

6.231 DYNAMIC PROGRAMMING

LECTURE 10

LECTURE OUTLINE

- Problems with imperfect state info
- Reduction to the perfect state info case
- Machine repair example

BASIC PROBLEM WITH IMPERFECT STATE INFO

- Same as basic problem of Chapter 1 with one difference: the controller, instead of knowing x_k , receives at each time k an observation of the form

$$z_0 = h_0(x_0, v_0), \quad z_k = h_k(x_k, u_{k-1}, v_k), \quad k \geq 1$$

- The observation z_k belongs to some space Z_k .
- The random observation disturbance v_k is characterized by a probability distribution

$$P_{v_k}(\cdot \mid x_k, \dots, x_0, u_{k-1}, \dots, u_0, w_{k-1}, \dots, w_0, v_{k-1}, \dots, v_0)$$

- The initial state x_0 is also random and characterized by a probability distribution P_{x_0} .
- The probability distribution $P_{w_k}(\cdot \mid x_k, u_k)$ of w_k is given, and it may depend explicitly on x_k and u_k but not on $w_0, \dots, w_{k-1}, v_0, \dots, v_{k-1}$.
- The control u_k is constrained to a given subset U_k (this subset does not depend on x_k , which is not assumed known).

INFORMATION VECTOR AND POLICIES

- Denote by I_k the *information vector*, i.e., the information available at time k :

$$I_k = (z_0, z_1, \dots, z_k, u_0, u_1, \dots, u_{k-1}), \quad k \geq 1,$$

$$I_0 = z_0.$$

- We consider policies $\pi = \{\mu_0, \mu_1, \dots, \mu_{N-1}\}$, where each function μ_k maps the information vector I_k into a control u_k and

$$\mu_k(I_k) \in U_k, \quad \text{for all } I_k, \quad k \geq 0.$$

- We want to find a policy π that minimizes

$$J_\pi = \underset{\substack{x_0, w_k, v_k \\ k=0, \dots, N-1}}{E} \left\{ g_N(x_N) + \sum_{k=0}^{N-1} g_k(x_k, \mu_k(I_k), w_k) \right\}$$

subject to the equations

$$x_{k+1} = f_k(x_k, \mu_k(I_k), w_k), \quad k \geq 0,$$

$$z_0 = h_0(x_0, v_0), \quad z_k = h_k(x_k, \mu_{k-1}(I_{k-1}), v_k), \quad k \geq 1$$

EXAMPLE: MULTIACCESS COMMUNICATION I

- Collection of transmitting stations sharing a common channel, are synchronized to transmit packets of data at integer times.
- x_k : backlog at the beginning of slot k .
- a_k : random number of packet arrivals in slot k .
- t_k : the number of packets transmitted in slot k .

$$x_{k+1} = x_k + a_k - t_k,$$

- At k th slot, each of the x_k packets in the system is transmitted with probability u_k (common for all packets). If two or more packets are transmitted simultaneously, they collide.
- So $t_k = 1$ (a success) with probability $x_k u_k (1 - u_k)^{x_k - 1}$, and $t_k = 0$ (idle or collision) otherwise.
- Imperfect state info: The stations can observe the channel and determine whether in any one slot there was a collision (two or more packets), a success (one packet), or an idle (no packets).

EXAMPLE: MULTIACCESS COMMUNICATION II

- Information vector at time k : The entire history (up to k) of successes, idles, and collisions. Mathematically, z_{k+1} , the observation at the end of the k th slot, is

$$z_{k+1} = v_{k+1}$$

where v_{k+1} yields an idle with probability $(1 - u_k)^{x_k}$, a success with probability $x_k u_k (1 - u_k)^{x_k - 1}$, and a collision otherwise.

- If we had perfect state information, the DP algorithm would be

$$J_k(x_k) = g_k(x_k) + \min_{0 \leq u_k \leq 1} E \left\{ p(x_k, u_k) J_{k+1}(x_k + a_k - 1) + (1 - p(x_k, u_k)) J_{k+1}(x_k + a_k) \right\},$$

$p(x_k, u_k)$ is the success probability $x_k u_k (1 - u_k)^{x_k - 1}$.

- The optimal (perfect state information) policy would be to select the value of u_k that maximizes $p(x_k, u_k)$, so $\mu_k(x_k) = \frac{1}{x_k}$, for all $x_k \geq 1$.
- Imperfect state info problem is much harder.

