

## Part II Dynamic optimization.

We consider a controlled dynamic system:

$$\underline{X_{n+1} = f_n(X_n, U_n)}$$

$$\left. \begin{array}{l} \text{Static optimization} \\ \min_{x \in K} f(x) \\ K \subseteq \mathbb{R}^n. \end{array} \right\}$$

$n$ : time.  $n = 0, 1, 2, \dots$

$X_n$ : state of the system at time  $n$ .

$$X_n \in X \subseteq \mathbb{R}^d$$

$U_n$ : the decision / the control chosen at time  $n$ .

$$U_n \in U_n$$

$x_0$ : the initial state. ( $X_0 = x_0$ )

$$f_n: X \times U_n \rightarrow X.$$

dynamic optimization problem,  
1. Finite Horizon

$$\min_{(U_n)_{n=0,1,\dots}} \left( \sum_{n=0}^{N-1} L_n(X_n, U_n) + g(X_N) \right)$$

2. Infinite horizon:

$$\min_{(U_n)_{n=0,1,\dots}} \left( \sum_{n=0}^{\infty} \beta^n L_n(X_n, U_n) \right), \text{ for } \beta \in (0, 1)$$

discount factor.

$$- L_n: X \times U_n \rightarrow \mathbb{R}.$$

$$- g: X \rightarrow \mathbb{R}.$$

$$\underline{X_{n+1}^{k,u} = f_n(X_n^{k,u}, U_n)}$$

# 1. Finite horizon problem.

$$X_k^{k,x,u} = x.$$

Value function:  $V(k, x) := \inf_{(u_n)_{n=k, k+1, \dots}} \left( \sum_{n=k}^{N-1} L_n(x_n, u_n) + g(x_N) \right)$

$X_k^{k,x,u} = x.$

Theorem: (Dynamic programming). For all  $k=0, 1, \dots, N-1$ ,  $x \in X$ .

one has. —  $V(k, x) = \inf_{u_k \in U_k} \left( L_k(x, u_k) + V(k+1, X_{k+1}^{k,x,u}) \right).$

$= \inf_{u_k \in U_k} \left( L_k(x, u_k) + V(k+1, f_k(x, u_k)) \right)$

$- V(N, x) = g(x) \rightarrow$  terminal value.

Proof: ① " $\geq$ " Let  $u = (u_n)_{n=k, k+1, \dots}$  be given arbitrarily.

Then:  $\sum_{n=k}^{N-1} L_n(x_n^{k,x,u}, u_n) + g(x_N^{k,x,u})$

$= L_k(x_k^{k,x,u}, u_k) + \left( \sum_{n=k+1}^{N-1} L_n(x_n^{k,x,u}, u_n) + g(x_N^{k,x,u}) \right)$

$$-L_k(x, u_k) + \sum_{n=k+1}^N L_n(X_n, u_n) + g(X_N) \geq L_k(x, u_k) + V(k+1, X_{k+1}^{k,x,u})$$

$$X_n^{k,x,u} = X_n^{k+1, X_{k+1}^{k,x,u}, u} \rightarrow \text{flow property,}$$

$$\left\{ \begin{array}{l} X_k = x \\ X_{k+1} = f_k(X_k, u_k) \\ X_{k+2} = f_{k+1}(X_{k+1}, u_{k+1}) \end{array} \right\} \quad \left\{ \begin{array}{l} X_{k+1} = X_{k+1}^{k,x,u} = f_k(X_k, u_k) \\ X_{k+2} = f_{k+1}(X_{k+1}, u_{k+1}) \end{array} \right.$$

Taking infimum on both sides over  $u$ , it follows that

$$V(k, x) \geq \inf_{u_k \in U_k} (L_k(x, u_k) + V(k+1, X_{k+1}^{k,x,u}))$$

② " $\leq$ ". Denote  $\bar{V}(k, x) := \inf_{u_k \in U_k} (L_k(x, u_k) + V(k+1, X_{k+1}^{k,x,u}))$

For any  $\varepsilon > 0$ , there exists  $u_k^\varepsilon \in U_k$  s.t.

