

# STOCHASTIC PROCESSES

## LECTURE 6: DTMC VALUE FUNCTION ANALYSIS

Hailun Zhang@SDS of CUHK-Shenzhen

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## Expected profit on day 3

- Expected profit on “day 3”:  $g(1) = -\$5$ ,  $g(2) = \$1$ ,  $g(3) = \$10$ .

$$\begin{pmatrix} 0.4 & 0.48 & 0.12 \\ 0.3 & 0.4 & 0.3 \\ 0.6 & 0.0 & 0.4 \end{pmatrix} \begin{pmatrix} -5 \\ 1 \\ 10 \end{pmatrix} = \begin{pmatrix} -0.32 \\ 1.90 \\ 1.00 \end{pmatrix}$$

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- Thus,

$$\begin{pmatrix} \mathbb{E}[g(X_3)|X_0 = 1] \\ \mathbb{E}[g(X_3)|X_0 = 2] \\ \mathbb{E}[g(X_3)|X_0 = 3] \end{pmatrix} = P^3 \begin{pmatrix} g(1) \\ g(2) \\ g(3) \end{pmatrix}$$

## Expected total profit in three days

- $\mathbb{E}[g(X_1) + g(X_2) + g(X_3)|X_0 = 1]$
- $v^3(i) = \mathbb{E}[g(X_1) + g(X_2) + g(X_3)|X_0 = i], i = 1, 2, 3.$

$$\begin{pmatrix} v^3(1) \\ v^3(2) \\ v^3(3) \end{pmatrix} = (P + P^2 + P^3) \begin{pmatrix} -5 \\ 1 \\ 10 \end{pmatrix} = \begin{pmatrix} 3.48 \\ 3.30 \\ -1.20 \end{pmatrix}$$

- Value function

## Expected total discounted profit in three days

- Expected total discounted profit in three days

$$\mathbb{E}[\beta g(X_1) + \beta^2 g(X_2) + \beta^3 g(X_3) | X_0 = 1)]$$

100 Yuan on day 1 is worthy 95 Yuan today if  $\beta = .95$ .

- $v^3(i) = \mathbb{E}[\beta g(X_1) + \beta^2 g(X_2) + \beta^3 g(X_3) | X_0 = i], i = 1, 2, 3.$

# Non-perishable Inventory models

- Assume i.i.d. demand

$d$	0	1	2	3
$\mathbb{P}(D = d)$	1/8	1/4	1/2	1/8

- Assume our inventory policy to be  $(s, S) = (2, 4)$ .
- Let  $X_n$  be inventory level at the end of week  $n$ . Note that values that  $X_n$  can take is in  $\{0, 1, 2, 3, 4\}$ .
- $c_v = .25$ ,  $c_f = .50$ ,  $c_p = 1$ ,  $c_s = -.10$ .
- Find the expected profit in three weeks given  $X_0 = 0$ .

## Expected total profit in multiple periods

- $g(i)$  is the expected profit in the next week, given the current week ends with  $i$  items.

$$g(0) = 11.1/8 - 1.5 = -0.1125$$

$$g(1) = 11.1/8 - 1.25 = 0.1375$$

$$g(2) = 2\left(\frac{4}{8} + \frac{1}{8}\right) + (1 - .1)\frac{2}{8} + (-.2)\frac{1}{8} = \frac{11.6}{8} = 1.45$$

$$g(3) = 3\frac{1}{8} + (2 - .1)\frac{4}{8} + (1 - .2)\frac{2}{8} + (-.3)\frac{1}{8} = 11.9/8 = 1.4875$$

$$g(4) = \quad = 1.3875$$

## Expected profit in each period

- The expected profit in three periods given  $X_0 = 0$  is

$$\mathbb{E}[g(X_0) + g(X_1) + g(X_2)|X_0 = 0]$$

- Recall

$$\begin{array}{c} \begin{array}{ccccc} & 0 & 1 & 2 & 3 & 4 \\ \begin{array}{c} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{array} & \begin{pmatrix} 0 & \frac{1}{8} & \frac{1}{2} & \frac{1}{4} & \frac{1}{8} \\ 0 & \frac{1}{8} & \frac{1}{2} & \frac{1}{4} & \frac{1}{8} \\ \frac{5}{8} & \frac{1}{4} & \frac{1}{8} & 0 & 0 \\ \frac{1}{8} & \frac{1}{2} & \frac{1}{4} & \frac{1}{8} & 0 \\ 0 & \frac{1}{8} & \frac{1}{2} & \frac{1}{4} & \frac{1}{8} \end{pmatrix} \end{array} \end{array}$$

