Adversarial Search

CS51A Joseph C. Osborn Spring 2020

: Material borrowed from David Kauchak, Sara Owsley Sood and others

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 35¹⁰⁰ states in search space, 10⁴⁰ legal states)

Some real-world problems fit this model

- game theory (economics)
- multi-agent problems

What are some of the games you've played?

Types of games: game properties

single-player vs. 2-player vs. multiplayer

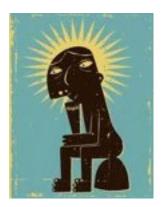
Fully observable (perfect information) vs. partially observable

Discrete vs. continuous

real-time vs. turn-based

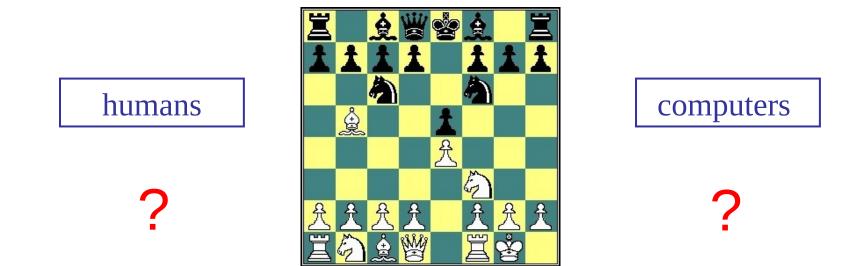
deterministic vs. non-deterministic (chance)

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?

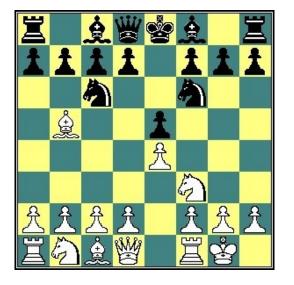
Humans and computers have different relative strengths in these games:



Humans and computers have different relative strengths in these games:



good at evaluating the strength of a board for a player





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

How could you figure out how humans approach playing chess?

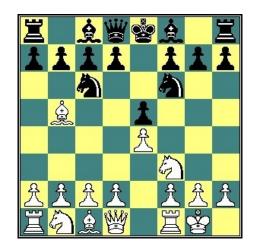


good at evaluating the strength of a board for a player



How humans play games...

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

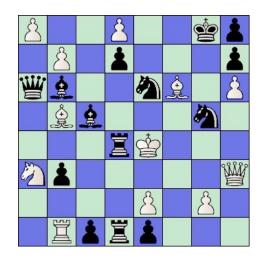


- experts could reconstruct these perfectly

- novice players did far worse...

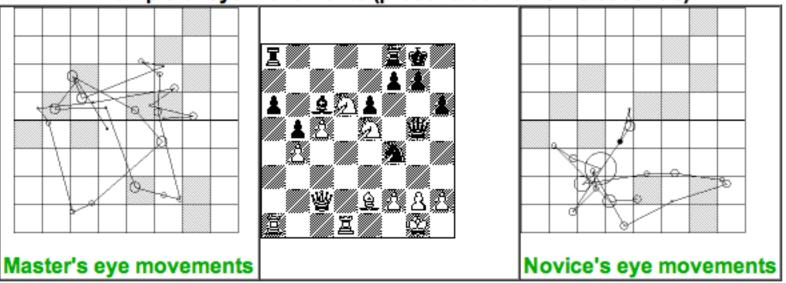
How humans play games...

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



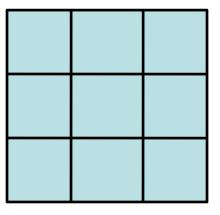
experts and novices did just as badly at reconstructing them!

People are still working on this problem...



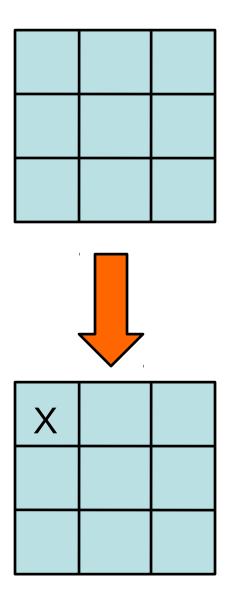
Example of eye movements (presentation time = 5 seconds)

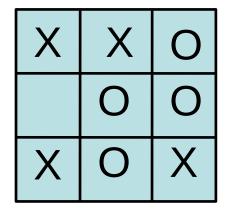
http://people.brunel.ac.uk/~hsstffg/frg-research/chess_expertise/

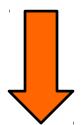


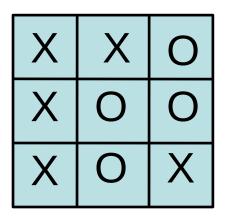
If we want to write a program to play tic tac toe, what question are we trying to answer?

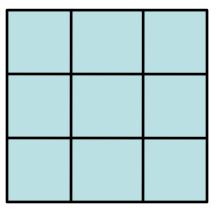
Given a state (i.e. board configuration), what move should we make!



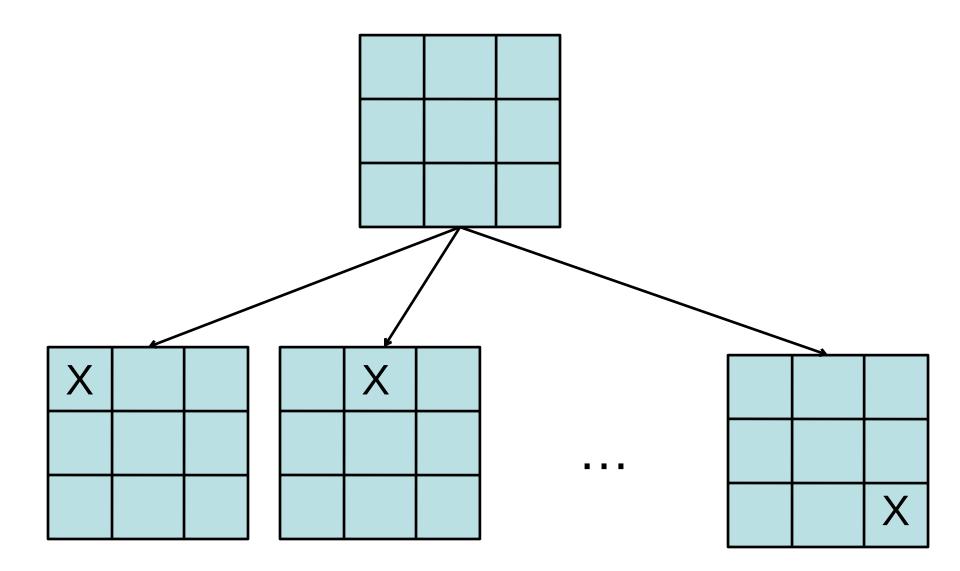


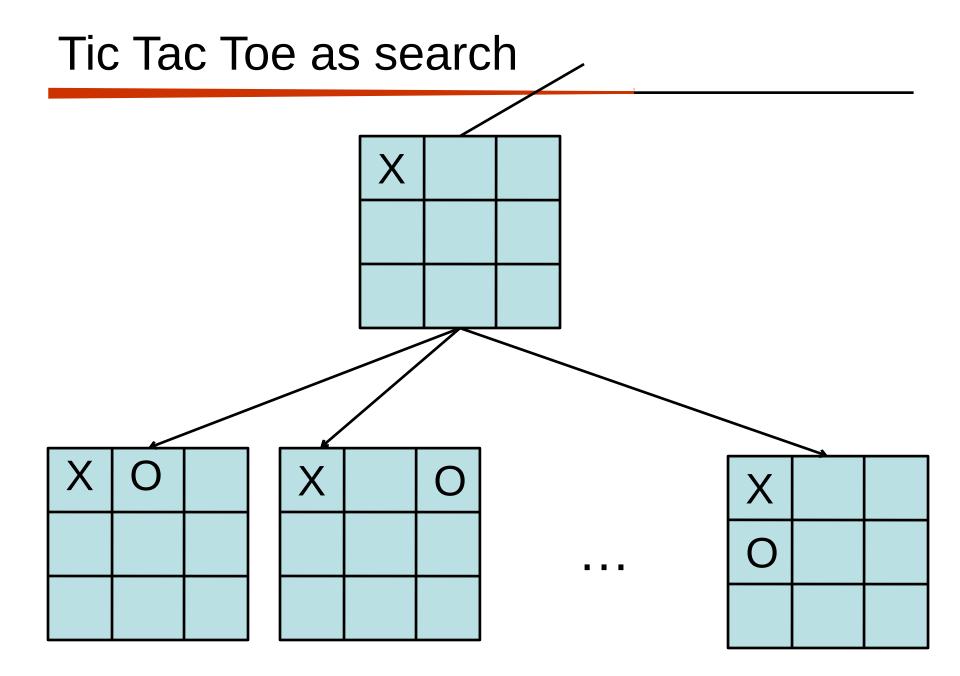




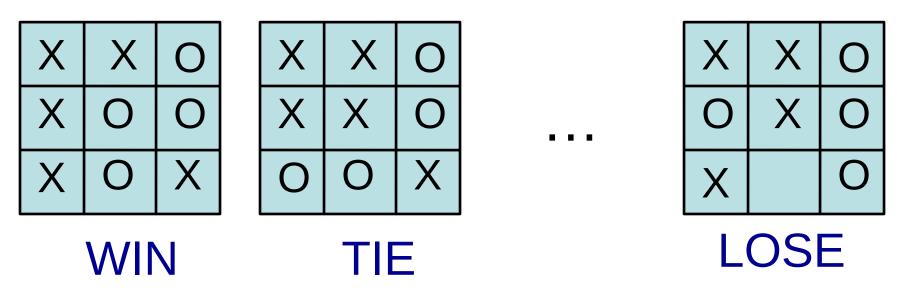


How can we pose this as a search problem?





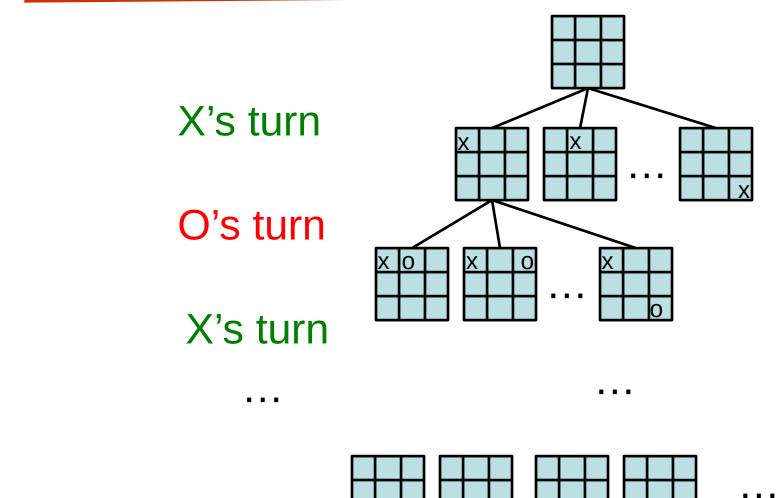
Eventually, we'll get to a leaf



How does this help us?

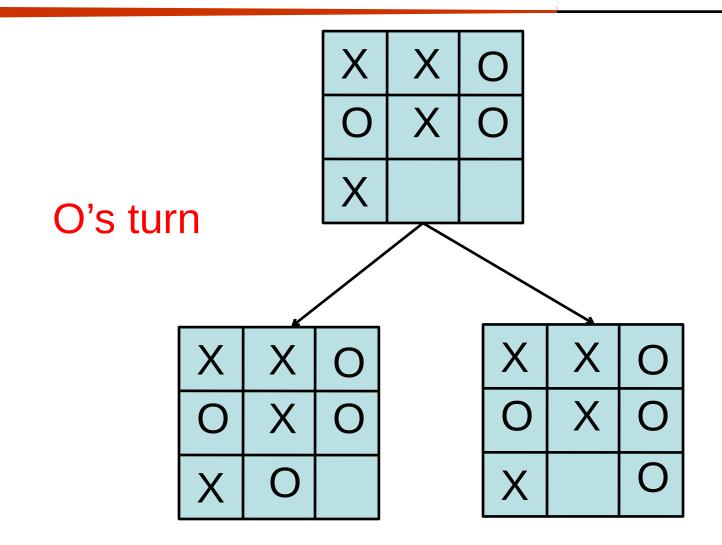
Try and make moves that move us towards a win, i.e. where there are leaves with a WIN.

Tic Tac Toe



Problem: we don't know what O will do

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

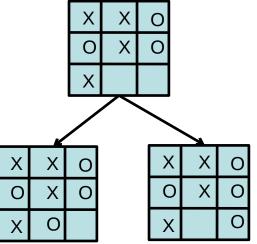


Minimizing risk

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

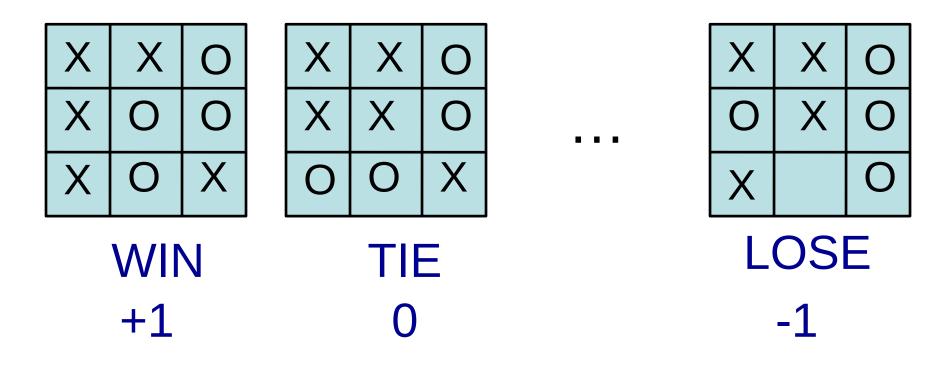
Even if O actually makes a different move, we're no worse off. Why?



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

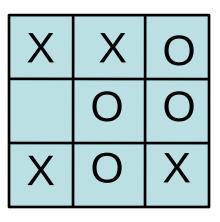
- Will only lose if there's no other option



Idea:

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

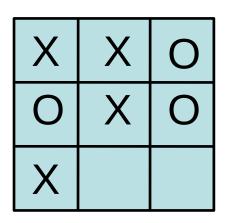
Our (X) turn



What should be the score of this state?

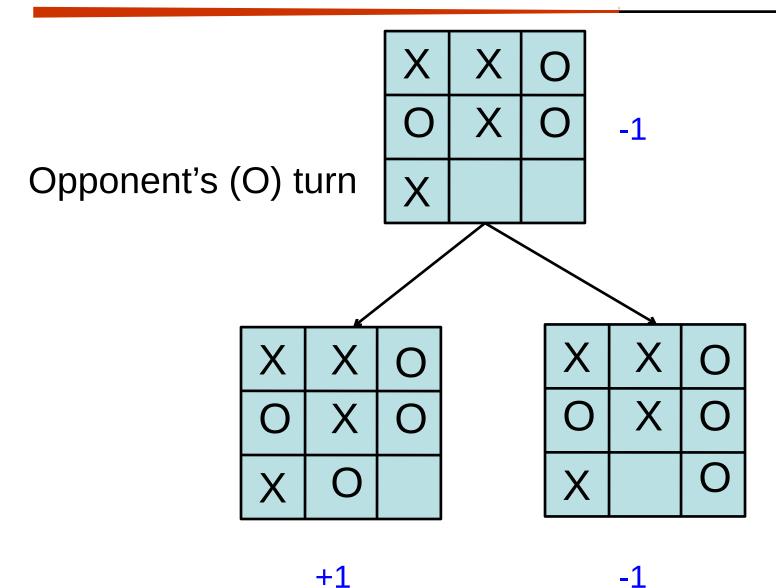
+1: we can get to a win

Opponent's (O) turn

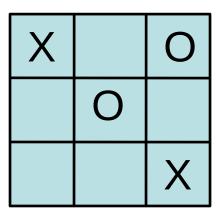


What should be the score of this state?

