

Project 2.1

# **The Abalone Game**

-

**Group 5**

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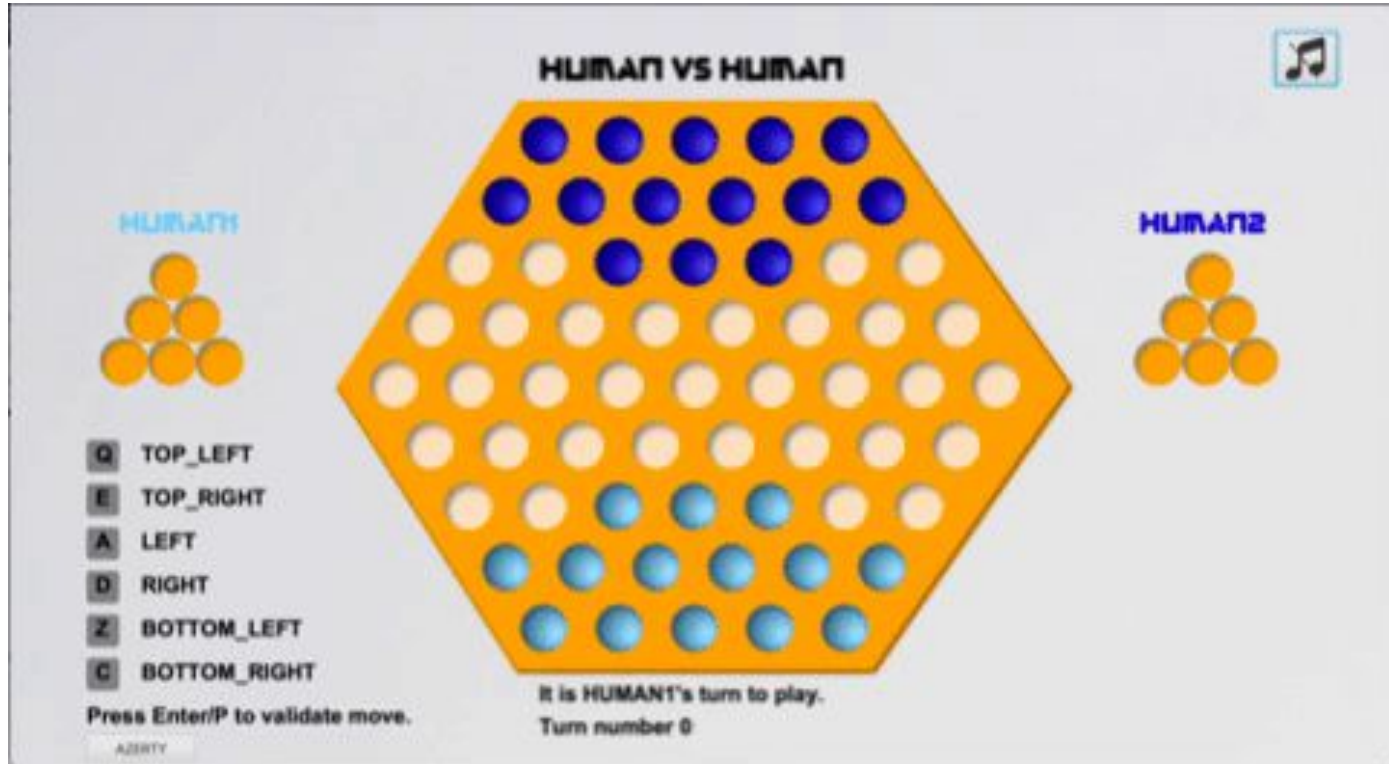
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# The Game - Rules



# Our GUI

## ABALONE

PLAYER 1'S NAME:

MCTS vs Human

PLAYER 2'S NAME:

