### **Adversarial Search**

CS51A David Kauchak, Joseph C. Osborn Fall 2019

## A quick review of search

#### Problem solving via search:

- To define the state space, define three things:
  - is\_goal
  - next\_states
  - starting state

#### Uninformed search vs. informed search

- what's the difference?
- what are the techniques we've seen?
- pluses and minuses?

## Why should we study games?

Clear success criteria

Important historically for AI

Fun

#### Good application of search

hard problems (chess 35<sup>100</sup> states in search space, 10<sup>40</sup> legal states)

#### Some real-world problems fit this model

- game theory (economics)
- multi-agent problems

## Types of games

What are some of the games you've played?

## Types of games: game properties

single-player vs. 2-player vs. multiplayer

Fully observable (perfect information) vs. partially observable

Discrete vs. continuous

real-time vs. turn-based

deterministic vs. non-deterministic (chance)

For reasons previously stated, two-player games have been a focus of AI since its inception...

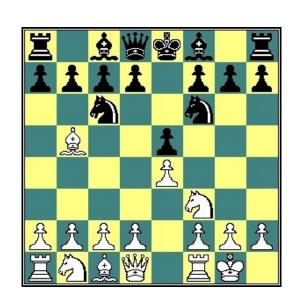


Important question: Is strategic thinking the same as intelligence?

Humans and computers have different relative strengths in these games:

humans

?



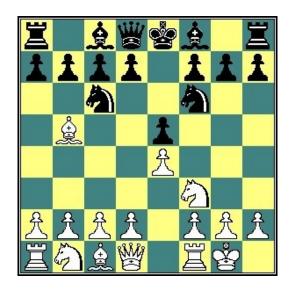
computers

?

Humans and computers have different relative strengths in these games:

#### humans

good at evaluating the strength of a board for a player



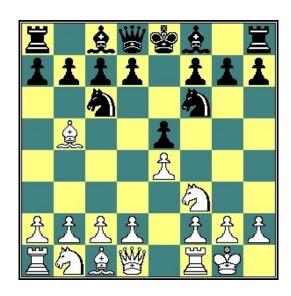
#### computers

good at looking ahead in the game to find winning combinations of moves

How could you figure out how humans approach playing chess?

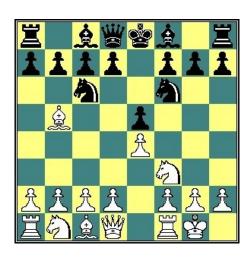
humans

good at evaluating the strength of a board for a player



## How humans play games...

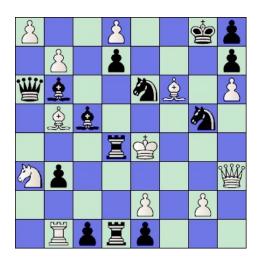
An experiment was performed in which chess positions were shown to novice and expert players...



- experts could reconstruct these perfectly
  - novice players did far worse...

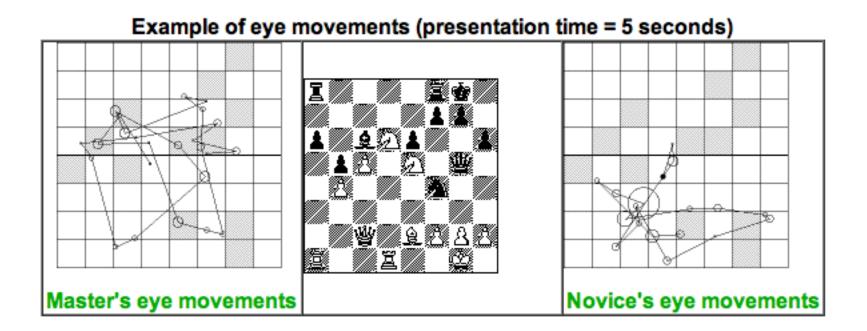
## How humans play games...

<u>Random</u> chess positions (not legal ones) were then shown to the two groups

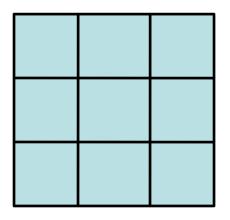


experts and novices did just as badly at reconstructing them!

### People are still working on this problem...

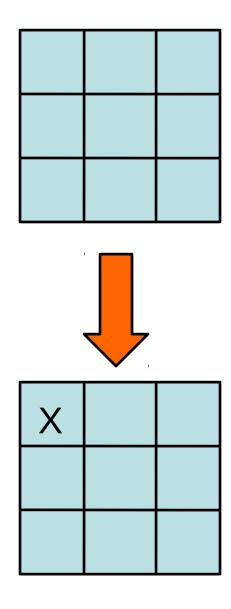


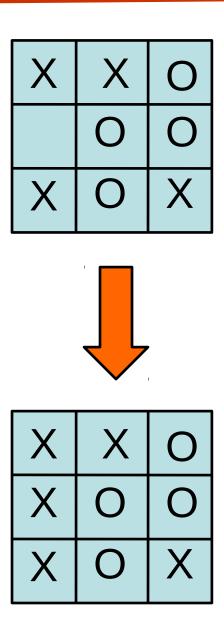
http://people.brunel.ac.uk/~hsstffg/frg-research/chess\_expertise/

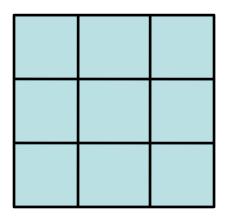


If we want to write a program to play tic tac toe, what question are we trying to answer?

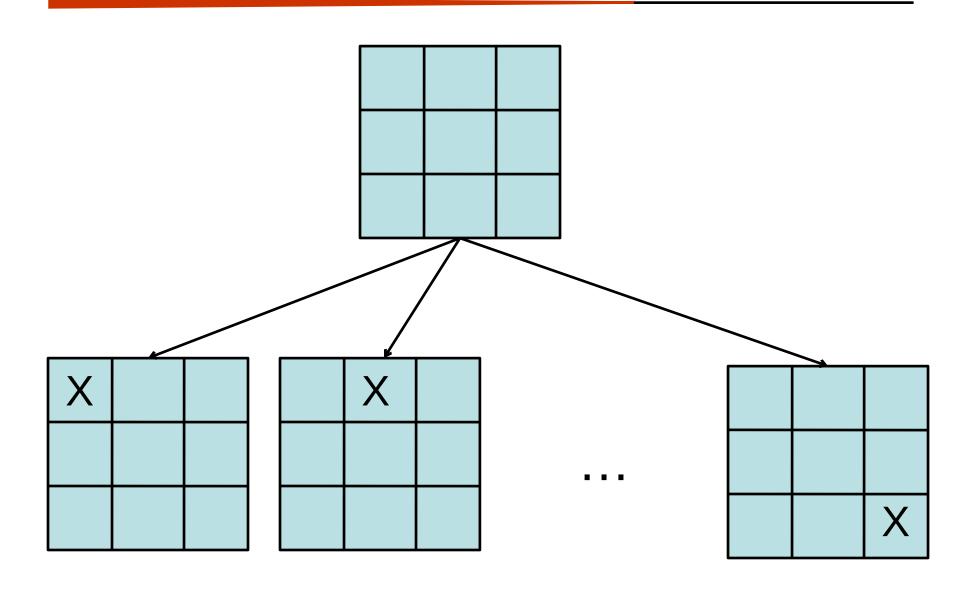
Given a state (i.e. board configuration), what move should we make!

