

# Foundations of Artificial Intelligence

## 42. Board Games: Alpha-Beta Search

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# Board Games: Overview

## chapter overview:

- 40. Introduction and State of the Art
- 41. Minimax Search and Evaluation Functions
- 42. Alpha-Beta Search
- 43. Stochastic Games
- 44. Monte-Carlo Tree Search Framework
- 45. Monte-Carlo Tree Search Configurations

# Limitations of Minimax

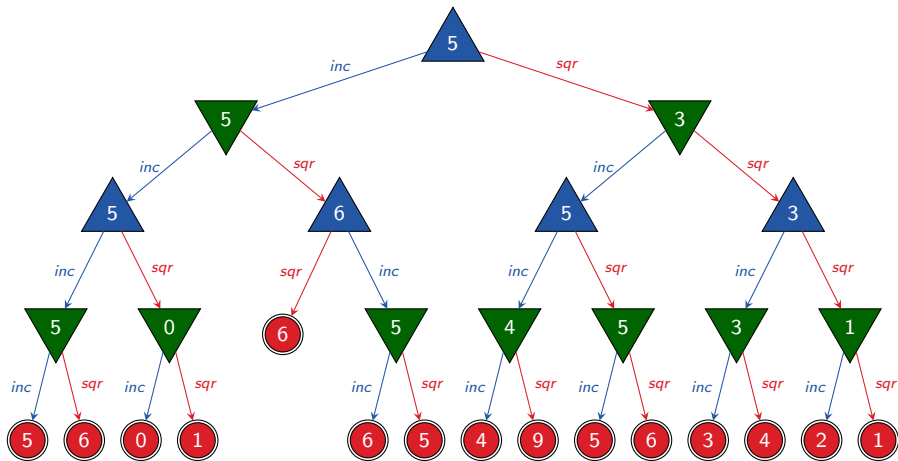


What if the size of the game tree is **too big for minimax**?

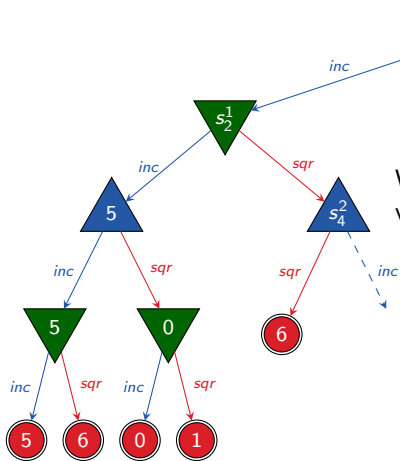
↪ Heuristic **Alpha-Beta Search**

# Alpha-Beta Search

# Can We Save Search Effort?



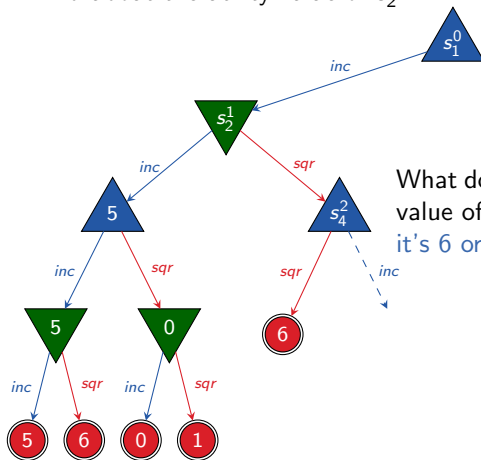
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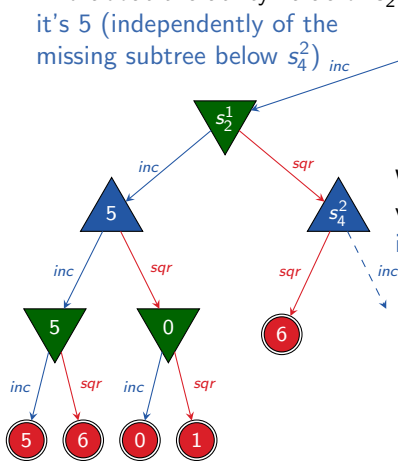
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What do we know about the utility value of  $s_4^2$  in this situation?  
it's 6 or higher

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it's 5 (independently of the  
missing subtree below  $s_4^2$ )

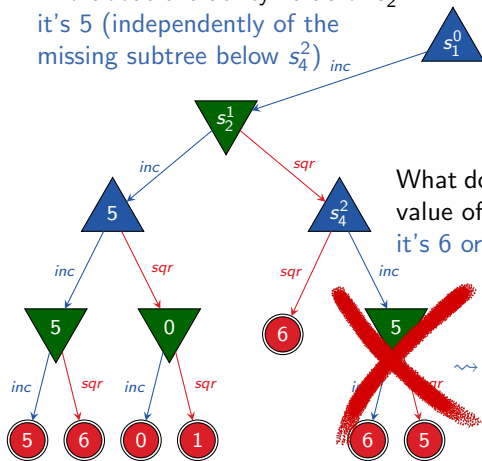


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What do we know about the utility  
value of  $s_4^2$  in this situation?  
it's 6 or higher

we don't have to look at this

# Idea

**idea:** use two values  $\alpha$  and  $\beta$  during minimax search such that

- $\alpha$  is known lower bound on utility of *max* and
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# Idea

