112-1 (Fall 2023) Semester

Reinforcement Learning

Assignment #3

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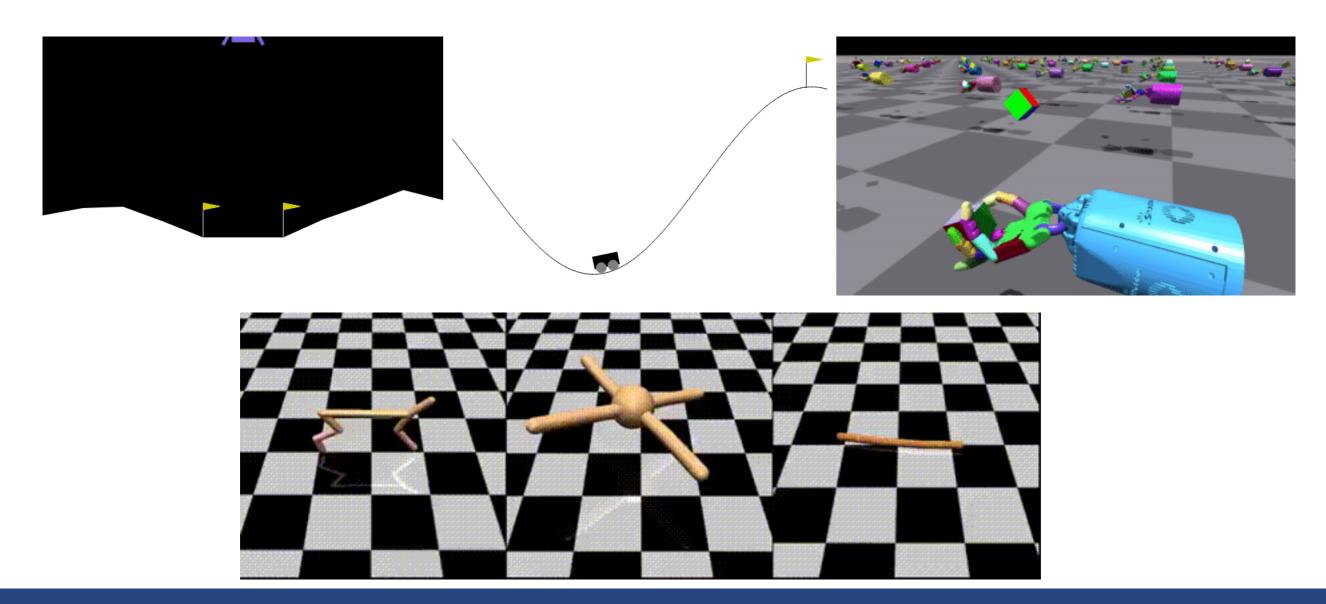
Outline

- Stable baseline 3
- · Tasks
 - Implement environment dynamic
 - Speed up environment
- Environment
- Code structure
- Report
- Grading
- Submission
- Policy
- Contact

Stable Baselines 3

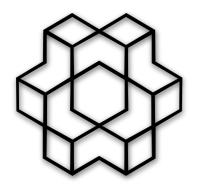
Environment Packages

- There are many open-source environment packages for RL training.
- Examples: "OpenAl Gym", "Deepmind Control", "Isaac Gym", etc.
- For this assignment, we will mainly focus on "OpenAl Gym" for its compatibility with "Stable Baselines 3".



OpenAl Gym / Gymnasium

- Gym is an open source package developed by OpenAI to provide a standard API to communicate between learning algorithms and environments.
- The Farm Foundation announced the release of Gymnasium by October 2022 and took place in the future maintenance of OpenAl Gym. (Announcing The Farama Foundation)
- For this assignment, we will be using Gymnasium.
- OpenAl Gym / Gymnasium





Environment Methods and Tips

Methods: (For <u>Gymnasium</u> or <u>Gym version after v0.26.0</u>)

• step(action): Interacts with the environment. Returns obs, rewards, termination,

truncation, info.

