

Lecture4 Game Playing

Type Turn-Based(Main Focused on) & Real-Time & Simultaneous

Task

Given

S start

IsEnd(s)

Action(s)

Utility(s)

Reward for the end state s

Succ(s,a) = s'

Players

Player(s) ∈ Players

Which player moves in s

Output

Path (Sequence of actions)

Example: Bin Selection

Goal: Maximize the chosen number

Opponent

Adversarial

Helpful

Stochastic

Node Type

Node

A state and policy(Π) of a player

Game

Min Node

Downward-pointing triangle



Max Node

Upward-pointing triangle



Chance Node

Cicle Probability ∈ (0,1)



Value Function

V(s) denotes the value (utility) of state s

计算方式类似求期望

$$V(s, d) = \begin{cases} \text{Utility} & \text{isEnd(s)} \\ \text{Eval(s)} & d=0 \\ \sum_{a \in \text{Actions}(s)} V(\text{Succ}(s, \pi_{\text{agent}}(s)), d-1) & \text{Player(s) = agent} \\ \sum_{a \in \text{Actions}(s)} V(\text{Succ}(s, \pi_{\text{opp}}(s)), d-1) & \text{Player(s) = opp} \end{cases}$$

Minimax

Zero-Sum Game

Other name: MinMax, MM, Saddle Point

Assumes opponent selects the worst action to an agent

Time Complexity

Advanced Methods

Evaluation Function

Approximate the utility [Eval(s)]

Can stop earlier

Require domain-specific knowledge

Alpha-Beta Pruning

Basic

Compute TRUE utility

Ignore unnecessary path

General-Purpose

Algorithm

α 下限初始 $-\infty$, β 上限初始 $+\infty$

max层改变 α , $\alpha = \max \{ \text{这层} \alpha, \text{下层} \alpha, \beta \}$

min层改变 β , $\beta = \min \{ \text{这层} \beta, \text{下层} \alpha, \beta \}$

顺序: 先左子树, 再返回父节点, 再右子树

一旦出现 $\alpha \geq \beta$, 即可对右侧分支进行剪枝操作

Simultaneous Game

Example

Two-Finger Morra

Prisoner's Dilemma