

# ECE 1508: Reinforcement Learning

## Course Logistics

Ali Bereyhi

[ali.bereyhi@utoronto.ca](mailto:ali.bereyhi@utoronto.ca)

Department of Electrical and Computer Engineering  
University of Toronto

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# Welcome to ECE 1508!

Great pleasure to see you in ECE 1508

## *Reinforcement Learning*

*Instructor:* Ali Bereyhi

- *Office:* BA 7208 at Bahen Centre for Information Technology
- *Email:* ali.bereyhi@utoronto.ca

*Teaching Assistants:* TBA

- There will be *tutorial lectures*

*Where and When?*

- Tuesdays at **5:00 PM till 7:00 PM** at **BA-1170**
- Fridays at **5:00 PM till 7:00 PM** at **BA-1180**

# Quercus and Piazza

We got a [Quercus page](#)

- You have been automatically enrolled
- Also you got registered at the [Piazza](#) page
  - ↳ You can login through the [Quercus](#) page
- We though use the [Course Page](#) to share the course materials

**Please!** Feel free to ask questions on Piazza!

# What Do We Learn?

*In nutshell:* we learn **Reinforcement Learning!**

You may wonder how do we learn it? Well! in 3 steps

- **Step 1: Fundamentals of Reinforcement Learning**
  - We try to get understand the underlying framework
  - We understand what the main problem is
    - ↳ *We see that it's a sort of optimization problem*
  - We get to look at some simple example

By the end of this step, we know in theory

- What kind of problems we are dealing with in Reinforcement Learning
- What **methods** are available to solve these problems

# What Do We Learn?

**In nutshell:** we learn Reinforcement Learning!

You may wonder how do we learn it? Well! in 3 steps

- **Step 2: Reinforcement Learning Methods**
  - Model-based Methods
    - In some toy-scenarios, we can write the underlying problem analytically
    - This is not really the case in practice though!
  - Model-free Methods
    - In reality, we cannot write the problem analytically!
    - We need to solve the problem directly from data by efficient algorithms

As we get over this part

- You have all background that you need on Reinforcement Learning
  - You can formulate a Reinforcement Learning problem
  - You can specify at least one algorithms to solve it

But! Your algorithm might take for ever to run! 😊

# What Do We Learn?

*In nutshell:* we learn Reinforcement Learning!

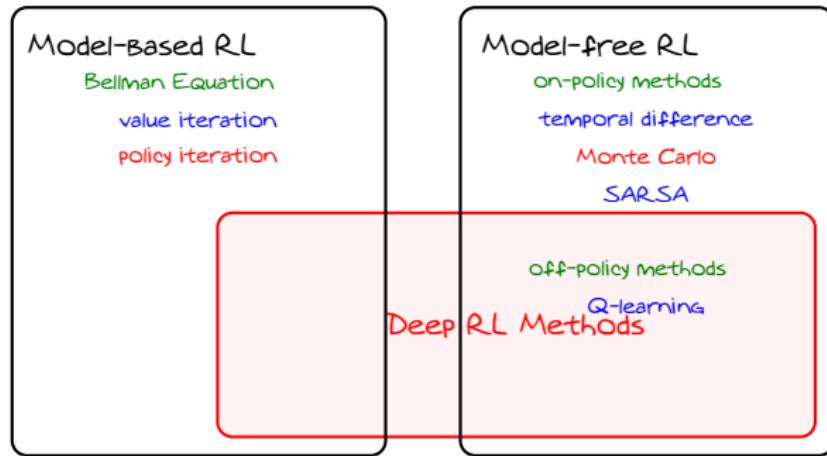
You may wonder how do we learn it? Well! in 3 steps

- **Step 3: Deep Reinforcement Learning**
  - We now apply deep learning to solve those hard problems
    - We use neural networks to learn the solution from few samples
    - We look into Deep Q-Learning and Policy Gradient Methods
    - We also learn actor-critic methods

This is the major part of the course  $\approx 50\%$

- You need good background on Deep Learning, i.e., to be fairly familiar with neural networks

# High Level Chart



We will see this chart couple of times in this course!

# How Do We Get Trained?

*There are three learning components in the course*

- Assignments
  - ↳ You get **three sets** of assignments
  - ↳ **No need to say** that they are the **best thing** to understand the course!
  - ↳ And, of course we do **lots of programming** in there!
    - ↳ Your code **should not return error!**
  - ↳ Each assignment will be solved in Tutorial **after the deadline**
    - ↳ Submission by **deadline at 11:59 PM**: full mark
    - ↳ Up to 3 days delay is allowed: **each day deducts 10%**
- Midterm Exam
- Final Project

# How Do We Get Trained?

*There are three learning components in the course*

- Assignments
- Midterm Exam
  - ↳ We will have one written exam in the middle of semester
    - ↳ Questions that can be solved by hand, so no programming in the exam
    - ↳ It is on **Tuesday, October 21, 2025**
    - ↳ It counts for 25% of the whole mark
- Final Project

