Open Source Macroeconomics Laboratory Boot Camp Week 5-6 Econ Homework

Ruby Zhang

2017

DSGE

Exercise 1

Let us guess and verify that the policy function for the Brock and Mirman model is $k_{t+1} = \phi(k_t, z_t) = Ae^{z_t}k_t^{\alpha}$ by using the Euler equation for the Brock and Mirman model:

$$\frac{1}{e^{z_t}k_t^{\alpha} - k_{t+1}} = \beta E_t \left\{ \frac{\alpha e^{z_{t+1}} k_{t+1}^{\alpha - 1}}{e^{z_{t+1}} k_{t+1}^{\alpha} - k_{t+2}} \right\}$$

Left side of the equation:

$$\frac{1}{e^{z_t}k_t^{\alpha} - Ae^{z_t}k_t^{\alpha}} = \frac{1}{e^{z_t}k_t^{\alpha}(1-A)}$$

Right side of the equation (note that the expected $z_{t+1} = \rho z_t$):

$$\beta E_t \left\{ \frac{\alpha e^{z_{t+1}} k_{t+1}^{\alpha - 1}}{e^{z_{t+1}} k_{t+1}^{\alpha} - k_{t+2}} \right\} = \beta \left(\frac{\alpha e^{\rho z_t} (A e^{z_t} k_t^{\alpha})^{\alpha - 1}}{e^{\rho z_t} (A e^{z_t} k_t^{\alpha})^{\alpha} (1 - A)} \right)$$
$$= \frac{\alpha \beta}{A e^{z_t} k_t^{\alpha} (1 - A)}$$

Since the left-hand side and right-hand side must equal each other, $\frac{\alpha\beta}{A} = 1 \implies A = \alpha\beta$. Therefore the policy function is $\phi(k_t, z_t) = \alpha\beta e^{z_t} k_t^{\alpha}$.

Exercise 2

$$\ell_t = L_t$$

$$k_t = K_t$$

$$w_t = W_t$$

$$r_t = R_t$$

$$c_t = (1 - \tau)[w_t \ell_t + (r_t - \delta)k_t] + k_t + T_t - k_{t+1}$$

$$\frac{1}{c_t} = \beta E_t \{ \frac{1}{c_{t+1}} [(r_{t+1} - \delta)(1 - \tau) + 1] \}$$

$$\frac{1}{1 - \ell_t} = \frac{1}{c_t} w_t (1 - \tau)$$

$$r_t = \alpha e^{z_t} (\frac{\ell_t}{k_t})^{1 - \alpha}$$

$$w_t = (1 - \alpha) e^{z_t} (\frac{k_t}{\ell_t})^{\alpha}$$

$$T_t = \tau [w_t \ell_t + (r_t - \delta)k_t]$$

$$z_t = (1 - \rho_z) \bar{z} + \rho_z z_{t-1} + \epsilon_t^z; \qquad \epsilon_t^z \sim \text{i.i.d.}(0, \sigma_z^2)$$

Exercise 3

$$\ell_t = L_t$$

$$k_t = K_t$$

$$w_t = W_t$$

$$r_t = R_t$$

$$c_t = (1 - \tau)[w_t \ell_t + (r_t - \delta)k_t] + k_t + T_t - k_{t+1}$$

$$\frac{1}{c_t^{\gamma}} = \beta E_t \left\{ \frac{1}{c_{t+1}^{\gamma}} [(r_{t+1} - \delta)(1 - \tau) + 1] \right\}$$

$$\frac{1}{1 - \ell_t} = \frac{1}{c_t^{\gamma}} w_t (1 - \tau)$$

$$r_t = \alpha e^{z_t} (\frac{\ell_t}{k_t})^{1 - \alpha}$$

$$w_t = (1 - \alpha)e^{z_t} (\frac{k_t}{\ell_t})^{\alpha}$$

$$T_t = \tau [w_t \ell_t + (r_t - \delta)k_t]$$

$$z_t = (1 - \rho_z)\bar{z} + \rho_z z_{t-1} + \epsilon_t^z; \qquad \epsilon_t^z \sim \text{i.i.d.}(0, \sigma_z^2)$$

Exercise 4

$$\ell_{t} = L_{t}$$

$$k_{t} = K_{t}$$

$$w_{t} = W_{t}$$

$$r_{t} = R_{t}$$

$$c_{t} = (1 - \tau)[w_{t}\ell_{t} + (r_{t} - \delta)k_{t}] + k_{t} + T_{t} - k_{t+1}$$

$$\frac{1}{c_{t}^{\gamma}} = \beta E_{t} \{ \frac{1}{c_{t+1}^{\gamma}} [(r_{t+1} - \delta)(1 - \tau) + 1] \}$$

$$\frac{a}{(1 - \ell_{t})^{\xi}} = \frac{1}{c_{t}^{\gamma}} w_{t} (1 - \tau)$$

$$r_{t} = \frac{\alpha e^{z_{t}} [\alpha k_{t}^{\eta} + (1 - \alpha)\ell_{t}^{\eta}]^{\frac{1}{\eta} - 1} k_{t}^{\eta - 1}}{\eta}$$

$$w_{t} = \frac{(1 - \alpha)e^{z_{t}} [\alpha k_{t}^{\eta} + (1 - \alpha)\ell_{t}^{\eta}]^{\frac{1}{\eta} - 1} \ell_{t}^{\eta - 1}}{\eta}$$

$$T_{t} = \tau [w_{t}\ell_{t} + (r_{t} - \delta)k_{t}]$$

$$z_{t} = (1 - \rho_{z})\bar{z} + \rho_{z}z_{t-1} + \epsilon_{t}^{z}; \qquad \epsilon_{t}^{z} \sim \text{i.i.d.}(0, \sigma_{z}^{2})$$

