

# CS 540 Introduction to Artificial Intelligence **Games I**

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**April 1, 2021** 

#### **Announcements**

- Homeworks:
  - None!
- Midterm: grading nearly done.

Class roadmap:

Thursday, April 1	Games I
Tuesday, April 6	Games II
Thursday, April 8	Search I
Tuesday, April 13	Search II

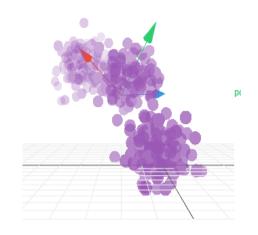
#### **Outline**

- Introduction to game theory
  - Properties of games, mathematical formulation
- Simultaneous Games
  - Normal form, strategies, dominance, Nash equilibrium
- Sequential Games
  - Game trees, minimax, search approaches

#### So Far in The Course

#### We looked at techniques:

- Unsupervised: See data, do something with it. Unstructured.
- **Supervised:** Train a model to make predictions. More structure.
  - Training: as taking actions to get a reward
- Games: Much more structure.



Victor Powell



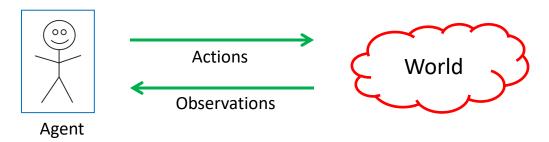


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#### More General Model

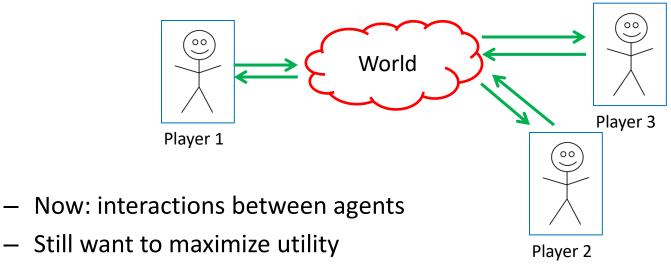
Suppose we have an agent interacting with the world



- Agent receives a reward based on state of the world
  - Goal: maximize reward / utility (\$\$\$)
  - Note: now data consists of actions & observations
  - Setup for decision theory, reinforcement learning, planning

## Games: Multiple Agents

#### Games setup: multiple agents



Strategic decision making.

## **Modeling Games: Properties**

- Number of agents/players
- State & action spaces: discrete or continuous
- Finite or infinite
- Deterministic or random
- Sum: zero or positive or negative
- Sequential or simultaneous



## Property 1: Number of players

Pretty clear idea: 1 or more players

- Usually interested in ≥ 2 players
- Typically a finite number of players





## Property 2: Discrete or Continuous

- Recall the world. It is in a particular state, from a set of states
- Similarly, the actions the player takes are from an action space
- How big are these spaces? Finite, countable, uncountable?







## Property 3: Finite or Infinite

- Most real-world games finite
- Lots of single-turn games; end immediately
  - Ex: rock/paper/scissors
- Other games' rules (state & action spaces) enforce termination
  - Ex: chess under FIDE rules ends in at most 8848 moves
- Infinite example: pick integers. First player to play a 5 loses



## Property 4: **Deterministic** or **Random**

- Is there chance in the game?
- Note: randomness enters in different ways



## Property 5: Sums

- Sum: zero or positive or negative
- Zero sum: for one player to win, the other has to lose
  - No "value" created

Blue Red	A	4	L	3	(	
1	30	-30	-10	10	20	-20
2	-10	10	20	-20	-20	20

- Can have other types of games: positive sum, negative sum.
  - Example: prisoner's dilemma

## **Property 6: Sequential or Simultaneous**

- Sequential or simultaneous
- Simultaneous: all players take action at the same time
- Sequential: take turns

- Simultaneous: players do not have information of others' moves. Ex: RPS
- Sequential: may or may not have perfect information (knowledge of all moves so far)





## Examples

#### Let's apply this to examples:

- 1. Chess: 2-player, discrete, finite, deterministic, zero-sum, sequential (perfect information)
- 2. RPS: **2-player, discrete, finite, deterministic, zero-sum, simultaneous**
- 3. Mario Kart: 4-player, continuous, infinite (?), random, zero-sum, simultaneous



## Another Example: Prisoner's Dilemma

Famous example from the '50s.

