Problem Set #1

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Exercise 1

In the Brock and Mirman's model the households solves the following dynamic program:

$$V(K_{t}, z_{t}) = \max_{K_{t+1}} \ln \left(e^{z_{t}} K_{t}^{\alpha} - K_{t+1} \right) + \beta E_{t} \left\{ V(K_{t+1}, z_{t+1}) \right\},\,$$

where the law of motion is:

$$z_{t+1} = \rho z_t + \varepsilon_t$$
; $\varepsilon_t \sim i.i.d\left(0, \sigma^2\right)$.

The associated Euler equation is:

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

In order to find an algebraic solution for the policy function we use the "guess and verify" method. We guess that the policy function is in the form of $K_{t+1} = Ae^{z_t}K_t^{\alpha}$ and we substitute it in the Euler equation. We obtain:

$$\frac{1}{e^{z_t}K_t^{\alpha} - Ae^{z_t}K_t^{\alpha}} = \beta E_t \left\{ \frac{\alpha e^{z_{t+1}}A^{\alpha-1}e^{(\alpha-1)z_t}K_t^{\alpha^2-\alpha}}{e^{z_{t+1}}A^{\alpha}e^{\alpha z_t}K_t^{\alpha^2} - A^{1+\alpha}e^{z_{t+1}}e^{\alpha z_t}K_t^{\alpha^2}} \right\},$$

$$\iff \frac{1}{(1-A)e^{z_t}K_t^{\alpha}} = \beta E_t \left\{ \frac{\alpha A^{-1}e^{-z_t}K_t^{-\alpha}A^{\alpha}e^{\alpha z_t}K_t^{\alpha}e^{z_{t+1}}}{(1-A)A^{\alpha}e^{\alpha z_t}K_t^{\alpha^2}e^{z_{t+1}}} \right\},$$

$$\iff \frac{1}{(1-A)e^{z_t}K_t^{\alpha}} = \beta E_t \left\{ \frac{\alpha}{(1-A)Ae^{z_t}K_t^{\alpha}} \right\},$$

$$\iff \alpha\beta = \frac{(1-A)Ae^{z_t}K_t^{\alpha}}{(1-A)e^{z_t}K_t^{\alpha}},$$

$$\iff A = \alpha\beta.$$

Therefore the policy function is $K_{t+1} = \alpha \beta e^{z_t} K_t^{\alpha}$.

Exercise 2

Our baseline model with:

$$u\left(c_{t},\ell_{t}\right)=\ln c_{t}+a\ln \left(1-\ell_{t}\right),$$

and

$$f(K_t, L_t, z_t) = e^{z_t} K_t^{\alpha} L_t^{1-\alpha},$$

has the following characterizing equations:

$$\begin{split} c_t &= (1 - \tau) \left[w_t \ell_t + (r_t - \delta) \, k_t \right] + k_t + T_t - k_{t+1}, \\ \frac{1}{c_t} &= \beta E_t \left\{ \frac{1}{c_{t+1}} \left[(r_{t+1} - \delta) \, (1 - \tau) + 1 \right] \right\}, \\ \frac{a}{1 - \ell_t} &= \frac{1}{c_t} w_t (1 - \tau), \\ r_t &= \alpha e^{z_t} k_t^{\alpha - 1} \ell_t^{1 - \alpha}, \\ w_t &= (1 - \alpha) e^{z_t} k_t^{\alpha} \ell_t^{- \alpha}, \\ \tau \left[w_t \ell_t + (r_t - \delta) \, k_t \right] &= T_t, \\ z_t &= (1 - \rho_z) \, \overline{z} + \rho_z z_{t-1} + \epsilon_t^z; \quad \epsilon_t^z \sim \text{ i.i.d. } \left(0, \sigma_z^2 \right). \end{split}$$

In this case we can't use the same tricks as in Exercise 1 to solve for the policy function because the model is too complex and it's impossible to guess a good algebraic solution.

Exercise 3

Our baseline model with:

$$u(c_t, \ell_t) = \frac{c_t^{1-\gamma} - 1}{1-\gamma} + a \ln(1 - \ell_t),$$

and

$$f(K_t, L_t, z_t) = e^{z_t} K_t^{\alpha} L_t^{1-\alpha},$$

has the following characterizing equations:

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

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

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

$$r_{t} = \alpha e^{z_{t}} k_{t}^{\alpha - 1} \ell_{t}^{1 - \alpha},$$

$$w_{t} = (1 - \alpha) e^{z_{t}} k_{t}^{\alpha} \ell_{t}^{-\alpha},$$

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

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

Exercise 4

Our baseline model with:

$$u(c_t, \ell_t) = \frac{c_t^{1-\gamma} - 1}{1-\gamma} + a \frac{(1-\ell_t)^{1-\xi} - 1}{1-\xi},$$

and

$$f(K_t, L_t, z_t) = e^{z_t} \left[\alpha K_t^{\eta} + (1 - \alpha) L_t^{\eta} \right]^{\frac{1}{\eta}},$$

has the following characterizing equations:

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

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

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

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

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

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

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

Exercise 5

Assume $\ell_t = 1$. For the market clearing conditions we also have that $L_t = \ell_t = 1$. Our baseline model with:

$$u\left(c_{t},\ell_{t}\right)=\frac{c_{t}^{1-\gamma}-1}{1-\gamma},$$

and

$$f(K_t, L_t, z_t) = K_t^{\alpha} (L_t e^{z_t})^{1-\alpha}$$

has the following characterizing equations:

$$\begin{split} c_t &= (1 - \tau) \left[w_t \ell_t + (r_t - \delta) \, k_t \right] + k_t + T_t - k_{t+1}, \\ c_t^{-\gamma} &= \beta E_t \left\{ c_{t+1}^{-\gamma} \left[(r_{t+1} - \delta) \, (1 - \tau) + 1 \right] \right\}, \\ c_t^{-\gamma} w_t (1 - \tau) &= 0, \\ r_t &= \alpha k_t^{\alpha - 1} \, (\ell_t e^{z_t})^{1 - \alpha}, \\ w_t &= (1 - \alpha) k_t^{\alpha} \ell_t^{-\alpha} e^{(1 - \alpha) z_t}, \\ \tau \left[w_t \ell_t + (r_t - \delta) \, k_t \right] &= T_t, \\ z_t &= (1 - \rho_z) \, \overline{z} + \rho_z z_{t-1} + \epsilon_t^z; \quad \epsilon_t^z \sim \text{ i.i.d. } \left(0, \sigma_z^2 \right). \end{split}$$

The steady state equations, considering $\ell_t = 1$, are:

$$\bar{c} = (1 - \tau) \left[\bar{w} + (\bar{r} - \delta) \bar{k} \right] + \bar{T},$$

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

$$\begin{split} \bar{c}^{-\gamma}\bar{w}(1-\tau) &= 0,\\ \bar{r} &= \alpha \bar{k}^{\alpha-1}e^{(1-\alpha)\bar{z}},\\ \bar{w} &= (1-\alpha)\bar{k}^{\alpha}e^{(1-\alpha)\bar{z}},\\ \tau\left[\bar{w} + (\bar{r} - \delta)\,\bar{k}\right] &= \bar{T}. \end{split}$$

Therefore, the steady state for the value of *k* is given by:

$$ar{k} = \left[\frac{1}{lpha} \left(\frac{1-eta}{eta(1- au)} + \delta \right) \right]^{\frac{1}{lpha-1}} e^{ar{z}}.$$

The numerical solution is presented below.

```
In [1]: # import packages
        import numpy as np
        from matplotlib import pyplot as plt
        from scipy import optimize
        # define system of characterizing equations
        def charact_eq(x, p):
            gamma, beta, alpha, delta, tau = p
            return [x[0] - x[1] - (x[2]-delta)*x[3],
                    x[0]**(-gamma) - beta*x[0]**(-gamma) * ((x[2]-delta)*(1-tau)+1),
                    x[2] - alpha * x[3] **(alpha-1),
                    x[1] - (1-alpha) * x[3] **alpha,
                    x[4] - tau*(x[1] + (x[2]-delta)*x[3])]
        # solve the system
        param = [2.5, .98, .4, .1, .05]
        X = optimize.root(charact_eq,[.5,.5,.5,.5], args=param)
        # present results
        c, w, r, k, T = X.x
        y = k ** .4
        i = .1 * k
        print("Steady state values:\n")
        print("{:<15}{:<5}{:<5}".format('Consumption','c',round(c,4)))</pre>
        print("{:<15}{:<5}\".format('Wage rate','w',round(w,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Rental rate','r',round(r,4)))</pre>
        print("{:<15}{:<5}\".format('Capital','k',round(k,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Transfer','T',round(T,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Output','y',round(y,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Investment','i',round(i,4)))</pre>
```

```
gamma, beta, alpha, delta, tau = param
print("\nThe algebraic solution for capital is k = ", round(((1/alpha)*((1-beta)
```

Steady state values:

Consumption	С	1.4845
Wage rate	W	1.328
Rental rate	r	0.1215
Capital	k	7.2875
Transfer	T	0.0742
Output	У	2.2133
Investment	i	0.7287

The algebraic solution for capital is k = 7.2875

Exercise 6

Our baseline model with:

$$u(c_t, \ell_t) = \frac{c_t^{1-\gamma} - 1}{1-\gamma} + a \frac{(1-\ell_t)^{1-\xi} - 1}{1-\xi},$$

and

$$f(K_t, L_t, z_t) = K_t^{\alpha} (L_t e^{z_t})^{1-\alpha},$$

has the following characterizing equations:

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

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

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

$$r_{t} = \alpha k_{t}^{\alpha - 1} (\ell_{t} e^{z_{t}})^{1 - \alpha},$$

$$w_{t} = (1 - \alpha) k_{t}^{\alpha} \ell_{t}^{-\alpha} e^{(1 - \alpha) z_{t}},$$

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

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

The steady state equations, considering $\ell_t = 1$, are:

$$\bar{c} = (1 - \tau) \left[\bar{w}\bar{\ell} + (\bar{r} - \delta)\bar{k} \right] + \bar{T},$$

