

# Webinar

# UCB Reinforcement Learning

## In R Programming



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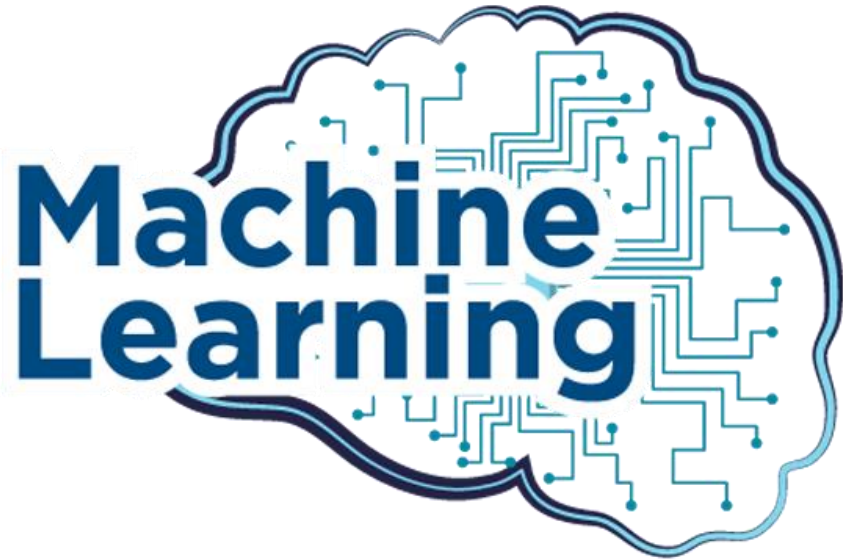
Summer 2022



## What is Machine Learning?



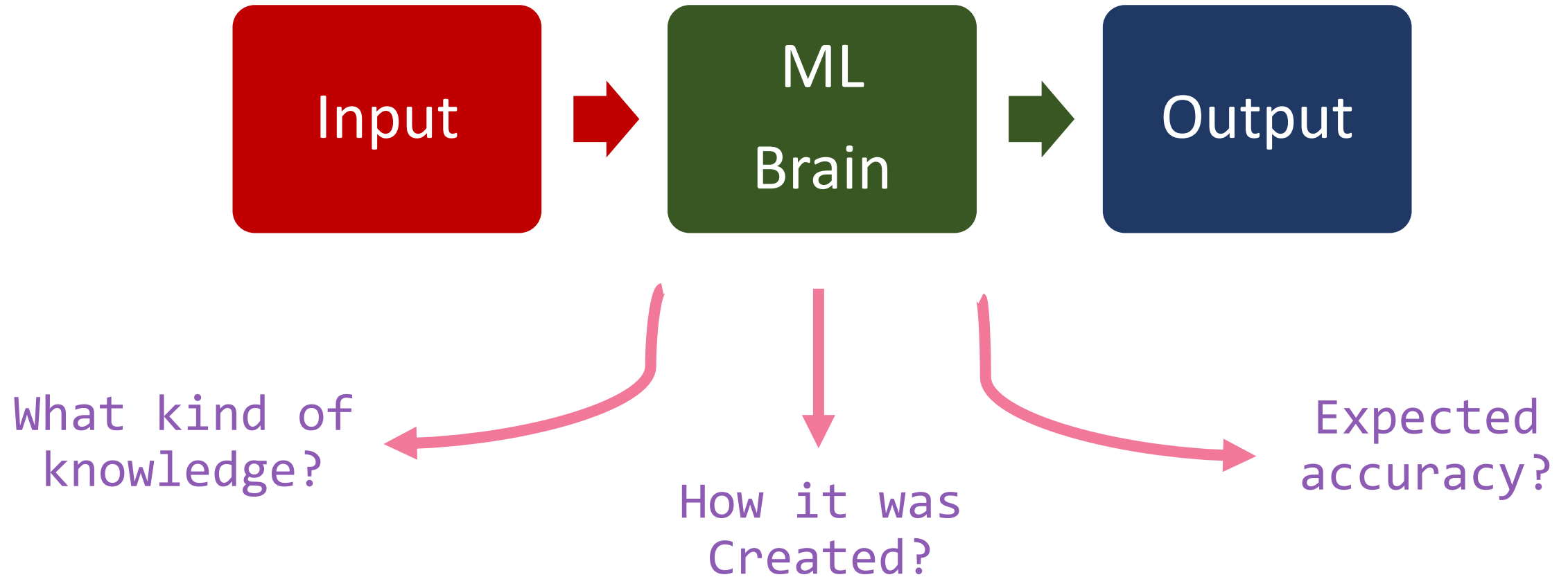
# What is Machine Learning?



