

Nim, part 2

CMPUT 355: Games, Puzzles, and Algorithms

Lecture Outline

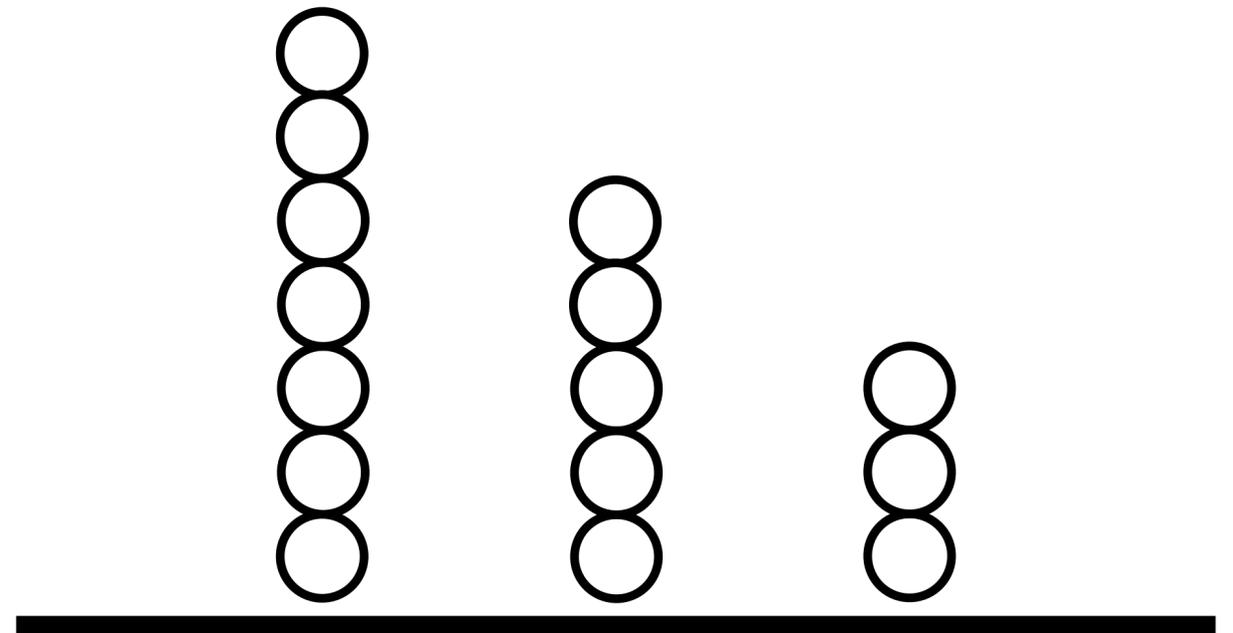
1. Logistics & Recap
2. Dynamic programming
3. Nim Formula & Theorem

Logistics

- **Quiz 2** marks are **released** on Canvas
 - A scan of your quiz is attached to the Canvas assignment
- **Practice questions #3** are available
 - Solutions released last Friday
- **Quiz 3** is **this Friday** (Feb 27)
 - *Coverage:* up to and including **Feb 13** (Move Ordering & Nim)
 - Bring your student ID!
 - No calculators or other devices
- **TA Office hours:** every Thursday 1pm-2pm in **UCOMM-3-136**

Recap: Nim

- There are k piles of stones
- A move is removing some **positive** number of stones from a **single pile**
- Players alternate moves
- Last player to move wins
 - i.e., if there are **no stones left** and you are player-to-move, you lose
- Example at right is position $7,5,3$
- Every position is either **winning** (i.e., player-to-move can force a win from here) or **losing** (every child position is a winning position for the opponent)



Dynamic Programming

- **Dynamic programming:** Solve a problem by first solving **overlapping subproblems**
 - **overlapping:** used more than once
 - so it's useful to store subproblem solutions rather than re-solving them
- **Memoization** (caching) is an easy way to speed up a recursive function with overlapping subproblems

```
def factorial(n):  
    if n == 0:  
        return 1  
    else:  
        return n * factorial(n - 1)
```

```
def fib(n):  
    if n <= 1:  
        return n  
    return fib(n - 2) + fib(n - 1)
```

```
def fib(n, D):  
    if n in D:  
        return D[n]  
    elif n <= 1:  
        return 1  
    else:  
        f = fib(n-1, D) + fib(n-2, D)  
        D[n] = f  
        return f
```

Questions:

1. Which of these two functions exhibit overlapping subproblems?
2. How long will `factorial(N)` take?
3. How long will `fib(N)` take?
4. How would each change if we **cached** the results?