Two prisoners A & B. Can choose to betray the other or not.

- A and B both betray, each of them serves two years in prison
- One betrays, the other doesn't: betrayer free, other three years
- Both do not betray: one year each

Properties: **2-player**, **discrete**, **finite**, **deterministic**, **zero-sum**, **simultaneous** 



## Why Do These Properties Matter?

#### Categorize games in different groups

- Can focus on understanding/analyzing/"solving" particular groups
- Abstract away details and see common patterns
- Understand how to produce a "good" overall outcome



## How Does it Connect To Learning?

Obviously, learn how to play effectively

Also: suppose the players don't know something

- Ex: the reward / utility function is not known
- Common for real-world situations
  - How do we choose actions?
- Model the reward function and learn it
  - Try out actions and observe the rewards



#### Simultaneous Games

#### Simpler setting, easier to analyze

- Can express reward with a simple diagram
- Ex: for prisoner's dilemma

Player 2	Stay silent	Betray
Player 1	,	,
Stay silent	-1, -1	-3, 0
Betray	0, -3	-2, -2

#### Normal Form

Mathematical description of simult. games. Has:

- *n* players {1,2,...,*n*}
- Player i strategy  $a_i$  from  $A_i$ . All:  $a = (a_1, a_2, ..., a_n)$
- Player i gets rewards  $u_i(a)$  for any outcome
  - Note: reward depends on other players!

• Setting: all of these spaces, rewards are known

## Example of Normal Form

#### Prisoner's Dilemma

Player 2	Stay silent	Betray
Player 1	_	-
Stay silent	-1, -1	-3, 0
Betray	0, -3	-2, -2

- 2 players: yields 2x2 matrix
- Strategies: {Stay silent, betray} (i.e, binary)
- Rewards: {0,-1,-2,-3}

## **Dominant Strategies**

Let's analyze such games. Some strategies are better

- Dominant strategy: if  $a_i$  better than  $a_i$  regardless of what other players do,  $a_i$  is **dominant**
- I.e.,

$$u_i(a_i, a_{-i}) \ge u_i(a'_i, a_{-i}) \forall a'_i \ne a_i \text{ and } \forall a_{-i}$$



All of the other entries of *a* excluding *i* 

Doesn't always exist!

## Dominant Strategies Example

#### Back to Prisoner's Dilemma

- Examine all the entries: betray dominates
- Check:

Player 2	Stay silent	Betray
Player 1		
Stay silent	-1, -1	-3, 0
Betray	0, -3	-2, -2

 Note: normal form helps locate dominant/dominated strategies.

## Equilibrium

a\* is an equilibrium if all the players do not have an incentive to unilaterally deviate

$$u_i(a_i^*, a_{-i}^*) \ge u_i(a_i, a_{-i}^*) \quad \forall a_i \in A_i$$

- All players dominant strategies 
   equilibrium
- Converse doesn't hold (don't need dominant strategies to get an equilibrium)

## Pure and Mixed Strategies

So far, all our strategies are deterministic: "pure"

Take a particular action, no randomness

#### Can also randomize actions: "mixed"

• Assign probabilities  $x_i$  to each action

$$x_i(a_i)$$
, where  $\sum_{a_i \in A_i} x_i(a_i) = 1, x_i(a_i) \ge 0$ 

Note: have to now consider expected rewards

## Nash Equilibrium

Consider the mixed strategy  $x^* = (x_1^*, ..., x_n^*)$ 

This is a Nash equilibrium if

$$u_i(x_i^*, x_{-1}^*) \geq u_i(x_i, x_{-i}^*) \quad \forall x_i \in \Delta_{A_i}, \forall i \in \{1, \dots, n\}$$

Better than doing Space of anything else, probability "best response" distributions

 Intuition: nobody can increase expected reward by changing only their own strategy. A type of solution!

