

# Games III: Alphabeta

15-150 M21

Lecture 0728-0  
28 July 2021



# 0 Optimizing Minimax

# The MiniMax Algorithm

Fix a search depth  $d$ .

- **evaluate**:
  - ▶ If at depth  $d$ , call the estimator to obtain the value of the current state
  - ▶ Otherwise, search through the available moves to find the best move (for the current player). The value of the current state is the value of the state resulting from that move.
- **search**: Make a sequence containing all the game states that would result from making a single (legal) move from the current state, evaluate all them, and pick the best one (for the current player)

**Demonstration:**

**Minimax**

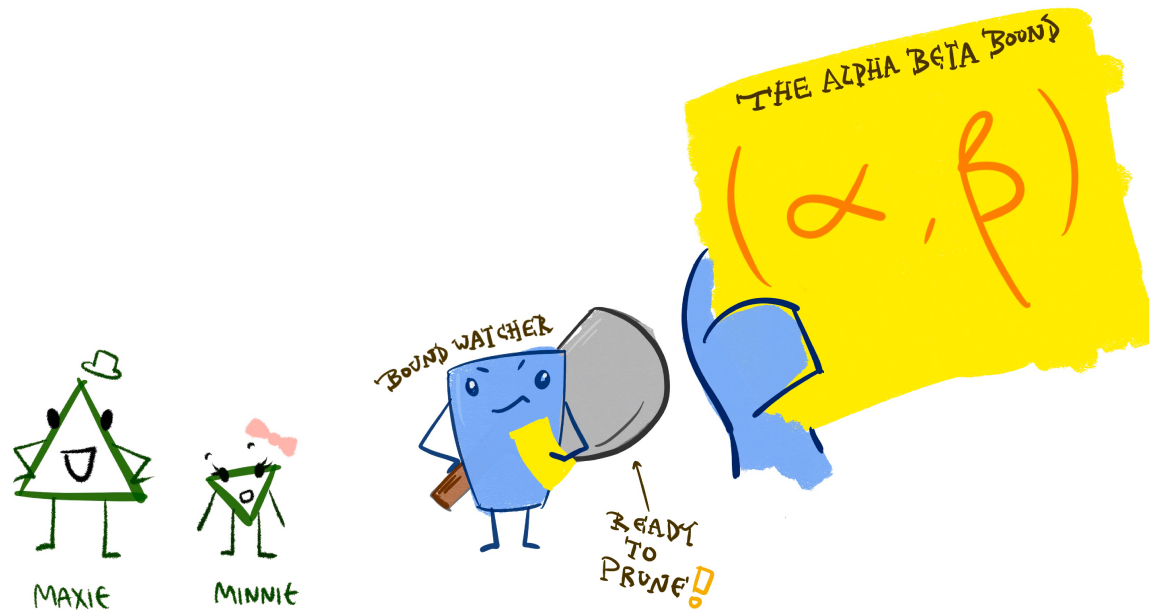
## **Advantages of Minimax:**

- Correctly determines optimal play
- Massively parallelizable

## **Disadvantages of Minimax:**

- Huge amount of work
- Indeed, often performs unnecessary computation

As we perform MiniMax, we want to keep track of “what can be guaranteed” to inform us when we’re exploring an irrelevant subtree.



# Idea

So, for every point along the minimax algorithm, there is some estimator guess value  $\alpha$ , which represents the greatest value that **Maxie** can *guarantee*.

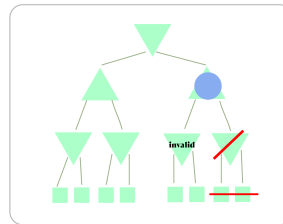
Analogously, we'll keep track of some value  $\beta$  representing the least value that **Minnie** can guarantee. It must be the case that  $\alpha \leq \beta$ .

When **Minnie** encounters a node whose value is  $\leq \alpha$ , then she can “prune” the rest of the current subtree: **Maxie** won't let the game get to this point. If **Maxie** encounters a node whose value is  $\geq \beta$ , then she prunes.

Once the bound becomes invalid,



Prune!



# The Alphabet Algorithm

Fix a search depth  $d$ .

- **evaluate:**
  - ▶ If at depth  $d$ , call the estimator to obtain the value of the current state
  - ▶ Otherwise, search through the available moves to find the best move (for the current player). The value of the current state is the value of the state resulting from that move.
- **search:** Try possible moves one-by-one,
  - ▶ evaluate the resulting game state (passing down the current  $(\alpha, \beta)$ ) to obtain  $x$
  - ▶ Update  $\alpha$  or update  $\beta$  or prune (as called for in the chart)
  - ▶ If we evaluate all the moves without pruning, then pick the best one (for the current player)

	$x \leq \alpha$	$\alpha < x < \beta$	$\beta \leq x$
Maxie	Ignore	Update $\alpha$	Prune
Minnie	Prune	Update $\beta$	Ignore



# Demonstration:

# AlphaBeta

In the SEQ signature:

```
datatype 'a lview =  
  Nil | Cons of 'a * 'a seq
```

```
val show1 : 'a seq -> 'a lview  
val hide1 : 'a lview -> 'a seq
```

**5 minute break**