Policy Evaluation Dynamic Programming

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

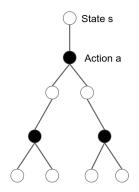
$$V_k^{\pi}(s) = r(s, \pi(s)) + \gamma \sum_{s' \in S} p(s' \mid s, \pi(s)) V_{k-1}^{\pi}(s')$$

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

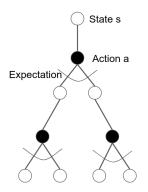
$$V^{\pi}(s) = \mathbb{E}_{\pi}[G_t \mid s_t = s] \approx \mathbb{E}[r_t + \gamma V_{k-1} \mid s_t = s]$$



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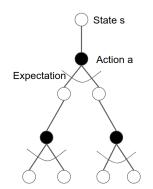


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- → DP computes one step update at a time
- \leadsto bootstrapping the rest of the expected return by the value estimate V_{k-1}



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
 - $V^{\pi}(s) = \mathbb{E}_{\pi}[r_t + \gamma V_{k-1} \mid s_t = s]$
 - ightharpoonup Requires model of MDP M
 - ▶ Bootstraps future return using value estimate
 - ► Requires Markov assumption: bootstrapping regardless of history



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- What if we don't know the dynamic model P and/or reward model R?
 (~→ see next videos)

