

Reinforcement Learning

Zahoor Tanoli (PhD)
CUI Attock

Algorithms

Supervised Learning

Linear Regression

Logistic Regression

Support Vector Machines

K Nearest Neighbors

Decision Tree

Unsupervised Learning

K Means Clustering

Hierarchical Clustering

DBSCAN

Principal Component Analysis

Reinforcement Learning

Q-Learning

SARSA

Monte Carlo

Deep Q Network

Approach

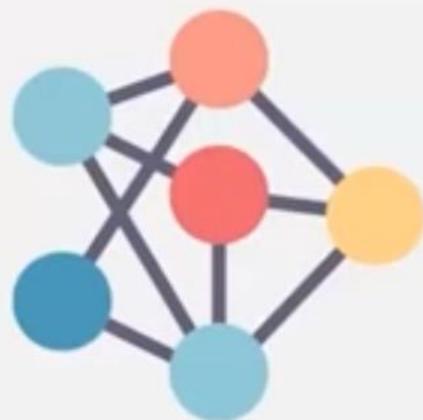
Supervised Learning

Takes labeled inputs and maps it to the known outputs



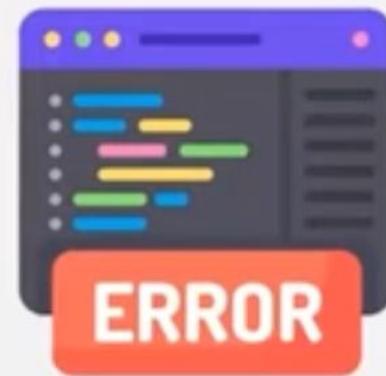
Unsupervised Learning

Understands patterns and trends in the data and discovers the output



Reinforcement Learning

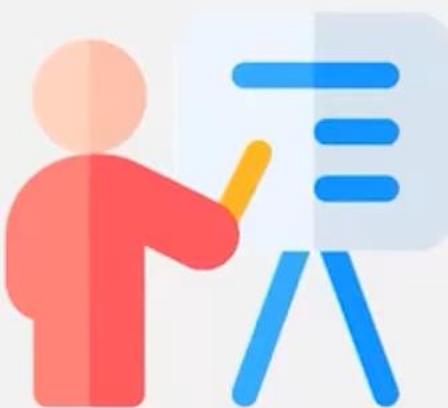
Follows trial and error method to arrive at the desired solution



Training

Supervised Learning

Supervised Learning techniques need external supervision to train models



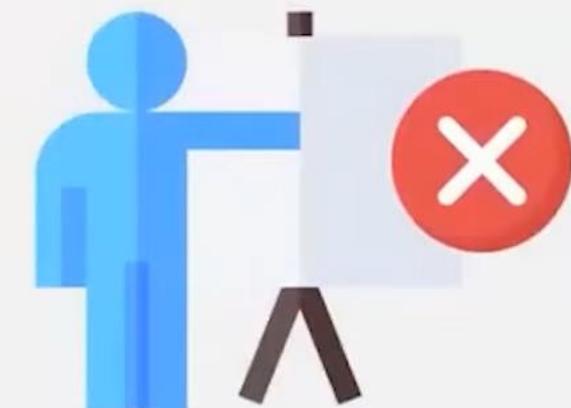
Unsupervised Learning

Unsupervised Learning techniques do not need any supervision to train models



Reinforcement Learning

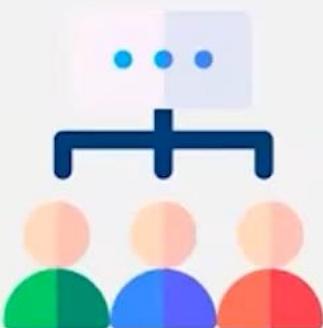
Reinforcement Learning techniques do not need any supervision to train models



