

Lecture 6: Alpha-Beta enhancements

AI For Traditional Games
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Total nodes expanded by minimax

$$N(b, d) = 1 + b + b^2 + \dots + b^{d-1} + b^d$$

$$b \cdot N(b, d) = b + b^2 + \dots + b^{d-1} + b^d + b^{d+1}$$

$$b \cdot N(b, d) - N(b, d) = b^{d+1} - 1$$

$$N(b, d) \cdot (b - 1) = b^{d+1} - 1$$

$$N(b, d) = \frac{b^{d+1} - 1}{b - 1}$$

$$N(b, d) \approx \frac{b^{d+1}}{b - 1} = b^d \frac{b}{b - 1}$$

Nodes expanded by iterative deepening

- Minimax work at depth d: $b^d \cdot b / (b - 1)$

$$b^d \frac{b}{b - 1} + b^{d-1} \frac{b}{b - 1} + b^{d-2} \frac{b}{b - 1} + \dots + b \frac{b}{b - 1} + \frac{b}{b - 1}$$

$$\frac{b}{b - 1} (b^d + b^{d-1} + b^{d-2} + \dots + b + 1)$$

$$\frac{b}{b - 1} b^d \frac{b}{b - 1} = b^d \left(\frac{b}{b - 1} \right)^2$$

Advantages of iterative deepening

- How do I order nodes?
 - Hand analysis/heuristics
 - Use information from the previous iterations
 - Start with principal variation
 - History Heuristic

Move ordering

- Can we learn a good move ordering?
- Killer Heuristic:
 - Find the move that is causing the most cutoffs at each depth of the tree
 - Try it first
 - S.G. Akl and M.M. Newborn, The Principle Continuation and the Killer Heuristic, ACM Annual Conference, 1977, 466-473.

Move Ordering

- Killer heuristic just finds a single move
 - Can we generalize to all moves?
- Simplistic “learning”

History Heuristic

- Each move has a score
 - Increment score whenever a move is the best move in a state (or causes a cutoff)
 - Score is 2^d , where d is the depth of the tree analyzed below the move
- Sort moves by score

Null move

- If one player is in a strong position, they could skip their turn and still win
 - Perform “null” move
 - Search to depth $2/3$ ply shallower than required
 - If the value of the game is still better on previous branches, it’s a win
 - Otherwise re-search with the full tree
- Hop Step?
- zugzwang -- sometimes it’s better not to move

Quiescence search

- Quiescence = quiet
 - Searching to a fixed depth may not be advantageous
 - eg if a capture has just been made, and the capture response hasn't
 - Extend search until position is quiet
 - eg no captures and no check

Horizon Effect

- A result of limited depth knowledge
- Something bad is about to happen, but find a way to delay it until it happens after the search depth
- May turn a minor problem into a catastrophic one
 - Make a bad move now to avoid a worse state by the horizon effect
 - The worse state still happens later

Using alpha-beta bounds

- When performing alpha-beta pruning, don't have to start with $\alpha = -\infty$, $\beta = \infty$
- If you have evidence that the value is in a smaller range, can search with a tighter bound
 - Produces smaller tree
 - But, if the value returned is at the bounds of your range you have to re-search

Graphs versus Trees

- A graph is not a tree
 - Turning a graph into a tree results in exponential blow-up
- Transposition (hash) Table
 - Key: Minimax uses (almost) no memory
 - Use some memory to detect tree
 - When two paths transpose into the same state

Breakthrough example

- When is the first transposition in Breakthrough?
- Assume you are searching 8 ply
 - 3-ply into the game
 - Save 5-ply ($20^5 \approx 3.2$ million states)

Transpositions

- When should we look for transpositions?
 - Near the top of the tree
 - Large savings
 - Likely to find transpositions
- When shouldn't we look for transpositions?
 - Near the bottom of the tree
 - Minimal savings
 - Unlikely to find transpositions

Transpositions

- What is needed to test for transpositions?
 - Naïve - list of states, and a linear search
 - Better - tree of states log(s) search
 - Best - hash table
- What hash function should we use?