

# Dynamic Programming Lecture #12

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Outline:

- Example: Machine repair
- Example: Hypothesis testing

## Set-Up

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- System:

$$x_{k+1} = f_k(x_k, u_k, w_k)$$

- Observations:

$$z_0 = h_0(x_0, v_0)$$

$$z_k = h_k(x_k, u_{k-1}, v_k)$$

- Initial condition is random.
- Measurement disturbance (noise)  $v_k$  is random, but can depend on  $(x_{k-1}, u_{k-1}, w_{k-1})$ .

- Solution: Apply DP to:

$$X_{k+1} = \Phi_k(X_k, u_k, z_{k+1})$$

# Machine Repair

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- Machine state:  $= \begin{cases} 1 & \text{up} \\ 0 & \text{down} \end{cases}$ .
- Disturbance: Machine failure.
- Control actions:  $= \begin{cases} C & \text{continue (do nothing)} \\ S & \text{stop \& repair if needed} \end{cases}$ .
- System probabilities: If up, stay up with probability  $2/3$  & fail with probability  $1/3$ .
- Measurement: State of machine.
- Measurement noise: Correct diagnosis with probability  $3/4$  & false diagnosis with probability  $1/4$ .
- Stage costs:
$$g(x, u) = \begin{cases} 2 & \text{continue with machine down} \\ 0 & \text{continue with machine up} \\ 1 & \text{stop \& repair} \end{cases}$$
- Let  $p_k =$  probability machine is down at time  $k$  given information up to time  $k$ . (Then  $(1 - p_k)$  is probability machine is up at time  $k$ .)
- What is  $p_1$  as a function of  $p_0, u_0,$  &  $z_1$ ? i.e., derive

$$X_1 = \Phi(X_0, u_0, z_1)$$

## Machine Repair, cont (2)

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- In each case, compute

$$\frac{P(\text{down} \ \& \ u_0, z_1)}{P(u_0, z_1)}$$

- $u_0 = S \ \& \ z_1 = U$ :

$$\begin{cases} (1/3)(1/4) & (\text{fail})(\text{false}) \\ (2/3)(3/4) & (\text{no fail})(\text{true}) \end{cases} \Rightarrow \frac{1/12}{1/12 + 6/12} = 1/7$$

- $u_0 = S \ \& \ z_1 = D$ :

$$\begin{cases} (1/3)(3/4) & (\text{fail})(\text{true}) \\ (2/3)(1/4) & (\text{no fail})(\text{false}) \end{cases} \Rightarrow \frac{1/4}{1/4 + 2/12} = 3/5$$

- $u_0 = C \ \& \ z_1 = U$ :

$$\begin{cases} (1 - p_0)(1/3)(1/4) & (\text{was up})(\text{fail})(\text{false}) \\ p_0(1/4) & (\text{was down})(\text{false}) \\ (1 - p_0)(2/3)(3/4) & (\text{was up})(\text{no fail})(\text{true}) \end{cases} \Rightarrow \frac{1 + 2p_0}{7 - 4p_0} = \frac{1 + 2}{1 + 2 + 3}$$

- $u_0 = C \ \& \ z_1 = D$ :

$$\begin{cases} (1 - p_0)(1/3)(3/4) & (\text{was up})(\text{fail})(\text{true}) \\ (1 - p_0)(2/3)(1/4) & (\text{was up})(\text{no fail})(\text{false}) \\ p_0(3/4) & (\text{was down})(\text{true}) \end{cases} \Rightarrow \frac{3 + 6p_0}{5 + 4p_0} = \frac{1 + 3}{1 + 2 + 3}$$

- In general:  $p^+ = \Phi(p, u, z^+)$

## Machine Repair, cont (3)

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- Now consider a horizon  $[0, 2]$  with no terminal cost.

$$J_2(X_2) = 0$$

$$J_1(X_1) = \min_u E \{g(x, u)\} = \begin{cases} 2p_1 + 0 * (1 - p_1) = 2p_1 & \text{for } u_1 = C \\ 1 & \text{for } u_1 = S \end{cases}$$

$$\mu_1(p_1) = \begin{cases} C & \text{if } 2p_1 < 1 \\ S & \text{otherwise} \end{cases}$$

$$J_1(p_1) = \min \{2p_1, 1\}$$

$$J_0(p_0) = \min_u E \{g(x, u) + J_1(\Phi(X_0, u_0, z_1))\}$$

- Case  $u_0 = S$ :

$$\begin{aligned} J_0(p_0) &= 1 + E \{J_1(\Phi(X_0, S, z_1))\} \\ &= 1 + J_1(1/7)P(z_1 = U) + J_1(3/5)P(z_1 = D) \\ &= 1 + J_1(1/7)(7/12) + J_1(3/5)(5/12) \\ &= 1 + (2 * 1/7)(7/12) + 1(5/12) \end{aligned}$$

- Case  $u_0 = C$ :

$$\begin{aligned} J_0(p_0) &= 0 * (1 - p_0) + 2p_0 + J_1\left(\frac{1 + 2p_0}{7 - 4p_0}\right)P(z_1 = U) + J_1\left(\frac{3 + 6p_0}{5 + 4p_0}\right)P(z_1 = D) \\ &= \text{etc} \end{aligned}$$

- Final step: Compare  $J_0(p_0)$  for  $u_0 = C$  vs  $u_0 = S$ .

