

CS 331, Fall 2024  
Lecture 6 (9/16)

Today: - Knapsack  
- LCS & friends  
- Palindromes  
- Game theory

## Knapsack (Part III, Section 3.3)

Recall **subset sum** from last time:

Input:  $L$ : list of  $n$  natural #'s

$V \in \mathbb{N}$ : target value



Output: True/False,  $\exists S \subseteq \{n\}, \sum_{i \in S} L[i] = V?$

Key idea:  $\omega$  memoization, prefix x value

General strategy applies to integer-constrained optimization problems, e.g. knapsack

Input:  $W$ :  $n$  natural #'s

$B \in \mathbb{N}$ : target value

Subset  
Sum

Output:  $\exists S \subseteq [n], \sum_{i \in S} W(i) = B?$

Input:  $W$ :  $n$  natural #'s (weights)

$B \in \mathbb{N}$ : weight budget

0-1  
knapsack

$V$ :  $n$  real positive #'s (values)

Output:  $\max_{S \subseteq [n]} \sum_{i \in S} V(i)$

$S \subseteq [n]$

$\sum_{i \in S} W(i) \leq B$

Applications: budget-constrained decision making problems.

Before:  $S(j)(b)$  = Can you hit target  $b$   
 $j \in (n), b \in (B)$  using first  $j$  items?

Intuition: decision to take/not take  $L(j)$   
affects remaining value, must store

Now:  $S(j)(b)$  = Max value w/ weight  
 $j \in (n), b \in (B)$  budget  $b$  using first  $j$  items

DP formula:

$$S(j)(b) = S(j-1)(b) \quad \text{subset sum}$$

OR  $S(j-1)(b-w(j))$

$$S(j)(b) = \max \left( S(j-1)(b), V(j) + S(j-1)(b-w(j)) \right) \quad \text{0-1 knapsack}$$

Also runs in  $O(nB)$  time. (row-by-row)

**Extension** unbounded knapsack

You can take multiple copies of any item.

Goal: maximize  $\sum_{i \in \mathcal{I}} c_i V(i)$  s.t.

$$c_i \in \mathbb{Z}_{\geq 0}^n \text{ (counts)}$$

$$\sum_{i \in \mathcal{I}} c_i W(i) \leq B$$

DP:  $S(b) = \max$  achievable w/ budget  $b$

$$S(b) = \max \left( \underbrace{0}_{\text{take nothing}}, \max_{\substack{i \in \mathcal{I} \\ W(i) \leq b}} \underbrace{S(b - W(i)) + V(i)}_{\text{take item } i} \right)$$

Runtime:  $O(B) \times O(n) = O(nB)$

# subprobs      time / subprob

# Longest Common Subsequence (Part II, Section 4.1)

Strings

$\Omega = \text{universe}$

e.g.  $\{ 'a', 'b', \dots, 'z', '1', \dots, '0' \}$

characters

String of length  $n$ : ordered list of  $n$  characters from  $\Omega$

"algorithms"  
 $\equiv \{ 'a', 'l', \dots, 'm', 's' \}$

Substring: contiguous sublist

"algo"

Subsequence: delete any characters, concatenate the rest

"alms"

LCS: natural distance measure on strings

Input:  $X$ , length- $m$  string (e.g. DNA)  
 $Y$ , length- $n$  string

Output:  $|Z|$ , largest possible length of common subsequence  $Z$  of  $X$  &  $Y$

Example

$X =$  "algorithms"

$Y =$  "complexity"

$$\text{LCS}(X, Y) = 3$$

Conclusion: they are both "lit"  
 $\arg\max_Z |Z|$

Key idea: 2-D DP again

$$S(i, j) = \text{LCS}(X(i), Y(j))$$

↑ prefixes ↑

2 cases: Can we match  $X(i), Y(j)$ ?

Case 1: No ( $X(i) \neq Y(j)$ )

e.g. "algor", "comp" ( $i=5, j=5$ )

What is last char of  $Z$ ?

$X(i), Y(j)$ , or neither. (not both)

If not  $X(i)$ , Plan A:  $S(i-1, j)$

If not  $Y(j)$ , Plan B:  $S(i, j-1)$

Case 2: Yes ( $X[i] = Y[j]$ )

e.g. "a", "comp" ( $i=2, j=5$ )

Now we can make  $X[i] = Y[j]$

last character in Z.

$$\text{Plan C: } 1 + S[i-1][j-1]$$

Summary:

$$S[i][j] = \max \left( \begin{array}{l} S[i-1][j], \quad \text{no } X[i] \\ S[i][j-1], \quad \text{no } Y[j] \end{array} \right)$$

Runtime:

$$O(mn)$$

$$+ \left( \begin{array}{l} S[i-1][j-1] \\ + \underline{1} (X[i] = Y[j]) \end{array} \right) \text{ both}$$



Extension

Multiple strings

$LCS(W, X, Y)$ : longest mutually  
common subsequence  
lengths  $l$   $m$   $n$

Same idea.

$$\begin{aligned} S(i, j, k) &= LCS(W(i), X(j), Y(k)) \\ &= \max \left( S(i-1, j, k), S(i, j-1, k), \right. \\ &\quad \left. S(i, j, k-1), S(i-1, j-1, k) \right. \\ &\quad \left. + 1 \left( W(i) = X(j) = Y(k) \right) \right) \end{aligned}$$

Extension Edit distance

How many ops needed to turn  $X$  to  $Y$ ?