Dynamic Programming in Nim

- Determining whether a Nim position is **winning** or **losing** is a dynamic programming problem
- We have **already used memoization** to solve this problem (**how?**)
- The recursion-with-memoization approach to dynamic programming has some drawbacks (**what?**)
- **Bottom-up approach:**
Solve the subproblems in a specific order such that you have already solved subproblem A,B,C,... before any other subproblems that depend on those solutions
- **Question:** What would that look like in the context of Nim?

Questions:

1. What are the **subproblems** in Nim?
2. Why are they **overlapping**?
3. Is the win/lose/draw computation in **Tic-tac-toe** a dynamic programming problem?
Why or why not?

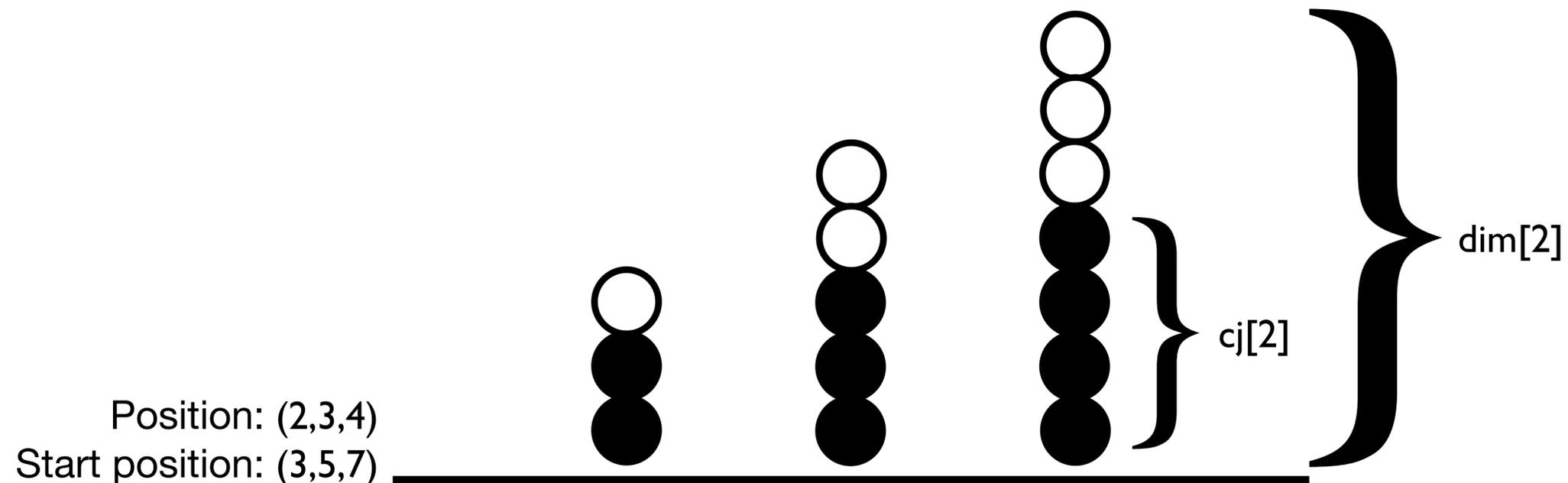
Canonical Representations

- Recall that we can convert any Nim position into a **canonical representation** just by **sorting** the piles in order of size
 - E.g., $(7,3,4) \rightarrow (3,4,7)$
- Just like in tic-tac-toe, we can then assign a **number** to each representative
 - E.g.1: $(3,4,7) \rightarrow 347$
 - E.g.2: $(3,4,7) \rightarrow 3 + 4 \times (3 + 1) + 7 \times ((3 + 1) \times (5 + 1)) = 120$
 - (assumes starting position of $(3,5,7)$) (**why?**)
- **Claim:** every child position will have a **smaller number** than its parent (**why?**)
- So if we compute subproblems in **increasing numeric order**, then every subproblem will be computed **before we need it**

Dynamic Programming Implementation: nim/nim.py

```
def solveall():
    # for each losing state, find winning states that reach it
    for j in range(len(self.wins) - 1): # nothing reaches starting state
        if not self.wins[j]: # loss, so find all psns that reach j
            cj = self.crd(j) # convert index to list of pile sizes
            for x in range(len(cj)):
                cscopy = deepcopy(cj)
                for t in range(1 + cj[x], 1 + self.dim[x]): # `dim` is starting pile

                    cscopy[x] = t
                    pjc = self.psn(cscopy) # convert pile sizes to index
                    self.wins[pjc], self.winmove[pjc] = True, j
```



XorSum

- **xorsum** operation: Cumulative bitwise XOR of **each binary digit** of the pile sizes

- Examples:

- $\text{xorsum}(1,2,3) = 00$

$$\begin{array}{l} 1 = 01 \\ 2 = 10 \\ 3 = 11 \end{array} \quad \text{and} \quad \begin{array}{l} 0 \oplus 1 \oplus 1 = 0 \\ 1 \oplus 0 \oplus 1 = 0 \end{array}$$

- $\text{xorsum}(3,5,3) = 101b = 5$

$$\begin{array}{l} 3 = 011 \\ 5 = 101 \\ 3 = 011 \end{array} \quad \text{and} \quad \begin{array}{l} 0 \oplus 1 \oplus 0 = 1 \\ 1 \oplus 0 \oplus 1 = 0 \\ 1 \oplus 1 \oplus 1 = 1 \end{array}$$

Nim Formula Theorem

Theorem: A Nim position with pile sizes p_1, \dots, p_k is **losing** iff

$$\text{xorsum}(p_1, \dots, p_k) = 0.$$

Proof sketch:

1. If $\text{xorsum}(P) = 0$, then $\text{xorsum}(C) \neq 0$ for **every** child C of P
2. If $\text{xorsum}(P) \neq 0$, then $\text{xorsum}(C) = 0$ for **some** child C of P
3. $\text{xorsum}(0, 0, \dots, 0) = 0$
4. Any position that has $(0, 0, \dots, 0)$ as a child is a **winning** position

Nim Formula Theorem Proof: Step 1

1. If $\text{xorsum}(P) = 0$, then $\text{xorsum}(C) \neq 0$ for every child C of P
 - (a) Consider any pile j with $p_j > 0$
 - (b) Then $\text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k) = p_j$ (**why?**)
 - (c) So if you remove any stones from pile j , $p'_j \neq p_j$, and so $p_j \oplus p'_j \neq 0$
 - (d) $\text{xorsum}(p_1, \dots, p_{j-1}, p'_j, p_{j+1}, \dots, p_k) = \text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k) \oplus p'_j = p_j \oplus p'_j \neq 0$
 - (e) But every child of (p_1, \dots, p_k) is made by removing some stones from a single pile j



Nim Formula Theorem Proof Step 2

2. If $\text{xorsum}(P) \neq 0$, then $\text{xorsum}(C) = 0$ for **some** child C of P

(a) Let d be the digit number of the most significant 1 in the binary representation of $\text{xorsum}(P)$

(b) There exists at least one j such that p_j has a 1 in digit d (**why?**)

(c) $\text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k)$ has a 0 in the d -th column (**why?**)

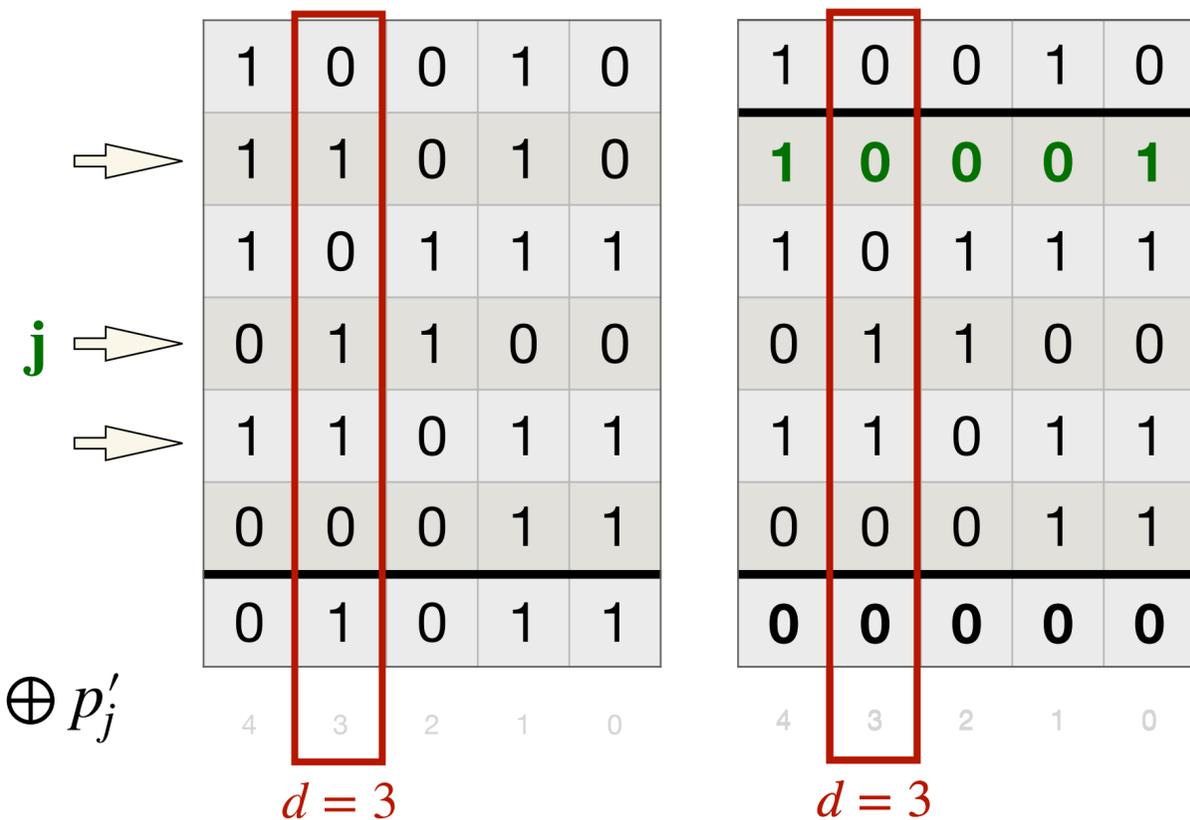
- and also in every more-significant column that p_j has a 0 in (**why?**)

