## Review Questions: Stochastic Growth Model

Econ720. Fall 2020. Prof. Lutz Hendricks

Useful problems from the texts:

- Acemoglu, "Introduction to Modern Economic Growth:" Exercises 17.2.1, 17.5.2, 17.9, 17.11, 17.12, 17.18, 17.19.
- Romer, "Advanced Macro:" Exercises 4.8, 4.9, 4.10, 4.11.

## 1 Stochastic growth model

[Due to Oksana Leukhina] Consider a neoclassical growth model with uncertainty. Time is discrete: t = 0, 1, 2, ...

Demographics: There is a unit mass of identical households.

Preferences:  $\mathbb{E} \sum_{t=0}^{\infty} \beta^t \ln c_t$ .

Endowments: The household is endowed with one unit of labor time in each period and with initial capital stock  $k_0$ .

Technology: Output per capita is given by  $\theta_t k_t^{\alpha}$  where k is the capital-labor ratio. The resource constraint is given by  $k_{t+1} = \theta_t k_t^{\alpha} - c_t$ .  $\theta_t$  is an i.i.d. random variable that takes on the values  $\theta^1$  with probability  $\pi$  and  $\theta^2$  with probability  $1 - \pi$ .

- a. State the social planner's problem in recursive form.
- b. Solve for the value function and the policy function. Start from the guess  $V(k, \theta^i) = A + B \ln k + C \ln \theta^i$ .
- c. Define a Competitive Equilibrium, assuming: (i)  $\theta_t$  is an aggregate shock; (ii) trading takes place in a sequence of markets; (iii) households trade a complete set of Arrow securities (an Arrow security pays one unit of the good if history  $\theta^t = (\theta_1, ..., \theta_t)$  occurs).

## 1.1 Answer: Stochastic growth model

[I have not checked the details.]

a. The Bellman equation is formulated as

$$V(k, \theta_i) = \max_{c, k' \ge 0} \{ \ln(c) + \beta \mathbb{E} V_1(\theta', k') \}$$
  
s.t.  $c + k' \le \theta_i k^{\alpha}$ ,

where  $\theta$  is a discrete i.i.d. r.v. that takes on two values  $\theta_1$  with probability  $\pi$  and  $\theta_2$  with probability  $(1-\pi)$ .

Find the value function using the method of undertermined coefficients. Use  $V(k, \theta_i) = A + B \ln k + C \ln \theta_i$  as a guess.

The Bellman equation is formulated as

$$V(k, \theta_i) = \max_{k'} \left\{ \ln(\theta_i k^{\alpha} - k') + \beta \left[ \pi V(k', \theta_h) + (1 - \pi) V(k', \theta_h) \right] \right\}.$$

Plugging the guess into the functional equation gives

$$V(k,\theta_i) = \max_{k'} \left\{ \ln(\theta_i k^{\alpha} - k') + \beta [\pi (A + B \ln k' + C \ln \theta_h) + (1 - \pi) (A + B \ln k' + C \ln \theta_l)] \right\} =$$

$$= \max_{k'} \left\{ \ln(\theta_i k^{\alpha} - k') + \beta B \ln k' + \beta A + \beta C [\pi \ln \theta_h + (1 - \pi) \ln \theta_l] \right\}.$$

Taking first-order conditions, which are necessary and sufficient, gives

$$\frac{-1}{(\theta_i k^{\alpha} - k')} + \frac{\beta B}{k'} = 0.$$

The optimal policy function is then

$$k' = \frac{\beta B}{1 + \beta B} \theta_i k^{\alpha},$$

i.e., a constant share of output is saved in each period.

We have

$$\begin{split} A + B \ln k + C \ln \theta_i &= \ln \left( \frac{\theta_i k^\alpha}{1 + \beta B} \right) + \beta B \ln \left( \frac{\beta B}{1 + \beta B} \theta_i k^\alpha \right) + \beta A + \beta C [\pi \ln \theta_h + (1 - \pi) \ln \theta_l] \\ A + B \ln k + C \ln \theta_i &= \ln \left( \frac{\theta_i k^\alpha}{1 + \beta B} \right) + \beta B \ln \left( \frac{\beta B}{1 + \beta B} \theta_i k^\alpha \right) + \beta A + \beta C [\pi \ln \theta_h + (1 - \pi) \ln \theta_l] \\ A + B \ln k + C \ln \theta_i &= \alpha \left( 1 + \beta B \right) \ln k + (1 + \beta B) \ln \theta_i - (1 + \beta B) \ln (1 + \beta B) + \beta B \ln \beta B + \beta A + \beta C [\pi \ln \theta_h + (1 - \pi) \ln \theta_l] \end{split}$$

We are now able to determine the coefficients:

$$\begin{array}{lll} B & = & \alpha \left( {1 + \beta B} \right), \text{ i.e. } B = \frac{\alpha }{{\left( {1 - \alpha \beta } \right)}} \\ C & = & \left( {1 + \beta B} \right) \\ A & = & \left( {\left( {1 + \beta B} \right)\ln \theta _i - \left( {1 + \beta B} \right)\ln \left( {1 + \beta B} \right) + \beta B\ln \beta B + \beta C[\pi \ln \theta _h + \left( {1 - \pi } \right)\ln \theta _l]} \right)/\left( {1 - \beta } \right) \end{array}$$