

# CS540 Introduction to Artificial Intelligence

## Lecture 21

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Based on lecture slides by Jerry Zhu, Yingyu Liang, and Charles Dyer

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# Lion Game Example

Quiz

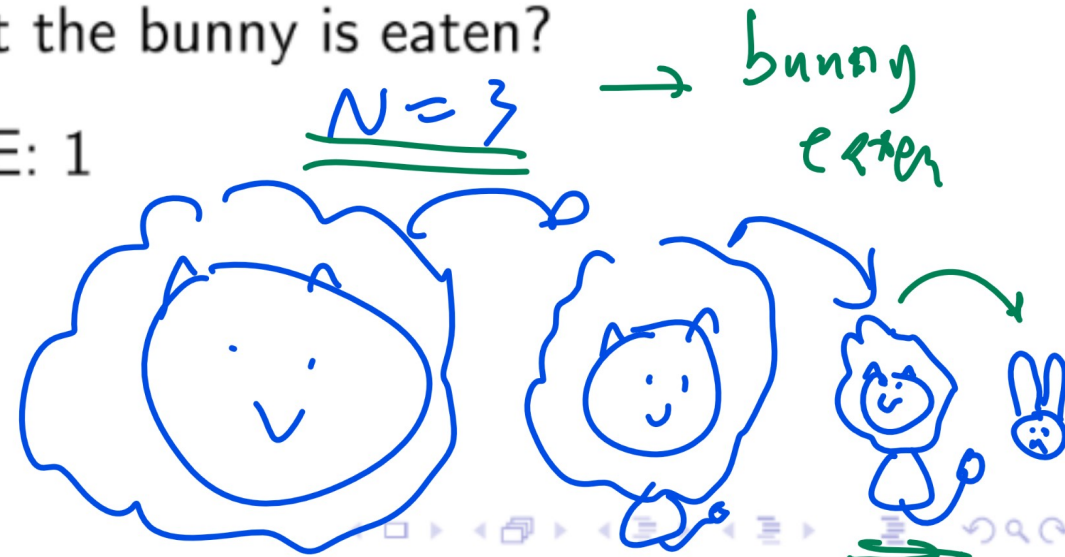
P3 solution  
country

→ P4 read data

- There are  $N$  lions, ordered by size,  $i = 1, 2, 3, \dots, N$ , and a bunny.  $N$  takes an integer between 1 and 10 with equal probability (known to all the lions). Each lion  $i$  can choose to jump out and eat the slightly smaller lion  $i - 1$ , or stay hidden, and only lion 1 can eat the bunny. Each lion prefers eating to staying hungry to being eaten.

- What is the probability that the bunny is eaten?

- A: 0, B:  $\frac{1}{3}$ , C:  $\frac{1}{2}$ , D:  $\frac{2}{3}$ , E: 1



# Lion Game Example Diagram

Quiz

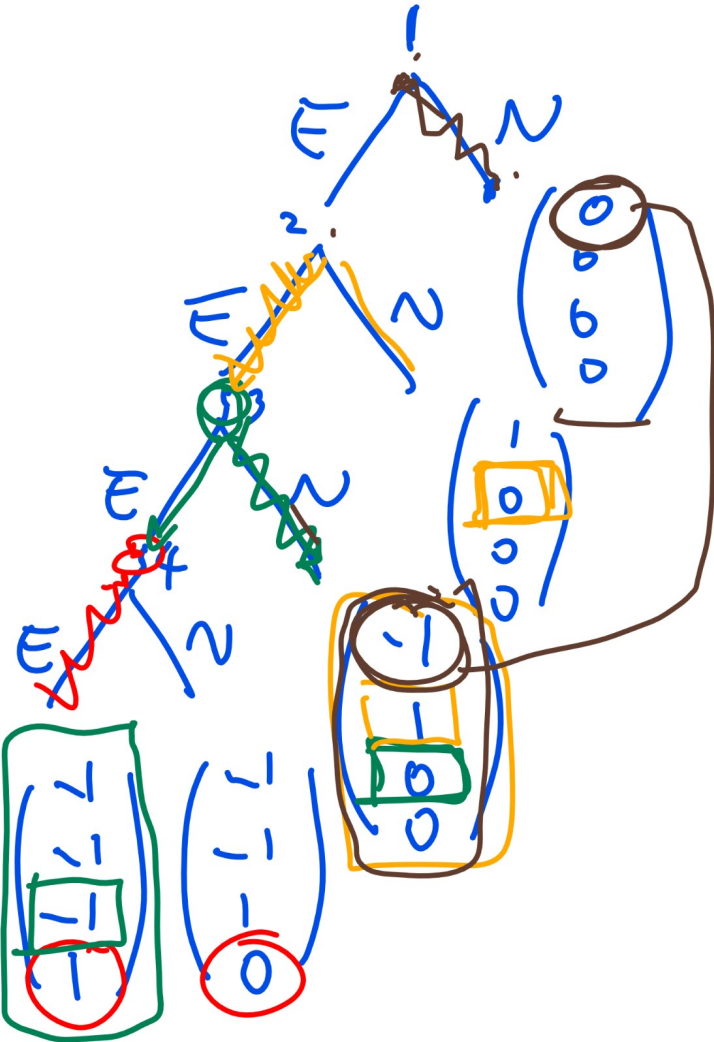
$\vec{E}at \rightarrow 1$   
 $Hungry \rightarrow 0$   
 $\vec{E}aten \rightarrow -1$

DFS

$N = 4$

bunny  
not eaten

$N$  odd  
 bunny eaten  
 $N$  even not



# Summary

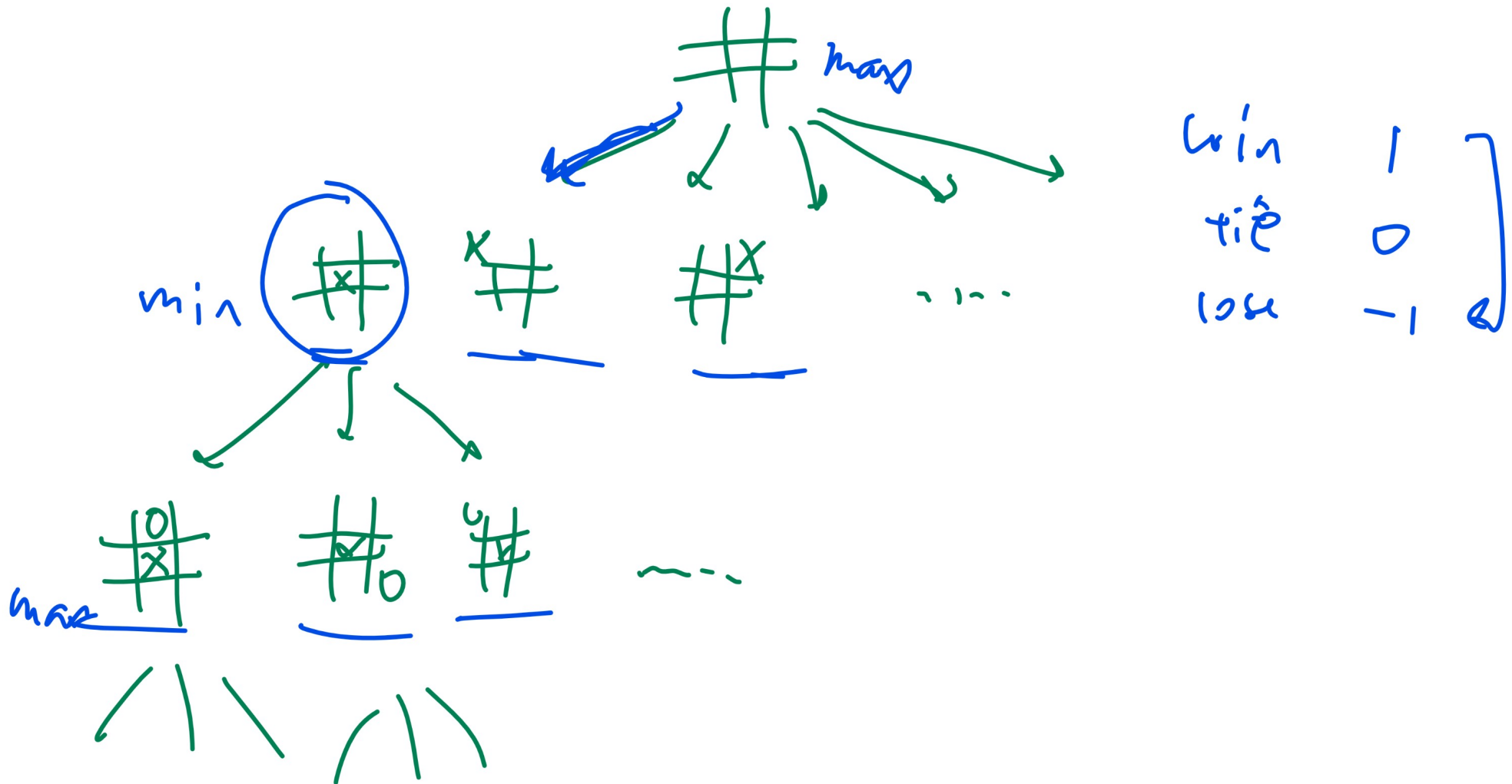
## Discussion

- Search:
  - 1 Uninformed.
  - 2 Informed.
  - 3 Local Search.
  - 4 Adversarial Search: Sequential move games. ←
  - 5 Adversarial Search: Simultaneous move games. ←



