











### The Planning Problem Applied To Chess

### Given:

- 1. An <u>initial state</u> of the world => The current state of the board
- 2. A set of <u>available actions</u>, their requirements, and their effects => Valid chess moves
- 3. A goal state => Checkmate conditions

### Compute:

A <u>valid sequence of actions</u> that starts from the initial state and terminates at the goal state with fewest actions => <u>Plan = sequence of chess moves</u>





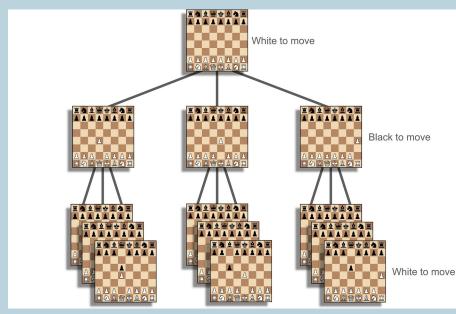
## Planning Via Search (Revision)

### Starting from initial state:

- 1. Enumerate all possible actions available, and the resulting states
- 2. Check if goal state reached
- 3. If not, for every possible outcome, repeat step 1 for all new states

## **Adversarial** Planning Via Search

- 1. Enumerate all possible actions available to you
- 2. Enumerate all possible actions available to opponent
- 3. Repeat 1, 2 until game ends
- 4. Rewind from game end back to now:
  - For each level pick action most favorable to the player (most games won)
  - 2. Stop when rewound back to now
- 5. Pick most favorable action (most games won)



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# **Adversarial** Planning Via Search

- 1. Enumerate all possible actions available to you
- 2. Enumerate all possible actions available to oppone Problem: There are too many possible game
- 3. Repe outcomes, exploring them all will take too long
- 4. Rewi (and exhaust computer memory)
  - 1. For each level pick action most tavorable to the player (most games won)
  - 2. Stop when rewound back to now
- 5. Pick most favorable action (most games won)

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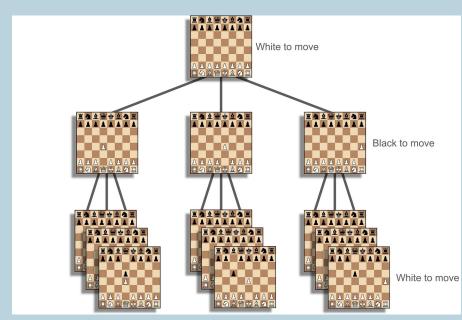


Black to move

Vhite to move

## Adversarial Planning Via Search +Heuristics

- 1. Enumerate all possible actions available to you
- 2. Enumerate all possible actions available to opponent
- 3. Repeat 1, 2 until game ends up to max d steps
- 4. Rewind from game end back to now:
  - 1. If not game end, use heuristic value of game state
  - 2. For each level pick action most favorable to the player (best value for player)
  - 3. Stop when rewound back to now
- 5. Pick most favorable action (most games won)



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### 1997: Deep Blue defeats Garry Kasparov

Algorithm: Tree search with some clever strategies to prune the search tree

"Secret Sauce": Special chips designed to speed up enumeration and evaluation of chess moves for tree search





Artificial Intelligence 134 (2002) 57-83

Artificial Intelligence

### Deep Blue

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### Abstract

Deep Blue is the chess machine that defeated then-reigning World Chess Champion Garry Kasparov in a six-game match in 1997. There were a number of factors that contributed to this success including:

- a single-chip chess search engine,
- a massively parallel system with multiple levels of parallelism
- a strong emphasis on search extensions

a complex evaluation function, and
 effective use of a Grandmaster game database

This paper describes the Deep Blue system, and gives some of the rationale that went into the design decisions behind Deep Blue. © 2001 Elsevier Science B.V. All rights reserved.

Keywords: Computer chess; Game tree search; Parallel search; Selective search; Search extensions; Evaluation function

### 1. Introduction

This paper describes the Deep Blue<sup>®</sup> computer chess system, developed at IBM<sup>®</sup> Research during the mid-1990s. Deep Blue is the culmination of a multi-year effort to build a world-class chess machine. There was a series of machine shat led up to Deep Blue, which we describe below. In fact there are two distinct versions of Deep Blue, one which lost to Garry Kasparov in 1996 and the one which defeated him in 1997. This paper will refer primarily to the 1997 version, and comparisons with the 1996 version, which we

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Campbell, Murray, A. Joseph Hoane Jr, and Feng-hsiung Hsu. "Deep blue." *Artificial intelligence* 134.1-2 (2002): 57-83.





