

## Solutions to Problem Set 14

ECON 772001 - Math for Economists  
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Practice for the Final – Not Collected or Graded

### 1. Stochastic Linear-Quadratic Dynamic Programming

The problem is to choose contingency plans for a flow variable  $z_t$  for all  $t = 0, 1, 2, \dots$  and a stock variable  $y_t$  for all  $t = 1, 2, 3, \dots$  to maximize the objective function

$$E_0 \sum_{t=0}^{\infty} \beta^t (Ry_t^2 + Qz_t^2),$$

subject to the constraints  $y_0$  given and

$$Ay_t + Bz_t + \varepsilon_{t+1} \geq y_{t+1} \quad (1)$$

for all  $t = 0, 1, 2, \dots$  and all possible realizations of  $\varepsilon_{t+1}$ , where  $0 < \beta < 1$ ,  $R < 0$ ,  $Q < 0$ ,  $A$ , and  $B$  are all constant, known parameters and  $\varepsilon_{t+1}$  is an independently and identically distributed random shock that satisfies  $E_t(\varepsilon_{t+1}) = 0$  and  $E_t(\varepsilon_{t+1}^2) = \sigma^2$ , that is, which has zero mean and variance  $\sigma^2$ .

- a. Using the guess that the value function for this problem depends only on  $y_t$  and not  $\varepsilon_t$  and takes the specific form

$$v(y_t, \varepsilon_t) = v(y_t) = Py_t^2 + d,$$

where  $P$  and  $d$  are unknown constants, the Bellman equation for this problem becomes

$$\begin{aligned} Py_t^2 + d &= \max_{z_t} Ry_t^2 + Qz_t^2 + \beta PE_t[(Ay_t + Bz_t + \varepsilon_{t+1})^2] + \beta d \\ &= \max_{z_t} Ry_t^2 + Qz_t^2 + \beta P(Ay_t + Bz_t)^2 + \beta P\sigma^2 + \beta d \end{aligned}$$

- b. Using the expression on the last line from part (a), the first-order condition for  $z_t$  becomes

$$2Qz_t + 2\beta BP(Ay_t + Bz_t) = 0$$

and the envelope condition for  $y_t$  becomes

$$2Py_t = 2Ry_t + 2\beta AP(Ay_t + Bz_t).$$

Notably, the introduction of uncertainty into the linear-quadratic problem does not alter the form of the optimality conditions for  $z_t$  and  $y_t$ . This special feature of the stochastic LQ model is often referred to as the property of “certainty equivalence,” and it does not apply more generally for problems that do not have the linear-quadratic structure.

c. Rewrite the first-order condition as

$$z_t = - \left( \frac{\beta ABP}{Q + \beta B^2 P} \right) y_t$$

and substitute this result into the envelope condition to obtain

$$P y_t = R y_t + \beta A^2 P y_t - \left( \frac{\beta^2 A^2 B^2 P^2}{Q + \beta B^2 P} \right) y_t$$

or, after combining the last two terms and dividing through by  $y_t$ ,

$$P = R + \frac{\beta A^2 Q P}{Q + \beta B^2 P}.$$

which is the same Riccati equation that helps characterize the problem's solution in the nonstochastic case, and therefore another reflection of the certainly equivalent property.

d. Substituting the first-order condition for  $z_t$  back into the Bellman equation yields

$$P y_t^2 + d = R y_t^2 + Q \left( \frac{\beta ABP}{Q + \beta B^2 P} \right)^2 y_t^2 + \beta P \left[ A - B \left( \frac{\beta ABP}{Q + \beta B^2 P} \right) \right]^2 y_t^2 + \beta P \sigma^2 + \beta d.$$

