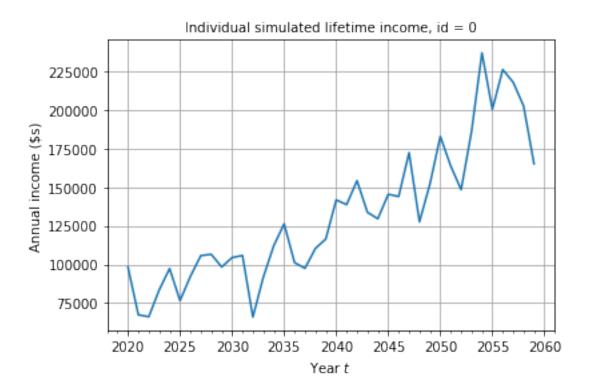
Assignment3 part2

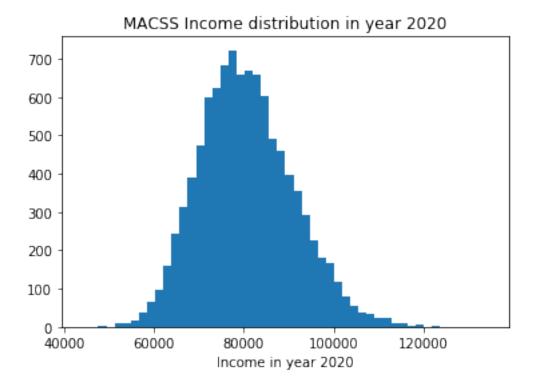
October 23, 2018

```
In [1]: import numpy as np
        import matplotlib.pyplot as plt
        from matplotlib.ticker import MultipleLocator
        from scipy import stats
        from numpy import cumsum
        def normal_sim(p):
           Requires a simulation profile, p, structures as a dictionary
                'inc0' : 80000
                                            #average starting income
                ' q '
                         : 0.025
                                           #rowth rate
                'rho'
                          : 0.4
                                            #positive deopendence
                         : [2020, 2059] #year t (t >= 2020)
                'num_draws': 10000
                                            #simulations
                'mean' : 0
                                            #mean
                'sd' : 0.13.
                                            #standard deviation
            }
            11 11 11
            #set random seed
           np.random.seed(524)
           normal_errors = np.random.normal(p['mean'], p['sd'], (p['t'], p['num_draws']))
            \#create \ a \ matrix \ of \ dim \ ((t - 2019), \ num\_draws)
            ln_inc_mat = np.zeros((p['t'] - 2019, p['num_draws']))
            #fill in the first row
            ln_inc_mat [0, :] = np.log(p['inc0']) + normal_errors[0, :]
            #loop and apply model
            for yr in range(1, p['t'] - 2019):
                 ln_inc_mat[yr, :] = (1 - p['rho']) * (np.log(p['incO']) + p['g'] * yr)
                     + p['rho'] * ln_inc_mat[ yr - 1, :] + normal_errors[yr, :]
```

```
income_mat = np.exp(ln_inc_mat)
            return income_mat
In [2]: sim_profile = {'inc0': 80000,
                       'g': 0.025,
                       'rho': 0.4,
                       't': 2059,
                       'num draws': 10000,
                       'mean': 0,
                       'sd': 0.13
                      }
        income_mat = normal_sim(sim_profile)
        income_mat
Out[2]: array([[ 66409.15585396, 98274.13534194, 101939.81109509, ...,
                 98720.39690442, 72404.51636886, 68710.32820307],
               [80020.53020329, 67383.19350738, 84557.85626308, ...,
                 68247.7770509 , 74518.33613244, 80555.96068584],
               [ 75805.26636606, 66134.42494243, 91458.20304692, ...,
                 67268.53350159, 90012.42673528, 80645.62355527],
               [272690.56519108, 217821.73027242, 184724.24512469, ...,
                159922.45424852, 253961.68337673, 209741.55004062],
               [231539.17420799, 202509.15149494, 197955.96626493, ...,
                199502.43481758, 210951.71828579, 205420.27946389],
               [197895.95201384, 165115.10025278, 172644.86927513, ...,
                248654.44847819, 234237.14656466, 221566.29879732]])
In [3]: # %matplotlib inline
       p = sim_profile
        year_vec = np.arange(2020, p['t']+1)
        individual = 1
        fig, ax = plt.subplots()
       plt.plot(year_vec, income_mat[:, individual])
       minorLocator = MultipleLocator(1)
        ax.xaxis.set_minor_locator(minorLocator)
        plt.grid(b=True, which='major', color='0.65', linestyle='-')
       plt.title('Individual simulated lifetime income, id = 0', fontsize=10)
       plt.xlabel(r'Year $t$')
       plt.ylabel(r'Annual income (\$s)')
Out[3]: Text(0,0.5,'Annual income (\\$s)')
```