Ops: Insert, Delete, Substitute

Example

$X = \text{"kitten"}$

$Y = \text{"sitting"}$

3 steps: "kitten"  $\rightarrow$  "kittin"  
 $\rightarrow$  "sittin"  $\rightarrow$  "sitting"

Observation: Suppose no substitutions.

Optimal edit sequence:

$X \xrightarrow{\text{(deletions)}} Z$   
 $Z \xrightarrow{\text{(insertions)}} Y$   
 $Z$  is LCS of  $X$  &  $Y$

Proof: 1) All deletions from  $X$  in optimal moves.

2) Rearrange so all deletions first.

3)  $X \rightarrow Z \rightarrow Y$  shortest if  $Z$  longest.

Edit distance: Same idea.

- All moves are
- 1) Delete from  $X$
  - 2) Insert from  $Y$
  - 3) Substitute  $X$  to  $Y$

or sequence can be made shorter.

$S(i, j) =$  Edit distance of  $X(1:i), Y(1:j)$

$$S(i, j) = \min \left( \begin{array}{l} S(i, j-1) + 1 \quad \text{insert } Y(j) \\ S(i-1, j) + 1 \quad \text{delete } X(i) \\ S(i-1, j-1) + \mathbb{1}(X(i) \neq Y(j)) \quad \text{sub } X(i) \rightarrow Y(j) \\ \quad \text{not necessary if equal} \end{array} \right)$$

Runtime: Again  $O(mn)$ .

# Longest palindromic substring (Part II, Section 4.2)

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Palindrome: "RACECAR" (odd length)

↑  
center

"TATTARRATTAT" (even length)

↖ ↗  
center x 2

Input: String  $X$

Output:  $|Z|$ , largest possible length of palindromic substring  $Z$  of  $X$

Example

$X =$ "bananana"  
                ↑  
                longest possible  $Z$

$X =$ "bananaana"  
                ↑  
                longest possible  $Z$

Idea 1: DP

To determine whether  $X[i:j]$  is palindrome  
 $= S[i](j)$

- either:
- just check  $(j \leq i+1)$
  - $S[i](j) = S[i+1](j-1)$   
AND  $(X[i] == X[j])$   
 $O(n^2)$  time

Idea 2 (or 1?): center growing

1) guess center  $c$  or  $cc$   $O(n)$

2) grow **left** & **right** pointers until exit  $\times O(n)$   
 $= O(n^2)$

Seems to match DP.

Manacher's algo:  $O(n)$  (see notes)

# Game theory (Part III, Section 5.1)

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Consider two-player win-lose game.

Alice vs. Bob. E.g. Tic-tac-toe, chess, ...

Move 1 Alice

Move 2 Bob

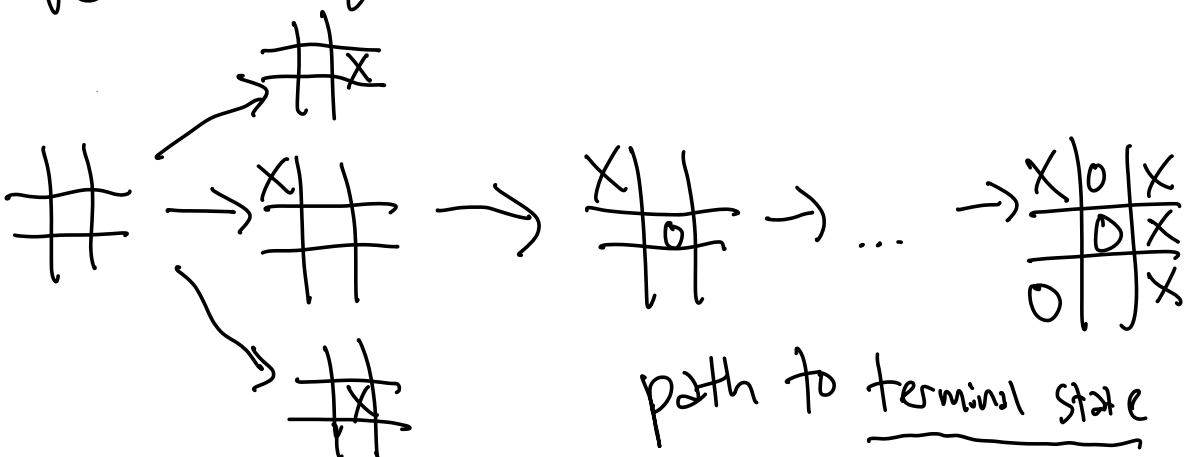
⋮

Move k (game over, Alice wins)

