## Assignment#3

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## 1 Assignment 3

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Due Wednesday, Oct.24 at 11:30 AM

## 1.0.3 1. Simulation in Sociology, Moretti (2002)

See the attached PDF.

## 1.0.4 2. Simulating your income

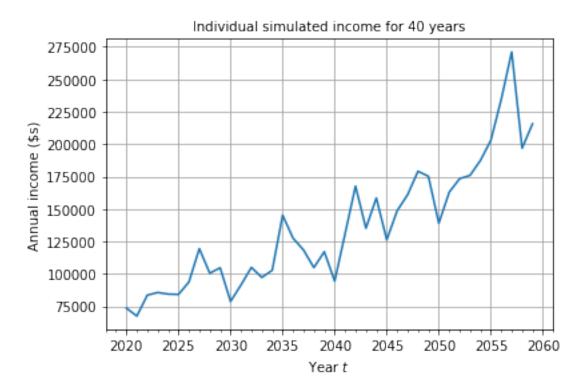
(a) Assume you will work for 40 years after you graduate (2020 to 2059). Simulate 10,000 different realizations of your lifetime income.

```
In [1]: # import initial packages
        import numpy as np
        import matplotlib.pyplot as plt
        from matplotlib.ticker import MultipleLocator
        import pandas as pd
In [2]: def lognormal_income_sim(p):
            np.random.seed(524)
            ln_errors_mat = np.zeros((p['num_draws'], p['work_years']))
            for draw in range(p['num_draws']):
                income_errors = np.random.lognormal(p['mean'], p['sd'], p['work_years'])
                ln_errors_mat[draw, :] = np.log(income_errors)
            ln_errors_mat = np.transpose(ln_errors_mat)
            #create a matrix of dim(lf_years, num_draws)
            ln_income_mat = np.zeros((p['work_years'], p['num_draws']))
            #fill the matrix
            ln_income_mat[0, :] = np.log(p['inc0']) + ln_errors_mat[0, :]
            #loop and apply model
```

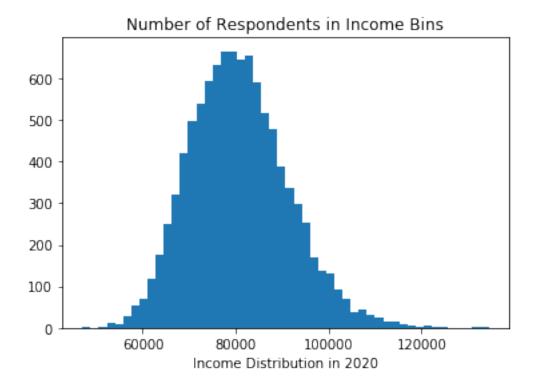
```
for yr in range(1, p['work_years']):
                                      ln_income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + p['gr'] * yr) + 0.4 * income_mat[yr, :] = (1 - 0.4) * (np.log(p['inc0']) + (np.l
                            income_mat = np.exp(ln_income_mat)
                            return income_mat
In [3]: simulation_profile = {
                            'mean' : 0,
                                                                                 # mean for ln(errors)
                            'sd' : 0.13,
                                                                                 # standard deviation for ln(errors)
                            'inc0': 80000,  # starting income
'gr': 0.025,  # growth rate
'work_years': 40,  # working years
'num_draws': 10000,  # simulations
                            'st_year' : 2020
                                                                                 # start year
                   }
                   income_mat = lognormal_income_sim(simulation_profile)
                   print(income_mat)
[[ 66409.15585396 87090.02361606 77964.9128787 ... 82615.14768903
      73060.280769 96185.73700534]
  [ 92599.54181011 110753.4836433 75396.13433933 ... 91033.55488888
      72981.70901205 93862.20664804]
  [111373.07206917 94963.7575539 79613.59599191 ... 89798.18450708
      88417.6480529 101046.98628708]
  [184381.71271416 225888.14811901 207083.55365427 ... 256618.44418765
    192277.34637423 220022.334453447
  [232334.57698127 193323.26525541 178471.60929258 ... 294250.66877047
    167032.88455955 234668.51878321]
   [224611.55424288 233696.52700494 190202.35706111 ... 290792.67259199
    196285.97182101 233683.64036871]]
      Plot one of the lifetime income paths.
In [4]: %matplotlib inline
                   p = simulation_profile
                   year_vec = np.arange(p['st_year'], p['st_year'] + p['work_years'])
                   individual = 500
                   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 income for 40 years', fontsize = 10)
                  plt.xlabel(r'Year $t$')
```

plt.ylabel(r'Annual income (\\$s)')

Out[4]: Text(0,0.5,'Annual income (\\\$s)')



(b) Plot a histogram with 50 bins of year t = 2020 initial income for each of the 10,000 simulations.