## Properties of Nash Equilibrium

#### Major result: (Nash '51)

- Every finite game has at least one Nash equilibrium
  - But not necessarily pure (i.e., deterministic strategy)
- Could be more than one!
- Searching for Nash equilibria: computationally hard!

Example: rock/paper/scissors has (1/3, 1/3, 1/3) as a mixed strategy NE.

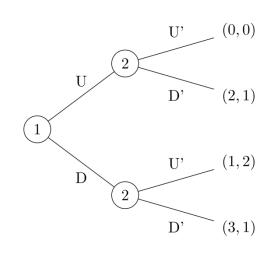


## Sequential Games

#### More complex games with multiple moves

- Instead of normal form, extensive for
- Represent with a tree
- Perform search over the tree

- Can still look for Nash equilibrium
  - Or, other criteria like minimax



## II-Nim: Example Sequential Game

- 2 piles of sticks, each with 2 sticks.
- Each player takes one or more sticks from pile
- Take last stick: lose (ii, ii)
- Two players: Max and Min
- If Max wins, the score is +1; otherwise -1
- Min's score is –Max's
- Use Max's as the score of the game

(ii, ii)

Max takes one stick from one pile

(i, ii)

(ii, ii)

Max takes one stick from one pile

(i, ii)

Min takes two sticks from the other pile

(i,-)

(ii, ii)

Max takes one stick from one pile

(i, ii)

Min takes two sticks from the other pile

(i,-)

Max takes the last stick

(-,-)

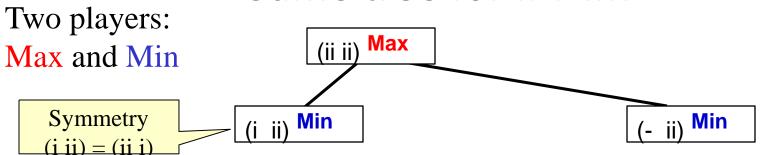
Max gets score -1

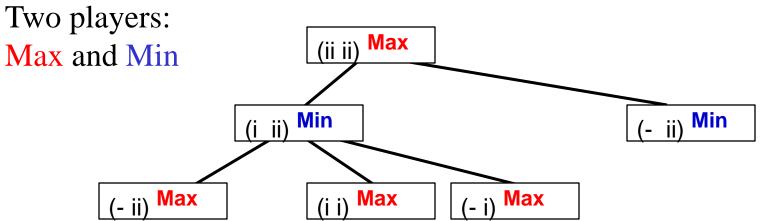
Two players:

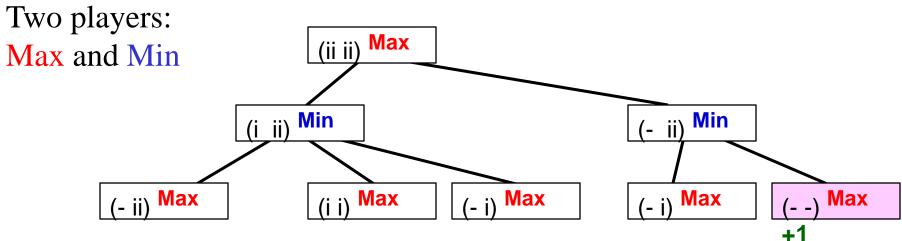
Max and Min

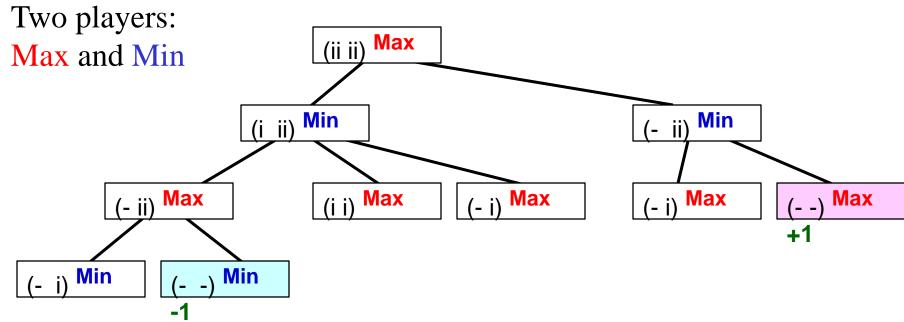


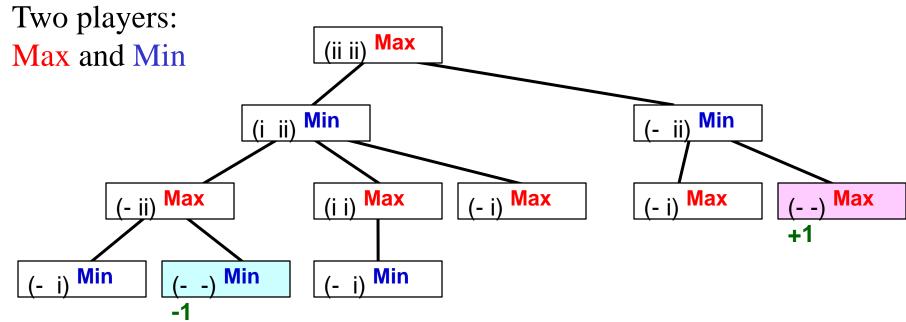
Convention: score is w.r.t. the first player Max. Min's score = -Max

