```
# Import Necessary Libraries
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import math
         import operator as op
         from functools import reduce
         from decimal import Decimal
In [2]: #Inputs
        x0_val = 95 #Current Stock Price
         x0_val_ans4 = 105 #Assumed Stock Price for Answer 4
         k_val = 105 #Strike Price
         t_val1 = 3 #Steps in the Pricing Process (No. of Steps per Year x Years to Maturity) (Ans 1 Part a)
         t val2 = 2 #Steps in the Pricing Process (No. of Steps per Year x Years to Maturity) (Ans 2 Part a)
         r_val = 0 #Risk-free Rate per Price Step (Calculated as Risk-free Rate per year/No. of Steps per year)
         r_val_ans4 = 0.01 #Assumed Risk-free Rate per Price Step for Answer 4
        u_val = 1.14 #Up Movement per Step in Binomial Model
In [3]: #Combinatorics Function
        def nCr(n, r):
            r = min(r, n-r)
            numer = reduce(op.mul, range(n, n-r, -1), 1)
            denom = reduce(op.mul, range(1, r+1), 1)
            return (numer // denom)
        #Asset Price Binomial Tree Function:
In [4]:
         def Asset_Px_Binomial_Tree(t, u, d, x0):
            tree_holder = np.zeros([t+1, t+1])
            freq = np.zeros(t+1)
            d = 1/u
            for i in range(t+1):
                for j in range(i+1):
                    tree_holder[j,i]=x0*(d**j)*(u**(i-j))
                freq[i]= Decimal(nCr(t, i))
            tree_holder = pd.DataFrame(tree_holder.round(2))
            return (tree_holder, freq)
In [5]:
        #Dataframe of Terminal Asset Prices from Binomial Tree:
        def Terminal_Px_Extractor(binomial_tree):
            terminal_px = pd.DataFrame(np.vstack([binomial_tree.iloc[:,-1].values]).T,
                                      columns = ['X(w)'])
            return terminal_px
        #European Option Price Function
         def European_Option_Price(x0, k, u, t, r, option_type):
            d = 1/u
            p = (1 - d)/(u - d)
            if option_type == "call":
                * (1/((1+r)**t)), [0]+list(range(t+1)))
                return price
            elif option_type == "put":
                * (1/((1+r)**t)), [0]+list(range(t+1)))
                return price
                price = print("Input format error")
                return price
        #European Option Value Tree Function:
In [7]:
        def European_Optionval_Tree(stockpx_binomial_tree, k, u, t, r, option_type):
            european_optionval_tree = pd.DataFrame()
            if option_type == "call":
                for i in range(stockpx_binomial_tree.shape[0]):
                    for j in range(stockpx_binomial_tree.shape[1]):
                        if i <= j:
                            european_optionval_tree.at[i, j] = European_Option_Price(stockpx_binomial_tree.at[i, j],
                                                                                 k, u, (t-j), r, option_type)
                           european_optionval_tree.at[i, j] = 0
                return european_optionval_tree.round(3)
            elif option_type == "put":
                for i in range(stockpx_binomial_tree.shape[0]):
                    for j in range(stockpx_binomial_tree.shape[1]):
                            european_optionval_tree.at[i, j] = European_Option_Price(stockpx_binomial_tree.at[i, j],
                                                                                 k, u, (t-j), r, option_type)
                           european optionval tree.at[i, j] = 0
                return european_optionval_tree.round(3)
                european_optionval_tree = print("Input format error")
                return european_optionval_tree.round(3)
In [8]:
        #Code Test - Verifying Values Using Example on Pages 6-7 of M5 Notes
         call_m5_eg = European_Option_Price(100, 110, 1.2, 2, 0, "call")
         put_m5_eg = European_Option_Price(100, 110, 1.2, 2, 0, "put")
        call_m5_eg, put_m5_eg
Out[8]: (7.024793388429751, 17.024793388429746)
```

