```
# Import necessary libraries:
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
```

### Answer 1 Binomial Tree for N = 6

```
In [2]: # Define u and d:
         u_{total} = 1.10 + 4/100
         d_total = 1/u_total
         S_0 = 100 #assume original stock px is 100
In [3]:
         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_px_binomial_tree takes 4 parameters to populate a binomial tree:
In [4]:
         # N - number of nodes in the tree
         # u - upper movement coef
         # d - downward movement coef
         # S0 - initial stock px
         def asset_px_binomial_tree(N, u, d, S0):
             tree_holder = np.zeros([N+1, N+1])
             freq = np.zeros(N+1)
             u_{increment} = u**(1/N)
             d_increment=1/u_increment
             # populate the tree:
             for i in range(N+1):
                 for j in range(i+1):
                     tree_holder[j,i]=S0*(d_increment**j)*(u_increment**(i-j))
                 freq[i]= Decimal(ncr(N, i))
             return (tree_holder, freq)
         def terminal_px_extractor(binomial_tree):
           size = binomial tree.shape[0]
           terminal_px_holder=[]
           for i in range(size):
             temp = binomial_tree[i,-1]
             terminal_px_holder.append(temp)
           return terminal_px_holder
```

#### Part (a) Binomial Tree for N = 6

```
In [5]: task_1, freq_task1 = asset_px_binomial_tree(6, u_total, d_total, S_0)

# Use int astype for better visability
print('Binomial tree for an asset price:\n', np.matrix(task_1.astype(int)))

Binomial tree for an asset price:
[[100 102 104 106 109 111 114]
[ 0 97 100 102 104 106 109]
[ 0 0 95 97 100 102 104]
[ 0 0 0 93 95 97 99]
[ 0 0 0 0 93 95 97 99]
[ 0 0 0 0 0 89 91]
[ 0 0 0 0 0 0 88 91]
```

# Part (b) Terminal Values

```
In [6]: #Terminal Values
    task_1_df = pd.DataFrame(task_1)

terminal_val_task1 = pd.DataFrame(np.vstack([task_1_df.iloc[:,-1].values,freq_task1]).T, columns = ['Terminal Value', 'Frequency'])
    terminal_val_task1
```

Out[6]:		Terminal Value	Frequency
	0	114.000000	1.0
	1	109.128093	6.0
	2	104.464393	15.0
	3	100.000000	20.0
	4	95.726398	15.0
	5	91.635432	6.0
	6	87.719298	1.0

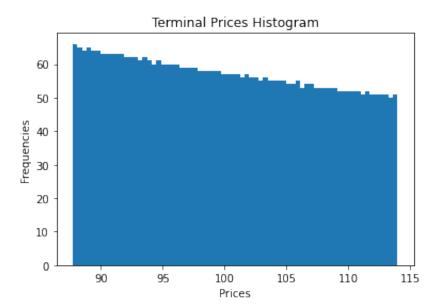
# Answer 2 Binomial Tree for N = 4000

```
In [7]: task_2, freq_task2 = asset_px_binomial_tree(4000, u_total, d_total, S_0)
    task_2_TV = terminal_px_extractor(task_2)
```

### Part (a) Histogram of Terminal Prices

```
In [8]: plt.hist(task_2_TV, bins=70)
    plt.title('Terminal Prices Histogram')
    plt.xlabel('Prices')
    plt.ylabel('Frequencies')
```

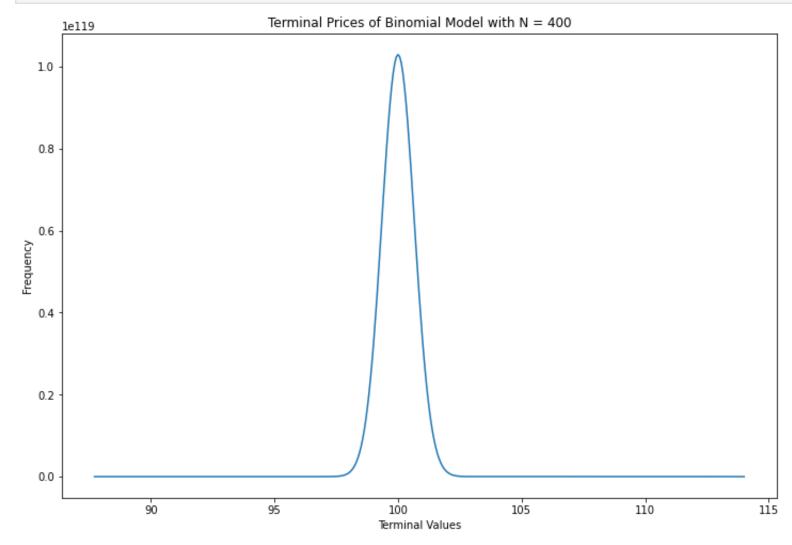
Out[8]: Text(0, 0.5, 'Frequencies')



```
In [9]: task_2_1000, freq_task2_1000 = asset_px_binomial_tree(400, u_total, d_total, S_0)
    task_2_1000 = pd.DataFrame(task_2_1000)
    terminal_val_task2 = pd.DataFrame(np.vstack([task_2_1000.iloc[:,-1].values,freq_task2_1000]).T, columns = ['Terminal_Value', 'Frequency'])
```

