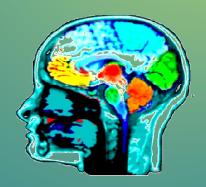


# Introduction To Artificial Intelligence

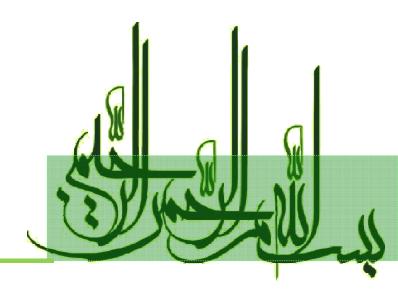
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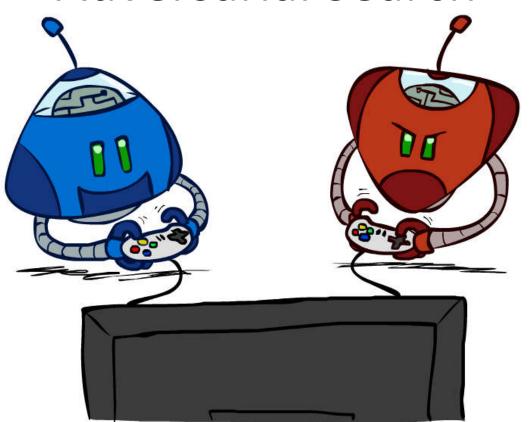
Adversarial Search

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[These slides were created by Dan Klein and Pieter Abbeel for CS188 Intro to AI at UC Berkeley.]



# **Adversarial Search**



[These slides were created by Dan Klein and Pieter Abbeel for CS188 Intro to AI at UC Berkeley. All CS188 materials are available at http://ai.berkeley.edu.]

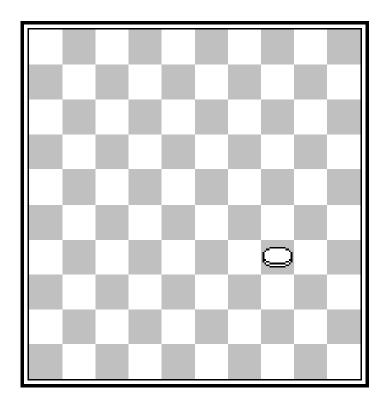
# Game Playing State-of-the-Art

Checkers: 1950: First computer player.

(Logical inference based)

1994: First computer champion: Chinook ended 40year-reign of human champion Marion Tinsley using complete 8-piece endgame.

2007: Checkers solved! (Win strategy found)



Gain des Blancs. S.R. 2003

## Game Playing State-of-the-Art

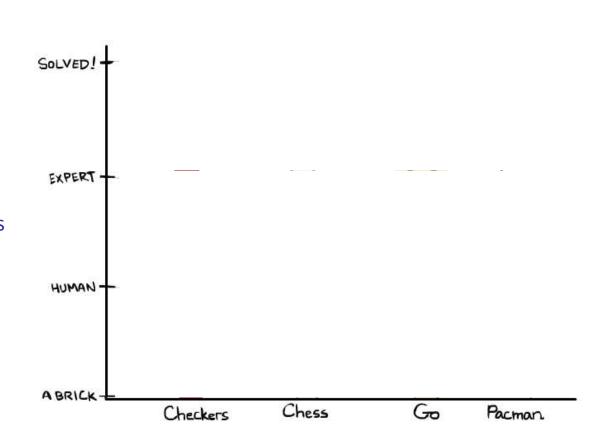
Checkers: 1950: First computer player.

(Logical inference based)

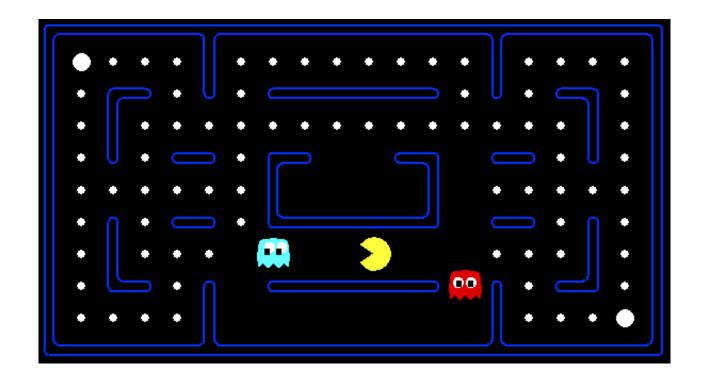
1994: First computer champion: CHINOOK ended 40year-reign of human champion Marion Tinsley using complete 8-piece endgame.