**idea:** use two values  $\alpha$  and  $\beta$  during minimax search such that

- $\alpha$  is known lower bound on utility of *max* and
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it holds for every recursive call that a subtree

- is **not interesting** if its utility value is  $\leq \alpha$   
because *max* will never enter it when playing optimally
- is **not interesting** if its utility value is  $\geq \beta$   
because *min* will never enter it when playing optimally
- rooted at a *max* node is **pruned** if utility  $\geq \beta$
- rooted at a *min* node is **pruned** if utility  $\leq \alpha$

# Alpha-Beta Search: Pseudo Code

- algorithm skeleton the same as minimax
- function signature extended by two variables  $\alpha$  and  $\beta$

```
function alpha-beta-main( $p$ )
```

```
   $\langle v, move \rangle := \text{alpha-beta}(p, -\infty, +\infty)$ 
```

```
return  $move$ 
```

# Alpha-Beta Search: Pseudo-Code

**function**  $\text{alpha-beta}(p, \alpha, \beta)$

**if**  $p$  is terminal position:

**return**  $\langle \text{utility}(p), \text{none} \rangle$

initialize  $v$  and  $\text{best\_move}$

[as in minimax]

**for each**  $\langle \text{move}, p' \rangle \in \text{succ}(p)$ :

$\langle v', \text{best\_move}' \rangle := \text{alpha-beta}(p', \alpha, \beta)$

    update  $v$  and  $\text{best\_move}$

[as in minimax]

**if**  $\text{player}(p) = \text{max}$ :

**if**  $v \geq \beta$ :

**return**  $\langle v, \text{none} \rangle$

$\alpha := \max\{\alpha, v\}$

**if**  $\text{player}(p) = \text{min}$ :

**if**  $v \leq \alpha$ :

**return**  $\langle v, \text{none} \rangle$

$\beta := \min\{\beta, v\}$

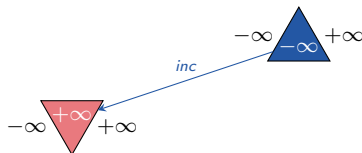
**return**  $\langle v, \text{best\_move} \rangle$

# Example



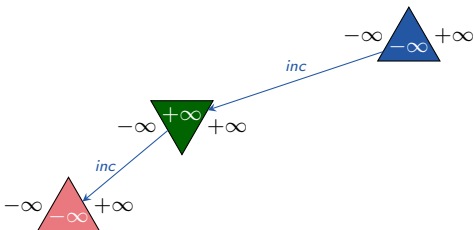
- $\alpha$  is lower bound on utility of *max*
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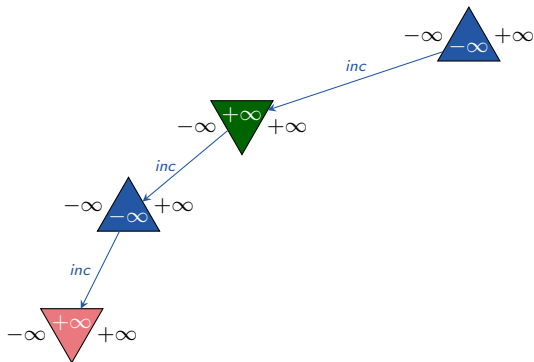
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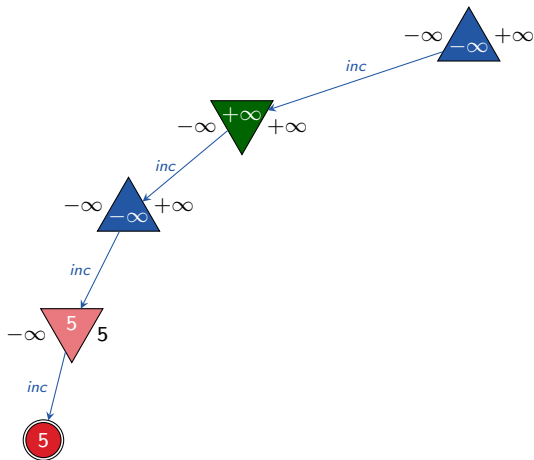


