

687 2017-10-17

Properties of First-Visit Monte Carlo

Intuition

- 1) generate many episodes
- 2) for each state average the return after the first time the state was visited in each episode

Pseudo-code:

Init $\pi \leftarrow$ policy to evaluate

$v \leftarrow$ arbitrary state-value f_v

Returns(s) \leftarrow empty list for each state

Forever

- 1) Generate an episode with π
- 2) for each state s in the episode
 $G \leftarrow$ return after first occurrence of s
Append G to Returns(s)
 $v(s) \leftarrow$ average (Returns(s))

Properties:

- Converges almost surely to v^π if each state is visited infinitely often.

Proof: Consider sequence of estimates $v_k(s)$ for $k \in \mathbb{N} (0, 1, \dots)$ for a particular state s .

$v_k(s) = \frac{1}{k} \sum_{i=1}^k G_i$ where G_i is the i^{th} return from s (not return at time i)

$$E[G_i] = E\left[\sum_{k=0}^{\infty} \gamma^k R_{t+k} \mid S_t = s, \pi\right] = v^\pi(s)$$

\hookrightarrow unbiased estimate of $v^\pi(s)$

(Bias(G_i) = $E[G_i] - \mu$ mean of distribution
Unbiased means Bias = 0.)

Apply Strong Law of Large Numbers:

Khinchin's S.L. of L.N.

Let x_1, x_2, \dots be i.i.d. random variables w/ expected value μ (finite), then

$\frac{1}{n} \sum_{i=1}^n x_i$ is a sequence of random variables converges to μ . I.e. $\Pr\left(\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n x_i = \mu\right) = 1$.
almost surely

By this Law of Large Numbers, with the x_i being G_i , we have $v_k(s) \rightarrow E[G_i]$ almost surely. What we have is convergence of V^k at each s , i.e. pointwise. Given that s is finite, $V_k \rightarrow V^*$, i.e., uniform convergence.

Rate of convergence

$$\text{Var}(v_k(s)) \approx \frac{1}{k}, \text{Stddev}(v_k(s)) \approx \frac{1}{\sqrt{k}}$$

$$\text{Var}(A+B) = \text{Var}(A) + \text{Var}(B) + 2 \text{Cov}(A, B) \rightarrow 0 \text{ since independent}$$

$$\text{So } \text{Var}\left(\frac{1}{k} \sum_{i=1}^k G_i\right) = \frac{1}{k^2} \sum_{i=1}^k \text{Var}(G_i)$$

$$= \frac{k}{k^2} \text{Var}(G) = \frac{1}{k} \text{Var}(G)$$

Every-Visit Monte Carlo

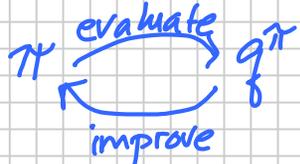
Only change: Add return from s at every place it occurs in an episode.

\Rightarrow The G_i in the same episode overlap so previous theorem does not directly apply.

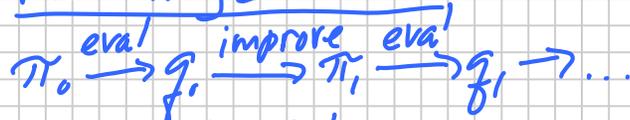
\Rightarrow However, the process still converges. (Likely with same rate, but Phil is not certain about it.)

Policy Improvement · Monte Carlo Control

- Generalized Policy Iteration (GPI)



MC Policy Iteration



- Evaluate with FV-MC
- Use q rather than v
(v is not adequate because we don't know P and R .)
- Same properties as PI if FV-MC is guaranteed to converge.
- Improve step is: $\pi'(s) = \arg \max_a q^{\pi}(s, a)$

What is the fix? More than one way:

- Exploring starts: randomly select s_0 and a_0
- Use stochastic policies

But this requires each s, a action pair to occur infinitely often!

Not quite: This is deterministic, while this is not!