# Answer 1 European Call Option (N = 3)

## Part (a) European Call Option Price

```
In [9]: call_price = European_Option_Price(x0_val, k_val, u_val, t_val1, r_val, "call")
    print("The price of the European call option with the given parameters is", "{:.3f}".format(call_price))
The price of the European call option with the given parameters is 4.799
```

### Part (b) Call Option Value H(w) for Each Price Path

```
In [10]:
          #Stock Price Evolution
          stockpx_binomial_tree1, freq_stockpx1 = Asset_Px_Binomial_Tree(t_val1, u_val, (1/u_val), x0_val)
          stockpx_binomial_tree1
Out[10]:
         0 95.0 108.30 123.46 140.75
             0.0
                 83.33
                        95.00 108.30
             0.0
                   0.00
                        73.10 83.33
             0.0
                  0.00
                         0.00
                               64.12
          #Pay-off of the European Call Option for Each Price Path
In [11]:
          call_payoffs = Terminal_Px_Extractor(stockpx_binomial_tree1)
          call_payoffs['H(w)'] = [max(i - k_val, 0) for i in call_payoffs['X(w)']]
          print('H(w) (call) = Max(Terminal Price - Strike, 0)')
          call_payoffs
         H(w) (call) = Max(Terminal Price - Strike, 0)
             X(w) H(w)
Out[11]:
         0 140.75 35.75
         1 108.30 3.30
            83.33 0.00
             64.12 0.00
          #Value of the European Call Option at Each Node
          call_optionval_tree = European_Optionval_Tree(stockpx_binomial_tree1, k_val, u_val, t_val1, r_val, "call")
          call_optionval_tree
           #def myfunc(x):
               c1=np.triu(np.ones(call_optionval_tree.shape),1).astype(np.bool)
               c2=np.tril(np.ones(call_optionval_tree.shape),-1).astype(np.bool)
               col1='color:black
               col2='color:white
               col3='color:black'
               df1 = pd.DataFrame(np.select([c1,c2],[col1,col2],col3),columns=x.columns,index=x.index)
          #call_optionval_tree.style.apply(myfunc,axis=None)
Out[12]:
         0 4.799 9.449 18.460 35.75
                       1.542 3.30
         1 0.000 0.720
         2 0.000 0.000 0.000
                               0.00
         3 0.000 0.000 0.000 0.00
```

# Answer 2 European Put Option (N = 2)

## Part (a) European Put Option Price

95.00 10.0
 73.10 31.9

```
In [13]: put_price = European_Option_Price(x0_val, k_val, u_val, t_val2, r_val, "put")
    print("The price of the European put option with the given parameters is", "{:.3f}".format(put_price))
The price of the European put option with the given parameters is 14.031
```