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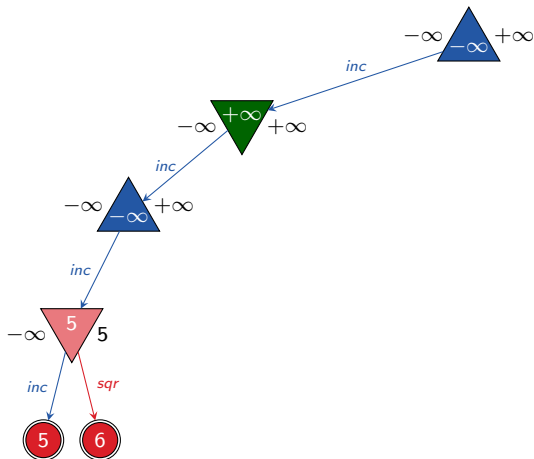
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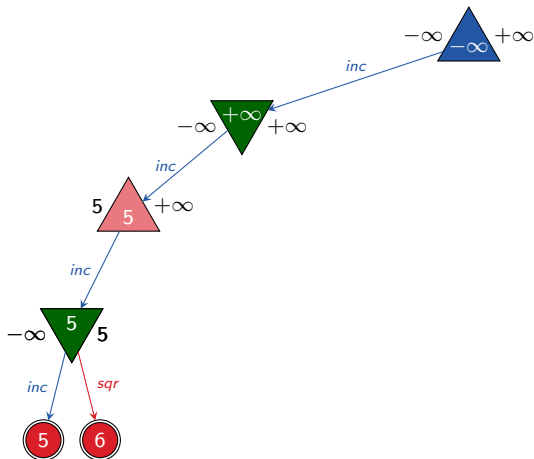
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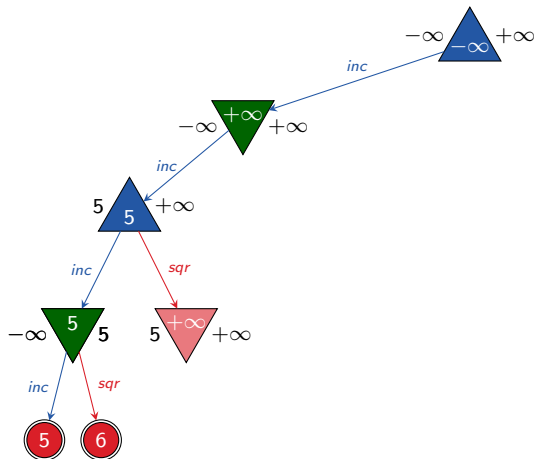
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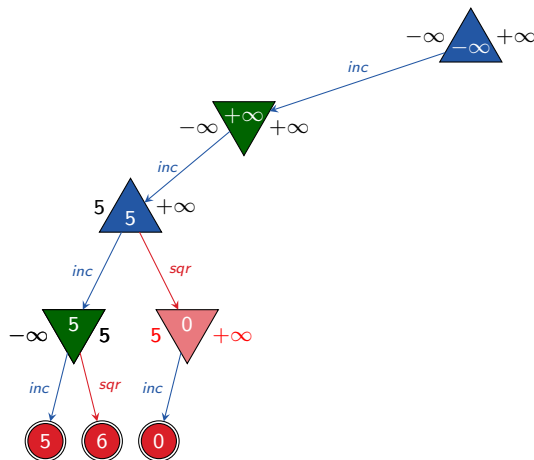
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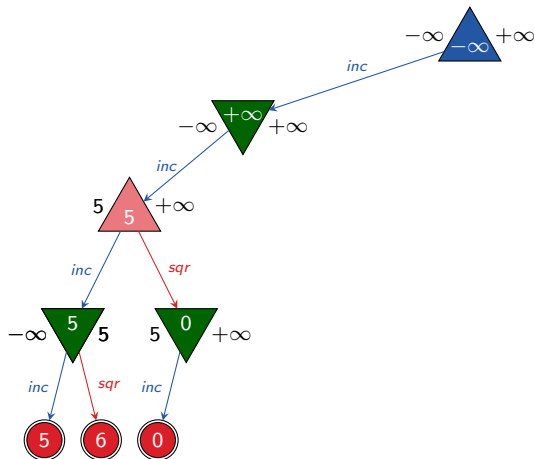
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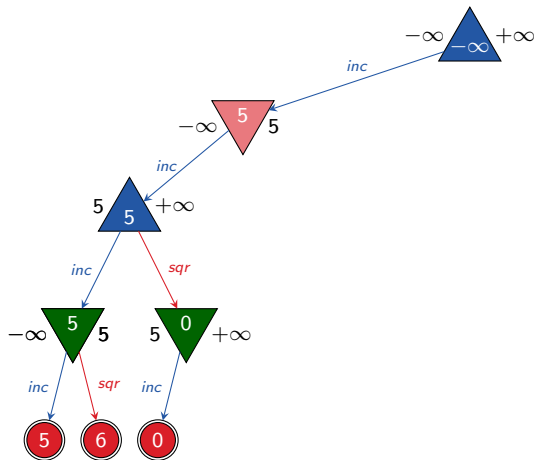
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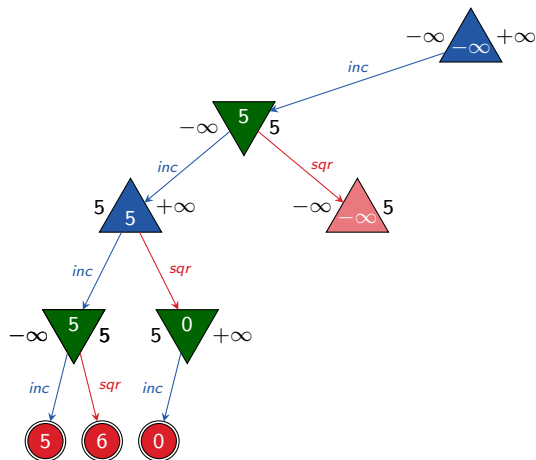
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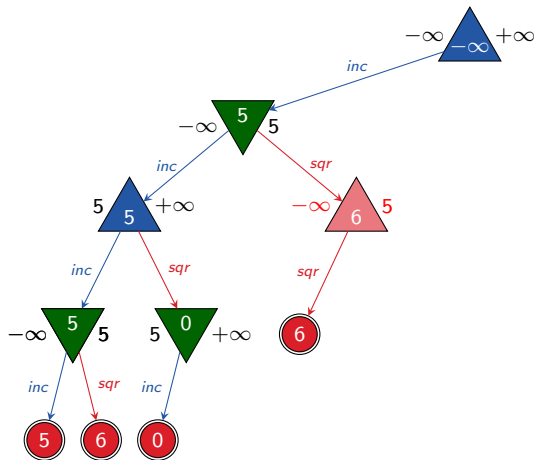


