

# Actor Critic Methods: From Paper to Code

Monte Carlo Control Problem

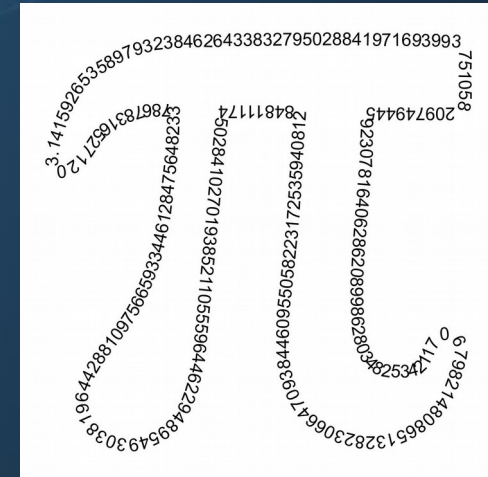
# Action Value Functions

- Without a model can't transition to next valuable state
- Replace  $V$  with  $Q$  and use first visit MC
- How to handle explore exploit dilemma?

# Exploring Starts



Pair random A, S  $\rightarrow$  exploring starts



Pick random A at episode start

Gives good coverage of state and action space

# Relaxing the Assumption of E.S.

- E.S. too limiting for some environments
- Use epsilon soft action selection

$$\text{Exploratory action} \rightarrow \frac{\epsilon}{|A(s)|}$$

$$\text{Greedy action} \rightarrow 1 - \epsilon + \frac{\epsilon}{|A(s)|}$$

- Greed increases over time

# Algorithm Overview

Initialize  $Q(s,a)$  arbitrarily for all  $s, a$ ; terminal  $\rightarrow 0$

Initialize arbitrary epsilon soft policy

Initialize list of Returns( $s,a$ ) for all states and actions

Repeat for large number of episodes:

    Generate episode using policy

    For each state  $s$  and action  $a$  in the agent's memory:

        Calculate the return that followed first visit to  $s, a$

        Append return  $G$  to list of Returns( $s,a$ )

        Update  $Q$  as the average of Returns( $s,a$ )

    For each state  $s$  in the agent's memory:

$$A^* \leftarrow \underset{a}{\operatorname{argmax}} Q(s, a)$$

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

200,000 games; plot cum. win ratio over 1000 games     $\epsilon \approx 0.001$

# Conclusion

- Performance not too bad
- Able to do it without exploring starts

