## Reinforcement Learning for Recommender Systems

From Contextual Bandits to Slate-Q

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#### Overview of the tutorial

25 min: Intro to RL and why use RL to solve Recommender Systems?

10 min: Break/ Q&A

25 min: RLlib's Contextual Bandits Algorithms Applied To Recommender

Systems

10min: Break/ Q&A

10 min: RLlib's SlateQ Applied To Recommender Systems





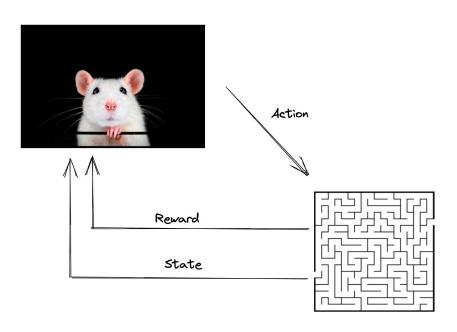
We want to teach a mouse to get to the center of a maze ... how can we do that?

Place bits of cheese in the maze along paths that we **do** want the mouse to take.

Place bits of poison along paths that we **don't** want the mouse to take.

Let the mouse repeatedly explore the maze for a fixed time from different random starting points.

Eventually the mouse learns all paths from everywhere in the maze that maximize cheese and minimize poison.



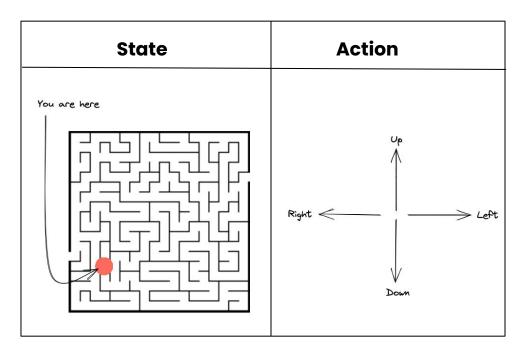


The mouse can be at different positions in the maze. We call this its **state** or observation.

The mouse can move up down left and right in the maze. We call this its **actions.** 

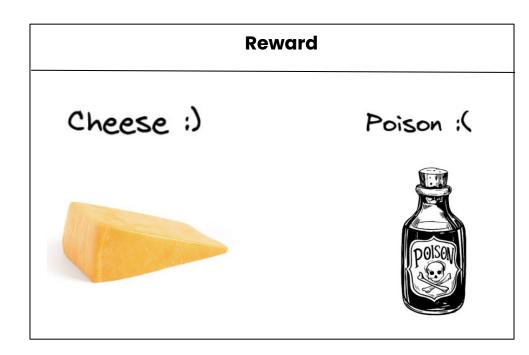
There are terminal states in which the mouse no longer needs to act, such as if the mouse enters the center of the maze.

For every action that mouse takes at its current state to transition into a new state we call this a **time step.** 





The mouse gets cheese or poison depending on whether its taking the best action at its state. We call this **reward.** 



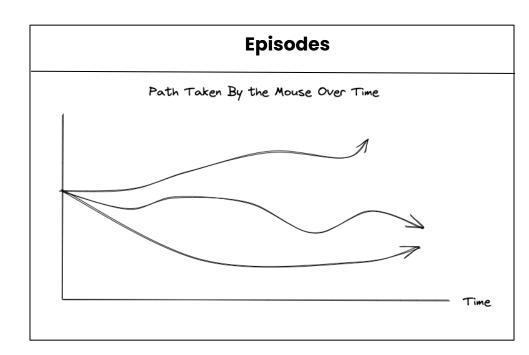


Mouse explores follows the cheese trail for some time.

**Reset** it to a new random state after n time steps.

The sequence of time steps mouse creates from starting state till reset is an **episode**.

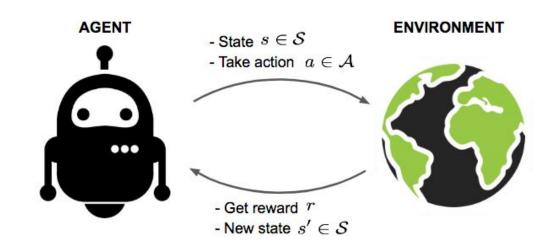
When episode terminates, **done** signal is true.