Machine learning is a branch of artificial intelligence and computer science which focuses on the use of data and algorithms to imitate the way that humans learn, gradually improving its accuracy.



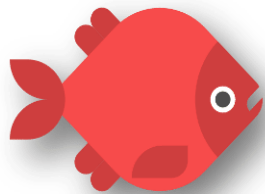
# Introduction to Machine Learning



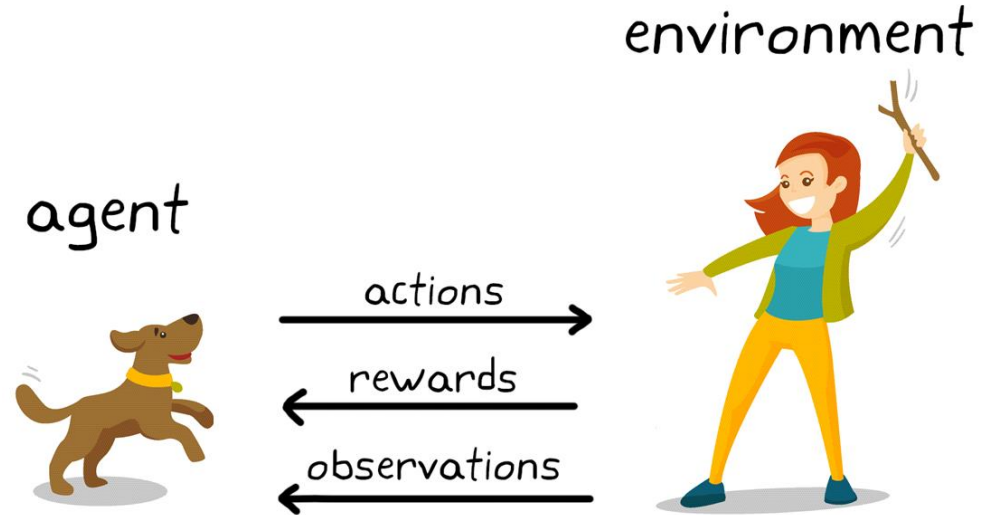
# ML Sections



**Supervised  
Learning**



**Unsupervised  
Learning**



**Reinforcement  
Learning**





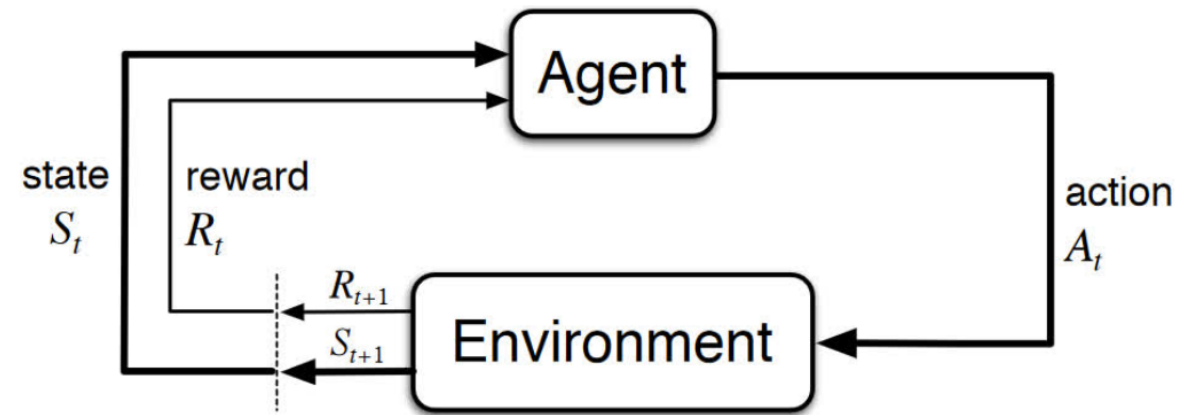
## What is Reinforcement Learning?



# What is Reinforcement Learning?

Reinforcement learning is a simulation-based optimization method which relies on interacting agents to find the optimal (or near-optimal) solution.

RL comprises two main elements, the agent (learner) and the learning environment



Let's break down this diagram into steps.