$$\bar{V}(k, x) + \varepsilon \geq L_k(x, u_k^\varepsilon) + V(k+1, X_{k+1}^\varepsilon)$$

with  $X_{k+1}^\varepsilon := X_{k+1}^{k,x,u_k^\varepsilon}$

Next, by the definition of  $V(k+1, \cdot)$ , there exists  $u_{k+1}^\varepsilon, \dots, u_{N-1}^\varepsilon$ ,

$$\text{s.t. } \underline{V}(k+1, \underline{x}_{k+1}^\varepsilon) + \varepsilon \geq \sum_{n=k+1}^{N-1} L_n(x_n^\varepsilon, u_n^\varepsilon) + g(x_N^\varepsilon)$$

$$\text{with } x_n^\varepsilon := x_n^{k+1}, \underline{x}_{k+1}^\varepsilon, u_n^\varepsilon$$

$$\Rightarrow \bar{V}(k, x) + \varepsilon + \underline{V}(k+1, \underline{x}_{k+1}^\varepsilon) + \varepsilon$$

$$\geq \underline{L}_k(x, u_k^\varepsilon) + \underline{V}(k+1, \underline{x}_{k+1}^\varepsilon) + \sum_{n=k+1}^{N-1} L_n(x_n^\varepsilon, u_n^\varepsilon) + g(x_N^\varepsilon)$$

$$\Rightarrow \bar{V}(k, x) + 2\varepsilon \geq \sum_{n=k}^{N-1} L_n(x_n^\varepsilon, u_n^\varepsilon) + g(x_N^\varepsilon) \geq \underline{V}(k, x)$$

By the arbitrariness of  $\varepsilon > 0$ , one has

$$\underline{V}(k, x) \leq \bar{V}(k, x) = \inf_{u_k \in U_k} ( \underline{L}_k(x, u_k) + \underline{V}(k+1, x_{k+1}^{kx, u_k}) )$$

Additional conditions:

$X, U_n, n=0, 1, \dots, N-1$  are all metric spaces.

$U_n, n=1, \dots, N-1$  are compact.

$$\left\{ \begin{array}{l} f_n: X \times U_n \rightarrow X, \\ L_n: X \times U_n \rightarrow \mathbb{R}. \end{array} \right.$$

are continuous.

$$g: X \rightarrow \mathbb{R}$$

Proposition: Assume the above additional conditions. Then.

①  $x \mapsto V(k, x)$  is continuous for each  $k=0, 1, \dots, N$ .

and there exists  $U_k^*(x)$ , s.t.  $V(k, x) = \underbrace{L_k(x, U_k^*(x)) + V(k+1, f_{k+1}(x, U_k^*(x)))}_{\text{for } k=0, 1, \dots, N-1}$ .

② Define  $\bar{X}_0 = x$ , and  $\bar{X}_{n+1} = f_n(\bar{X}_n, \bar{u}_n)$  with  $\bar{u}_n := U_n^*(\bar{X}_n)$ .

Then,  $(\bar{X}_n, \bar{u}_n)_{n=0, 1, \dots, N-1}$  is an optimal solution to the dynamic optimization problem. i.e.

$$V(0, x) = \sum_{n=0}^{N-1} L_n(\bar{X}_n, \bar{u}_n) + g(\bar{X}_N).$$

Proof: ①. -  $V(N, x) := g(x)$  is a continuous function.

- Assume that  $x \mapsto V(k+1, x)$  is continuous. then by Thm. of DPP.

$$V(k, x) = \inf_{u \in U_k} J_k(x, u) \quad \text{where } J_k(x, u) := L_k(x, u) + V(k+1, f_k(x, u))$$

is a continuous function in  $(x, u)$

Recall that  $U_k$  is compact.

Then, for all  $x \in X$ , there exists  $U_k^*(x) \in U_k$ .

$$\text{s.t. } \underline{V(k, x) = \mathcal{J}_k(x, U_k^*(x))}$$

and we claim that  $x \mapsto V(k, x)$  is also continuous

Indeed, let  $x_n \rightarrow x$ , then  $\underline{V(k, x_n)} \leq \mathcal{J}_k(x_n, \underline{U_k^*(x)}) \xrightarrow{n \rightarrow +\infty} \mathcal{J}_k(x, U_k^*(x))$

$$\text{and } \underline{\lim_{n \rightarrow +\infty} V(k, x_n)} = \underline{\lim_{n \rightarrow +\infty} \mathcal{J}_k(x_n, \underline{U_k^*(x_n)})} = \mathcal{J}_k(x, \bar{U}^*) \geq V(k, x).$$

$\Rightarrow \lim_{n \rightarrow +\infty} V(k, x_n) = V(k, x)$ , i.e.  $x \mapsto V(k, x)$  is continuous.

