

$$= \max_s \left| \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot V(s')) \right.$$

$$\left. - \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot V'(s')) \right|$$

$$= \max_s \left| \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot V(s')) - \sum_{s'} \dots + \gamma \cdot V'(s') \right|$$

$$\text{Claim} \Rightarrow \left| \max_x f(x) - \max_x g(x) \right| \leq \max_x |f(x) - g(x)|$$

$$\forall x \quad f(x) - g(x) \leq |f(x) - g(x)|$$

$$f(x) \leq |f(x) - g(x)| + g(x)$$

$$\max_x f(x) \leq \max_x |f(x) - g(x)| + g(x)$$

$$\leq \max_x |f(x) - g(x)| + \max_x g(x)$$

$$\therefore \max_x f(x) - \max_x g(x) \leq \max_x |f(x) - g(x)|$$

Lecture - 11

- $T_v(s) = \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot v(s'))$

- Contraction mapping

$f: X \rightarrow X$ is a contraction mapping if there exists a $\lambda \in [0, 1]$ such that $\forall x, y$

$$d(f(x), f(y)) \leq \lambda d(x, y)$$

- Show

$$\|T_v - T_{v'}\| \leq \gamma \|v - v'\|$$

$$\begin{aligned} \|T_v - T_{v'}\| &= \max_s |T_v(s) - T_{v'}(s)| \\ &= \max_s \left| \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot v(s')) \right. \\ &\quad \left. - \max_a \sum_{s'} P(s, a, s') (R(s, a, s') + \gamma \cdot v'(s')) \right| \\ &\leq \max_s \left| \max_a \left| \sum_{s'} P(s, a, s') (\gamma \cdot v(s') - \gamma \cdot v'(s')) \right| \right| \\ &\leq \gamma \cdot \max_s \max_a \left| \max_{s'} |v(s') - v'(s')| \right| \end{aligned}$$

$$\leq \gamma \cdot \max_{s'} |V(s') - V'(s')|$$

$$\leq \gamma \cdot \|V - V'\|$$

Overview till today —

- MDP. — environment
- CFM — first agent
- Value function, Bellman eqⁿ
- Dynamic programming
 - Policy evaluatⁿ
 - Policy improvement
 - — iteratⁿ
 - Value iteratⁿ
 - Bellman operator

Monte - Carlo method —

- Named after a Casino: Monte Carlo Casino

Idea — Use randomness to solve problems that are deterministic in principle.

Policy evaluation using Monte-Carlo —

— History $H = (S_0, A_0, R_0, \dots)$
generated by policy π

Estimate of $V^\pi(s)$

$$= \sum_{t=0}^{\infty} \gamma^t R_t$$

first visit MC

- Generate many episodes
- For each state, average discounted returns after it was visited the first time in each episode

Pseudo code —

- Initialize : $\pi \leftarrow$ policy to be evaluated
 $V \leftarrow$ an arbitrary value-function estimate
 $R \leftarrow$ an empty list for each state s
- Repeat forever

① Generate an episode using π

② for each state s appearing in the episode

$G_t \leftarrow$ Return following the first occurrence of s

if t is the first where $s_t = s$, then

$$G = G_t \triangleq \sum_{k=0}^{\infty} \gamma^k R_{t+k}$$

Append G to $\text{Returns}(s)$

$v(s) \leftarrow \text{average}(\text{Returns}(s))$