



# Q-Learning + Project 4 Introduction

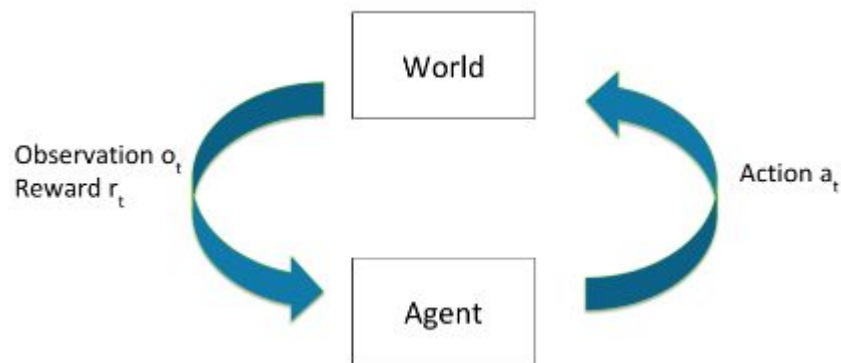
TAs: Shimian Zhang, Keaton Kraiger  
03/21/2023



First, let's talk about Q-learning

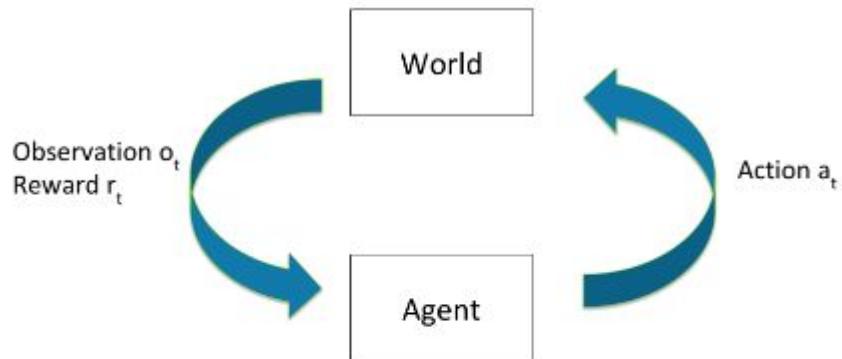
# Q-Learning for RL

- Q-learning is a classical Reinforcement Learning (RL) approach



- Each time step  $t$ :
  - Agent takes an action  $a_t$
  - World updates given action  $a_t$ , emits observation  $o_t$  and reward  $r_t$
  - Agent receives observation  $o_t$  and reward  $r_t$

# History



- History  $h_t = (a_1, o_1, r_1, \dots, a_t, o_t, r_t)$
- Agent chooses action based on history
- State is information assumed to determine what happens next
  - Function of history:  $s_t = (h_t)$



# Policy

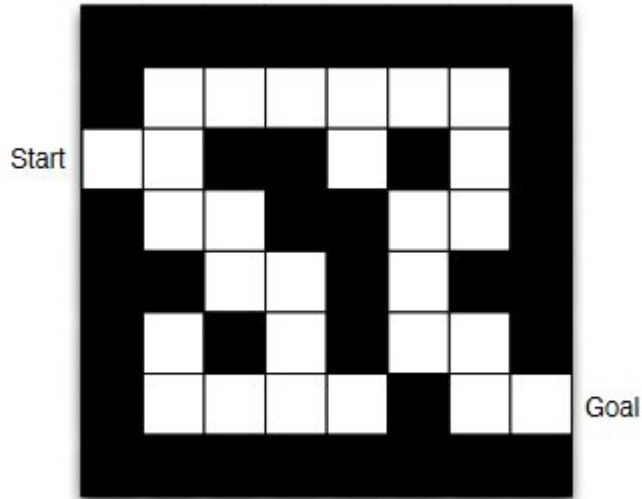
- Want to learn a policy (plan/strategy) for agent to use in environment
  - Policy  $\pi$  determines how the agent chooses actions
  - $\pi: S \rightarrow A$ , mapping from states to actions
- Deterministic policy:
  - $\pi(s) = a$
- Note you could instead have a stochastic policy



# Q-Learning

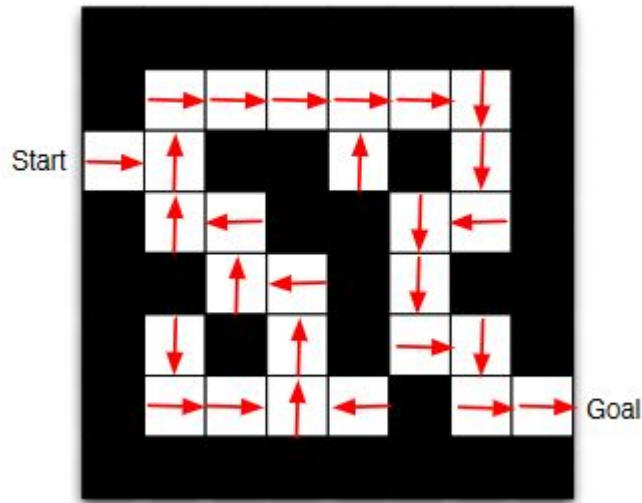
- Q-learning is a model-free based algorithm
  - Model-based
    - Explicit: Model
    - May or may not have a policy and/or value function
  - Model-free
    - Explicit: Value function and/or policy function
    - No model

# Maze Example



- Rewards: -1 per time-step
- Actions: N, E, S, W
- States: Agent's Location

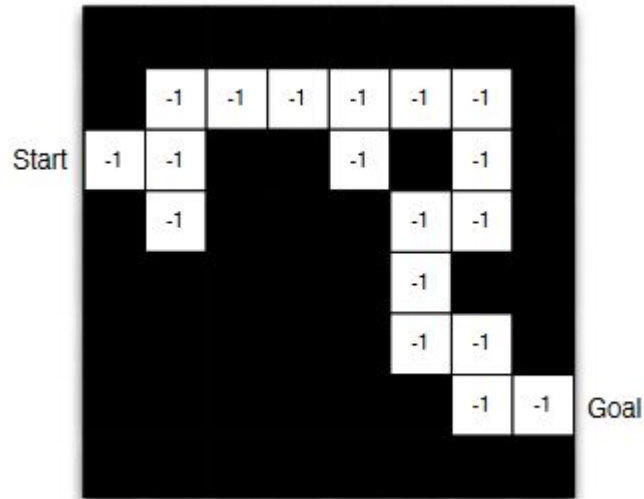
## Maze Example - Policy



- Arrows represent policy  $\pi(s)$  for each state  $s$

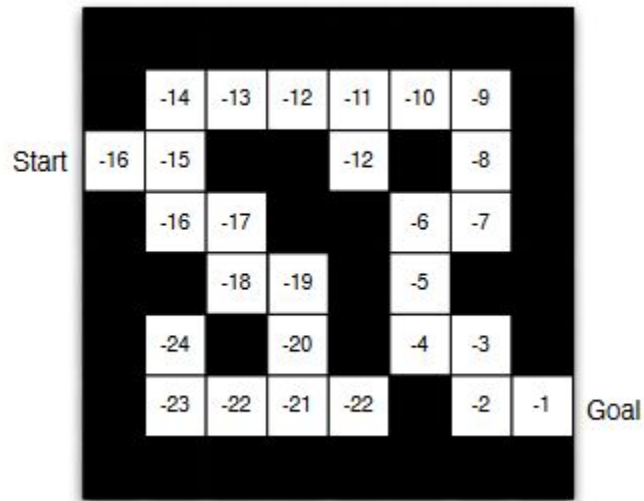


## Maze Example - Model



- Grid layout represents transition model
- Numbers represent immediate reward for each state  $s$ 
  - Model-based approach

## Maze Example: Value Function



- Numbers represent value  $v_{\pi}(s)$  of each state  $s$ 
  - Model-free approach

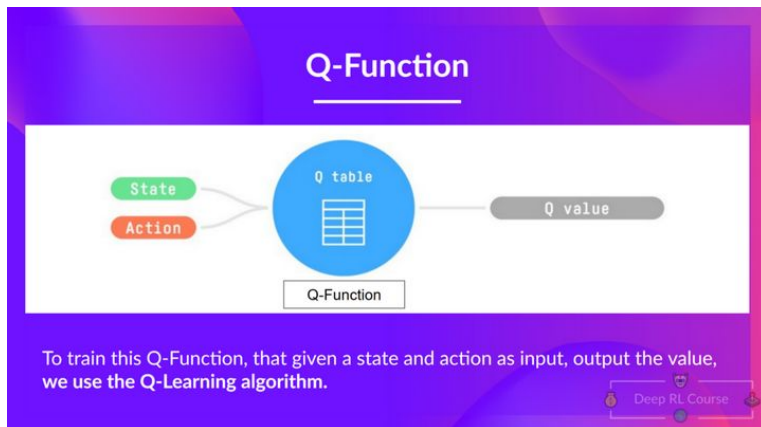


# Quickly, On and Off-Policy Learning

- Off-policy
  - Learn to estimate and evaluate a policy using experience gathered from following a different policy
- On-policy
  - Direct experience
  - Learn to estimate and evaluate a policy form experience obtained from following that policy

# Q-Learning Algorithm

- Q-learning is an off-policy method for training a Q-function
  - Finds policy indirectly by training a value or action-value function that will tell us the value of each state or each state-action pair



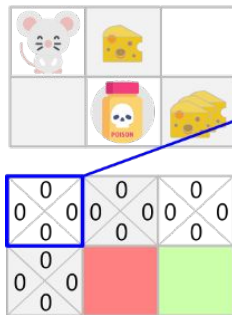
Given a state and action, our Q-function outputs a state-action value (or Q-value)

$$Q(S_t, A_t) \leftarrow Q(S_t, A_t) + \alpha [R_{t+1} + \gamma \max_a Q(S_{t+1}, a) - Q(S_t, A_t)]$$

New Q-value estimation      Former Q-value estimation      Learning Rate      Immediate Reward      Discounted Estimate optimal Q-value of next state      Former Q-value estimation

# Q-Table

- Our Q-function has a Q-table where each cell corresponds to a state-action value pair value.



States

Actions

	0	0	0	0
	0	0	0	0
	0	0	0	0
	0	0	0	0
	0	0	0	0

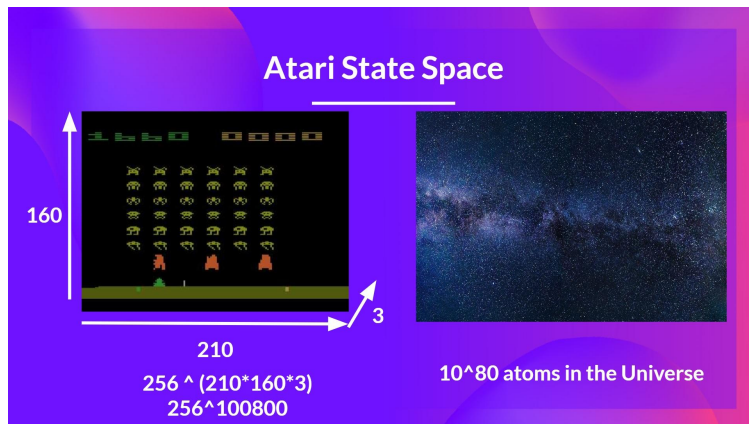
## Q-Table

- Initially, the Q-table is useless as it's arbitrarily initialized.
- Need to explore to update it!



# Problem with (classical) Q-Learning

- Q-learning is a tabular method
- Internally, our Q-function has a Q-table, a table where each cell corresponds to a state-action pair value. Think of this Q-table as the memory or cheat sheet of our Q-function.

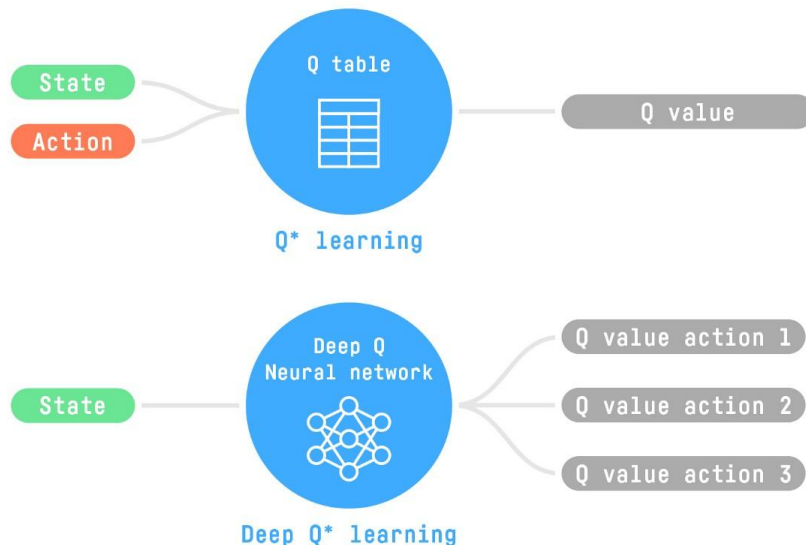


Difficult  
to scale!

# Deep Q-Learning

- Learn to approximate Q-values

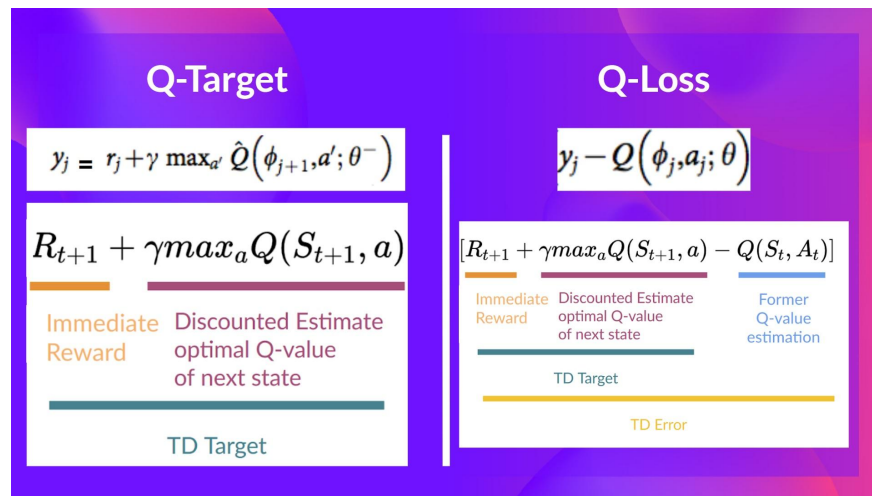
Original DQN Paper:  
[Human-level control through deep reinforcement learning](#)





# Deep Q-Learning

- Deep Q-learning uses a deep neural network to approximate different Q-values for each possible action at a state (value estimation)





Let's now discuss project 4!



# Introduction

- This project will explore regularity discovery using PRML techniques.
- You will investigate an existing regularity-related method, work it a baseline method, and then further explore the topic
- Release Date: Tuesday, March 21
- Submission Due Date: **3, April 2023**

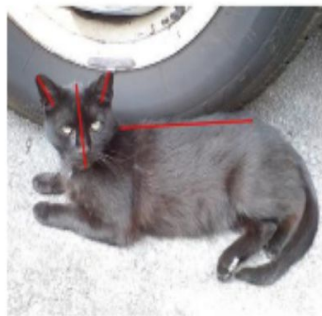


# Regularity Discovery Topics

- You will choose to investigate one of six regularity-related topics from real data
  - Reflection Symmetry Detection
  - Rotation Symmetry Detection
  - Translation Symmetry Detection
  - Glide-Reflection Symmetry Detection
  - Recurring Pattern Detection
  - Reinforcement Learning-Based Regularity Understanding
- For each method you will...

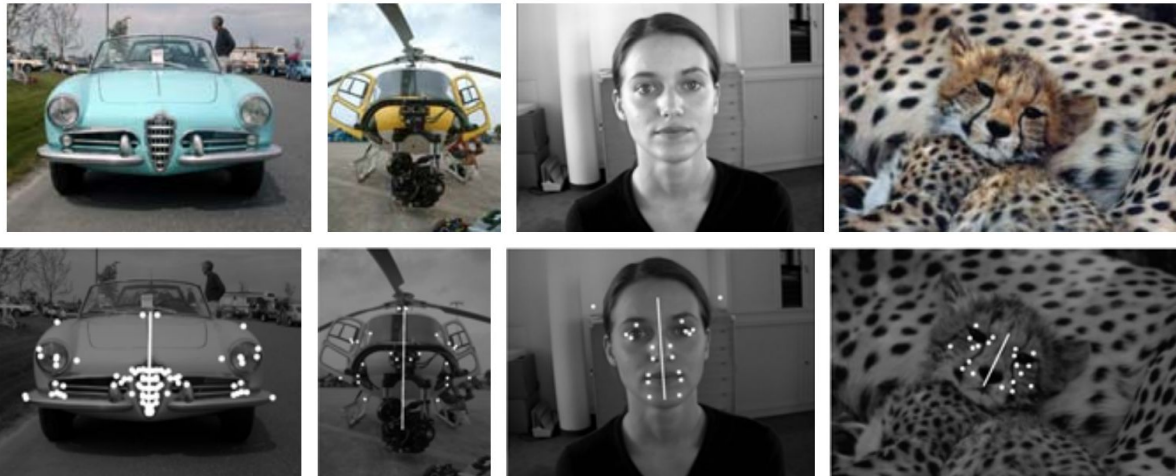
# Reflection Symmetry Detection

The goal of reflection symmetry detection aims to discover a reflection axis/plane that remains invariant under reflection.



## Baseline Method

Loy, Gareth, and Jan-Olof Eklundh. "Detecting symmetry and symmetric constellations of features." Computer Vision–ECCV 2006: 9th European Conference on Computer Vision, Graz, Austria, May 7-13, 2006. Proceedings, Part II 9. Springer Berlin Heidelberg, 2006.



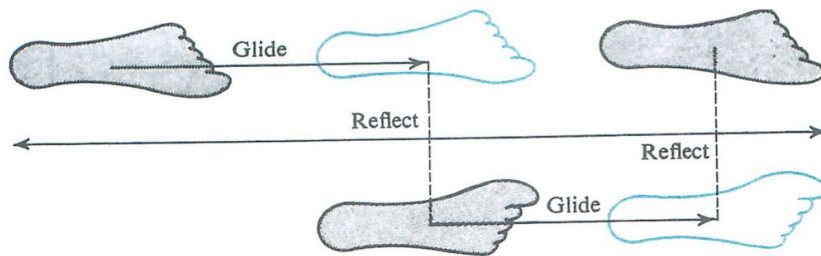
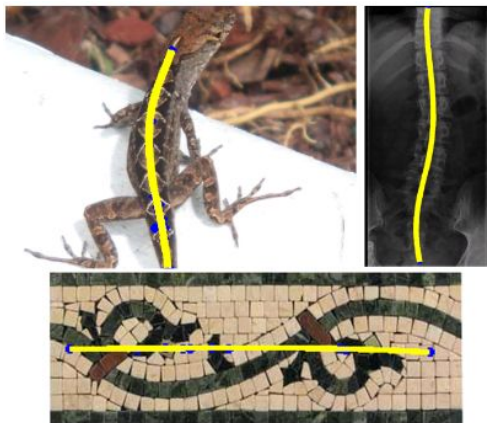
## Methods for Investigation

Seo, Ahyun, Byungjin Kim, Suha Kwak, and Minsu Cho. "Reflection and Rotation Symmetry Detection via Equivariant Learning." In Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition, pp. 9539-9548. 2022.



# Glide-Reflection Symmetry Detection

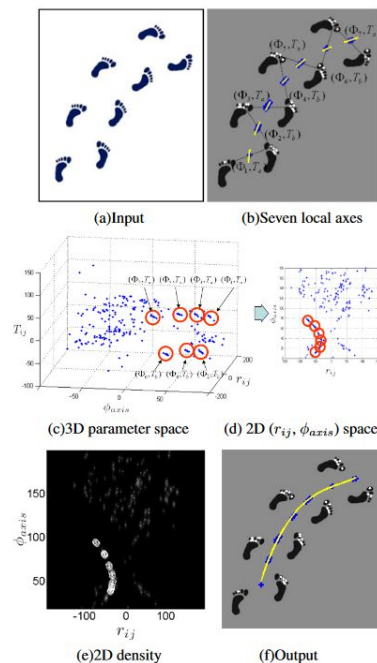
Glide-reflection is another symmetry primitive. Glide-reflection is a combination of two symmetries that consists of an axis of reflection as well as translation along that axis.





# Baseline Method

Lee, Seungky, and Yanxi, Liu. "Curved Glide-Reflection Symmetry Detection". Proceedings of the IEEE Transactions on Pattern Analysis and Machine Intelligence. 2011.



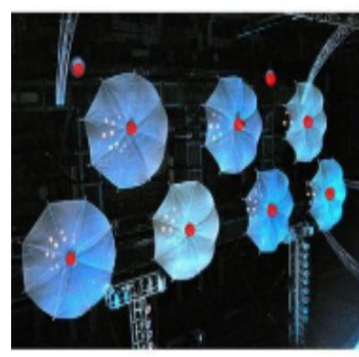
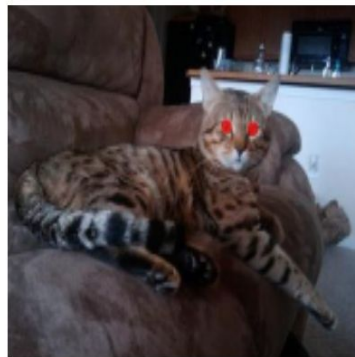
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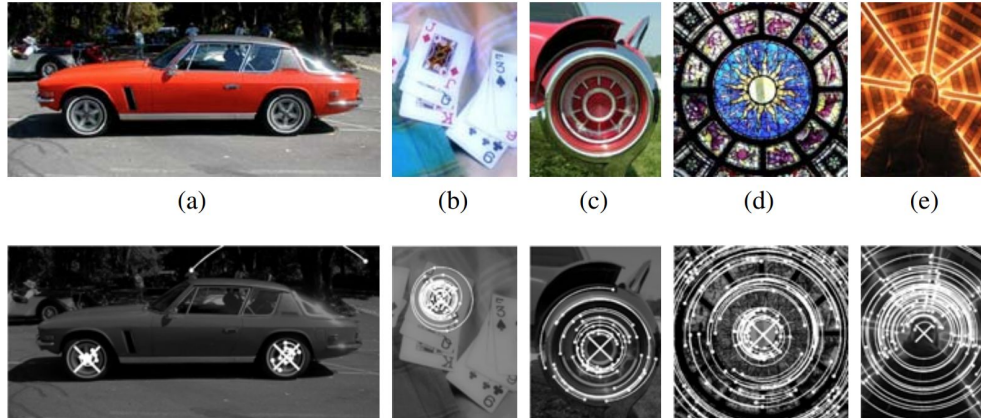
# Rotation Symmetry Detection

The goal of rotation symmetry detection is to find a rotation center invariant of rotation.



## Baseline Method

Loy, Gareth, and Jan-Olof Eklundh. "Detecting symmetry and symmetric constellations of features." Computer Vision–ECCV 2006: 9th European Conference on Computer Vision, Graz, Austria, May 7-13, 2006. Proceedings, Part II 9. Springer Berlin Heidelberg, 2006.



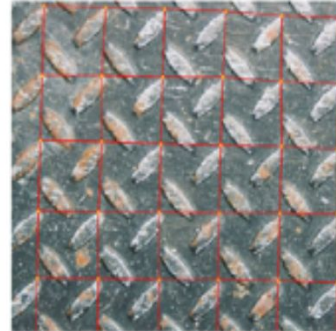
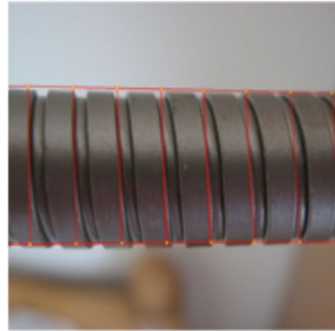
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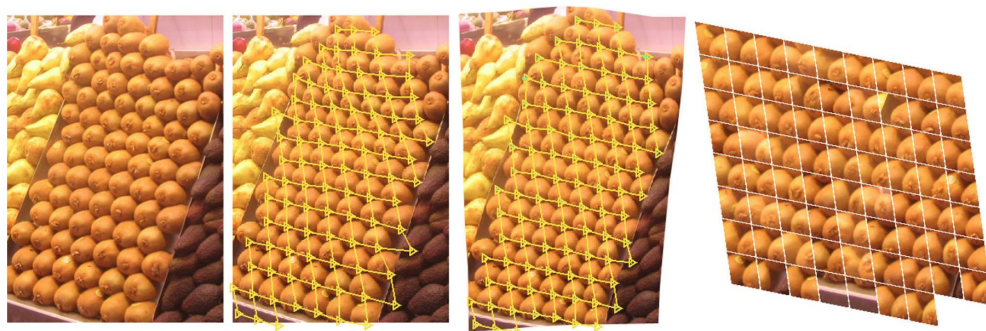
# Translation Symmetry Detection

There are two types of translation (1) Translation with repetition in one direction, also known as **Frieze Pattern**. (2) Translation with repetition in two directions, also known as **Wallpaper Pattern**.



## Baseline Method

- Leordeanu, Marius, et al. "Discovering texture regularity as a higher-order correspondence problem." (2006).
- Park, Minwoo, et al. "Translation-symmetry-based perceptual grouping with applications to urban scenes." Computer Vision–ACCV 2010: 10th Asian Conference on Computer Vision, Queenstown, New Zealand, November 8-12, 2010, Revised Selected Papers, Part III 10. Springer Berlin Heidelberg, 2011.

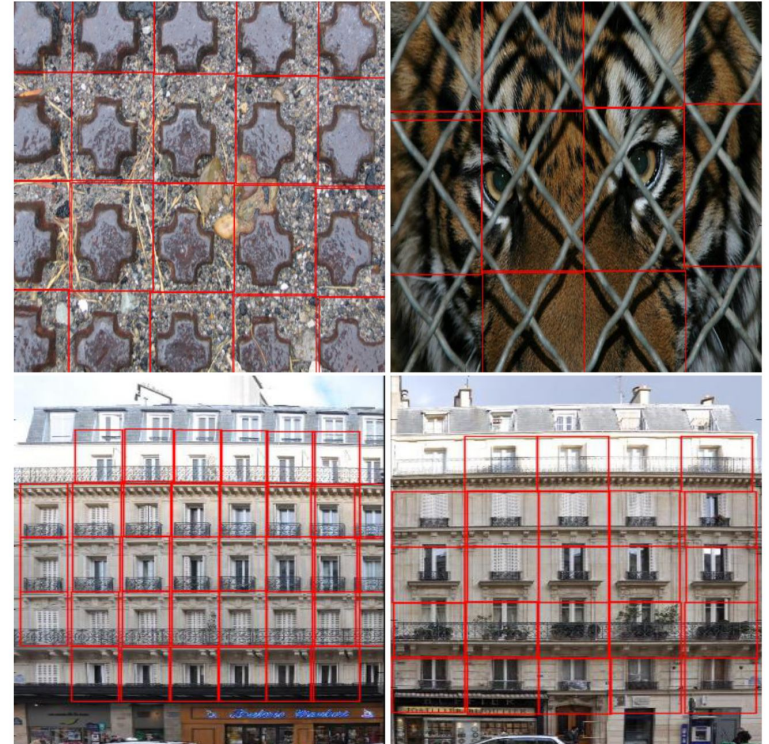


**Fig. 4.** From left to right: input, final lattice, warped texture, and extracted texels.



# Methods for Investigation

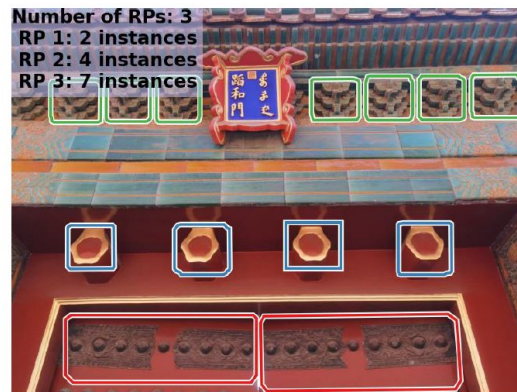
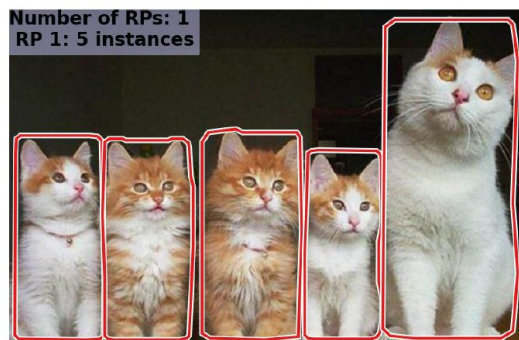
Rodriguez-Pardo, Carlos, et al. "Automatic extraction and synthesis of regular repeatable patterns." Computers & Graphics 83 (2019): 33-41.





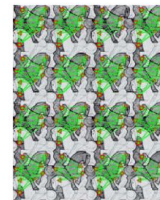
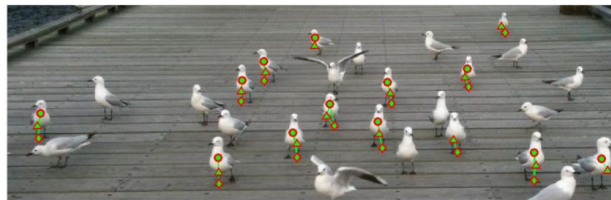
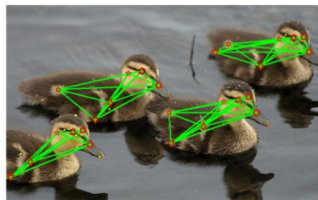
# Recurring Pattern Discovery

Recurring pattern (RP) detection refers to the detection of "things that recur" in images, ranging from perfect symmetry to real-world similar objects.



## Baseline Method

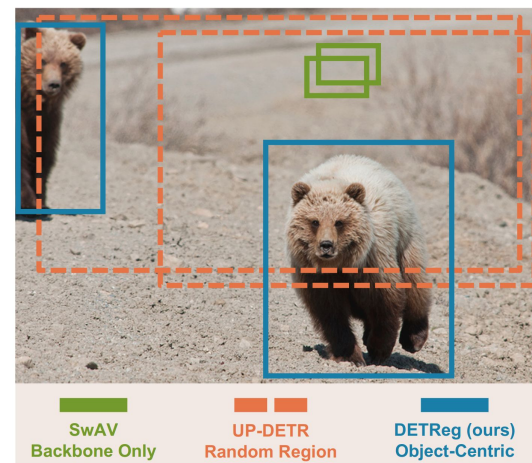
Liu, Jingchen, and Yanxi Liu. "Grasp recurring patterns from a single view." Proceedings of the IEEE conference on computer vision and pattern recognition. 2013.



# Methods for Investigation

## Unsupervised Object Detection:

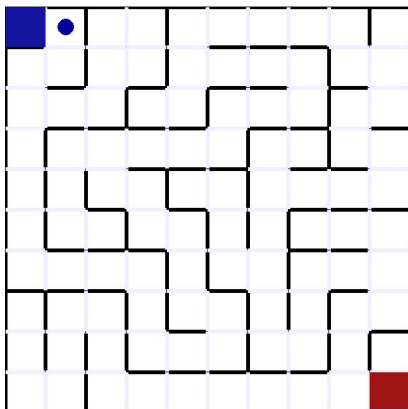
- Wang, Xudong, et al. "Cut and Learn for Unsupervised Object Detection and Instance Segmentation." arXiv preprint arXiv:2301.11320 (2023).
- Bar, Amir, et al. "Detreg: Unsupervised pretraining with region priors for object detection." Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition. 2022.



# Regularity in Reinforcement Learning

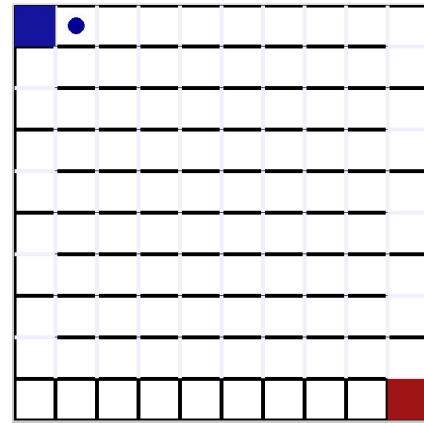
We will be working with Q-learning, where an agent attempts to learn an optimal policy when acting in an environment.

“Normal maze”



How does an agent perform  
→  
in a maze with regularity?

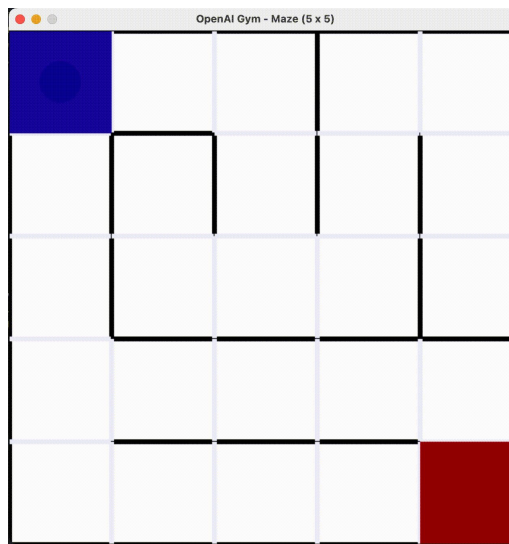
Maze with regular elements



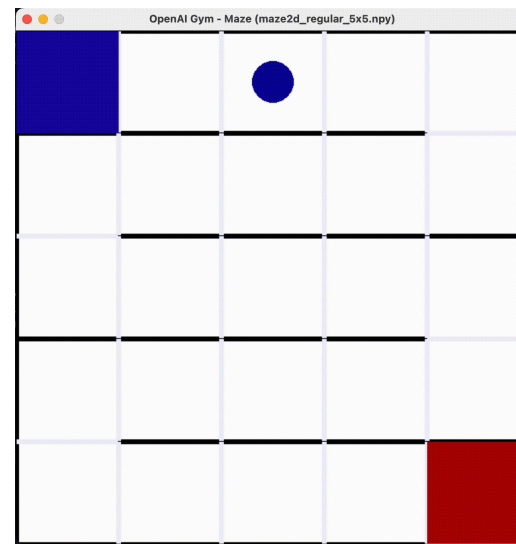
# Regularity in RL

- We will be evaluating agent performance on
  - Random/non-regular maps
  - Maps with regular elements

Random Layout



Regular Layout





# Grading Criteria

- Part **one** (all methods):
  - Summarize the baseline
  - Try it out
  - Report on it
- Part **two** (option **one**) Symmetry or RP detection:
  - Apply new/different method
  - Report on it and compare it
- Part **two** (option **two**) RL:
  - Compare classical Q-learning to deep Q-learning
  - Measure performance on maps with regular elements
  - Report on results



## Q & A