② To prove that  $(\bar{X}_n, \bar{U}_n)_{n=0,1,\dots}$  is an optimal solution, it is enough to prove that

$$\textcircled{*} \quad \underline{V(0, x_0) = \sum_{n=0}^{m-1} L_n(\bar{X}_n, \bar{U}_n) + \underline{V(m, \bar{X}_m)}}, \quad \forall m=1, 2, \dots, N.$$

Indeed,  $\textcircled{*}$  is true for  $m=0$

Next, assume that  $\textcircled{*}$  is true for  $\underline{m}$ . i.e.  $V(0, x_0) = \sum_{n=0}^{m-1} L_n(\bar{X}_n, \bar{U}_n)$

Since  $V(m, \bar{X}_m) = \mathcal{J}_m(\bar{X}_m, \bar{U}_m) + \underline{V(m, \bar{X}_m)}$

$$= \underline{L_m(\bar{X}_m, \bar{u}_m) + V(m+1, \bar{X}_{m+1})}$$

$$\Rightarrow V(x_0) = \sum_{n=0}^{m-1} L_n(\bar{X}_n, \bar{u}_n) + L_m(\bar{X}_m, \bar{u}_m) + V(m+1, \bar{X}_{m+1})$$

$$= \sum_{n=0}^m L_n(\bar{X}_n, \bar{u}_n) + V(m+1, \bar{X}_{m+1})$$

$\Rightarrow$   $(*)$  is also true for  $m+1$ .

By induction,  $(*)$  is true for all  $m=0, 1, \dots, N$   $\#$

2. Infinite horizon problem:

$$X_{n+1} = f(X_n, u_n),$$

Assume that  $(f_n, L_n, U_n)$  are independent  $n$ .  
 $= (f, L, U)$  for all  $n=0, 1, \dots$   
 $u_n \in U$ .  $L$  is uniformly bounded.

$$\inf_{(u_n)_{n=0,1,\dots}} \sum_{n=0}^{+\infty} \beta^n L(X_n, u_n)$$

$$\beta \in (0, 1)$$

Value function:  $V(x) := \inf_{(u_n)_{n=0,1,\dots}} \sum_{n=0}^{+\infty} \beta^n L(X_n^{oxu}, u_n)$