"True" if episode ends because it reaches max episode length.

"True" if episode ends because (ex: Agent made illegal moves or finished a task)

• reset(): Resets the environment. Returns obs, info.

A dictionary of information that might be useful. (ex: {"success": False, "score": 42})

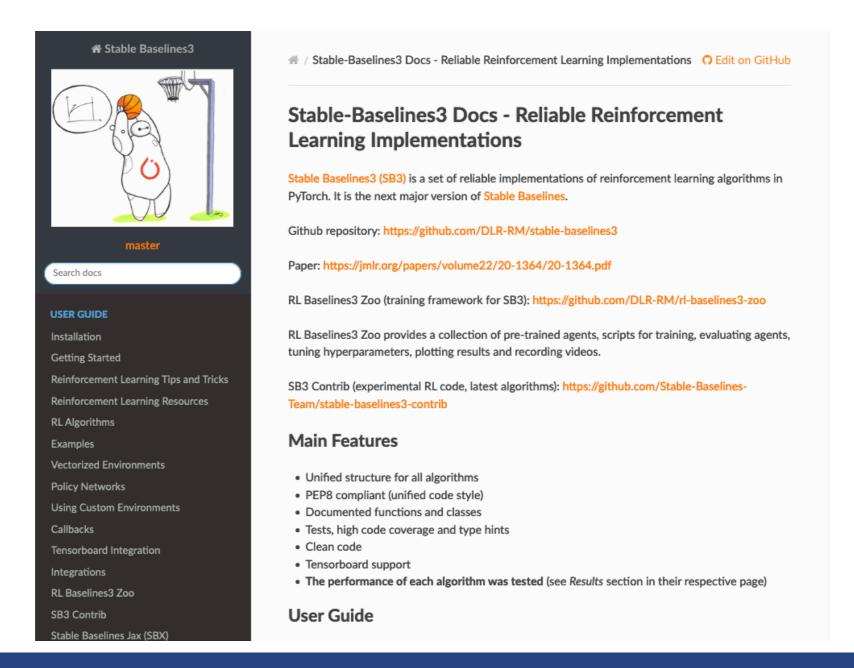
 Note: For a customized Gym environment, you need to first register your environment with gym.envs.register() to use gym.make("ENV_NAME") for returning env object.

Environment Methods and Tips

- There was a major update for the Gym API after version 0.26.0
- If you're using environments from older versions of Gym, you might want to watch out for compatibility issues.
- Release notes for v0.26.0

Stable Baselines 3

- Stable Baselines is a package that provides training APIs for RL training.
- Useful for training Gym environment on basic RL algorithms.
- Documentation



Vectorized Environments

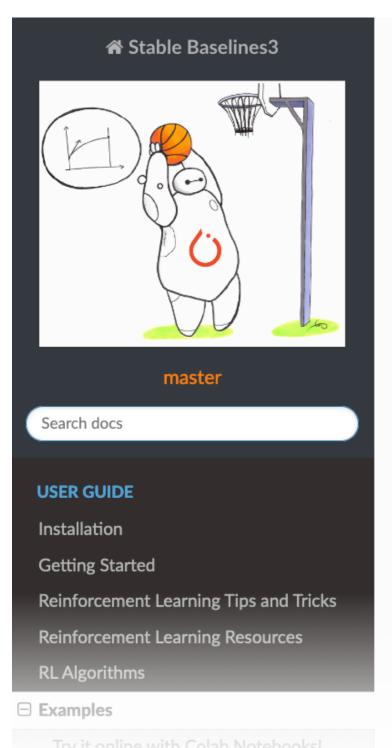
- <u>Vectorized Environments</u> are used to stack multiple independent environments into a single environment.
- Training the agent on a parallel environment helps speed up the training process.
- You can use VecEnv for training and still use the original Env for evaluation.
- Note: VecEnv wraps the environment with Gym 0.21 API logic, so:
 - Output of <u>VecEnv.step()</u> would be <u>obs, reward, termination, info</u> instead.
 - Output of <u>VecEnv.reset()</u> would be <u>obs</u> only.
 - (See VecEnv API vs Gym API)