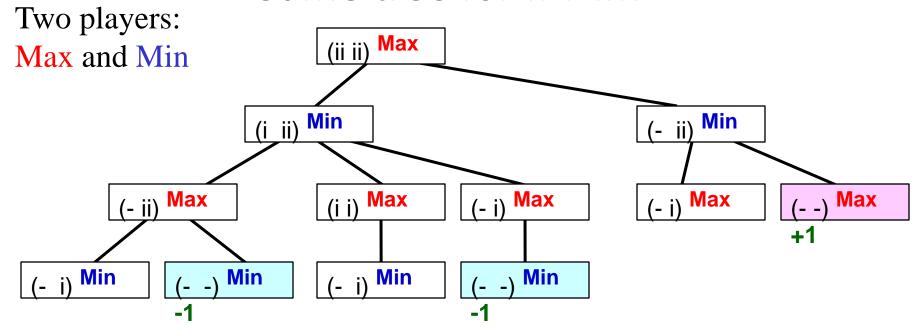


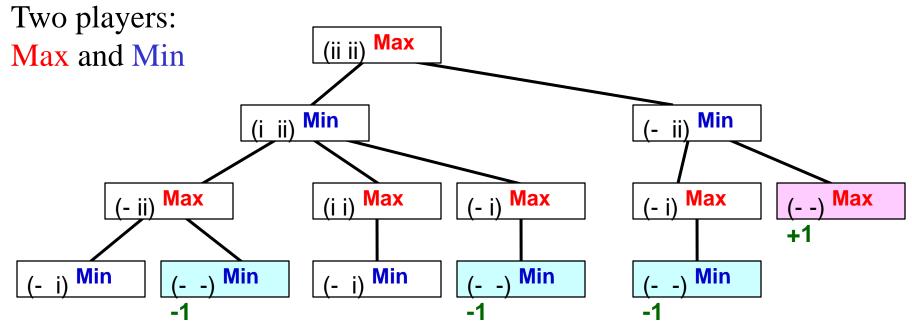


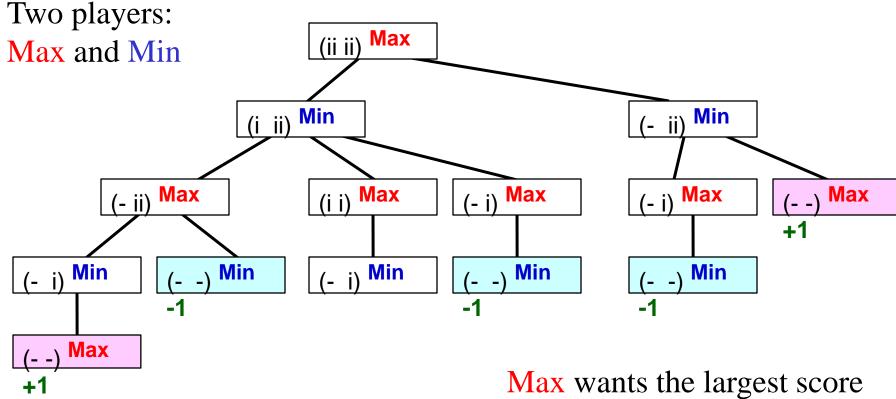


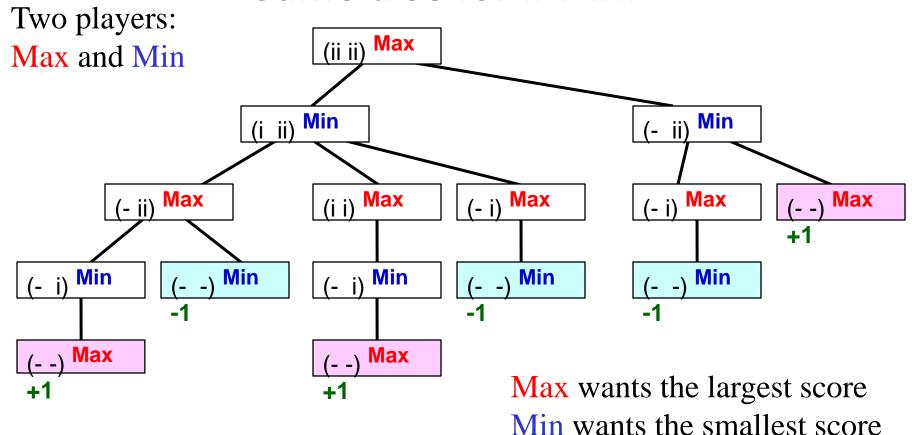












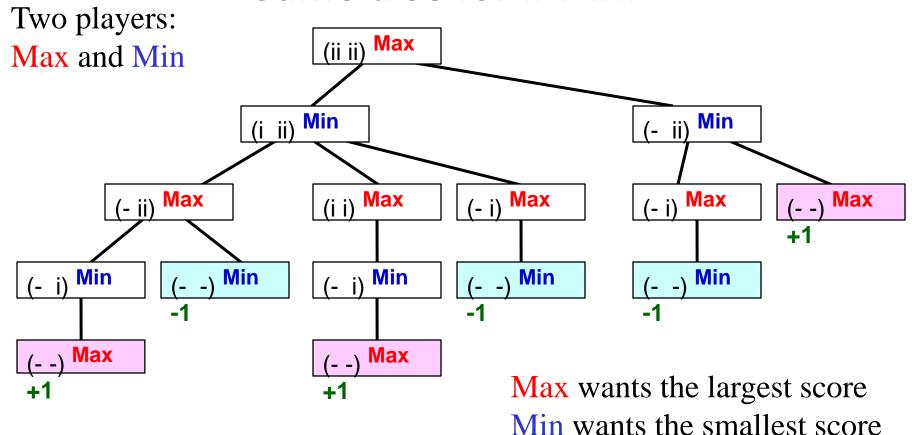