# Tic Tac Toe Example

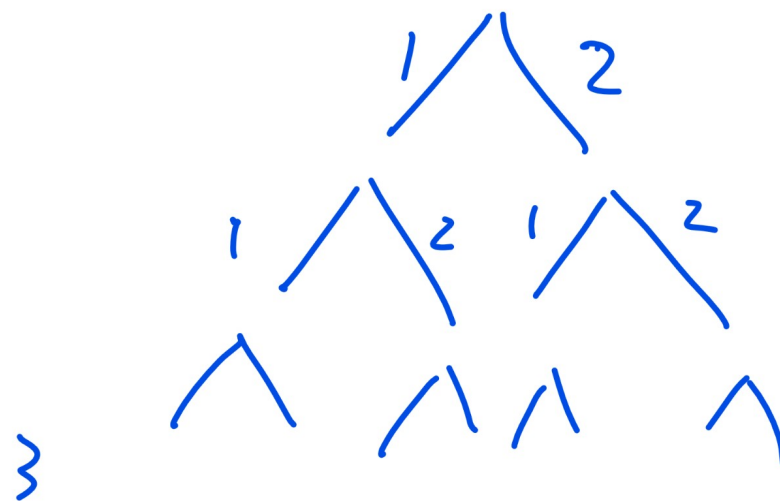
## Motivation



# Nim Game Example

## Motivation

- Ten objects. Pick 1 or 2 each time. Pick the last one to win.
- *A* : Pick 1.
- *B* : Pick 2.
- *C, D, E* : Don't choose.



# Minimax Algorithm

## Description

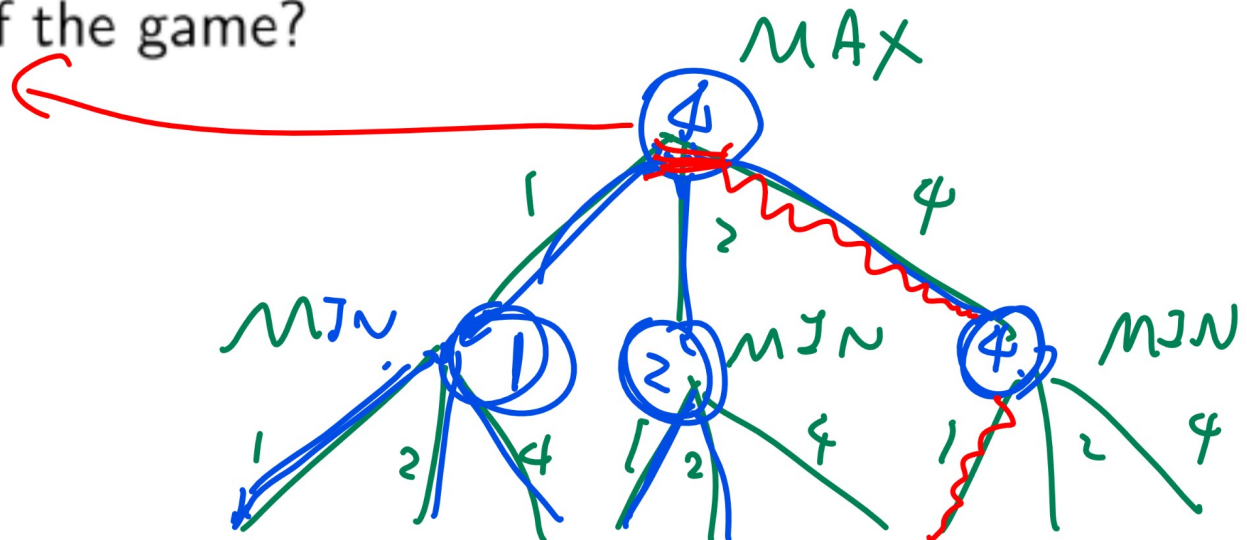
- Use DFS on the game tree.

