

# Problem set 1

Due Tuesday, January 20, 11.59pm by email. In addition to your codes, please compile a pdf that contains all requested figures and numbers.

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## 1 Steady state of the standard incomplete markets model

Consider the standard incomplete markets model from the lecture with the following parametrization:

- $u(c) = \log c$
- $\log s_{it} = \rho \log s_{it-1} + \epsilon_{it}$  with
  - $\epsilon_{it} \sim \mathcal{N}\left(0, \sigma_\epsilon \sqrt{1 - \rho^2}\right)$
  - $\rho = 0.975$  and  $\sigma_\epsilon = 0.7$
- $Y = 1, r = 0.01/4, \underline{a} = 0, \beta = 0.988$

Write a program that solves the Bellman equation and the steady state distribution! Specifically, do the following:

1. Set up grids for  $a$  and  $s$ , and define the transition matrix  $\Pi_{ss'}$ . Specifically:
  - (a) discretize  $s$  using Rouwenhorst's method to 7 grid points, yielding the grid for  $s$  and  $\Pi_{ss'}$ . Normalize the resulting grid so that  $\mathbb{E}_i s_{it} = 1$ .
  - (b) discretize  $a$  to 500 grid points between  $\underline{a} = 0$  and  $a^{\max} = 200$ , using the double-exponential grid from the lecture.

If you use Python, you can download the library `utils.py` from the `codes/` folder in my Dropbox repository and use the following code to set up the grids:

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```
import utils # Import helpers library (utils.py must be in same directory).

num_s, num_a = 7, 500 # Set size of income and asset grid.
a_max = 200 # Set upper bound of asset grid.
rho, sigma = 0.975, 0.7 # Set persistence and unconditional s.d. of log(s).

# Set up grids.
s_grid, pi_s, Pi = utils.rouwenhorst(rho, sigma, num_s)
a_grid = utils.exponential_grid(0, a_max, num_a)
```

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2. Write a function `backward_step` that takes  $\partial V/\partial a$  as input, and returns  $\partial V/\partial a_-$  as well as the savings and consumption policies  $a(s, a_-)$  and  $c(s, a_-)$ . Use the endogenous gridpoint method discussed in class to do so. In Python, you can use the provided function `utils.interpolate` to do the inverting/interpolation from  $a_-(s, a)$  to  $a(s, a_-)$ .

*Hint:* besides  $\partial V/\partial a$ , you will also need to pass various parameters to `backward_step`, including the transition matrix  $\Pi_{ss'}$ , the grids for  $a$  and  $s$ , the interest rate  $r$ , aggregate income  $Y$ , discount factor  $\beta$ , and—optionally—the intertemporal elasticity of substitution  $\sigma$  in case you choose to implement a more general version where  $u(c) = \frac{1}{1-\sigma^{-1}} c^{1-\sigma^{-1}}$ .

3. Initialize  $\partial V/\partial a_-$  guessing that for each  $(s, a_-)$  households consume 10% of their available “cash on hand”:

$$c(s, a_-) = 0.1 \cdot ((1+r)a_- + sY).$$

Then substitute this guess into the envelope condition to obtain an initial guess for the marginal value of assets:

$$\frac{\partial V}{\partial a_-}(s, a_-) = (1+r)u'(c(s, a_-)).$$

4. Starting from this initial guess for  $\partial V/\partial a_-$ , iterate on `backward_step` until convergence.
5. Plot the resulting net saving policy,  $a(s, a_-) - a_-$ , and consumption function  $c(s, a_-)$  as a function of  $a_-$ . (Your plots should have 7 lines, one for each income level).

Congratulations, you have solved the Bellman equation! Next, use the resulting savings policy  $a(s, a_-)$  to compute the stationary distribution over  $(s, a)$ .

6. Use Young’s method to discretize the savings policy  $a(s, a_-)$  that solves the Bellman equation. Specifically, for each point in the state space  $(s, a_-)$ , the discretization should give you an index  $\mathbf{i}_a$  and weight  $\pi_a$  on the lower gridpoint bracketing  $a(s, a_-)$ , such that:

$$\mathbf{a} = \pi_a \cdot \mathbf{a\_grid}[\mathbf{i}_a] + (1 - \pi_a) \cdot \mathbf{a\_grid}[\mathbf{i}_a + 1].$$

If you use Python, you can use the provided function `utils.youngs_method` to do so, using the following code:

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```
a_i, a_pi = utils.youngs_method(a_grid, a) # a is savings policy from Q1.4
```

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7. Write a function `forward_step` that takes a probability mass distribution over  $(s, a_-)$  as input, and returns a probability mass distribution over  $(s', a)$ .

