Research Presentation

Active Learning with V-Learning

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Pool-Based Active Learning



 $\mathcal{L} \in \mathbb{R}^{\lambda imes m}$ Labeled Set $\mathcal{U} \in \mathbb{R}^{\mu imes m}$ Unlabeled Set $\phi_{ heta} := \mathbb{R}^m o \mathbb{R}^c$ Classifier $\mathcal{K} \in \mathbb{R}^{k imes m}$ Unlabeled Sample $\psi := \mathbb{R}^{k imes m} o \mathbb{R}^k$ Active Learning Heuristic $\pi_{\psi} := argmax \ \psi(\mathcal{S})$ Active Learning Policy

Active Learning with Q-Learning



Problems:

- Fixed Sample size
- Expensive Transitions
- Same actions in different places

$$\mathcal{S} \in \mathbb{R}^{k \times \sigma}$$
 State Space $\mathcal{A} \in [0, \dots, k]$ Action Space $\mathcal{R} := \mathcal{S} \times \mathcal{A} \to \mathbb{R}$ Reward Function $\tau := \{\mathcal{S}, \mathcal{A}, \mathcal{S}, \mathcal{R}, \mathbb{R}\}$ Transition

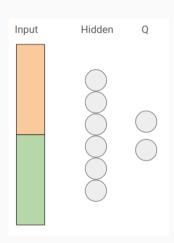
Active Learning with Q-Learning



Sample size k = 2

The same datapoint can appear in multiple places in the input

Both output nodes essentially learn the same function



Pool-Based Active Learning



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