

20 days of # 100 days of learning.

Reinforcement learning \Rightarrow

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

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Supervised

Given x , predict y

Unsupervised

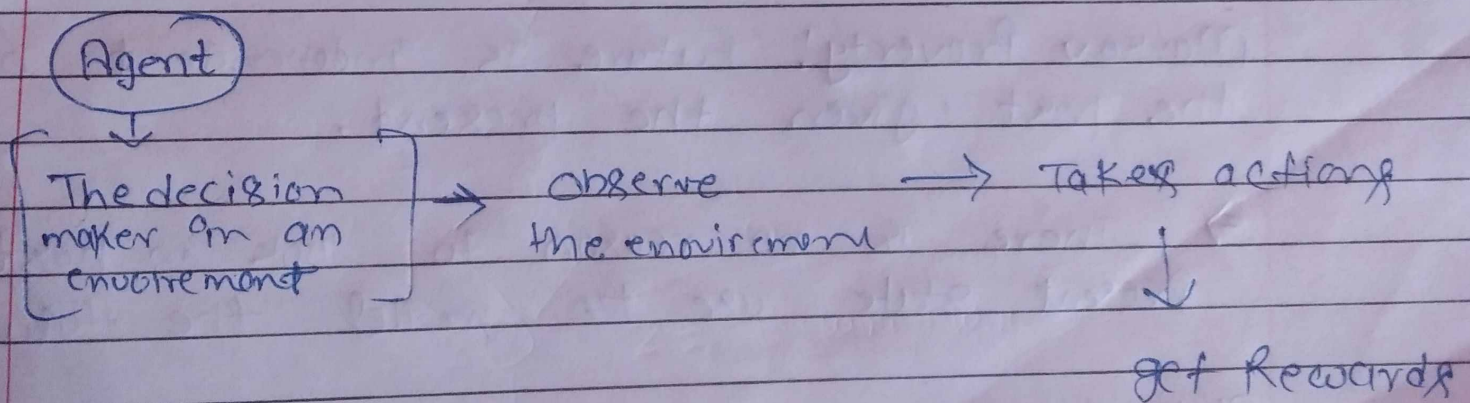
Given x , simplify x

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

Definition \Rightarrow

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

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



Reinforcement Learning	Supervised / Unsupervised Learning
Objective: choice "best" actions	Objective: Predict, Classify or Simplify
Environment is uncertain	Environment is known
Training involves exploring the environment	Training involves finding pattern in data.
Training process involved determining the best policy	Training process involves fitting the "best models"
Explicit dependency of rewards on previous action	Individual points 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 ⇒ S_t
↳ Decision maker can choose an action ⇒ a
- ② moves environment to new state ⇒ S_{t+1}
↳ Decision maker receives reward ⇒ $R_a(S_t, S_{t+1})$
- ③ S_{t+1} depends only a and S_t .

Markov Property \Rightarrow

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

* Policy Search Algorithms

- 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 Implementation \Rightarrow

- Q-learning
- SARSA
- Monte Carlo