REFORMULATION AS A PERFECT INFO PROBLEM

- We have

$$I_{k+1} = (I_k, z_{k+1}, u_k), \quad k = 0, 1, \dots, N-2, \quad I_0 = z_0.$$

View this as a dynamic system with state I_k , control u_k , and random disturbance z_{k+1} .

- We have

$$P(z_{k+1} \mid I_k, u_k) = P(z_{k+1} \mid I_k, u_k, z_0, z_1, \dots, z_k),$$

since z_0, z_1, \dots, z_k are part of the information vector I_k . Thus the probability distribution of z_{k+1} depends explicitly only on the state I_k and control u_k and not on the prior “disturbances” z_k, \dots, z_0 .

- Write

$$E \{ g_k(x_k, u_k, w_k) \} = E \left\{ E \left\{ g_k(x_k, u_k, w_k) \mid I_k, u_k \right\} \right\}$$

so the cost per stage of the new system is

$$\tilde{g}_k(I_k, u_k) = E_{x_k, w_k} \left\{ g_k(x_k, u_k, w_k) \mid I_k, u_k \right\}$$

DP ALGORITHM

- Writing the DP algorithm for the (reformulated) perfect state info problem and doing the algebra:

$$J_k(I_k) = \min_{u_k \in U_k} \left[\begin{array}{l} E \\ x_k, w_k, z_{k+1} \end{array} \left\{ g_k(x_k, u_k, w_k) \right. \right. \\ \left. \left. + J_{k+1}(I_k, z_{k+1}, u_k) \mid I_k, u_k \right\} \right]$$

for $k = 0, 1, \dots, N - 2$, and for $k = N - 1$,

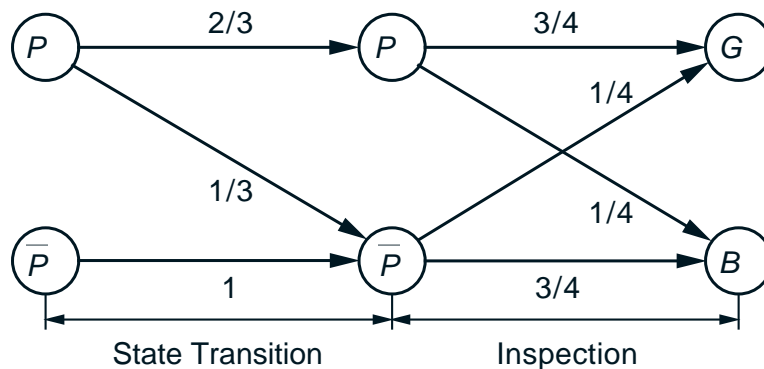
$$J_{N-1}(I_{N-1}) = \min_{u_{N-1} \in U_{N-1}} \left[\begin{array}{l} E \\ x_{N-1}, w_{N-1} \end{array} \left\{ g_N(f_{N-1}(x_{N-1}, u_{N-1}, w_{N-1})) \right. \right. \\ \left. \left. + g_{N-1}(x_{N-1}, u_{N-1}, w_{N-1}) \mid I_{N-1}, u_{N-1} \right\} \right],$$

- The optimal cost J^* is given by

$$J^* = E_{z_0} \{ J_0(z_0) \}.$$

MACHINE REPAIR EXAMPLE I

- A machine can be in one of two states denoted P (good state) and \bar{P} (bad state).
- At the end of each period the machine is inspected.
- Two possible inspection outcomes: G (probably good state) and B (probably bad state).
- Transition probabilities:



- Possible actions after each inspection:
 - C : Continue operation of the machine.
 - S : Stop the machine, determine its state, and if in \bar{P} bring it back to the good state P .
- Cost per stage:

$$g(P, C) = 0, \quad g(P, S) = 1, \quad g(\bar{P}, C) = 2, \quad g(\bar{P}, S) = 1.$$

MACHINE REPAIR EXAMPLE II

- The information vector at times 0 and 1 is

$$I_0 = z_0, \quad I_1 = (z_0, z_1, u_0),$$

and we seek functions $\mu_0(I_0), \mu_1(I_1)$ that minimize

$$E_{\substack{x_0, w_0, w_1 \\ v_0, v_1}} \left\{ g(x_0, \mu_0(z_0)) + g(x_1, \mu_1(z_0, z_1, \mu_0(z_0))) \right\}.$$