2007: Checkers solved! (Win strategy found)

- Chess: 1997: Deep Blue defeats human champion GARY KASPAROV in a six-game match. Deep Blue examined 200M positions per second, used very sophisticated evaluation and undisclosed methods for extending some lines of search up to 40 ply. Current programs are even better, if less historic.
- More branching factor
- Go: Human champions are now starting to be challenged by machines, though the best humans still beat the best machines. In go, branchfacotr > 300! Classic programs use pattern knowledge bases, but big recent advances use Monte Carlo (randomized) expansion methods.
- Pacman



# **Behavior from Computation**

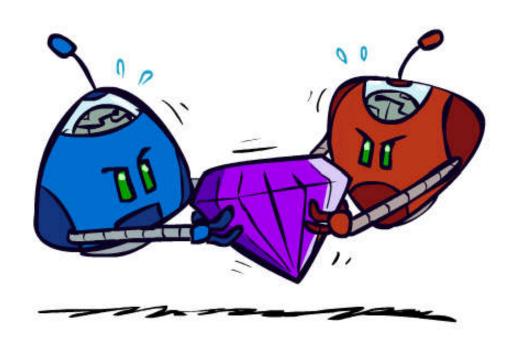


[Demo: mystery pacman (L6D1)]

# Video of Demo Mystery Pacman



# **Adversarial Games**



## Types of Games

Many different kinds of games!

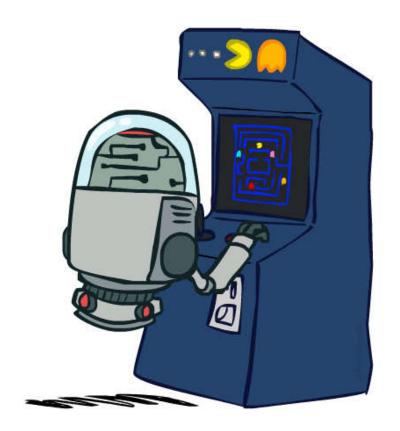
- Axes:
  - Deterministic or stochastic?
  - One, two, or more players?
  - Zero sum?
  - Perfect information (can you see the state)?



 Want algorithms for calculating a strategy (policy) which recommends a move from each state

#### **Deterministic Games**

- Many possible formalizations, one is:
  - States: S (start at  $s_0$ )
  - Players: P={1...N} (usually take turns)
    - here N=2
  - Actions: A (may depend on player / state)
  - Transition Function:  $S \times A \rightarrow S$
  - Terminal Test:  $S \rightarrow \{t, f(finish game)\}$
  - Terminal Utilities:  $S \times P \rightarrow R$  (utility)
    - Bigger R means better Game
    - Utility define for terminal states.
- Solution(goal) is not Path (why?)
- Solution for a player is a policy:  $S \rightarrow A$ 
  - Find policy to win game in every game

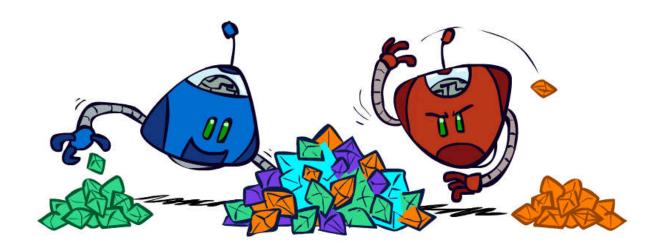


#### Game vs. Search Problem

- "Unpredictable" opponent ⇒ solution is a strategy specifying a action(move) for every possible state/opponent reply(actions)
- Time limits ⇒ unlikely to find goal, must approximate

#### **Zero-Sum Games**





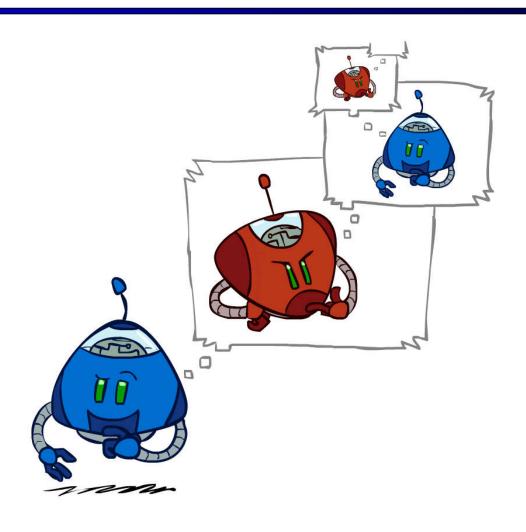
#### 1.Zero-Sum Games

- Same utility for opposite
- Agents have opposite utilities (values on outcomes)
- Lets us think of a single value that one maximizes and the other minimizes
- Adversarial, pure competition