- oxu

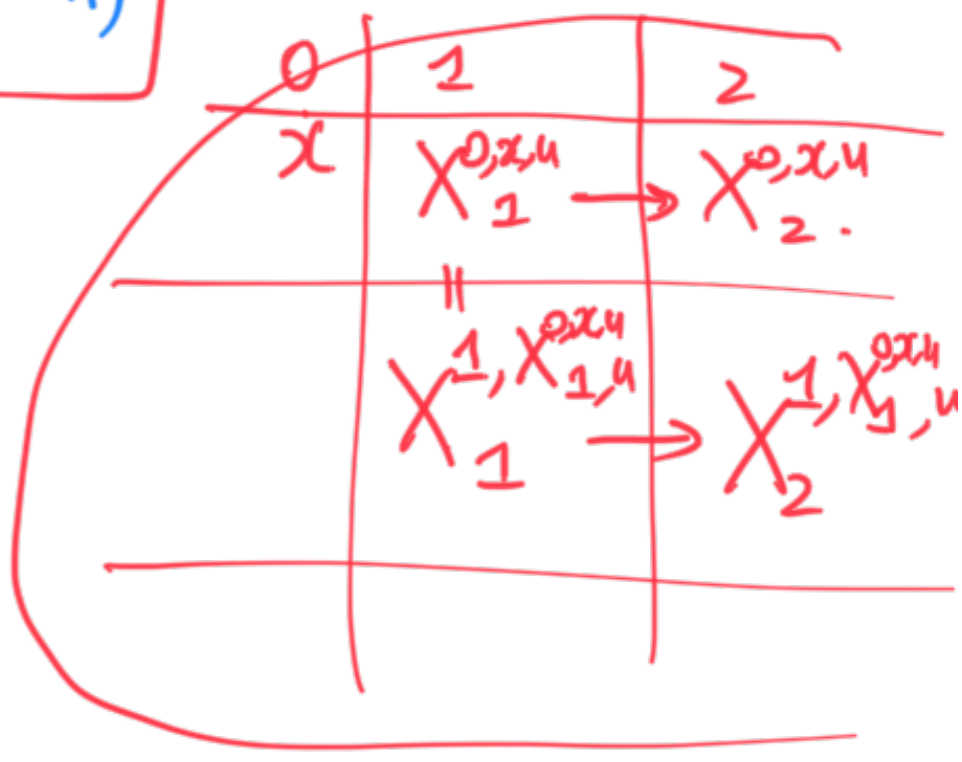
$$X_0 = x.$$

$$X_{n+1} = f(X_n, u_n)$$

Theorem. (DPP) For all  $x \in X$ , one has

$$V(x) = \inf_{u_0 \in U} (L(x, u_0) + \beta V(X_1^{0,x,u_0}))$$

$$= \inf_{u_0 \in U} (L(x, u_0) + \beta V(f(x, u_0))).$$



Proof: ① Let  $(u_n)_{n=0,1,\dots}$  be an arbitrary control process.

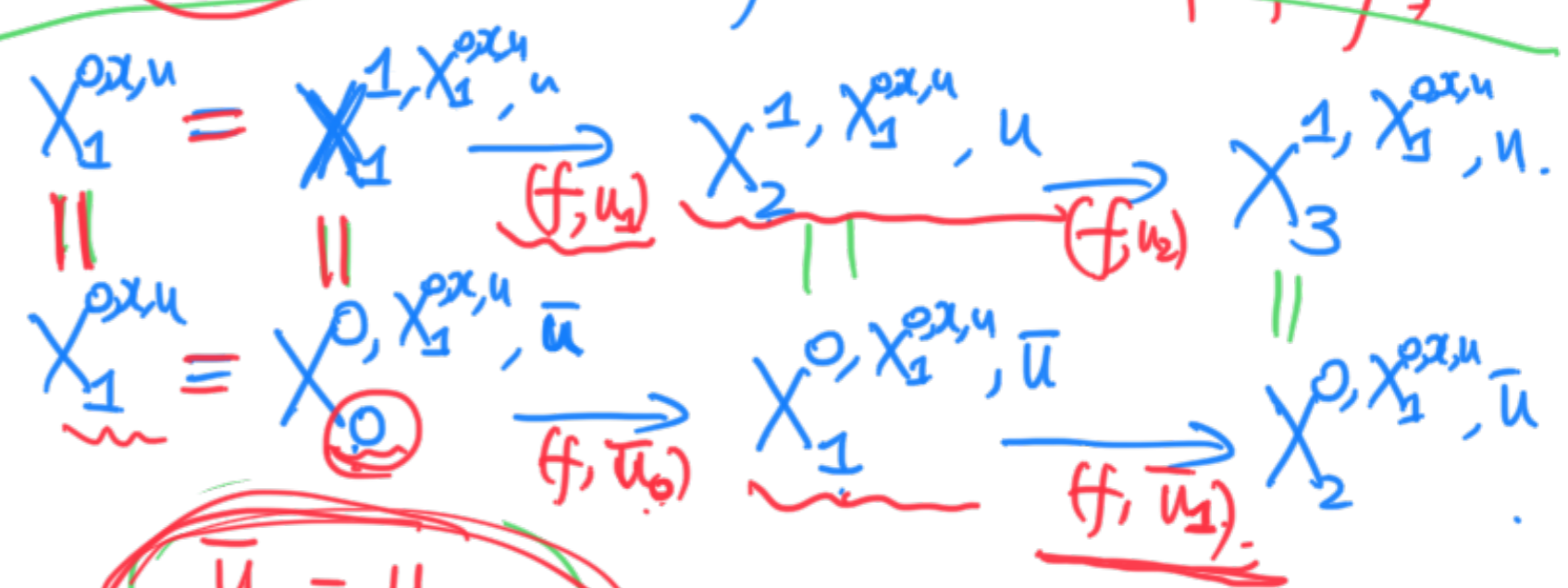
$$\text{Then } \sum_{n=0}^{+\infty} \beta^n L(X_n^{0,x,u}, u_n) = L(x, u_0) + \sum_{n=1}^{+\infty} \beta^n L(X_n^{0,x,u}, u_n)$$

$$= L(x, u_0) + \beta \left( \sum_{n=1}^{+\infty} \beta^{n-1} L(X_n^{1,X_1^{0,x,u}}, u_n) \right) \quad (\text{flow property})$$

$$X_2^{1,X_1^{0,x,u}, u} = f(X_1^{0,x,u}, u_1)$$

$$X_1^{0,x,u}, \bar{u} = f(X_1^{0,x,u}, \bar{u}_0)$$

$$= f(X_1^{0,x,u}, u_1)$$



$$\bar{u}_n = u_{n+1}$$



$$\begin{aligned}
&= L(x, u_0) + \beta \left( \sum_{n=1}^{+\infty} \beta^{n-1} L(X_{n-1}^0, X_1^{\alpha x, u}, \bar{u}, \bar{u}_{n-1}) \right) \\
&= L(x, u_0) + \beta \left( \sum_{n=0}^{+\infty} \beta^n L(X_n^0, X_1^{\alpha x, u}, \bar{u}, \bar{u}_n) \right) \\
&\geq L(x, u_0) + \beta \cdot \underbrace{V(X_1^{\alpha x, u})}
\end{aligned}$$