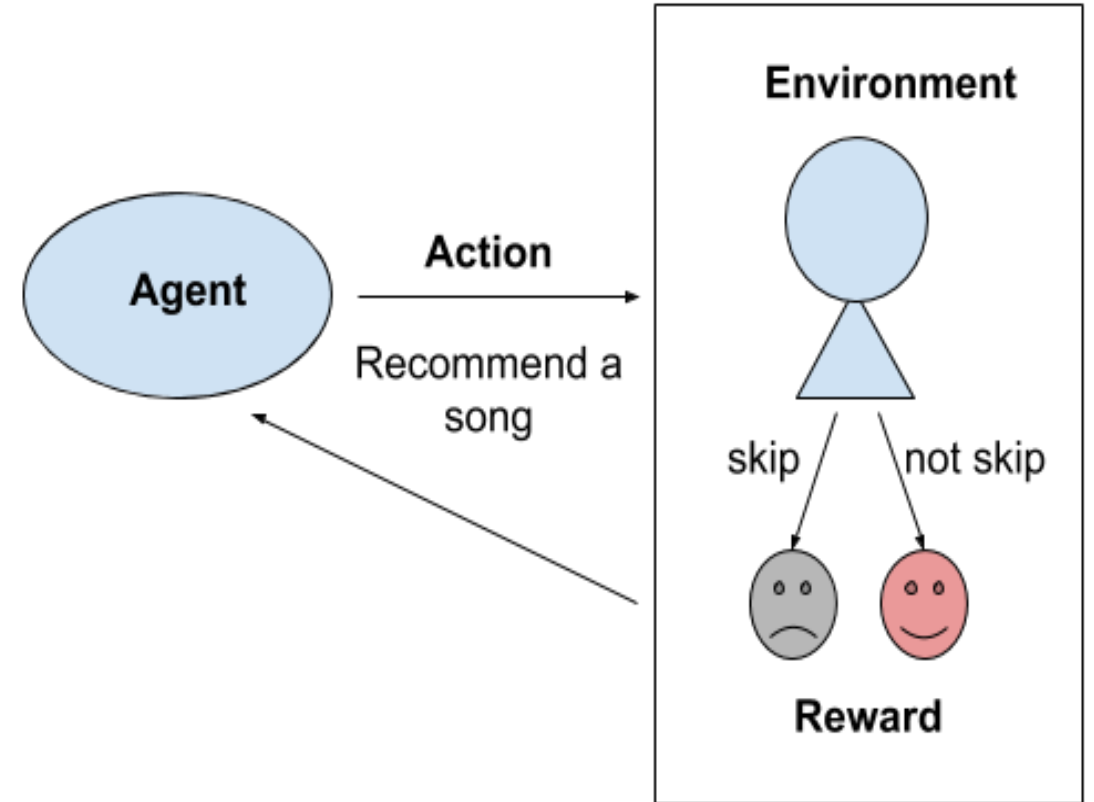
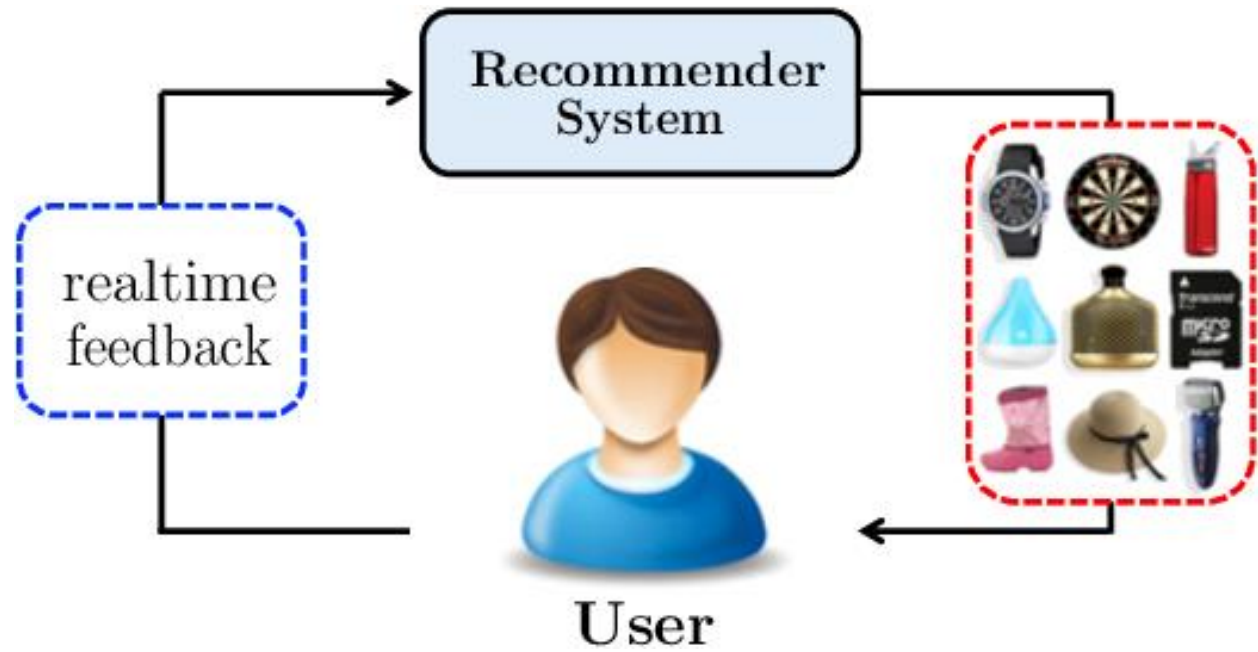
1. At time  $t$ , the environment is in state  $S_t$ .
2. The agent observes the current state and selects action  $A_t$ .
3. The environment transitions to state  $S_{t+1}$  and grants the agent reward  $R_{t+1}$ .
4. This process then starts over for the next time step,  $t + 1$ .
  - Note,  $t + 1$  is no longer in the future, but is now the present. When we cross the dotted on the bottom left, the diagram shows  $t + 1$  transforming into the current time step  $t$  so that  $S_{t+1}$  and  $R_{t+1}$  are now  $S_t$  and  $R_t$ .