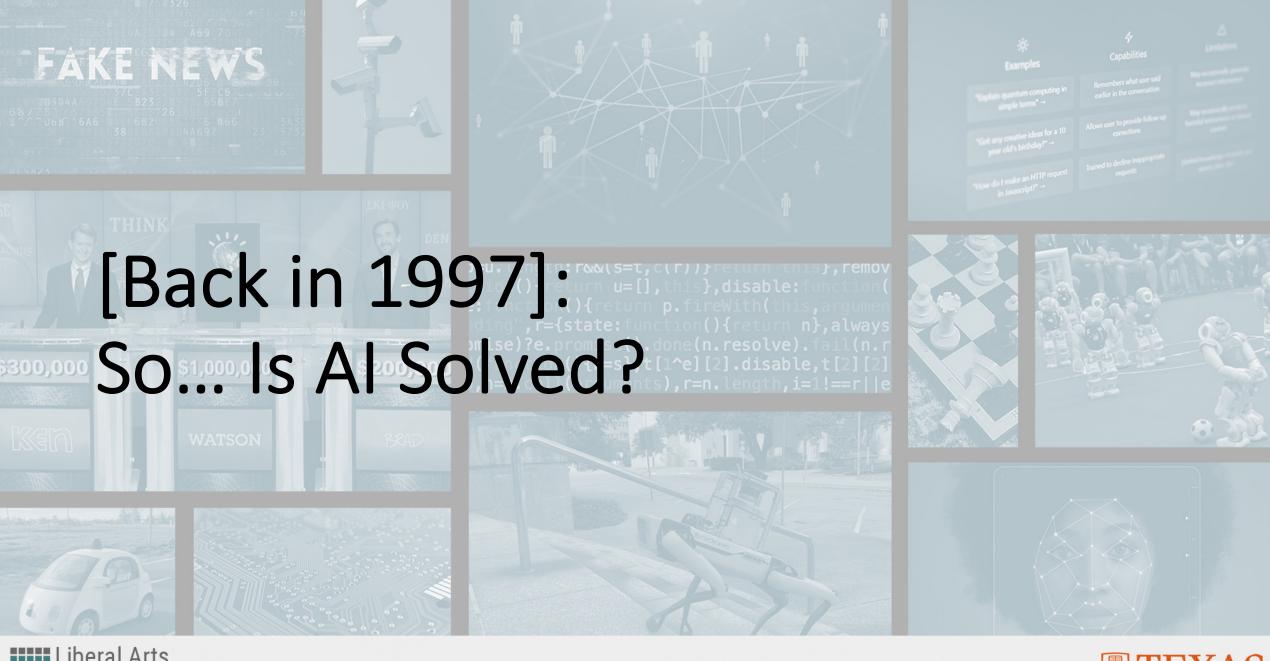
# And on 8/21/2023...



https://twitter.com/IJCAlconf/status/1693798967920746711











### Defining Artificial Intelligence

- A science and a set of computational technologies that are inspired by, but typically operate quite differently from, the ways people use their nervous systems and bodies to sense, learn, reason, and take action
- NOT one thing
  - More than just deep learning
  - RL, NLP, vision, **planning**, symbolic reasoning, algorithmic game theory, computational social choice, human computation
- Getting Computers to do the things they can't do yet
  - Once it works, it's engineering





### Adversarial Planning Via Search + Heuristics

- 1. Enumerate all possible actions available to you
- 2. Enumerate all possible actions available to opponent
- 3. Repeat 1, 2 until game ends up to max d steps
- 4. Rewind from game end back to now:
  - 1. If not game end, use heuristic value of game state
  - 2. For each level pick action most favorable to the player (best value for player)
  - 3. Stop when rewound back to now
- Pick most favorable action (most games won)

- 1. There are games (e.g., Go) where 1,2 are intractable even for a few steps.
- 2. We don't want to create specialpurpose computer chips <u>for every</u> <u>specific problem</u>









In the computer vision community, deep neural networks gain steam...

[Story for another lecture: Computer Vision]





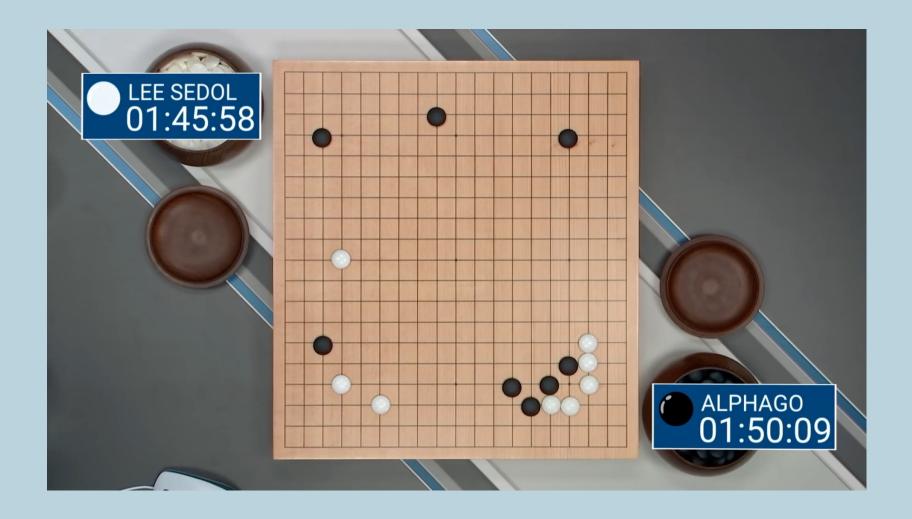
### 2016: AlphaGo Defeats Lee Sedol at Go







### 2016: AlphaGo Defeats Lee Sedol at Go







### Adversarial Planning Via Tree Search + Deep Learning

- 1. Enumerate promising actions [policy network] available to you
- 2. Enumerate *promising actions* [policy network] available to opponent
- 3. Repeat 1, 2 until game ends up to max d steps
- 4. Rewind from game end back to now:
  - 1. If not game end, use *learned value [value network]* of game state
  - 2. For each level pick action most favorable to the player (best value for player)
  - 3. Stop when rewound back to now
- 5. Pick most favorable action (most games won)