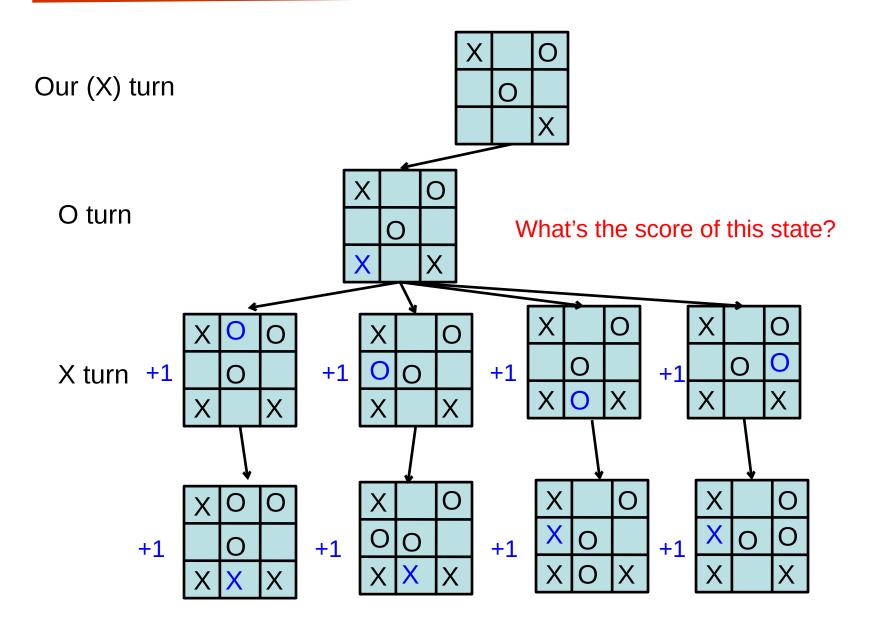
-1: we can get to a win

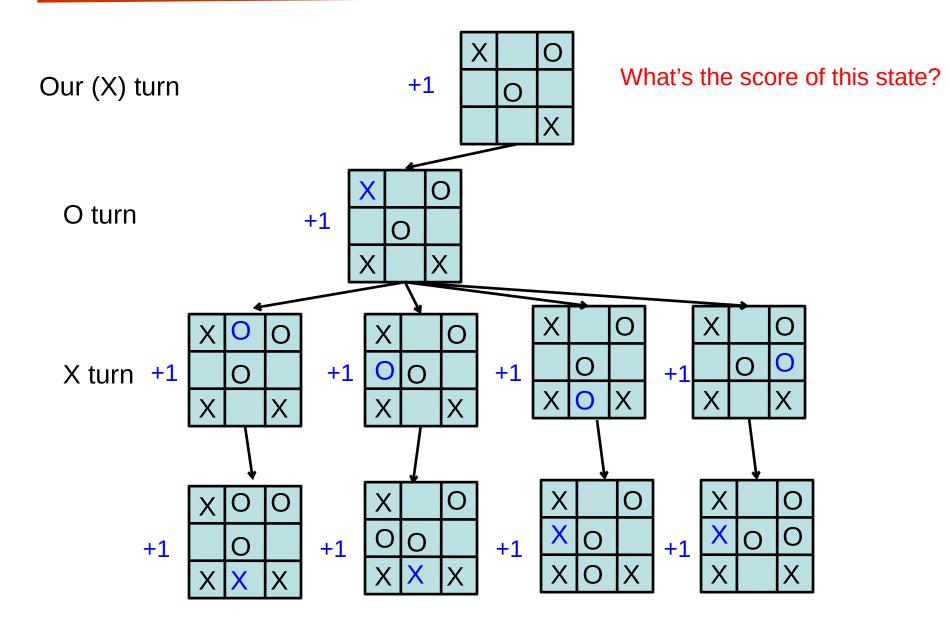


Our (X) turn

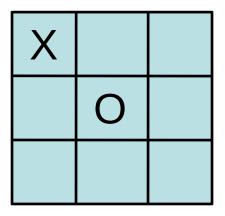


What should be the score of this state?





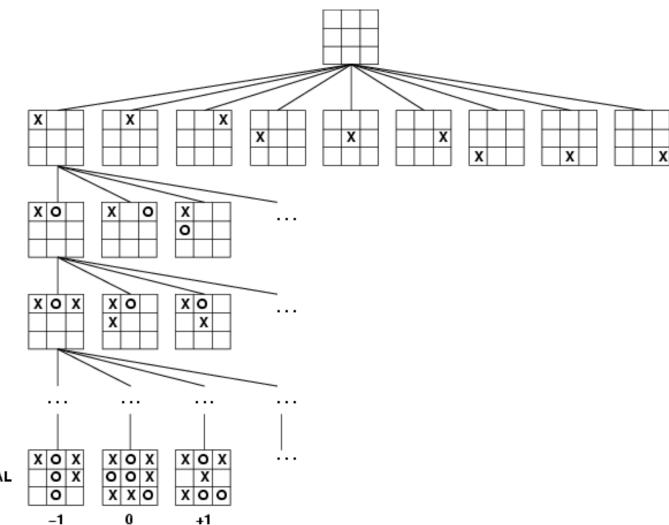
Our (X) turn



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)

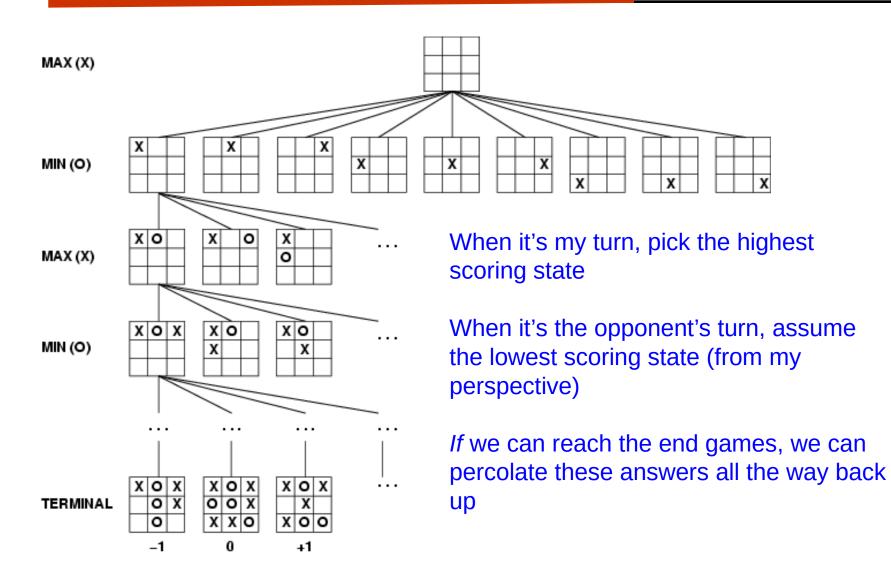
How can X play optimally?