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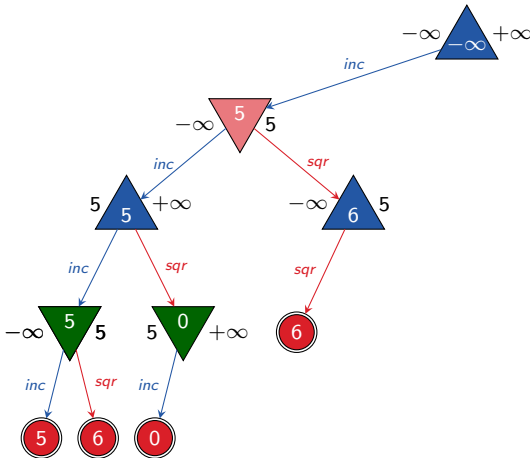
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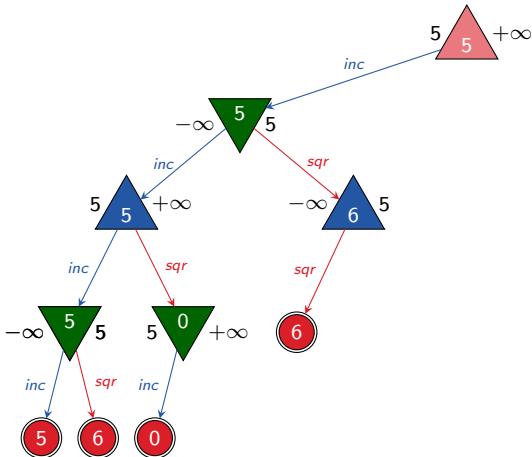
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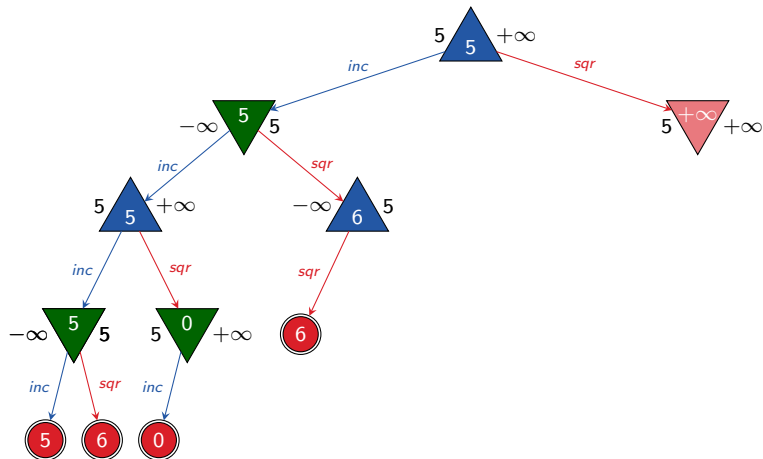
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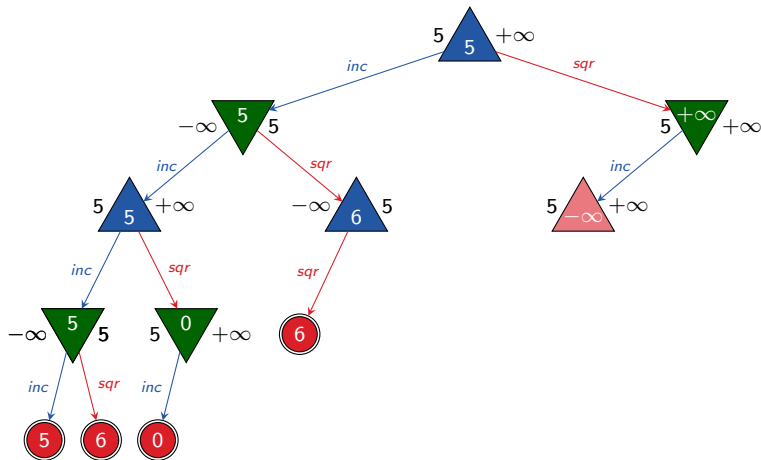
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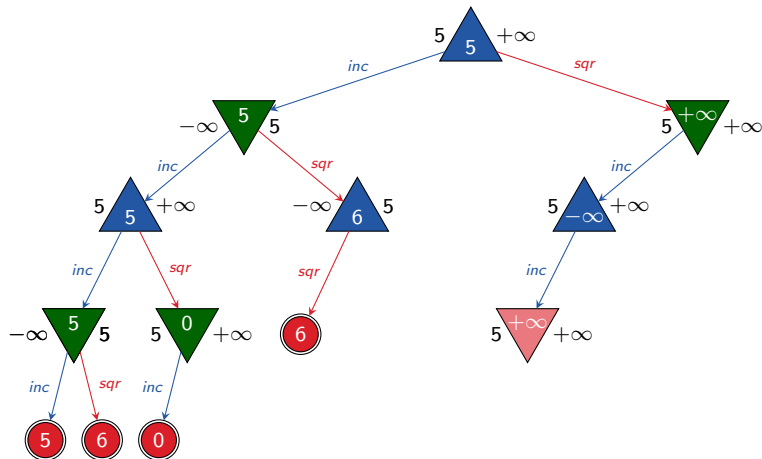
