Exploring Value Function Transfer Between On-Policy and Off-Policy Methods in Tabular Gridworld



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TL;DR

- Value function transfer between SARSA (on-policy) and Q-Learning (off-policy).
- Using two different exploration strategies ε-greedy and softmax.
- Cross-paradigm transfer often leads to trade off between return reward and training error.

Motivation & Problem Setting

Motivation

- Transferring value function can accelerate learning in tabular RL.
- Exploration strategies like εgreedy and softmax may influence transfer success.

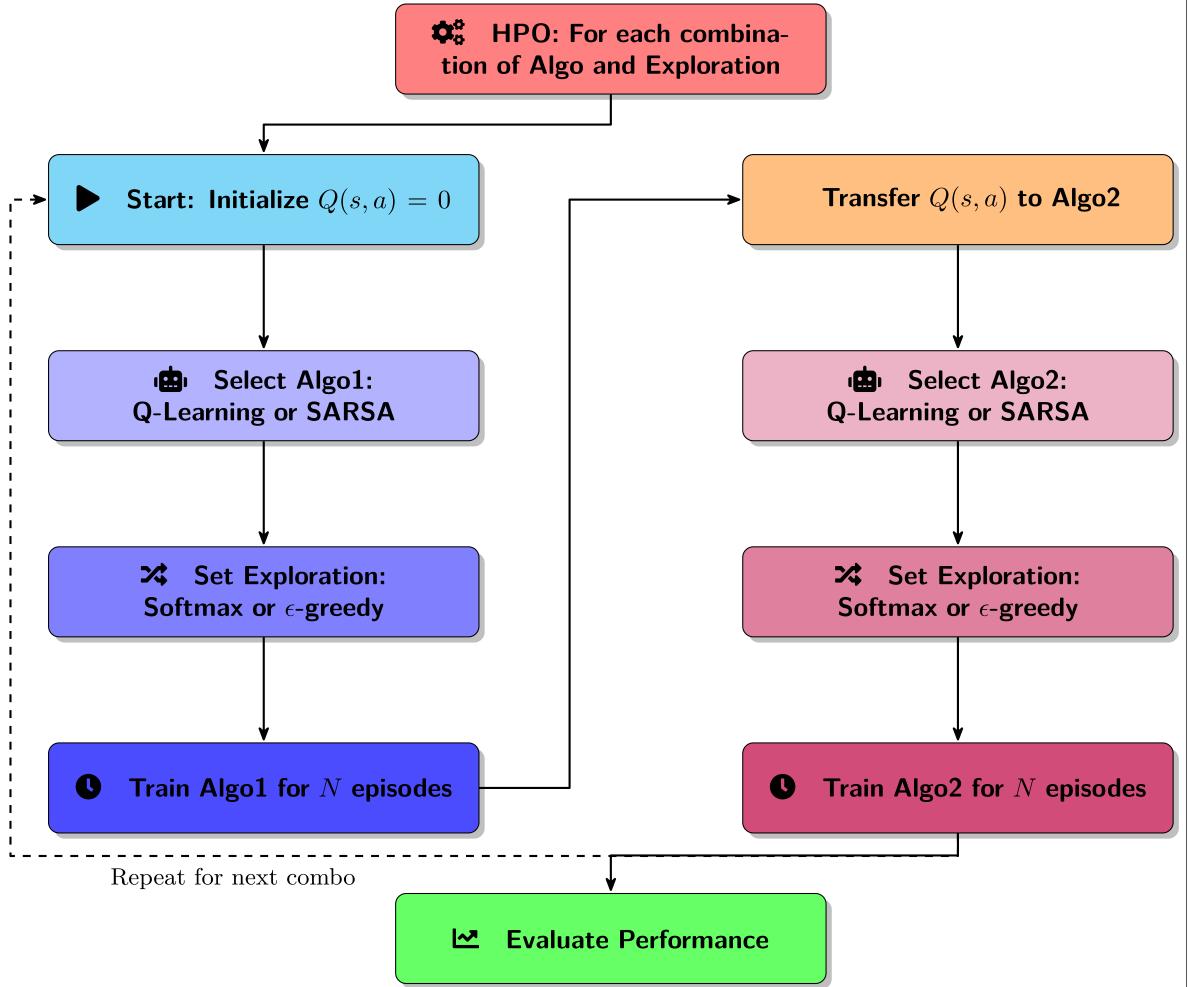
Cumulative Return

Problem Setting

- Can value functions be effectively transferred between off- and on-policy tabular RL methods?
- How does the exploration strategy (ε-greedy vs. softmax) influence such transfer?