Supported Algorithms

- A2C (Asynchronous Advantage Actor Critic)
- DDPG (Deep Deterministic Policy Gradient)
- DQN (Deep Q Network)
- PPO (Proximal Policy Optimization)
- SAC (Soft Actor Critic)
- TD3 (Twin Delayed DDPG)
- Note that some algorithms don't support every action and observation space type.
- Make sure that the algorithm you pick fits the environment for this assignment.
- Read <u>RL Algorithms</u> and the documentation of each algorithm for more detail.

Tutorials

Stable Baselines 3 Examples



Assignment #3

/ Examples C Edit on GitHub

Examples

Note

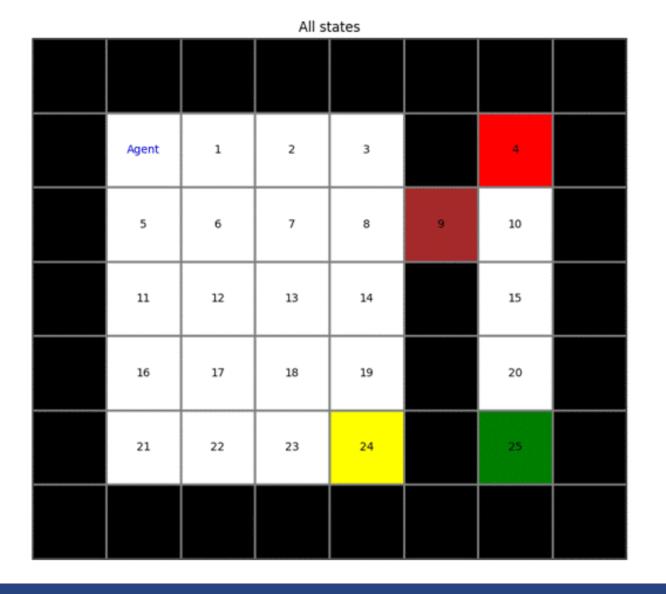
These examples are only to demonstrate the use of the library and its functions, and the trained agents may not solve the environments. Optimized hyperparameters can be found in the RL Zoo repository.

```
import gymnasium as gym
from stable_baselines3 import DQN
from stable_baselines3.common.evaluation import evaluate_policy
# Create environment
env = gym.make("LunarLander-v2", render_mode="rgb_array")
# Instantiate the agent
model = DQN("MlpPolicy", env, verbose=1)
# Train the agent and display a progress bar
model.learn(total_timesteps=int(2e5), progress_bar=True)
# Save the agent
model.save("dgn_lunar")
del model # delete trained model to demonstrate loading
```

Tasks

Embed the GridWorld to SB3 and speed up

- We have learned how to implement algorithms on GridWorld environments. SB3 has some well-known Deep RL algorithms. Make the GridWorld SB3 trainable.
- Add more types of states to the GridWorld and have some interesting findings.
- You can train some algorithms on the GridWorld environment (not for grading)