Type of Problems

Supervised Learning

Classification and Regression



Unsupervised Learning

Clustering and Association



Reinforcement Learning

Reward based



Applications

Supervised Learning



Weather Prediction



Sales Forecast

Stock Price Analysis

Unsupervised Learning



Customer Segmentation



Churn Analysis

Reinforcement Learning



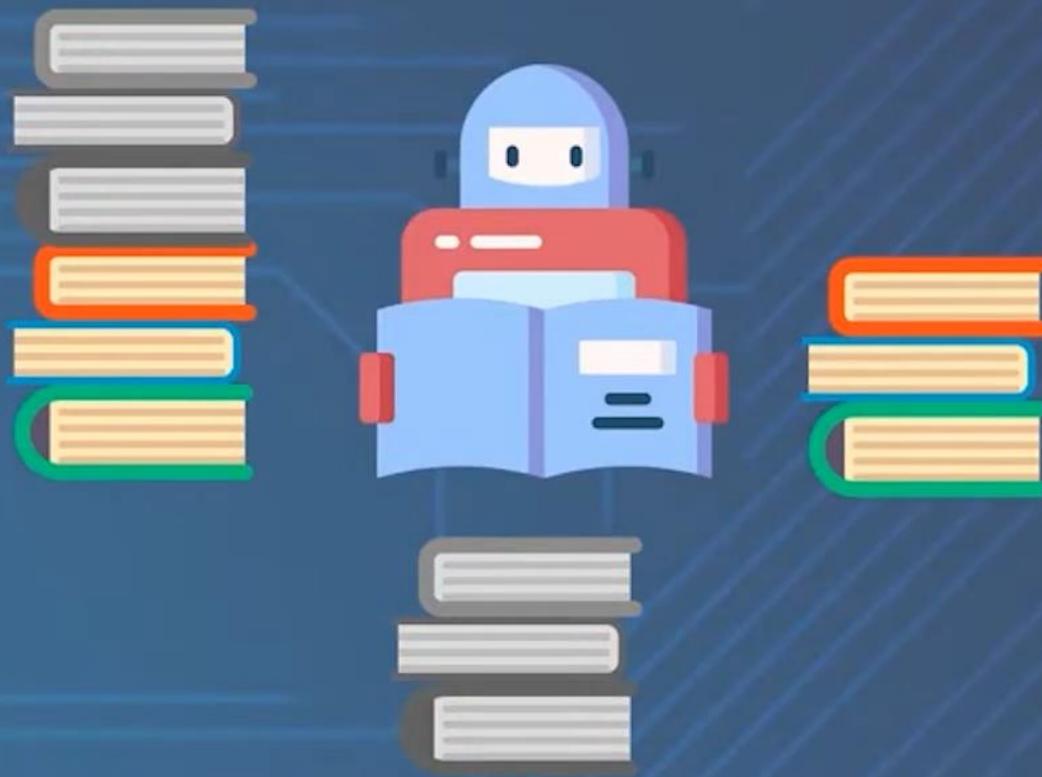
Building Games



Training Robots

Why Reinforcement Learning?

Training a Machine Learning model requires a lot of data which might not always be available to us. Further, the data provided might not be reliable



Why Reinforcement Learning?

Learning from a small subset of actions will not help expand the vast realm of solutions that may work for a particular problem



Learn: To Walk

What is Reinforcement Learning?

Reinforcement learning is a sub-branch of Machine Learning that trains a model to return an optimum solution for a problem by taking a sequence of decisions by itself. Consider a robot learning to go from one place to another



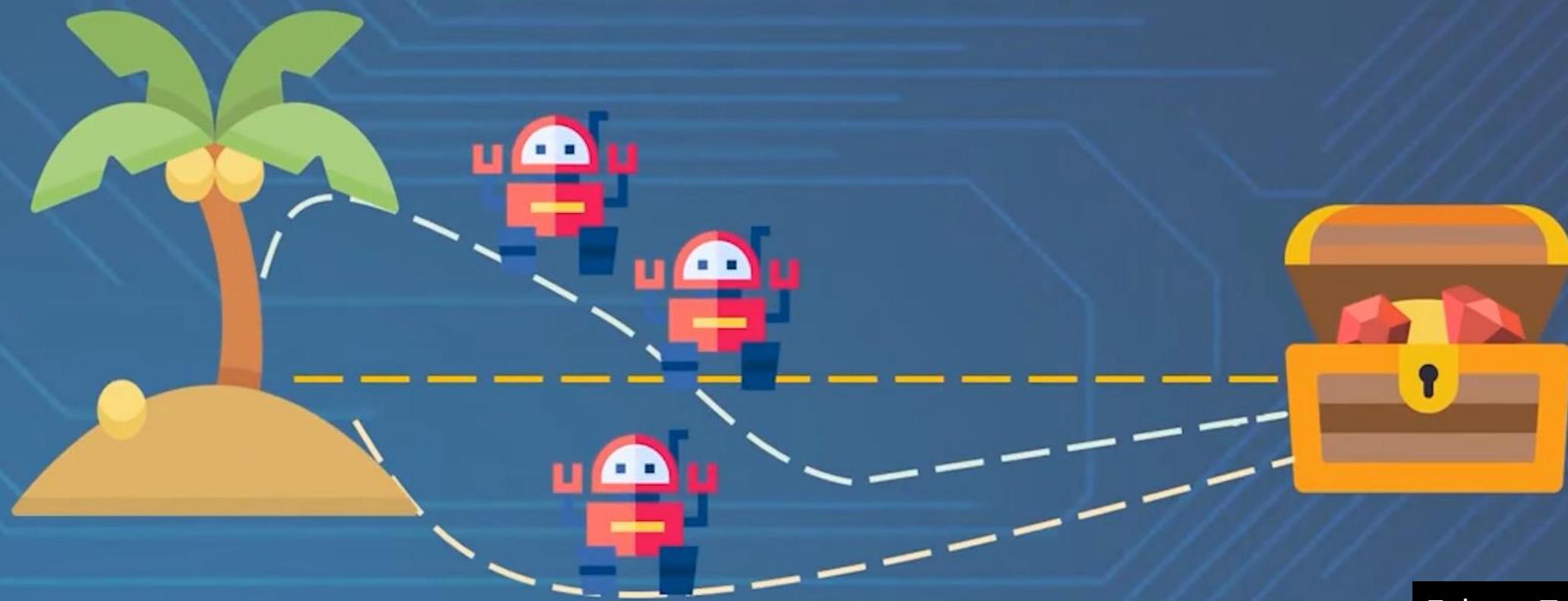
What is Reinforcement Learning?

