

Linear Programming (LP) for MDPs.

Primal form. $\min_{V \in \mathbb{R}^S} d^T V$ $d_i > 0, \sum_s d_i(s) = 1$
 s.t. $V \geq \mathcal{T}V$ Bellman optimality op.

① Why LP? Constraints: $\forall s \in S$.

$$V(s) \geq \max_{a \in A} (R(s,a) + \gamma \mathbb{E}_{s' \sim P(\cdot | s, a)} [V(s')]).$$

Nonlinear! Can convert to $|A|$ linear constraints:

$$\underbrace{V(s) \geq R(s,a) + \gamma \mathbb{E}_{s' \sim P(\cdot | s, a)} [V(s')]}_{\Delta} \quad \Delta \text{ linear!}, \quad \forall a \in A.$$

∴ Primal form has $|S|$ decision variables, & $|S| \times |A|$ constraints.

② Why it solves planning problem?

Claim: if $d_i(s) > 0, \forall s \in S$, then opt. sol. to primal is $V = V^*$.

Proof: constraint $V \geq \mathcal{T}V$.

Lemma. Monotone property of \mathcal{T} . $\forall V \geq V'$. $\mathcal{T}V \geq \mathcal{T}V'$.

We will show: $\underline{V \geq \mathcal{T}V} \Rightarrow V \geq V^*$.

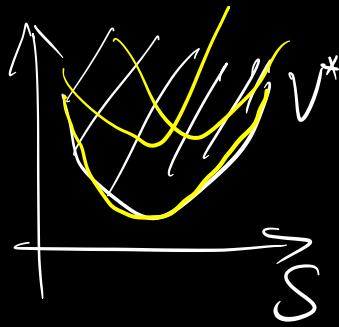
Invoke monotonicity on V & $V' = \mathcal{T}V$. $\Rightarrow \mathcal{T}V \geq \mathcal{T}(\mathcal{T}V)$

$$V \geq \mathcal{T}V \geq \mathcal{T}^2V \Rightarrow V \geq \mathcal{T}^2V.$$

$$\text{Now let } V' = \mathcal{T}^2V \Rightarrow V \geq \mathcal{T}V \geq \mathcal{T}^3V \Rightarrow V \geq \mathcal{T}^3V.$$

$$\dots \forall n, V \geq \mathcal{T}^nV.$$

Take limit $n \rightarrow \infty$. $V = \mathcal{T}^\infty V = V^*$.



Therefore for any feasible $V \geq V^*$.

② V^* is feasible : $V^* = \mathcal{T}V^*$

$\Rightarrow \min_V d^T V$ is achieved w/ $V = V^*$.

Dual form

$$\max_{d \in \mathbb{R}^{S \times A}, d \geq 0} d^T R \rightarrow \text{reward function}$$

$$\text{s.t. } \forall s \in S, \sum_a d(s, a) = \gamma \sum_{\tilde{s}, \tilde{a}} d(\tilde{s}, \tilde{a}) P(s|\tilde{s}, \tilde{a})$$

Interpretation: d plays the role of occupancy.

In fact, dual constraint characterizes all possible occupancies in the MDP.

$$\text{if } d = d^\pi. \Rightarrow d^T R = (d^\pi)^T R = \mathbb{E}_{(s, a) \sim d^\pi} [R(s, a)] \\ = \mathbb{E}_{s \sim d_0} [V^\pi(s)].$$

Verify $d = d^\pi$ is feasible for any π :

$$d^\pi(s, a) = (1-\gamma) \sum_{t=1}^{\infty} \gamma^{t-1} d_t^\pi(s, a)$$

distribution of
 s_t, a_t under π, d_0 .

$$d^\pi(s) = (1-\gamma) \sum_{t=1}^{\infty} \gamma^{t-1} d_t^\pi(s)$$

$=$

\downarrow

Dual constraint : $\sum_a d(s, a) = \gamma \sum_{\tilde{s}, \tilde{a}} d(\tilde{s}, \tilde{a}) P(s|\tilde{s}, \tilde{a}) + (1-\gamma)d_0(s)$

$$LHS = \sum_{a \in A} d^\pi(s, a) = d^\pi(s) = (1-\gamma) \sum_{t=1}^{\infty} \gamma^{t-1} d_t^\pi(s).$$

$$RHS (\text{first term}) = \gamma (1-\gamma) \sum_{t=1}^{\infty} \underbrace{\sum_{\substack{s, a \\ \Delta}} \gamma^{t-1} d_t^\pi(s, a)}_{\text{a function } s.} P(s | \tilde{s}, \tilde{a})$$

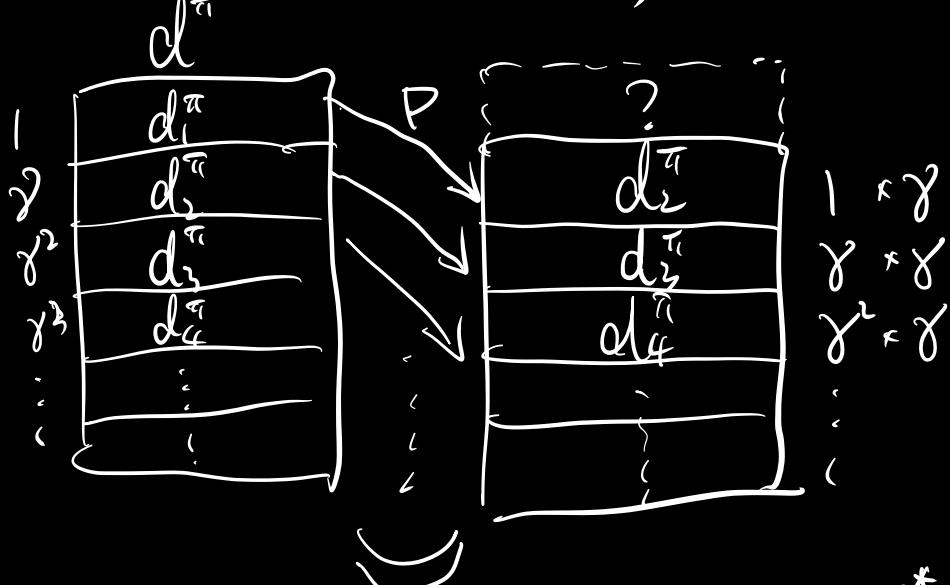
a distribution of state.

↪ s drawn from the dist $\Rightarrow (\tilde{s}, \tilde{a}) \sim d_t^\pi, s \sim P(\cdot | \tilde{s}, \tilde{a})$

\Rightarrow the distribution is $d_{t+1}^\pi(s)$.

$$\begin{aligned} \Rightarrow RHS (\text{first term}) &= \gamma (1-\gamma) \sum_{t=1}^{\infty} \gamma^{t-1} d_{t+1}^\pi(s) \\ &= (1-\gamma) \sum_{t=1}^{\infty} \gamma^t d_{t+1}^\pi(s) \\ &= (1-\gamma) \sum_{t=2}^{\infty} \gamma^{t-1} d_t^\pi(s) \end{aligned}$$

$$RHS (\text{second term}) = (1-\gamma) d_0(s) = (1-\gamma) d_0^\pi(s) \quad \forall \pi.$$



Remark: solution to dual is $d = \underline{d}^{\pi^*} \rightarrow \pi^*$. back out

$\forall \pi$ (stat. possibly stochastic). Given $\underline{d}^\pi(s, a) \Rightarrow \pi(a|s) = \frac{d^\pi(s, a)}{\sum a' d^\pi(s, a')}$