

2.1 a naive model-based approach

Central question:

Do we require an accuracy model of the world to find a near optimal policy?

A naive model to learn P : after sampling N times, let $\hat{P}(s'|s, a) = \frac{\text{count}(s', s, a)}{N}$

here we view \hat{P} as a matrix of size $|S||A| \times |S|$

Expectation: $O(|S|^2|A|)$ observation is enough for an accurate model.

Proposition 2.1 Assume $\epsilon \in (0, \frac{1}{1-\gamma})$, $\exists c > 0$ s.t.

samples from generative model

$$= |S||A|N \geq \frac{4c^2}{(1-\gamma)^4} \frac{|S|^2|A|\log(1/\delta)}{\epsilon^2} \quad [\text{different from book}]$$

where (s, a) is sampled uniformly, and with prob. $> 1-\delta$ we have

① (Model accuracy)

$$\max_{s, a} \|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \leq (1-\gamma)^2 \epsilon$$

② (Uniform value accuracy)

$$\|Q^\pi - \hat{Q}^\pi\|_\infty \leq \frac{\epsilon}{2} \quad \text{for all } \pi$$

③ (Near optimal planning) Suppose $\hat{\pi}$ is optimal w.r.t. \hat{M}

$$\|\hat{Q}^* - Q^*\|_\infty \leq \frac{\epsilon}{2}, \quad \|Q^{\hat{\pi}} - Q^*\|_\infty \leq \epsilon$$

To show this, we need following lemmas.

Lemma 2.2 [Simulation lemma] For all π :

$$Q^\pi - \hat{Q}^\pi = \gamma (I - \gamma \hat{P}^\pi)^{-1} (P - \hat{P}) V^\pi.$$

$$\begin{aligned} \text{Pf: } Q^\pi - \hat{Q}^\pi &= Q^\pi - (I - \gamma \hat{P}^\pi)^{-1} r \\ &= (I - \gamma \hat{P}^\pi)^{-1} ((I - \gamma \hat{P}^\pi) Q^\pi - r) \\ &= (I - \gamma \hat{P}^\pi)^{-1} ((I - \gamma \hat{P}^\pi) - (I - \gamma P^\pi)) Q^\pi \\ &= \gamma (I - \gamma \hat{P}^\pi)^{-1} (P^\pi - \hat{P}^\pi) Q^\pi \\ &= \gamma (I - \gamma \hat{P}^\pi)^{-1} (P - \hat{P}) V^\pi \quad \square \end{aligned}$$

Lemma 2.3 For any policy π , MDP M and $v \in \mathbb{R}^{|S||A|}$

$$\|(I - \gamma P^\pi)^{-1} v\|_\infty \leq \frac{1}{1-\gamma} \|v\|_\infty$$

$$\text{Pf: } v = (I - \gamma P^\pi)(I - \gamma P^\pi)^{-1} v =: (I - \gamma P^\pi) w$$

$$\begin{aligned} \Rightarrow \|v\|_\infty &= \|(I - \gamma P^\pi) w\|_\infty \\ &\geq \|w\|_\infty - \gamma \|P^\pi w\|_\infty \\ &\geq \|w\|_\infty - \gamma \|w\|_\infty \\ &= (1-\gamma) \|w\|_\infty \end{aligned}$$

$$\text{i.e. } \|(I - \gamma P^\pi)^{-1} v\| \leq \frac{1}{1-\gamma} \|v\|_\infty \quad \square$$

Lemma A.8 [Concentration for discrete distributions]

Let z be r.v. of $\{1, \dots, d\}$, distributed according to q , where $\bar{q} = [P_r(z=j)]_{j=1}^d$. Assume we have N i.i.d. samples and that our empirical estimate is $[\hat{q}]_j = \frac{\sum_{i=1}^N \mathbb{1}_{\{z_i=j\}}}{N}$,

we have $\forall \varepsilon > 0$:

$$P_r(\|\hat{q} - q\|_2 \geq \frac{1}{\sqrt{N}} + \varepsilon) \leq e^{-N\varepsilon^2},$$

which implies:

$$P_r(\|\hat{q} - \bar{q}\|_1) \geq \sqrt{d} \left(\frac{1}{\sqrt{N}} + \varepsilon \right) \leq e^{-N\varepsilon^2}.$$

this proof is ignored

Pf of Proposition 2.1 :

with ℓ_1 norm in lemma A.8, for fixed s, a , with prob. $\geq 1 - \delta$, we have

$$\|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \leq c \sqrt{\frac{|S| \log(1/\delta)}{N}} \quad (*)$$

where N is the number of samples used to estimate $\hat{P}(\cdot|s, a)$. just let $\delta = e^{-N\varepsilon^2} \Rightarrow \varepsilon = \sqrt{\frac{\log(1/\delta)}{N}}$, $d = |S|$

and let c satisfy $c \sqrt{\frac{|S| \log(1/\delta)}{N}} \geq \sqrt{|S|} \left(\frac{1}{\sqrt{N}} + \varepsilon \right) \Rightarrow c = 1 + \sqrt{\log(1/\delta)}$

$$\textcircled{1} \|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \leq (1 - \gamma)^2 \varepsilon$$

since $N \geq \frac{4c^2}{(1-\gamma)^4} \frac{|S| \log(1/\delta)}{\varepsilon^2}$, by (*) we have

$$\|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \leq (1 - \gamma)^2 \varepsilon / 2 \quad \text{with prob. } \geq 1 - \delta$$

$$\textcircled{2} \|Q^\pi - \hat{Q}^\pi\|_\infty \leq \frac{\varepsilon}{2}$$

By Lemma 2.2:

$$\|Q^\pi - \hat{Q}^\pi\|_\infty = \|\gamma(I - \gamma P^\pi)^{-1}(P - \hat{P})V^\pi\|_\infty$$

$$\text{Lemma 2.3} \leq \frac{\gamma}{1-\gamma} \|(P - \hat{P})V^\pi\|_\infty$$

$$\begin{aligned} \text{H\"older ineq.} &\leq \frac{\gamma}{1-\gamma} \left(\max_{s, a} \|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \right) \|V^\pi\|_\infty \\ &\leq \frac{\gamma}{(1-\gamma)^2} \max_{s, a} \|P(\cdot|s, a) - \hat{P}(\cdot|s, a)\|_1 \\ &\leq \gamma \varepsilon / 2 \leq \varepsilon / 2 \end{aligned}$$

$$\textcircled{3} \|\hat{Q}^* - Q^*\|_\infty \leq \frac{\varepsilon}{2}, \quad \|Q^\pi - Q^{\pi^*}\|_\infty \leq \varepsilon$$

observe that $|\sup_x f(x) - \sup_x g(x)| \leq \sup_x |f(x) - g(x)|$

$$\begin{aligned} \Rightarrow |\hat{Q}^*(s, a) - Q^*(s, a)| &= \left| \sup_{\pi} \hat{Q}^\pi(s, a) - \sup_{\pi} Q^\pi(s, a) \right| \\ &\leq \sup_{\pi} |\hat{Q}^\pi(s, a) - Q^\pi(s, a)| \end{aligned}$$

$$\begin{aligned}
\|Q^{\hat{\pi}} - Q^{\pi^*}\|_{\infty} &\leq \|Q^{\hat{\pi}} - \hat{Q}^*\|_{\infty} + \|\hat{Q}^* - Q^{\pi^*}\|_{\infty} \\
&= \|Q^{\hat{\pi}} - \hat{Q}^{\hat{\pi}}\|_{\infty} + \|\hat{Q}^* - Q^*\|_{\infty} \\
&\leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon \quad \square
\end{aligned}$$