

Foundations of Artificial Intelligence

43. Board Games: Stochastic Games

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Board Games: Overview

chapter overview:

- ▶ 40. Introduction and State of the Art
- ▶ 41. Minimax Search and Evaluation Functions
- ▶ 42. Alpha-Beta Search
- ▶ 43. Stochastic Games
- ▶ 44. Monte-Carlo Tree Search Framework
- ▶ 45. Monte-Carlo Tree Search Configurations

43.1 Expected Value

Discrete Random Variable

- ▶ a **random event** (like the result of a die roll)
 - ▶ is described in terms of a **random variable** X
 - ▶ with associated **domain** $\text{dom}(X)$
 - ▶ and a **probability distribution** over the domain
- ▶ if the number of outcomes of a random event is **finite** (like here), the random variable is a **discrete random variable**
- ▶ and the probability distribution is given as a **probability** $P(X = x)$ that the **outcome** is $x \in \text{dom}(X)$

Discrete Random Variable: Example



informal description:

- ▶ you plan to **invest** in **stocks**
- ▶ your analyst **expects** these **stock price changes**:

Bellman Inc.	Howard Corp.
+2 with 30%	+3 with 40%
+1 with 60%	±0 with 10%
±0 with 10%	-1 with 50%

Markov Tec.
 +4 with 20%
 +2 with 30%
 -1 with 50%

formal model:

- ▶ discrete random variables B , H and M

$\text{dom}(B) = \{2, 1, 0\}$
 $\text{dom}(H) = \{3, 0, -1\}$
 $\text{dom}(M) = \{4, 2, -1\}$

▶ $P(B = 2) = 0.3$ $P(H = 3) = 0.4$
 $P(B = 1) = 0.6$ $P(H = 0) = 0.1$
 $P(B = 0) = 0.1$ $P(H = -1) = 0.5$

$P(M = 4) = 0.2$
 $P(M = 2) = 0.3$
 $P(M = -1) = 0.5$

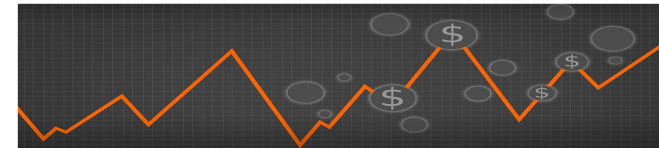
Expected Value

- ▶ the **expected value** $\mathbb{E}[X]$ of a random variable X is a **weighted average** of its outcomes
- ▶ it is computed as the **probability-weighted sum** of all outcomes $x \in \text{dom}(X)$, i.e.,

$$\mathbb{E}[X] := \sum_{x \in \text{dom}(X)} P(X = x) \cdot x$$

- ▶ in stochastic environments, it is **rational** to deal with uncertainty by **optimizing expected values**

Expected Value: Example



formal model:

- ▶ discrete random variables B , H and M

▶ $\text{dom}(B) = \{2, 1, 0\}$
 $\text{dom}(H) = \{3, 0, -1\}$
 $\text{dom}(M) = \{4, 2, -1\}$

▶ $P(B = 2) = 0.3$
 $P(B = 1) = 0.6$
 $P(B = 0) = 0.1$

$P(H = 3) = 0.4$ $P(M = 4) = 0.2$
 $P(H = 0) = 0.1$ $P(M = 2) = 0.3$
 $P(H = -1) = 0.5$ $P(M = -1) = 0.5$

expected gain:

$\mathbb{E}[B] = P(B = 2) \cdot 2 + P(B = 1) \cdot 1 + P(B = 0) \cdot 0$
 $= 0.3 \cdot 2 + 0.6 \cdot 1 + 0.1 \cdot 0 = 1.2$

$\mathbb{E}[H] = P(H = 3) \cdot 3 + P(H = 0) \cdot 0 + P(H = -1) \cdot (-1)$
 $= 0.4 \cdot 3 + 0.1 \cdot 0 + 0.5 \cdot (-1) = 0.7$

$\mathbb{E}[M] = P(M = 4) \cdot 4 + P(M = 2) \cdot 2 + P(M = -1) \cdot (-1)$
 $= 0.2 \cdot 4 + 0.3 \cdot 2 + 0.5 \cdot (-1) = 0.9$

rational decision: buy Bellman Inc.

43.2 Stochastic Games

Definition

Definition (stochastic game)

A **stochastic game** is a

7-tuple $\mathcal{S} = \langle S, A, T, s_I, S_*, \text{utility}, \text{player} \rangle$ with

- ▶ finite set of **positions** S
- ▶ finite set of **moves** A
- ▶ **transition function** $T : S \times A \times S \mapsto [0, 1]$ that is **well-defined for $\langle s, a \rangle$** (see below)
- ▶ **initial position** $s_I \in S$
- ▶ set of **terminal positions** $S_* \subseteq S$
- ▶ **utility function** $\text{utility} : S_* \rightarrow \mathbb{R}$
- ▶ **player function** $\text{player} : S \setminus S_* \rightarrow \{\max, \min\}$

A transition function is **well-defined for $\langle s, a \rangle$** if $\sum_{s' \in S} T(s, a, s') = 1$ (then a is **applicable** in s) or $\sum_{s' \in S} T(s, a, s') = 0$.

Reminder: Bounded Inc-and-square Game

informal description:

- ▶ Players alternatingly apply a
 - ▶ **increment-mod10** (*inc*) or
 - ▶ **square-mod10** (*sqr*) move
- ▶ on the natural numbers from 0 to 9
- ▶ starting from the number 1;
- ▶ if the game reaches the number 6 or 7
- ▶ or **after a fixed number of moves n**
- ▶ \max obtains utility r (\min : $-r$) where x is the current number.

formal model:

- ▶ $S = \{s_0^k, s_1^k, \dots, s_9^k \mid 0 \leq k \leq n\}$
- ▶ $A = \{\text{inc}, \text{sqr}\}$
- ▶ for $0 \leq i \leq 9$ and $0 \leq k \leq n$:
 - ▶ $\langle s_i^k, \text{inc}, s_{(i+1) \bmod 10}^{k+1} \rangle \in T$
 - ▶ $\langle s_i^k, \text{sqr}, s_{i^2 \bmod 10}^{k+1} \rangle \in T$
- ▶ $s_I = s_1^0$
- ▶ $S_* = \{s_6^k, s_7^k \mid 0 \leq k \leq n\} \cup \{s_i^n \mid 0 \leq i \leq 9\}$
- ▶ $\text{utility}(s_i^k) = i$ for all $s_i^k \in S_*$
- ▶ $\text{player}(s_i^k) = \max$ if k even and $\text{player}(s_i^k) = \min$ otherwise

Example: Stochastic Inc-and-square Game

informal description:

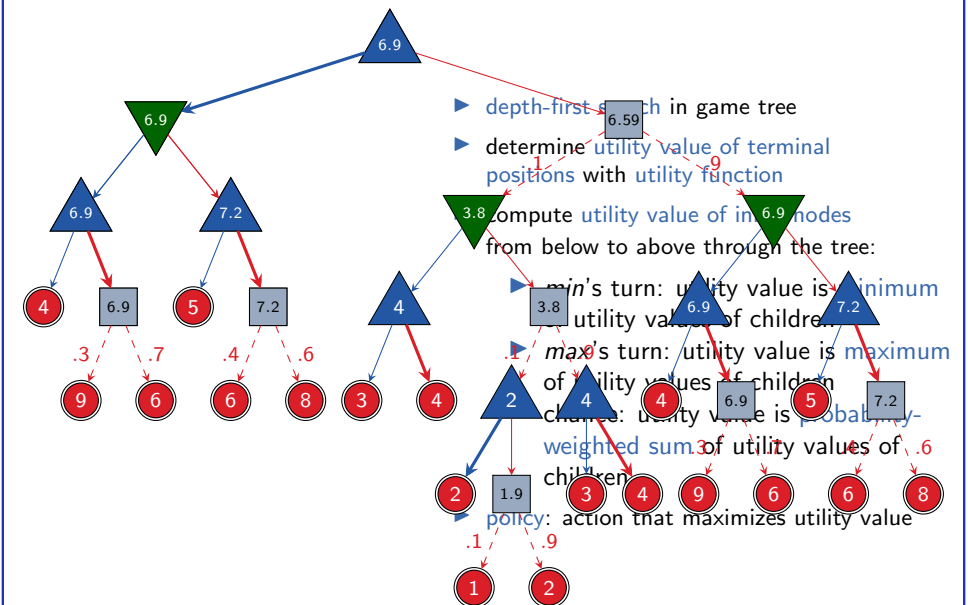
- ▶ rules like **bounded inc-and-square game**
- ▶ except applying **square-mod10** move in state s
- ▶ results **with probability $\frac{s}{10}$** in $(s^2 \bmod 10)$
- ▶ and otherwise in $((2 \cdot s) \bmod 10)$

formal model:

- ▶ $S = \{s_0^k, s_1^k, \dots, s_9^k \mid 0 \leq k \leq n\}$
- ▶ $A = \{\text{inc}, \text{sqr}\}$
- ▶ for $0 \leq i \leq 9$ and $0 \leq k < n$:
 - ▶ $T(s_i^k, \text{inc}, s_{(i+1) \bmod 10}^{k+1}) = 1$
 - ▶ $T(s_i^k, \text{sqr}, s_{i^2 \bmod 10}^{k+1}) = \frac{s}{10}$
 - ▶ $T(s_i^k, \text{sqr}, s_{(2 \cdot i) \bmod 10}^{k+1}) = \frac{10-s}{10}$
- ▶ $(T(s, a, s') = 0 \text{ for all other } \langle s, a, s' \rangle)$
- ▶ $s_I = s_1^0$
- ▶ $S_* = \{s_6^k, s_7^k \mid 0 \leq k \leq n\} \cup \{s_i^n \mid 0 \leq i \leq 9\}$
- ▶ $\text{utility}(s_i^k) = i$ for all $s_i^k \in S_*$
- ▶ $\text{player}(s_i^k) = \max$ if k even and $\text{player}(s_i^k) = \min$ otherwise

43.3 Expectiminimax

Idea and Example



Discussion

- **expectiminimax** is the simplest (decent) search algorithm for stochastic games
- yields optimal policy* (in the game-theoretic sense, i.e., under the assumption that the opponent plays perfectly)
- **max** obtains **at least** the state value computed for the root **in expectation**, no matter how **min** plays
- if **min** plays perfectly, **max** obtains **exactly** the computed value **in expectation**
- the reward for **max** that is actually obtained may be **a higher or lower** than the state value computed for the root, independently of how **min** plays

(*) for finite trees; otherwise things get more complicated

43.4 Summary

Summary

- ▶ **Stochastic games** are board games with an additional element of **chance**.
- ▶ **Expectiminimax** is a minimax variant for stochastic games with identical behavior in *max* and *min* nodes.
- ▶ In **chance nodes**, it propagates the **probability-weighted sum** of all successors.
- ▶ Expectiminimax has **similar guarantess** as minimax but **in expectation**.