The robot is given a scenario and must arrive at a solution by itself. The robot can take different paths to reach the destination



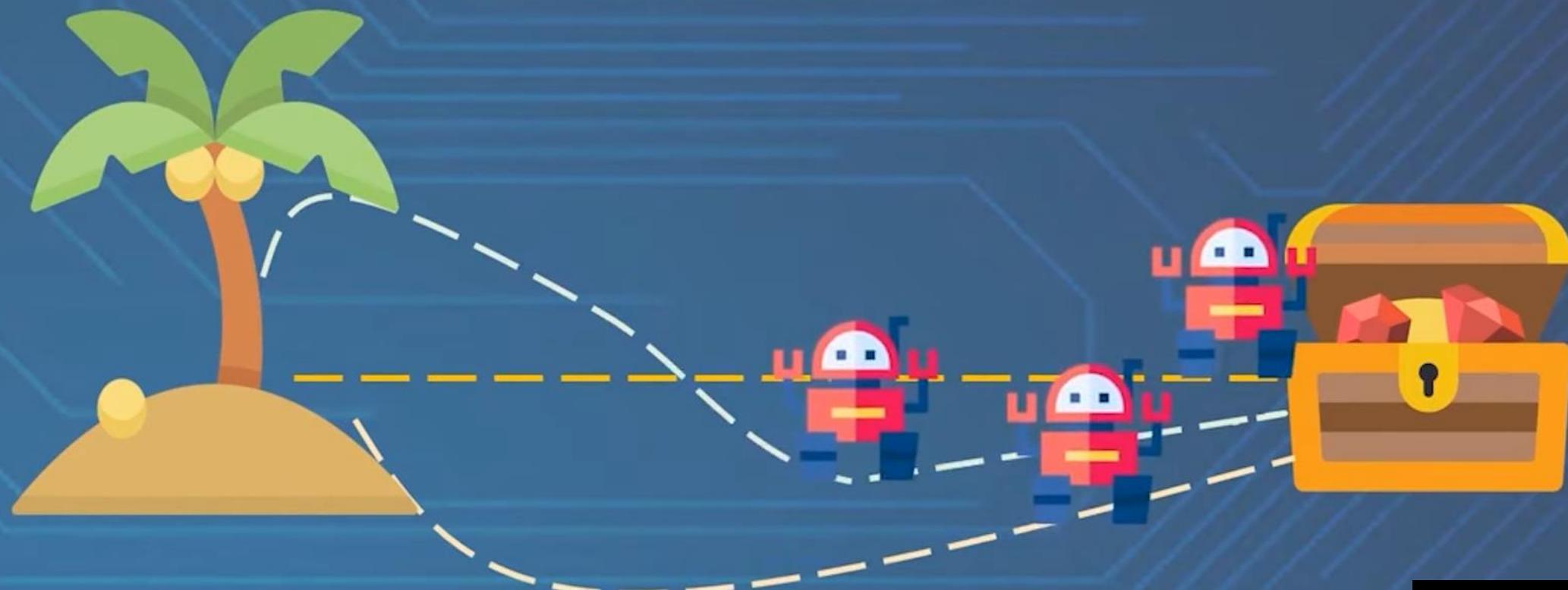
What is Reinforcement Learning?

It will know the best path by the time taken on each path. It might even come up a unique solution all by itself



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It will know the best path by the time taken on each path. It might even come up a unique solution all by itself



Supervised vs Unsupervised vs Reinforcement Learning

Supervised Learning	Unsupervised Learning	Reinforcement Learning
Data provided is labeled data, with output values specified	Data provided is unlabeled data, the outputs are not specified, machine makes its own prediction	The machine learns from its environment using rewards and errors
Used to solve Regression and classification problems	Used to solve Association and clustering problems	Used to solve Reward based problems
Labeled data is used	Unlabeled data is used	No predefined data is used
External Supervision	No supervision	No supervision
Solves problems by mapping labeled input to known output	Solves problems by understanding patterns and discovering output	Follows Trail and Error problem solving approach