- $\mathbb{E}[g(X_1)|X_0 = 0]$
- $\mathbb{E}[g(X_2)|X_0 = 0]$



# Why value function is important?

- AlphaGo Zero

$$v(i) = \mathbb{E}\left[g(X_1) + g(X_2) + g(X_3) + \dots g(X_{\tau-1}) + h(X_{\tau}) | X_0 = i\right]$$

- For airline yield management

$$v^{(10)}((10, 90))$$

expected profit if there are 10 days left for sale, 10 empty first class seats and 90 economic class seats.

# Computation of the value function

- Suppose that  $g(1) = -\$5$ ,  $g(2) = \$1$ ,  $g(3) = \$10$ .

- $\mathbb{E}\left[g(X_1) + g(X_2) + g(X_3)|X_0 = 1\right]$

$$v(i) = \mathbb{E}\left[g(X_0) + g(X_1) + g(X_2) + g(X_3)|X_0 = i\right], \quad i = 1, 2, 3.$$

- value function

$$\begin{pmatrix} v(1) \\ v(2) \\ v(3) \end{pmatrix} = (I + P + P^2 + P^3) \begin{pmatrix} -5 \\ 1 \\ 10 \end{pmatrix} = \begin{pmatrix} -1.52 \\ 4.30 \\ 8.80 \end{pmatrix}$$

# Complexity: a naive algorithm

- Let  $m = |S|$  be the number of states.
- Computing  $Pg$ :  
for  $i = 1$  to  $m$

$$(Pg)(i) = \sum_{\ell=1}^m P_{i\ell}g(\ell),$$

requiring  $m^2$  operations.

- Computing  $P^2$  requires  $m^3$  operations.
- Knowing  $P$  and  $P^2$ , computing  $P^3$  requires  $m^3$  operations.
- Knowing  $P$ ,  $P^2$ , and  $P^3$ , computing  $P + P^2 + P^3$  requires  $2m^2$  operations.
- Knowing  $P + P^2 + P^3$ , computing  $(P + P^2 + P^3)g$  requires  $m^2$  operations.
- One algorithm to compute  $(P + P^2 + P^3)g$  requires

$$2m^3 + 3m^2$$

operations. Think of  $m = 1$  billion.

## Complexity: Can we do better?

- Cost-to-go function  $v^k$ :  $k = 0, 1, 2, 3$

$$v^k(j) = \mathbb{E}\left[g(X_k) + \dots g(X_3) | X_k = j\right], \quad j \in S.$$

- Then  $v = v^0$ .
- Algorithm

$$v^0 = g + Pv^1,$$

$$v^1 = g + Pv^2,$$

$$v^2 = g + Pv^3, \quad m^2 + m \text{ operations}$$

$$v^3 = g$$

# Bellman equation

## THEOREM

$$v^k = g + Pv^{k+1}, \quad k = 0, 1, 2$$

Proof.

# Dynamic programming algorithm

- Backward induction: computing  $v^3, v^2, v^1, v^0$  in this order.
- Complexity:  $3(m^2 + m)$  operations.

## When the problem is infinite horizon...

Expected total discounted profit over an infinite horizon

$$\begin{aligned} v(i) &= \mathbb{E}\left[g(X_0) + \gamma g(X_1) + \gamma^2 g(X_2) + \dots | X_0 = i\right] \\ &= \mathbb{E}\left[\sum_{n=0}^{\infty} \gamma^n g(X_n) | X_0 = i\right], \quad i \in \mathcal{S}. \end{aligned}$$

### THEOREM (BELLMAN EQUATION)

*Assume that  $g$  is bounded.*

*(a) The value function satisfies that*

$$v = g + \gamma P v. \tag{1}$$

*(b) Solution  $v$  to (1) exists and is unique.*

The theorem holds even if  $|\mathcal{S}|$  is infinite.

# Proof of the theorem (a)