

Exploration strategies in the Multi-Arm bandit testbed.

This is the course's first assignment. We will use the exploration strategies together with the 10-arm bandit testbed to help you get familiar with the main tools used in a reinforcement learning project.

1. Implement a 10-armed bandit in Python, each arm following a normal distribution $\mathcal{N}(\mu, \sigma^2)$ with $\sigma = 1$ and $\mu \sim \mathcal{N}(0, 1)$.
2. Conduct 500 different learning runs, each initializing a distinct 10-armed bandit setup.
3. Produce two plots and corresponding tables:
 - (a) Plot 1. Plot the average reward learning curve per step for the ϵ -greedy algorithm with $\epsilon = 0$, and $\epsilon = 0.1$.
 - Utilize the sample-averaged method for the action-value updates.
 - Initialize the starting estimation of action values at 0 for each action a in the action set \mathcal{A} ($Q_0(a) = 0, \forall a \in \mathcal{A}$).
 - Run the simulation for 1000 steps.

The plot should resemble Figure 1. Note that your version should have legend, smoothing, mean, and standard deviation, as in Figure 2.

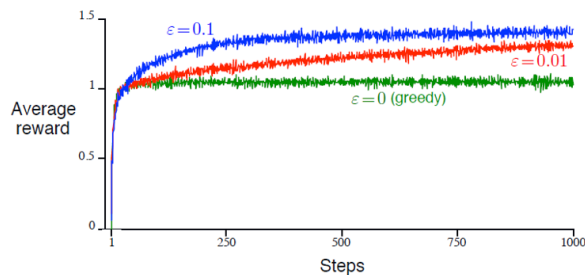


Figure 1: Example of a learning curve.

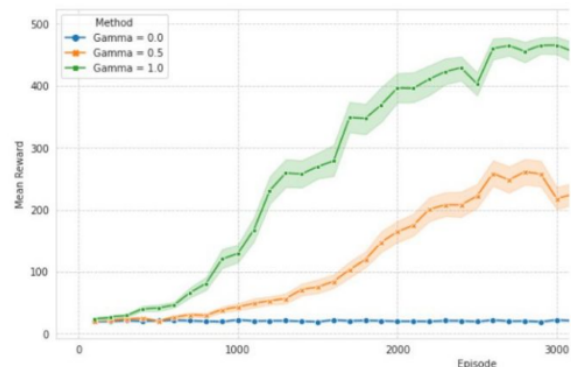


Figure 2: Example of smoothing, legend, mean, and standard deviation in a learning curve.

- (b) Plot 2. Compare the average rewards over the first 1000 steps across different exploration strategies with varying hyperparameters.
 - Analyze the following algorithms:
 - Greedy with optimistic initialization and weighted-average method. ($Q_0, \alpha = 0.1$)
 - ϵ -greedy with sample-averaged method. (ϵ)
 - Upper-Confidence Bound with sample-averaged method.(UCB) (c)
 - SoftMax with sample-averaged method.(τ)
 - Gradient Bandit (Action preferences with baseline) (α)
 - Include a plot with the **average and standard deviation** of rewards.

The plot should resemble Figure 3. Note that your version should have legend, smoothing, mean, and standard deviation, as in Figure 2, and that the example does not provide a curve for the SoftMax policy.

4. This assignment is designed to familiarize you with various tools utilized in a reinforcement learning project:
 - (a) Conduct experiments on Hábrók.

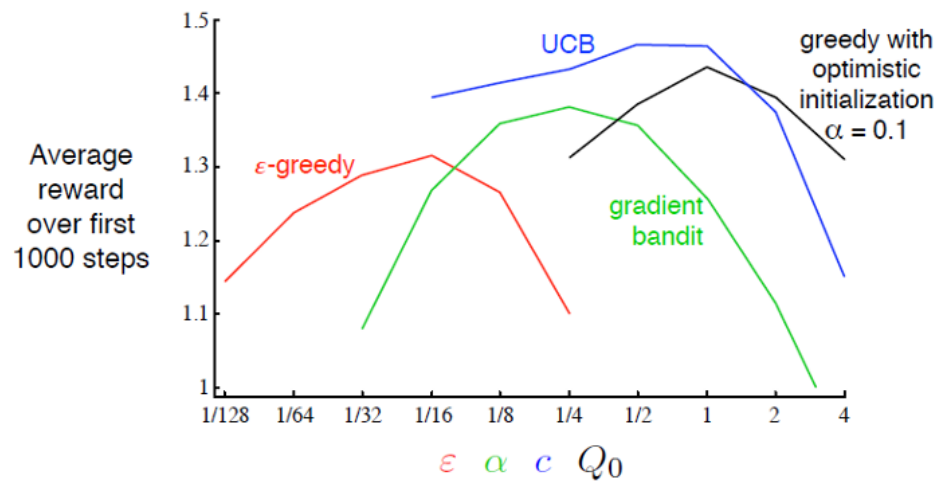


Figure 3: Example of curve for the second plot of the assignment.

- (b) Manage file transfers between Hábrók and GitHub using git commands.
 - (c) Utilize TensorBoard for data visualization during training.
 - (d) Develop a method to convert data into a Pandas dataframe and plot it using Seaborn or Matplotlib, considering data smoothing, mean, and standard deviation.
 - (e) Each plot should include a caption inspired by “Experimentation in RL” from Tutorial 3 of the theoretical RL course.
 - (f) Ensure reproducibility by selecting a consistent starting seed.
 - (g) Follow the “bachelor’s project” report template.
5. The report should include:
- (a) A section describing the algorithms.
 - (b) A section detailing the plots, tables, and discussions of results.
 - (c) A conclusion section.
 - (d) The report should not exceed five pages.

For additional support, refer to the materials provided in the theoretical reinforcement learning course:

- Theory: Lecture 1 or Chapter 2 of “Reinforcement Learning: An Introduction”.
- Coding: Tutorial 2, Part 2.
- Hábrók usage: Tutorial 7, Part 2.
- Presenting statistical results: Tutorial 3, Part 2.
- Captions’ plot: Tutorial 3, Part 2.