

# SEED RL: SCALABLE AND EFFICIENT DEEP-RL WITH ACCELERATED CENTRAL INFERENCE

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# Outline

- Purpose
- What is Reinforcement Learning ?
- What is Deep Reinforcement Learning?
- What is this special project's goal ?
- What is SEED-RL ?
- Experiments

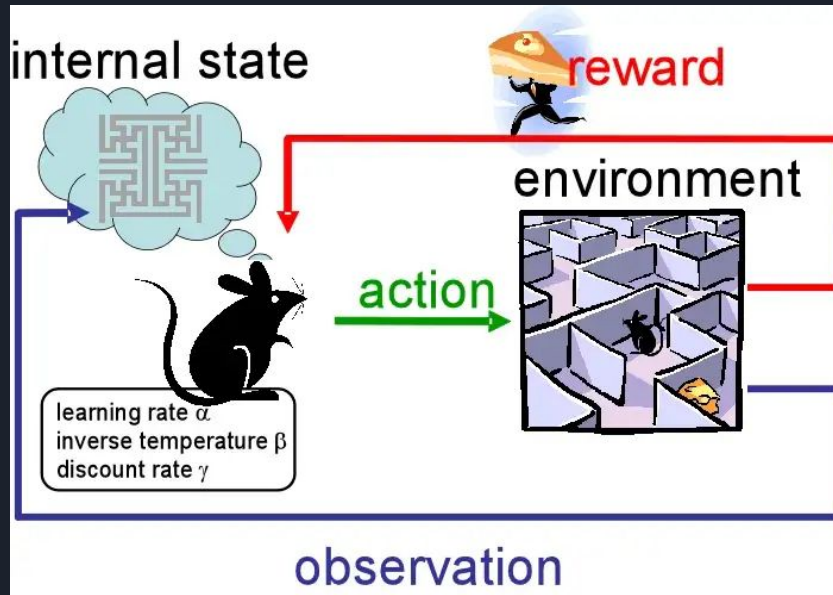
# Purpose

- Altiscan Crawler
  - Used for inspection and maintenance of marine vessels
  - Attached by an umbilical that bundles data, power, and water supply cables
- Irregular behavior as the crawler traverses larger structures
  - Currently controlled by inverse velocity kinematics
- Velocity control is necessary for localization, defect detection, and coordination



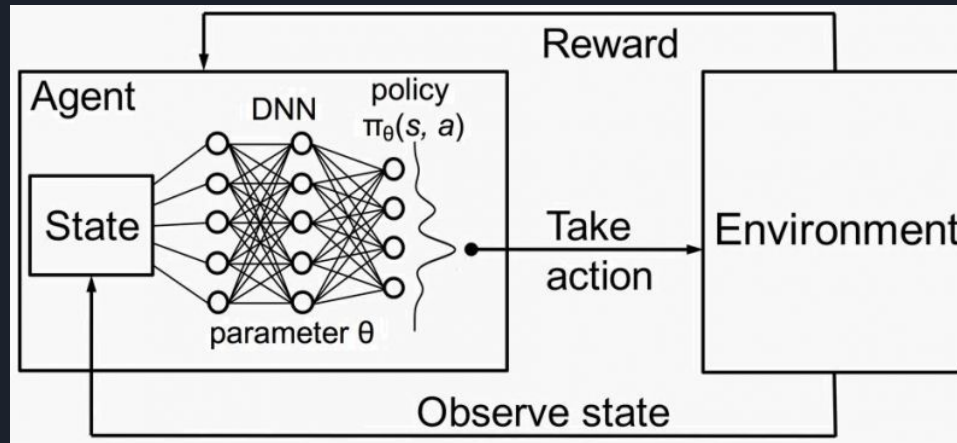
Source: <https://www.directindustry.fr/prod/roboplanet/product-64387-1763534.html>