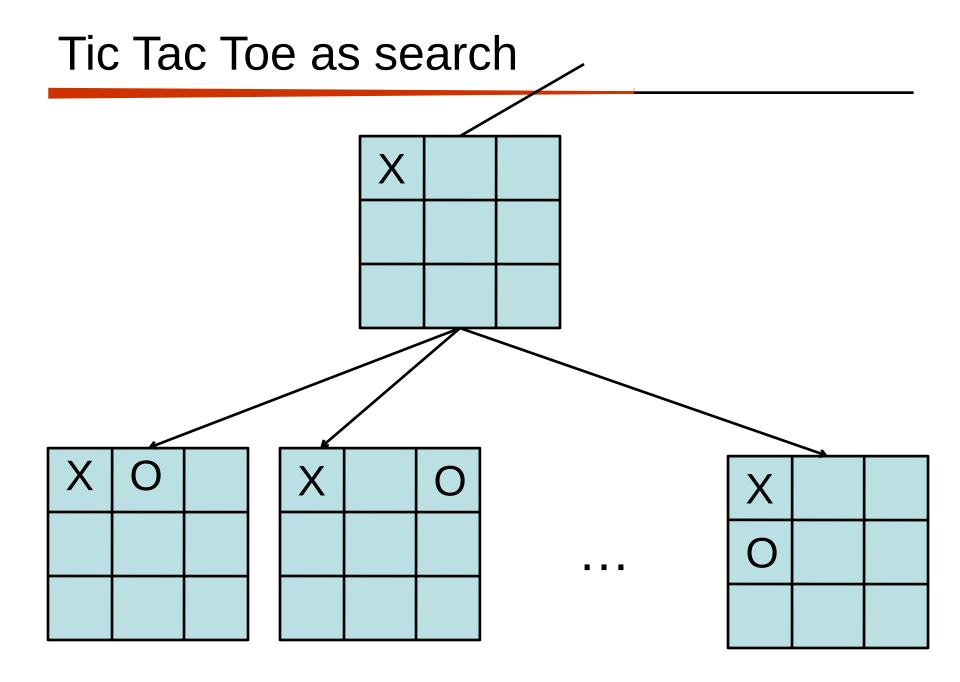






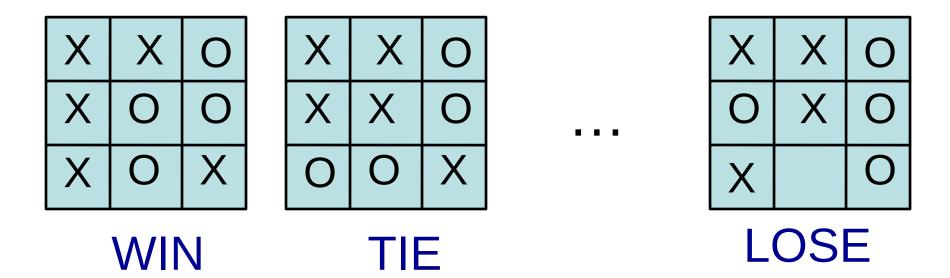
How can we pose this as a search problem?





. . .

#### Eventually, we'll get to a leaf



How does this help us?

Try and make moves that move us towards a win, i.e. where there are leaves with a WIN.

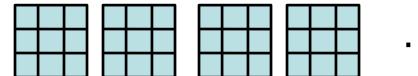
### Tic Tac Toe

X's turn

O's turn

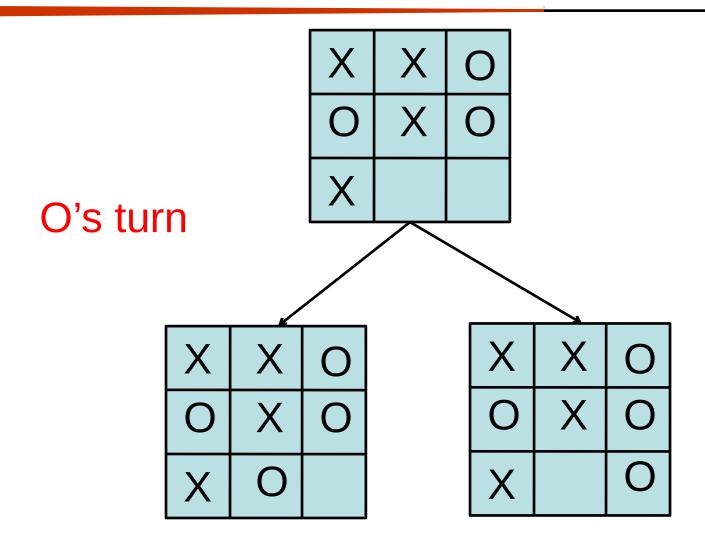
X's turn

Problem: we don't know what O will do





## I'm X, what will 'O' do?

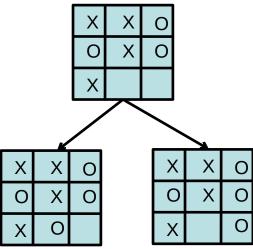


## Minimizing risk

The computer doesn't know what move O (the opponent) will make

It can assume that it will try and make the best move possible

Even if O actually makes a different move, we're no worse off. Why?



## **Optimal Strategy**

An Optimal Strategy is one that is at least as good as any other, no matter what the opponent does

- If there's a way to force the win, it will

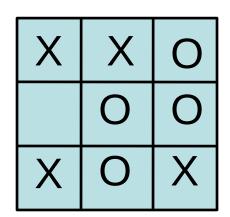
Will only lose if there's no other option

| X   | X | 0 |  | X | X   | 0 |   |   | X    | X | 0 |  |
|-----|---|---|--|---|-----|---|---|---|------|---|---|--|
| X   | 0 | 0 |  | X | X   | 0 |   |   | 0    | X | О |  |
| X   | 0 | X |  | 0 | 0   | X |   |   | X    |   | 0 |  |
| WIN |   |   |  |   | TIE |   | • | , | LOSE |   |   |  |
| +1  |   |   |  | 0 |     |   |   |   | -1   |   |   |  |

#### Idea:

- define a function that gives us a "score" for how good each state is
- higher scores mean better

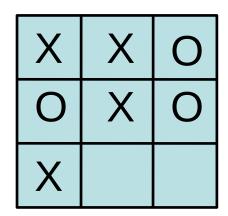
Our (X) turn



What should be the score of this state?

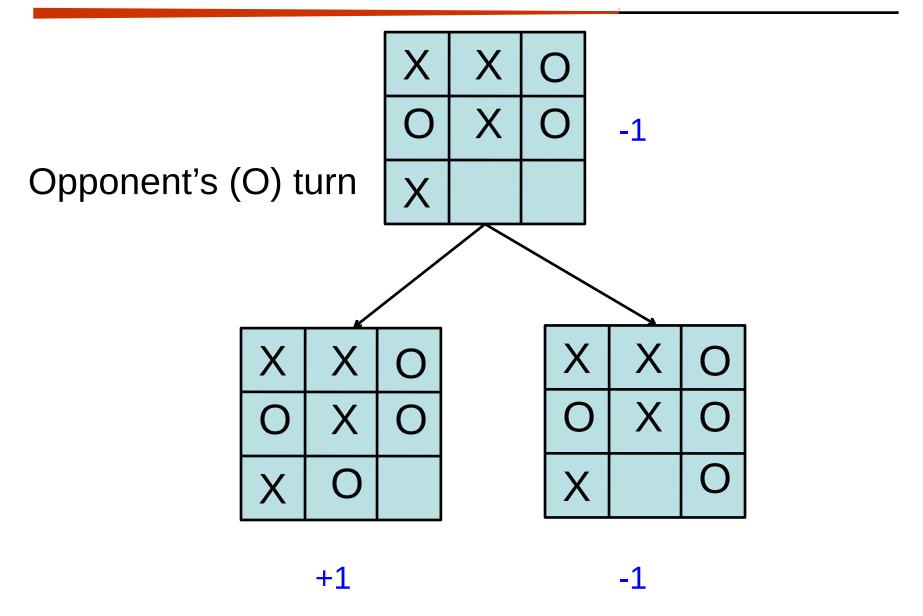
+1: we can get to a win

Opponent's (O) turn

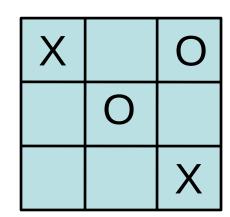


What should be the score of this state?

