

PDE  $\Rightarrow$  MRP

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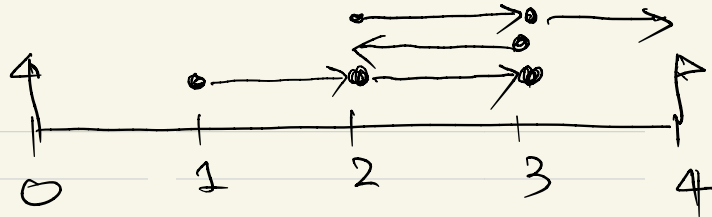
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ex random walk, ( $p = \frac{1}{2}$ )

episode



$S_0$	$R_1$	$S_1$	$R_2$	$S_2$	$R_3$	$S_3$	$R_4$	$S_4$	$R_5$	$S_5$
1	1	2	1	3	1	2	1	3	1	4

Gain with discount  $\gamma \in (0, 1)$

$$G_2 = R_3 + \gamma R_4 + \gamma^2 R_5$$

$$G_0 = \dots$$

value

$$v(1) = \mathbb{E}[G_t | S_t = 1] = ?$$

## MRP (Markov Reward Process)

An MRP is a random sequence (episode)

$$\{s_0, R_1, s_1, R_2, s_2, \dots, R_T, s_T\}$$

S.t.

(\*)  $\mathcal{S}$  is a state space, i.e.  $s_t \in \mathcal{S}$

(\*)  $\mathcal{P}$  is transition probability matrix. i.e.  
$$\mathcal{P}(s, s') = \mathbb{P}(s_{t+1} = s' | s_t = s)$$

(\*)  $\mathcal{R}$  is reward function  
$$\mathcal{R}(s) = \mathbb{E}[R_{t+1} | s_t = s]$$

(\*)  $\gamma$  is discount factor

(\*)  $T$  is stopping time, (maybe  $T = \infty$ )

## Rk on MRP

(\*) Terminal time  $T = \inf \{t \geq 0 : S_t \in \underline{\partial S}\}$

↑  
absorbing state space

(\*) Gain (random)

$$G_t = \sum_{k=0}^{T-t-1} \gamma^k R_{t+k+1}$$

$$= R_{t+1} + \gamma R_{t+2} + \dots + \gamma^{T-t-1} R_T$$

(\*) value (deterministic)

$$v(s) = \mathbb{E}[G_t \mid S_t = s]$$

Bellman eqn ①  $\forall s \in S \setminus \partial S$

$$\begin{aligned} V(s) &= \mathbb{E}[R_{t+1} + \gamma V(s_{t+1}) \mid s_t = s] \\ &= R(s) + \gamma \cdot \sum_{s' \in S} P(s, s') V(s') \end{aligned}$$

②  $V(s) = 0$ , if  $s \in \partial S$ .

Pf

$$\begin{aligned} V(s) &= \mathbb{E}[R_{t+1} + \gamma R_{t+2} + \dots + \gamma^{T-t-1} R_T \mid s_t = s] \\ \text{Tower property} \quad &= \mathbb{E}\left[R_{t+1} + \gamma \cdot \underbrace{\mathbb{E}[R_{t+2} + \gamma R_{t+3} + \dots + \gamma^{T-t-2} R_T \mid s_{t+1}]}_{\substack{S_{t+1} \\ \boxed{S}}} \mid s_t = s\right] \\ &= \mathbb{E}[R_{t+1} + \gamma \cdot V(s_{t+1}) \mid s_t = s] \end{aligned}$$

⊛ 2-d PDE

$$\begin{cases} \frac{1}{2} \Delta V - V + x_1^2 + x_2^2 - x_1 - x_2 - \frac{3}{2} = 0, & x \in (0, 1)^2 \\ V(x) = (x_1 - \frac{1}{2})^2 + (x_2 - \frac{1}{2})^2 & x \notin (0, 1)^2 \end{cases}$$

⊛ GFD with  $h$ .

Bellman

$$\begin{cases} V(x) = \gamma \ell^h(x) + \gamma \sum_{i=1}^2 \left( p^h(x + h e_i | x) V(x + h e_i) + \right. \\ \left. p^h(x - h e_i | x) V(x - h e_i) \right) \\ V(x) = (x_1 - \frac{1}{2})^2 + (x_2 - \frac{1}{2})^2 \end{cases}$$

$x \in O^h$   
 $x \notin O^h$

⊛ para

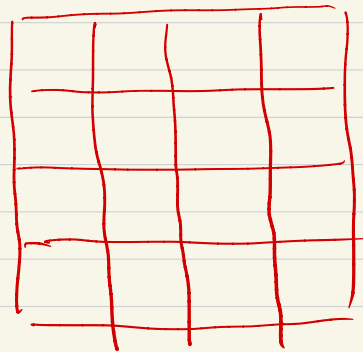
$$\gamma = \frac{2}{2 + h^2}, \quad p^h(x \pm h e_i | x) = \frac{1}{4}, \quad \ell^h(x) = \frac{h^2}{2} (x_1^2 + x_2^2 - x_1 - x_2 - \frac{3}{2})$$

MRP for 2-d

$$h = \frac{1}{4}$$

$$\textcircled{1} S = \mathbb{R}^{2,h}$$

$$\partial S = S \setminus (0,1)^2$$



$\textcircled{2}$  If  $s \notin \partial S$ , then

$$P(s, s') = \begin{cases} \frac{1}{4} \\ 0 \end{cases}$$

if  $\|s' - s\|_1 = h$   
otherwise

$\textcircled{3}$  Reward  $R(s) = \gamma Q^h(s)$

$$\textcircled{4} \gamma = \frac{2}{2+h^2} = \frac{h^2}{2+h^2} \left( x_1^2 + x_2^2 - x_1 - x_2 - \frac{3}{2} \right)$$

$$\|x\|_1 \triangleq \sum_{i=1}^d |x_i|$$

A few ways to solve Bellman

① Lin Alg.

② value iteration / Dynamic Programming

③ MC

④ TD (Temporal Differences)