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

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

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

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

$$\tau \left[\bar{w}\bar{\ell} + (\bar{r} - \delta)\bar{k}\right] = \bar{T}.$$

The numerical solution is presented below.

```
In [2]: # import packages
        import numpy as np
        from matplotlib import pyplot as plt
        from scipy import optimize
        # define system of characterizing equations
        def charact_eq(x, p):
            gamma, xi, beta, alpha, a, delta, tau = p
            return [x[0] - x[1]*x[5] - (x[2]-delta)*x[3],
                     x[0]**(-gamma) - beta*x[0]**(-gamma) * ((x[2]-delta)*(1-tau)+1),
                     x[0]**(-gamma)*x[1]*(1-tau) - a*(1-x[5])**(-xi),
                     x[2] - alpha * x[3] **(alpha-1) * x[5] **(1-alpha),
                     x[1] - (1-alpha) * x[3] **alpha * x[5] **(-alpha),
                     x[4] - tau*(x[1]*x[5] + (x[2]-delta)*x[3])]
        # solve the system
        param = [2.5, 1.5, .98, .4, .5, .1, .05]
        X = \text{optimize.root}(\text{charact\_eq}, [.5, .5, .5, .5, .5], \text{ args=param})
        # present results
        c, w, r, k, T, 1 = X.x
        y = (k ** .4)*(1 ** .6)
        i = .1 * k
        print("Steady state values:\n")
        print("{:<15}{:<5}{:<5}".format('Consumption','c',round(c,4)))</pre>
        print("{:<15}{:<5}\".format('Wage rate','w',round(w,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Rental rate','r',round(r,4)))</pre>
        print("{:<15}{:<5}\".format('Capital','k',round(k,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Transfer','T',round(T,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Labor','l',round(1,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Output','y',round(y,4)))</pre>
        print("{:<15}{:<5}{:<5}".format('Investment','i',round(i,4)))</pre>
Steady state values:
Consumption
                     0.8607
Wage rate
                     1.328
               W
Rental rate r 0.1215
Capital
              k 4.2252
```

```
Transfer T 0.043
Labor 1 0.5798
Output y 1.2832
Investment i 0.4225
```

Exercise 7

```
In [3]: # import packages
        import numpy as np
        from matplotlib import pyplot as plt
        from scipy import optimize
        # define system of characterizing equations
        def charact_eq(x, p):
            gamma, xi, beta, alpha, a, delta, tau = p
            return [x[0] - x[1]*x[5] - (x[2]-delta)*x[3],
                    x[0]**(-gamma) - beta*x[0]**(-gamma) * ((x[2]-delta)*(1-tau)+1),
                    x[0]**(-gamma)*x[1]*(1-tau) - a*(1-x[5])**(-xi),
                    x[2] - alpha * x[3] **(alpha-1) * x[5] **(1-alpha),
                    x[1] - (1-alpha) * x[3] **alpha * x[5] **(-alpha),
                    x[4] - tau*(x[1]*x[5] + (x[2]-delta)*x[3])
        # define parameters and initial value
        param = [2.5, 1.5, .98, .4, .5, .1, .05]
        x0 = [.5, .5, .5, .5, .5]
        # model at the steady state
        X = optimize.root(charact_eq,x0, args=param)
        c, w, r, k, T, 1 = X.x
        y = (k ** .4)*(1 ** .6)
        i = .1 * k
        # compute derivatives
        h = 1e-5
        dd = np.empty([len(x0)+2,len(param)])
        for ip in range(len(param)):
            h_param = param.copy()
            h_param[ip] += h
            h_X = optimize.root(charact_eq,x0, args=h_param)
            for ix in range(len(x0)):
                dd[ix,ip] = (h_X.x[ix] - X.x[ix]) / h
            h_y = (h_x.x[3] ** .4)*(h_x.x[5] ** .6)
            h_i = .1 * h_X.x[3]
```

```
dd[-2,ip] = (h_y - y) / h
    dd[-1,ip] = (h_i - i) / h

namevar = ['c', 'w', 'r', 'k', 'T', 'l', 'y', 'i']

# present results
print("{:<5}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}
```

print("{:<5}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}{:>10}.format(namevar[ix],r

	gamma	xi	beta	alpha	a	delta	tau
С	0.0283	-0.1633	1.751	2.0853	-0.3767	-3.5111	-0.2344
W	-0.0	0.0	7.9879	4.3961	0.0	-7.287	-0.1648
r	-0.0	-0.0	-1.096	0.0	-0.0	1.0	0.0226
k	0.1387	-0.8017	65.4385	25.9858	-1.8492	-48.3453	-2.3232
T	0.0014	-0.0082	0.0876	0.1043	-0.0188	-0.1756	0.849
1	0.019	-0.11	0.2603	-0.7694	-0.2538	1.3197	-0.1389
у	0.0421	-0.2435	8.2949	2.135	-0.5616	-4.1209	-0.4667
i	0.0139	-0.0802	6.5438	2.5986	-0.1849	-4.8345	-0.2323