Important Terms in Reinforcement Learning

Agent

Agent is the model that
is being trained via
reinforcement learning



Important Terms in Reinforcement Learning

Environment

The training situation that the model must optimize to is called its environment



Output

Important Terms in Reinforcement Learning

Action

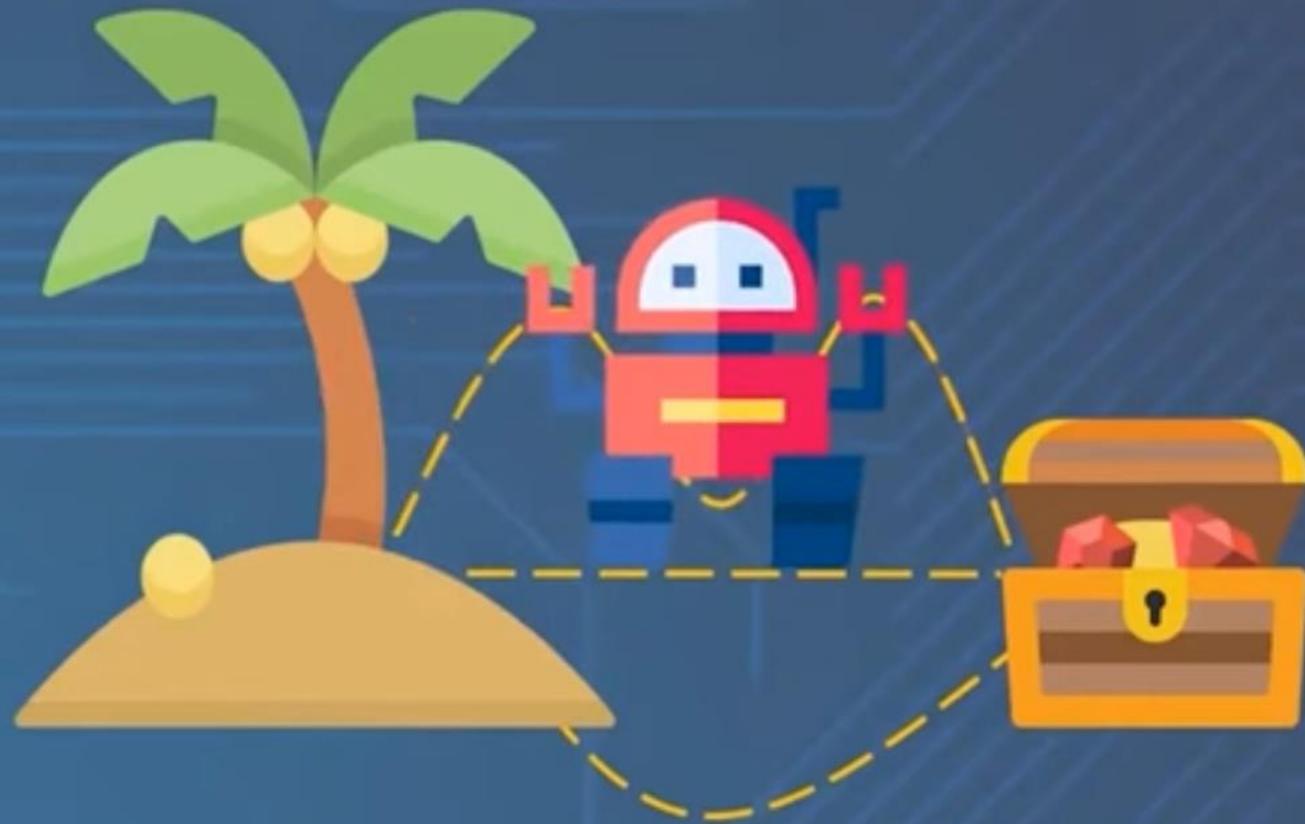
All possible steps
that can be taken by
the model



Important Terms in Reinforcement Learning

State

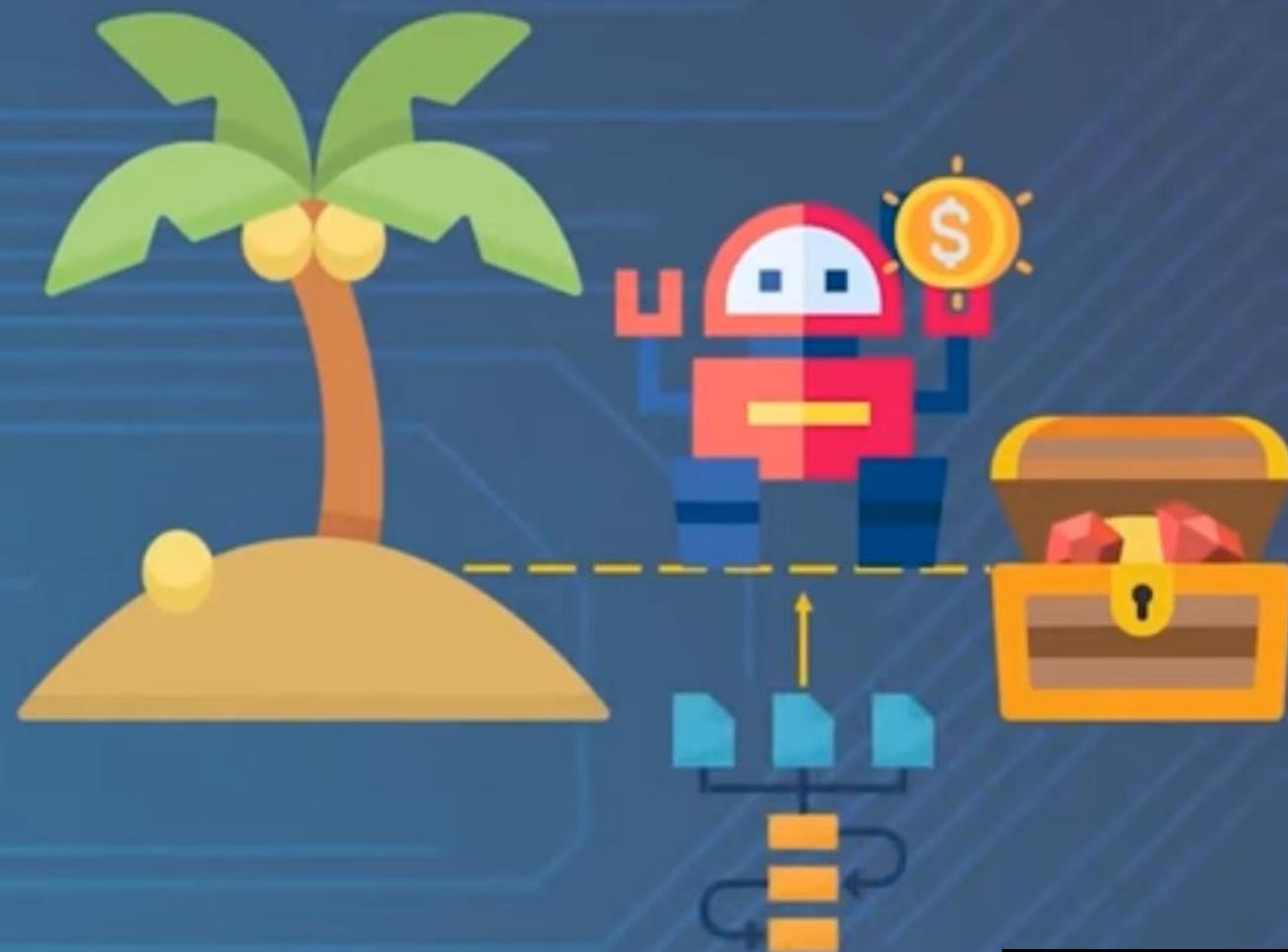
The current position/
condition returned
by the model



Important Terms in Reinforcement Learning

Policy

Policy determines how an agent will behave at anytime. It acts as a mapping between Action and present State



Reinforcement Learning Example

We can get the dog to perform various actions by offering incentives such as dog biscuits as reward



Reinforcement Learning Example

The dog will follow a policy to maximize its reward and hence, will follow every command and might even learn a new action, like begging, by itself



Fetching



Handshake



Begging

Reinforcement Learning Example

The dog will also want to run around and play and explore its environment. This quality of a model is called Exploration



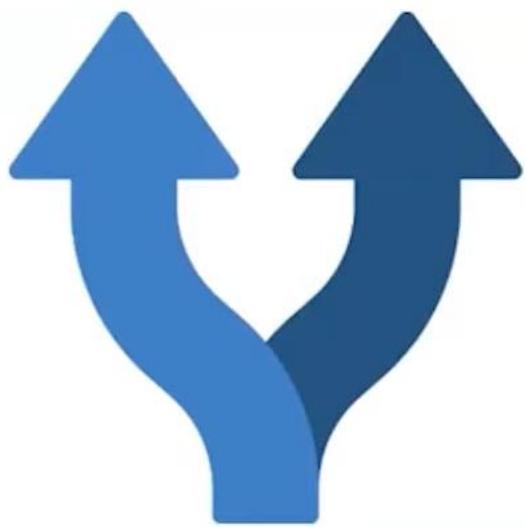
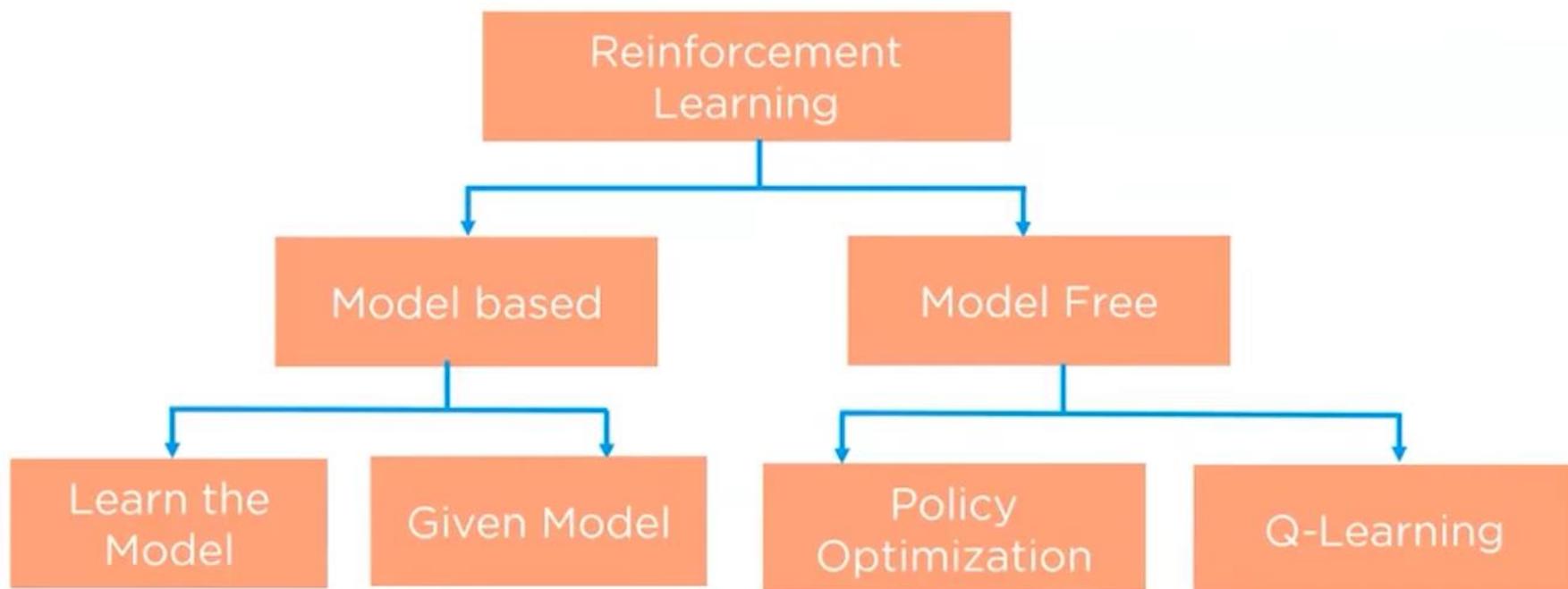
Reinforcement Learning Example

The tendency of the dog to maximize rewards is called Exploitation. There is always a tradeoff between exploration and exploitation as exploration actions may lead to lesser rewards



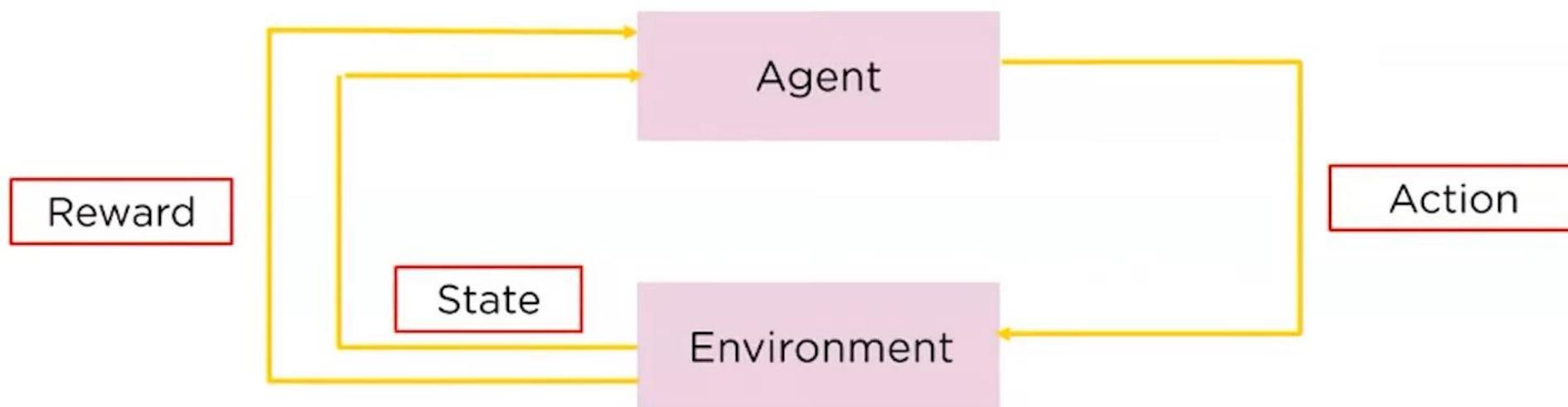
Introduction to Reinforcement Learning

Reinforcement learning can be divided based on whether the machine uses a model to learn or learns by itself