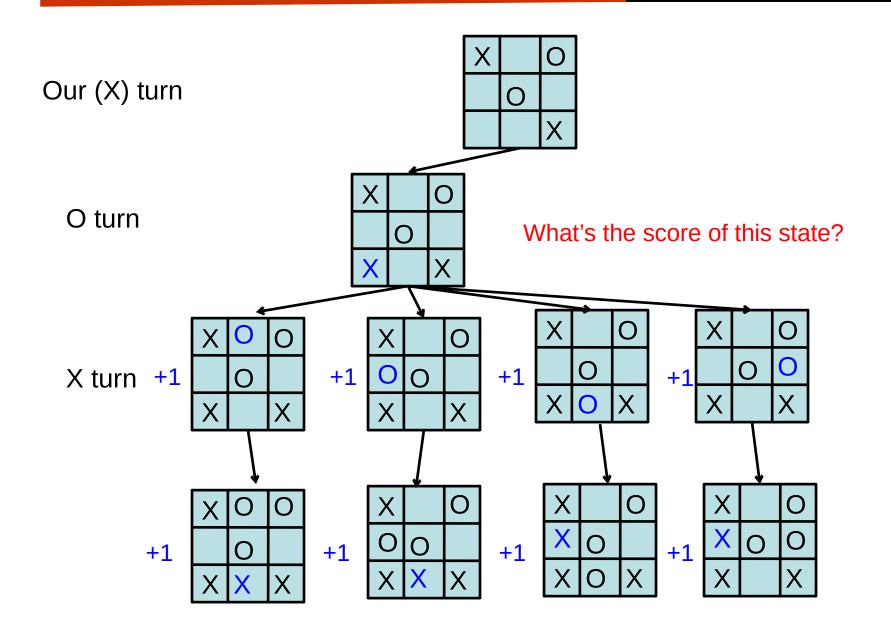
-1: we can get to a win

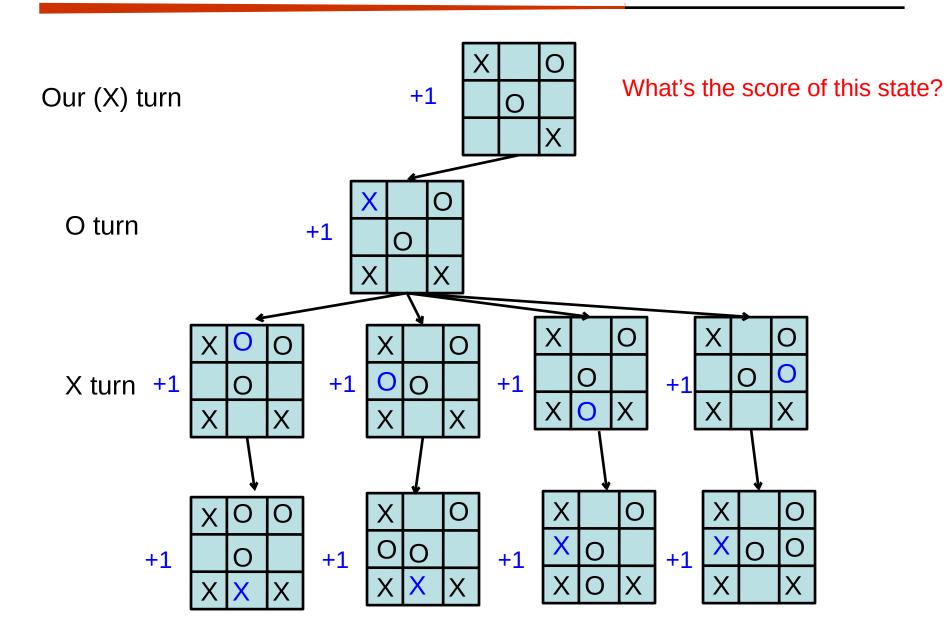


Our (X) turn

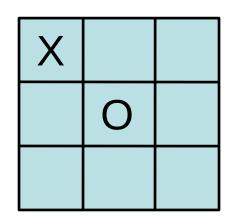


What should be the score of this state?





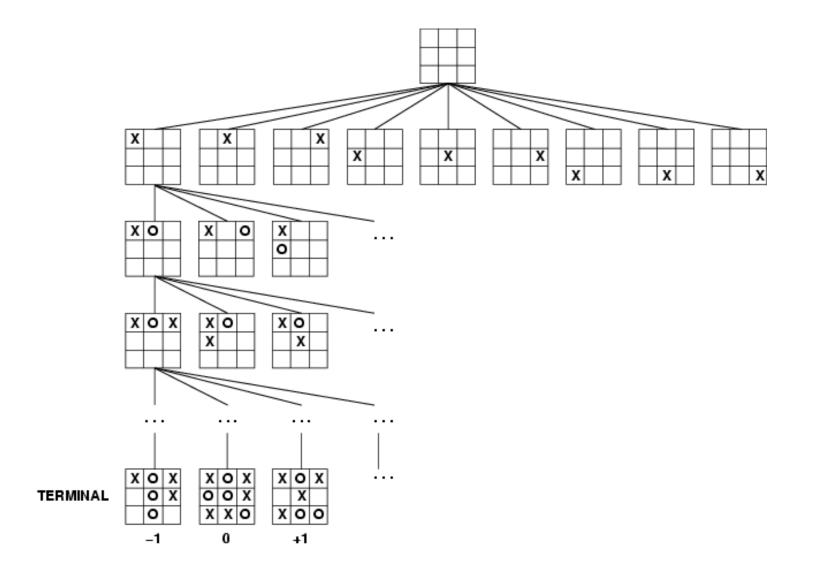
Our (X) turn



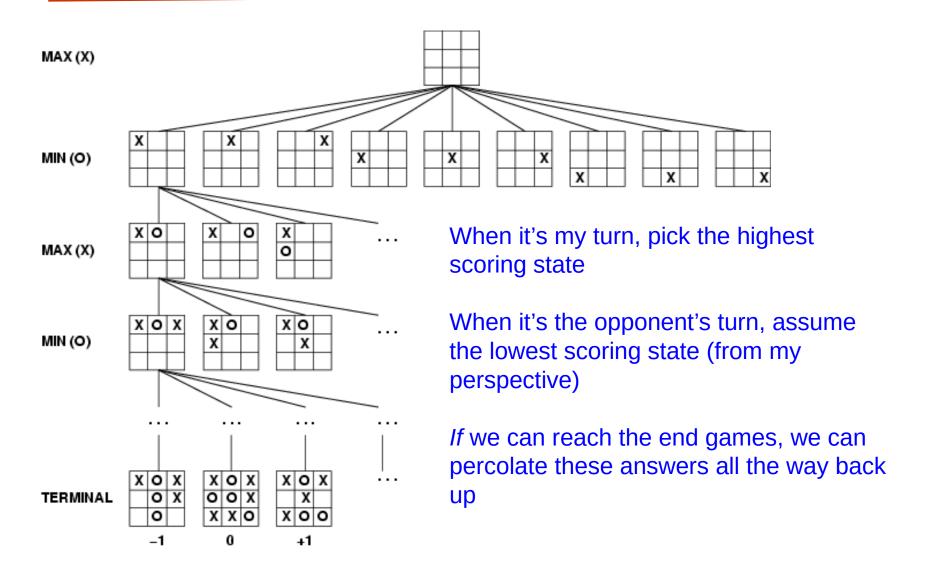
What should be the score of this state?

0: If we play perfectly and so does O, the best we can do is a tie (could do better if O makes a mistake)

# How can X play optimally?



## How can X play optimally?

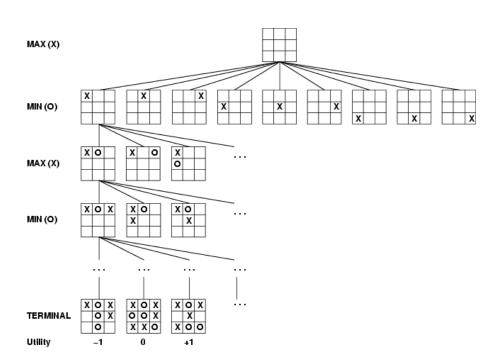


## How can X play optimally?

Start from the leaves and propagate the score up:

- if X's turn, pick the move that maximizes the utility
- if O's turn, pick the move that minimizes the utility

Is this optimal?



## Minimax Algorithm: An Optimal Strategy

```
minimax(state) =
   if state is a terminal state
      score(state)
   else if MY turn
      over all next states, s: return the maximum of minimax(s)
   else if OPPONENTS turn
      over all next states, s: return the minimum of minimax(s)
```

Uses recursion to compute the "value" of each state

Searches down to the leaves, then the values are "backed up" through the tree as the recursion finishes

What type of search is this?

What does this assume about how MIN will play? What if this isn't true?

### Nim

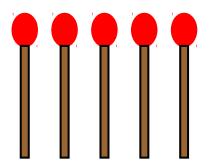
K piles of coins

On your turn you must take one or more coins from one pile

Player that takes the last coin wins

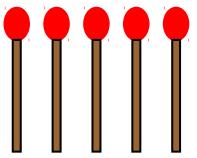
Example:

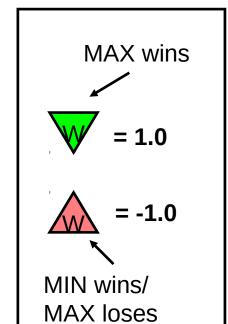
https://www.goobix.com/games/nim/

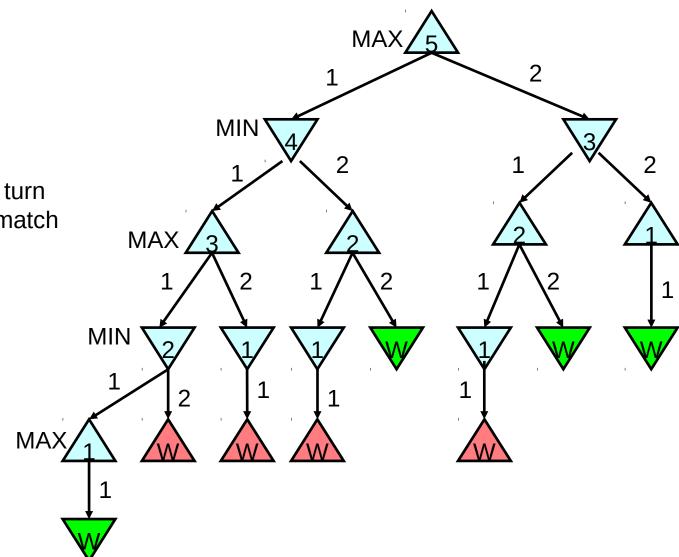


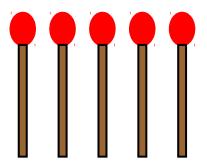
Take 1 or 2 at each turn Goal: take the last match

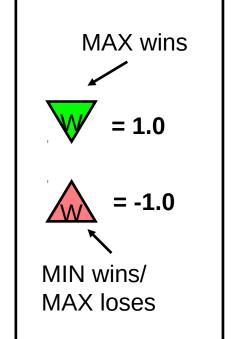
What move should I take?

