

# Reinforcement Learning

## Exercise 5

Jim Mainprice, Philipp Kratzer  
Machine Learning & Robotics lab, U Stuttgart  
Universitätsstraße 38, 70569 Stuttgart, Germany

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### Submission Instructions:

The submission deadline for this exercise sheet is 26.05., 23:55.

Put your answers into a single pdf. Your python code should be a single python script. Upload both files to ilias. Make sure that the code runs with `python3 yourscrip.py` without any errors.

Group submissions of up to three students are allowed.

### 1 Random Walk (2P)

Recall the Random walk example presented in the lecture. From the results shown in the left graph (estimated value) it appears that the first episode results in a change in only  $V(A)$ . What does this tell you about what happened on the first episode? Why was only the estimate for this one state changed? By exactly how much was it changed (assuming  $\alpha = 0.1$ )?

The episode ended on the far left end-state.  $V(A)$  reduced 0.05 at the end of the episode (considering  $\gamma=1$ )

### 2 Sarsa and Q-learning on the FrozenLake (8P)

The code template can be found on github (<https://github.com/humans-to-robots-motion/rl-course>) in `ex05-td/ex05-td.py`.

- Implement Sarsa and obtain and plot the state-value function, action-value function, and policy for the FrozenLake environment. Plot the average episode length as training continues. Put your plots into your submission pdf. (3P)
- Implement Q-learning and obtain and plot the optimal state-value function, action-value function, and policy for FrozenLake. Put your plots into your submission pdf. What can you say about on-line performance during training in comparison to the performance of the optimal policy? (3P)
- Explore how your results for a) and b) change if you switch to the non-slippy version (i.e. deterministic environment). Put your plots into your submission pdf. (1P)
- Rerun your code for the larger FrozenLake environment. Put your plots into your submission pdf. (1P)