Interaction between Game theory and Reinforcement Learning

Interaction between Recommendation Systems and Reinforcement Learning



# What is Reinforcement Learning?





# Elements of RL

1. **Agent** — the learner and the decision maker.
2. **Environment** — where the agent learns and decides what actions to perform.
3. **Action** — a set of actions which the agent can perform.
4. **State** — the state of the agent in the environment.
5. **Reward** — for each action selected by the agent the environment provides a reward. Usually a scalar value.



# Bellman Equation

Source: Martin L. Puterman. Markov Decision Processes: Discrete Stochastic Dynamic Programming. John Wiley & Sons, Inc., New York, NY, USA, 1st edition, 1994. ISBN 0471619779.

## Dynamic Programming Iterations:

$$\begin{aligned} v_{k+1}(s) &\doteq \max_a \mathbb{E}[R_{t+1} + \gamma v_k(S_{t+1}) \mid S_t = s, A_t = a] \\ &= \max_a \sum_{s', r} p(s', r \mid s, a) [r + \gamma v_k(s')] \end{aligned}$$

## Bellman Equation:

$$\begin{aligned} v_*(s) &= \max_a \mathbb{E}[R_{t+1} + \gamma v_*(S_{t+1}) \mid S_t = s, A_t = a] \\ &= \max_a \sum_{s', r} p(s', r \mid s, a) [r + \gamma v_*(s')] \end{aligned}$$