# Minimax Example

## Quiz

- For a zero-sum game, the value to the MAX player if MAX plays  $x_1 \in \{1, 2, 4\}$  and MIN plays  $x_2 \in \{1, 2, 4\}$  is  $x_1 \cdot x_2$ . What is the value of the game?

- A : 1
- B : 2
- **C : 4**
- D : 8
- E : 16



1	2	4	2	4	8	4	8	16
-1	-2	-4	-2	-4	-8	-4	-8	-16

# Minimax Performance

## Discussion

- The time and space complexity is the same as DFS. Note that  $D = d$  is the maximum depth of the terminal states.

$$T = 1 + b + b^2 + \dots + b^d$$

$$S = (b - 1) \cdot d$$

} ←

# Non-deterministic Game

## Discussion

- For non-deterministic games in which chance can make a move (dice roll or coin flip), use expected reward or cost instead.

payoff score

- The algorithm is also called expectiminimax.



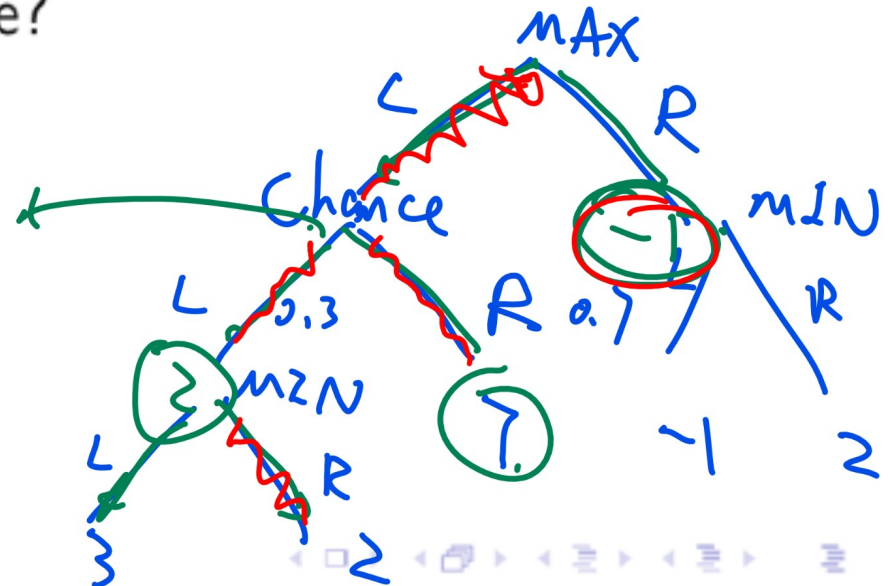
# Game Tree with Chance Example 1

## Quiz

- Max can pick  $L$  or  $R$ . If Max picks  $L$ , Chance picks  $L$  with probability 0.3 and  $R$  with probability 0.7. If Chance picks  $L$ , Min picks  $L$  to get 3,  $R$  to get 2, and if Chance picks  $R$ , Min gets 7. If Max picks  $R$ , Min picks  $L$  to get -1 and  $R$  to get 2. What is the value of the game?

5.5

$$\begin{array}{r}
 7 \cdot 0.7 + 2 \cdot 0.3 \\
 \hline
 4.9 + 0.6 \\
 = 5.5
 \end{array}$$



# Game Tree with Chance Example 1 Diagram

## Quiz

# Pruning

## Motivation

- Time complexity is a problem because the computer usually has a limited amount of time to "think" and make a move.
- It is possible to reduce the time complexity by removing the branches that will not lead the current player to win. It is called the Alpha-Beta pruning.