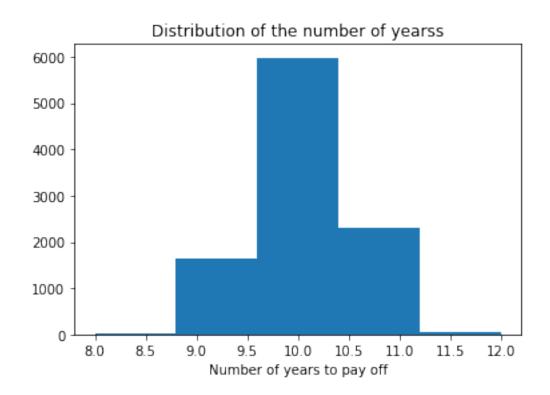


From my simulation, around 4.51% of our class will earn more than 100,000 dollars in the first year out of this program. And about 15.19% of our class will earn less than 70,000 dollars in the first year. I think this distribution is roughly normal distribution, with a little bit skewed. We may use larger number of simulation or more bins to make it more normally ditributed.

(c) Plot the histogram of how many years it takes to pay off the loan in each of your 10,000 simulations. In what percent of the simulations are you able to pay off the loan in 10 years (on or before t = 2029).

```
In [14]: years_to_pay_off = []
         for i in range(p['num_draws']):
             year = 0
             amount = 0
             income = income_mat[:, i]
             for inc in income:
                 amount += inc * 0.1
                 if amount < 95000:
                     year = year + 1
                 else:
                     break
             years_to_pay_off.append(year)
         years_to_pay_off = np.asarray(years_to_pay_off)
         plt.hist(years_to_pay_off, bins = 5)
         plt.xlabel("Number of years to pay off")
         plt.title("Distribution of the number of yearss")
         from scipy import stats
         percent = stats.percentileofscore(years_to_pay_off, 10.5)
         print(percent, "percent of the simulations will be able to pay off the loan in 10 year
```

76.46 percent of the simulations will be able to pay off the loan in 10 years.



(d) Plot the new histogram of how many years it takes to pay off your loan of \$95,000 in your new 10,000 simulations with the new standard deviation and the new average initial salary. In what percent of the simulations are you able to pay off the loan in 10 years (on or before t = 2029)?

```
In [10]: simulation_profile_new = {
              'mean' : 0,
                                       # mean for ln(errors)
                                 # standard deviation for ln(errors)
# starting income
              'sd' : 0.17,
              'inc0': 90000,
              'gr' : 0.025,
                                     # growth rate
             'work_years' : 40,  # working years
'num_draws' : 10000,  # simulations
              'st_year' : 2020
                                     # start year
         }
         income_mat_new = lognormal_income_sim(simulation_profile_new)
In [16]: years_to_pay_off_new = []
         for i in range(p['num_draws']):
             year = 0
             amount = 0
             income_new = income_mat_new[:, i]
             for inc in income_new:
                  amount += inc * 0.1
                  if amount < 95000:
                      year = year + 1
                  else:
                      break
             years_to_pay_off_new.append(year)
         years_to_pay_off_new = np.asarray(years_to_pay_off_new)
         plt.hist(years_to_pay_off_new, bins = 6)
         plt.xlabel("Number of years to pay off")
         plt.title("Distribution of the number of yearss")
         from scipy import stats
         percent = stats.percentileofscore(years_to_pay_off_new, 10.5)
         print("In the new circumstance,", percent, "percent of the simulations will be able to
```

In the new circumstance, 98.16 percent of the simulations will be able to pay off the loan in