Training Error

Approach

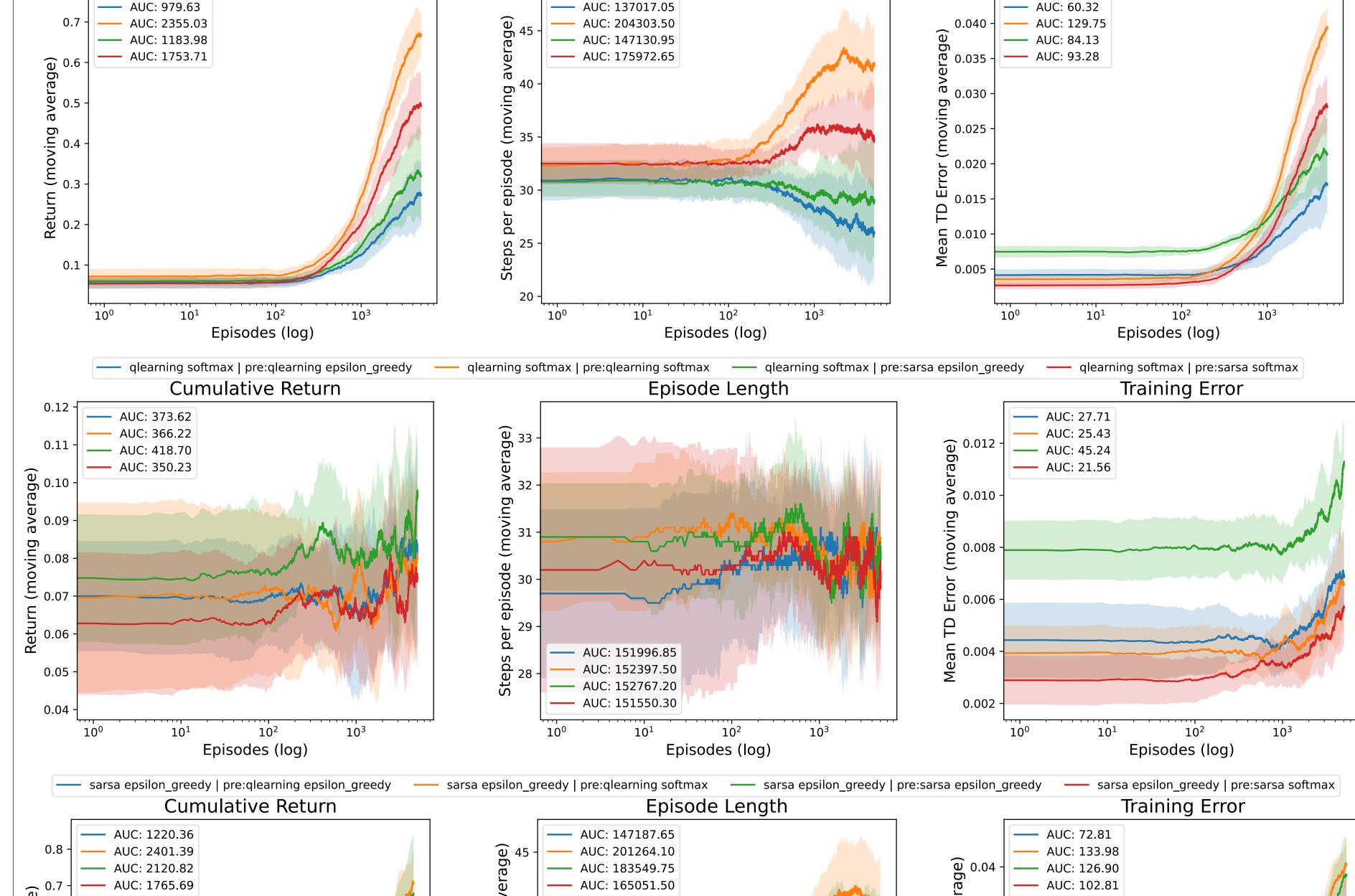


- FrozenLake Environment with a grid-size of 7x7.
- Transferring Q-Table.
- 5000 Episodes per run for pre- and transfer-training.
- Resetting Policy after transfer (e.g. εgreedy).

