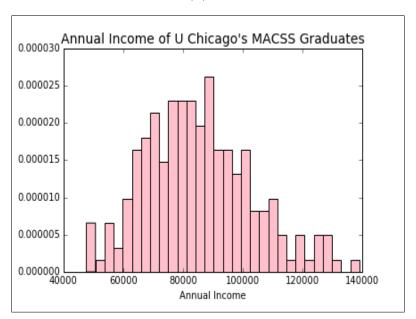
Problem Set #3

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Problem 1: Some income data, lognormal distribution, and GMM

Part (a). The histogram is shown in Figure 1.

Figure 1: Histogram of the MACSS Graduates' Income of 1(a)



Part (b). The plot of estimated lognormal PDF is shown in Figure 2. The value of my GMM criterion function at the estimated parameter values is: 1.84965511266e-14.

The data moments are: mean = 85276.8236063, standard deviation = 17992.542128. The model moments are: mean = 85276.81387602663, standard deviation = 17992.5414622. The comparison of data moments and model moments shows great similarity, which means the GMM estimation is good.

Part (c). The plot of the estimated lognormal PDF against the PDF from part(b) and the histogram from part(a) is shown in Figure 3.

The value of my GMM criterion function at the estimated parameter values is: 0.765715166926.

The data moments are: mean = 85276.8236063, standard deviation = 17992.542128. The model moments are: mean = 63620.09639498967, standard deviation = 21224.5908045. Given the values of data moments, model moments and the plot, the new GMM estimation with optimal weighting matrix does not perform very well. The model moments are less close to the data moments.

0.000030 Annual Income of U Chicago's MACSS Graduates
0.000025
0.000020
0.000015

Figure 2: Estimated Lognormal PDF by GMM of 1(b)

Part (d). The value of my GMM criterion function at the estimated parameter values is: 0.23818173309.

Annual Income

80000 100000 120000 140000 160000

The data moments are:

0.000005

0.000000

Percent of individuals who earn less than \$75,000 = 0.3

20000

40000

Percent of individuals who earn between \$75,000 and \$100,000 = 0.5

Percent of individuals who earn more than \$100,000 = 0.2

The model moments are:

Percent of individuals who earn less than \$75,000 = 0.29927245367786354

Percent of individuals who earn between \$75,000 and \$100,000 = 0.4980574552151655

Percent of individuals who earn more than \$100,000 = 0.19966278324796635

The data moments and model moments are very close, which means that the GMM estimation performs well.

Part (e). The value of my GMM criterion function at the estimated parameter values is: -5.05831569339e-10.

The data moments are:

Percent of individuals who earn less than \$75,000 = 0.3

Percent of individuals who earn between \$75,000 and 100,000 = 0.5

Percent of individuals who earn more than \$100,000 = 0.2

The model moments are:

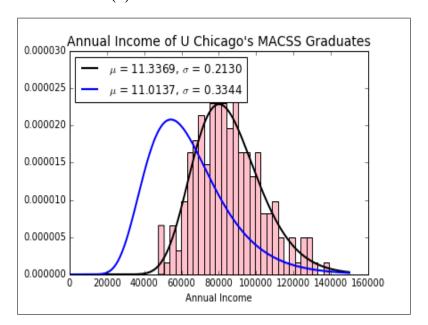
Percent of individuals who earn less than \$75,000 = 0.03709302071777423

Percent of individuals who earn between \$75,000 and \$100,000 = 0.38825270349916574

Percent of individuals who earn more than \$100,000 = 0.5550946448869221

The data moments and model moments vary a lot, which means the estimation with the optimal weighting matrix does not provide estimation close enough to the data.

Figure 3: Estimated Lognormal PDF by GMM of 1(c)



Part (f). According to the comparisons between data moments and model moments, as well as the plots, the GMM estimations in part(b) and part(d) seem to fit the data best, for their estimations are close to the data. Their estimated parameters are very close, thus it is hard to say which one is better. If I have to choose one, I would say the estimation in part(e) is better, since it includes more detailed consideration to the data. The other two estimations with weighting matrix do not provide very accurate estimation to the data.

Problem 2: Linear regression and GMM

Part (a). The value of estimated parameter values by GMM is: 0.00182128981773. And the estimated parameters are:

$$\begin{split} \beta_0^{GMM} &= 0.251644726617, \beta_1^{GMM} = 0.012933451003, \\ \beta_2^{GMM} &= 0.400501185497, \beta_3^{GMM} = -0.00999169581195 \end{split}$$

Figure 4: Estimated Lognormal PDF by GMM of 1(d)

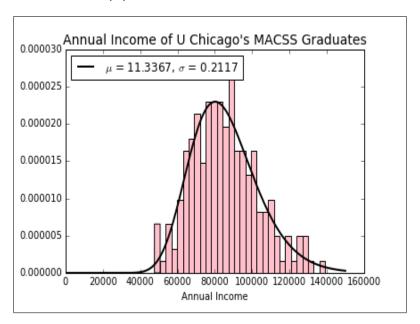


Figure 5: Estimated Lognormal PDF by GMM of 1(e)

