

# Lecture14

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## CSE 421, Introduction to Algorithms

Lecture 14, Winter 2024

Dynamic Programming

Subset Sum, Longest Common  
Subsequence

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# Announcements

- **Dynamic Programming Reading:**
  - 6.1-6.2, **Weighted Interval Scheduling**
  - 6.3 **Segmented Least Squares**
  - 6.4 Knapsack and Subset Sum
  - 6.6 String Alignment
  - 6.8 Shortest Paths (Bellman-Ford)
- **Midterm, Friday, Feb 9**
  - Material through 6.3 and HW 5
  - Feb 8 Section will be Midterm review

What is the largest sum you can make of the following integers that is  $\leq 20$

{4, 5, 8, 10, 13, 14, 17, 18, 21, 23, 28, 31, 37}

What is the largest sum you can make of the following integers that is  $\leq 2000$

{78, 101, 122, 133, 137, 158, 189, 201, 220, 222, 267, 271, 281, 289, 296, 297, 301, 311, 315, 321, 322, 341, 349, 353, 361, 385, 396 }

# Subset Sum Problem

- Given integers  $\{w_1, \dots, w_n\}$  and an integer  $K$
- Find a subset that is as large as possible that does not exceed  $K$
- Dynamic Programming: Express as an optimization over sub-problems.
- New idea: Represent at a sub problems depending on  $K$  and  $n$ 
  - Two dimensional grid

# Subset Sum Optimization

$\text{Opt}[j, K]$  the largest subset of  $\{w_1, \dots, w_j\}$  that sums to at most  $K$

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + w_j)$$

# Subset Sum Grid

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + w_j)$$

4																	
3	0	2	2	4	4	6	7	7	9								
2	0	2	2	4	4	6	6	6									
1	0	2	2	2	2												
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		1	2	3	4	5	6	7	8	9							

{2, 4, 7, 10}

# Subset Sum Grid

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + w_j)$$

4	0	2	2	4	4	6	7	7	9	10	11	12	13	14	14	16	17
3	0	2	2	4	4	6	7	7	9	9	11	11	13	13	13	13	13
2	0	2	2	4	4	6	6	6	6	6	6	6	6	6	6	6	6
1	0	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

{2, 4, 7, 10}



# Subset Sum Code

```
for j = 1 to n
  for k = 1 to W
    Opt[j, k] = max(Opt[j-1, k], Opt[j-1, k-wj] + wj)
```

# Knapsack Problem

- Items have weights and values
- The problem is to maximize total value subject to a bound on weight
- Items  $\{I_1, I_2, \dots, I_n\}$ 
  - Weights  $\{w_1, w_2, \dots, w_n\}$
  - Values  $\{v_1, v_2, \dots, v_n\}$
  - Bound  $K$
- Find set  $S$  of indices to:
  - Maximize  $\sum_{i \in S} v_i$  such that  $\sum_{i \in S} w_i \leq K$

# Knapsack Recurrence

Subset Sum Recurrence:

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + w_j)$$

Knapsack Recurrence:

$$\text{Opt}[i, K] = \max(\text{Opt}[i-1, K], \text{Opt}[i-1, K - w_j] + v_j)$$

# Knapsack Grid

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + v_j)$$

4																	
3																	
2	0	3	3	5	5	8	—————										
1	0	3	3	3	3	3	—————										
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		1	2	3	4	5	6										

Weights {2, 4, 7, 10} Values: {3, 5, 9, 16}

# Knapsack Grid

$$\text{Opt}[j, K] = \max(\text{Opt}[j - 1, K], \text{Opt}[j - 1, K - w_j] + v_j)$$

4	0	3	3	5	5	8	9	9	12	16	16	18	18	21	21	24	25
3	0	3	3	5	5	8	9	9	12	12	14	14	17	17	17	17	17
2	0	3	3	5	5	8	8	8	8	8	8	8	8	8	8	8	8
1	0	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Weights {2, 4, 7, 10} Values: {3, 5, 9, 16}