**Game graph**

Vertices: game states

(potentially huge,  
pruning in practice)



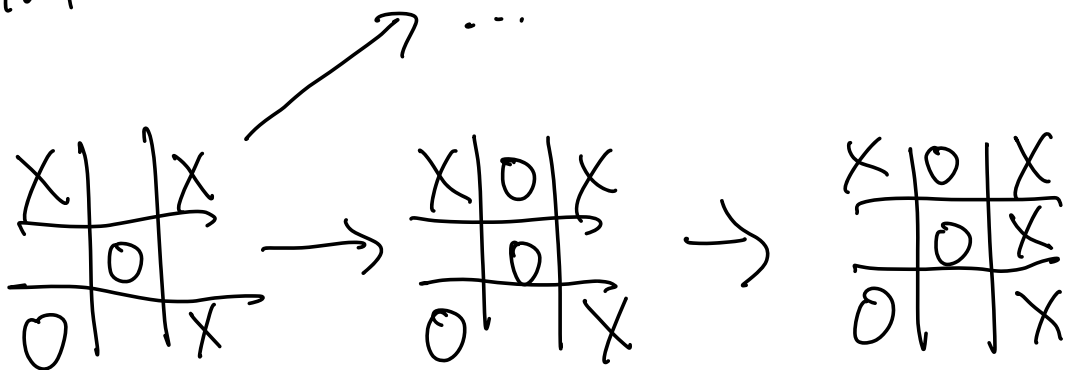
Terminal state: Vertex where game is over.

- 1) Alice wins
- 2) Bob wins
- 3) Tie (we'll mostly ignore)

Game moves: directed edges  $\# \rightarrow \#$

Q: Can Alice always force a win?

Intuition:



"forced win"

Whatever Bob does, Alice can win.

DP solution: label all vertices  $v$  True "forced win"  
False

Work backwards from leaves  $\equiv$  terminal states  
(next class) (with ties, label them False.)

$$S(v) = \begin{cases} \text{OR}_{\text{edge}(v,u)} (S(u)) & \text{Alice plays} \\ \text{AND}_{\text{edge}(v,u)} (S(u)) & \text{Bob plays} \end{cases}$$

Alice wins if she can move to another winning state.

Bob wins if he can move to another losing state.

So Alice needs all possible moves to be True.



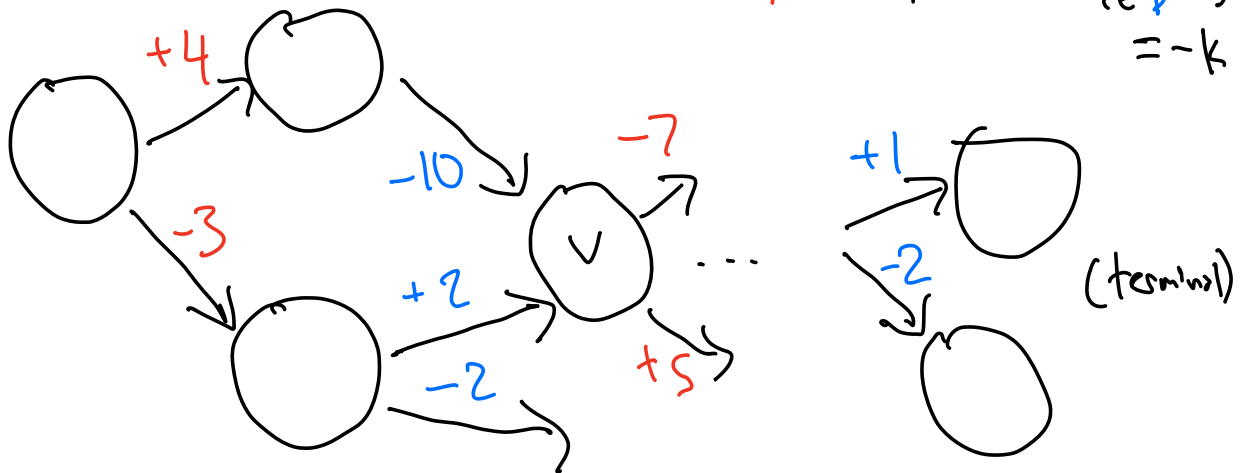
# Extension Zero-sum games

Alice and Bob have scores.

(2) End of game, sum = 0.

e.g. win-lose (score at end is Winner +1, Loser -1)

dividing items ( $V_1 + \dots + V_n = T$  Split to A, B)  
 Score:  $\sum_{i \in A} V_i = k, -T + \sum_{i \in B} V_i = -k$



$$S(v) = \begin{cases} \text{Max} & S(u) + \text{score change } (v,u) & \text{Alice} \\ \text{Min} & S(u) + \text{score change } (v,u) & \text{Bob} \end{cases}$$

Max guaranteed score if dropped in at state v.