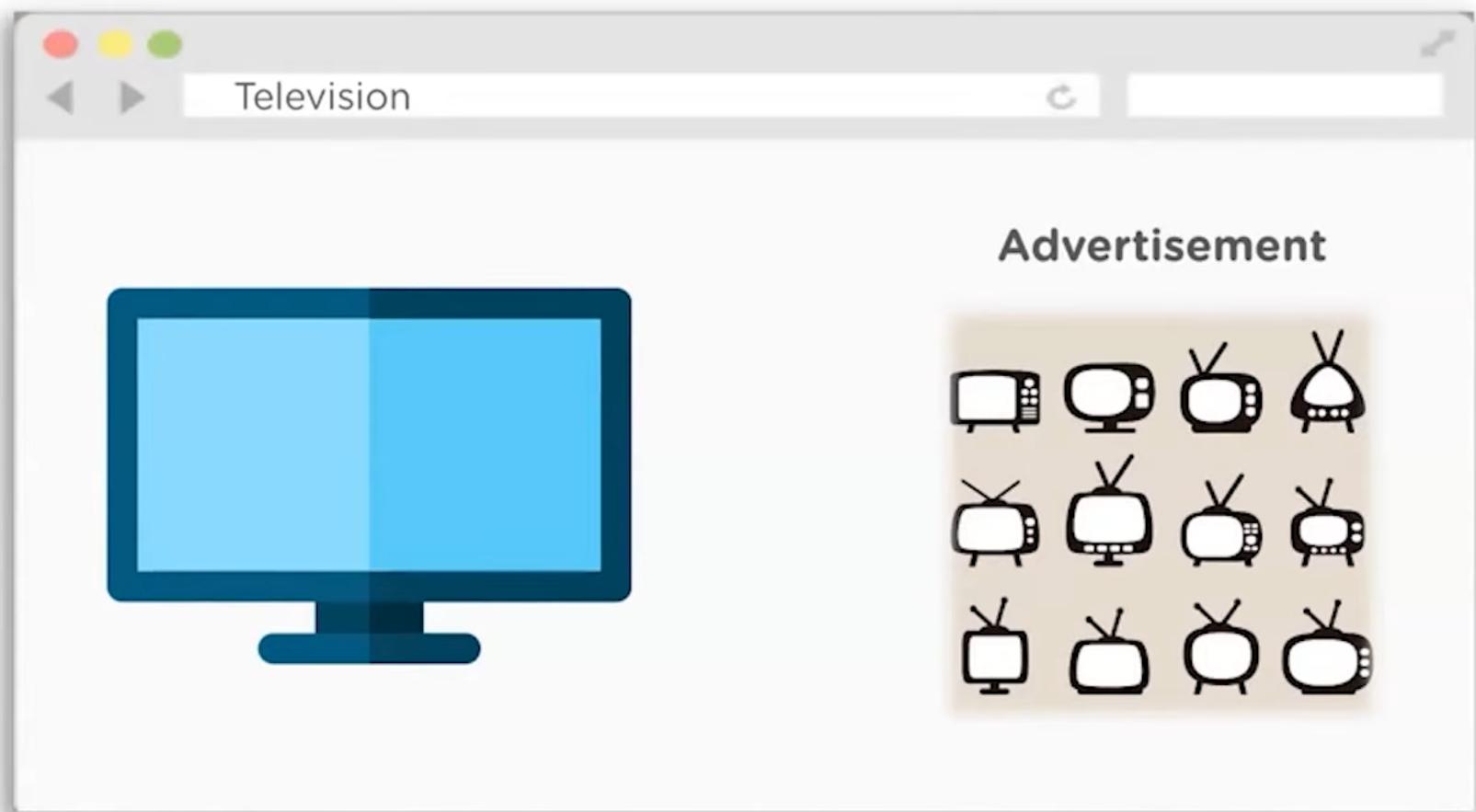
What is Q-Learning?

Q-Learning is a Reinforcement learning policy which will find the next best action, given a current state. It chooses this action at random and aims to maximize the reward