However, the Riccati equation implies that all of the terms involving  $y_t^2$  cancel out, leaving

$$d = \beta P \sigma^2 + \beta d,$$

which can be solved for

$$d = \left( \frac{\beta}{1 - \beta} \right) P \sigma^2.$$

## 2. Stochastic Growth

The problem is to choose contingency plans for consumption  $c_t$  for all  $t = 0, 1, 2, \dots$  and physical capital  $k_t$  for all  $t = 1, 2, 3, \dots$  to maximize the expected utility function

$$E_0 \sum_{t=0}^{\infty} \beta^t \ln(c_t),$$

with  $0 < \beta < 1$ , subject to the constraints  $k_0$  given and

$$z_t k_t^\alpha \geq c_t + k_{t+1}, \tag{2}$$

for all  $t = 0, 1, 2, \dots$  and all possible realizations of  $z_t$ , where  $E_t[\ln(z_{t+1})] = 0$  for all  $t = 0, 1, 2, \dots$

a. Using the guess that the value function for this problem takes the form

$$v(k_t, z_t) = E + F \ln(k_t) + G \ln(z_t),$$

where  $E$ ,  $F$ , and  $G$  are unknown constants, the Bellman equation for this problem can be written as

$$E + F \ln(k_t) + G \ln(z_t) = \max_{c_t} \ln(c_t) + \beta E + \beta F \ln(z_t k_t^\alpha - c_t),$$

where the term involving  $G E_t[\ln(z_{t+1})]$  drops out of the right-hand-side of the Bellman equation in light of the assumption that  $E_t[\ln(z_{t+1})] = 0$ .

b. Using the Bellman equation as it appears above, the first-order condition for  $c_t$  is

$$\frac{1}{c_t} - \frac{\beta F}{z_t k_t^\alpha - c_t} = 0$$

and the envelope condition for  $k_t$  is

$$\frac{F}{k_t} = \frac{\alpha \beta F z_t k_t^{\alpha-1}}{z_t k_t^\alpha - c_t}.$$

c. Rewrite the first-order condition as

$$c_t = \left( \frac{1}{1 + \beta F} \right) z_t k_t^\alpha$$

and substitute this expression into the envelope condition to obtain

$$F z_t k_t^\alpha - F \left( \frac{1}{1 + \beta F} \right) z_t k_t^\alpha = \alpha \beta F z_t k_t^\alpha,$$

which implies that

$$\frac{1}{1 + \beta F} = 1 - \alpha \beta.$$

In light of this last expression, the first-order condition for  $c_t$  implies

$$c_t = (1 - \alpha \beta) z_t k_t^\alpha$$

and the binding constraint for  $k_{t+1}$  implies

$$k_{t+1} = \alpha \beta z_t k_t^\alpha.$$

These last two expressions confirm that the key result from the perfect foresight case, that with complete depreciation it is optimal to consume the fixed fraction  $1 - \alpha \beta$  of output and to save the remaining fraction  $\alpha \beta$ , carries over to the stochastic case as well.

d. The expression

$$\frac{1}{1 + \beta F} = 1 - \alpha\beta$$

from above implies that

$$F = \frac{\alpha}{1 - \alpha\beta}.$$

Substituting this result, along with the solution for  $c_t$ , back into the Bellman equation yields

$$\begin{aligned} E + F \ln(k_t) + G \ln(z_t) &= \ln(1 - \alpha\beta) + \ln(z_t) + \alpha \ln(k_t) \\ &\quad + \beta E + \beta F \ln(\alpha\beta) + \beta F \ln(z_t) + \alpha\beta F \ln(k_t). \end{aligned}$$

The solution for  $F$  implies that the terms involving  $k_t$  drop out of this last expression, leaving

$$E + G \ln(z_t) = \ln(1 - \alpha\beta) + \ln(z_t) + \beta E + \beta F \ln(\alpha\beta) + \beta F \ln(z_t).$$