# What is Reinforcement Learning ?



# What is Deep Reinforcement Learning (DRL)?

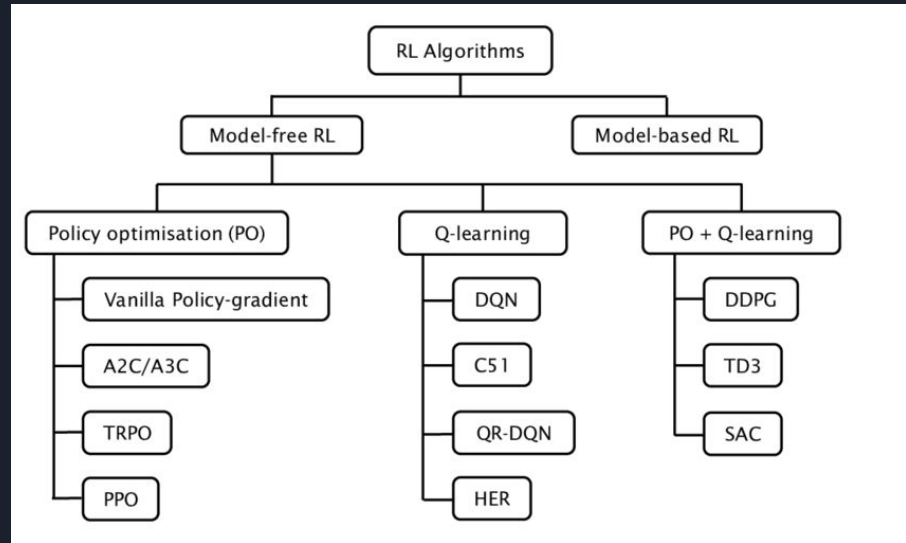
- Extends RL using deep neural networks (DNNs) to tasks with complex environments
  - Continuous state and action spaces



Source: <https://medium.com/@vishnuvijayanpv/deep-reinforcement-learning-value-functions-dqn-actor-critic-method-backpropagation-through-83a277d8c38d>

# What is DRL?

- Policy Gradient
  - How do we update the DNN params to choose better actions?
- Actor-Critic
  - How do we improve learning stability?
- Trust-Region Policy Optimization (TRPO)
  - How do we account for large gradient steps?
- Proximal Policy Optimization (PPO)
  - How do we use less calculus?



