

# Day 21

## Dynamic Programming

Kerry Back  
BUSI 521–ECON 505  
Rice University  
Spring 2022

## Optimal Capital Investment

A firm combines capital  $k$  and labor  $\ell$  to produce output  $q = k^\alpha \ell^\psi$  with  $\alpha + \psi \leq 1$ . The wage rate  $w$  is constant.

The firm is a monopolist facing a constant elasticity demand curve:

$$\frac{d \log q}{d \log p} = \gamma \quad \Rightarrow \quad p = a q^{-1/\gamma}$$

for a constant (soon to be stochastic process)  $a$ . We call  $pq - w\ell$  the firm's operating cash flow.

Capital depreciates over time. Also, adjusting the capital stock is costly. To adjust by  $\Delta k$  costs  $\Delta k + (1/2)(\Delta k)^2 b$ . We call  $-\Delta k - (1/2)(\Delta k)^2 b$  the firm's investment cash flow.  $\Delta k$  can be negative.

Total cash flow is operating cash flow plus financing cash flow.

We formulate the problem in continuous time. The firm chooses an investment rate  $I_t$  and capital  $K_t$  evolves as

$$dK_t = -\delta K_t dt + I_t dt$$

Given the capital stock  $K$  at time  $t$ , the firm chooses the labor input  $L_t$  to maximize the operating cash flow  $pq - w\ell$ , i.e.,

$$L_t = \operatorname{argmax}_{\ell} a (K_t^{\alpha} \ell^{\psi})^{1-1/\gamma} - w\ell$$

There is random time-varying demand: the constant  $a$  is a stochastic process  $A$ . The firm chooses an investment policy to maximize

$$E \int_0^{\infty} M_t [\text{Operating Cash Flow} - I_t - (1/2)bl_t^2] dt$$

where  $M$  is the SDF process.

Assume the risk-free rate is constant. By switching to the risk-neutral probability, we can formulate the firm's objective as

$$E^* \int_0^{\infty} e^{-rt} [\text{Operating Cash Flow} - I_t - (1/2)bl_t^2] dt$$

Assume  $A$  is a geometric Brownian motion (hence Markov) under the risk-neutral probability. The state variables are  $A_t$  and  $K_t$ .

The problem is time-stationary, so the firm's value function is of the form  $J(A_t, K_t)$ .

What sort of solution can we find?

We want to find the optimal investment rate  $I_t$  as a function of  $A_t$  and  $K_t$ .

We can show that the optimum satisfies

$$I_t^* = \frac{1}{b} \left[ \frac{\partial J(A_t, K_t)}{\partial K_t} - 1 \right]$$

Can we compute  $\partial J / \partial K$ ?

Not analytically. But, we can show that  $J$  satisfies a partial differential equation (PDE), which can be solved numerically.

The PDE is the Bellman equation, called Hamilton-Jacobi-Bellman (HJB) equation in continuous time.

The HJB equation is

$$\max_{\text{controls}} \left\{ \text{current reward} - rJ + \text{drift of } J \right\} = 0$$

The optimal controls must achieve the max here.

Let  $\pi(A_t, K_t)$  denote the operating cash flow maximized over  $L_t$ . Then the current reward is

$$\pi(A_t, K_t) - l_t - \frac{1}{2} b l_t^2$$

The control is  $l_t$ .

The drift of  $J$  is

$$\frac{\partial J}{\partial A} \text{drift of } A + \frac{\partial J}{\partial K} (-\delta K + I) + \frac{1}{2} \frac{\partial^2 J}{\partial A^2} (dA)^2$$

dropping the  $dt$  from  $(dA)^2$ .

Substituting into the HJB equation and dropping terms that do not depend on  $I$ , the maximization problem is

$$\max_I \left\{ -I - \frac{1}{2}bI^2 + \frac{\partial J}{\partial K}I \right\}$$

The solution is

$$I = \frac{\partial J / \partial K - 1}{b}$$

Write  $J_k$  for the partial derivative. So, the solution is  $I = (J_k - 1)/b$ .

At the max  $I = (J_k - 1)/b$ , the HJB equation tells us

$$\text{current reward} - rJ + \text{drift of } J = 0$$

This is a PDE in  $J$ .

Write  $J_a$  for  $\partial J / \partial A$  and  $J_{aa}$  for the second partial. Then, suppressing the arguments  $(A, K)$ , we have

$$\begin{aligned} \pi - \frac{J_k - 1}{b} - \frac{1}{2}b \left( \frac{J_k - 1}{b} \right)^2 - rJ \\ + J_a \text{drift of } A + \left( \frac{J_k - 1}{b} - \delta K \right) J_k + \frac{1}{2} J_{aa} (dA)^2 = 0 \end{aligned}$$

This is an equation that must hold for  $A > 0$  and  $K > 0$ , with boundary conditions at 0 and  $\infty$ .



## HJB Equation

Discrete-time, no uncertainty, control  $c$ , state variable  $x$ , and state evolution equation  $x' = f(x, c)$ :

$$J(x) = \max_c u(c) + e^{-\delta} J(x')$$

With uncertainty and expected utility

$$J(x) = \max_c u(c) + e^{-\delta} E[J(x') \mid x, c]$$

Equivalently,

$$0 = \max_c u(c) + E[J(x') - J(x) \mid x, c] - (1 - e^{-\delta}) E[J(x') \mid x, c]$$

In continuous time,

$$0 = \max_c u(c) + \text{drift of } J - \delta J$$

## Another Derivation

Let  $C$  control process,  $X$  = state process,  $u$  = utility function,  $\delta$  = discount rate ( $r$  in capital investment problem),  $E$  = appropriate expectation. Problem is

$$\max_C E \int_0^\infty e^{-\delta t} u(C_t, X_t) dt$$

where evolution of  $X$  depends on current  $X$  and  $C$ .

Let  $J$  denote value function. Suppose there is an optimal policy  $C^*$ , which induces the state process  $X^*$ , so

$$J(X_0) = E \int_0^\infty e^{-\delta t} u(C_t^*, X_t^*) dt$$

Bellman's principle of optimality states that, for any  $s$ ,

$$J(X_0) = E \left[ \int_0^s e^{-\delta t} u(C_t^*, X_t^*) dt + \max_C E_s \int_s^\infty e^{-\delta t} u(C_t, X_t) dt \right]$$

In other words, if a plan is optimal, then if we re-optimize at some later date, we cannot improve upon the original plan.

So,

$$J(X_0) = E \left[ \int_0^s e^{-\delta t} u(C_t^*, X_t^*) dt + e^{-\delta s} J(X_s) \right]$$

So, the expectation is constant with respect to  $s$ . It is not very much extra work to show that, in fact,

$$\int_0^s e^{-\delta t} u(C_t^*, X_t^*) dt + e^{-\delta s} J(X_s)$$

is a martingale.

A martingale has no drift. Calculate the drift and set to zero:

$$e^{-\delta s} u(C_s^*, X_s^*) + e^{-\delta s} \text{drift of } J - \delta e^{-\delta s} J = 0$$

Multiply by  $e^{\delta s}$  to cancel the  $e^{-\delta s}$  factor:

$$u(C_s^*, X_s^*) + \text{drift of } J - \delta J = 0$$

This is almost the HJB equation. The actual HJB equation states that the maximum drift is zero.

The maximum drift is zero, because if you could get a positive drift, you would contradict the principle of optimality (re-optimizing would improve upon the original plan).