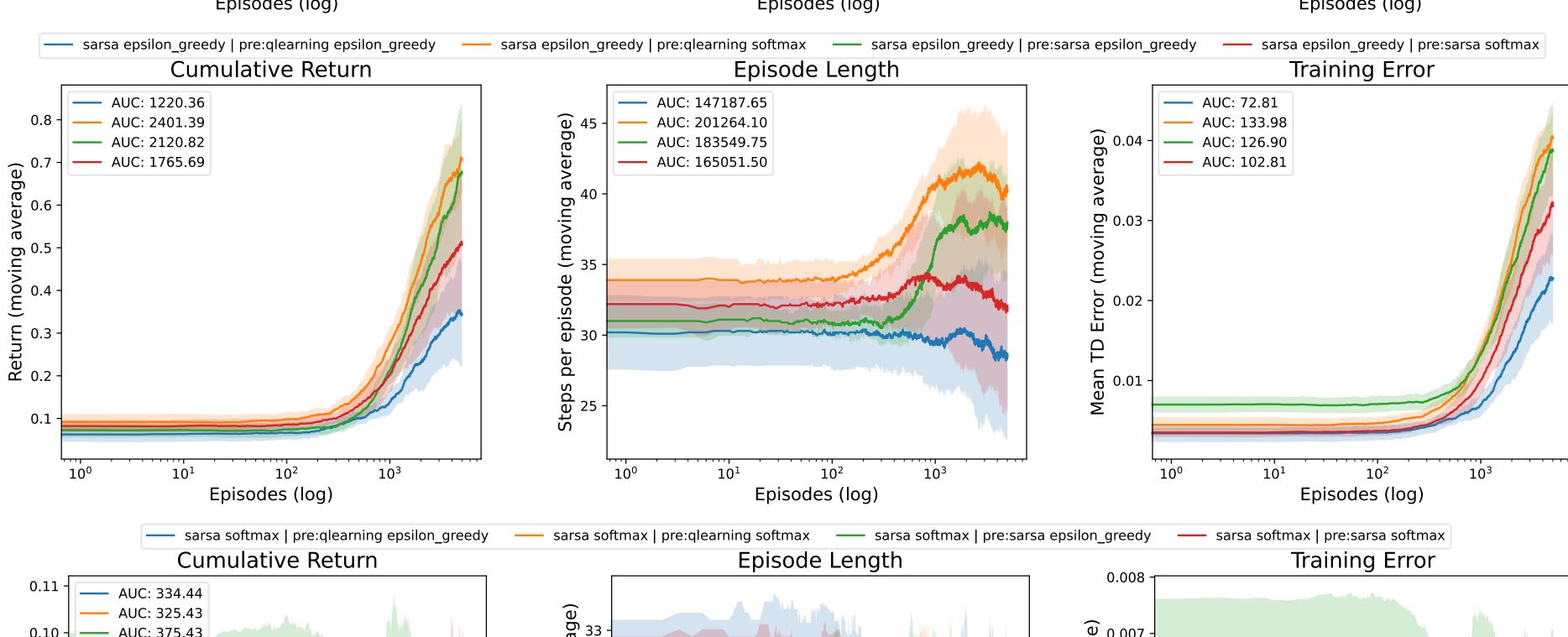
5 **Future Works**

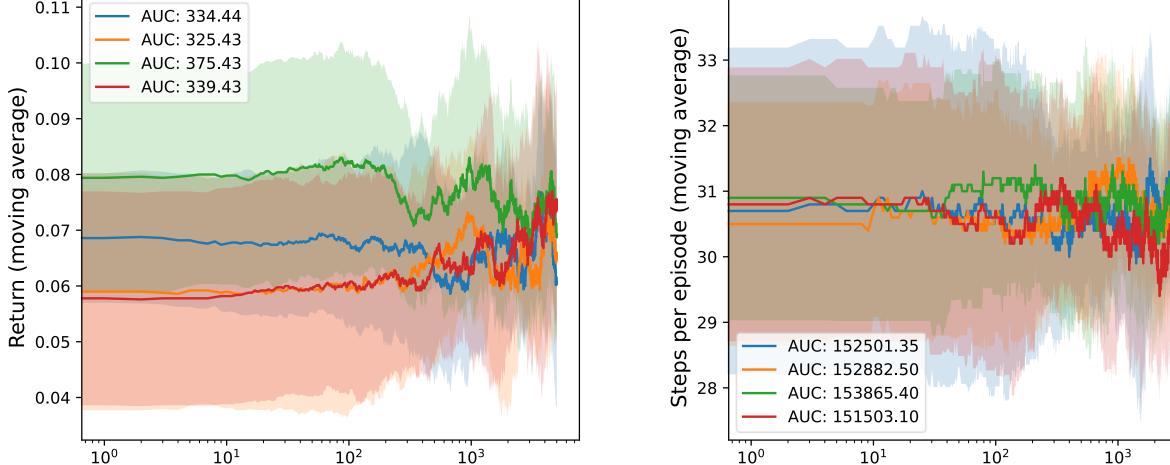
- Explore more complex environments.
- Test additional RL algorithms and exploration strategies.
- Exploring the optimal timing or scope of transfer.
- Explore generalization with switching the environment.

Key Insights



Episode Length





- Episodes (log)
- 0.007 Episodes (log)
- Off- to On-Policy: Sharp Q-values transferred, but overestimation needs correction.

Episodes (log)

- On- to Off-Policy: Conservative values get overwritten fast.
- ε-greedy to softmax: Only if Qvalues ranked well. Otherwise, it causes noisy probabilities.
- softmax to ε-greedy: Smooth values confuse greedy selection.