- DP algorithm: Start with $J_2(I_2) = 0$. For $k = 0, 1$, take the min over the two actions, C and S,

$$\begin{aligned}
 J_k(I_k) = \min & \left[P(x_k = P \mid I_k)g(P, C) \right. \\
 & + P(x_k = \bar{P} \mid I_k)g(\bar{P}, C) \\
 & + E_{z_{k+1}} \left\{ J_{k+1}(I_k, C, z_{k+1}) \mid I_k, C \right\}, \\
 & P(x_k = P \mid I_k)g(P, S) \\
 & + P(x_k = \bar{P} \mid I_k)g(\bar{P}, S) \\
 & \left. + E_{z_{k+1}} \left\{ J_{k+1}(I_k, S, z_{k+1}) \mid I_k, S \right\} \right]
 \end{aligned}$$

MACHINE REPAIR EXAMPLE III

• *Last Stage*: Compute $J_1(I_1)$ for each of the eight possible information vectors $I_1 = (z_0, z_1, u_0)$. We have

cost of $C = 2 \cdot P(x_1 = \bar{P} \mid I_1)$, cost of $S = 1$,

and therefore $J_1(I_1) = \min[2P(x_1 = \bar{P} \mid I_1), 1]$.
The probabilities $P(x_1 = \bar{P} \mid I_1)$ are computed using Bayes' rule:

(1) For $I_1 = (G, G, S)$

$$\begin{aligned} P(x_1 = \bar{P} \mid G, G, S) &= \frac{P(x_1 = \bar{P}, G, G \mid S)}{P(G, G \mid S)} \\ &= \frac{\frac{1}{3} \cdot \frac{1}{4} \cdot \left(\frac{2}{3} \cdot \frac{3}{4} + \frac{1}{3} \cdot \frac{1}{4}\right)}{\left(\frac{2}{3} \cdot \frac{3}{4} + \frac{1}{3} \cdot \frac{1}{4}\right)^2} = \frac{1}{7}. \end{aligned}$$

Hence

$$J_1(G, G, S) = \frac{2}{7}, \quad \mu_1^*(G, G, S) = C.$$

MACHINE REPAIR EXAMPLE IV

(2) For $I_1 = (B, G, S)$

$$P(x_1 = \bar{P} \mid B, G, S) = P(x_1 = \bar{P} \mid G, G, S) = \frac{1}{7},$$

$$J_1(B, G, S) = \frac{2}{7}, \quad \mu_1^*(B, G, S) = C.$$

(3) For $I_1 = (G, B, S)$

$$\begin{aligned} P(x_1 = \bar{P} \mid G, B \mid S) &= \frac{P(x_1 = \bar{P}, G, B, S)}{P(G, B \mid S)} \\ &= \frac{\frac{1}{3} \cdot \frac{3}{4} \cdot \left(\frac{2}{3} \cdot \frac{3}{4} + \frac{1}{3} \cdot \frac{1}{4}\right)}{\left(\frac{2}{3} \cdot \frac{1}{4} + \frac{1}{3} \cdot \frac{3}{4}\right) \left(\frac{2}{3} \cdot \frac{3}{4} + \frac{1}{3} \cdot \frac{1}{4}\right)} \\ &= \frac{3}{5}, \end{aligned}$$

$$J_1(G, B, S) = 1, \quad \mu_1^*(G, B, S) = S.$$

- Similarly, for all possible I_1 , we compute $J_1(I_1)$, and $\mu_1^*(I_1)$, which is to continue ($u_1 = C$) if the last inspection was G , and to stop otherwise.

MACHINE REPAIR EXAMPLE V

- *First Stage*: Compute $J_0(I_0)$ for each of the two possible information vectors $I_0 = (G)$, $I_0 = (B)$. We have

$$\begin{aligned}\text{cost of } C &= 2P(x_0 = \bar{P} \mid I_0) + E_{z_1} \{ J_1(I_0, z_1, C) \mid I_0, C \} \\ &= 2P(x_0 = \bar{P} \mid I_0) + P(z_1 = G \mid I_0, C)J_1(I_0, G, C) \\ &\quad + P(z_1 = B \mid I_0)J_1(I_0, B, C),\end{aligned}$$

$$\begin{aligned}\text{cost of } S &= 1 + E_{z_1} \{ J_1(I_0, z_1, S) \mid I_0, S \} \\ &= 1 + P(z_1 = G \mid I_0)J_1(I_0, G, S) \\ &\quad + P(z_1 = B \mid I_0)J_1(I_0, B, S),\end{aligned}$$

using the values of J_1 from the previous stage.

- We have

$$J_0(I_0) = \min[\text{cost of } C, \text{cost of } S]$$

- The optimal cost is

$$J^* = P(G)J_0(G) + P(B)J_0(B).$$