Exercise 5

$$\begin{split} &\ell_{t} = L_{t} = 1 \\ &k_{t} = K_{t} \\ &w_{t} = W_{t} \\ &r_{t} = R_{t} \\ &c_{t} = (1 - \tau)[w_{t}\ell_{t} + (r_{t} - \delta)k_{t}] + k_{t} + T_{t} - k_{t+1} \\ &\frac{1}{c_{t}^{\gamma}} = \beta E_{t} \{ \frac{1}{c_{t+1}^{\gamma}} [(r_{t+1} - \delta)(1 - \tau) + 1] \} \\ &r_{t} = \alpha (\frac{\ell_{t}e^{z_{t}}}{k_{t}})^{1 - \alpha} \\ &w_{t} = (1 - \alpha)e^{z_{t}} (\frac{k_{t}}{\ell_{t}e^{z_{t}}})^{\alpha} \\ &T_{t} = \tau [w_{t}\ell_{t} + (r_{t} - \delta)k_{t}] \\ &z_{t} = (1 - \rho_{z})\bar{z} + \rho_{z}z_{t-1} + \epsilon_{t}^{z}; \qquad \epsilon_{t}^{z} \sim \text{i.i.d.}(0, \sigma_{z}^{2}) \end{split}$$

Making all the necessary substitutions and letting the variables be represented by their steady state values, we then have the following equations (note that the stochastic shocks are always zero so the expectation disappears):

$$\bar{c} = \bar{w} + (\bar{r} - \delta)\bar{k} \tag{1}$$

$$\frac{1}{\overline{c}^{\gamma}} = \beta \frac{1}{\overline{c}^{\gamma}} [(\overline{r} - \delta)(1 - \tau) + 1] \tag{2}$$

$$\bar{r} = \alpha \left(\frac{e^{\bar{z}}}{\bar{k}}\right)^{1-\alpha} \tag{3}$$

$$\bar{w} = (1 - \alpha)e^{\bar{z}}(\frac{\bar{k}}{e^{\bar{z}}})^{\alpha} \tag{4}$$

We solve for \bar{k} by solving for \bar{r} from equation (2) since \bar{c} cancels out, then substituting that into equation (4). We can then get the steady state output and investment from $\bar{Y} = \bar{k}^{\alpha} (e^{\bar{z}})^{1-\alpha}$ and

 $\bar{I} = \delta \bar{k}$:

$$\bar{r} = \frac{(1-\beta) + \delta\beta(1-\tau)}{\beta(1-\tau)}$$

$$\therefore \bar{k} = e^{\bar{z}} \left(\frac{\alpha\beta(1-\tau)}{(1-\beta) + \delta\beta(1-\tau)}\right)^{\frac{1}{1-\alpha}}$$

$$\therefore \bar{Y} = e^{\bar{z}} \left(\frac{\alpha\beta(1-\tau)}{(1-\beta) + \delta\beta(1-\tau)}\right)^{\frac{\alpha}{1-\alpha}}$$

$$\therefore \bar{I} = \delta e^{\bar{z}} \left(\frac{\alpha\beta(1-\tau)}{(1-\beta) + \delta\beta(1-\tau)}\right)^{\frac{1}{1-\alpha}}$$

If we substitute in the given parameters, we get the following steady state values using the algebaic formulations:

$$\bar{k} = 7.2875$$

$$\bar{Y} = 2.213$$

$$\bar{I} = 0.7287$$

In the python file, we solve for the steady states numerically using the given equations. We get the exact same steady state values:

$$\bar{k} = 7.2875$$

$$\bar{Y} = 2.213$$

$$\bar{I} = 0.7287$$

Exercise 6

$$\ell_t = L_t$$

$$k_t = K_t$$

$$w_t = W_t$$

$$r_t = R_t$$

$$c_t = (1 - \tau)[w_t \ell_t + (r_t - \delta)k_t] + k_t + T_t - k_{t+1}$$

$$\frac{1}{c_t^{\gamma}} = \beta E_t \left\{ \frac{1}{c_{t+1}^{\gamma}} [(r_{t+1} - \delta)(1 - \tau) + 1] \right\}$$

$$\frac{a}{(1 - \ell_t)^{\xi}} = \frac{1}{c_t^{\gamma}} w_t (1 - \tau)$$

$$r_t = \alpha \left(\frac{e^{z_t} \ell_t}{k_t}\right)^{1 - \alpha}$$

$$w_t = (1 - \alpha)e^{z_t} \left(\frac{k_t}{e^{z_t} \ell_t}\right)^{\alpha}$$

$$T_t = \tau [w_t \ell_t + (r_t - \delta)k_t]$$

$$z_t = (1 - \rho_z)\bar{z} + \rho_z z_{t-1} + \epsilon_t^z; \qquad \epsilon_t^z \sim \text{i.i.d.}(0, \sigma_z^2)$$