## Part (b) Stock Price Evolution, Option Pay-off, Option Value, Hedging Strategy

```
#Stock Price Evolution
In [14]:
          stockpx_binomial_tree2, freq_stockpx2 = Asset_Px_Binomial_Tree(t_val2, u_val, (1/u_val), x0_val)
          stockpx_binomial_tree2
                     1
Out[14]:
         0 95.0 108.30 123.46
         1 0.0 83.33 95.00
             0.0
                  0.00 73.10
          #Pay-off of the European Put Option for Each Price Path
In [15]:
          put_payoffs = Terminal_Px_Extractor(stockpx_binomial_tree2)
          put_payoffs['H(w)'] = [max(k_val - i, 0) for i in put_payoffs['X(w)']]
          print('H(w) (put) = Max(Strike - Terminal Price, 0)')
          put_payoffs
         H(w) (put) = Max(Strike - Terminal Price, 0)
             X(w) H(w)
Out[15]:
         0 123.46
                    0.0
```

```
In [16]:
           #Value of the European Put Option at Each Node
           put_optionval_tree = European_Optionval_Tree(stockpx_binomial_tree2, k_val, u_val, t_val2, r_val, "put")
           put_optionval_tree
                        1 2
Out[16]:
          0 14.031 5.327 0.0
          1 0.000 21.670 10.0
          2 0.000 0.000 31.9
In [17]:
           #Table of Stock Price Xt(w) Evolution and Option Pay-off
           stockpx\_table = pd.DataFrame(columns = ['w', 'X0(w)', 'X1(w)', 'X2(w)', 'H(w) = Max(K - X2(w), 0)'])
           stockpx table.at[0, 'w'] = '(u,u)'
           stockpx_table.at[1, 'w'] = '(u,d)'
           stockpx_table.at[2, 'w'] = '(d,u)'
           stockpx\_table.at[3, 'w'] = '(d,d)'
           \#Stock\ Price\ at\ t=0
           for i in range(4):
               stockpx_table.at[i, 'X0(w)'] = stockpx_binomial_tree2.at[0,0]
           #Stock Price at t = 1
           for i in range(2):
               stockpx_table.at[i, 'X1(w)'] = stockpx_binomial_tree2.at[0,1]
           for i in range(2,4):
               stockpx_table.at[i, 'X1(w)'] = stockpx_binomial_tree2.at[1,1]
           \#Stock\ Price\ at\ t=2
           stockpx_table.at[0, 'X2(w)'] = stockpx_binomial_tree2.at[0,2]
           stockpx_table.at[1, 'X2(w)'] = stockpx_binomial_tree2.at[1,2]
stockpx_table.at[2, 'X2(w)'] = stockpx_binomial_tree2.at[1,2]
           stockpx_table.at[3, 'X2(w)'] = stockpx_binomial_tree2.at[2,2]
           #Put Option Pay-off H(w)
           for i in range(4):
               stockpx\_table.at[i, 'H(w) = Max(K - X2(w), 0)'] = max(k\_val - stockpx\_table.at[i, 'X2(w)'], 0)
           #stockpx_table.index += 1
           stockpx_table
Out[17]:
               W \times XO(w) \times X1(w) \times X2(w) + H(w) = Max(K - X2(w), 0)
                     95 108.3 123.46
          0 (u,u)
          1 (u,d)
                     95 108.3
                                   95
                                                          10
          2 (d,u)
                     95 83.33
                                   95
                                                          10
          3 (d,d)
                     95 83.33
                                 73.1
                                                        31.9
In [18]:
           #Table of Put Option Value Vt(w) Evolution
           putval_table = pd.DataFrame(columns = ['w', 'V0(w)', 'V1(w)', 'V2(w)'])
           putval_table.at[0, 'w'] = '(u,u)'
           putval_table.at[1, 'w'] = '(u,d)'
           putval_table.at[2, 'w'] = '(d,u)'
           putval_table.at[3, 'w'] = '(d,d)'
           \#Put\ Option\ Price\ at\ t=0
           for i in range(4):
               putval_table.at[i, 'V0(w)'] = put_optionval_tree.at[0,0].round(3)
           #Put Option Price at t = 1
           for i in range(2):
               putval table.at[i, 'V1(w)'] = put optionval tree.at[0,1].round(3)
           for i in range(2,4):
               \verb|putval_table.at[i, 'V1(w)'| = \verb|put_optionval_tree.at[1,1].round(3)|
           #Put Option Price at t = 2
           putval_table.at[0, 'V2(w)'] = put_optionval_tree.at[0,2].round(3)
           putval_table.at[1, 'V2(w)'] = put_optionval_tree.at[1,2].round(3)
           putval_table.at[2, 'V2(w)'] = put_optionval_tree.at[1,2].round(3)
                                'V2(w)'] = put_optionval_tree.at[2,2].round(3)
           putval_table.at[3,
           putval_table
Out[18]:
               w V0(w) V1(w) V2(w)
          0 (u,u) 14.031
                         5.327
          1 (u,d) 14.031 5.327
                                   10
          2 (d,u) 14.031 21.67
                                   10
          3 (d,d) 14.031 21.67
                                 31.9
           #Hedging Strategy at Each Node
In [19]:
           hedging_strat_table = pd.DataFrame(columns = ['w', 'Q1(w)', 'Q2(w)'])
           hedging_strat_table.at[0, 'w'] = '(u,u)'
           hedging_strat_table.at[1, 'w'] = '(u,d)'
hedging_strat_table.at[2, 'w'] = '(d,u)'
           hedging_strat_table.at[3, 'w'] = '(d,d)'
           for j in range(1, hedging_strat_table.shape[1]):
               for i in range(hedging_strat_table.shape[0]):
                   hedging_strat_table.iloc[i, j] = ((putval_table.iloc[i, j+1] - putval_table.iloc[i, j])/
                                                        (stockpx_table.iloc[i, j+1] - stockpx_table.iloc[i, j])).round(3)
           hedging_strat_table
               w Q1(w) Q2(w)
Out[19]:
          0 (u,u) -0.654 -0.351
          1 (u,d) -0.654 -0.351
          2 (d,u) -0.655
```