Environment

Terminal states

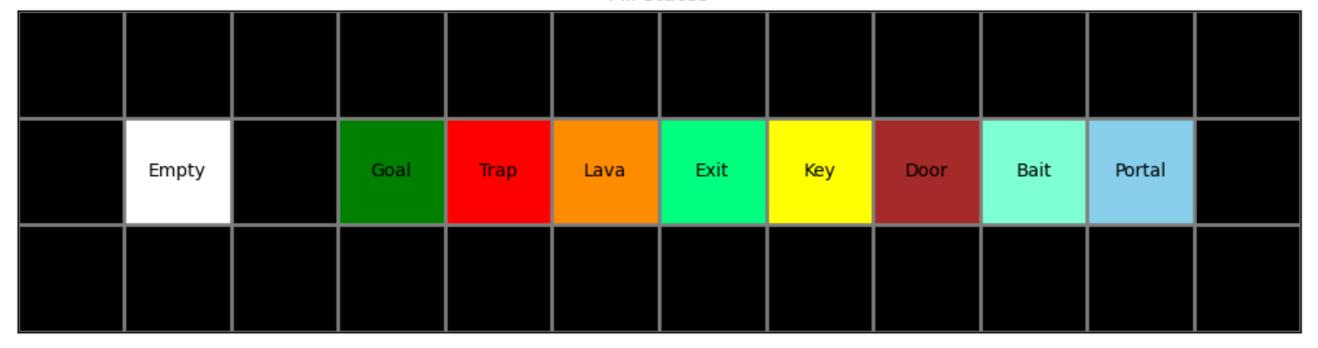
- Original terminal states
 - Trap and Goal: Terminate after taking one step at the state
- New terminal states
 - Exit: Exit the GridWorld with a much less positive reward. The reward is given after taking any step at the Exit step.
 - Lava: Terminate the trajectory when entering the lava.

Non-terminal state

- Original non-terminal states
 - Empty and Wall: The setting of Empty and Wall remains the same as in PA1 and PA2
- New non-terminal states
 - Bait: Enter bait state can get a position bait reward but will have step penalty afterward.
 - Door and Key: The Door will be opened when entering the Key state.
 - Portal: There can be only two portal states in GridWorld. We have to hit the wall at one portal state to teleport to the other portal state.

All states and its color

All states



Reward

- Step reward is given at every transition
- Goal reward is given after reaching goal state
- Trap reward is given after reaching trap state
- Exit reward is given after reaching exit state
- Reach Key can get step reward.
- Going through the opened door can get step reward.
- Reach Lava can get step reward.
- Enter the portal and get a step reward.
- If the agent is baited, the reward is the sum of the step reward and bait penalty.

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Initial state and reset

Initial

- You can not initialize the state in door, key, bait, lava states.
- In PA3, the tasks are left-to-right. You should initialize the state at the left of the Lava and door, and the goal state will be at the right of the Lava and door.
- Door and key and portal-pair are unique.

Reset

- Select one valid initial state.
- Close the door if the door is opened.
- Place the bait back in the bait state if it is bitten.

Inner State and Outer State

- The outer state is returned by the environment for Deep RL policy.
- The inner state indicates the current agent position in GridWorld.
- The outer state is initially the same as the inner state but can differ when door is opened. If the door is opened, the outer state is *inner_state + len(self._state_list)*.

```
next_state = get_next_state(self._current_state)
if door_is_open:
    next_state += len(self._state_list)
```