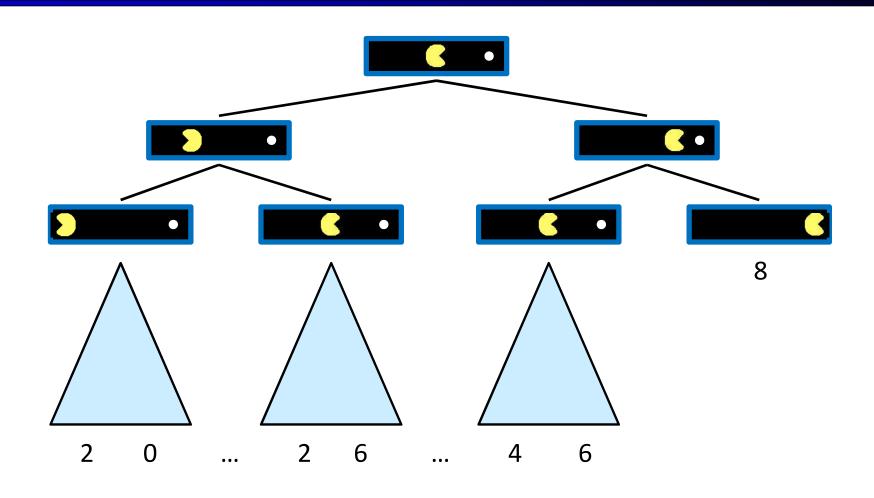
#### 2. General Games

- Agents have independent utilities (values on outcomes)
- Cooperation, indifference, competition, and more are all possible
- More later on non-zero-sum games