Use **reinforcement learning** to update the mouse's, or our **agent's**, using the episodes the mouse collects.

- 1. Mouse plans a path to take.
- 2. Mouse acts optimally, collecting episodes.
- 3. We update the mouse using it using collected episodes.
- 4. Go back to step 1.



## Can We Use The Framework Of The Toy Problem And Relate It To A Real World Problem?





#### **RL and Self Driving**

#### **Autonomous Vehicle vision-based controller**

**State:** camera images, lidar sensor readings

Action: steering, brake, gas

Reward: +1 safe driving, 0 for unsafe driving

**Done:** trip ends or if an unsafe driving incident

occurs.

Using RL, train the car-agent to maximize the amount of safe driving behavior during a trip



## Recommender Systems and RL



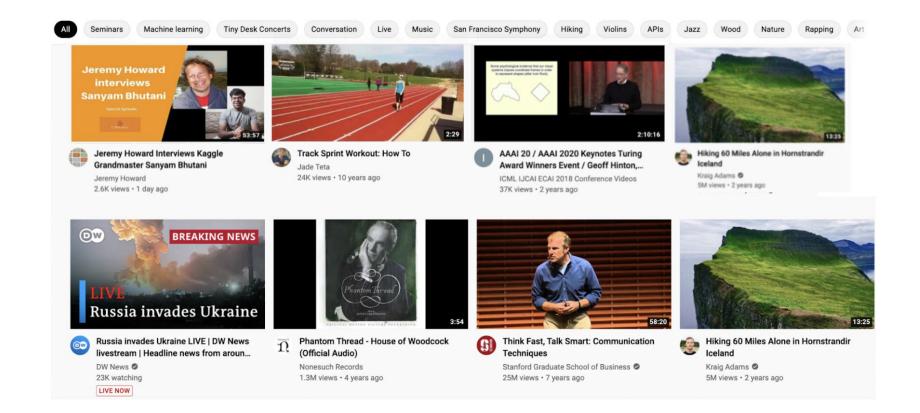


## **RLIP** What Are Recommender Systems?

What	Where
Serve up personalized content, based on users interactions, in order to provide an improved experience for each user.	<ul> <li>Games</li> <li>Rideshare</li> <li>Purchasing apps (B2B, B2C)</li> <li>Websites, web apps</li> <li>Mobile apps</li> <li>Chatbots</li> <li>Call centers</li> </ul>



### **RL and Recommender Systems**





#### RL and Recommender Systems

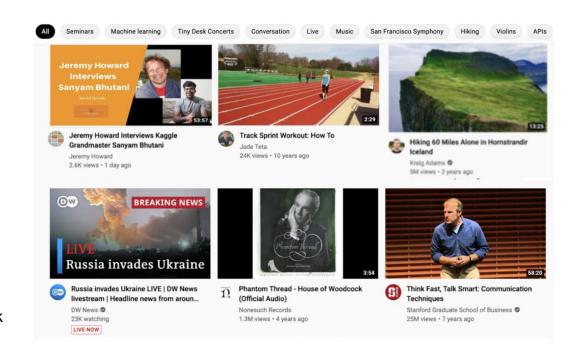
**State:** user actions, features about the user, user watch history

Action: recommended videos

**Reward:** +1 if user clicks a piece of content AND long-term satisfaction has increased at the current time step, 0 otherwise

**Done:** True if we've made the maximum amount of recommendations in a user's session on the website.

Using RL, train a recsys agent to maximize the click through rate and long term satisfaction of users.