# Alpha Beta Pruning

## Description

- During DFS, keep track of both  $\alpha$  and  $\beta$  for each vertex.
- Prune the subtree with  $\alpha \geq \beta$ .

# Alpha Beta Example 1

## Quiz

# Alpha Beta Example 1 Continued

## Quiz

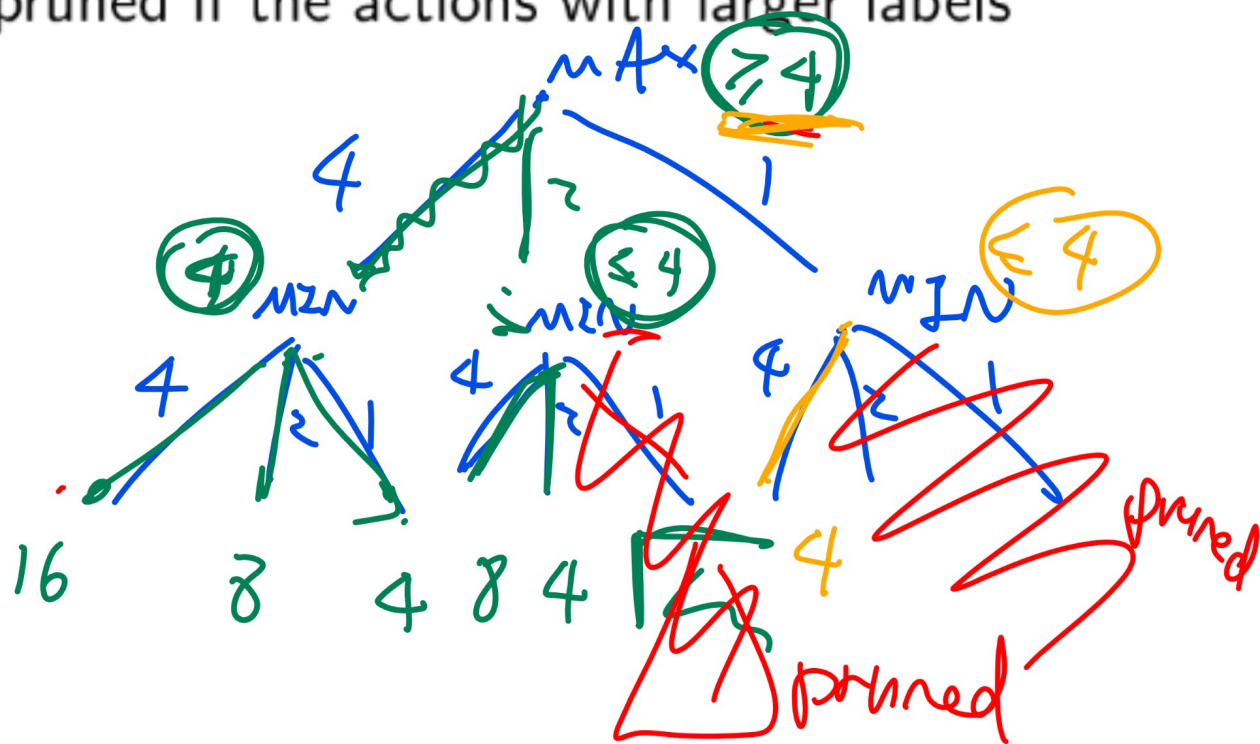


# Alpha Beta Example 2

## Quiz

• For a zero-sum game, the value to the MAX player if MAX plays  $x_1 \in \{1, 2, 4\}$  and MIN plays  $x_2 \in \{1, 2, 4\}$  is  $x_1 \cdot x_2$ . Alpha-Beta pruning is used. What is the number of branches (states) that can be pruned if the actions with larger labels are searched first?