TERMINAL

l

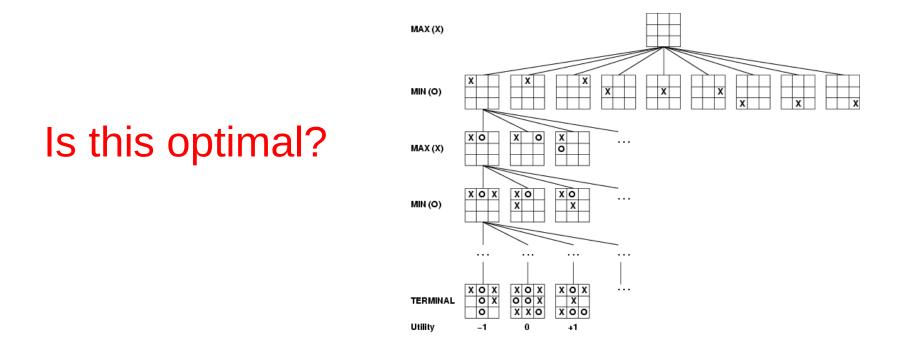
How can X play optimally?



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



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?

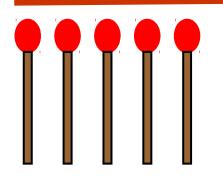
What does this assume about how MIN will play? What if this isn't true?

K piles of coins

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

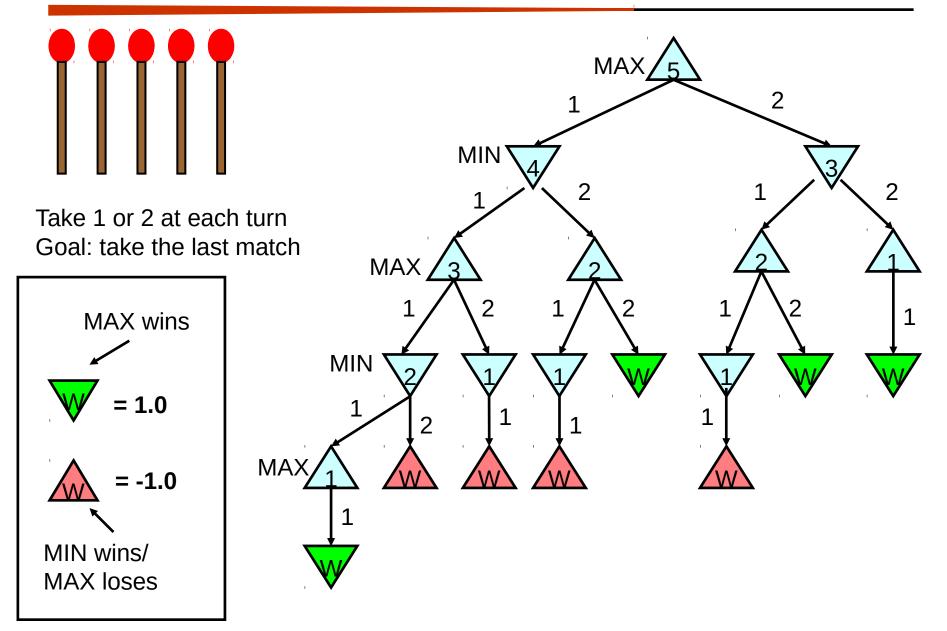
Player that takes the last coin wins

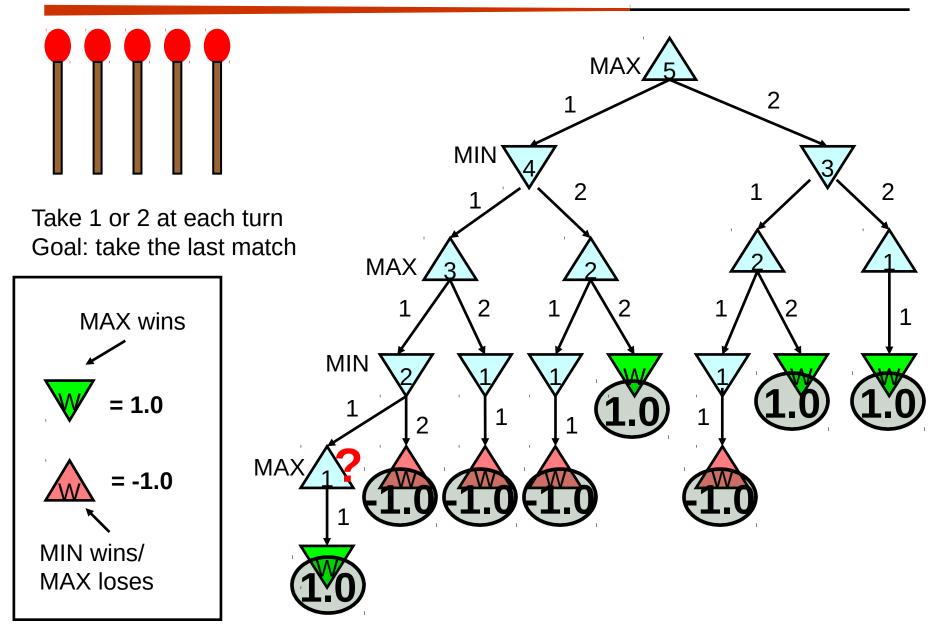
Example: https://www.goobix.com/games/nim/

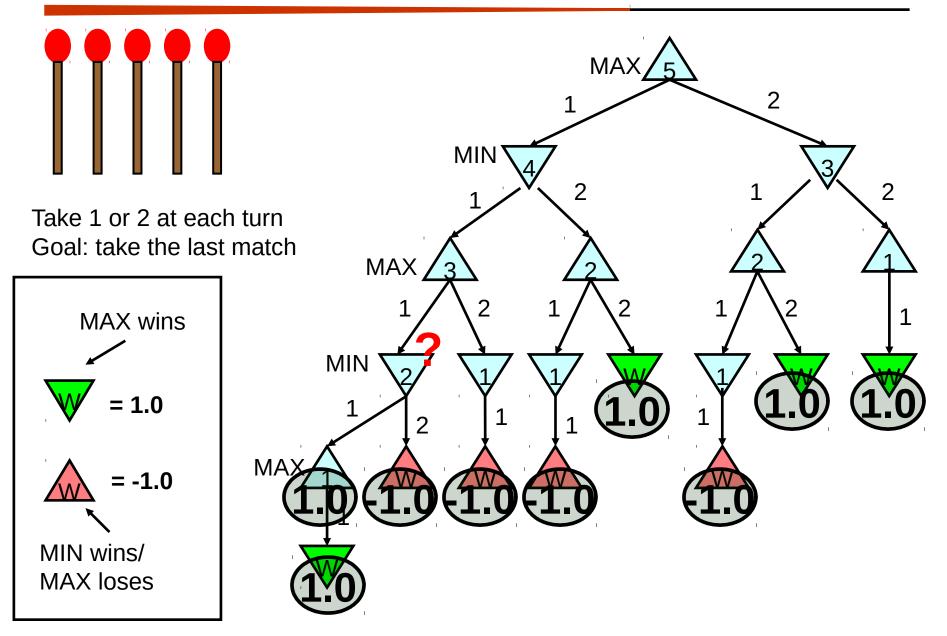


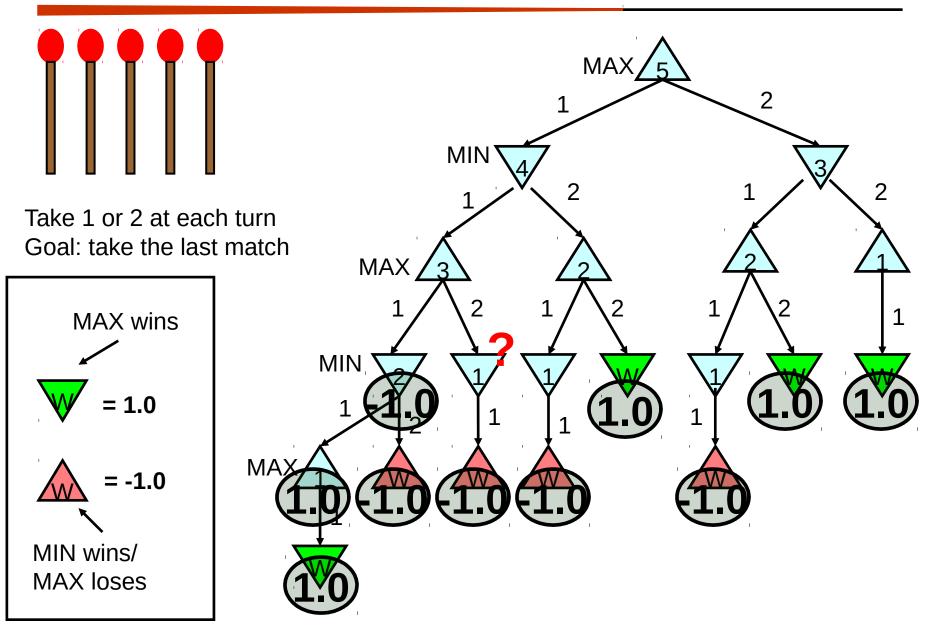
Take 1 or 2 at each turn Goal: take the last match

