Game Tree Searching by Min / Max Approximation*

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Summary of paper's goals:

Paper discussed current MinMax with alpha beta pruning method and suggested an improvement over it. In one-line purpose of the paper *is to explain a method which identifies, in a game tree which node to expand next*.

The improvement was done in two ways

- a) Instead of expanding every relevant node (irrelevant are pruned nodes by alpha beta pruning) paper suggested approximating which *node* to expand next. It is done by assigning the derivatives of generalized means at each node. Algorithm then uses that value to determine which node to expand next. Unlike Alpha beta; this algorithm falls under the category of iterative heuristic; i.e. it grows tree one step at a time.
- b) While implementing; algorithm approximates the formulae and uses a weight>0(starts with weight 0) termed as Penalties to assign to each node; Penalties of a node is the sum of penalties from that node up until the root node. Using that value; algorithm determines which current leaf node has maximum impact on root node. Key difference here as opposed to regular AB search is that node at a different depth could be expanded.

Important points to note about this and this class (iterative heuristic) of algorithms.

- a) Unlike depth first search schemes; this requires trees to be explicitly stored (as the expansion isn't uniform) hence memory requirements are large.
- b) Penalty based scheme like this one are more geared towards improving the value of estimate at the root rather than towards selecting the best move.

Summary of paper's results:

Experiments were conducted with 49 different starting positions of Connect-Four game. Game was based under two resources constraints one is Time bound 1 to 5 seconds with 1 second interval and Moves bound 1000 – 5000 moves with 1000 moves interval.

In Time based categorie AB came out superior by roughly 15 % however in moves based bound MM approximation came out on top by roughly 25%. In addition to that, number of distinct positions considered by AB was roughly 3 times of that of approximation method.