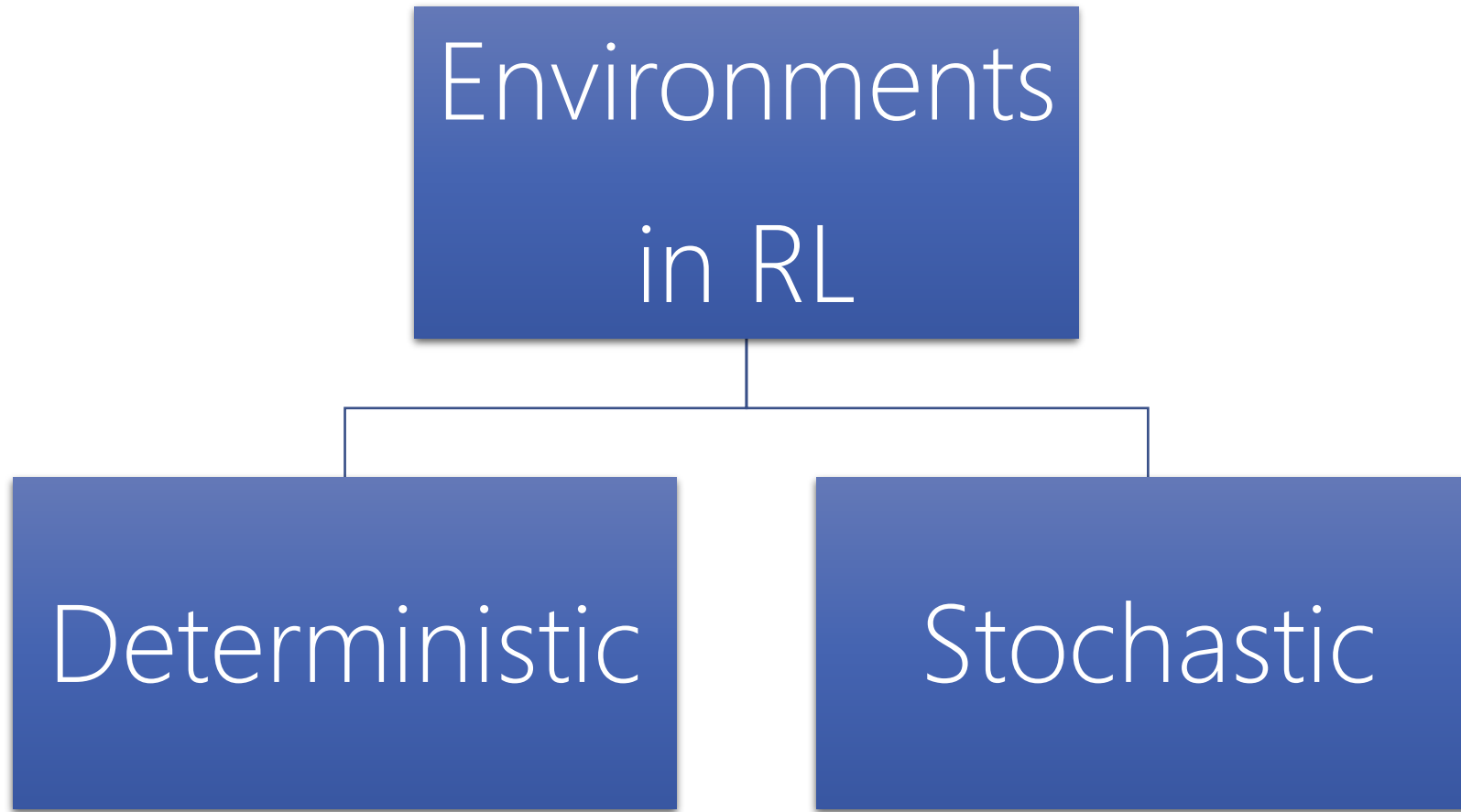




## Exploitation Vs Exploration

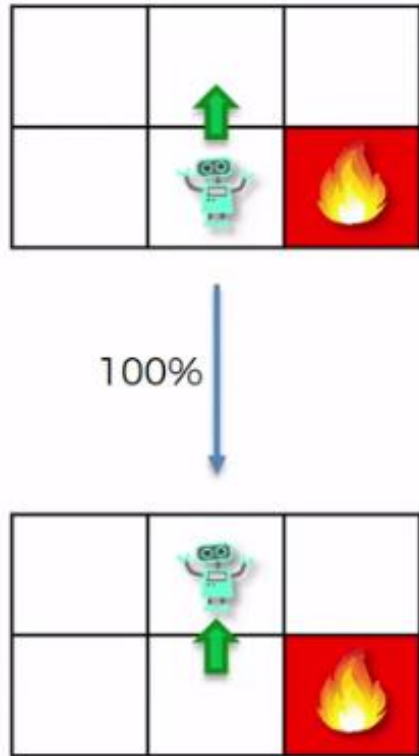


# Environments