**3** (d,d) -0.655

# **Answer 3 Market Completeness Analysis**

# Part (a) (2x2) A Matrix Construction

## Part (b) Populating Matrix A

#### Part (c) Constructing and Populating the b Matrix

## Part (e) Solving for x

31.9

# **Answer 4 Put-Call Parity**

# Part (a) Comparing Portfolio with Long Call and Short Put to Portfolio with Long Stock and Short Bond

```
#Stock Price Binomial Tree with X0 = 105
In [24]:
          stockpx_binomial_tree3, freq_stockpx3 = Asset_Px_Binomial_Tree(t_val2, u_val, 1/u_val, x0_val_ans4)
          stockpx_binomial_tree3
Out[24]:
               0
                     1
                           2
         0 105.0 119.70 136.46
             0.0 92.11 105.00
             0.0 0.00 80.79
          #European Call Option Value Binomial Tree
In [25]:
          call_optionval_tree2 = European_Optionval_Tree(stockpx_binomial_tree3, k_val, u_val,
                                                         t val2, r val ans4, "call")
          call_optionval_tree2
                     1
Out[25]:
         0 6.734 14.554 31.46
         1 0.000 0.002 0.00
         2 0.000 0.000 0.00
          #European Put Option Value Binomial Tree
In [26]:
          put_optionval_tree2 = European_Optionval_Tree(stockpx_binomial_tree3, k_val, u_val,
                                                        t_val2, r_val_ans4, "put")
          put_optionval_tree2
```

```
    Out[26]:
    0
    1
    2

    0
    6.734
    0.000
    0.00

    1
    0.000
    12.765
    0.00

    2
    0.000
    0.000
    24.21
```