HUMAN

MCTS



START

Neutral evaluation function

Easy

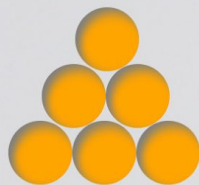
Rules

Contact

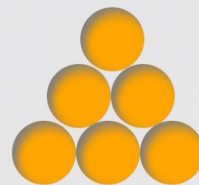
# HUMAN VS HUMAN



CLEMENT



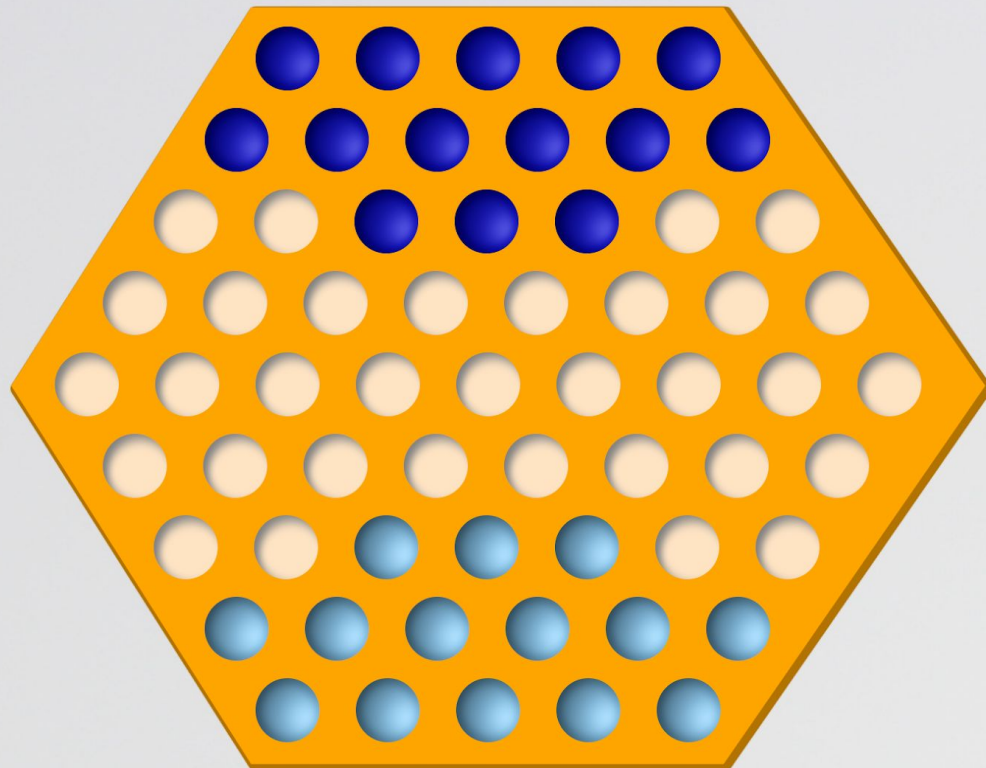
MATHIAS



- Q** TOP\_LEFT
- E** TOP\_RIGHT
- A** LEFT
- D** RIGHT
- Z** BOTTOM\_LEFT
- C** BOTTOM\_RIGHT

Press Enter/P to validate move.

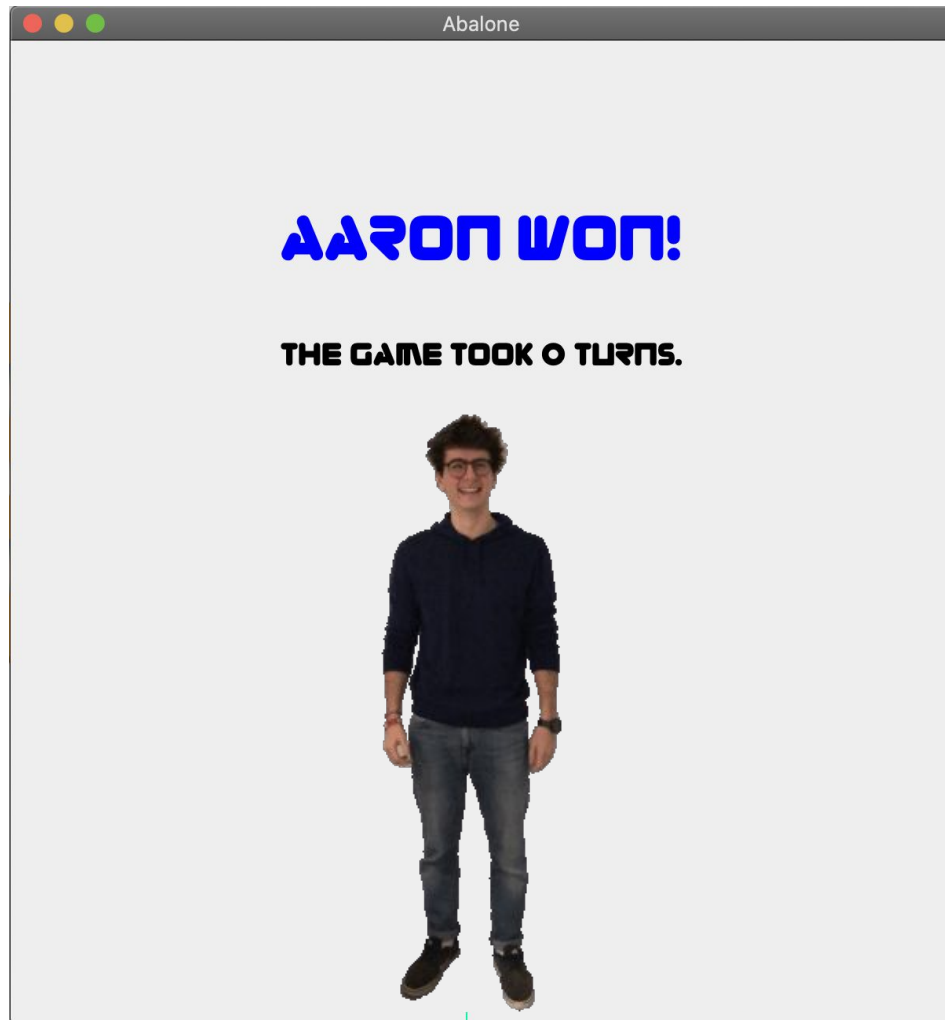
AZERTY



It is CLEMENT's turn to play.

Turn number 0

# Win Page



# Rule-Based Algorithm

```
if there is a sumito move:
    pick random sumito move;
else:
    if there is a pushing move:
        pick random pushing move;
    else:
        if there is a triple move:
            pick random triple move;
        else:
            if there is a double move:
                pick random double move;
            else:
                pick random single move;
```

Sumito move:

any move that  
ejects an  
opponent's marble

Pushing move:

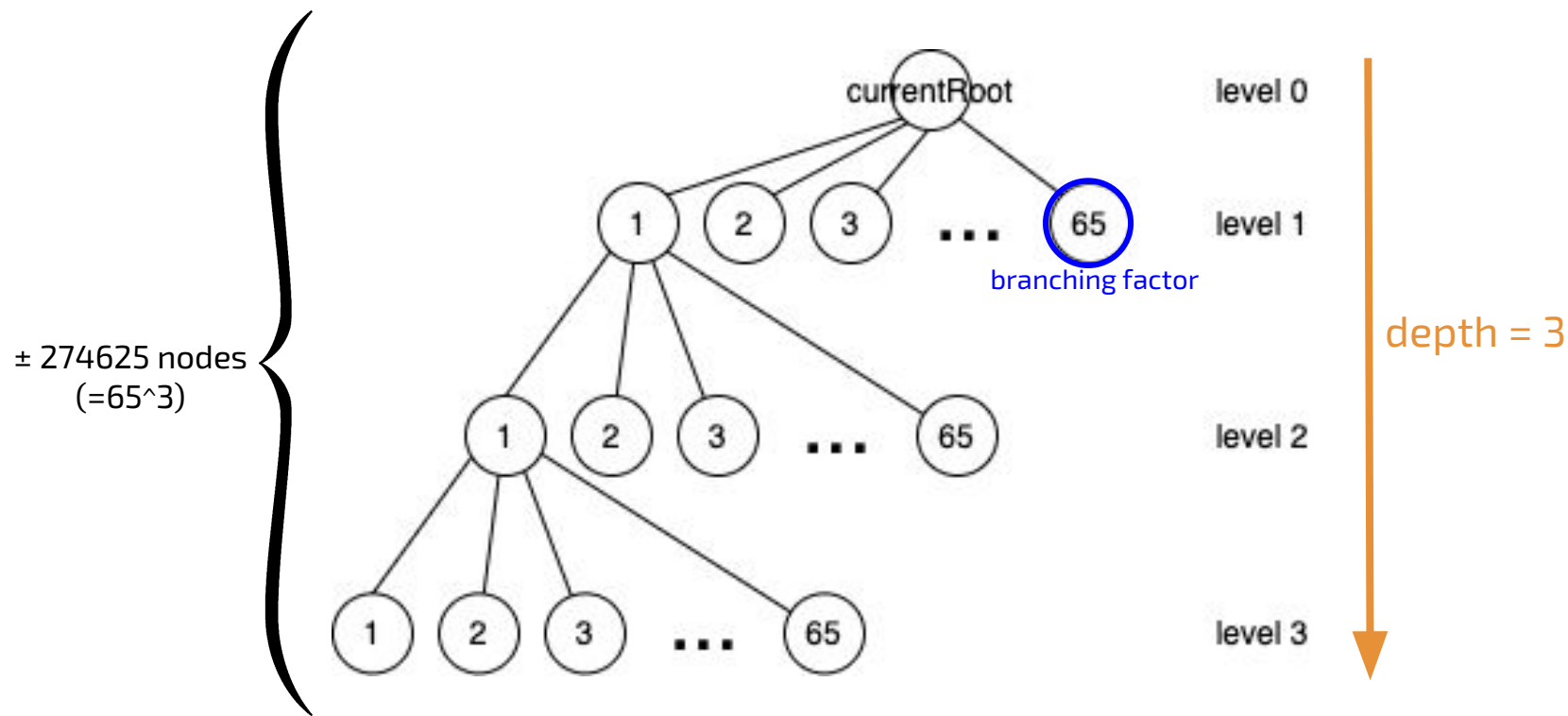
any move that  
displace an  
opponent's marble

Super fast at  
playing but not  
super performant...

# Game Tree Structure

*used by ABTS*

- creation in BFS :  $O(b^d)$
- visit in DFS



**IMPOSSIBLE TO COMPUTE ENTIRE GAME TREE !**



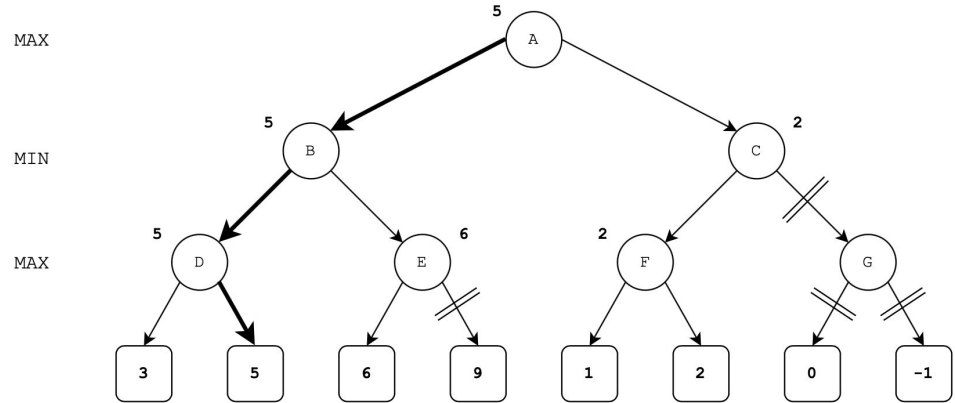
# Alpha-Beta Tree Search

## Minimax

- Min player and Max player
- Evaluate positions
- Costs a lot of computational time
- Complexity of  $O(M^D)$ .

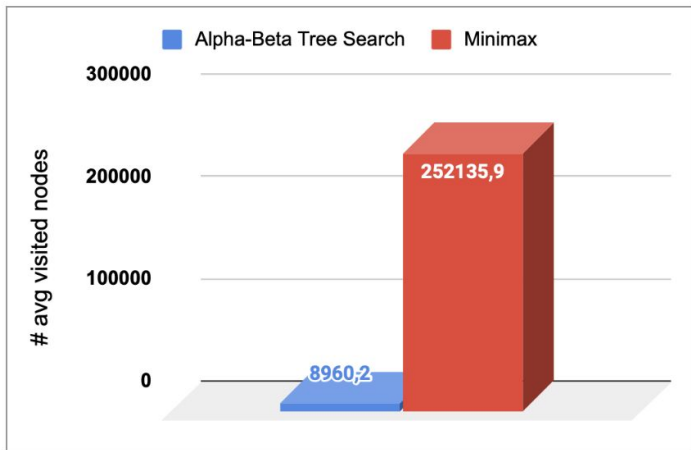
## Alpha-Beta pruning

- Will result in the same outcome
- Prunes nodes not worth checking
- Allows the AI player to search at depth 3 in reasonable time
- improves with a better move ordering



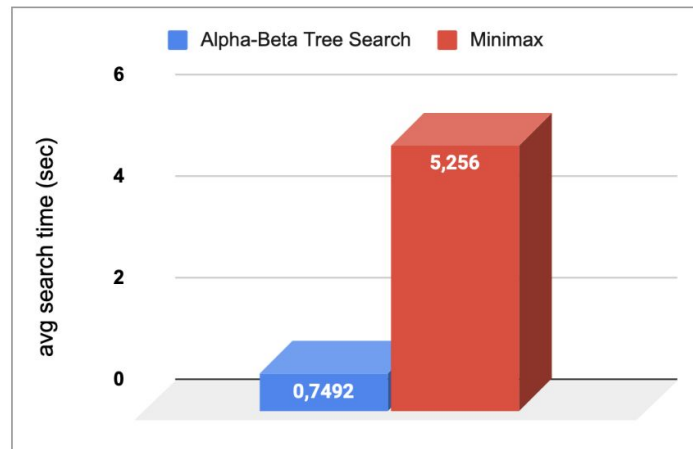
# → Experiment

## Alpha-Beta pruning vs MiniMax



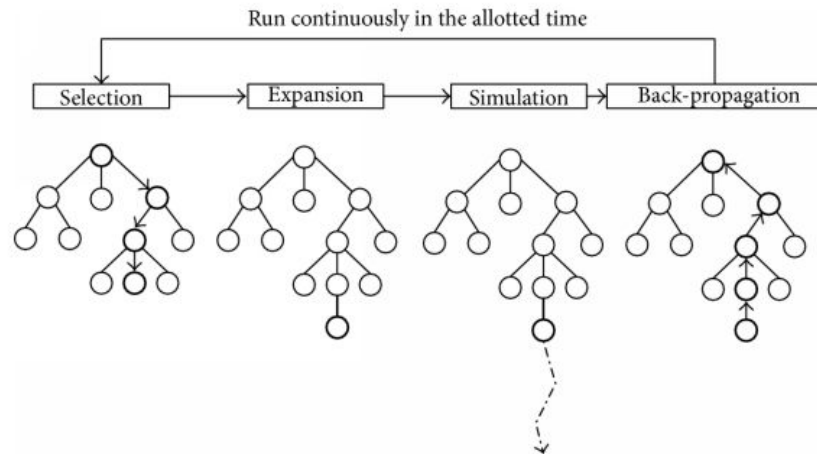
- Search algorithm runs faster
- 4.5sec saved on average

- Huge investigated nodes difference
- Up to 96% saved nodes



# Monte-Carlo Tree Search

- **Selection**
  - Start with the root
  - Select the best child until a leaf is reached
- **Expansion**
  - Expands the tree with current node children
- **Simulation**
  - Select a random child
  - Simulate it and get the results
- **Back-propagation**
  - Back-propagate the score till the root

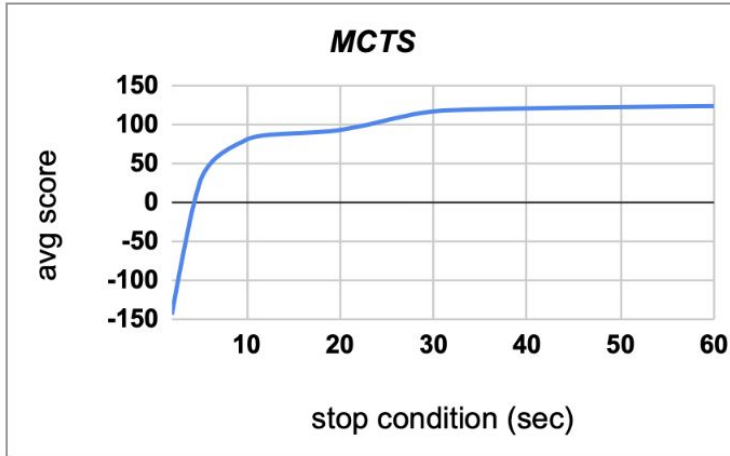


→ Repeat until **stop condition** is met, then output the best move

└→ Fixed amount of time

# → Experiment

MCTS stop condition



- Efficiency threshold at 10 sec
- Inefficient below it
- Constant efficiency above

# Evaluation Functions

...all based on heuristics

## Neutral

- Most efficient
- Weights adapting
- From offensive to defensive (and vice versa)

## Offensive

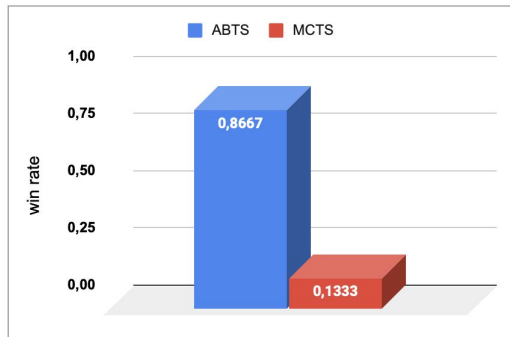
- Focuses on ejecting
- Risked approach

## Defensive

- Focuses on consolidating
- Safed approach

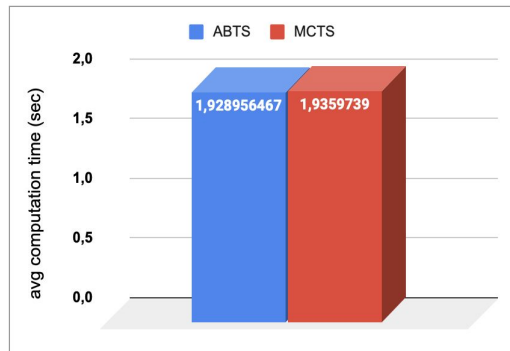
# → Experiment

ABTS vs MCTS



ABTS better  
win rate

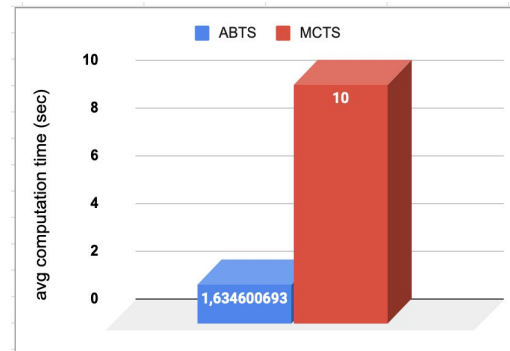
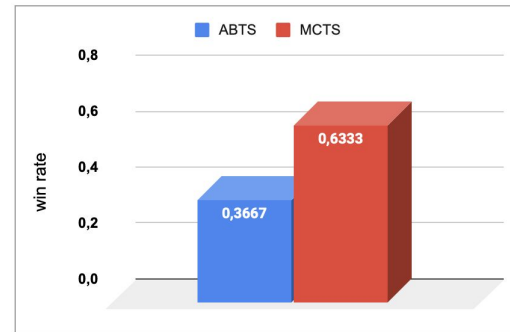
100% fairness



MCTS better  
win rate



No fairness

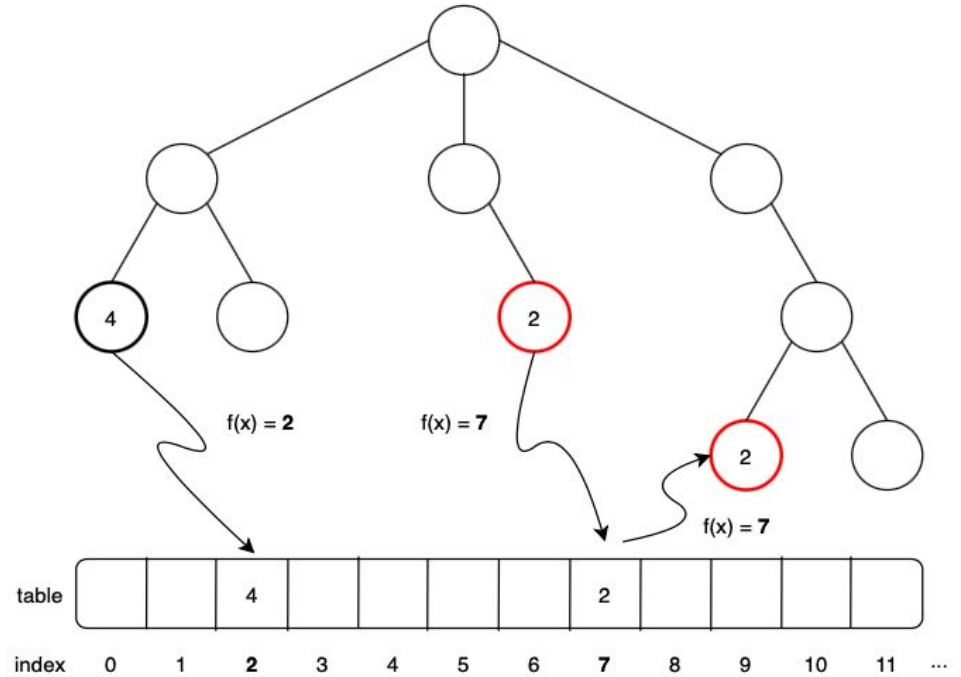


# Genetic Algorithm

- **Goal:** advanced evaluation function weights assessment
- Two optimizations
  - Applying both Rank and Tournament selection
  - “Islands” strategy for better performance
- Weights could later be used in MCTS
- Turns out a huge number of games end in a draw

# Transposition Table

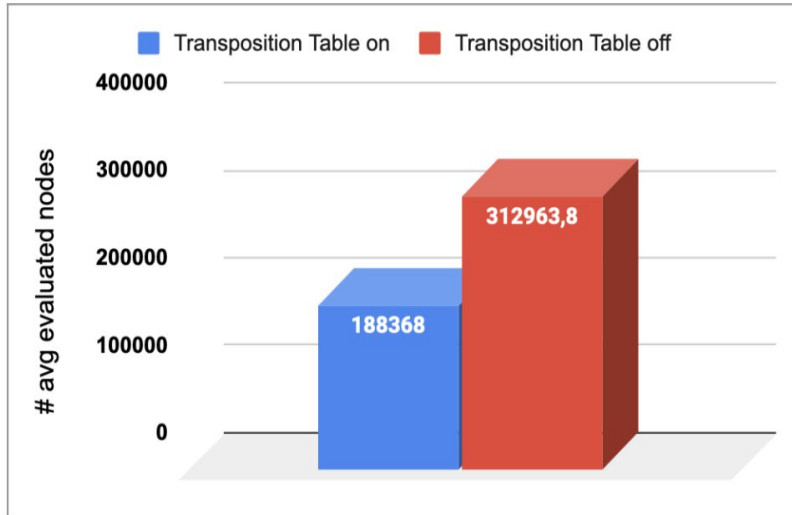
- Transposition Table = Hash Table
- Hash Function  $f(x) \rightarrow$  unique key
- Key  $\rightarrow$  index of the table
- Table contains already computed node score
- Transposed node reused previous evaluation





