

Policy Evaluation

Dynamic Programming

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Dynamic Programming for Evaluating Value of Policy π

- ▶ Initialize $V_0^\pi(s) = 0$ for all $s \in S$
- ▶ For $k = 1$ until convergence
 - ▶ For all s in S

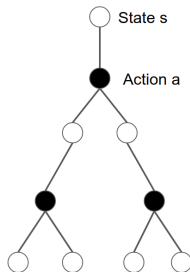
$$V_k^\pi(s) = r(s, \pi(s)) + \gamma \sum_{s' \in S} p(s' \mid s, \pi(s)) V_{k-1}^\pi(s')$$

- ▶ $V_k^\pi(s)$ is exact value of k -horizon value of state s under policy π
- ▶ $V_k^\pi(s)$ is an estimate of infinite horizon value of state s under policy π

$$V^\pi(s) = \mathbb{E}_\pi[G_t \mid s_t = s] \approx \mathbb{E}[r_t + \gamma V_{k-1} \mid s_t = s]$$

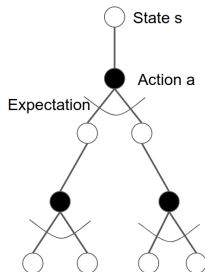
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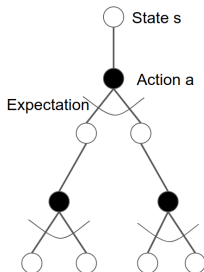
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- ~> DP computes one step update at a time
- ~> bootstrapping the rest of the expected return by the value estimate V_{k-1}

Policy Evaluation: $V^\pi = \mathbb{E}[G_t \mid s_t = s]$

- ▶ $G_t = r_t + \gamma r_{t+1} + \gamma^2 r_{t+2} + \dots$ in MDP M under policy π
- ▶ Dynamic Programming
 - ▶ $V^\pi(s) = \mathbb{E}_\pi[r_t + \gamma V_{k-1} \mid s_t = s]$
 - ▶ Requires model of MDP M
 - ▶ Bootstraps future return using value estimate
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- ▶ What if we don't know the dynamic model P and/or reward model R ?