#### Part (a-i.) Value of Portfolio with 1 Long Call and 1 Short Put

#### Part (a-ii.) Value of Portfolio with 1 Long Stock and K dollars Borrowing (Short Bond)

# 1 0.000 -11.85 0.00 2 0.000 0.00 -24.21

#### Value of the Two Portfolios across the Three States

```
In [29]:
           #Portfolio Value for the 3 Final States
           portfolio_val_comparison_table = pd.DataFrame(columns = ['w', 'X2(w)', 'K', '(C-P)2(w)', '(S-B)2(w)'])
           portfolio_val_comparison_table.at[0, 'w'] = '(u,u)'
           portfolio_val_comparison_table.at[1, 'w'] = '(u,d)'
           portfolio_val_comparison_table.at[2, 'w'] = '(d,u)'
           portfolio_val_comparison_table.at[3, 'w'] = '(d,d)'
           #Terminal Stock Price
           portfolio\_val\_comparison\_table.at[0, 'X2(w)'] = stockpx\_binomial\_tree3.at[0,2].round(3)
           portfolio_val_comparison_table.at[1, 'X2(w)'] = stockpx_binomial_tree3.at[1,2].round(3)
portfolio_val_comparison_table.at[2, 'X2(w)'] = stockpx_binomial_tree3.at[1,2].round(3)
           portfolio_val_comparison_table.at[3, 'X2(w)'] = stockpx_binomial_tree3.at[2,2].round(3)
           #Strike Price
           for i in range(4):
               portfolio_val_comparison_table.at[i, 'K'] = k_val
           #Option Portfolio Value at t = 2
           portfolio\_val\_comparison\_table.at[0, '(C-P)2(w)'] = option\_portfolio\_val.at[0,2].round(3)
           portfolio_val_comparison_table.at[1, '(C-P)2(w)'] = option_portfolio_val.at[1,2].round(3)
           portfolio\_val\_comparison\_table.at[2, \ '(C-P)2(w)'] = option\_portfolio\_val.at[1,2].round(3)
           portfolio_val_comparison_table.at[3, '(C-P)2(w)'] = option_portfolio_val.at[2,2].round(3)
           \#Option Portfolio Value at t = 2
           portfolio_val_comparison_table.at[0, '(S-B)2(w)'] = stock_bond_portfolio_val.at[0,2].round(3)
                          _comparison_table.at[1,
                                                                         _bond_portfolio_val.at[1,2].round(3
           portfolio_val_comparison_table.at[2, '(S-B)2(w)'] = stock_bond_portfolio_val.at[1,2].round(3)
           portfolio_val_comparison_table.at[3, '(S-B)2(w)'] = stock_bond_portfolio_val.at[2,2].round(3)
           portfolio_val_comparison_table
Out[29]:
               w X2(w) K (C-P)2(w) (S-B)2(w)
          0 (u,u) 136.46 105
```

```
      Out[29]:
      W X2(W) K (C-P)2(W) (S-B)2(W)

      0 (u,u) 136.46 105 31.46 31.46

      1 (u,d) 105 105 0 0

      2 (d,u) 105 105 0 0

      3 (d,d) 80.79 105 -24.21 -24.21
```

# Part (b) Verifying Put-Call Parity for Options in Questions 1 and 2 (Taking N = 2)

```
Out[31]:
         0 3.952 8.542 18.46
         1 0.000 0.000 0.00
         2 0.000 0.000 0.00
In [32]: #Value of the European Put Option at Each Node
          put_optionval_tree = European_Optionval_Tree(stockpx_binomial_tree2, k_val, u_val, t_val2,
                                                       r_val_ans4, "put")
          put_optionval_tree
               0
                    1 2
Out[32]:
         0 13.755 5.274 0.0
         1 0.000 21.455 10.0
         2 0.000 0.000 31.9
In [33]: #Option Portfolio Value at Each Node
          option_portfolio_val2 = call_optionval_tree3 - put_optionval_tree
          option_portfolio_val2
Out[33]:
         0 -9.803 3.268 18.46
         1 0.000 -21.455 -10.00
         2 0.000
                   0.000 -31.90
          #Stock-Bond Portfolio Value at Each Node
In [34]:
          stock_bond_portfolio_val2 = pd.DataFrame()
          for j in range(stockpx_binomial_tree2.shape[1]):
              for i in range(stockpx_binomial_tree2.shape[0]):
                  if i <=j:
                      stock_bond_portfolio_val2.at[i,j] = (stockpx_binomial_tree2.at[i,j]
                      - k_val/((1+r_val_ans4)**(t_val2-j))).round(3)
                  else:
                      stock_bond_portfolio_val2.at[i,j] = 0
          stock_bond_portfolio_val2
                      1
                           2
Out[34]:
         0 -7.931 4.34 18.46
         1 0.000 -20.63 -10.00
         2 0.000 0.00 -31.90
          #Difference between Option Portfolio Value and Stock-Bond Portfolio Value
In [35]:
          diff = option_portfolio_val2 - stock_bond_portfolio_val2
          diff.round(3)
               0
Out[35]:
         0 -1.872 -1.072 0.0
         1 0.000 -0.825 0.0
         2 0.000 0.000 0.0
```