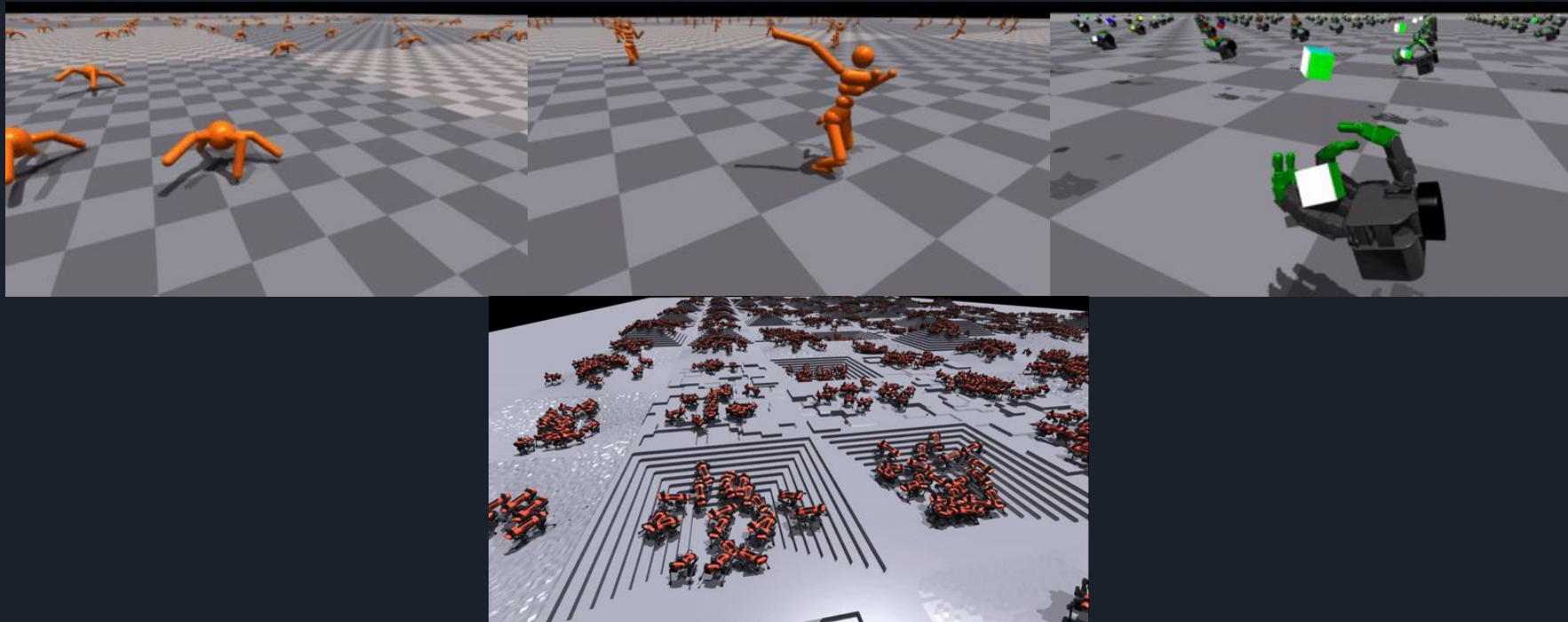
Source: [https://www.researchgate.net/figure/Class-of-model-free-RL-algorithms-adapted-from-21\\_fig2\\_351105046](https://www.researchgate.net/figure/Class-of-model-free-RL-algorithms-adapted-from-21_fig2_351105046)



# Learning Robotic Control with DRL

- DRL needs to take observations from the environment and produce actions to control a robot
- Training agents on physical robots can be costly and dangerous
- How do we efficiently train an agent for robotic control?
  - Use simulation!