Making all the necessary substitutions and letting the variables be represented by their steady state values, we then have the following equations (note that the stochastic shocks are always zero so the expectation dissapears):

$$\bar{c} = \bar{w}\bar{l} + (\bar{r} - \delta)\bar{k}$$

$$\frac{1}{\bar{c}^{\gamma}} = \beta \frac{1}{\bar{c}^{\gamma}}[(\bar{r} - \delta)(1 - \tau) + 1]$$

$$\frac{a}{(1 - \bar{l})^{\xi}} = \frac{1}{\bar{c}^{\gamma}}\bar{w}(1 - \tau)$$

$$\bar{r} = \alpha(\frac{e^{\bar{z}}\bar{l}}{\bar{k}})^{1 - \alpha}$$

$$\bar{w} = (1 - \alpha)e^{\bar{z}}(\frac{\bar{k}}{e^{\bar{z}\bar{l}}})^{\alpha}$$

In the python file, we solve for the steady states numerically using the given equations. We get

the following steady state values:

$$\bar{k} = 4.225$$

$$\bar{\ell} = 0.580$$

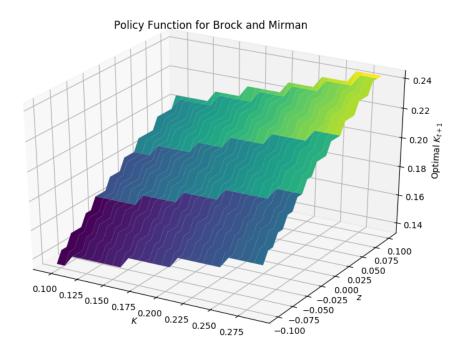
$$\bar{Y} = 1.283$$

$$\bar{I}=0.4225$$

Exercise 8

As seen in the following picture, the investment in capital next period increases when the current capital stock and the productivity factors increase. This coincides with the closed form solution of the policy function.

Figure 1: Optimal Policy Function Given Current Capital Stock and Productivity



Linearization

Exercise 1

Using section 1.4 and the fact that $\bar{K} = A^{\frac{1}{1-\alpha}} = (\alpha\beta)^{\frac{1}{1-\alpha}}$, we have that:

$$F = \frac{\alpha \beta \bar{K}^{\alpha - 1}}{\bar{K}^{\alpha} - \bar{K}} = 2.708$$

$$G = -F(\alpha + \bar{K}^{\alpha - 1}) = -8.843$$

$$H = F(\alpha \bar{K}^{\alpha - 1}) = 2.763$$

$$L = -F\bar{K} = -0.522$$

$$M = F(\bar{K}^{\alpha}) = 1.522$$

$$N = \rho - 0.95$$

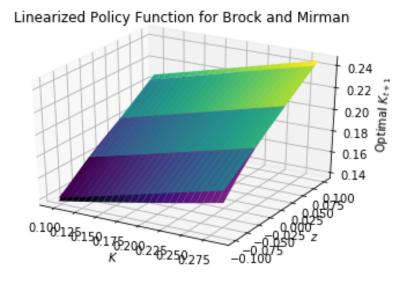
Now we can solve for P and Q:

$$P = \frac{-G \pm \sqrt{G^2 - 4FH}}{2F} = 0.35$$

$$Q = -\frac{LN + M}{FN + FP + G} = 0.193$$

Using the following linearized policy function $K_{t+1} = H(K_t, z_t)\bar{K} + P(K_t - \bar{K}) + Qz_t$, we plot the three dimensional surface plot:

Figure 2: Linearized Policy Function Given Current Capital Stock and Productivity Compared with Closed Form



As seen, the linearized policy function intersects the real policy function (magma color scheme) at around the steady state ($K \approx 0.19, z = 0$) but deviates slightly at the non-linear parts.

Exercise 2

We replace all values of K in the Euler Equation with e^k . When taking the log-linearization of the new Euler Equation we get the following matrix values:

F = -0.380254

G = 0.468559

H = -0.117414

L = 2.569321

M = -2.266718

N = 0.95

P = 0.35

Q = 6.756849

Using the fact that $E\{\varepsilon_t\}=0$, we substitute equations (6), $\tilde{Z}_t=N\tilde{Z}_{t-1}$, and (7), $\tilde{X}_t=P\tilde{X}_{t-1}+Q\tilde{Z}_t$, into equation (5):

$$\begin{split} 0 &= E_{t} \{ F \tilde{X}_{t+1} + G \tilde{X}_{t} + H \tilde{X}_{t-1} + L \tilde{Z}_{t+1} + M \tilde{Z}_{t} \} \\ &= F P \tilde{X}_{t} + F Q N \tilde{Z}_{t} + G P \tilde{X}_{t-1} + G Q \tilde{Z}_{t} + H \tilde{X}_{t-1} + L N \tilde{Z}_{t} + M \tilde{Z}_{t} \\ &= F P P \tilde{X}_{t-1} + F P Q \tilde{Z}_{t} + F Q N \tilde{Z}_{t} + G P \tilde{X}_{t-1} + G Q \tilde{Z}_{t} + H \tilde{X}_{t-1} + L N \tilde{Z}_{t} + M \tilde{Z}_{t} \\ &= [(F P + G) P + H] \tilde{X}_{t-1} + [(F Q + L) N + (F P + G) Q + M] \tilde{Z}_{t} \end{split}$$

Exercise 4

Note that in the steady state, $\bar{z} = 0$. Therefore, there is no uncertainty and the steady states when solved numerically are (this is the same as the DSGE exercise):

$$k = 4.225, \quad \ell = 0.580, \quad r = 0.121, \quad w = 1.328, \quad c = 0.861, \quad T = 0.043, \quad y = 1.283, \quad i = 0.423$$

Exercise 5

Using a first order central difference to estimate the derivatives, we have the following derivative matrix in the jupyter notebook output (refer to problem 5).