# Sequential Hypothesis Testing

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- Set-up: Measure  $\{z_0, z_1, \dots, z_{N-1}\}$ , i.e.,  $z_i \in Z$ .
- Objective: Are  $z$ 's generated by distribution  $f_0$  or  $f_1$ ?
- Decision: At time  $k$ 
  - Stop observing and conclude  $f_0$  or  $f_1$ .
  - Take an additional observation at a cost  $C$ .
- Losses: If we choose to stop
  - Cost =  $L_0$  if  $f_0$  is chosen and is wrong.
  - Cost =  $L_1$  if  $f_1$  is chosen and is wrong.
- State equations:  $x_{k+1} = x_k$ , where state is either 0 or 1.
- Initial condition:  $P(x_0 = 0) = p$

## Hypothesis Testing cont, (2)

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- Notation: Let  $p_k = P(x_k = 0 | I_k)$ .
- $p_0 = ?$ :

$$\begin{aligned} P(x_0 = 0 | z_0 = z^*) &= \frac{P(x_0 = 0 \ \& \ z_0 = z^*)}{P(z_0 = z^*)} \\ &= \frac{pf_0(z^*)}{pf_0(z^*) + (1 - p)f_1(z^*)} \end{aligned}$$

So

$$p_0 = \frac{pf_0(z_0)}{pf_0(z_0) + (1 - p)f_1(z_0)}$$

- $p_1 = ?$ : Same analysis

$$p_{k+1} = \frac{p_k f_0(z_{k+1})}{p_k f_0(z_{k+1}) + (1 - p_k) f_1(z_{k+1})}$$

## Probability Propagation Illustration

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- Suppose

$$Z = \{A, B\}$$

$$f_0 \simeq \{0.99, 0.01\}$$

$$f_1 \simeq \{0.01, 0.99\}$$

- Suppose *a priori*  $p = 1/3$ :

- Case measure  $z = A$ :

$$p_0^+ = \frac{(1/3)(0.99)}{(1/3)(0.99) + (2/3)(0.01)} = 0.98$$

- Case measure  $z = B$ :

$$p_0^+ = \frac{(1/3)(0.01)}{(1/3)(0.01) + (2/3)(0.99)} = 0.005$$



## Hypothesis Testing, cont (3)

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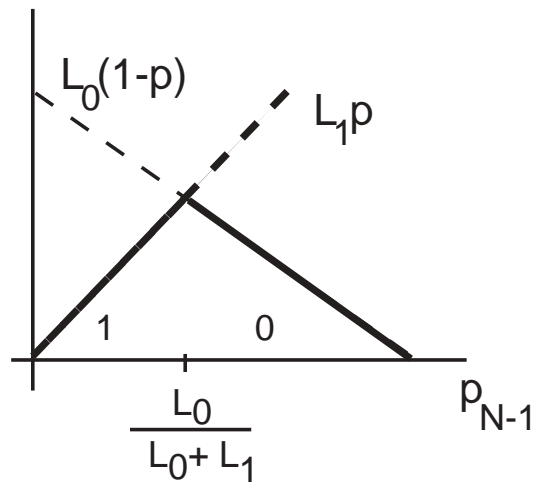
- Now apply DP

$$J_N(p_N) = \min_{\text{choose 0 or 1}} E \{ \text{cost of choice} \}$$

– Choose 0:  $L_0(1 - P_N)$

– Chose 1:  $L_1 P_N$

- $J_N(p_N) = \min \{ L_0(1 - P_N), L_1 P_N \}$



- If  $L_0 \gg L_1 \Rightarrow$  always choose 1 unless VERY confident.

## Hypothesis Testing, cont (4)

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- DP recursions:

$$J_k(p_k) = \min \begin{cases} (1 - p_k)L_0 \\ p_k L_1 \\ C + A_k(p_k) \end{cases}$$

where

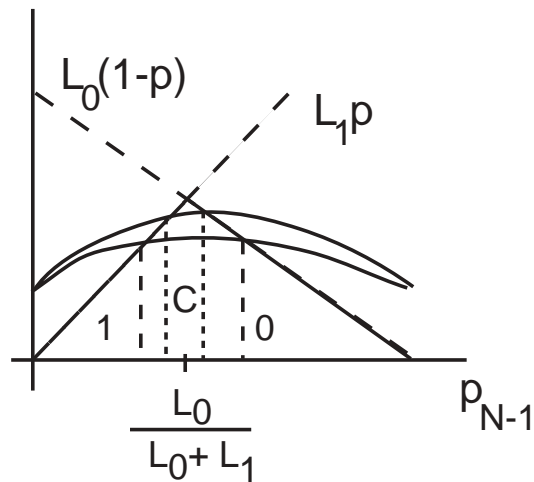
$$\begin{aligned} A_k(p_k) &= E \left\{ J_{k+1} \left( \frac{p_k f_0(z_{k+1})}{p_k f_0(z_{k+1}) + (1-p_k) f_1(z_{k+1})} \right) \right\} \\ &= \left( \sum_i f_0(z^i) J_{k+1} \left( \frac{p f_0(z^i)}{p f_0(z^i) + (1-p) f_1(z^i)} \right) \right) p \\ &\quad + \left( \sum_i f_1(z^i) J_{k+1} \left( \frac{p f_0(z^i)}{p f_0(z^i) + (1-p) f_1(z^i)} \right) \right) (1 - p) \\ &= \sum_i (p f_0(z^i) + (1 - p) f_1(z^i)) J_{k+1} \left( \frac{p f_0(z^i)}{p f_0(z^i) + (1-p) f_1(z^i)} \right) \end{aligned}$$

## Hypothesis Testing, cont (5)

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- Facts:

- $A_k(0) = A_k(1) = 0$ .
- $A_{k-1}(p) \leq A_k(p)$  by monotonicity.
- $A_k(p)$  concave



- Structure of optimal thresholds:

- Pick 1 if in left region.
- Pick 0 if in right region.
- Pick  $C$  if in middle.
- Thresholds converge as horizon length approaches infinity.