# Isaac Gym



Source: <https://developer.nvidia.com/isaac-gym>

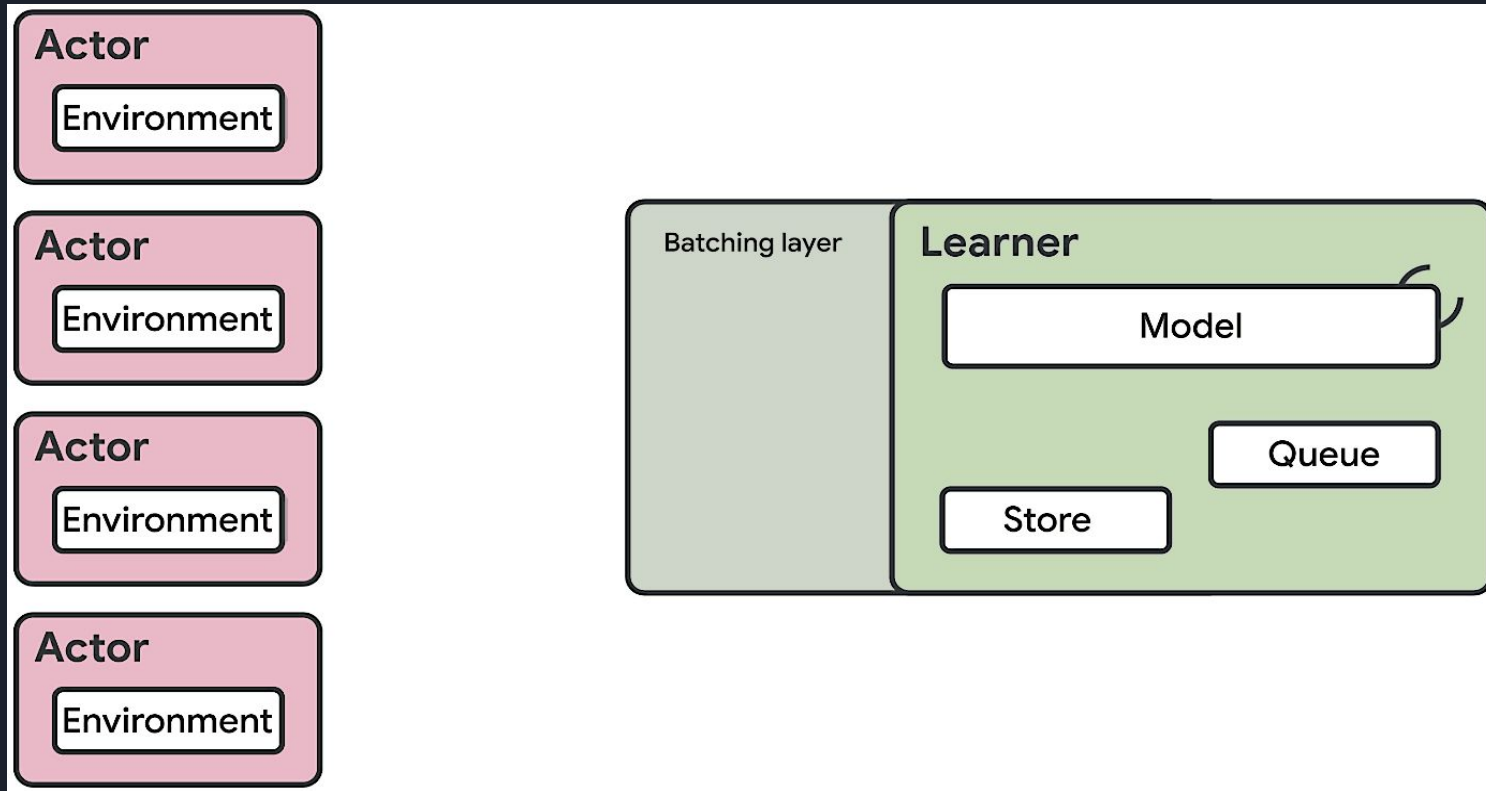




# What is this special project's goal ?

- Reuse existing infrastructure
- Speed-up CPU-based environments
- Create a benchmark for GPU-based environments
- Make a reusable framework for future GTL projects

# What is SEED-RL ?



# SEED-RL vs Isaac Gym

- Isaac Gym
  - 1024 actors
  - cost < 0.296 \$
  - 21 917 steps per second
  - 204 hours of simulation
- Small SEED-RL
  - 3.5 actors
  - cost = 0.807 \$
  - 260 steps per second
  - 42 hours of simulation
- Big SEED-RL
  - Started with 26 actors, but we lost many
  - cost = 0.517 \$
  - 2 988 steps per second
  - 44 hours of simulation

Each actor uses ~70% CPU

Resource	Cost per hour
CPU core	\$0.0475
Nvidia Tesla P100	\$1.46
TPU v3 core	\$1.00

Table 3: Cost of cloud resources as of Sep. 2019.

Table 6.2: Runtime performance.

Simulation Software	DRL Library	Training Time
Isaac Gym	rl-games	12m10s
PyBullet	SEED RL	17h3m16s
PyBullet	SEED-RLv2	1h39m12s



Figure 6.2: Training Reward for Isaac Gym and PyBullet



# References

- SEED RL: Scalable and Efficient Deep-RL with Accelerated Central Inference. Lasse Espeholt and Raphael Marinier and Piotr Stanczyk and Ke Wang and Marcin Michalski. 2019
- The very basics of Reinforcement Learning, Aneek Das.  
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- N. Rudin, D. Hoeller, P. Reist, and M. Hutter, *Learning to walk in minutes using massively parallel deep reinforcement learning*, 2021.
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