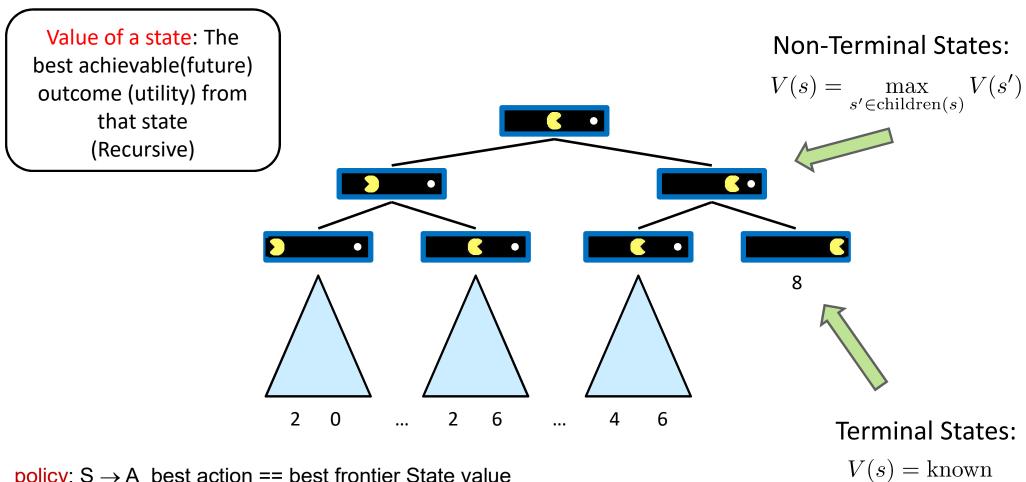
# **Adversarial Search**



# Single-Agent Trees

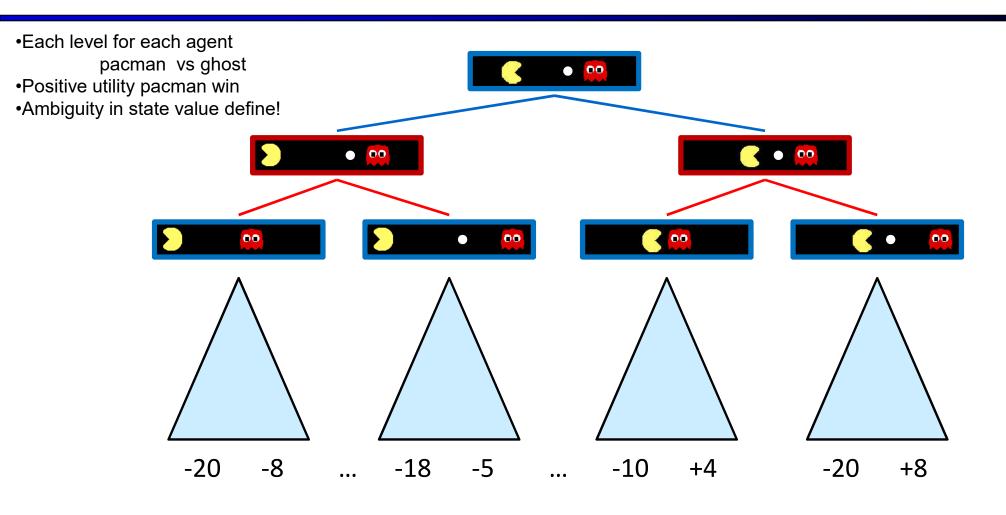


#### Value of a State



policy:  $S \rightarrow A$  best action == best frontier State value

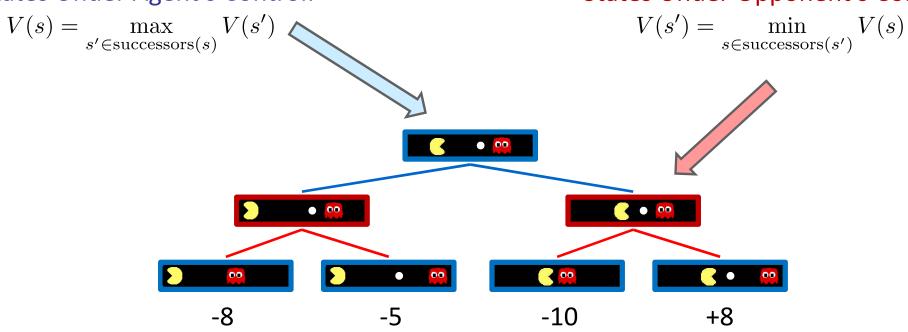
#### **Adversarial Game Trees**



#### Minimax Values

#### States Under Agent's Control:

#### States Under Opponent's Control:



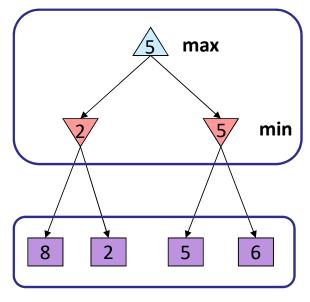
#### **Terminal States:**

$$V(s) = \text{known}$$

## Adversarial Search (Minimax)

- Deterministic, zero-sum games:
  - Tic-tac-toe, chess, checkers
  - One player maximizes result
  - The other minimizes result
- Minimax search:
  - A state-space search tree
  - Players alternate turns
  - Compute each node's minimax value: the best achievable utility against a rational (optimal) adversary

## Minimax values: computed recursively

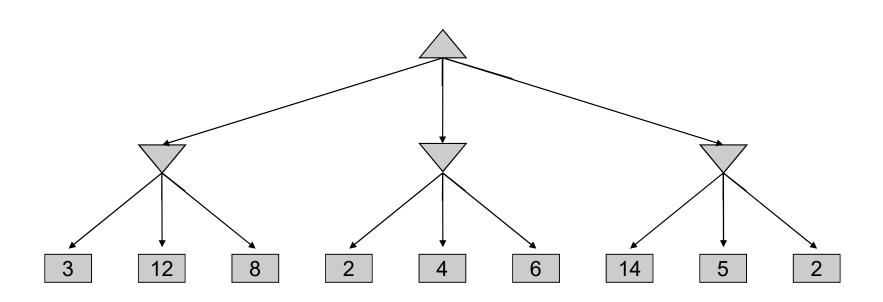


Terminal values: part of the game

#### MinMax Strategy

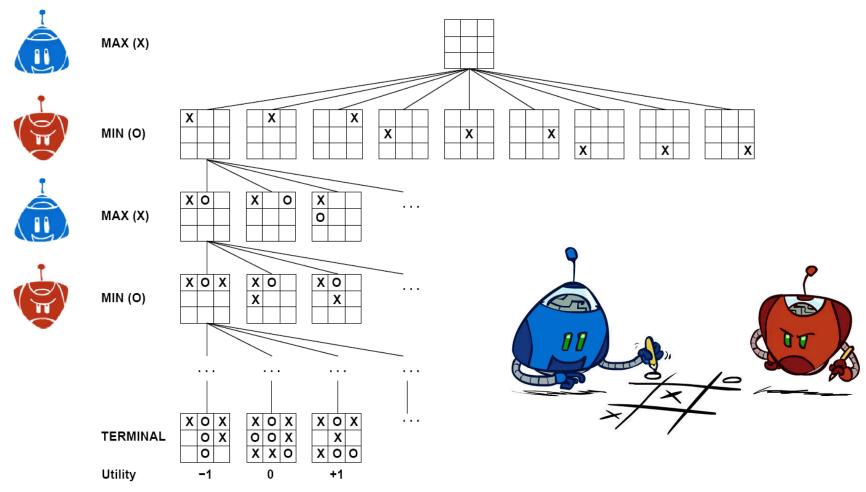
- Perfect play for deterministic, perfect-information games
- Idea: choose move to position with highest minimax value = best achievable payoff against best play

# Minimax Example



Best value solution obtain Again Best opponent actions(Min Select) What happen if opponent Select non minimum action?

#### Tic-Tac-Toe Game Tree



What is Max value?