Exercise 6

In order to coincide with the problem, we will use $X_t = \{k_t, \ell_{t-1}\}$. In which case, the Uhlig's matrices are (for the log-linearized problem):

$$F = \begin{bmatrix} -17.857 & 3.155 \\ 0 & 0 \end{bmatrix}$$

$$G = \begin{bmatrix} 36.196 & -3.254 \\ -22.527 & 8.638 \end{bmatrix}$$

$$H = \begin{bmatrix} -18.24 & 0 \\ 22.277 & 0 \end{bmatrix}$$

$$L = \begin{bmatrix} 3.155 \\ 0 \end{bmatrix}$$

$$M = \begin{bmatrix} -3.254 \\ 3.004 \end{bmatrix}$$

$$P = \begin{bmatrix} 0.915 & 0 \\ -0.192 & 0 \end{bmatrix}$$

$$Q = \begin{bmatrix} 0.129 \\ -0.011 \end{bmatrix}$$

Figure 3: GDP after 10,000 Simulations

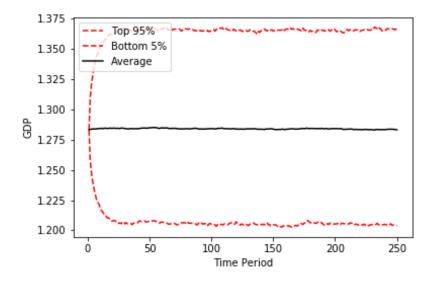


Figure 4: Consumption after 10,000 Simulations

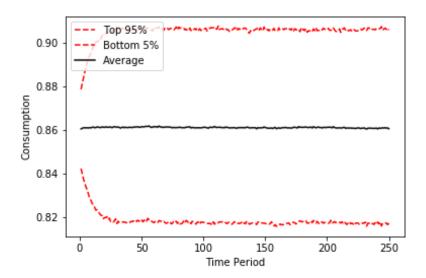


Figure 5: Investment after 10,000 Simulations

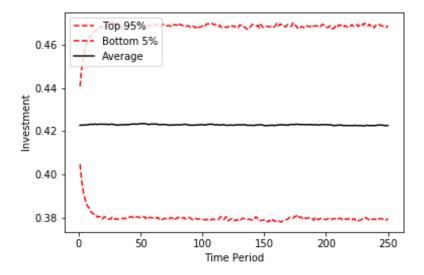
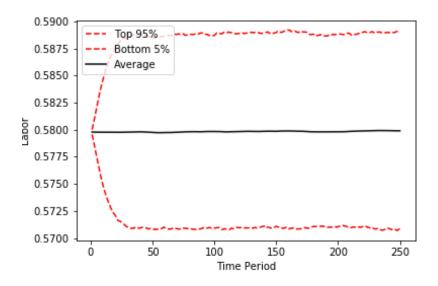


Figure 6: Labor after 10,000 Simulations



Please refer to the juputer notebook for the result with all the moments.

Figure 7: Impulse Response of GDP

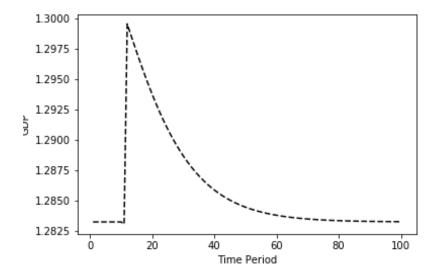


Figure 8: Impulse Response of Consumption

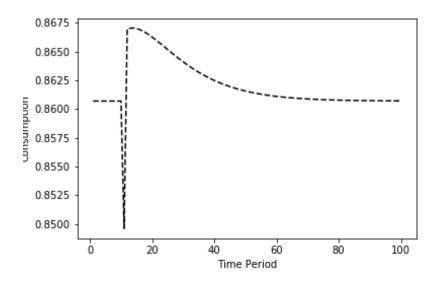


Figure 9: Impulse Response of Investment

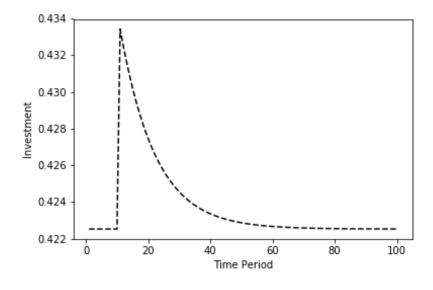


Figure 10: Impulse Response of Labor

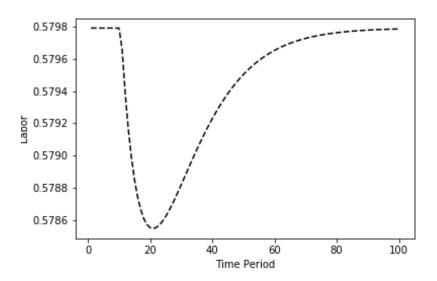


Figure 11: Transition Path for K2

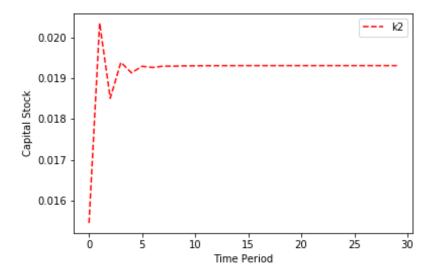


Figure 12: Transition Path for K3

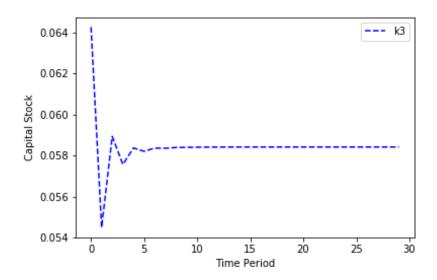
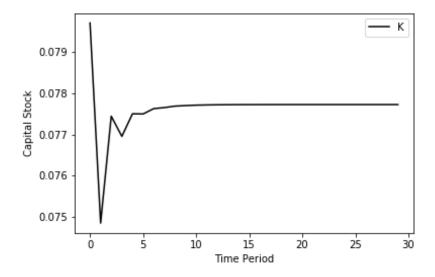


