

5. Adversarial Search

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Artificial Intelligence

Overview 概述

- □ The minimax algorithm generates the entire game search space.

 minimax算法生成整个博弈搜索空间。
- □ The alpha—beta pruning algorithm allows us to prune large parts of it.
 alpha—beta剪枝算法允许我们将其剪去大部分。
- □ However, alpha—beta still has to search all the way to terminal states for at least a portion of the search space.

然而, alpha-beta仍然需要搜索抵达终端状态的所有途径、至少是搜索空间的一部分。

This depth is usually not practical, because moves must be made in a reasonable amount of time.

这个深度通常是不实际的, 因为移动必须在合理的时间内完成。

Apply a Heuristic Evaluation Function 应用启发式评价函数

☐ Claude Shannon proposed instead that programs should cut off the search earlier and apply a heuristic evaluation function to states in the search, effectively turning nonterminal nodes into terminal leaves. The suggestion is in two ways:

克劳德·香农提出:程序应该早一些剪断搜索,并在搜索中对状态应用启发式评价函数,有效地将非终端节点转换为终端叶节点。该建议是用如下两种方法:

Use EVAL instead of UTILITY

用EVAL来代替UTILITY

EVAL: a heuristic evaluation function, which estimates the position's utility.

EVAL: 一个启发式评价函数,用于估计位置的效用。

■ Use Cutoff-Test instead of Terminal-Test

用CUTOFF-TEST来代替TERMINAL-TEST

CUTOFF-TEST: decides when to apply EVAL.

CUTOFF-TEST: 确定何时应用EVAL函数。

Heuristic H-Minimax vs. Minimax 启发式H-Minimax与Minimax

```
function H-MINIMAX(s, d) returns an action if CUTOFF-TEST(s, d) then return EVAL(s) if PLAYER(s) = MAX then return \max_{a \in \text{ACTIONS}(s)} \text{H-MINIMAX}(\text{RESULT}(s, a), d+1) if PLAYER(s) = MIN then return \min_{a \in \text{ACTIONS}(s)} \text{H-MINIMAX}(\text{RESULT}(s, a), d+1)
```

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function MINIMAX(s) returns an action if Terminal-Test(s) then return Utility(s) if Player(s) = Max then return \max_{a \in ACTIONS(s)} MINIMAX(RESULT(s, a)) if Player(s) = MIN then return \min_{a \in ACTIONS(s)} MINIMAX(RESULT(s, a))
```

Evaluation Functions 评价函数

- An evaluation function returns an *estimate* of the expected utility of the game from a given position. How to design good evaluation functions?
 - 一个评价函数返回从一个给定位置该博弈的期望效用估计。如何设计好的评价函数?
 - It should order the terminal states in the same way as the true utility function:
 - a) the states that are wins must evaluate better than draws,
 - b) the states that are draws must be better than losses.

它应该与真实效用函数相同的方式对终端状态进行整理:

- a) 获胜状态必须评价为优于平局,
- b) 平局状态必须评价为优于失败。
- The computation must not take too long. 计算的时间一定不能太长。
- Nonterminal states should be strongly correlated with actual chances of winning. 非终端状态应该与实际获胜的机会密切相关。

Weighted Linear Function 加权线性函数

-- A kind of evaluation function 一种评价函数

EVAL(s) =
$$w_1 f_1(s) + w_2 f_2(s) + ... + w_n f_n(s)$$

= $\sum_{i=1}^{n} w_i f_i(s)$

where 其中

- \mathbf{w}_i -- a weight $\sqrt{2}$
- **I** f_i -- a feature of the position 位置的特征

For Chess 对于国际象棋

- f_i could be the numbers of each kind of piece on the board 应该为棋盘上每种棋子的数量
- **w**_i could be the values of the pieces: 1 for pawn, 3 for bishop, etc. 应该为棋子的数值: 1表示兵、3表示象、等等。

Chess 国际象棋



(a) White to move 白棋要走子



(b) White moved 白棋走子后

Two chess positions that differ only in the position of the rook at lower right:

- (a) Black has an advantage of a knight and two pawns, which should be enough to win the game.
- (b) White will capture the queen, giving it an advantage that should be strong enough to win.

两盘国际象棋布局仅在右下角车的位置不同:

(a) 黑棋多一个马和两个兵,应该赢得这盘棋。(b) 白棋将捕获皇后,使得它处于足以获胜的态势。

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