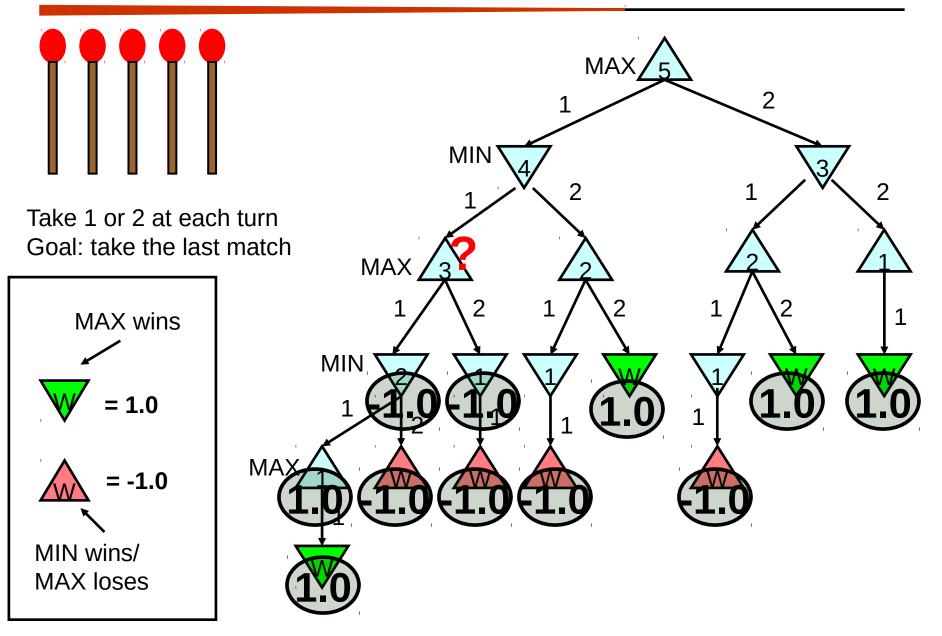
What move should I take?

