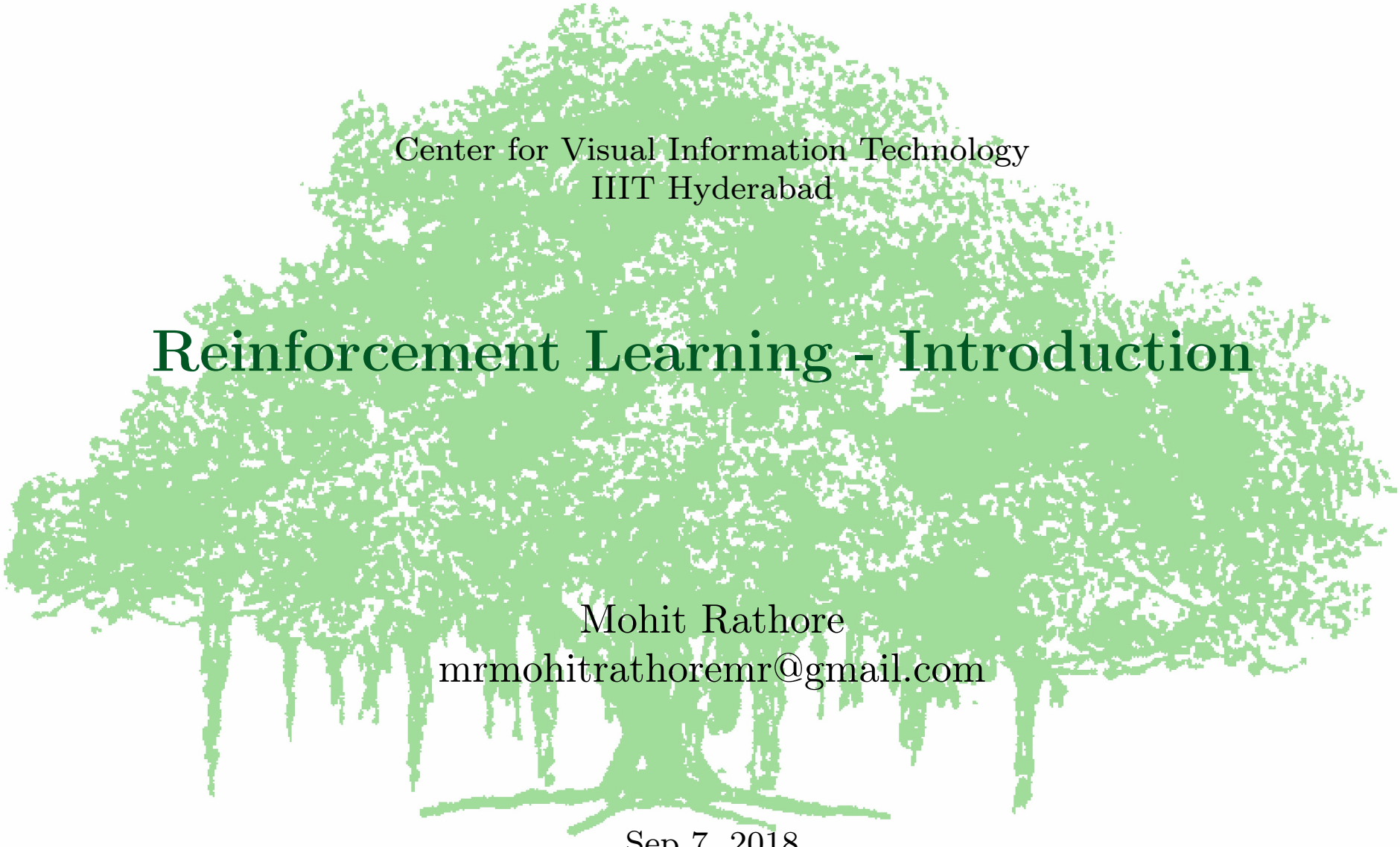

Center for Visual Information Technology
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Reinforcement Learning - Introduction

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Sep 7, 2018

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Trivia

- Motivation
- News

Deep Q Learning

- Understanding the terms.
- The approach.
- Flaws with DQN.

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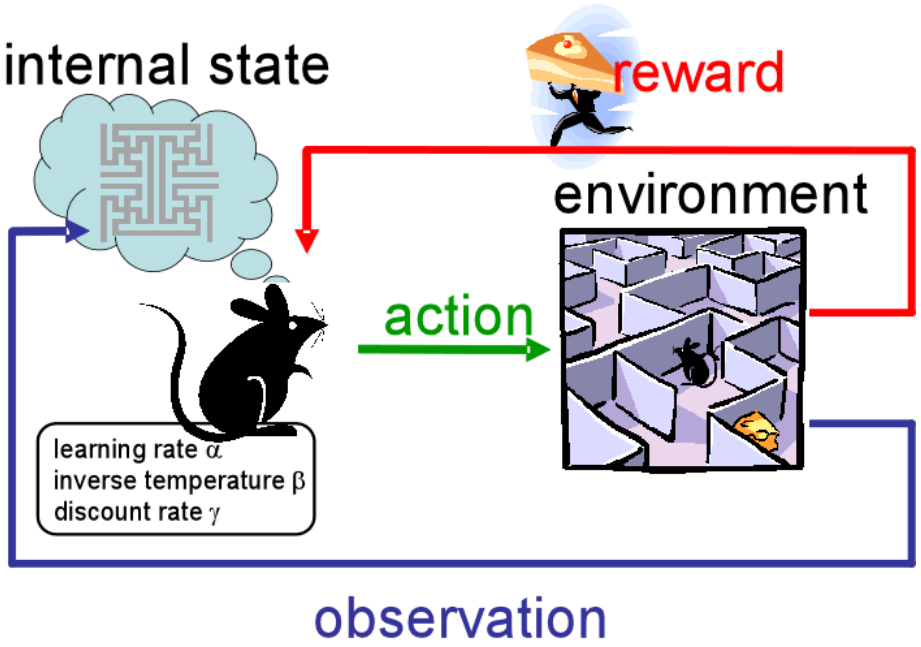
Trivia
Why should I care?

- At the very core deep learning can solve problem of classification, regression i.e. approximating complex functions.
- Reinforcement learning on the other hand help's us devise an optimum strategy to follow in our feat of maximizing reward.

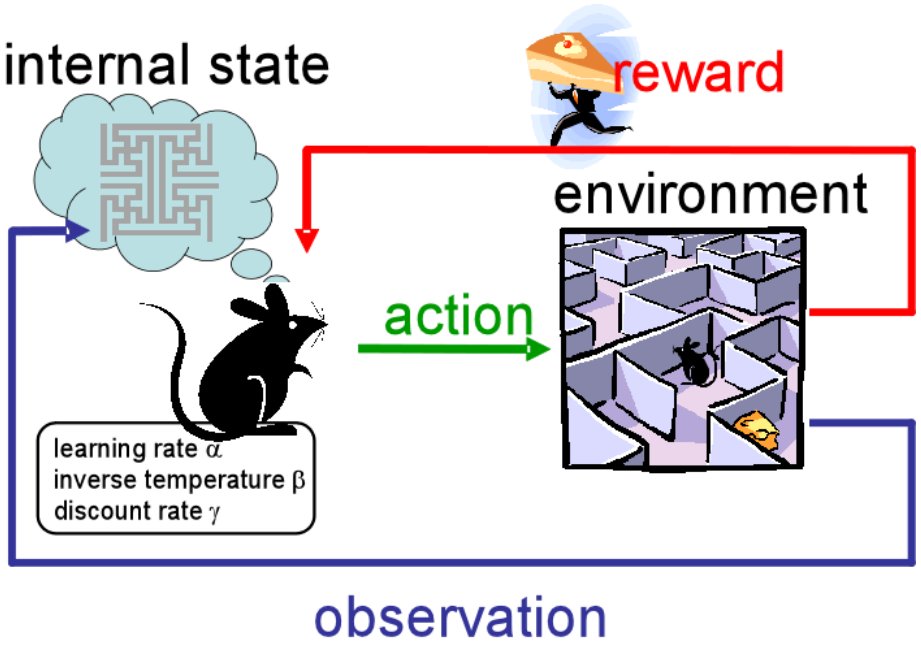
Trivia
News

- RL Beats Humans at Dota 2. Link
- AlphaGo beats 18-time world champion Lee Sedol. Link
- Google has developed an approach to hyperparameter tuning using reinforcement learning that they call AutoML. Link

- **State** - It can be seen as a variable which directly affect the action taking mechanism. In an environment such as a maze, a state can be seen as the x, y coordinates of the agent.



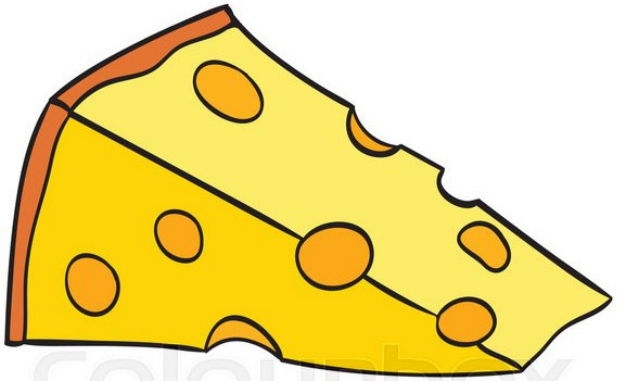
- **Action** - Performing a particular action in a particular state takes you to the next state.
- The mouse can move in either of the four directions to reach the next block.



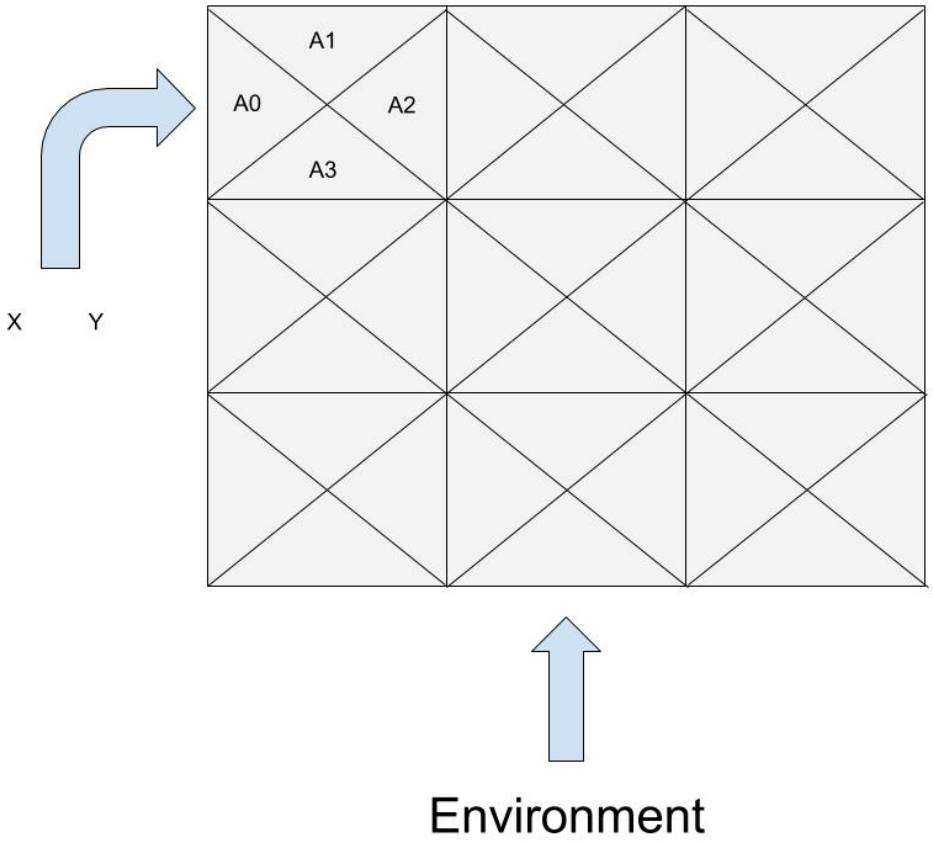
Introduction
Jargons

- **Agent** - It is an abstraction of the entity which acts and learns from previous experiences in hope to maximize reward.
- In this case our agent is - mouse.

- **Reward** - It is the utility that our agent is trying to maximize.
- For our mouse reward is - cheese.

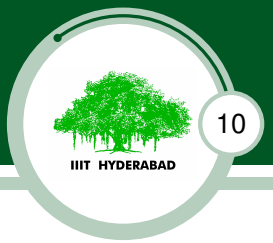


- **Environment** - Environment can be considered as a entity which takes in space-action and returns the next state and reward for that action.



Reinforcement Learning
Deep Q learning

- 0.08 - 0.08 0.88 - 0.09	- 0.04 - 0.07 0.92 - 0.04	- 0.04 - 0.04 0.96 - 0.03	
0.84 - 0.10 - 0.10 - 0.10		0.51 - 0.05 - 0.41 - 0.05	
0.79 - 0.12 - 0.12 - 0.12	- 0.10 - 0.12 - 0.10 - 0.10	0.08 - 0.09 - 0.08 - 0.09	- 0.27 - 0.08 - 0.08 - 0.08



What can be the best case scenario?

What can be the best case scenario?

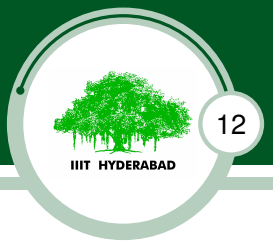
Deep Q-Network Training

- Bellman Equation:

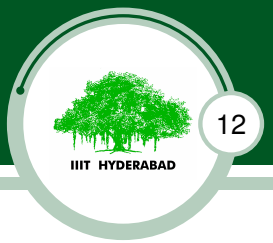
$$Q(s,a) = r + \gamma max_{a'} Q(s',a')$$

- Loss function (squared error):

$$L = \mathbb{E}[(\underbrace{\boldsymbol{r + \gamma max_{a'} Q(s',a')}}_{\textbf{target}} - Q(s,a))^2]$$



Why discount the reward by gamma?



Why discount the reward by gamma?

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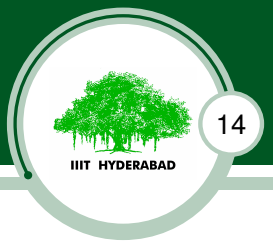
Why discount the reward by gamma?

- So that rewards do not diverge as the time steps increase.
- We will be running in circles.

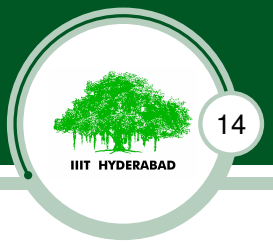
Exploration!

- We take the optimum action but sometimes we take a random action with exploration probability ϵ so that surroundings are investigated for possible rewards.
- We decay this ϵ over time exponentially.

$$\epsilon = \epsilon_{min} + (\epsilon_{max} - \epsilon_{min}) * \exp^{(-\lambda * t)}$$



- **Explore** - Take random actions and create a queue of (current state, action, next state, reward).



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  - Taking a random action with a exploration probability (which decrease over time).
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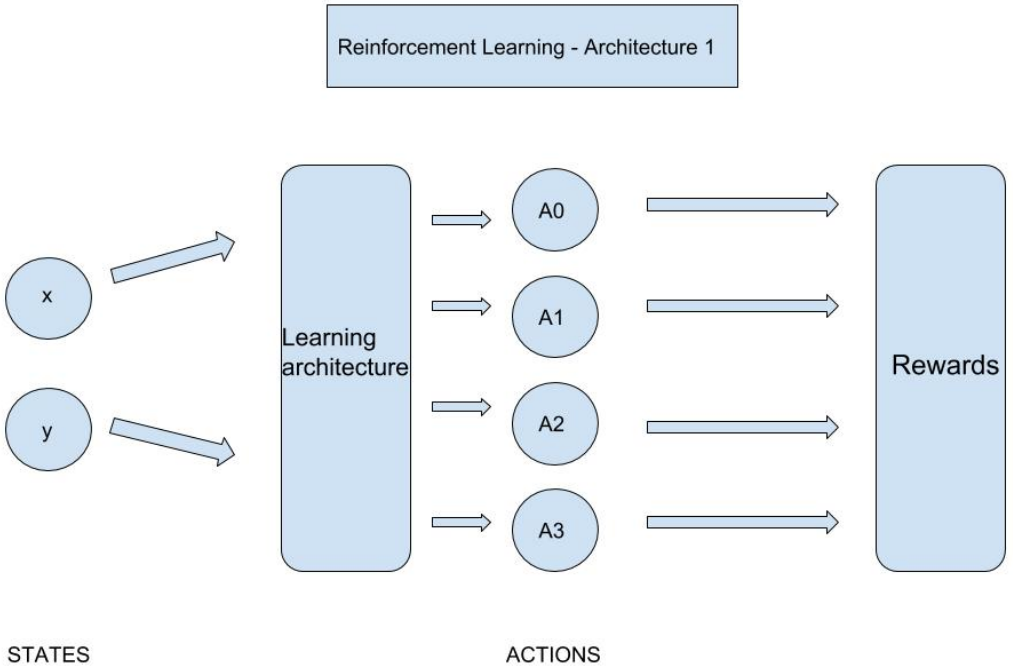
- **Play** - The model is tested by keeping the exploration probability as zero and stopping learning from experiences.

Reinforcement Learning
Deep Q learning



Deep Q-Network Training

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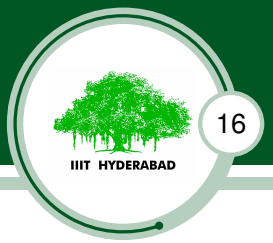
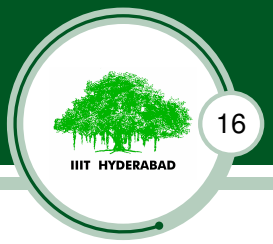
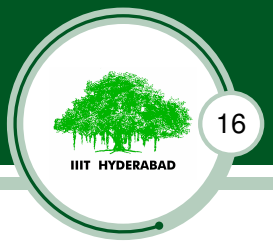


There is a flaw in this approach.



There is a flaw in this approach.

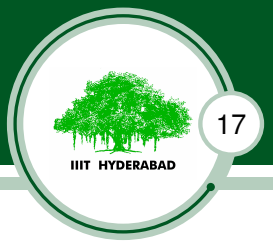
- Our Q values shift but the target value also shifts.



There is a flaw in this approach.

- Our Q values shift but the target value also shifts.
- Due to the max in the formula for setting targets, the network suffers from maximization bias, possibly leading to overestimation of the Q function's value and poor performance.

$$\underbrace{Q(s,a)}_{\text{TD target}} = r(s,a) + \gamma Q(s', \underbrace{argmax_a Q(s',a)}_{\substack{\text{DQN Network choose action for next state} \\ \text{Target network calculates the Q value of taking that action at that state}}})$$



Like most all machine learning algorithms all this comes at a price.

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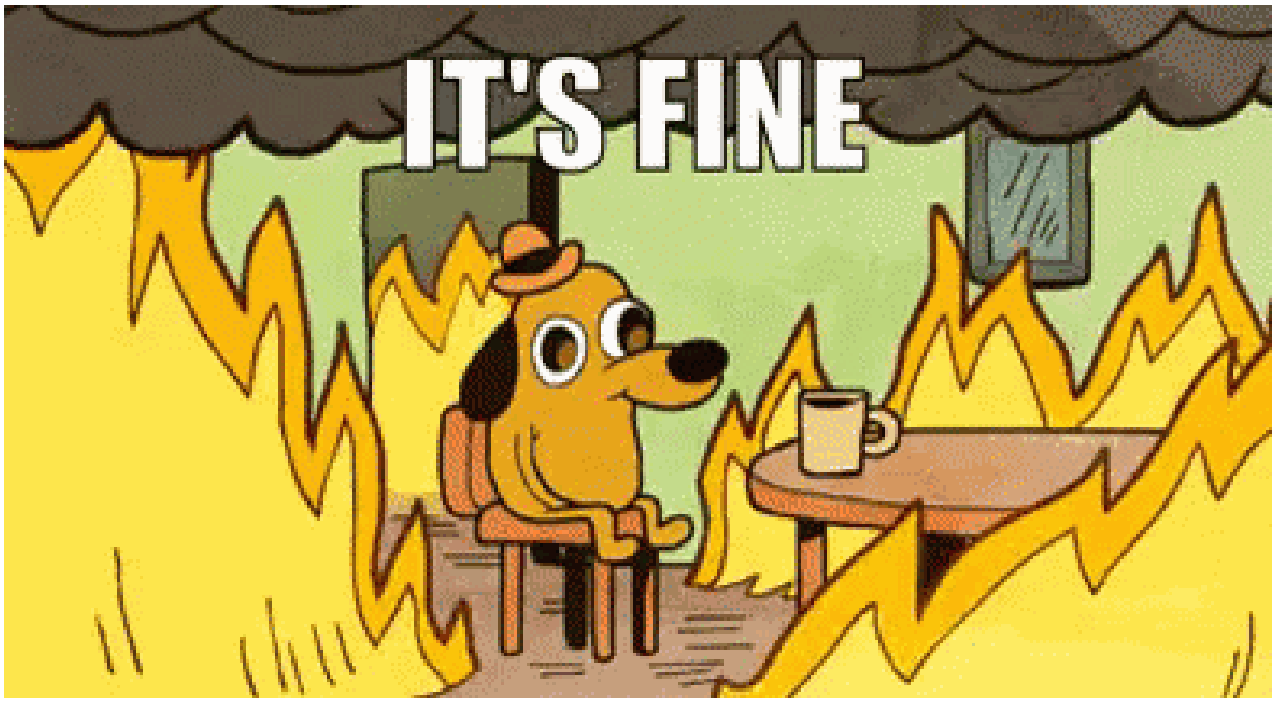

It uses a lot of system resources!

Breakout by deepmind ▸ Link

Examples
openAI gym and fromscratchtoml

openAI gym provides an environment for your agent to learn and play around. Link

Code for machine learning algorithms from scratch - Link


Thank You