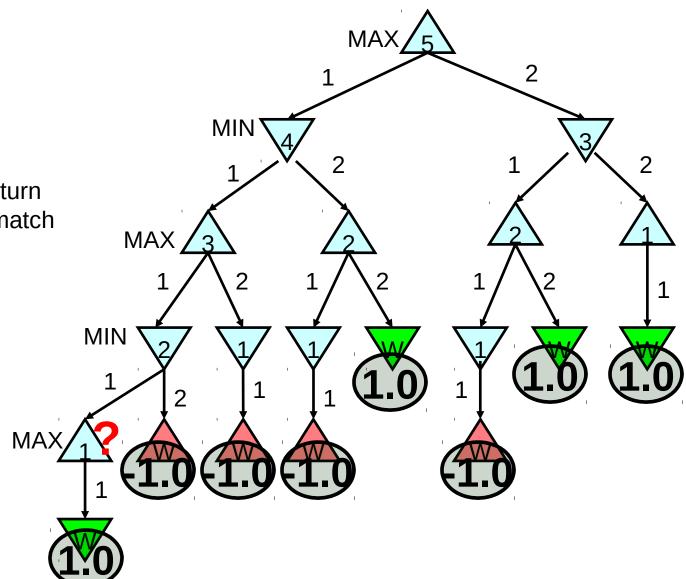


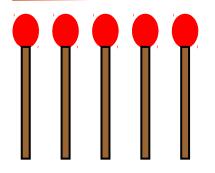


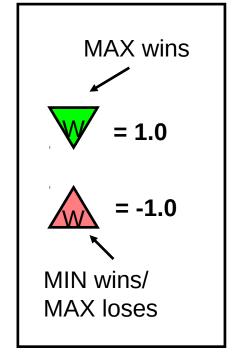


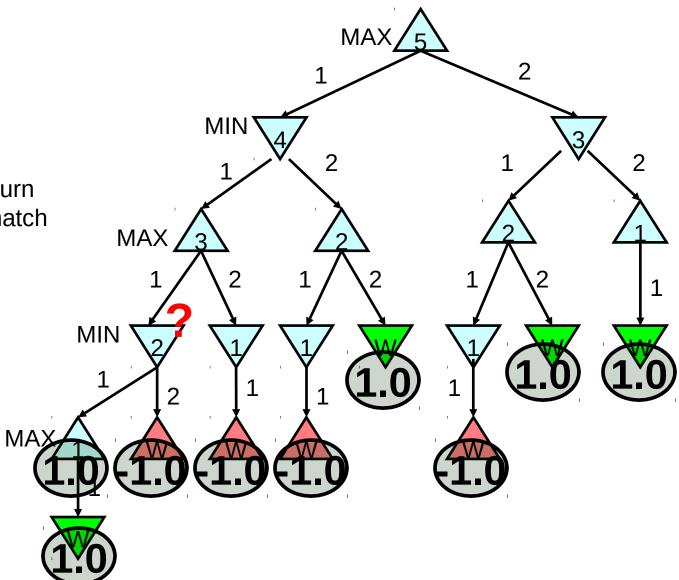


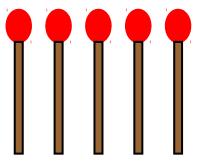


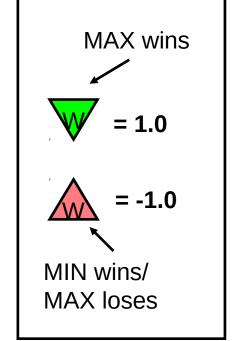


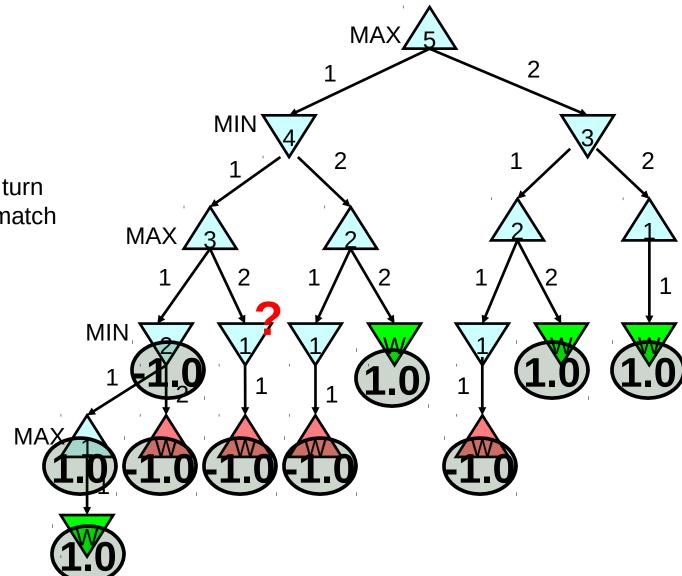


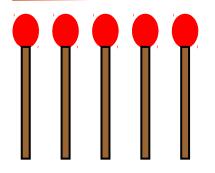


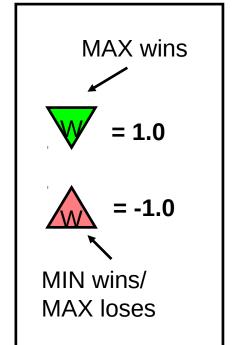


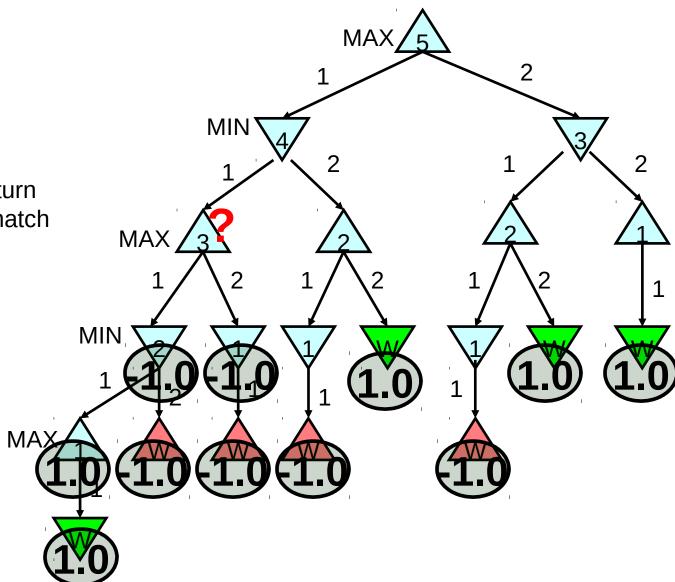


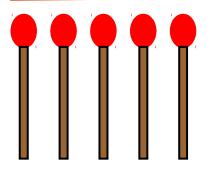


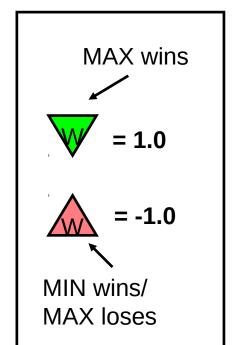


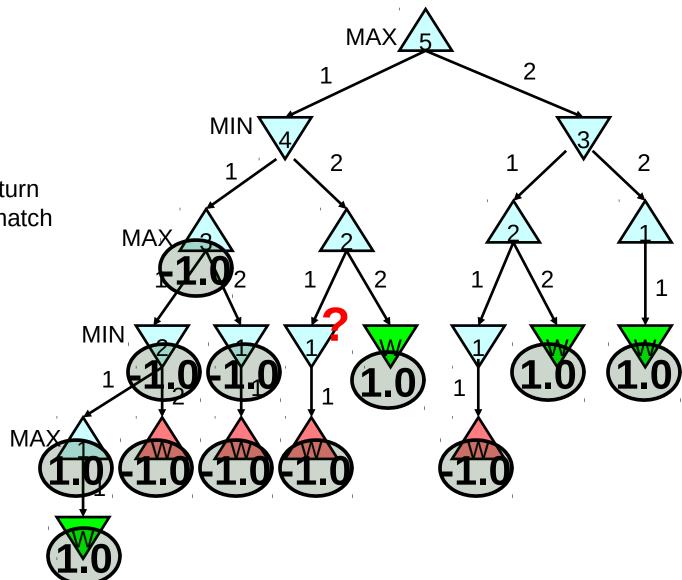


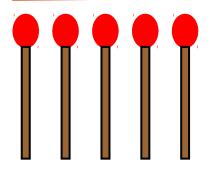


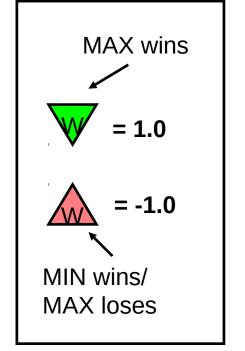


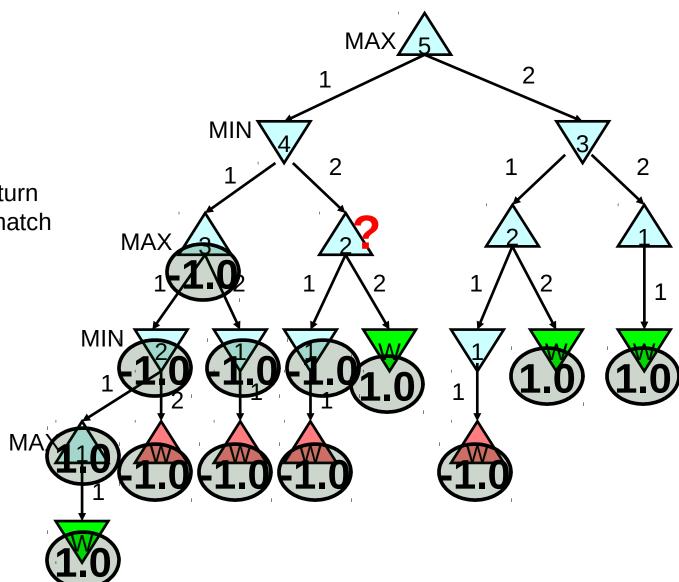


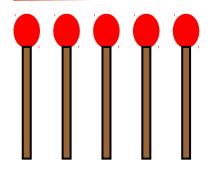


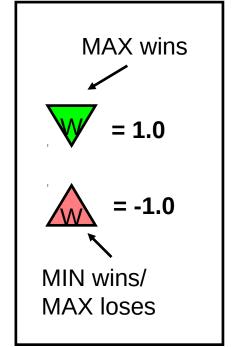


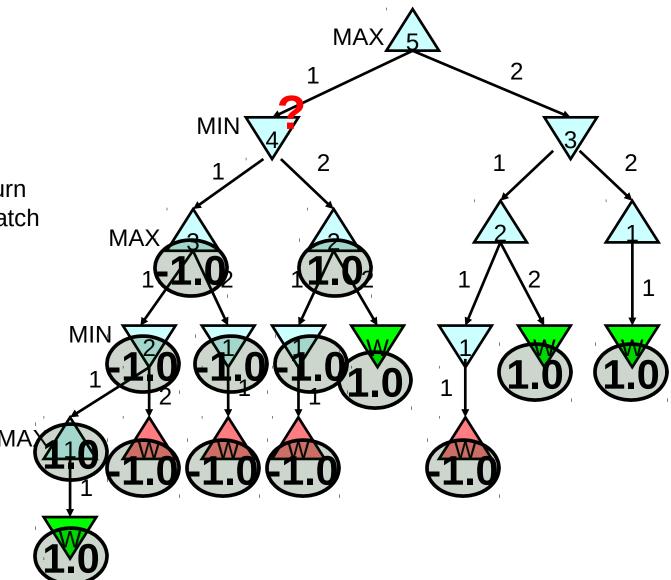


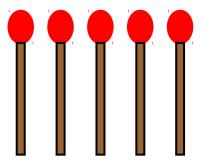


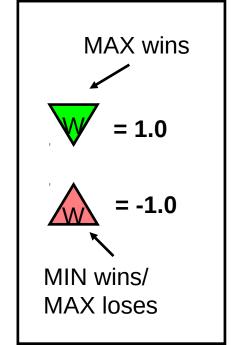


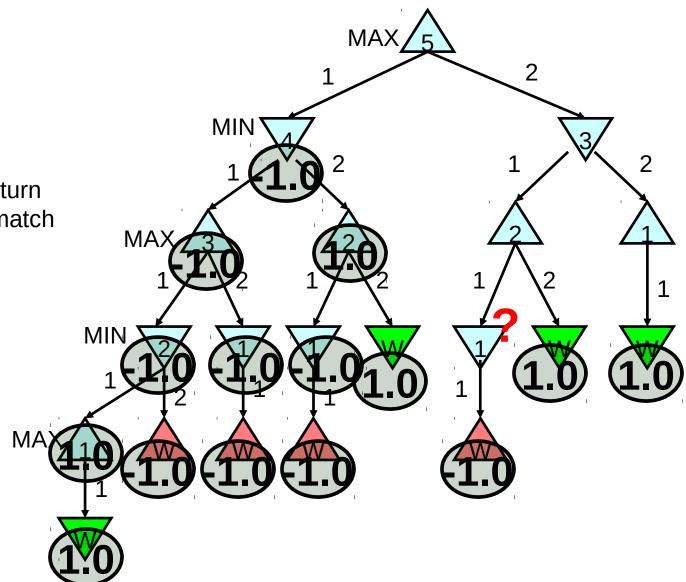


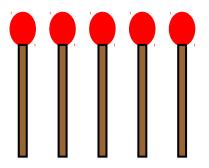


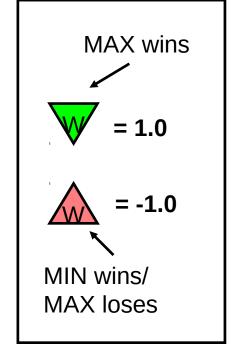


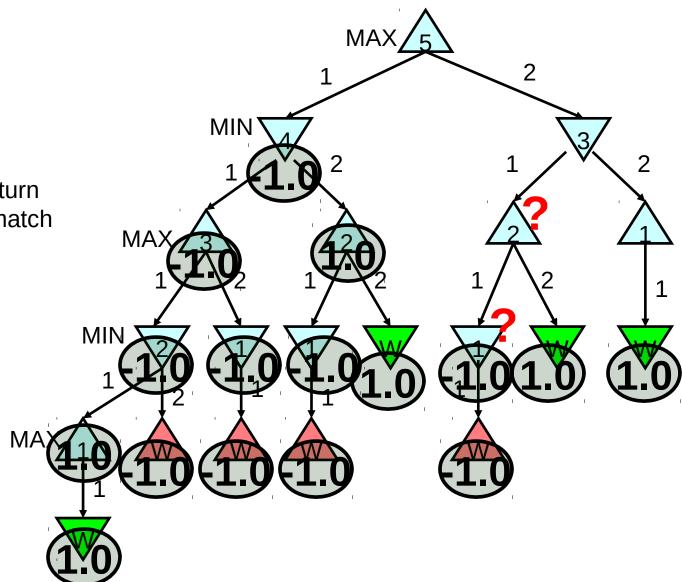


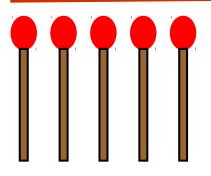


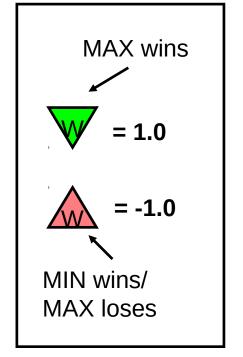


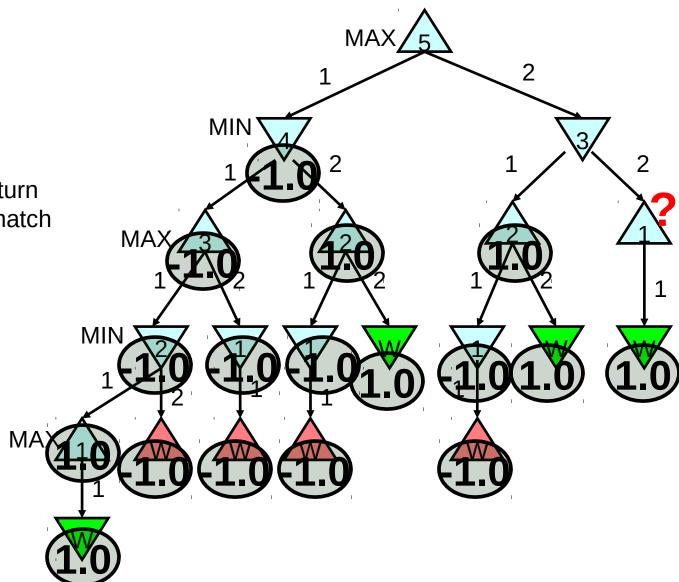


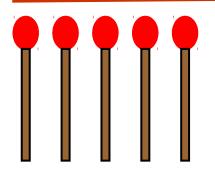


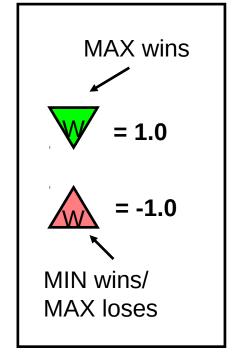


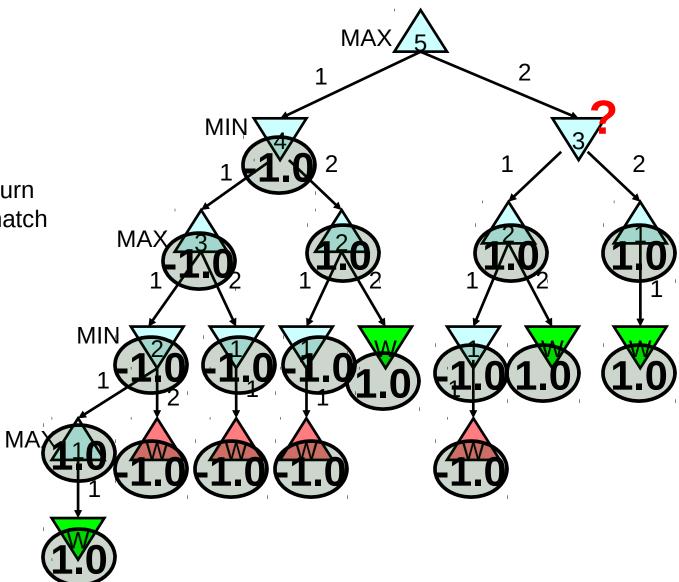


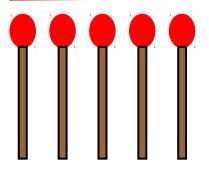


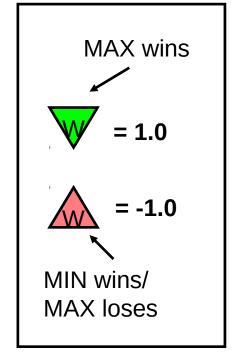


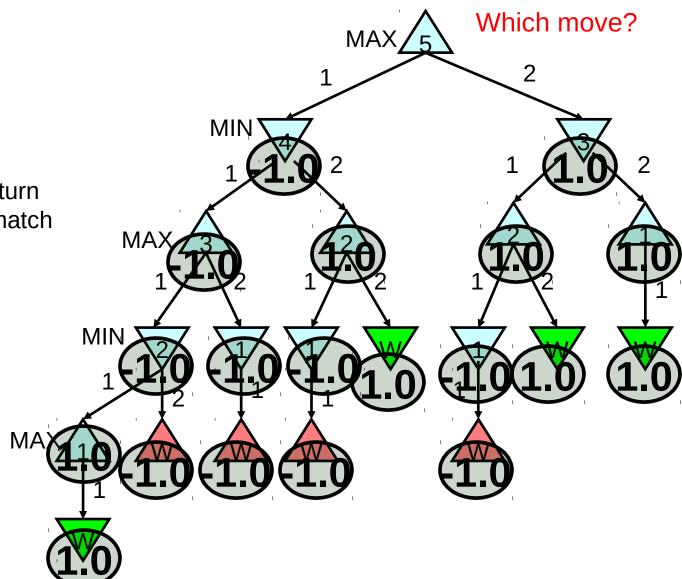


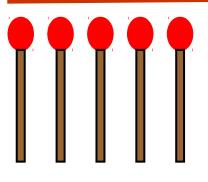


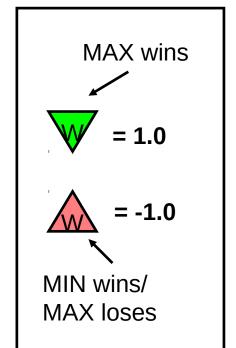


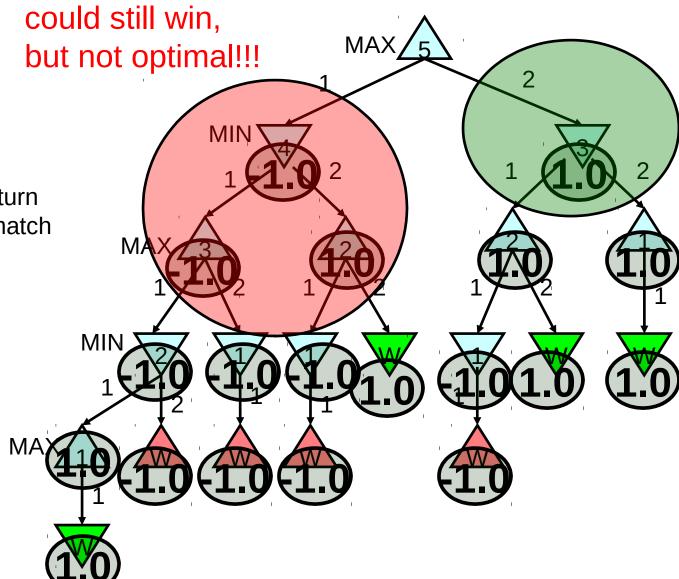




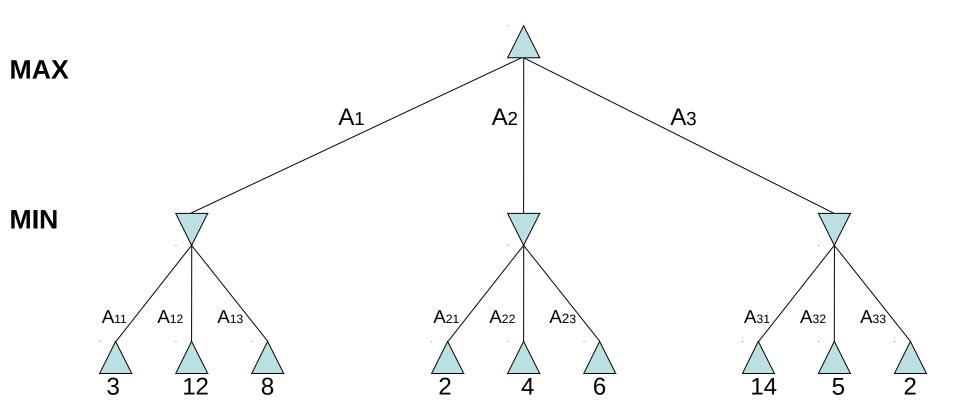