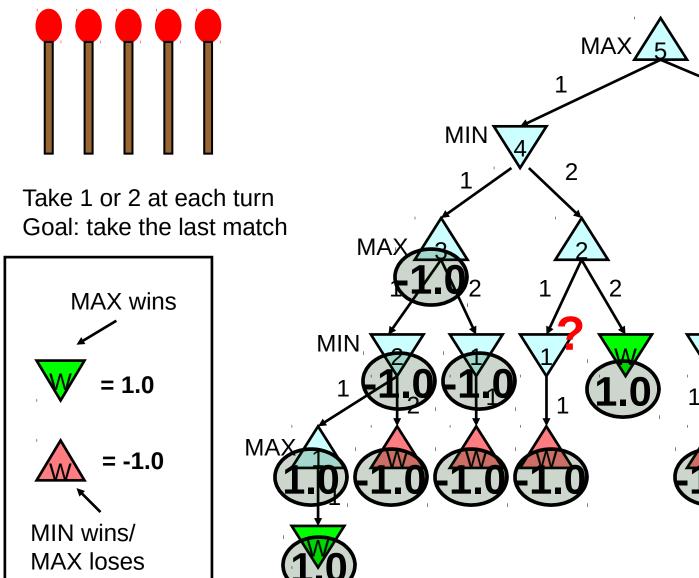


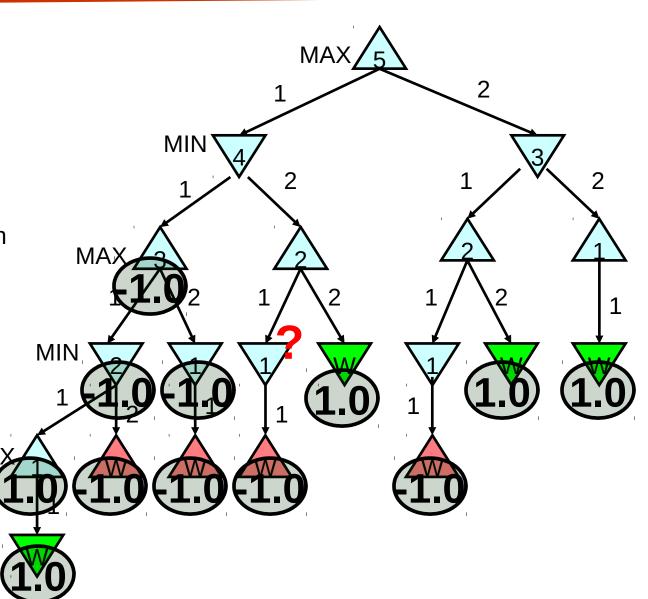


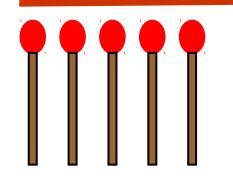


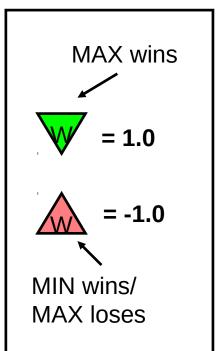


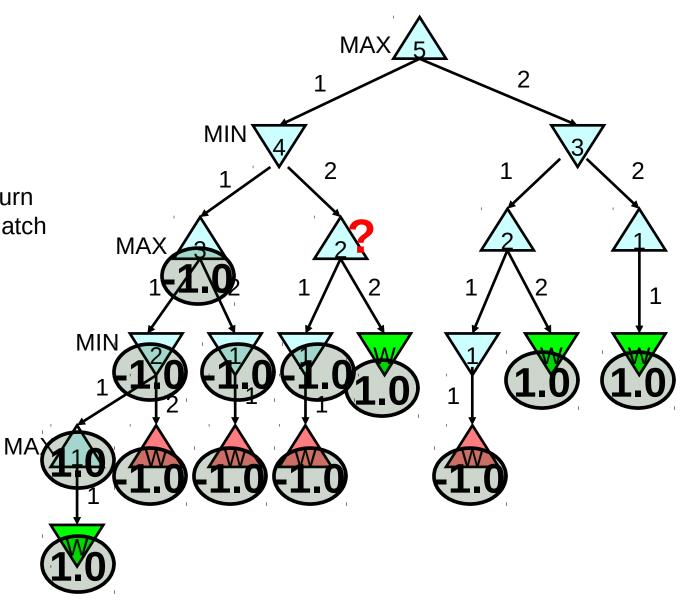


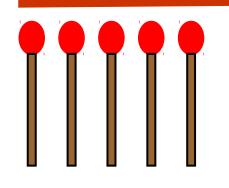


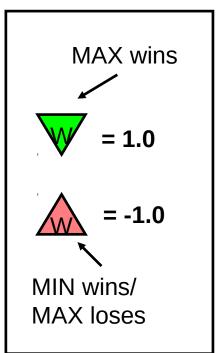


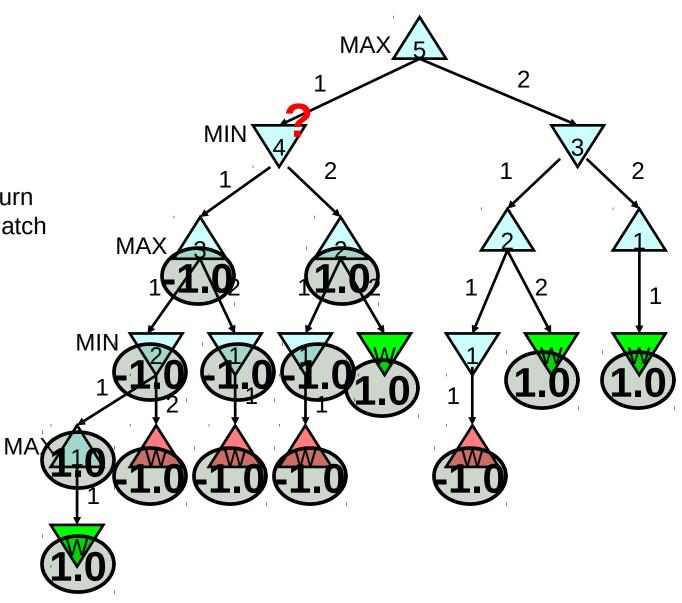


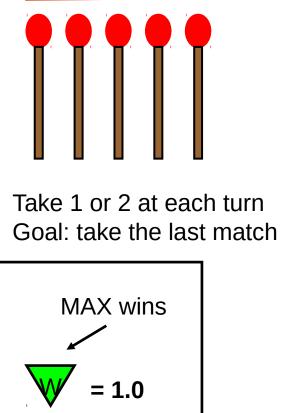








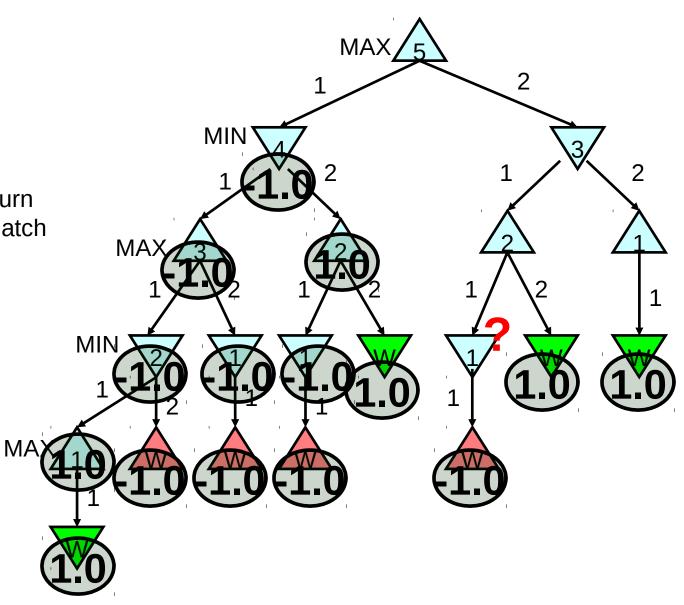


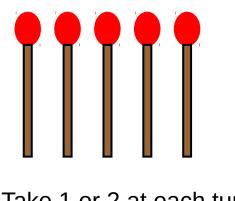


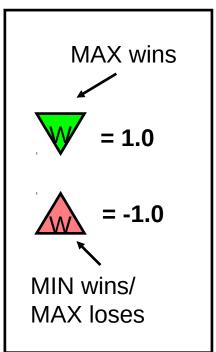
= -1.0

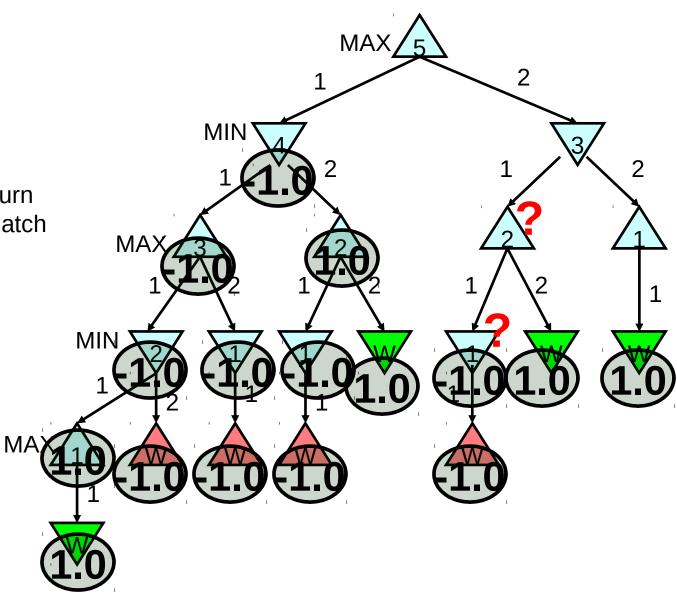
MIN wins/

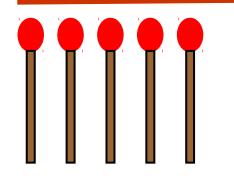
MAX loses