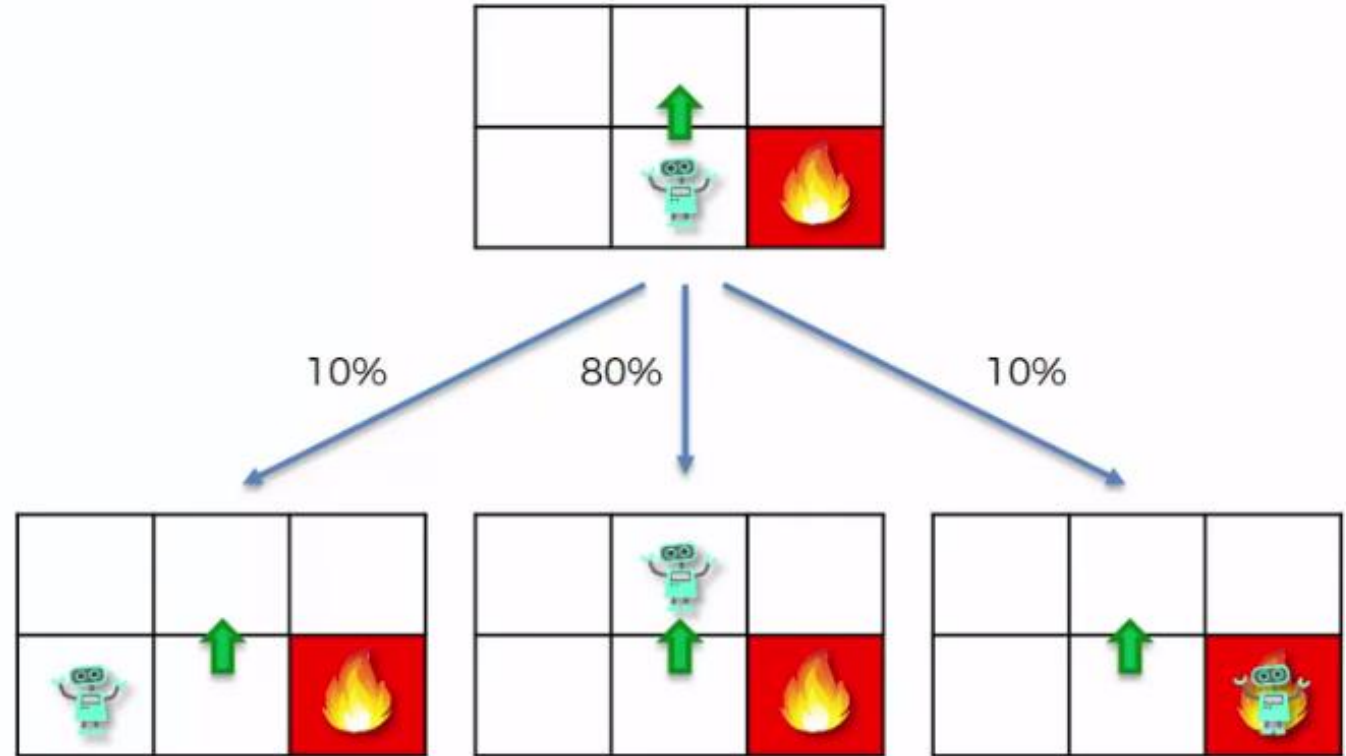
## Exploitation

Deterministic Search



## Exploration

Non-Deterministic Search

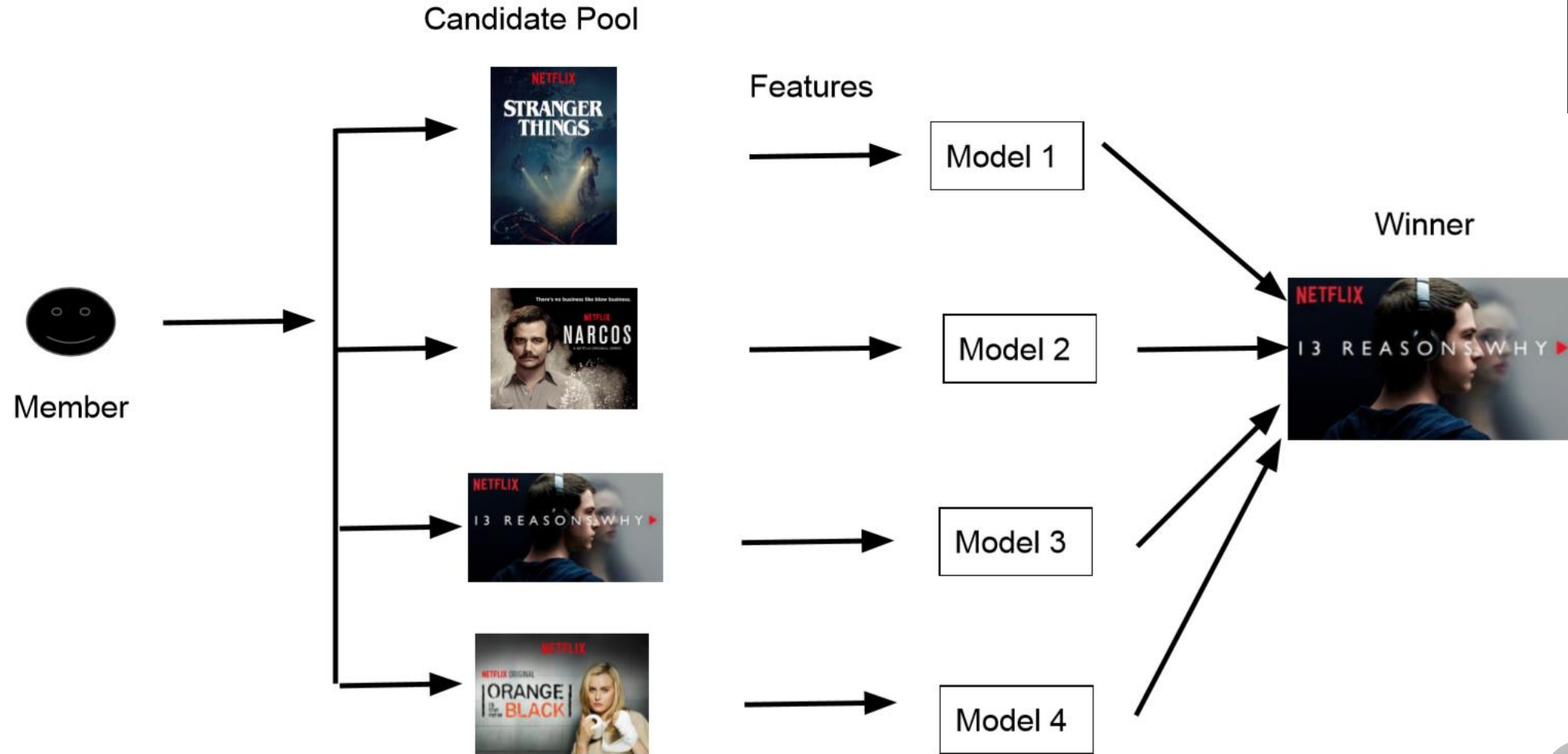