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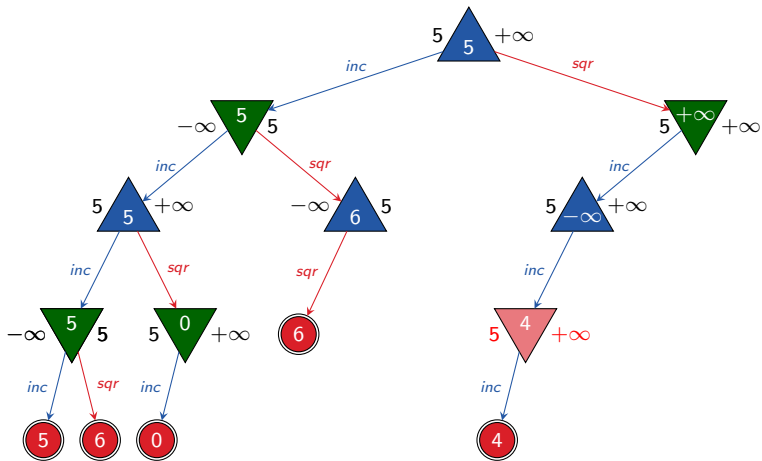
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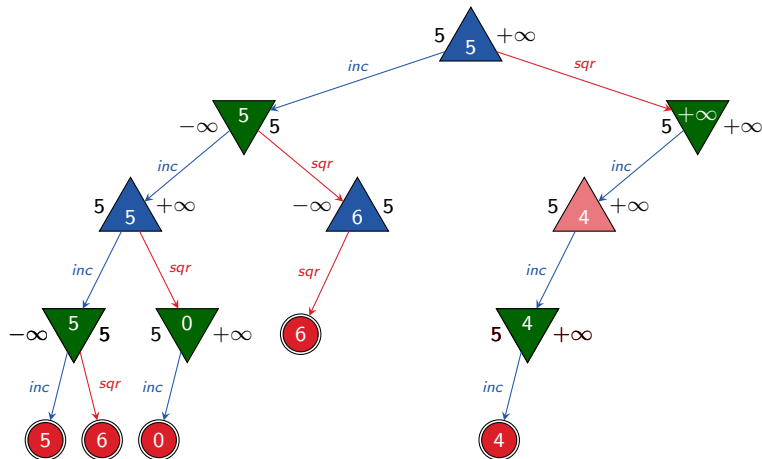
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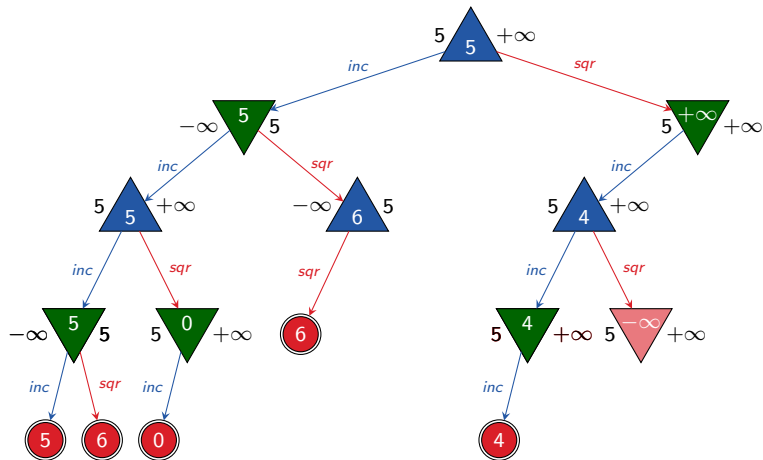
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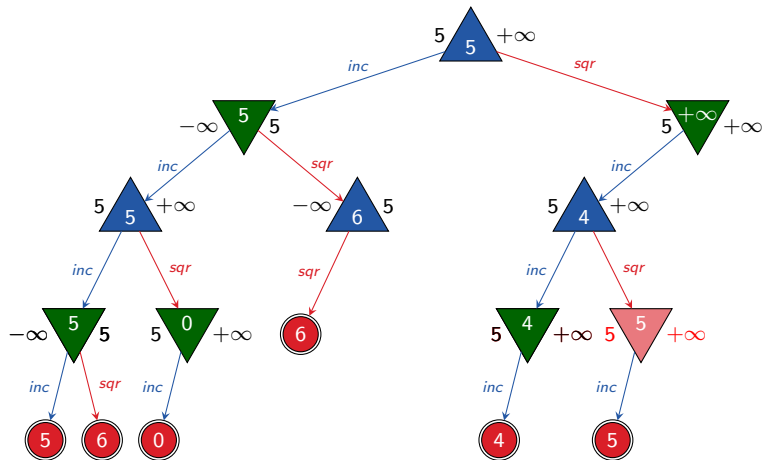
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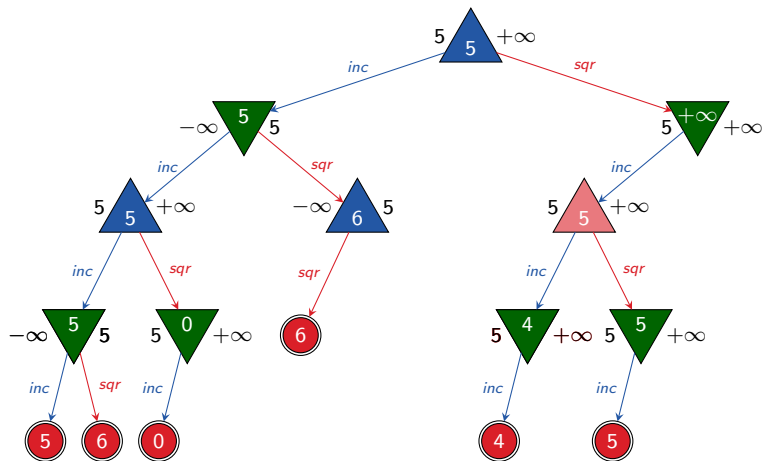
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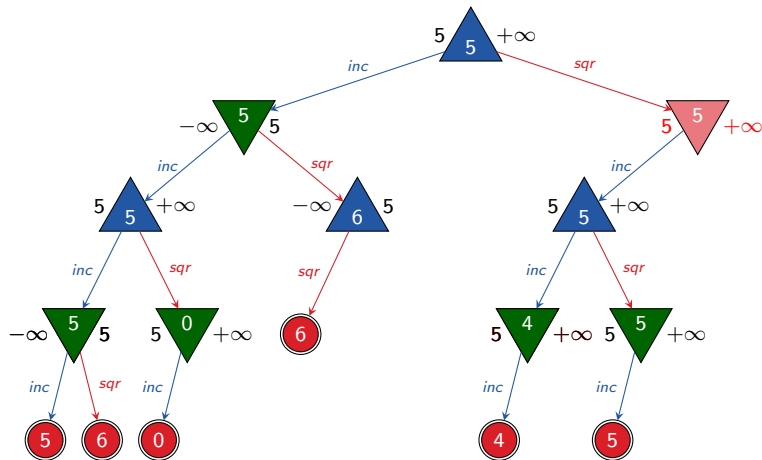
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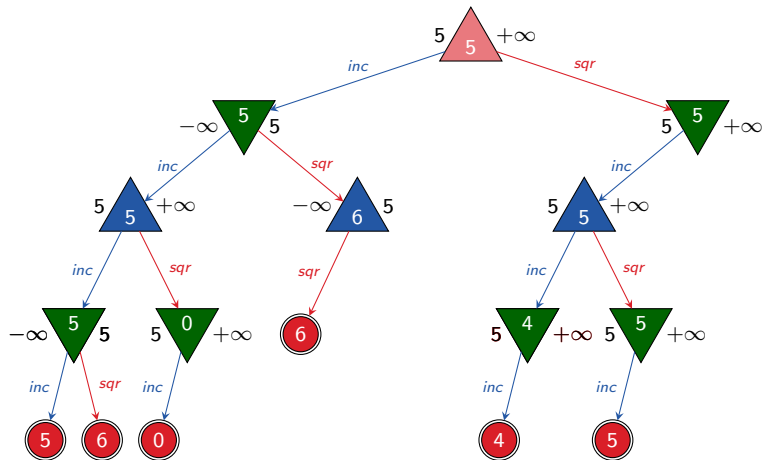
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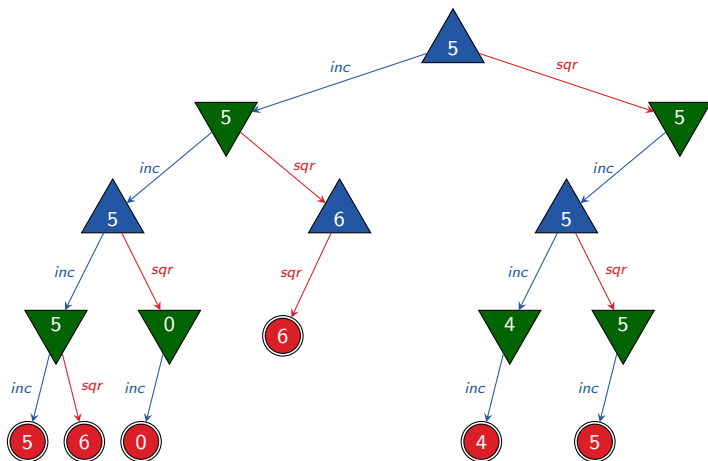
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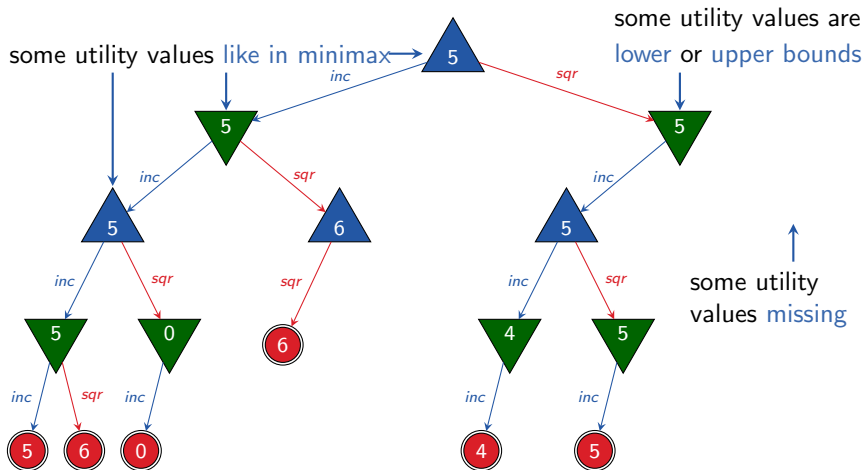
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# Discussion