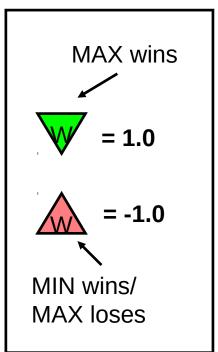


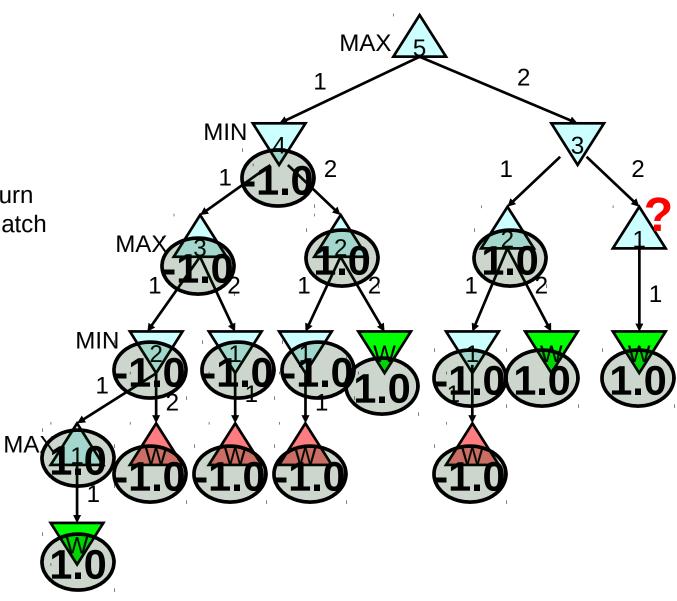


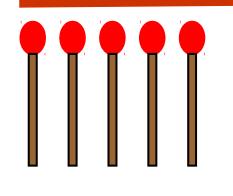


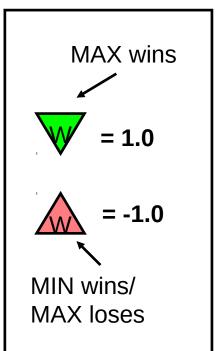


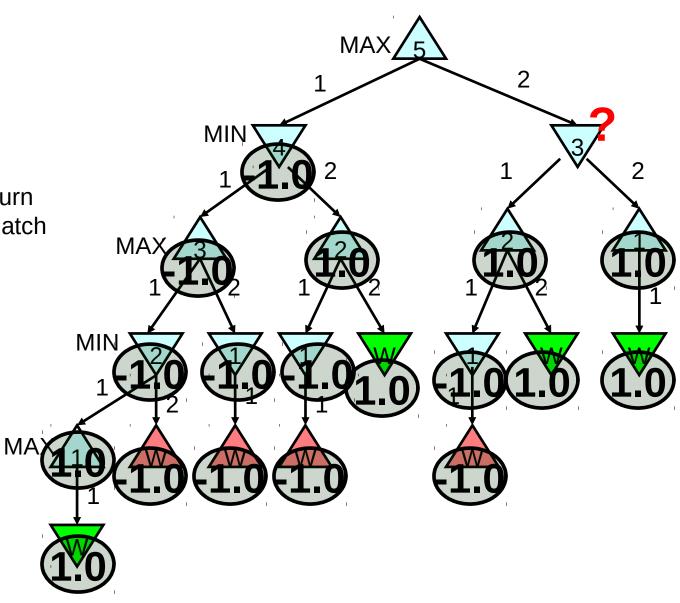


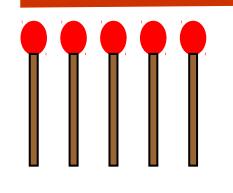


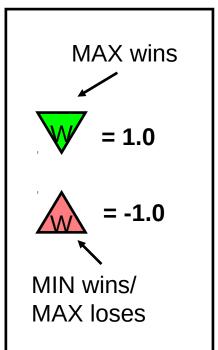


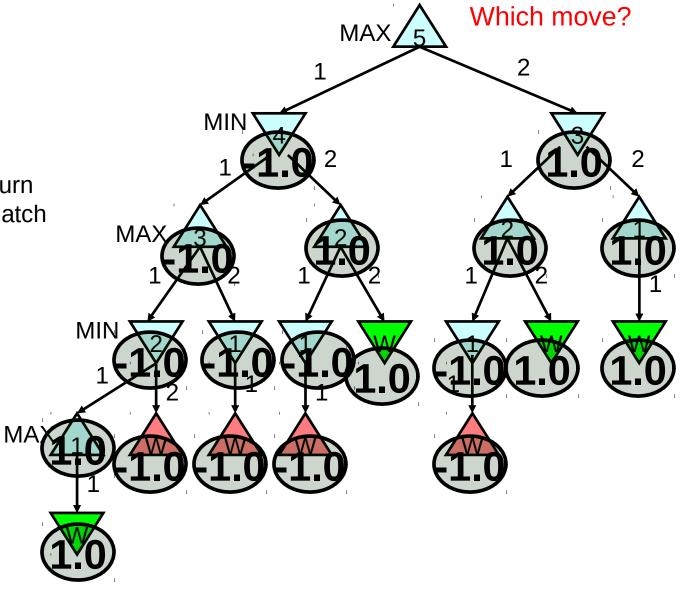


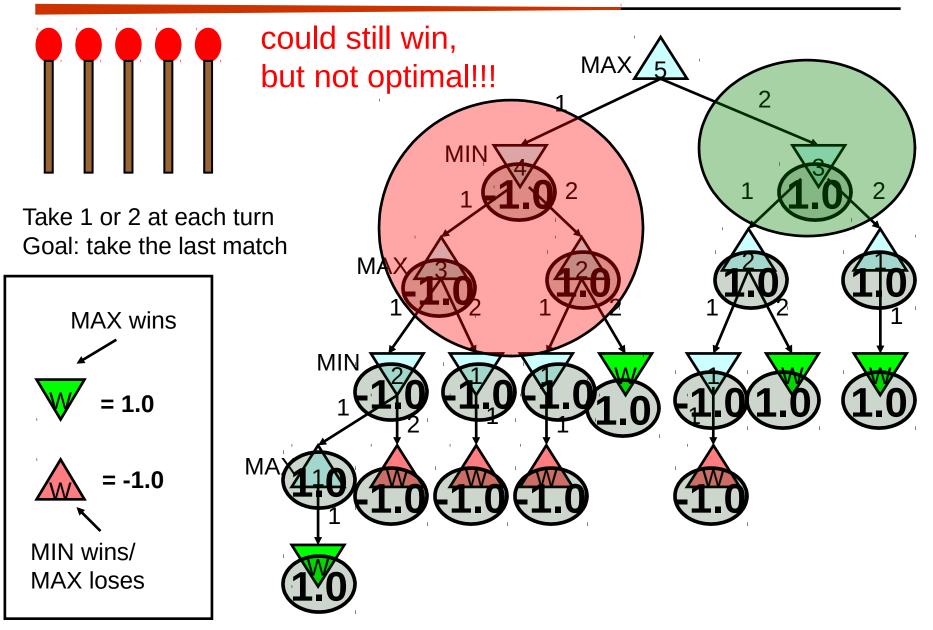




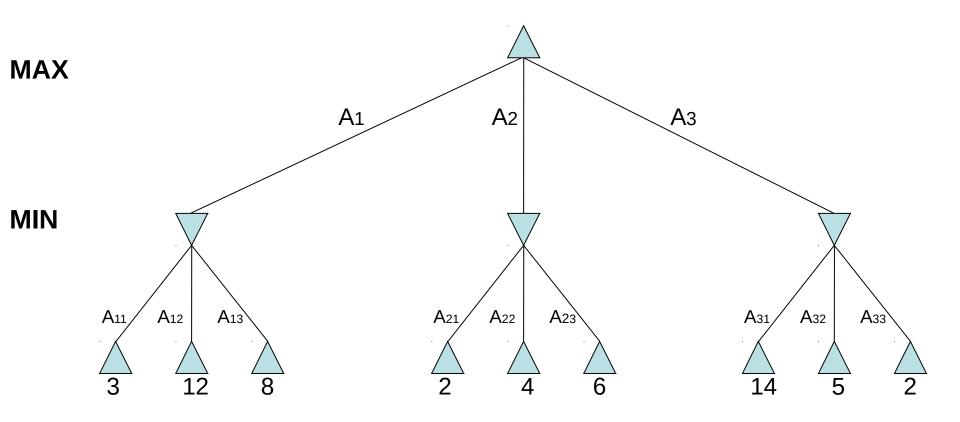






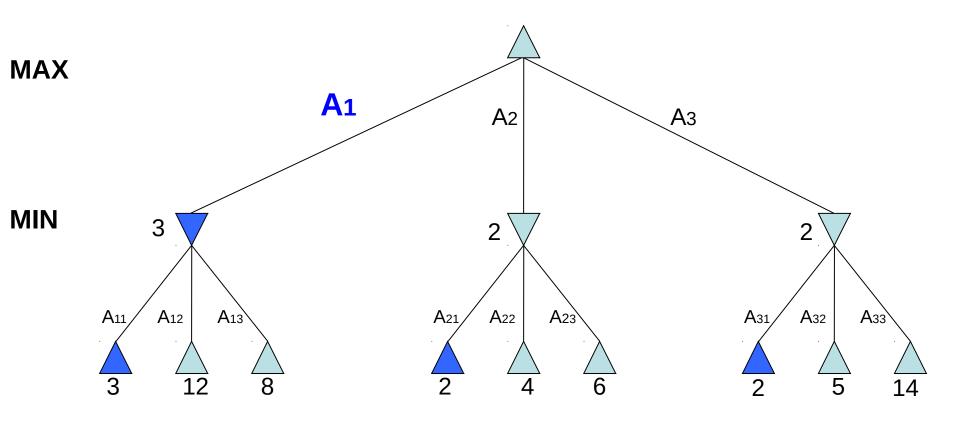


Minimax example 2



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

Minimax example 2



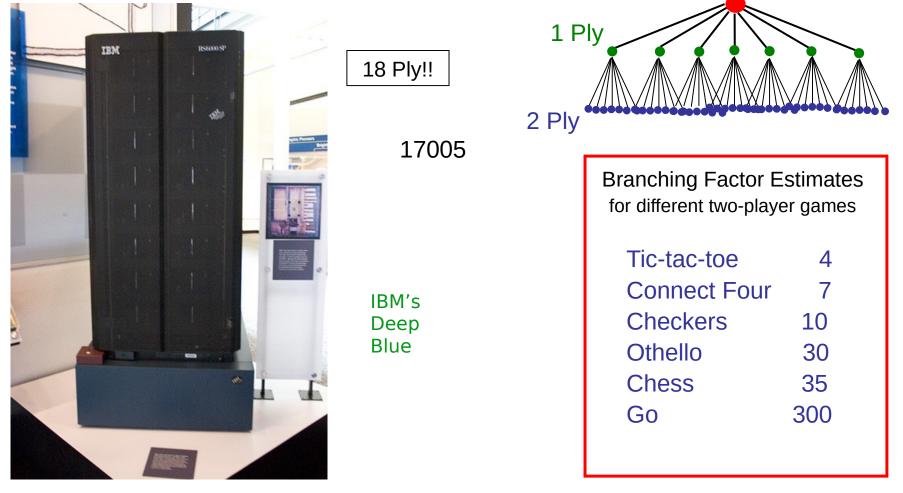
Properties of minimax

Minimax is optimal!

