

On the Generalization Gap in Reparameterizable Reinforcement Learning

Huan Wang, Stephan Zheng, Caiming Xiong, Richard Socher

Salesforce Research

June 12, 2019



Summary



- ▶ **Reparameterize** RL to decouple randomness of the environment from the policy.
- ▶ Bring in supervised learning theory to RL thanks to reparameterization.
- ▶ Theoretical guarantees on the generalization gap in RL.
- ▶ Generalization gap is related to:
 - ▶ Number of training episodes
 - ▶ Smoothness of the environment, policy and reward
 - ▶ Discrepancy between training and test environment
 - ▶ “Complexity” of the reward and transition function class.

Reparameterization using Gumbel-max trick



- ▶ Treat each episode, $\mathbf{s}^i = [s_0^i, s_1^i, \dots, s_T^i] \sim \mathcal{D}_\pi$, as a sample.
- ▶ Reparameterization using Gumbel noise $g_t^i \sim G$:

$$s_{t+1}^i = \arg \max [\log \mathcal{P}(s_t^i, \pi(s_t^i)) + g_t^i]$$

- ▶ Empirical reward:

$$\begin{aligned}\hat{\pi} &= \arg \max_{\pi \in \Pi, \mathbf{s}^i \sim \mathcal{D}_\pi} \frac{1}{n} \sum_i R(\mathbf{s}^i) \\ &= \arg \max_{\pi \in \Pi, g^i \sim G} \frac{1}{n} \sum_i R(\mathbf{s}^i(\pi, g^i))\end{aligned}$$

- ▶ The distribution G is static and independent of learned model $\pi \rightarrow$ **as in supervised learning!**

Generalization Gap in Episodic RL



$$\begin{aligned} & \left| \frac{1}{n} \sum_{i=1}^n R(\mathbf{s}^i) - \mathbb{E}_{\mathbf{s} \sim \mathcal{D}'} R(\mathbf{s}) \right| \\ & \leq \underbrace{\left| \frac{1}{n} \sum_{i=1}^n R(\mathbf{s}^i) - \mathbb{E}_{\mathbf{s} \sim \mathcal{D}_{\hat{\pi}}} R(\mathbf{s}) \right|}_{\epsilon_{intrinsic}} + \underbrace{\left| \mathbb{E}_{\mathbf{s} \sim \mathcal{D}_{\hat{\pi}}} R(\mathbf{s}) - \mathbb{E}_{\mathbf{s} \sim \mathcal{D}'} R(\mathbf{s}) \right|}_{\epsilon_{external}} \end{aligned}$$

- ▶ Intrinsic Gap: internal randomness from the same “reparameterizable MDP”:

$$\epsilon_{intrinsic} \leq Rad(R_{\pi, \mathcal{T}, \mathcal{I}}) + O\left(c \sqrt{\frac{\log(1/\delta)}{n}}\right)$$

- ▶ External Gap: between “different reparameterizable MDPs”:

$$\epsilon_{extrinsic} \leq L_r \zeta \sum_{t=0}^T \gamma^t \frac{\nu^t - 1}{\nu - 1} + L_r \epsilon \sum_{t=0}^T \gamma^t \nu^t$$



<http://proceedings.mlr.press/v97/wang19o/wang19o.pdf>

<https://arxiv.org/abs/1905.12654>

- ▶ Come see our poster!

Wed Jun 12th 6:30 – 9:00 PM @ Pacific Ballroom #43

- ▶ Contact us:

huan.wang@salesforce.com
stephan.zheng@salesforce.com