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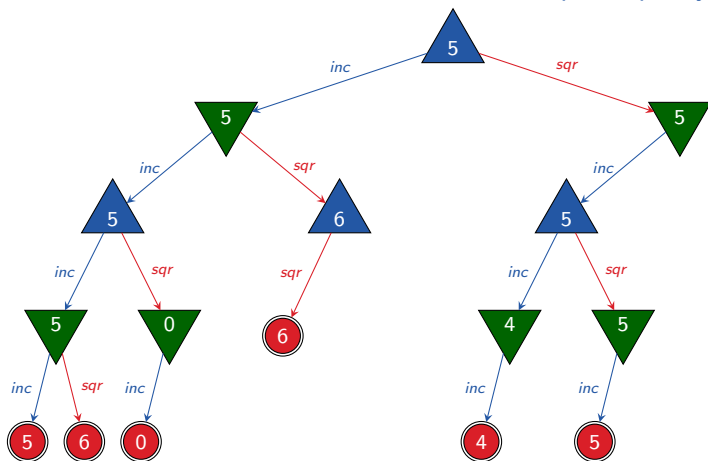


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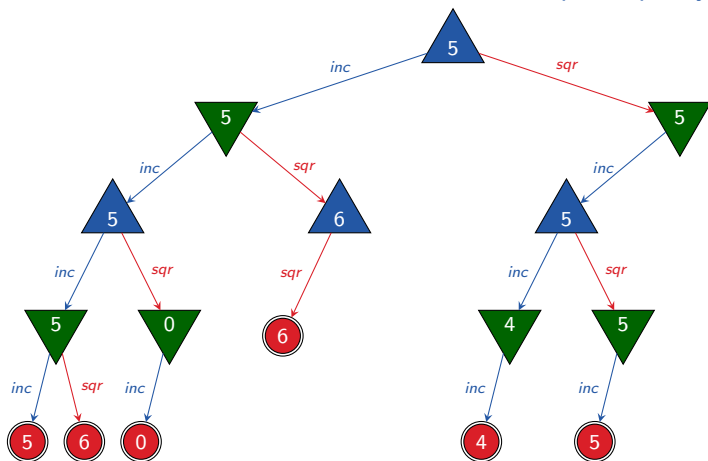
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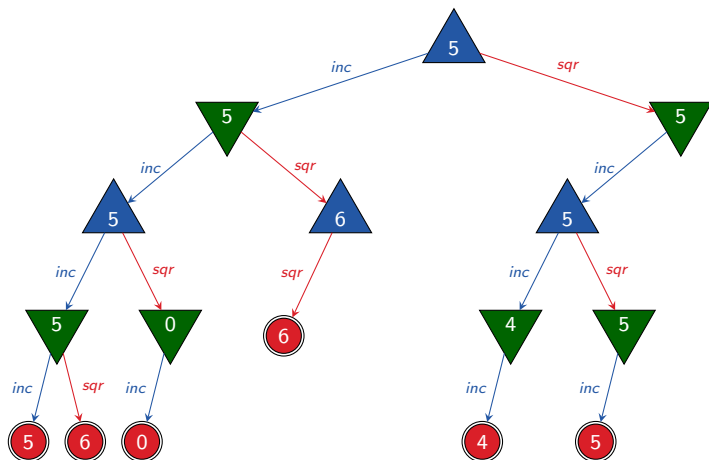
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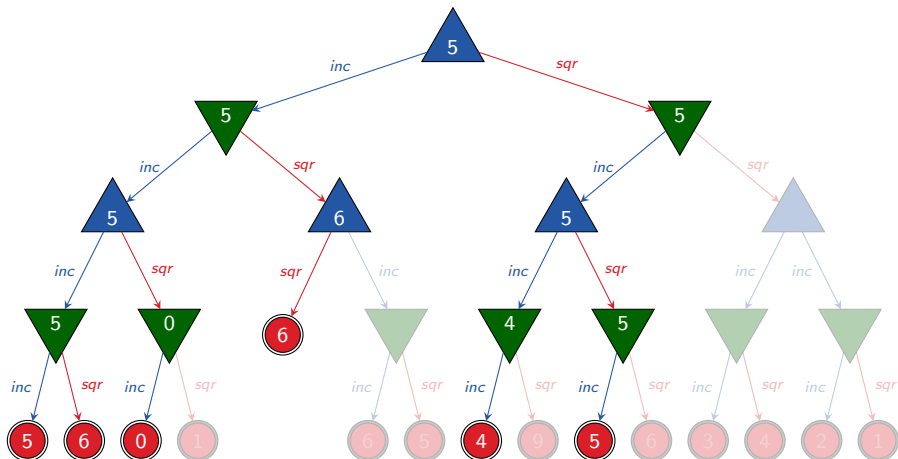
- only partial
- optimal in positions reachable under optimal play
- need to take earliest move in case of ties