Taking infimum on both sides over  $u$ , it follows that

$$V(x) \geq \underbrace{\inf_{u_0 \in U} \left( L(x, u_0) + \beta V(\underbrace{f(x, u_0)}_{X_1^{\alpha x, u_0}}) \right)}_{\triangleq \bar{V}(x)}$$

(2) For any  $\varepsilon > 0$

there exists  $u_0^\varepsilon$  s.t.  $\bar{V}(x) + \varepsilon \geq L(x, u_0^\varepsilon) + \beta V(X_1^{\alpha x, u_0^\varepsilon})$

Further, there exist  $(u_1^\varepsilon, u_2^\varepsilon, \dots)$  s.t.

$$V(X_1^{\alpha x, u_0^\varepsilon}) + \varepsilon \geq \underbrace{\sum_{n=1}^{+\infty} \beta^{n-1} L(X_{n-1}^0, X_1^{\alpha x, u_0^\varepsilon}, u^\varepsilon, u_n^\varepsilon)}$$

$$\Rightarrow \bar{V}(x) + 2\varepsilon \geq \sum_{n=0}^{+\infty} \beta^n L(X_n^0, X_1^{\alpha x, u^\varepsilon}, u_n^\varepsilon)$$

Since  $\Sigma$  is arbitrary, it follows that

$$\underline{V}(x) \leq \sum_{n=0}^{+\infty} \beta^n L(x_n^{\sigma x, u^n}, u_n^\Sigma) \leq \bar{V}(x) = \inf_{u_0 \in U} \left( L(x, u_0) + \beta V(x_1^{\sigma x, u_0}) \right)$$

(DP equation): 
$$\underline{V}(x) = \inf_{u_0 \in U} \left( L(x, u_0) + \beta V(x_1^{\sigma x, u_0}) \right)$$

Theorem: Assume that  $L$  is uniformly bounded,  $\beta \in (0, 1)$ .

Then, there exists a unique solution  $V: X \rightarrow \mathbb{R}$ , to the DP equation in the space of  $B(X) := \{ \text{all functions } h: X \rightarrow \mathbb{R} \text{ bounded} \}$ .

Remark: let  $\|h\|_\infty := \sup_{x \in X} |h(x)|$ . then  $(B(X), \|\cdot\|_\infty)$  is a Banach space.

Proof: ① Let  $T: B(X) \rightarrow B(X)$  be defined by

$$\underline{T(h)}(x) := \inf_{u_0 \in U} \left( \underline{L}(x, u_0) + \beta \underline{h}(f(x, u_0)) \right)$$

So that the DP equation is

$V = T(V)$ ,  
i.e. a fixed point of  $T$ .

(2) It is enough to prove that  $T: B(X) \rightarrow B(X)$  is a contraction mapping.

Let  $h_1, h_2 \in B(X)$ . then  $T(h_1)(x) - T(h_2)(x)$

$$= \inf_{u \in U} (L(x, u) + \beta h_1(f(x, u)))$$

$$- \inf_{u \in U} (L(x, u) + \beta h_2(f(x, u))).$$

$$\leq L(x, u_0^{2\varepsilon}) + \beta h_1(f(x, u_0^{2\varepsilon}))$$

$$- \left( L(x, u_0^{2\varepsilon}) + \beta h_2(f(x, u_0^{2\varepsilon})) \right) + \varepsilon.$$

$$\leq \beta \|h_1 - h_2\|_\infty + \varepsilon.$$

Similarly,  $T(h_2)(x) - T(h_1)(x) \leq \beta \|h_1 - h_2\|_\infty$

$$\Rightarrow \|T(h_1) - T(h_2)\|_\infty \leq \beta \|h_1 - h_2\|_\infty \quad \text{for } \beta \in (0, 1)$$

Therefore,  $T$  is a contraction mapping.  $\#$

