# Final Projects for the RL course

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Some guidelines

#### When will the exams be?

#### QLS diploma students:

July 1<sup>st</sup>, individual projects

All exams will be in presence. **Define your project by June 17 at the latest**QLS students must attend all presentations

#### **UNITS Master students:**

- Whenever you are ready. First opportunity, July 1st.

Group projects are possible

Write us in advance (min. 10 days). Exams can be in presence or online

### Two main flavours of projects (A degree of hands-on is always recommended).

#### I - Hands-on focus

What: Construct the solution to a specific problem/game using RL.

Why: Understand the different approaches to a problem, discuss problems and solutions.

#### II - Theory focus (harder!)

What: Delve deeper into one particular subject

Why: Explore a theme outside the class material (but with some connection to it)

#### - All projects -

The exam can cover also theory questions both on the project *and* on the rest of the program

# How to choose the subject of the project

Choose something that interests you

It can be an *extension of the theory* seen in class.

It can be an algorithm used on a different problem.

Once you have an idea (and especially if you *do not* have any ideas):

- write to us, so that we can give the project some "boundaries"

# What should you produce at the end of the project?

1) Presentation of 20/30 minutes

2) GitHub repositories / Jupyter notebook with code

### What is the aim of the hands-on side of a project?

Show that you fully understand the process of applying RL methods to a problem.

- How to "translate" a problem in a formal setting (generic problem to MDP...)
  - How to tackle the problem (methods/algorithm...)
  - How to overcome issues (simplify the problem / use learning tricks...)

#### What is *not* its aim?

Spending weeks on the coding part of the problem at the expense of understanding

- Avoid choosing too complex problems
- Avoid solutions which which require huge amounts of data/GPUs

### **Results: How to present**

- i) **Describe and motivate** the problem.
- ii) How did you **translate** the problem **into a RL framework**?
- iii) **How** did you (try to) solve it? **Why** did you use one algorithm or another?
- iv) Is there **something else** could you have done?
- v) Did it work? (**Can you interpret** the optimal policy?)
- vi) Are the results consistent with your expectations? Can you explain them?

# What is the aim of the theory part of a project?

Show that you can navigate the literature in order to build on your knowledge and learn a new method, and get a deeper understanding of something treated in class.

#### Examples:

- a chapter of the Sutton and Barto's book that was not covered
- a deep dive on Actor-Critic methods
- applications of RL in a specific sector (advertisement, medical science...)

### **Success stories:**

https://github.com/lorenzobasile/RLProject

https://github.com/mariagraziaberni/Reinforcement\_Learning

### Can you use AI?

Yes, but...

- you must declare it explicitly.
   (this part of the code was done by, the text / graphs / analysis, ..., the idea ...)
- ChatGPT is **not** a good scientist:
  - . use with caution
  - . double-check everything
  - . question everything
- ChatGPT *may be* good at:
  - . translating code from one language to another
  - . code faster (assuming you already know what it is)

Probability ChatGPT (or related) makes your project better: low

# Best/Worst practice

- **DO** share the code with clear annotations
- **DO** show your reasoning in the presentation
- **DO** question your results
- **DO** show final policies vs only learning curves
- DON'T talk only about code
- DON'T present learning curves without "target score"
- **DON'T** say/write things you don't understand

### Can we work in groups?

Yes (for Master students)

Note that:

- We expect that the "project size" scales with "group size"
- Presentation is collective
- Make sure that individual contributions are clearly defined