

ORF 418 : OPTIMAL LEARNING

LECTURE 4 : September 15 , 2025

RICATTI EQUATIONS

- 1) Dynamic Programming
- 2) Linear Ansatz
- 3) Solution



I. DYNAMIC PROGRAMMING

We introduce dynamic programming in the LQ setting as a general solution technique. It will be revisited later in the context of MDPs.

We use the same notation as before and consider the infinite horizon problem.

$$x_{k+1} = Ax_k + Bu_k,$$
$$J(x_0, u) = \sum_{k=0}^{\infty} \rho^k (x_k^T N x_k + u_k^T N u_k)$$

and define

$$v(x_0) := \inf_u J(x_0, u) \quad (= v_0(x_0) \text{ but } \infty \text{ is dropped})$$

THEOREM (2.2.1) Suppose $v(x_0) < \infty$. Then it is the unique solution to

$$v(x_0) = x_0^T N x_0 + \inf_{u \in \mathbb{R}^e} \{ u^T N u + \rho v(Ax_0 + Bu) \}$$

Moreover, the optimal control is given by

$$u_0^*(x_0) \in \operatorname{argmin}_{u \in \mathbb{R}^e} \{ u^T N u + \rho v(Ax_0 + Bu) \}.$$

Basic idea of the proof is this:

$$J(x_0, u) = x_0^T M x_0 + u_0^T N u_0$$

$$+ \rho \sum_{k=1}^{\infty} \rho^{-k} (x_k^T M x_k + u_k^T N u_k)$$

$= J(x_1, \bar{u})$

where $\bar{u} = (u_1, u_2, \dots)$, i.e., u shifted by one.

The formal proof is given in the notes.

II. LINEAR ANSATZ (In this section, we take $\rho=1$)

We postulate that

$$v(x) = x^T V x, \quad x \in \mathbb{R}^d$$

for some $(d \times d)$, symmetric, positive-definite matrix V ,

where x^T is the transpose. We substitute this form

into the dynamic programming equation:

$$x^T V x = x^T M x + \inf_u \{ u^T N u + (Ax + Bu)^T V (Ax + Bu) \}.$$

The minimizer u^* satisfies

$$N u^* + B^T V A x + B^T V B u^* = 0$$

Hence, $u^* = -F \cdot x$, $F = (N + B^T V B)^{-1} B^T V A$

Substitute this back into the equation to obtain

$$x^T V x = x^T M x + x^T F^T N F x + x^T (A - B F)^T V (A - B F) x.$$

Since above holds for every x , we conclude that

$$\begin{aligned} V &= M + F^T N F + (A^T - F^T B^T) V (A - B F) \\ &= M + F^T (N + B^T V B) F + A^T V A \\ &\quad - F^T B^T V A - A^T V B F. \end{aligned}$$

We substitute the formula for F into the above equation. First observe that

$$F^T (N + B^T V B) F = A^T V B (N + B^T V B)^{-1} B^T V A,$$

$$F^T B^T V A = A^T V B (N + B^T V B)^{-1} B^T V A,$$

$$A^T V B F = A^T V B (N + B^T V B)^{-1} B^T V A.$$

Hence, V satisfies

$$V = M + A^T V A - A^T V B (N + B^T V B)^{-1} B^T V A$$

This equation is known as the **homogenous Riccati** equation.

III. SOLUTION.

We summarize the above calculations as follows:

(i) Suppose that the homogenous Riccati equation has a solution V .

(ii) Then $v(x) = \pi^T V \pi$ solves the dynamic programming equation. Since this equation has a unique solution which is equal to the value function, V is unique.

(iii) Optimal feedback control is given by

$$u^*(x) = -Fx, \quad F = p(N + pB^T V B)^{-1} B^T V A.$$

a) One-dimensional example.

There $d=l=1$, $A=4$, $B=M=N=1$, $p=\frac{1}{2}$. Then the dynamic programming equation is

$$v(x) = x^2 + \inf_u \left\{ u^2 + \frac{1}{2} v(4x+u) \right\}.$$

We postulate that $v(x) = \pi x^2$. Then,

$$\pi x^2 = x^2 + \inf_u \left\{ u^2 + \frac{1}{2} \pi (4x+u)^2 \right\}$$

By differentiating at the minimizer u^* :

$$2u^* + \pi(4x+u^*) = 0 \quad \Rightarrow \quad u^* = -\frac{4\pi}{2+\pi} x =: -fx$$

Hence $f = 4v(r+v)^{-1}$ which agrees with the

general formula. Also

$$\begin{aligned} 8vx^2 &= x^2 + f^2 x^2 + \frac{1}{2} v (4-f)^2 x^2 \\ &= x^2 \left[1 + f^2 \left(1 + \frac{1}{2} v \right) + 8v - 4fv \right] \end{aligned}$$

\Rightarrow

$$v = 1 + 8v - \frac{8v^2}{2+v}$$

This equation also agrees with the general formula.

Moreover, it reduces to the following quadratic equation:

$$8v^2 = (2+v)(7v+1) = 7v^2 + 15v + 2$$

$$\Rightarrow v^2 - 15v - 2 = 0$$

Its positive solution $v \approx 3.533$ agrees with the previous calculation from the previous lecture.