#### Minimax Value

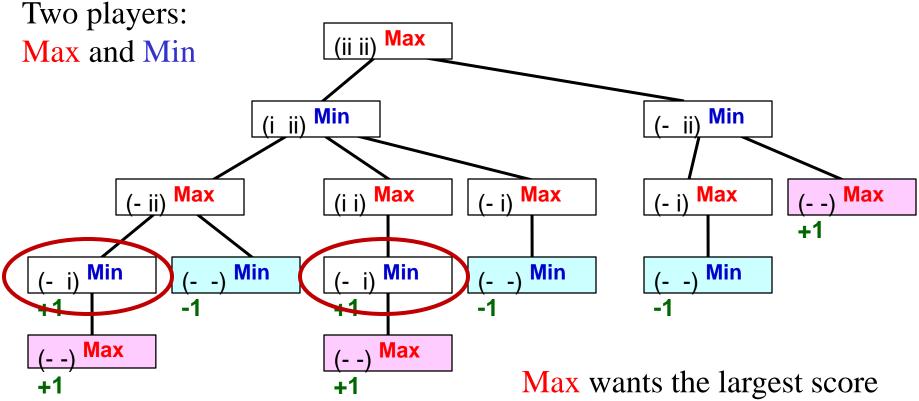
Also called **game-theoretic value**.

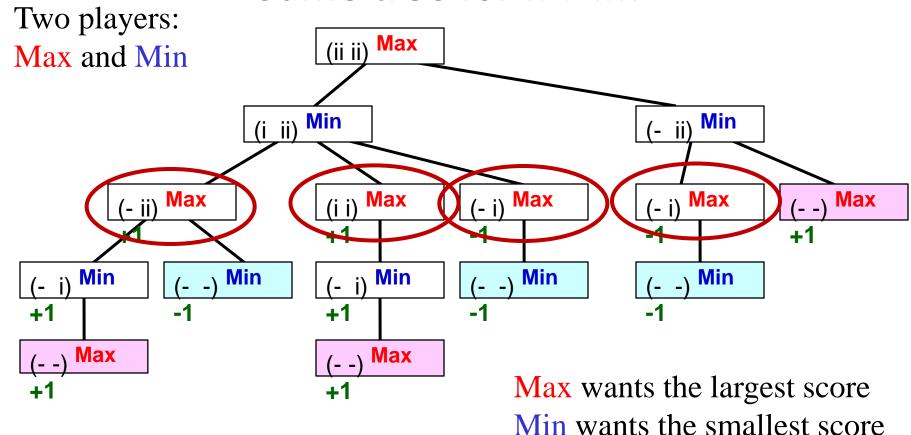
- Score of terminal node if both players play optimally.
- Computed bottom up; basically search