### Part (b) Statistical Distribution of Prices Generated by the Binomial Model

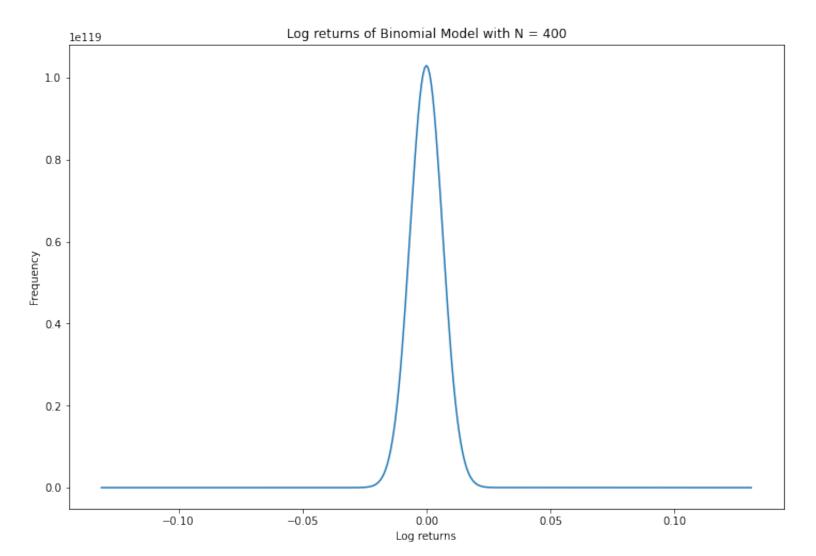
```
fig, ax = plt.subplots(figsize = (12,8))
    ax.plot(terminal_val_task2['Terminal Value'].values, terminal_val_task2['Frequency'].values, label = 'Terminal Values')
    ax.set_xlabel('Terminal Values')
    ax.set_ylabel('Frequency')
    ax.set_title('Terminal Prices of Binomial Model with N = 400');
```



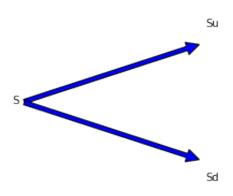
## Part (c) Statistical Distribution of Log Returns

```
In [11]: terminal_val_task2['Log Returns'] = np.log(terminal_val_task2['Terminal Value']/100)

fig, ax = plt.subplots(figsize = (12,8))
   plt.plot(terminal_val_task2['Log Returns'].values, terminal_val_task2['Frequency'].values, label = 'Log returns')
   ax.set_xlabel('Log returns')
   ax.set_ylabel('Frequency')
   ax.set_title('Log returns of Binomial Model with N = 400');
```



```
np.log(task_2[1,1]/task_2[0,0])
In [12]:
Out[12]: -3.275706560166144e-05
In [13]:
          np.log(task_2[0,1]/task_2[0,0])
Out[13]: 3.275706560164976e-05
          task 3, freq task3 = asset px binomial tree(400, u total, d total, S 0)
In [14]:
          task_3_TV = terminal_px_extractor(task_3)
          plt.hist(task_3_TV, bins=10)
In [15]:
Out[15]: (array([46., 43., 43., 41., 41., 39., 38., 37., 37., 36.]),
           array([ 87.71929825, 90.34736842, 92.9754386 , 95.60350877,
                  98.23157895, 100.85964912, 103.4877193 , 106.11578947,
                 108.74385965, 111.37192982, 114.
                                                         ]),
           <a list of 10 Patch objects>)
          40
          30
          20
          10
                  90
                           95
                                   100
                                            105
                                                     110
                                                              115
          plt.xlim(0,1)
          plt.figtext(0.18,0.5,'S')
          plt.figtext(0.6,0.5+0.25,'Su')
```



plt.figtext(0.6,0.5-0.25,'Sd')

plt.axis('off')
plt.show()

 $\texttt{plt.annotate('',xy=(0.6,0.5+0.25), xytext=(0.1,0.5), arrowprops=dict(facecolor='b',shrink=0.01)) } \\ \texttt{plt.annotate('',xy=(0.6,0.5-0.25), xytext=(0.1,0.5), arrowprops=dict(facecolor='b',shrink=0.01))} \\ \texttt{plt.annotate('',xy=(0.6,0.5-0.25), xytext=(0.1,0.5), arrowprops=dict(facecolor='b',shrink=0.01))} \\ \texttt{plt.annotate('',xy=(0.6,0.5+0.25), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=(0.1,0.5), xytext=$ 

<Figure size 432x288 with 0 Axes>

Tn [ 1: