

Machine Learning in Practice

#7-1: Combinatorial Games

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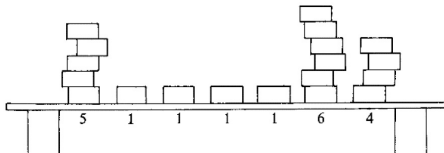
Outline

- 1 **Combinatorial Games**
- 2 Python Implementation

Combinatorial Games: Informal Characterization

2-player **sequential** games with **perfect information**

- e.g.: Nim, Go, chess, checkers (but not poker, tetris, ...)



- There are **2** players who **alternate** moves.
 - ▶ extension: N -player games (e.g. Blokus with $N = 4$)
- There are **no chance** devices (e.g. dice).
- Every information is **known** to both.
 - ▶ information: possible moves, history of game states, ...
- Both players play **optimally** (well-defined?).
 - ▶ Optimal strategies of 2 players are **interdependent**!

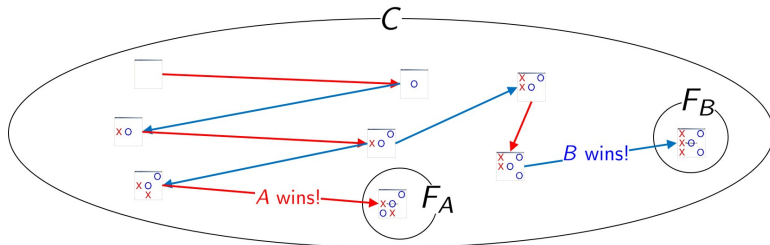
Combinatorial Games: Formulation

A combinatorial game is a tuple (C, F_A, F_B, m_A, m_B) where

- C : the set of all possible **positions** (**configurations**)
- $F_i \subset C$: the set of **ending** positions of player $i \in \{A, B\}$
- $m_i : C \setminus F_{i-1} \rightarrow 2^C$: possible **moves** of player i

Starting from a starting position $c_0 \in C$, the first player that reaches its **ending** position **wins** the game:

$$c^0 \Rightarrow c_A^1 \in m_A(c^0) \Rightarrow c_B^1 \in m_B(c_A^1) \Rightarrow c_A^2 \in m_A(c_B^1) \Rightarrow \dots \Rightarrow c_i^k \in F_i$$



Example: Go

(C, F_A, F_B, m_A, m_B) (player A/B : Black/White)

- $C = \{\text{백}, \text{흑}, \text{무}\}^{[19]^2} (= \{c : [19]^2 \rightarrow \{\text{백}, \text{흑}, \text{무}\}\})$
 - ▶ Under the repetition rule, $C = \{\text{백}, \text{흑}, \text{무}\}^{[19]^2} \times 2^{[19]^2}$
- $F_i = \{c \in C \mid m_A(c) = m_B(c) = \emptyset\}$

and player i 's score is higher at $c\}$

- $m_i(c) = \{\tau(c_{pq}) \mid c(p, q) = \text{무}\}$ where

$$c_{pq}(x, y) = \begin{cases} i & \text{if } (x, y) = (p, q) \\ c(x, y) & \text{otherwise} \end{cases}$$

- ▶ where $\tau : C \rightarrow C$; $(\tau(c))(x, y) =$

$$\begin{cases} \text{무} & \text{if } c(x, y) \text{ is a dead stone} \\ c(x, y) & \text{otherwise} \end{cases}$$

Winning/Losing Positions

- $W_A = \{c \in C \mid \alpha_A(c)\}$ winning positions of player A
- $L_A = \{c \in C \mid \beta_A(c)\}$ losing positions of player A

where

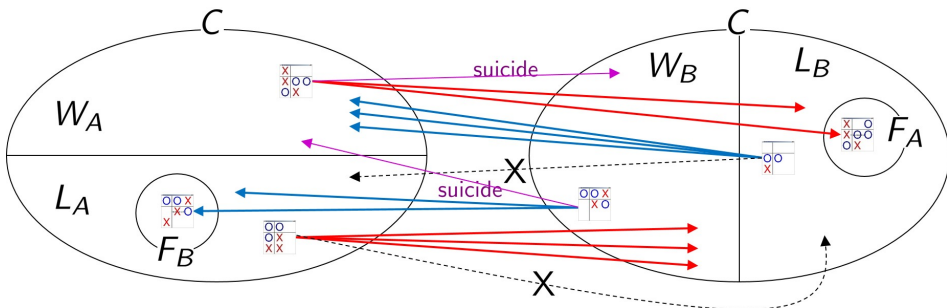
- $\alpha_A(c) = \text{"}\exists c_A^1 \in m_A(c), \forall c_B^1 \in m_B(c_A^1),$
 $\exists c_A^2 \in m_A(c_B^1), \forall c_B^2 \in m_B(c_A^2), \dots,$
 $\exists c_A^k \in m_A(c_B^{k-1}), c_A^k \in F_A\text{"}$
 - ▶ i.e. starting/resuming from c , A can force a win
- $\beta_A(c) = \text{"}\forall c_A^1 \in m_A(c), \exists c_B^1 \in m_B(c_A^1),$
 $\forall c_A^2 \in m_A(c_B^1), \exists c_B^2 \in m_B(c_A^2), \dots,$
 $\forall c_A^k \in m_A(c_B^{k-1}), \exists c_B^k \in m_B(c_A^k), c_B^k \in F_B\text{"}$
 - ▶ i.e. A cannot force a win if B does its best
- $W_A \cap L_A = \emptyset$ can be easily proved by De Morgan's laws.
 - ▶ winning/losing positions are well-defined
- W_B and L_B can be defined similarly.

Winning/Losing Positions \Rightarrow Winning Strategy

Proposition (easily derivable from the definition of W_i/L_i)

- $c \in W_A \iff \exists c' \in m_A(c), c' \in L_B$
- $c' \in L_B \iff \forall c \in m_B(c'), c \in W_A$

- From every **winning** position, a player **can**(\exists) move to the other player's **losing** position
- From every **losing** position, a player **should**(\forall) move to the other player's **winning** position

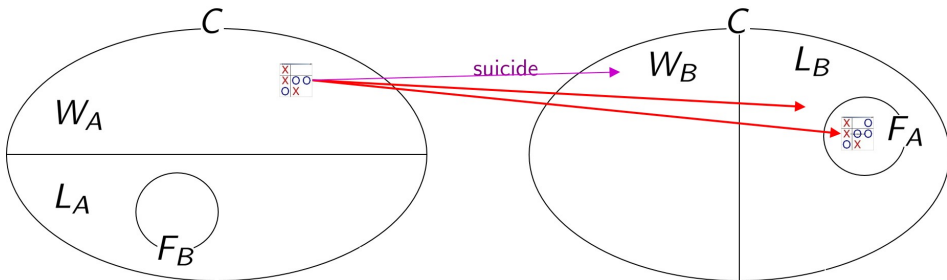


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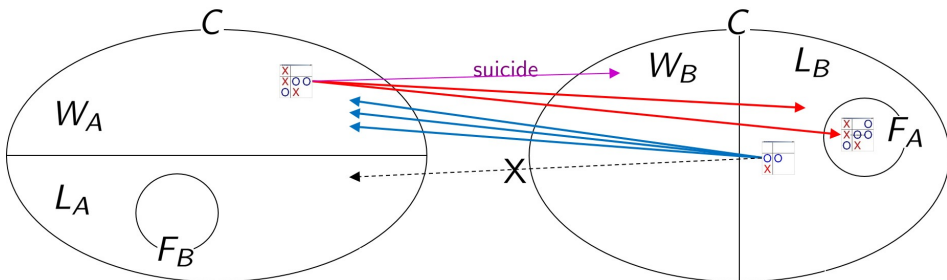


Winning/Losing Positions \Rightarrow Winning Strategy

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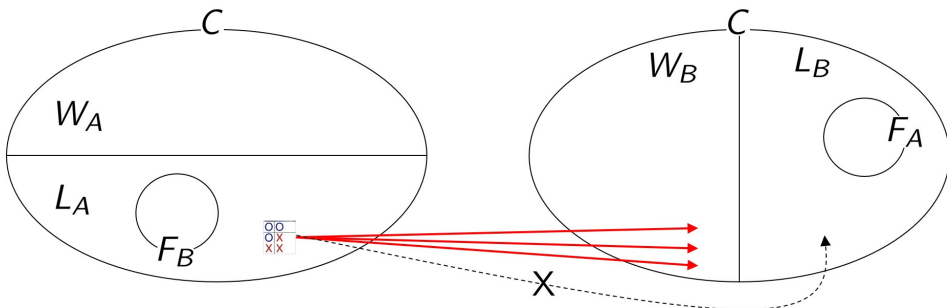
- From every **winning** position, a player **can**(\exists) move to the other player's **losing** position
- From every **losing** position, a player **should**(\forall) move to the other player's **winning** position



Winning/Losing Positions \Rightarrow Winning Strategy**Proposition (easily derivable from the definition of W_i/L_i)**

$$\bullet \quad c' \in L_B \iff \forall c \in m_B(c'), c \in W_A$$

- From every **losing** position, a player **should**(\forall) move to the other player's **winning** position

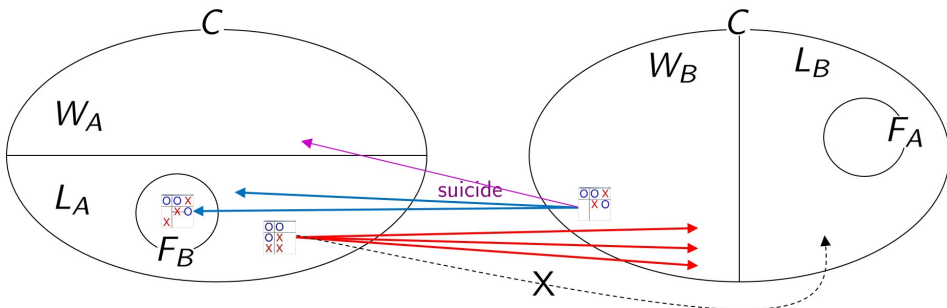


Winning/Losing Positions \Rightarrow Winning Strategy

Proposition (easily derivable from the definition of W_i/L_i)

- $c \in W_A \iff \exists c' \in m_A(c), c' \in L_B$
- $c' \in L_B \iff \forall c \in m_B(c'), c \in W_A$

- From every **winning** position, a player **can**(\exists) move to the other player's **losing** position
- From every **losing** position, a player **should**(\forall) move to the other player's **winning** position



How to Compute Winning/Losing Positions & Winning Strategy

Recursive Definition of Winning/Losing Positions

- $W_A = \{c \in C \mid \exists c' \in m_A(c), c' \in L_B\}$
- $L_B = \{c' \in C \mid \forall c \in m_B(c'), c \in W_A\}$
- $W_B = \{c' \in C \mid \exists c \in m_B(c'), c \in L_A\}$
- $L_A = \{c \in C \mid \forall c' \in m_A(c), c' \in W_B\}$
- $F_A \subseteq L_B$
- $F_B \subseteq L_A$

```
def in_WA(c):  
    if c ∈ FB:  
        return False  
    for each c' ∈ mA(c):  
        if in_WB(c') = False:  
            return True  
    return False
```

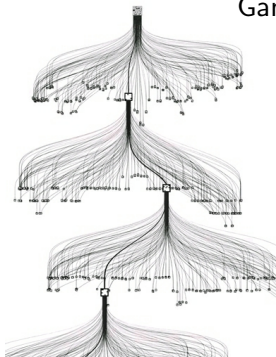
```
def in_WB(c'):  
    if c' ∈ FA:  
        return False  
    for each c ∈ mB(c'):  
        if in_WA(c) = False:  
            return True  
    return False
```

Exact Exhaustive Algorithm

```
def in_WA(c):
    if c ∈ F_B:
        return False
    for each c' ∈ m_A(c):
        if in_WB(c') = False:
            return True
    return False
```

```
def in_WB(c'):
    if c' ∈ F_A:
        return False
    for each c ∈ m_B(c'):
        if in_WA(c) = False:
            return True
    return False
```

Game-tree of Go: about $5 \cdot 10^{359}$ nodes!



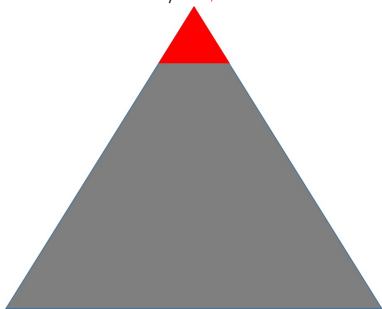
- The above algorithm is exact, but impractical

Inexact Fast Heuristics: Search Space Pruning (to be covered later)

```
def max_A(c, depth):  
    if depth ≥ dth:  
        return value(c)  
    for each  $c' \in m_A(c)$ :  
         $v := \min\_B(c', \text{depth}+1)$   
         $\text{max\_v} := \max\{\text{max\_v}, v\}$   
    return max_v
```

```
def min_B(c', depth):  
    if depth ≥ dth:  
        return value(c)  
    for each  $c \in m_B(c')$ :  
         $v := \max\_A(c, \text{depth}+1)$   
         $\text{min\_v} := \min\{\text{min\_v}, v\}$   
    return min_v
```

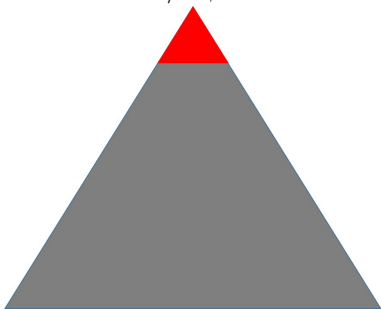
Minimax/ α - β Search



Inexact Fast Heuristics: Search Space Pruning (to be covered later)

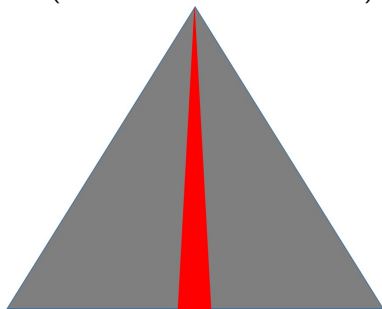
```
def max_A(c, depth):  
    if depth ≥ dth:  
        return value(c)  
    for each c' ∈ mA(c):  
        v := min_B(c', depth+1)  
        max_v := max{max_v, v}  
    return max_v
```

Minimax/ α - β Search



```
def min_B(c', depth):  
    if depth ≥ dth:  
        return value(c)  
    for each c ∈ mB(c'):  
        v := max_A(c, depth+1)  
        min_v := min{min_v, v}  
    return min_v
```

MCTS (Monte-Carlo Tree Search)



Any Hope for Solving Game Exactly in Reasonable Time?