# How Do We Get Trained?

*There are three learning components in the course*

- Assignments
- Midterm Exam
- *Final Project*
  - ↳ The **most interesting** part of the course
  - ↳ We build **groups** by **Week 5**
    - ↳ A list of predefined project topics is provided  $\approx$  Week 3
    - Topics are all on Deep Reinforcement Learning*
      - ↳ Each group chooses a topic
      - ↳ **Each group member** submits a progress report by **Week 10**
      - ↳ Each group submits a final paper
      - ↳ Each group submits its implementation
      - ↳ Each group presents its poster in a poster session on last week

# Lots of Programming in Python

*We are going to do lots of programming in Python*

- ↳ Basic knowledge in Python is necessary
- ↳ We use PyTorch and NumPy a lot
  - ↳ *Don't run away if you haven't used them too much*
  - ↳ *If you know Python and have good programming skills, you're fine!*
- ↳ It's important to mention *knowing Deep Learning is a must!*
  - ↳ You may follow without Deep Learning till midterm
  - ↳ After midterm we need to use deep learning
- ↳ We will learn a new library
  - ↳ *Gymnasium*: standard API widely used for Reinforcement Learning

# Course Calendar

Week #	Date	Notes	Posted	Deadline
1	Sep 01 - Sep 05			
2	Sep 08 - Sep 12		Assignment 1: Basics	
3	Sep 15 - Sep 19			
4	Sep 22 - Sep 26			Assignment 1: Basics
5	Sep 29 - Oct 03		Assignment 2: Tabular RL	Project: Flyer and Proposal
6	Oct 06 - Oct 10			
7	Oct 13 - Oct 17			Assignment 2: Tabular RL
8	Oct 20 - Oct 24	Midterm Exam -- Oct 21		
9	Oct 27 - Oct 31	Reading Week-- No Lectures		
10	Nov 03 - Nov 07		Assignment 3: Deep RL	Project: Progress Briefing
11	Nov 10 - Nov 14			
12	Nov 17 - Nov 21			Assignment 4: Deep RL
13	Nov 24 - Nov 28			
14	Dec 01 - Dec 05	Last Lecture on Dec 2		Project: Final Submission and Presentation

- You could check it out [here](#)

# No Major Prerequisites

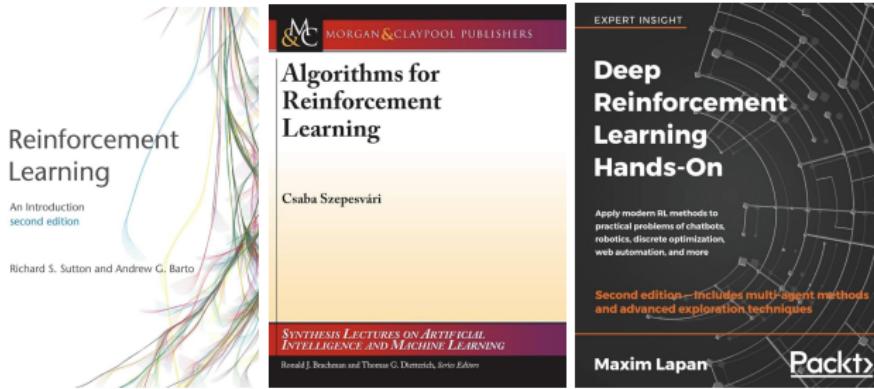
Except Deep Learning, the course is **self-containing** meaning that  
you will learn **all other background!**

We assume that we all have some **basic math** in mind

- ↳ linear algebra, calculus, and probability theory

But we review **whatever we need** from these topics **whenever needed!**

# Textbooks



All materials **are provided in the course**. It's however good to know some texts!

- Sutton abd Brato can be accessed online [at this link](#)
- Szepesvári is available online [here](#)
- Lapan is a good source for Part 3

# Terms and Conditions!



*The instructor keeps the right reserved for himself to modify the slides*

- *last minute before the lecture* ☺
- *after the lecture has been given*
  - *Typically happens due to typos*

*The instructor keeps the right reserved for himself to deliver the lecture-notes*

- *in form of mini-batches* ☺

Date and Signature .....

# Introducing Glum

*Glum does not buy my words! e.g.,*

- + Well! I know Deep Learning, you think you can teach me Reinforcement Learning!
- Sure! Let's try!

*So, please excuse me if I explain things sometimes in too much detail! I need to convince Glum!*



# No such thing as a stupid question!

Did you know that we got [a Wikipedia page](#) on this?

- ↳ Trust me! Your question will **never** sound stupid!
- ↳ **If you don't ask**; then, I need to ask!
  - ↳ **Interaction** is the best tool to avoid getting bored!

Any Questions? ☺