- A : 0
- B : 1
- C : 2
- **D : 3**
- E : 4



# Alpha Beta Example 3

## Quiz

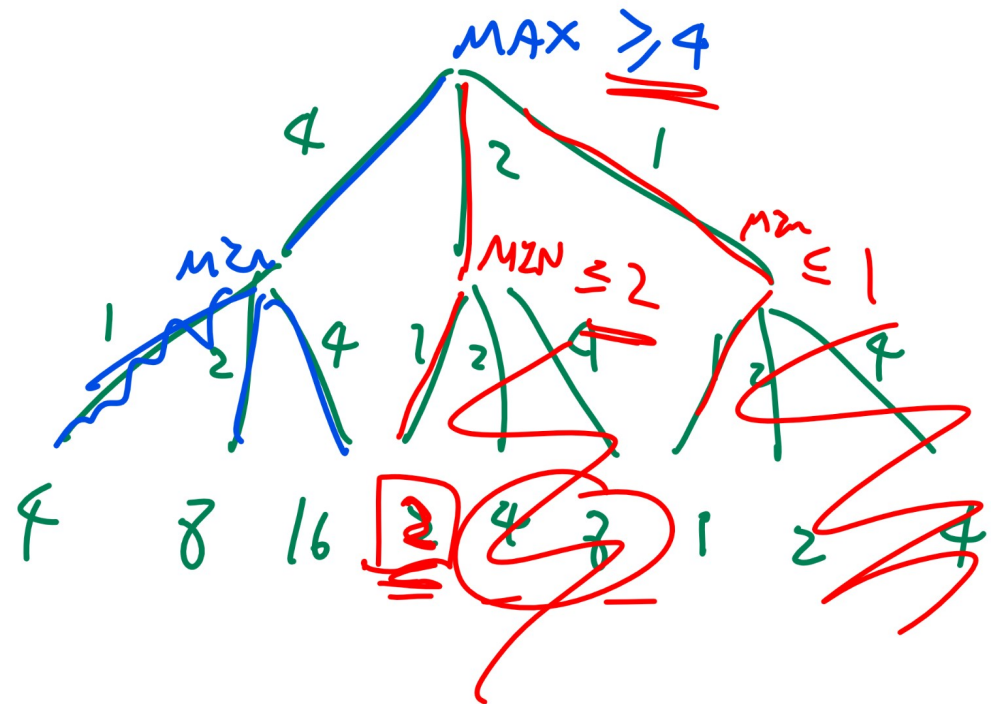
- For a zero-sum game, the value to the MAX player if MAX plays  $x_1 \in \{1, 2, 4\}$  and MIN plays  $x_2 \in \{1, 2, 4\}$  is  $x_1 \cdot x_2$ . Alpha-Beta pruning is used. What is the number of branches (states) that can be pruned if the actions with smaller labels are searched first?
- A : 0
- B : 1
- C : 2
- D : 3
- E : 4

# Alpha Beta Example 4

## Quiz

- For a zero-sum game, the value to the MAX player if MAX plays  $x_1 \in \{1, 2, 4\}$  and MIN plays  $x_2 \in \{1, 2, 4\}$  is  $x_1 \cdot x_2$ . Alpha-Beta pruning is used. What is the maximum number of branches (states) that can be pruned if the actions can be searched in any order?

- A : 2
- B : 3
- C : 4
- D : 5
- E : 6



# Alpha Beta Example 4

## Quiz

# Alpha Beta Example 4 Continued

## Quiz

# Alpha Beta Performance

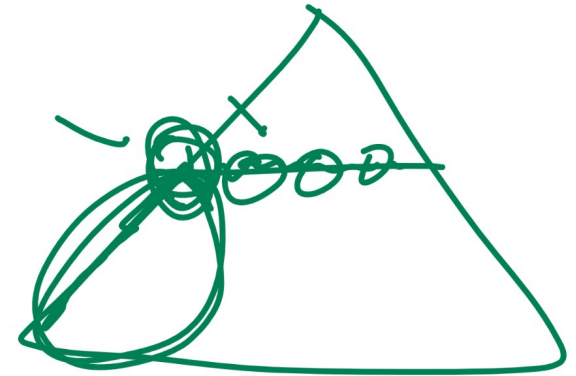
## Discussion

- In the best case, the best action of each player is the leftmost child.
- In the worst case, Alpha Beta is the same as minimax.



# Static Evaluation Function

## Definition



- A static board evaluation function is a heuristics to estimate the value of non-terminal states.
- It should reflect the player's chances of winning from that vertex.
- It should be easy to compute from the board configuration.