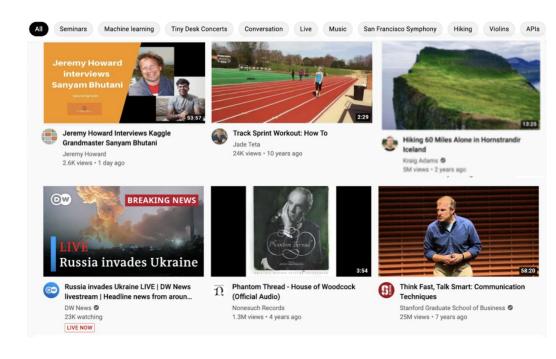




## Supervised Learning (SL) Approach to Recommender Systems

## High-level steps to build a Recommender System using SL

- 1) Train a model using supervised learning that takes as input some features about the user (demographic, interests) and outputs some categories or topics the user would be interested in.
- Use a search ranking algorithm on the categories to produce video recommendations.





## Why use RL over SL for Recommendations?

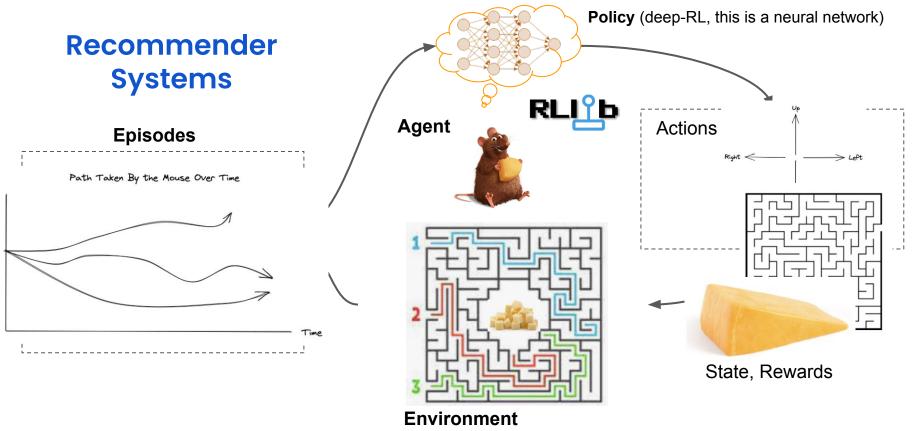
RL Algorithms will train recsys agents that maximize the quality of all the recommendations that are given. SL trains agents to give pointwise recommendations.

With SL there is no concept of long term satisfaction of a user embedded in to training. This must be hardcoded into the recommender system as a heuristic.

RL can be used to train an end to end recommendation system whereas SL will have brittle hand-engineered features in the system.



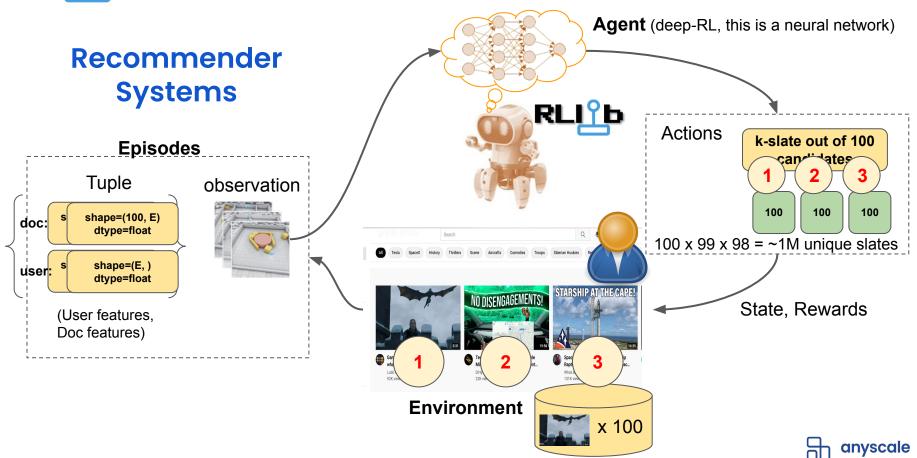
#### Deep RL and RLlib code abstractions