Figure 13: Transition Path for K (Capital Stock)



The graph for capital stock is very similar to the time path iteration result for capital, thus the linear approximation seems to be a good one. In this case, since the compute time is faster and more straight forward, it is a good tradeoff between accuracy and compute time.

Figure 14: GDP for OLG after 10,000 Simulations

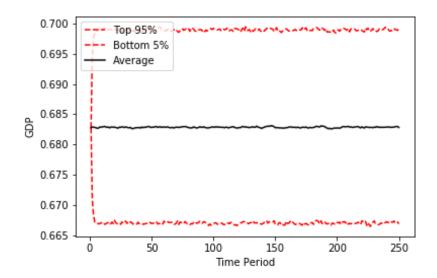


Figure 15: Consumption for OLG after 10,000 Simulations

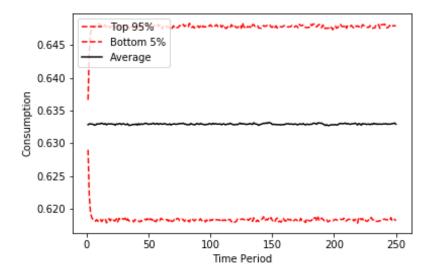
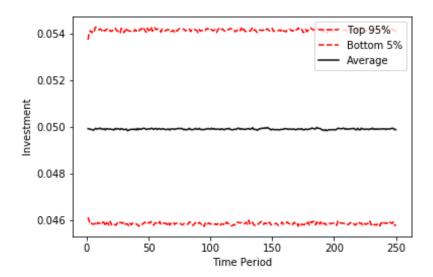


Figure 16: Investment for OLG after 10,000 Simulations



Perturbation

Exercise 1

For the sake of clarity, we will suppress the function arguments. After deriving equation (5) again with respect to u we have that:

$$x_{uuu}(u) = -\frac{F_{xxx}x_u^3 + 3(F_{xxu}x_u^2 + F_{xx}x_{uu}x_u + F_{xuu}x_u + F_{xu}x_{uu}) + F_{uuu}}{F_x}$$

Exercise 2

The market clearing wage rate at k=5 is 0.63. The first and second order approximations are as given:

Figure 17: Approximation About k=5

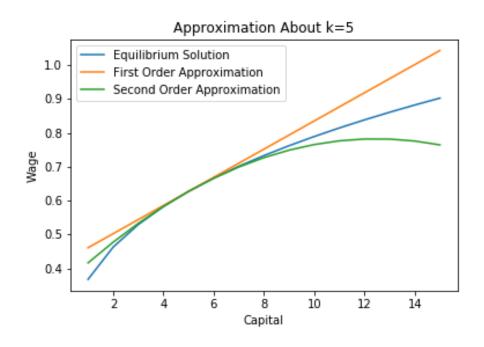
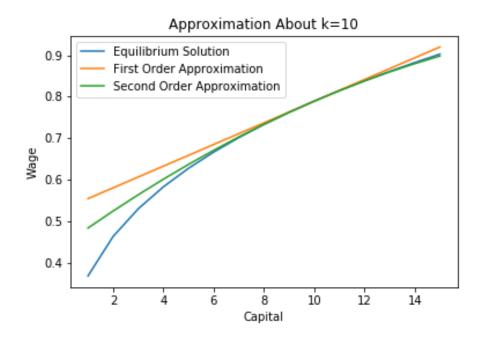


Figure 18: Approximation About k=10



Simplifying the function and plugging in y = G(x), we have the following functional form:

$$0.95G(x)^{0.35} + 1.855G(x) = x^{0.35} + 0.9x$$

We have the following first, second, and third order approximations and their respective differences from the actual solution:

Figure 19: Approximation About x=100

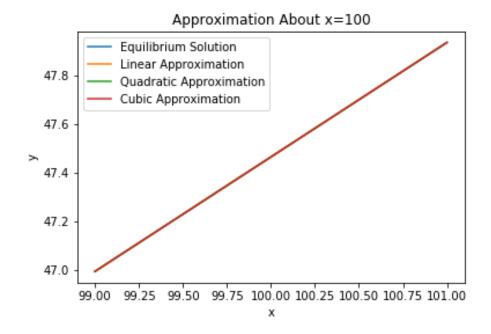
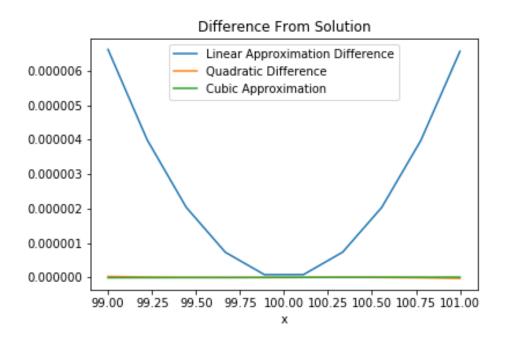