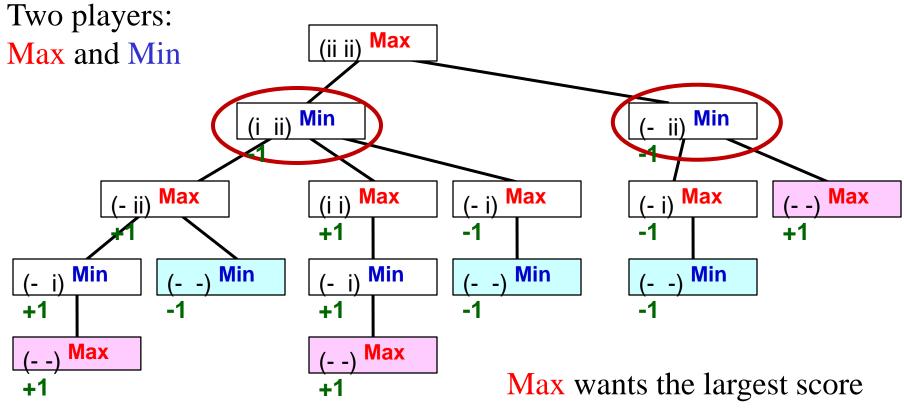
Let's see this for example game

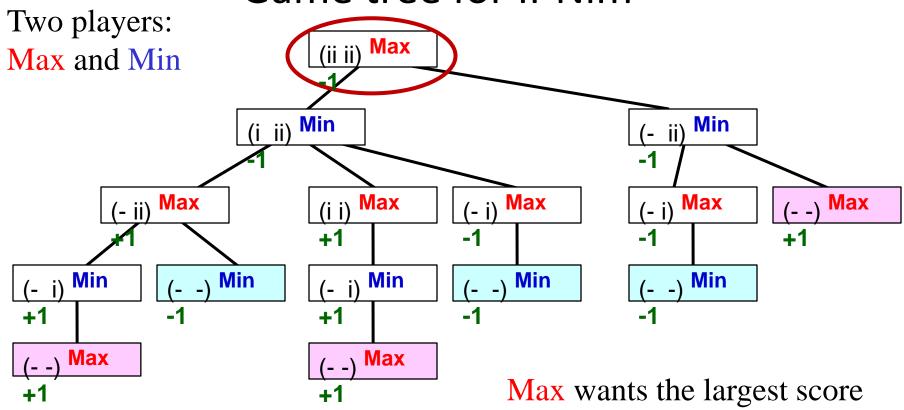


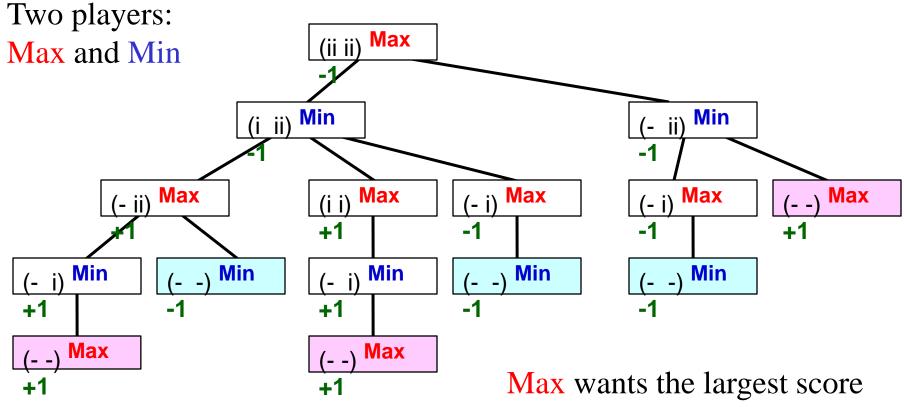


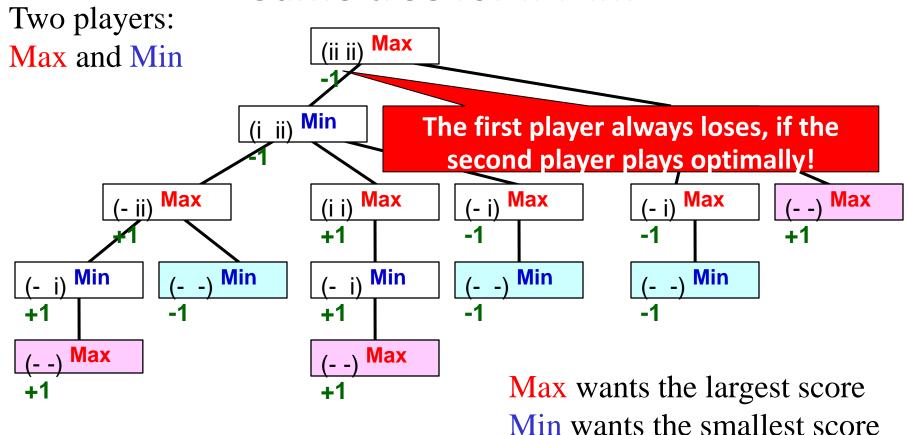












# Summary

- Intro to game theory
  - Characterize games by various properties
- Mathematical formulation for simultaneous games
  - Normal form, dominance, equilibria, mixed vs pure
- Sequential games
  - Game trees, game-theoretic/minimax value



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