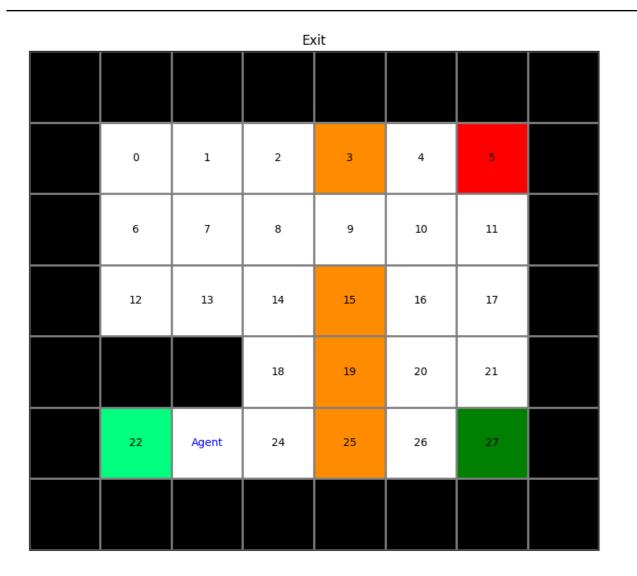
Bait and Door&Key

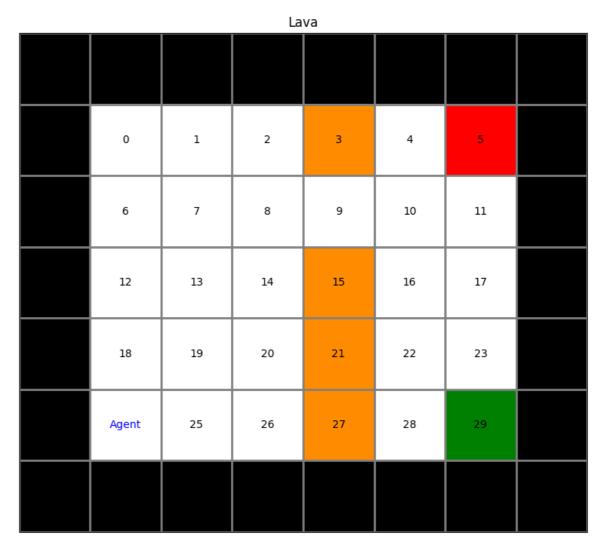
Bait											
	0	1	2	3		4					
	Agent	6	7	8	9	10					
	11	12	13	14		15					
	16	17	18	19		20					
	21	22	23	24		25					

Door											
	Agent	1	2	3		4					
	5	6	7	8	9	10					
	11	12	13	14		15					
	16	17	18	19		20					
	21	22	23	24		25					

- Any transition in the bait state can get a bait reward but will be penalized for every step afterward.
- Going to the Key state can only get a step reward, but the environment changes.

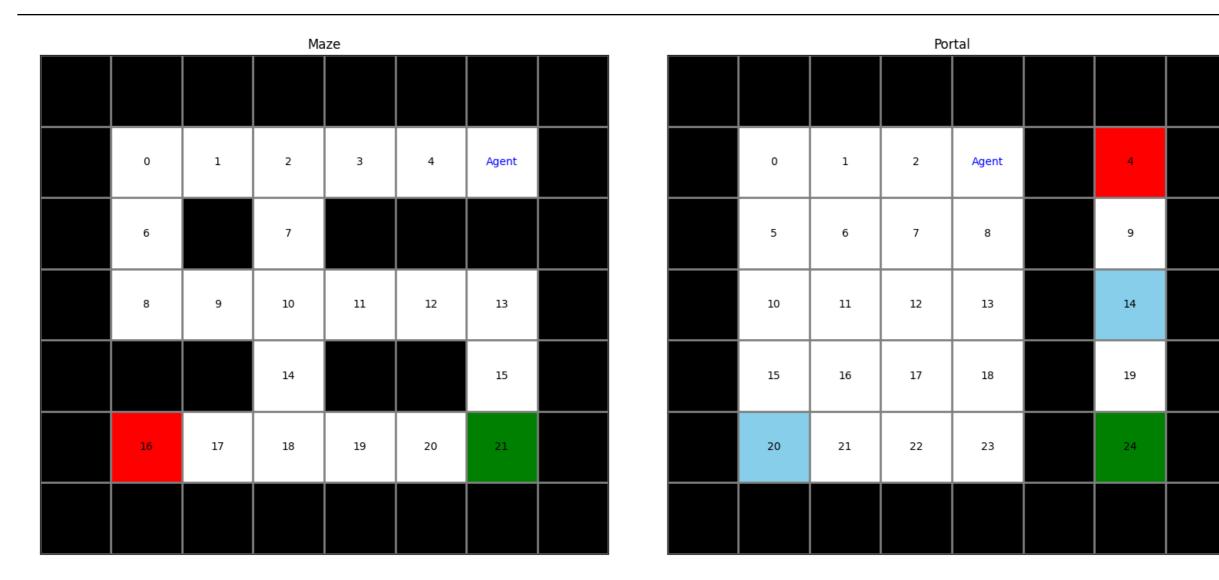
Exit and Lava





- You can not stay in lava states at any time.
- You should initial the Agent left to the Lava.