- Most impartial games can be solved in polynomial-time
 - ▶ They are poly-time reducible to Nim by Sprague-Grundy thm
 - ▶ Nim is solvable in poly-time
- Most partizan games (e.g. Go, chess) are provably hard
 - ▶ PSPACE-hard, in general
- Some (artifact) partizan games can be solved in poly-time
 - ▶ Those that appear in math/programming competitions
 - ▶ For them, the boundaries bet'n winning/losing positions that are computationally easy are derivable by math analysis
- Small-sized instances of hard games can be solved in reasonable time
 - ▶ e.g. 3×3 Go

Outline

- 1 Combinatorial Games
- 2 Python Implementation**

How to (exhaustively) compute the winner and a winning strategy?

- $F_B \subseteq L_A$, $F_A \subseteq L_B$
- $W_A = \{c \in C \mid \exists c' \in m_A(c), c' \in L_B\}$
- $L_A = \{c \in C \mid \forall c' \in m_A(c), c' \in W_B\}$

Exhaustive bottom-up algorithm

(exponential space)

```
 $L_A := L_B := F;$      $W_A := W_B := \{\}$   
do {  
     $W_A := W_A \cup \{c \mid m_A(c) \cap L_B \neq \emptyset\}$   
     $W_B := W_B \cup \{c \mid m_B(c) \cap L_A \neq \emptyset\}$   
     $L_A := L_A \cup \{c \mid m_A(c) \subseteq W_B\}$   
     $L_B := L_B \cup \{c \mid m_B(c) \subseteq W_A\}$   
} while (there is a change in one of  $W_A, L_A, W_B, L_B$ )  
  
if (starting position is in  $L_A$ )  
    A is the winner  
else  
    B is the winner
```

Exhaustive top-down recursive algorithm

(poly-space)

- $F_B \subseteq L_A$, $F_A \subseteq L_B$
- $W_A = \{c \in C \mid \exists c' \in m_A(c), c' \in L_B\}$
- $L_A = \{c \in C \mid \forall c' \in m_A(c), c' \in W_B\}$

```
def AcanWin(c): # True iff  $c \in W_A$ 
    if  $c \in F_B$ : return False

    for each  $c' \in m_A(c)$ :
        if not BcanWin( $c'$ ):
            return True      #  $c \rightsquigarrow c'$  guarantee A's winning
    return False # whichever position A selects, A loses

def BcanWin(c): # True iff  $c \in W_B$ 
    if  $c \in F_A$ : return False

    for each  $c' \in m_B(c)$ :
        if not AcanWin( $c'$ ):
            return True      #  $c \rightsquigarrow c'$  guarantee B's winning
    return False # whichever position B selects, B loses

# A is the winner iff AcanWin(starting_position) is True
```

Exhaustive top-down recursive algorithm for impartial games

(C, F_A, F_B, m_A, m_B) is called an **impartial** game (대칭 게임) if

- $F_A = F_B$ and $m_A = m_B$

Otherwise, it is called a **partizan** game (비대칭 게임)

```
def canWin(c): # True iff  $c \in W_A (= W_B)$ 
    if  $c \in F_B (= F_A)$ : return False
    for each  $c' \in m_A(c) (= m_B(c))$ :
        if not canWin(c'):
            return True #  $c \rightsquigarrow c'$  guarantee winning
    return False # whichever position selects, it loses

# A is the winner iff canWin(starting_position) is True
```

Rules of Go

- <http://gall.dcinside.com/board/view/?id=baduk&no=30064>
- 남은 수업 기간 동안 규칙을 단순화 시킨 5×5 바둑 문제를 다루므로 위 규칙을 어느 정도 이해는 하고 있어야 함
 - ▶ 규칙 자체는 구현 완료된 형태로 제공됨
- 다음 수업: Monte-Carlo tree search & minimax/ α - β search
 - ▶ Exhaustive search에서 width/height를 가지치기(pruning)한 것으로 오늘 내용을 완벽히 숙달해야 함
- 다음² 수업: Monte-Carlo policy iteration
- 최종과제: MCTS/ α - β + 강화학습
 - ▶ 폴리그전 결과로 점수 결정