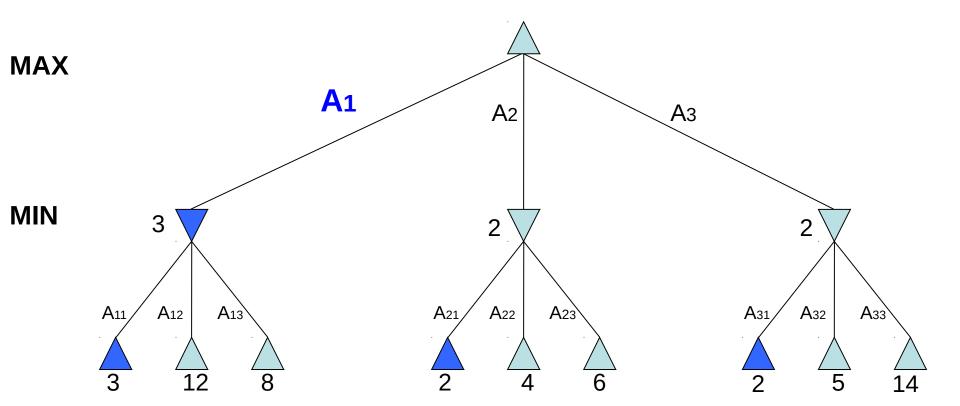


### Minimax example 2



Which move should be made:  $A_1$ ,  $A_2$  or  $A_3$ ?

## Minimax example 2



### Properties of minimax

Minimax is optimal!

Are we done?



On average, there are ~35 possible moves that a chess player

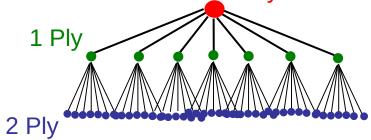
can make from any board configuration...



18 Ply!!

17005

IBM's Deep Blue

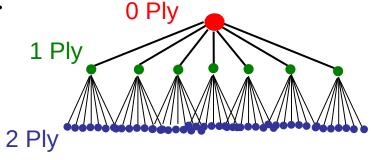


Branching Factor Estimates for different two-player games

0 Ply

| Tic-tac-toe  | 4   |
|--------------|-----|
| Connect Four | 7   |
| Checkers     | 10  |
| Othello      | 30  |
| Chess        | 35  |
| Go           | 300 |

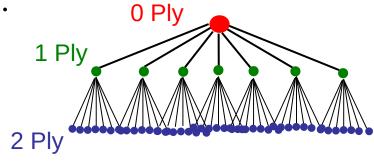
On average, there are ~35 possible moves that a chess player can make from any board configuration...