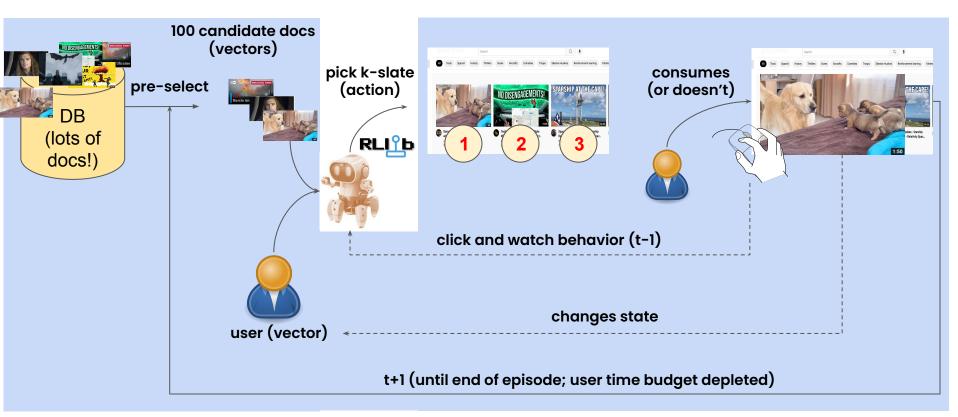


#### RL Ingredients and RLlib code abstractions





#### A Recommender System in Action



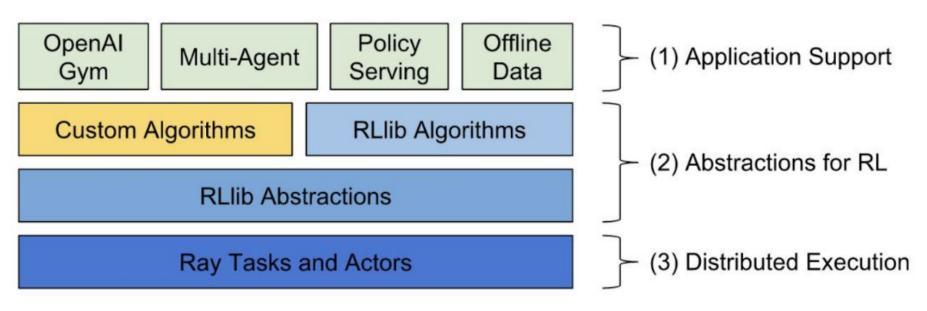
## What Software Should I Use To Do RL?





#### RLlib Parallelize and Distribute at Scale

Ray provides a common framework for distributed RL algorithms using asynchronous parallel Tasks and Actors.





Support for both TF2 and PyTorch





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- Benchmarks and tuned examples. Gives you a leg-up, someplace to start with hyperparameter choices.





#### And why should I use RLlib?

Because these companies here do!
Thx for presenting at Ray- and RL Summits!













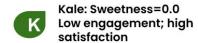






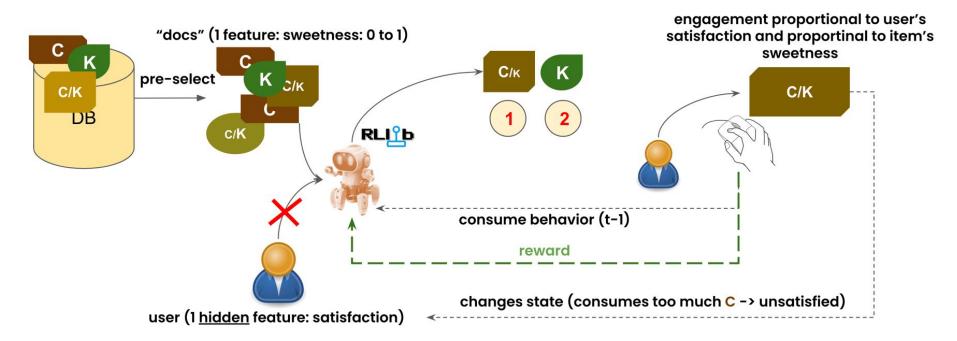






## Google RecSim The "Long Term Satisfaction" Problem

Chocolate: Sweetness=1.0
High engagement; low
satisfaction





# 10 min Break :) Then ... our Colab Notebook

Google Colab: https://tinyurl.com/odsc2022rllibtutorial

Github: bit.ly/odsc\_rllib\_github