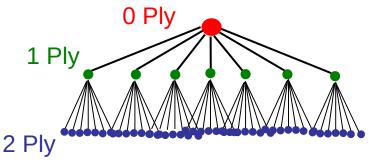
Are we done?



On average, there are ~35 possible moves that a chess player can make from any board configuration... 0 Ply



On average, there are \sim 35 possible moves that a chess player can make from any board configuration... 0 Ply

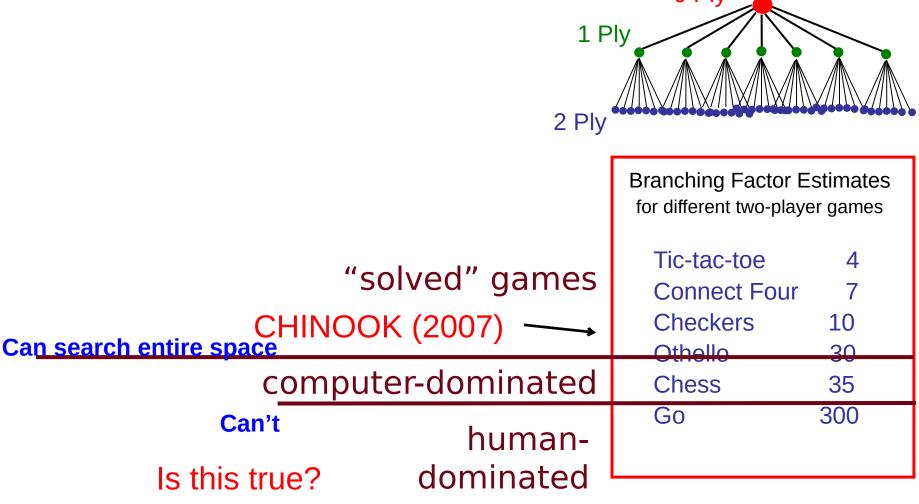


Branching Factor Estimates for different two-player games

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

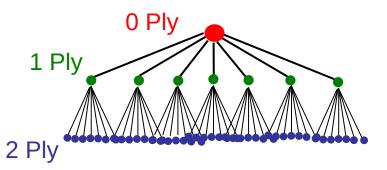
Boundaries for qualitatively different games...

On average, there are \sim 35 possible moves that a chess player can make from any board configuration... O Ply



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

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



Branching Factor Estimates for different two-player games

"solved" games Can search entire space HINOOK (2007) →	Tic-tac-toe Connect Four Checkers Othello	4 7 10 30
computer-dominated	Chess Go	35 300
Can't What do we do?		

Alpha-Beta pruning

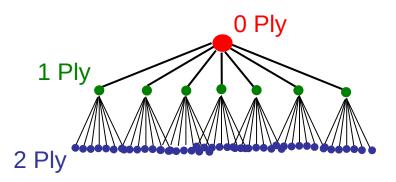
An optimal pruning strategy

- 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

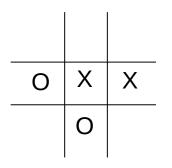
	Tic-tac-toe Connect Four Checkers Othello	4 7 10 30
computer-dominated	Chess Go	35 300

0 Ply

Pruning helps get a bit deeper

For many games, still can search the entire tree	1 P 2 Ply		
Go as deep as you can: - estimate the score/qua		Branching Factor Estimates for different two-player games	
the state (called an eva	aluation	Tic-tac-toe	4
function)		Connect Four	7
- use that instead of the	real	Checkers	10
score		Othello	30
	er-dominated	Chess	35
		Go	300

Tic Tac Toe evaluation functions



Ideas?

Example Tic Tac Toe EVAL

Tic Tac Toe Assume MAX is using "X"

EVAL(state) =

if state is win for MAX:

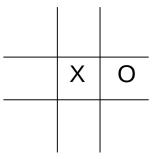
 $+\infty$

if state is win for MIN:

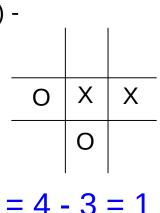
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else:

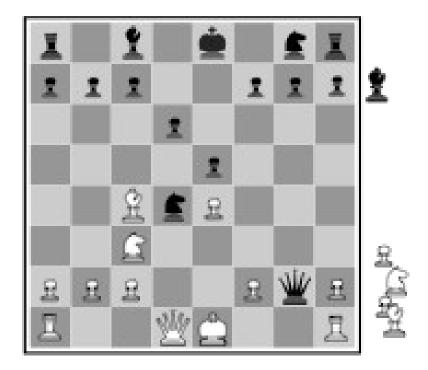
(number of rows, columns and diagonals available to MAX) - (number of rows, columns and diagonals available to MIN)







Chess evaluation functions



Ideas?

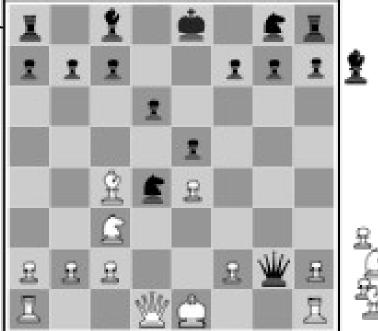
Chess EVAL

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

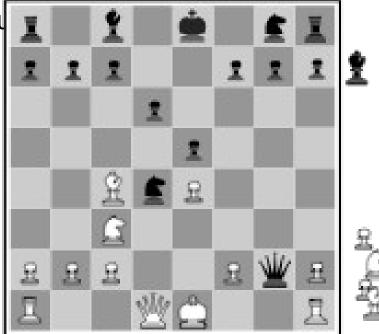
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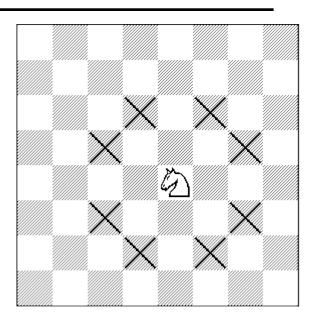
Any problems with this?

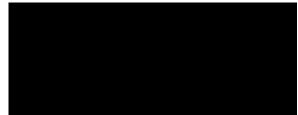
Chess EVAL

Ignores actual positions!

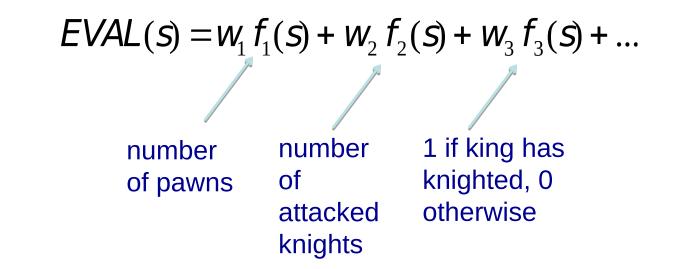
Actual heuristic functions are often a weighted combination of features

$$EVAL(S) = W_1 f_1(S) + W_2 f_2(S) + W_3 f_3(S) + \dots$$









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

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

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/