## Reinforcement Learning Exercise 1

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## **Submission Instructions:**

The submission deadline for this exercise sheet is 28.04., 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 yourscript.py* without any errors.

Group submissions of up to three students are allowed.

## 1 Multi-armed Bandits (4 Points)

- a) Consider  $\epsilon$ -greedy action selection for a bandit with two actions (k=2) and  $\epsilon=0.5$ . What is the probability that the greedy action is selected? (2P)
- b) Consider a k-armed bandit problem with k=4 actions, denoted 1, 2, 3, and 4. Consider applying to this problem a bandit algorithm using  $\epsilon$ -greedy action selection, sample-average action-value estimates, and initial estimates of  $Q_1(a)=0$ , for all a. Suppose, you observe the following sequence of actions and rewards:  $A_1=1, R_1=1, A_2=2, R_2=1, A_3=2, R_3=2, A_4=2, R_4=2, A_5=3, R_5=0$ . On some of these time steps the  $\epsilon$  case may have occurred, causing an action to be selected at random. (2P)
  - 1. On which time steps did this definitely occur?
  - 2. On which time steps could this possibly have occurred?

## 2 Action Selection Strategies (6 points)

The source code for programming exercises will be published on github: https://github.com/humans-to-robots-motion/rl-course. The first exercise can be found as python script in ex01-bandits/ex01-bandits.py. The code implements a 10-armed Gaussian bandit.

- a) Implement the greedy action selection strategy in the function *greedy*. Initialize the values by playing each arm once. (3P)
- b) Implement the  $\epsilon$ -greedy strategy in the function epsilon\_greedy. Use  $\epsilon = 0.1$ . (1P)
- c) In the main function set n\_episodes=10000 to create a plot with less noise (this might take some time). Add the plot into your submission pdf (The code template already stores it as an eps file). Which of the 2 methods performs better, why? (1P)
- d) Think about possible ways to improve the implemented methods. What changes could you make to the strategies in order to improve them? (1P)