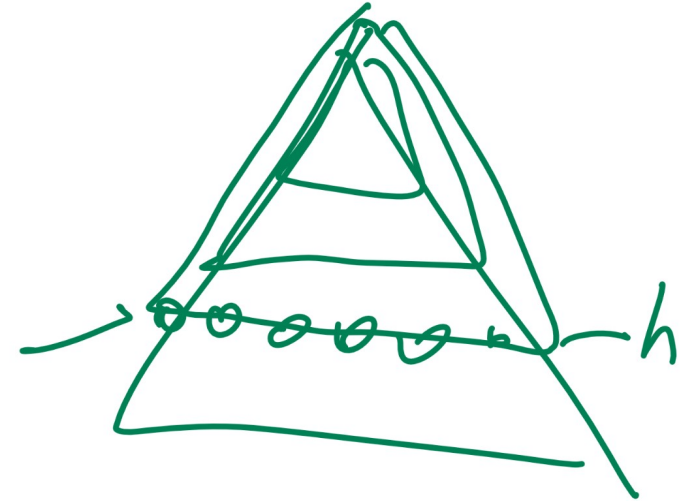
# Linear Evaluation Function Example

## Definition

- For Chess, an example of an evaluation function can be a linear combination of the following variables.
  - 1 Material.
  - 2 Mobility.
  - 3 King safety.
  - 4 Center control.
- These are called the features of the board.

# Iterative Deepening Search

## Discussion



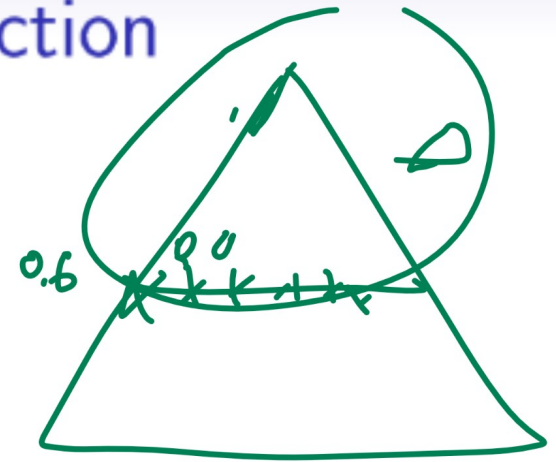
- IDS could be used with SBE.
- In iteration  $d$ , the depth is limited to  $d$ , and the SBE of the non-terminal vertices are used as their cost or reward.

# IDS with SBE Diagram

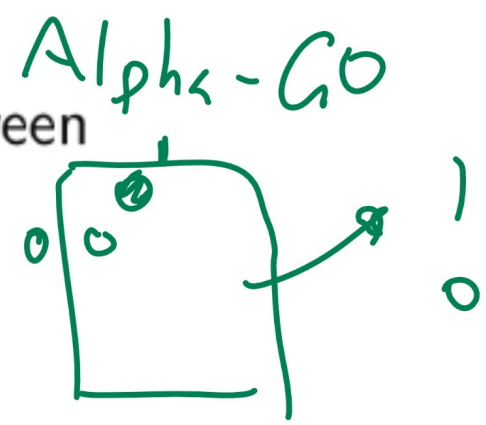
## Discussion

# Non Linear Evaluation Function

## Discussion



- The SBE can be estimated given the features using a neural network.
- The features are constructed using domain knowledge, or a possibly a convolutional neural network.
- The training data are obtained from games between professional players.



# Monte Carlo Tree Search

## Discussion



- Simulate random games by selecting random moves for both players.
- Exploitation by keeping track of average win rate for each successor from previous searches and picking the successors that lead to more wins.
- Exploration by allowing random choices of unvisited successors.



# Monte Carlo Tree Search Diagram

## Discussion

back 7:40 ,

# Alpha GO Example

## Discussion

- MCTS with  $> 10^5$  play-outs.
- Convolutional neural network to compute SBE.

# Summary

## Discussion

- Adversarial Search:
  - 1 Sequential Move Games: Minimax  $\rightarrow$  DFS on the game tree.
  - 2 Sequential Move Games: Alpha-Beta Pruning  $\rightarrow$  DFS to keep track  $\alpha$  and  $\beta \rightarrow$  prune the subtree with  $\alpha \Rightarrow \beta$ .
  - 3 Simultaneous Move Games: Iterated Elimination of Strictly Dominated Strategies (Rationalizability).
  - 4 Simultaneous Move Games: Nash Equilibrium.