*Hint:* In addition to the current guess for the distribution  $D_t$ , you will also need to pass the transition matrix  $\Pi_{ss'}$  and the discretized asset policy  $(\mathbf{i}_a, \pi_a)$  from the previous step as parameters to `forward_step`. To arrive at  $D_{t+1}$ , it’s best to proceed in 2 steps:

- (a) Initialize  $D_t^{\text{end}} = \mathbf{0}_{\text{num.s} \times \text{num.a}}$ . Then update the asset position by iterating through all  $(s, a_-)$ . Letting  $(\mathbf{i}_s, \mathbf{i}_{a_-})$  denote the corresponding indices of the grid, for each  $(\mathbf{i}_s, \mathbf{i}_{a_-})$ :
  - i. fix  $D = D_t(\mathbf{i}_s, \mathbf{i}_{a_-})$  as the mass of the originating state.
  - ii. look up the index  $\mathbf{i}_a$  and weight  $\pi_a$  for next periods’ asset position at  $(\mathbf{i}_s, \mathbf{i}_{a_-})$ .
  - iii. then add  $\pi_a \cdot D$  to  $D^{\text{end}}(\mathbf{i}_s, \mathbf{i}_a)$  and add  $(1 - \pi_a) \cdot D$  to  $D^{\text{end}}(\mathbf{i}_s, \mathbf{i}_a + 1)$ .

- (b) Update the income to arrive at  $D_{t+1}$  by premultiplying  $D_t^{\text{end}}$  from the previous step with the transpose of  $\Pi_{ss'}$ . In Python:

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```
D_next = Pi.T @ D_end
```

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8. Initialize some initial distribution over  $(s, a)$  and iterate on **forward\_step** until convergence. A decent initial guess for  $D$  is to assume that  $a$  and  $s$  are independent, with  $a$  being uniformly distributed across its grid, and  $s$  being distributed according to its stationary distribution. (If you used my code above to implement Rouwenhorst's method, the stationary distribution over  $s$  is given by `pi_s`).

9. Use the resulting invariant distribution to compute:

- (c) the aggregate asset stock, normalized by annual earnings,

$$A = \sum_{s, a_-} D(s, a_-) \cdot \frac{a_-}{4Y},$$

- (d) the average marginal propensity to consume out of current income (which equals  $(1 + r)^{-1}$  times the MPC out of incoming assets  $a_-$ ):

$$\overline{MPC} = \frac{1}{1 + r} \sum_{s, a_-} D(s, a_-) \cdot \frac{\partial c(s, a_-)}{\partial a_-}.$$

(*Hint:* Use forward differences on `a_grid` to numerical differentiate the consumption policy.)

10. How do the aggregate asset stock and the average MPC vary with  $\beta$ ?

11. Plot a histogram of the cross-sectional asset distribution.

Congratulations, you solved the standard incomplete markets model. Feel free to play around with your code and explore other parametrization and implications. Bonus question: How long does it take to solve for the steady state on your computer? In Python, you can use the provided tic-toc functions to time your code:

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```
utils.tic()
# add your code here
utils.toc()
```

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## 2 Homogeneity of the aggregate savings functions

Prove that the aggregate asset stock  $A$  is homogenous of degree 1 in  $(Y, \underline{a})$  for  $u(c) = \frac{c^{1-\gamma}-1}{1-\gamma}$  with  $\gamma > 0$ .

*Hint:* This question is intended to be done by pen & paper. If you are trying to verify this on your computer, you will also need to adjust the upper bound of the asset grid accordingly.