Figure 20: Difference from Solution



From the linearization exercises, we know the analytical forms of the following derivatives (function arguments have been suppressed for clarity):

$$F_y = F = \frac{\alpha \beta \bar{K}^{\alpha - 1}}{\bar{K}^{\alpha} - \bar{K}} = 2.708$$

$$F_x = G = -F(\alpha + \bar{K}^{\alpha - 1}) = -8.843$$

$$F_u = H = F(\alpha \bar{K}^{\alpha - 1}) = 2.763$$

$$x_u = P = \frac{-G \pm \sqrt{G^2 - 4FH}}{2F} = 0.35$$

Taking the analytical derivatives, we have the following forms for the second derivatives:

$$F_{yy} = \frac{2\alpha\beta\bar{K}^{\alpha-1}}{(\bar{K}^{\alpha} - \bar{K})^{2}} = 14.667$$

$$F_{yx} = \frac{\alpha\beta\bar{K}^{\alpha}(\bar{K}^{\alpha-1}(1 - (\alpha + 1)\bar{K}^{\alpha-1}) + \alpha)}{(\bar{K}^{\alpha} - \bar{K})^{3}} = -31.431$$

$$F_{yu} = \frac{\alpha^{2}\beta\bar{K}^{2(\alpha-1)}}{(\bar{K}^{\alpha} - \bar{K})^{2}} = 7.483$$

$$F_{xx} = \frac{\alpha\beta\bar{K}^{\alpha}(2\bar{K}^{3(\alpha-1)} + (\alpha^{2} + 3\alpha - 4)\bar{K}^{2(\alpha-1)} - 2(2\alpha - 1)\bar{K}^{\alpha-1} - \alpha(\alpha - 1))}{(\bar{K}^{\alpha} - \bar{K})^{3}} = 105.309$$

$$F_{xu} = -\frac{\alpha^{2}\beta\bar{K}^{2(\alpha-1)}(\bar{K}^{\alpha-1} + \alpha - 1)}{(\bar{K}^{\alpha} - \bar{K})^{2}} = -16.953$$

$$F_{uu} = \frac{(\alpha - 1)\alpha^{2}\beta\bar{K}^{2\alpha-3}}{\bar{K}^{\alpha} - \bar{K}} = -9.317$$

Now we can find x_{uu} using equation (12):

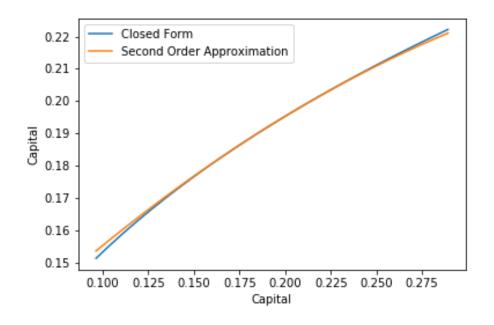
$$x_{uu} = -\frac{F_{yy}x_u^4 + 2F_{yx}x_u^3 + 2F_{yu}x_u^2 + F_{xx}x_u^2 + 2F_{xu}x_u + F_{uu}}{F_yx_u^2 + F_yx_u + F_x}$$

$$= -\frac{14.667 \times 0.35^4 - 2 \times 31.431 \times 0.35^3 + 2 \times 7.483 \times 0.35^2 + 105.309 \times 0.35^2 - 2 \times 16.953 \times 0.35 - 9.317}{2.708 \times 0.35^2 + 2.708 \times 0.35 - 8.843}$$

$$= -1.18$$

Therefore $H_X = x_u = 0.35$ and $H_{XX} = x_{uu} = -1.18$. Using the approximate policy function $K_{t+1} = \bar{K} + x_u(K_t - \bar{K}) + 0.5x_{uu}(K_t - \bar{K})^2$, we have:

Figure 21: Brock and Mirman Second Order



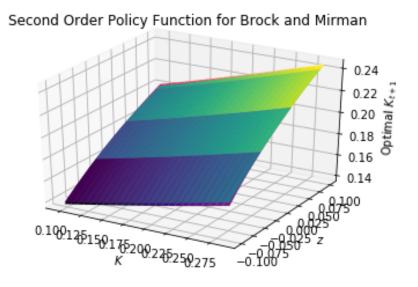
Taking numerical derivatives and finding the Jacobian and Hessian of Γ , we have solve for the following values:

$$H_X = 0.35$$
 $H_Z = 0.1928$
 $H_v = 0$
 $H_{XX} = -1.18$
 $H_{XZ} = 0.35$
 $H_{ZZ} = 0.1827$
 $H_{vv} = -2.8 \times 10^{-17}$

We now have the second order approximation of the policy function, which we plot:

$$K_{t+1} = \bar{K} + H_X(K_t - \bar{K}) + H_Z z_t + \frac{1}{2} (H_{XX}(K_t - \bar{K})^2 + H_{ZZ} z_t^2) + H_{XZ}(K_t - \bar{K}) z_t + H_{vv}$$

Figure 22: Brock and Mirman Second Order Approximation



As seen compared to the actual policy function (which is underneath in purple, barely visible), it approximates the ends better than the linear approximation since the second order approximation is more appropriately curved at the ends.

Filtering

Exercise 3

We use U.S. quarterly and seasonally adjusted data, in billions of chained 2009 dollars, and plot the natural log of the periodogram to get the spectral density estimates from 1960 to 2017.