## RL Real Cases



# Case Netflix

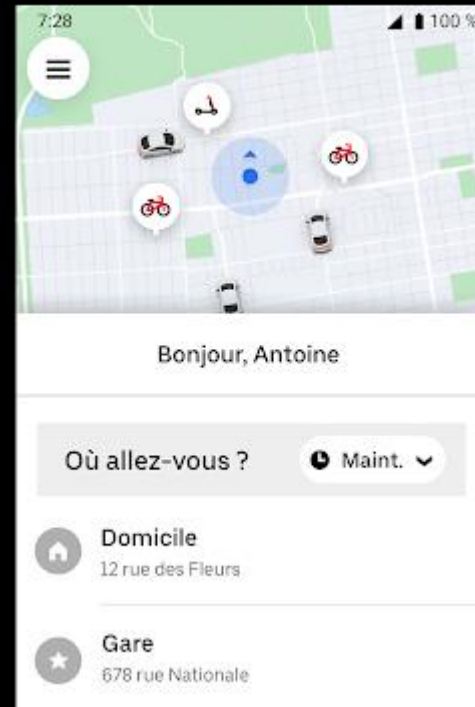
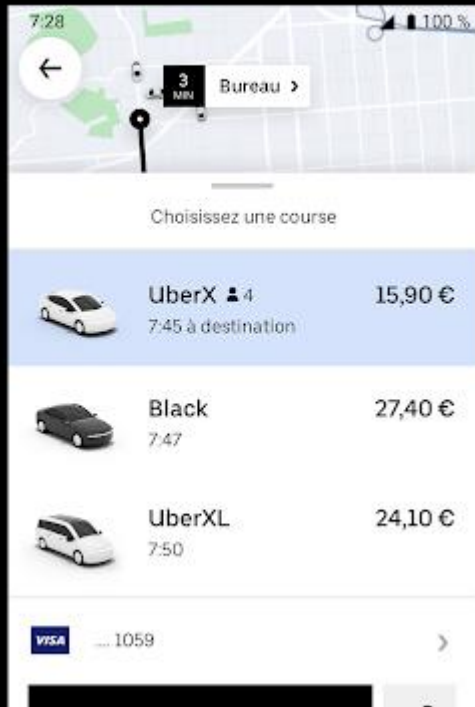