# Move Ordering

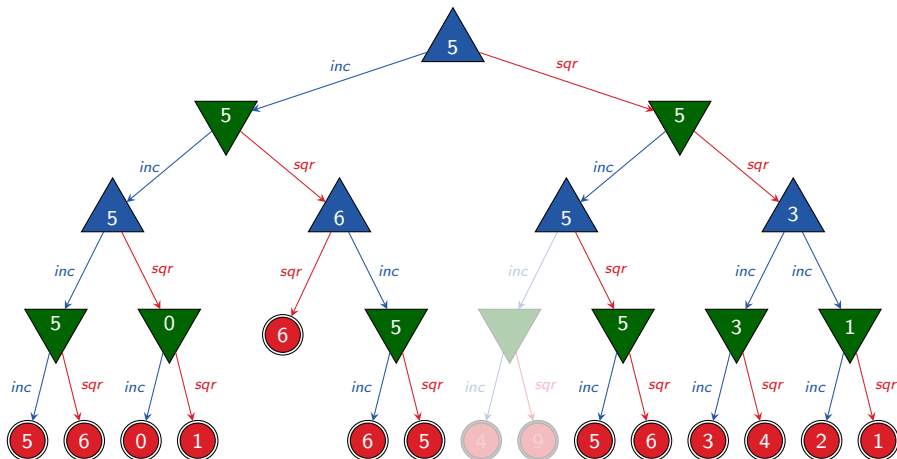
# How Much Effort Did We Save?



# How Much Effort Did We Save?



# Have We Been Lucky?



if successors are considered in **reverse order**, we prune only a few positions

# Move Ordering

**idea:** first consider the successors that are likely to be best

- **domain-specific ordering function**  
e.g. chess: captures < threats < forward moves < backward moves
- **dynamic move-ordering**
  - first try moves that have been good in the past
  - e.g., in iterative deepening search:  
best moves from previous iteration

# How Much Do We Gain with Alpha-Beta Pruning?

**assumption:** uniform game tree, depth  $d$ , branching factor  $b \geq 2$ ;  
*max* and *min* positions alternating

- perfect move ordering

- best move in every position is considered first  
(this cannot be done in practice)
- effort reduced from  $O(b^d)$  (minimax) to  $O(b^{d/2})$
- doubles the search depth that can be achieved in same time

- random move ordering

- effort still reduced to  $O(b^{3d/4})$  (for moderate  $b$ )

in practice, it is often possible to get close to the optimum



# Heuristic Alpha-Beta Search

- combines **evaluation function** and **alpha-beta search**
- often uses additional techniques, e.g.
  - quiescence search
  - transposition tables
  - forward pruning
  - specialised sub-procedure for critical parts of the game (e.g., endgame database in chess)
  - ...

⇒ reaches expert level of play in chess

# Summary

# Summary

## alpha-beta search

- stores which utility both players can force somewhere else in the game tree
- exploits this information to **avoid unnecessary computations**
- can have significantly **lower search effort than minimax**
- best case: search **twice as deep** in the same time