Figure 23: Real GDP

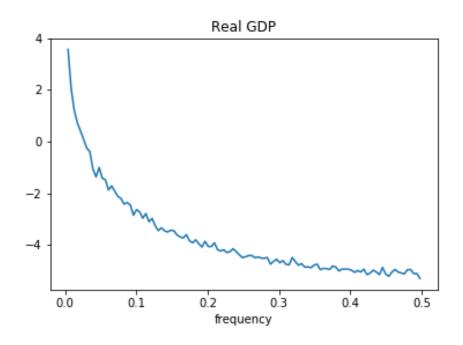


Figure 24: Real Personal Consumption

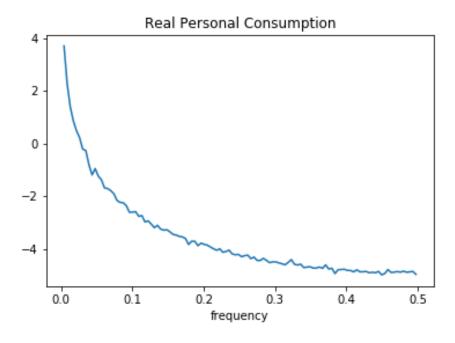


Figure 25: Real Gross Private Domestic Investment

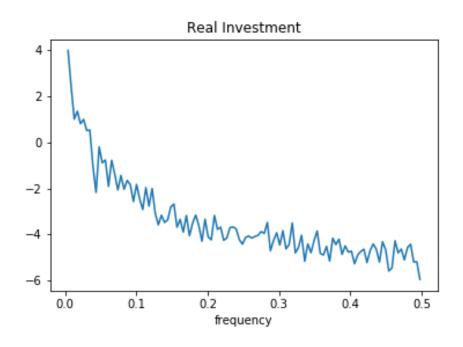
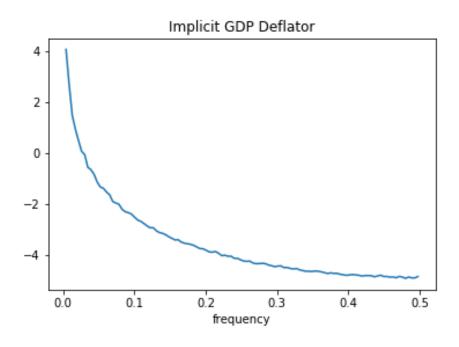


Figure 26: Implicit GDP Deflator



We plot the spectral estimates for first the cyclical then the trend parts of the HP-filtered time series:

Figure 27: Real GDP Cyclical

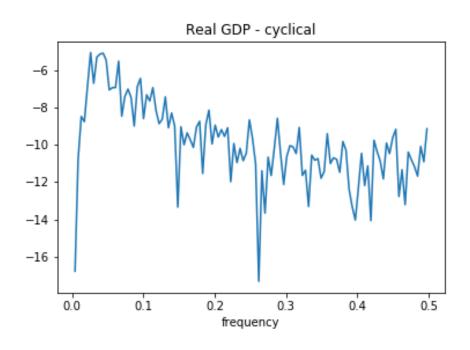


Figure 28: Real Personal Consumption Cyclical

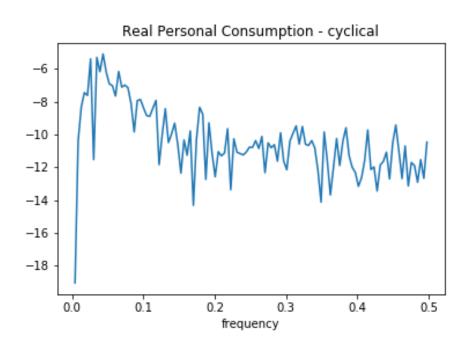


Figure 29: Real Gross Private Domestic Investment Cyclical

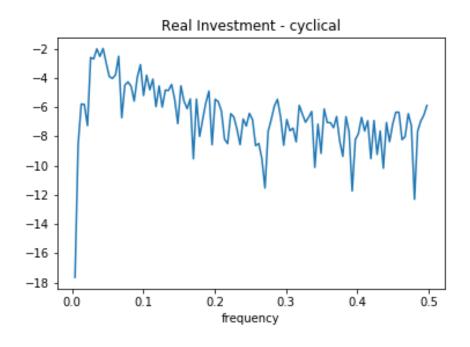


Figure 30: Implicit GDP Deflator Cyclical

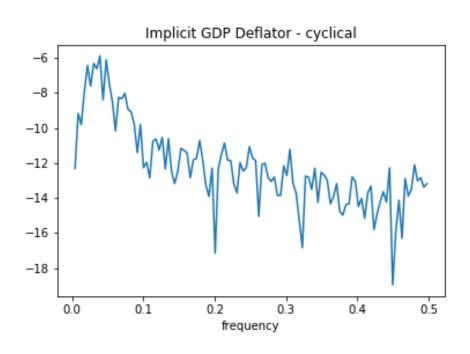


Figure 31: Real GDP Trend

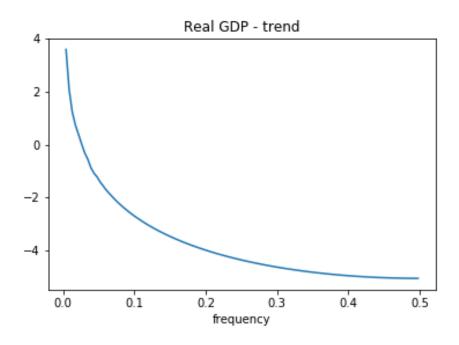


Figure 32: Real Personal Consumption Trend

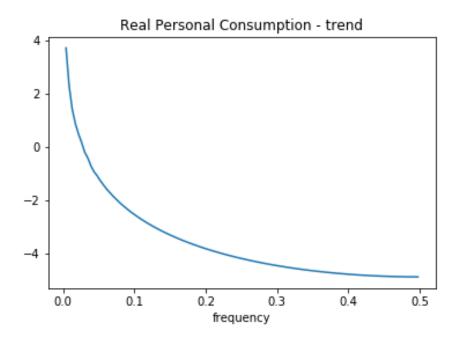


Figure 33: Real Gross Private Domestic Investment Trend

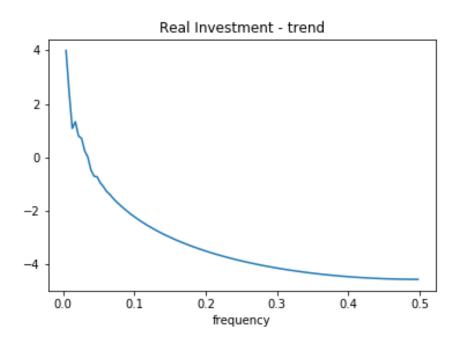
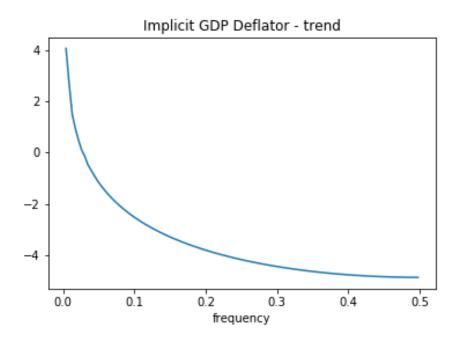


Figure 34: Implicit GDP Deflator Trend



Please refer to the jupyter notebook for the standard deviation, autocorrelation, and correlation with GDP results for each time series and corresponding λ . Note that the standard deviation and

correlation with GDP decrease as λ increases, while autocorrelation increases as λ increases.

Figure 35: Real Investment HP-Filter $\lambda = 100$

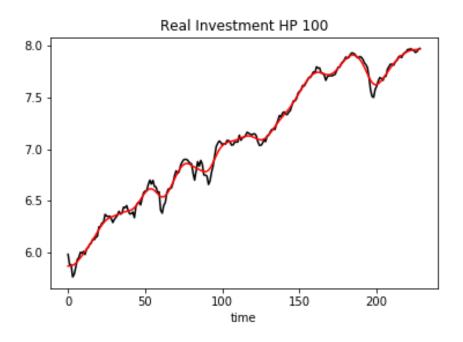


Figure 36: Real Investment HP-Filter $\lambda = 400$

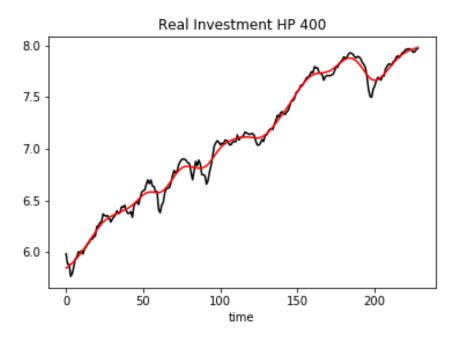


Figure 37: Real Investment HP-Filter $\lambda = 1600$

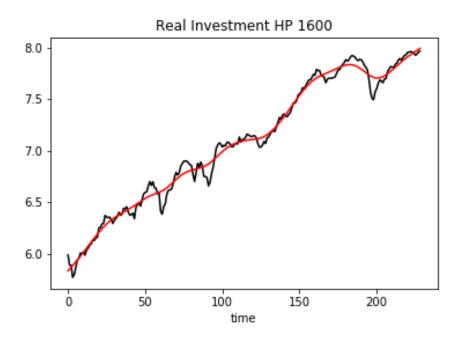


Figure 38: Real Investment HP-Filter $\lambda = 6400$

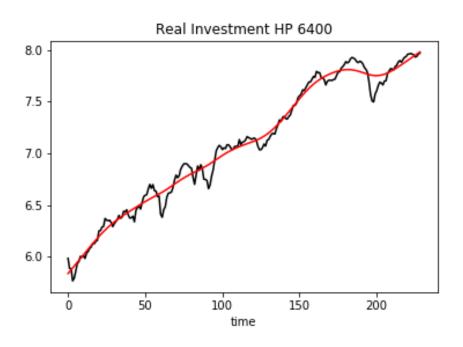
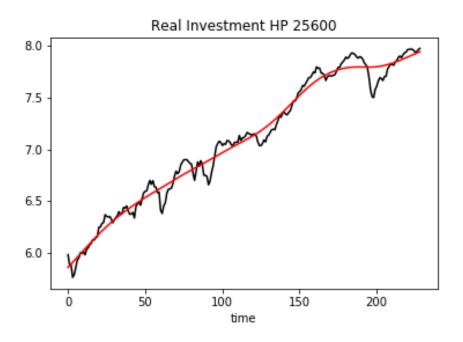


Figure 39: Real Investment HP-Filter $\lambda = 25600$



Replicate Table ?? using data collected from the sites mentioned in Section ??. Use quarterly data going back as far as 1947, if possible. Do this for each of the following filters:

- A linear trend filter
- the first difference $(y_t y_{t-1})$ filter
- $HP(\lambda = 1600)$
- BP(6,32,K=8)