Maze and Portal



- Hit the wall at portal can go to the other portal with step reward.
- The initial state is not limited to these tasks.

Code Structure

requirement.txt

Conda

```
conda create -n rl_assignment_3 python=3.10
conda activate rl_assignment_3
pip install -r requirements.txt
```

venv

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• Sorry for virtual environment users. Stable Baseline 3 is hard to install using venv with pip. Do not use venv in PA3.



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main.py

- print_sa(obs, action, env)
 - Write state-action pair to console
- test_task(filename, algorithm)
 - Test the termination of your environment.
- test_speed(filename)
 - Test the speed of your implementation
- test_correctness(filename)
 - Test the correctness of your environment
- write_gif(filename, algorithm)
 - Write out one trajectory to gif.

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gridworld.py

- class GridWorld
 - TODO: step(action).
 - TODO: reset(). Unlike PA1 and PA2, reset will not be called inside the class when terminated. SB3 will call reset from outside.
 - ONLY the return values of step(action) and reset() are used for grading. Feel free to change any other parts except the arguments in __init__.
- class GridWorldEnv
 - It is the wrapper function for the gymnasium. Do not change any part of the class.

Other folders

- assets/
 - The assets directory contains trained Deep RL algorithms.
- tasks/
 - There are six sample tasks.
- grading/

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 The grading directory has two types of data. JSON files are the testing files for speed, and CSV files are for correctness.

Submission

Report and Submission

- Deadline 2023/11/09 Thu 09:30am
- No late submission is allowed
- The submission format and report questions will be announced in Assignment #3-2

Grading

Grading

- Correctness (15%)
 - Public Test cases (1% x 6 public cases)
 - Private Test cases (1% x 9 private cases)

```
The correctness of task lava: 100.0 %
The correctness of task exit: 100.0 %
The correctness of task bait: 100.0 %
The correctness of task door: 100.0 %
The correctness of task portal: 100.0 %
The correctness of task maze: 100.0 %
```

- Speed (10%)
 - Public Test cases (1% x 6 cases)
 - Private Test cases (1% x 4 cases)

```
time for task lava: 1.9245 seconds time for task exit: 1.9291 seconds time for task bait: 1.9396 seconds time for task door: 1.9516 seconds time for task portal: 1.9387 seconds time for task maze: 1.9353 seconds
```

Criteria

- Test cases
 - Only check the output of step and reward function.
 - 5 minutes for each test case to avoid infinite loops.
 - You have to be faster than TA's implementation to get points.
 - The private tasks will be more complex and bigger than the sample tasks.
 - The testing sequence of correctness will be pretty long.
 - The truncation will not grade for correctness.
- Sample agents trained on correct dynamics are provided for reference.
 - The trained policy may not be optimal.
 - Some test tasks are sequences of trajectory rolled out by trained agents.

Policy

Policy

Package

- You can use any Python standard library (e.g., heap, queue...)
- System level packages are prohibited (e.g., sys, multiprocess, subprocess...) for security concern

Collaboration

- Discussions are encouraged
- Write your own codes

Plagiarism & cheating

- All assignment submissions will be subject to duplication checking (e.g., MOSS)
- Cheater will receive an **F** grade for this course

Grade appeal

Assignment grades are considered finalized two weeks after release

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Contact

Questions?

- General questions
 - Use channel **#assignment 3** in slack as first option
 - Reply in thread to avoid spamming other people
- Personal questions
 - ・ DM us on Slack: TA 李威緒 Wei-Hsu Lee

TA 林辰宇 Chen-Yu Lin