# ML Cases



Optimize Waiting Time and  
Idle Vehicle

Find Best Price



UBER

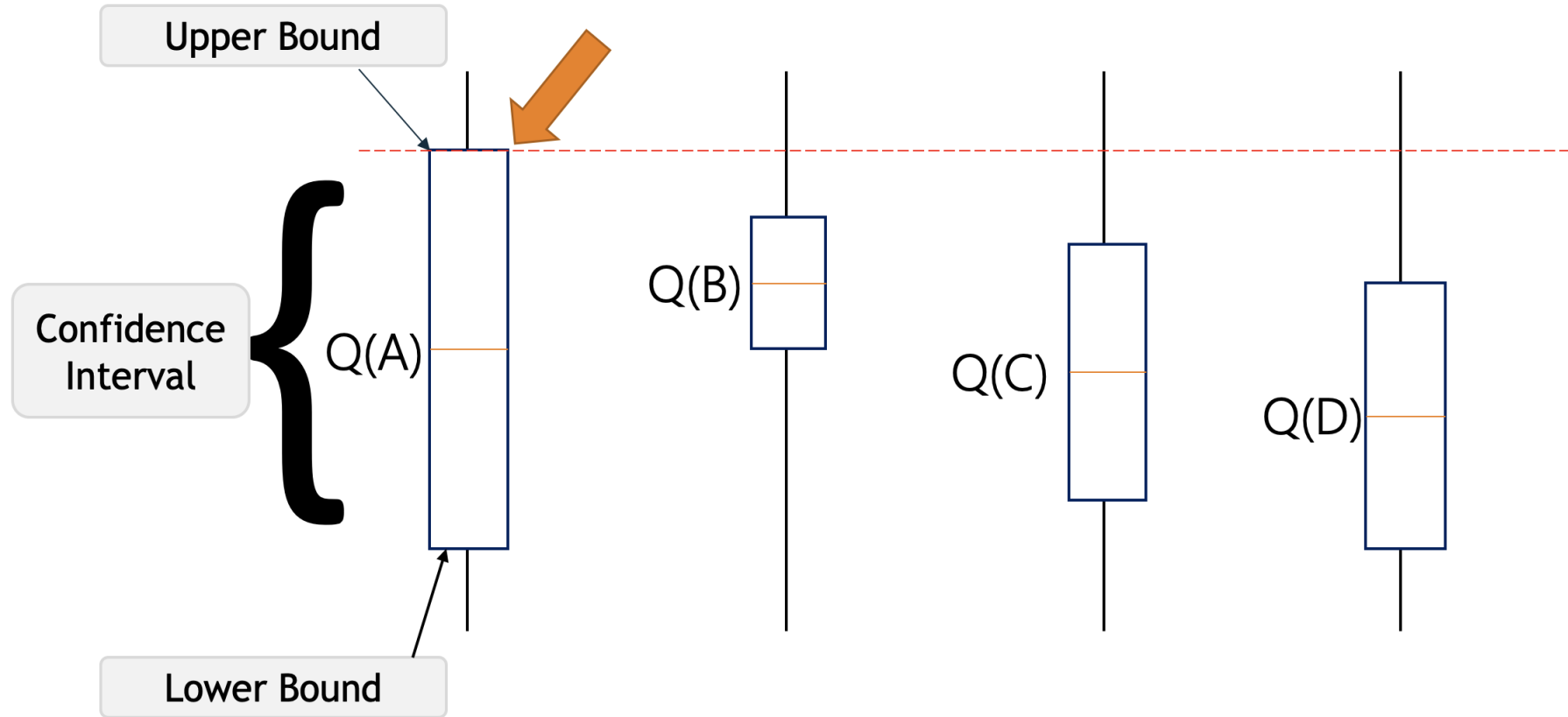




Upper Confidence Bound (UCB)



# Upper Confidence Bound (UCB)



Upper Confidence Bound is a Reinforcement Learner



# Upper Confidence Bound (UCB)

1. *At each round  $n$ , we consider two numbers for machine  $m$ .*
  - >  $N_m(n)$  = number of times the machine  $m$  was selected up to round  $n$ .
  - >  $R_m(n)$  = number of rewards of the machine  $m$  up to round  $n$ .
2. *From these two numbers we have to calculate,*
  - a. The average reward of machine  $m$  up to round  $n$ ,  $r_m(n) = R_m(n) / N_m(n)$ .
  - b. The confidence interval  $[ r_m(n) - \Delta_m(n), r_m(n) + \Delta_m(n) ]$  at round  $n$  with,  $\Delta_m(n) = \text{sqrt}( 1.5 * \log(n) / N_m(n) )$
3. *We select the machine  $m$  that has the maximum UCB,  $( r_m(n) + \Delta_m(n) )$*

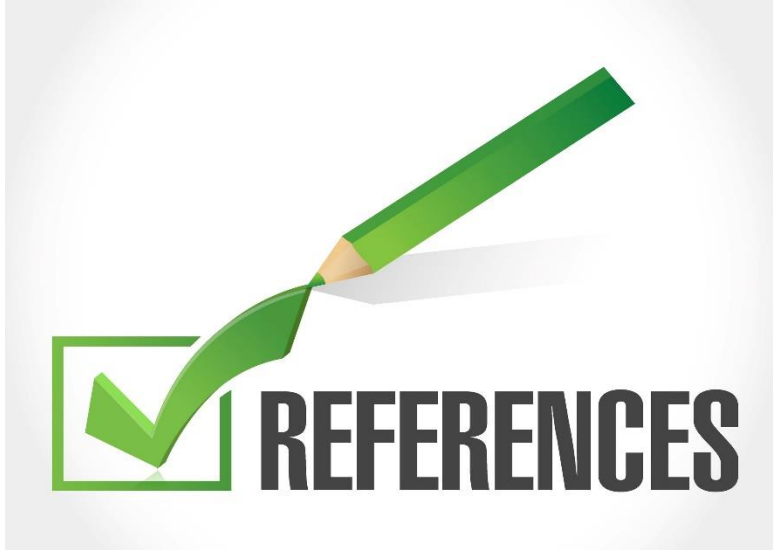


# References

Martin L. Puterman. Markov Decision Processes: Discrete Stochastic Dynamic Programming. John Wiley & Sons, Inc., New York, NY, USA, 1st edition, 1994. ISBN 0471619779.

Decision Theory: Single & Sequential Decisions. VE for Decision Networks .Alan Mackworth, UBC CS 322 – Decision Theory, 2013.

Sutton R. Barto, A. Reinforcement Learning: An Introduction. The MIT Press., Cambridge, Massachusetts, UK, 2nd edition, 2018.



# Conclusion and Questions?

A photograph of ancient stone ruins, possibly a temple or palace, silhouetted against a dark, twilight sky. The ruins feature tall columns and a long wall with intricate carvings. The scene is dimly lit, with a warm glow from the setting or rising sun on the horizon.

**THANK YOU!**