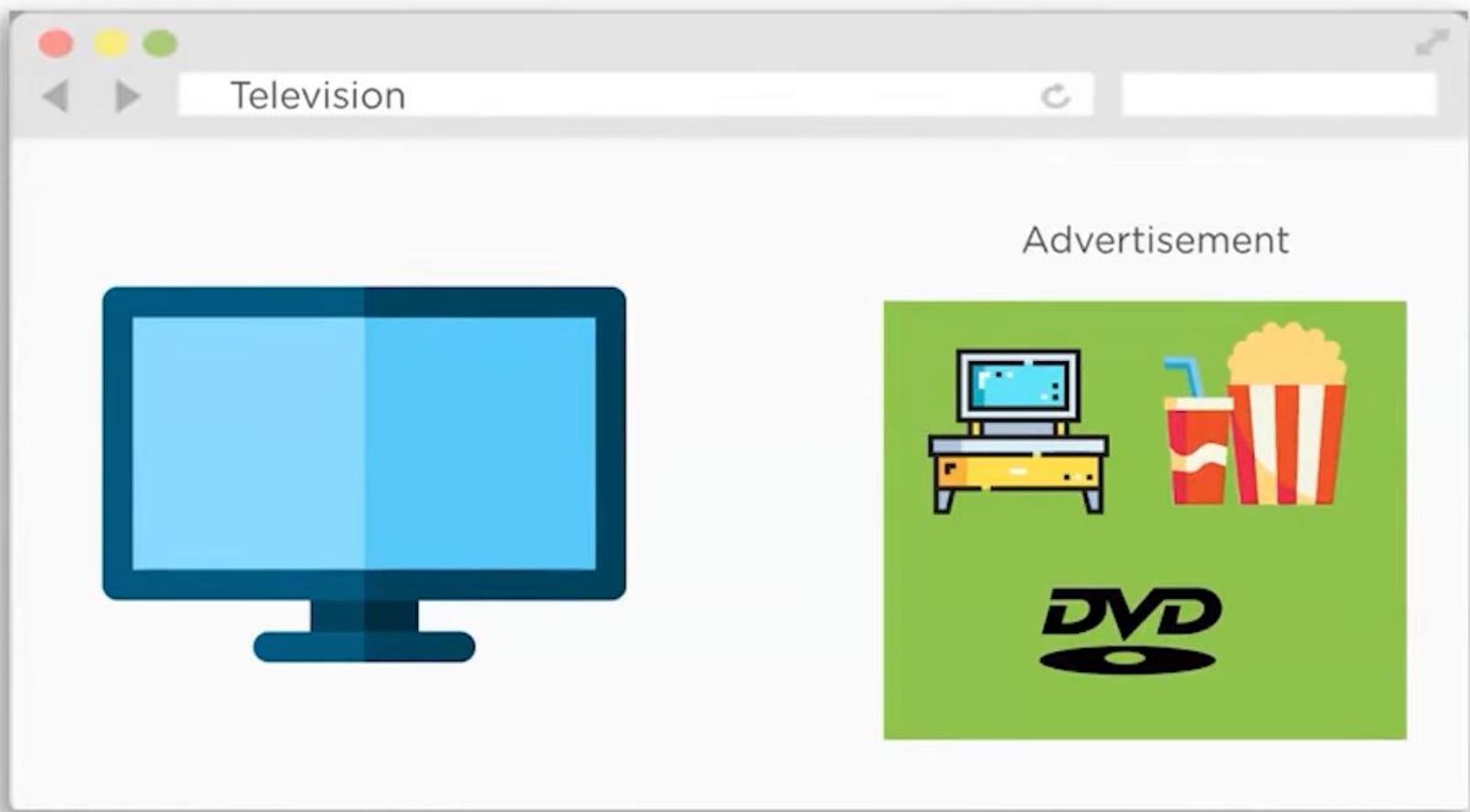
What is Q-Learning? : Ad suggestion

Consider an Ad recommendation system. Usually, when you look up a product online, you get ads which will suggest the same product over and over again



What is Q-Learning? : Ad suggestion

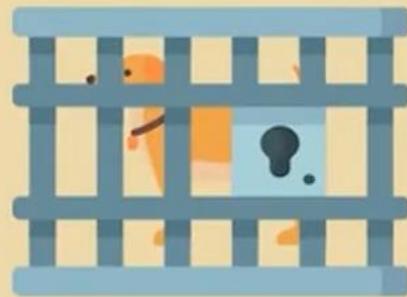
Using Q-learning, we can make an Ad recommendation system which will suggest related products to our previous purchase. The reward will be if user clicks on the suggested product



Some Important Terms

Episodes

When an agent ends up in a terminating state and can't take a new action



Q-Values

Used to determine how good an Action, A, taken at a particular state, S, is $Q(A,S)$



Temporal Difference

A formula used to find the Q-Value by using the value of current state and action and previous state and action

Bellman's Equation

The Bellman Equation is used to determine the value of a particular state and deduce how good it is to be in/ take that state.

The optimal state will give us the highest optimal value

$$\text{New } Q(S, A) = Q(S, A) + \alpha [R(S, A) + \gamma \text{Max } Q'(S', A') - Q(S, A)]$$

Diagram illustrating the components of the Bellman Equation:

- Current Q Value: $Q(S, A)$ (orange box)
- Learning Rate: α (blue box)
- Reward: $R(S, A)$ (orange box)
- Discount Rate: γ (blue box)
- Maximum Expected Future Reward: $\text{Max } Q'(S', A')$ (orange box)
- Current Q Value: $Q(S, A)$ (orange box)
- Discount Rate: γ (blue box)
- Maximum Expected Future Reward: $\text{Max } Q'(S', A')$ (orange box)

Bellman's Equation

Factors Influencing Q-values :

- Current State and Action (S, A)
- Previous State and Action (S', A')
- Reward for Action, R
- Maximum expected future reward

$$\text{New } Q(S, A) = Q(S, A) + \alpha [R(S, A) + \gamma \text{Max } Q'(S', A') - Q(S, A)]$$

↓ ↓ ↓
Current Q Value Learning Rate Reward

↑ ↑
Discount Rate Maximum Expected Future Reward

Steps in Q-Learning

Step 1

Create an initial Q-Table with all values initialized to 0

Action	Fetching	Sitting	Running
Start	0	0	0
Idle	0	0	0
Wrong Action	0	0	0
Correct Action	0	0	0
End	0	0	0

Steps in Q-Learning

Step 2

Choose an action and perform it. Update values in the table

Action	Fetching	Sitting	Running
Start	0	1	0
Idle	0	0	0
Wrong Action	0	0	0
Correct Action	0	0	0
End	0	0	0

Steps in Q-Learning

Step 3

Get the value of the reward and calculate the value Q-Value using Bellman Equation

Action	Fetching	Sitting	Running
Start	0	1	0
Idle	0	0	0
Wrong Action	0	0	0
Correct Action	0	34	0
End	0	0	0

Steps in Q-Learning

Step 4

Continue the same until the table is filled or an episode ends

Action	Fetching	Sitting	Running
Start	5	7	10
Idle	2	5	3
Wrong Action	2	6	1
Correct Action	54	34	17
End	3	1	4

Steps in Q-Learning

The below table gives us an idea of how many times an action has been taken and how positively(correct action) or negatively(wrong action) it is going to affect the next state

Action	Fetching	Sitting	Running
Start	5	7	10
Idle	2	5	3
Wrong Action	2	6	1
Correct Action	54	34	17
End	3	1	4