# → Experiment

Transposition Table ON vs OFF

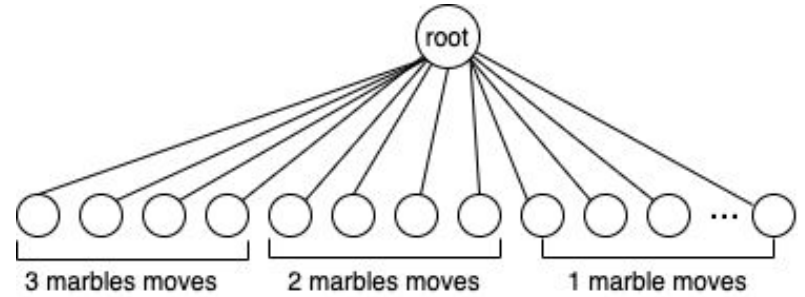


- Almost half of evaluated nodes saved
- Computational gain

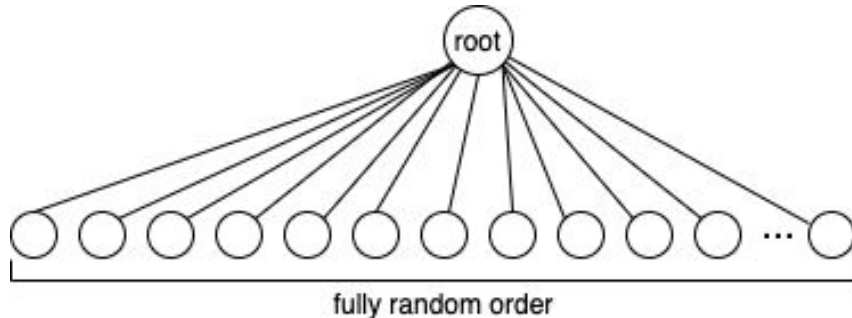
# Move ordering

- Improves alpha-beta pruning
- Better moves first
- Three different orderings tested:

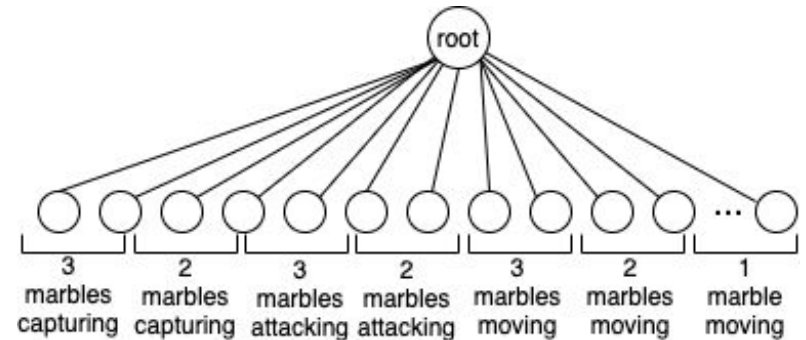
Simple ordering



Random ordering

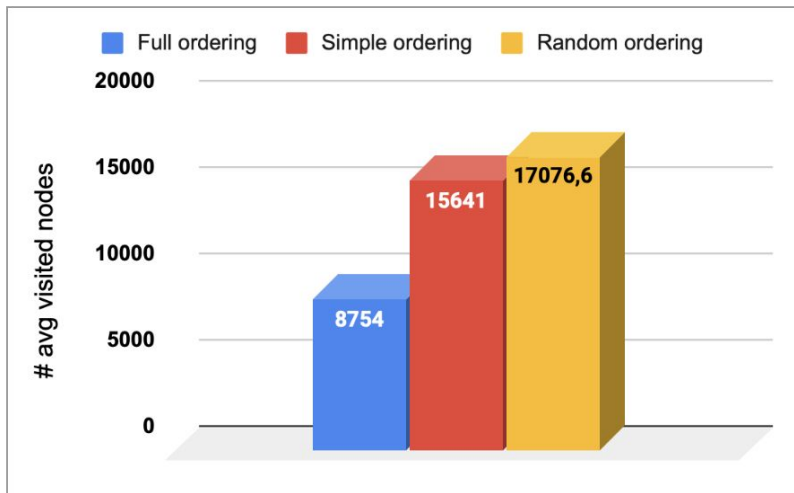


Full ordering



# → Experiment

Move ordering



- Full ordering greatly outperforms the other two orderings
- Simple ordering still outperforms random ordering.
- Prioritizing capturing and pushing is best.

# Conclusion

## Bots

- Rule Based
- Monte-Carlo Tree Search
- Alpha-Beta Tree Search

## Evaluation functions

- Neutral
- Offensive
- Defensive

## Pruning techniques

- Move ordering
- Transposition table

**THANK YOU FOR LISTENING!**

