

# 20 days of # 100 days of learning.

Reinforcement Learning  $\Rightarrow$

"If you ask the wrong question, you will never get the right answer."



[Supervised]

Given X, predict Y

(Unsupervised)

Given X, Simplify Y

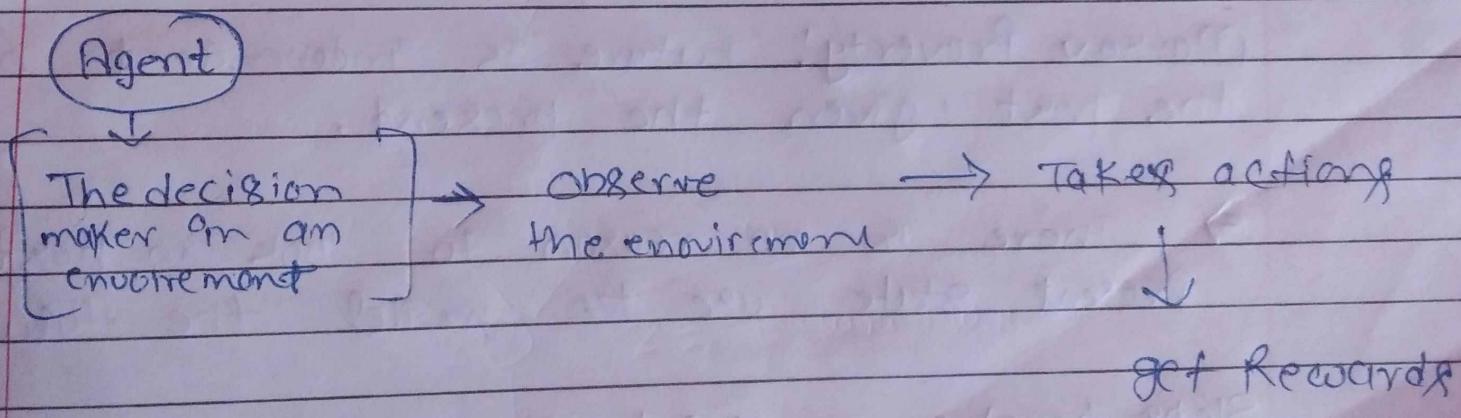


Neither supervised nor unsupervised learning will work in an unknown environment.

Definition:

Train decision makers to take actions to maximize rewards in uncertain environment.

e.g.: How long self-drive car go without accident



## Reinforcement Learning

Objective: choose "best" actions  
Environment is uncertain

Training involves exploring the environment

Training process involved determining the best policy

Explicit dependency of rewards on previous action

## Supervised / Unsupervised Learning

Objective: Predict, Classify or Simplify  
Environment is Known

Training involved finding pattern in data.

Training process involved fitting the "best model"

Individual point are independent of each other

## \* Environment as a Markov Decision Process (MDP)

Markov Property: Future is independent of the past, given the present.

⇒ There is no reason to look in past  
· Present state use to model the future.

Steps involve in MDP ⇒ At each time step

① environment in some state  $\Rightarrow S_t$

↳ Decision maker can choose an action  $\Rightarrow a$

② moves environment to new state  $\Rightarrow S_{t+1}$

↳ Decision maker receives reward  $\Rightarrow R_a(S_t, S_{t+1})$

③  $S_{t+1}$  depends only on  $a$  and  $S_t$ .

Markov Property  $\Rightarrow$

future ( $s_{t+1}, s_{t+2}, \dots$ ) is independent  
of the past ( $s_{t-1}, s_{t-2}, \dots$ ), given the present ( $a_t, s_t$ )

## \* Policy Search Algorithm

- ① Brute force methods  $\rightarrow$  Evaluate every possible policy over every possible state
- ② Policy gradient methods  $\rightarrow$  Explore state space to find best policy
- ③ Value function methods  $\rightarrow$ 
  - Explicitly model environment as markov decision process.
  - Popular and robust.

Several implementations  $\Rightarrow$

- Q-learning
- SARSA
- Monte Carlo