(output)
#print(Euro_result)

return Euro_result, option_value

```

## Performance

In [ ]:



```
In [14]: y = 1
crr_arr = np.zeros(y)
rb_arr = np.zeros(y)
lr_arr = np.zeros(y)
x = 150

for i in range(x, x+1):

    crr_arr[i-x] = crr_European(S0, K, T, r, sigma, i, D, TD)[0][0]
    rb_arr[i-x] = rb_European(S0, K, T, r, sigma, i, D, TD)[0][0]
    lr_arr[i-x] = lr_European(S0, K, T, r, sigma, i, D, TD)[0][0]

print("crr_arr:", "\n", crr_arr, "\n", "\n"
      "rb_arr: ", "\n", rb_arr, "\n", "\n"
      "lr_arr: ", "\n", lr_arr)
```

```
crr_arr:
[10.12180064]
```

```
rb_arr:
[10.12263718]
```

```
lr_arr:
[2.95496845]
```

```
In [15]: def bsf(0, S, K, r, sigma, q, T, disFac):  
  
    0 = str(0)  
  
    d1 = (np.log(S/K) + (r - q + (0.5 * sigma**2)) * T) / (sigma * np.sqrt(T))  
    d2 = (np.log(S/K) + (r - q - (0.5 * sigma**2)) * T) / (sigma * np.sqrt(T))  
  
    if "Call" in 0:  
        result = disFac * (S * np.exp((r - q)*T) * ss.norm.cdf(d1, 0,1) - K * ss.norm.cdf(d2, 0, 1))  
    elif "Put" in 0:  
        result = disFac * (K * ss.norm.cdf(-d2, 0, 1) - S * np.exp((r - q)*T) * ss.norm.cdf(-d1, 0,1))  
    elif "Dp" in 0:  
        result = disFac * (ss.norm.cdf(-d2, 0, 1))  
    else:  
        result = disFac * (-ss.norm.cdf(d2, 0, 1))  
  
    return result
```

```
In [16]: import scipy.stats as ss
```

```
In [ ]:
```

```
In [21]: disFac = np.exp(-r*T)  
  
    bsm = bsf("Put", S0, K, r, sigma, 0, T, disFac)  
    bsm
```

```
Out[21]: 10.10930334495747
```

```
In [22]: crr_arr = np.zeros(951)
rb_arr = np.zeros(951)
lr_arr = np.zeros(951)
x = 50

for i in range(x, 1001):

    crr_arr[i-x] = crr_European(S0, K, T, r, sigma, i, D, TD)[0][0] - bsm
    rb_arr[i-x] = rb_European(S0, K, T, r, sigma, i, D, TD)[0][0] - bsm
    lr_arr[i-x] = lr_European(S0, K, T, r, sigma, i, D, TD)[0][0] - bsm

#print("crr_arr:", "\n", crr_arr, "\n", "\n",      "rb_arr: ", "\n", rb_arr, "\n", "\n",      "lr_arr: ", "\n", lr_arr)
```

```
In [31]: df = pd.DataFrame({"N": np.arange(50,1001), "CRR": crr_arr, "R&B": rb_arr, "L&R": lr_arr})
df
```

```
Out[31]:
```

	N	CRR	R&B	L&R
0	50	0.007624	-0.002123	-6.919757
1	51	0.028756	0.032954	-6.927965
2	52	0.003946	-0.006070	-6.933158
3	53	0.029440	0.033117	-6.943487
4	54	0.000443	-0.009828	-6.946154
...	...	...	...	...
946	996	0.001747	0.000330	-7.188441
947	997	-0.000287	0.001441	-7.188441
948	998	0.001728	0.000288	-7.188441
949	999	-0.000241	0.001462	-7.188441
950	1000	0.001708	0.000246	-7.188441

951 rows × 4 columns

```
In [32]: fig, ax1 = plt.subplots(figsize=(20,10))
ax1.plot(df['N'], df["CRR"],color="gray",linestyle="-",label="CRR")
ax1.plot(df['N'], df["R&B"],color="brown",linestyle="-",label="CRR")
ax1.set_ylabel("error CRR R&B")
ax1.legend();

ax2 = ax1.twinx()
ax2.plot(df['N'], df["L&R"], color="forestgreen", linestyle = "-.", label = "L&R", linewidth = 3)
ax2.set_ylabel("error R&B");
```



