

Topic 5.3 Improving the Performance of the MINIMAX search Strategy

A. Common Factors:

- 1) Database of experience like 'best opening move', 'great winning [/loosing] combination', ...
- 2) Maintaining transposition table that marks repeated states
- 3) Forward pruning, that excludes equivalent states
- 4) Singular expansion, where only 'clearly better' moves are considered

B. Using 'cutoff test' and heuristic evaluation function

i) To have a reduced search space 'cutoff tests' are applied in place of 'terminal tests':

- Simple: Set a fixed depth; Not so good.
- Sophisticated: quiescence search; [A quiescent state does not exhibit wild swing in near future.]

ii) To apply on states at cutoff depths, utility functions are replaced with heuristic functions, which return expected utility of states. [In Chess, material values like pawn = 1, knight/bishop = 3, queen = 9, etc. are used.]

C. Using alpha-beta pruning:

- Cutting off in compliance with already calculated minimax values, that is, values of
 - best choice for the maximizing player, α ,
 - best choice for the minimizing player, β .

➤ Example:

Say, in absence of true minimax values, we are given the following sequence of numbers from which we are supposed to assign a value to each newly generated 'terminal' state, that is, a state at the cutoff depth:

4, 3, -1, 4, 5, 2, 1, -2, -5, 3, 2, 1.

[-5, +5]

Further:

Branching factor = 2;

Cutoff depth = 2 moves or 4 plies;

MAX makes the opening move;

Left to right expansion of the tree.

✓ Try to prune a branch in every occasion and go up;

✓ To go up, one must know at least the range of all the siblings, and it is not enough.

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4, 3, -1, 4, 5, 2, 1, -2, -5, 3, 2, 1.