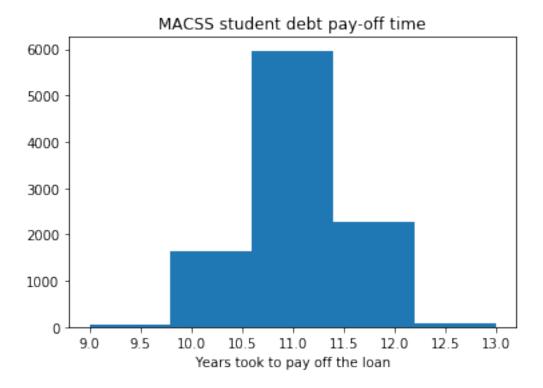


Out[4]: Text(0.5,1,'MACSS Income distribution in year 2020')



From the above calculation, we can see that 4.17% of my class will earn more than \$100000 in the first year out of the program and 15.12% will earn less than \$70000. From the graph above, the distribution is bell curved and normally distributed with a mean of 80000.

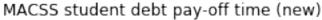
Out[8]: Text(0.5,1,'MACSS student debt pay-off time')

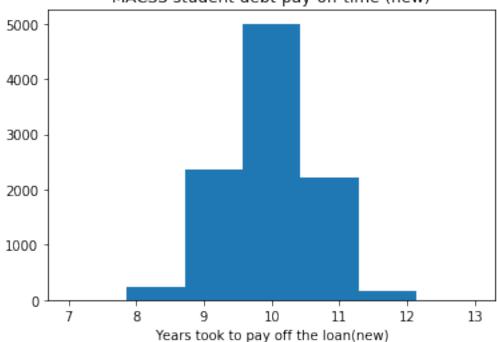


Out[9]: 0.1678

In 16.78% of the simulations, I can pay off the debt in ten years.

```
72778.88084775, 81644.3347736 , 90400.57899801],
                [82955.30101689, 69396.06916251, 106035.55593099, ...,
                  70956.3661129 , 103848.93176006, 89949.09077038],
                [338309.11761165, 252187.52025149, 203293.03644369, ...,
                 168361.21927259, 308250.29858492, 240024.49205936],
                [271061.07048342, 227502.32436192, 220836.5697397, ...,
                 223095.32811759, 239983.96514044, 231788.44418303],
                [219057.46748997, 172865.33333479, 183245.71710131, ...,
                 295275.8618388 , 273090.00167035, 253934.86273481]])
In [11]: a2 = np.zeros((40, 10000))
         for i in range(10000):
             a2[:, i] = 0.1 * cumsum(income_mat2[:, i])
        b2 = np.zeros((10000, 1))
        for i in range(10000):
             b2[i, 0] = np.sum(a2[:, i] < 95000) + 1
        stats.describe(b2)
Out[11]: DescribeResult(nobs=10000, minmax=(array([7.]), array([13.])), mean=array([9.9733]),
In [12]: plt.hist(b2, bins = 7)
        plt.xlabel("Years took to pay off the loan(new)")
        plt.title("MACSS student debt pay-off time (new)")
Out[12]: Text(0.5,1,'MACSS student debt pay-off time (new)')
```





```
In [13]: num2 = (b2 \le 10).sum()
num2/10000
```

Out[13]: 0.7602

In the new simulation, in 76.02% of the simulations, I can pay off the debt in ten years.