Since this last equation must hold for all possible realizations of  $z_t$ , it requires that

$$G \ln(z_t) = \ln(z_t) + \beta F \ln(z_t)$$

or, using the solution for  $F$  again,

$$G = \frac{1}{1 - \alpha\beta}.$$

Hence, the Bellman equation also requires that

$$E = \ln(1 - \alpha\beta) + \beta E + \beta F \ln(\alpha\beta)$$

or, using the solution for  $F$  one last time,

$$E = \frac{1}{1 - \beta} \left[ \ln(1 - \alpha\beta) + \left( \frac{\alpha\beta}{1 - \alpha\beta} \right) \ln(\alpha\beta) \right].$$

### 3. Saving with a Random Return

The problem takes initial assets  $A_0$  as given and chooses contingency plans for  $s_t$  for all  $t = 0, 1, 2, \dots$  and  $A_t$  for  $t = 1, 2, 3, \dots$  to maximize the expected utility function

$$E_0 \sum_{t=0}^{\infty} \beta^t u(c_t) = E_0 \sum_{t=0}^{\infty} \beta^t u(A_t - s_t),$$

with  $0 < \beta < 1$ , subject to the constraints

$$R_{t+1}s_t \geq A_{t+1},$$

which must hold for all  $t = 0, 1, 2, \dots$  and all possible realizations of  $R_{t+1}$ .

- a. Allowing for the possibility of serial correlation in the random return on saving, the Bellman equation for this problem can be written as

$$v(A_t, R_t) = \max_{s_t} u(A_t - s_t) + \beta E_t[v(R_{t+1}s_t, R_{t+1})].$$

Write down the Bellman equation for this problem, using  $A_t$  as the state variable,  $s_t$  and the control variable, and allowing the value function for time  $t$  to depend on  $R_t$  as well as  $A_t$ .

- b. Using the assumed form

$$u(A_t - s_t) = \frac{(A_t - s_t)^{1-\sigma}}{1-\sigma}$$

for the single-period utility function, the additional assumptions that the random interest rate  $R_{t+1}$  is independently and identically distributed with

$$E_t(R_{t+1}^{1-\sigma}) = 1$$

for all  $t = 0, 1, 2, \dots$ , and the guess that under these conditions, the value function depends only on  $A_t$  and takes the specific form

$$v(A_t) = \frac{K A_t^{1-\sigma}}{1-\sigma},$$

where  $K$  is an unknown constant, the Bellman equation can be rewritten as

$$\frac{K A_t^{1-\sigma}}{1-\sigma} = \max_{s_t} \frac{(A_t - s_t)^{1-\sigma}}{1-\sigma} + \frac{\beta K s_t^{1-\sigma}}{1-\sigma}.$$

The first-order condition for  $s_t$  then becomes

$$-(A_t - s_t)^{-\sigma} + \beta K s_t^{-\sigma} = 0$$

and the envelope condition for  $A_t$  becomes

$$K A_t^{-\sigma} = (A_t - s_t)^{-\sigma}.$$

- c. There are a number of ways to derive this result, but one is to start by rewriting the first-order condition for  $s_t$  as

$$s_t = \left[ \frac{(\beta K)^{1/\sigma}}{1 + (\beta K)^{1/\sigma}} \right] A_t$$

and substitute this result into the envelope condition to obtain

$$K A_t^{-\sigma} = \left[ \frac{1}{1 + (\beta K)^{1/\sigma}} \right]^{-\sigma} A_t^{-\sigma},$$

which can be solved for

$$K = \left( \frac{1}{1 - \beta^{1/\sigma}} \right)^\sigma.$$

d. Since the solution for  $K$  just derived implies that

$$(\beta K)^{1/\sigma} = \frac{\beta^{1/\sigma}}{1 - \beta^{1/\sigma}},$$

the first-order condition for  $s_t$  implies that the optimal choice for  $s_t$  is given by

$$s_t = \beta^{1/\sigma} A_t.$$

Hence, the optimal choice for  $c_t$  is

$$c_t = (1 - \beta^{1/\sigma}) A_t.$$