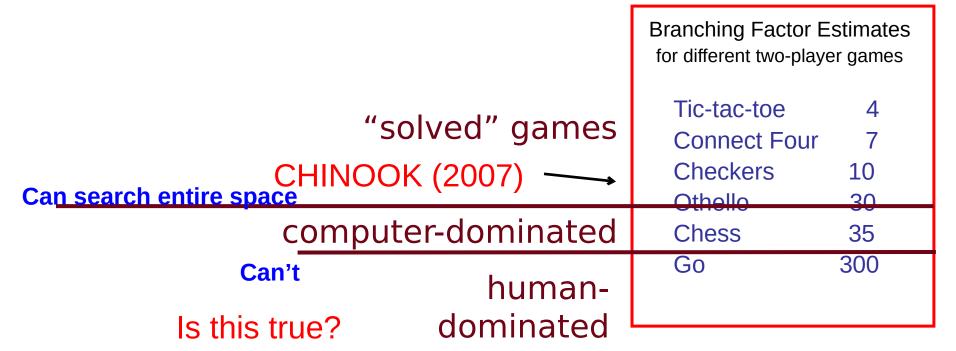


Boundaries for qualitatively different games...

| Branching Factor Estimates for different two-player games |     |
|---|-----|
| Tic-tac-toe   | 4   |
| Connect Four  | 7   |
| Checkers  | 10  |
| Othello   | 30  |
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| Go  | 300 |
|   |     |

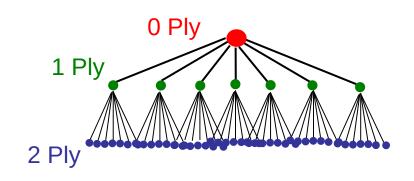
On average, there are ~35 possible moves that a chess player can make from any board configuration...





AlphaGo (created by Google), in April 2016 beat one of the best Go players:

http://www.nytimes.com/2016/04/05/science/google-alphago-artificial-intelligence.html



|  | Branching Factor Estimates for different two-player games |  |
|--|---|--|
| "solved" games                         | Tic-tac-toe 4 Connect Four 7                              |  |
| Can search entire space CHINOOK (2007) | Checkers 10   |  |
| computer-dominated                     | Chess 35<br>Go 300  |  |
| Can't What do we do?                   |   |  |

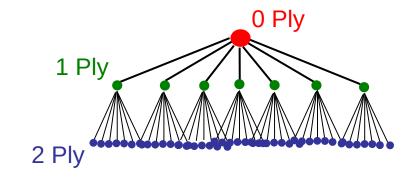
#### Alpha-Beta pruning

#### An optimal pruning strategy

- only prunes paths that are suboptimal (i.e. wouldn't be chosen by an optimal playing player)
- returns the *same* result as minimax, but faster

Pruning helps get a bit deeper

For many games, still can't search the entire tree



Now what?

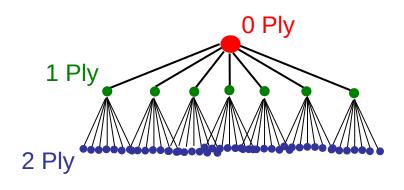
Branching Factor Estimates for different two-player games

Tic-tac-toe 4
Connect Four 7
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Chess 35
Go 300

computer-dominated

Pruning helps get a bit deeper

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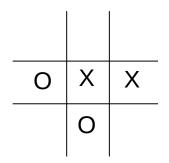
#### Go as deep as you can:

- estimate the score/quality of the state (called an evaluation function)
- use that instead of the real score

computer-dominated

| Branching Factor Estimates for different two-player games |     |  |
|---|-----|--|
| Tic-tac-toe   | 4   |  |
| Connect Four  | 7   |  |
| Checkers  | 10  |  |
| Othello   | 30  |  |
| Chess   | 35  |  |
| Go  | 300 |  |
|   |     |  |

#### Tic Tac Toe evaluation functions



Ideas?

#### Example Tic Tac Toe EVAL

#### Tic Tac Toe

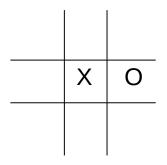
Assume MAX is using "X"

if state is win for MAX:

 $+\infty$ 

if state is win for MIN:

**-** ∞



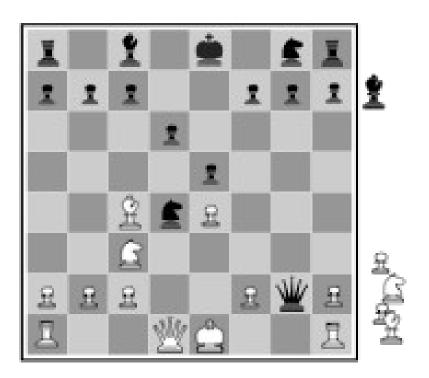
$$= 6 - 4 = 2$$

#### else:

(number of rows, columns and diagonals available to MAX) -(number of rows, columns and diagonals available to MIN)

$$= 4 - 3 = 1$$

#### Chess evaluation functions

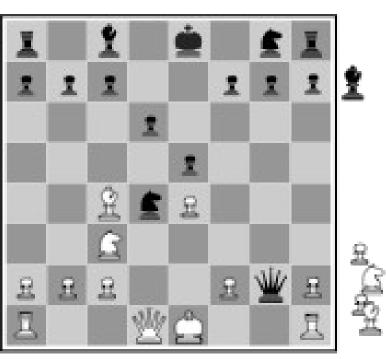


Ideas?

Assume each piece has the following value

```
pawn = 1;
knight = 3;
bishop = 3;
rook = 5;
queen = 9;
```

EVAL(state) =
 sum of the value of white pieces sum of the value of black pieces

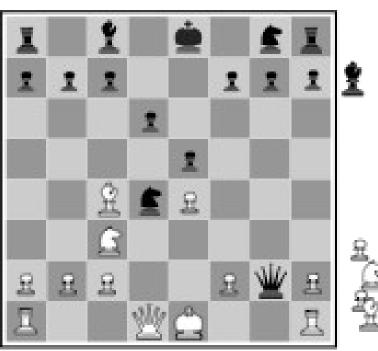


$$= 31 - 36 = -5$$

Assume each piece has the following value

```
pawn = 1;
knight = 3;
bishop = 3;
rook = 5;
queen = 9;
```

EVAL(state) =
 sum of the value of white pieces sum of the value of black pieces



Any problems with this?

#### Ignores actual positions!

Actual heuristic functions are often a weighted combination of features

$$EVAL(s) = W_1 f_1(s) + W_2 f_2(s) + W_3 f_3(s) + ...$$





A feature can be any numerical information about the board

- as general as the number of pawns
- to specific board configurations

Deep Blue: 8000 features!

#### history/end-game tables

#### History

- keep track of the quality of moves from previous games
- use these instead of search

#### end-game tables

- do a reverse search of certain game configurations, for example all board configurations with king, rook and king
- tells you what to do in any configuration meeting this criterion
- if you ever see one of these during search, you lookup exactly what to do

#### end-game tables

#### Devastatingly good

#### Allows much deeper branching

- for example, if the end-game table encodes a 20-move finish and we can search up to 14
- can search up to depth 34

#### Stiller (1996) explored all end-games with 5 pieces

– one case check-mate required 262 moves!

#### Knoval (2006) explored all end-games with 6 pieces

– one case check-mate required 517 moves!

Traditional rules of chess require a capture or pawn move within 50 or it's a stalemate

#### Opening moves

At the very beginning, we're the farthest possible from any goal state

People are good with opening moves

Tons of books, etc. on opening moves

Most chess programs use a database of opening moves rather than search

#### Nim

K piles of coins

On your turn you must take one or more coins from one pile

Player that takes the last coin wins

Example:

https://www.goobix.com/games/nim/