

## **MONTE CARLO**

### **Aim:**

Aim to analyze the stability and convergence properties of Monte Carlo methods in reinforcement learning.

### **Description:**

Monte Carlo

The agent collects experience by interacting with the environment then uses this experiences to estimate the value of states or state-action pairs by averaging the returns obtained from different trajectories or episodes.

A Monte Carlo's stimulation is a model used to project the probability of a variety of outcomes the potential for random variables is present. Monte carlo simulations helps to explain the impact of risk & uncertainty in predictions & forecasting models.

## Algorithm.

Algorithm:

Initialize, for all  $s \in S, a \in A(s)$

$Q(s, a) \leftarrow \text{arbitrary}$

$\text{Returns}(s, a) \leftarrow \text{empty list}$

$\pi \leftarrow \text{an arbitrary } \epsilon\text{-soft policy}$

Repeat forever:

(a) Generate an episode using  $\pi$

(b) For each pair  $s, a$  appearing in the episode:

$R \leftarrow \text{return following the first occurrence of } s, a$

Append  $R$  to  $\text{Returns}(s, a)$

$Q(s, a) \leftarrow \text{average}(\text{Returns}(s, a))$

(c) For each  $s$  in the episode:

$a^* \leftarrow \arg \max_a Q(s, a)$

For all  $a \in A(s)$ :

$$\pi(s, a) \leftarrow \begin{cases} 1 - \epsilon + \epsilon / |A(s)| & \text{if } a = a^* \\ \epsilon / |A(s)| & \text{if } a \neq a^* \end{cases}$$

## Result:

Successfully Implemented the Monte carlo Algorithm.