(d) So $\text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k) < p_j$ (**why?**)

(e) Set $p'_j = \text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k)$

$$\text{xorsum}(p_1, \dots, p_{j-1}, \mathbf{p}'_j, p_{j+1}, \dots, p_k) = \text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k) \oplus p'_j$$

$$\begin{aligned} \text{(f)} \quad &= p'_j \oplus p'_j \\ &= 0 \end{aligned}$$



Nim Formula Theorem Proof Step 5

1. If $\text{xorsum}(P) = 0$, then $\text{xorsum}(C) \neq 0$ for **every** child C of P
2. If $\text{xorsum}(P) \neq 0$, then $\text{xorsum}(C) = 0$ for **some** child C of P
3. $\text{xorsum}(0,0,\dots,0) = 0$
4. Any position that has $(0,0,\dots,0)$ as a child is a **winning** position

Inductive hypothesis: suppose that for all positions P within k steps of $(0,\dots,0)$,
($\text{xorsum}(P) \neq 0 \implies P$ is winning) and ($\text{xorsum}(P) = 0 \implies P$ is losing)

Then for all P that are within $k + 1$ steps of $(0,\dots,0)$:

(a) $\text{xorsum}(P) \neq 0$:

- by (2), $\text{xorsum}(C) = 0$ for **some** child C
- C is within k of $(0,\dots,0)$, so by IH C is **losing**
- So P has **at least one losing child**, and is therefore **winning**

(b) $\text{xorsum}(P) = 0$:

- by (1), $\text{xorsum}(C) \neq 0$ for **all** children C
- every C is within k of $(0,\dots,0)$, so by IH every C is **winning**
- So **all of P 's children are winning**, hence P is **losing**

Base case:

(a) By (3), every position P within 0 of $(0,\dots,0)$ has $\text{xorsum}(C) = 0$, and by definition it is a losing position

(b) By (1), every position P within 1 of $(0,\dots,0)$ has $\text{xorsum}(C) \neq 0$, and by (4) they are all winning ■

Using the Nim Formula

- **Checking** if a position P is **winning** or **losing**: Simply compare $\text{xorsum}(P)$ to 0
- Finding a **winning move**:
 1. Compute $\text{xorsum}(P)$
 2. Find d and a pile j with a 1 in digit d
 - Equivalently: a pile j with $p_j \geq \text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k)$
 3. Set $p'_j = \text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k)$
 4. Take $p_j - p'_j$ stones from pile j
 - i.e., move to position $(p_1, \dots, p_{j-1}, \mathbf{p'_j}, p_{j+1}, \dots, p_k)$

Nim Formula Implementation: nim/nim.py

```
def xorsum(L):  
    xsum = 0  
    for j in L:  
        xsum ^= j  
    return xsum
```

($x \wedge y$) is the Python syntax for ($x \oplus y$)

```
def nimreport(P): # report all winning nim moves from P, use formula  
    total = xorsum(P)  
    if total==0:  
        print(' loss')  
        return  
    for j in P:  
        tj = total^j  
        if j >= tj:  
            print(' win: take', j - tj, 'from pile with', j)
```

Note: $\text{xorsum}(p_1, \dots, p_{j-1}, p_{j+1}, \dots, p_k) = \text{xorsum}(p_1, \dots, p_k) \oplus p_j$ (why?)

Summary

- Nim can be solved using **dynamic programming**
 - recursion with memoization (i.e., recursion with transposition table)
 - bottom-up
- The **Nim formula** allows us to **directly compute** whether a Nim move is **winning**
 - *Without* having to solve all the subpositions!
- Can directly compute a **winning move** from any winning position
 - Again, without having to solve subpositions
 - Or even consider every possible child