# Alternate approach for Subset Sum

- Alternate formulation of Subset Sum dynamic programming algorithm
  - $\text{Sum}[i, K] = \text{true}$  if there is a subset of  $\{w_1, \dots, w_i\}$  that sums to exactly  $K$ , false otherwise
  - $\text{Sum}[i, K] = \text{Sum}[i-1, K] \text{ OR } \text{Sum}[i-1, K - w_i]$
  - $\text{Sum}[0, 0] = \text{true}$ ;  $\text{Sum}[i, 0] = \text{false}$  for  $i \neq 0$
  
- To allow for negative numbers, we need to fill in the array between  $K_{min}$  and  $K_{max}$

# Run time for Subset Sum

- With  $n$  items and target sum  $K$ , the run time is  $O(nK)$
- If  $K$  is 1,000,000,000,000,000,000,000,000,000 this is very slow
- Alternate brute force algorithm: examine all subsets:  $O(2^n)$
- Point of confusion: Subset sum is NP Complete

# Two dimensional dynamic programming

Subset sum and knapsack

$O(Kn)$  time

$O(K)$  space

$$\text{Opt}[j, K] = \max(\text{Opt}[j-1, K], \text{Opt}[j-1, K - w_j] + w_j)$$

$$\text{Opt}[j, K] = \max(\text{Opt}[j-1, K], \text{Opt}[j-1, K - w_j] + v_j)$$

4	0																
3	0																
2	0																
1	0																
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

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# Reducing dimensions

- Computing values in the array only requires the previous row
  - Easy to reduce this to just tracking two rows
  - And sometimes can be implemented in a single row
- Space savings is significant in practice
- Reconstructing values is harder

# Longest Common Subsequence

- $C=c_1\dots c_g$  is a subsequence of  $A=a_1\dots a_m$  if  $C$  can be obtained by removing elements from  $A$  (but retaining order)
- $LCS(A, B)$ : A maximum length sequence that is a subsequence of both  $A$  and  $B$

ocurraneec

attacggct

occurrence

tacgacca

ocurranc

Determine the LCS of the following strings

BARTHOLEMEWSIMPSON

KRUSTYTHECLOWN



# LCS Optimization

- $A = a_1 a_2 \dots a_m$
- $B = b_1 b_2 \dots b_n$
- $\text{Opt}[j, k]$  is the length of  $\text{LCS}(a_1 a_2 \dots a_j, b_1 b_2 \dots b_k)$

$$\text{Opt}[j, k] = \begin{cases} \text{Opt}[j-1, k] & \text{if } (a_j \neq b_k) \\ \text{Opt}[j-1, k-1] + 1 & \text{if } (a_j = b_k) \end{cases}$$

Handwritten notes in red ink:

- $\text{Opt}[j-1, k]$
- $\text{Opt}[j, k-1]$
- $\text{Opt}[j-1, k-1]$
- $\text{Opt}[j, k]$
- $\text{Opt}[j-1, k]$
- $\text{Opt}[j, k-1]$
- $\text{Opt}[j-1, k-1]$
- $\text{Opt}[k-1, j-1] + 1$
- $\max \left\{ \begin{array}{l} \text{Opt}[j-1, k] \\ \text{Opt}[j, k-1] \end{array} \right.$

# Optimization recurrence

$$\text{If } a_j = b_k, \text{ Opt}[j,k] = 1 + \text{Opt}[j-1, k-1]$$

$$\text{If } a_j \neq b_k, \text{ Opt}[j,k] = \max(\text{Opt}[j-1,k], \text{Opt}[j,k-1])$$

# Code to compute Opt[ n, m]

```
for (int i = 0; i < n; i++)
  for (int j = 0; j < m; j++)
    if (A[ i ] == B[ j ] )
      Opt[ i, j ] = Opt[ i-1, j-1 ] + 1;
    else if (Opt[ i-1, j ] >= Opt[ i, j-1 ] )
      Opt[ i, j ] := Opt[ i-1, j ];
    else
      Opt[ i, j ] := Opt[ i, j-1];
```

# Storing the path information

$A[1..m]$ ,  $B[1..n]$

for  $i := 1$  to  $m$      $Opt[i, 0] := 0$ ;

for  $j := 1$  to  $n$      $Opt[0, j] := 0$ ;

$Opt[0, 0] := 0$ ;

for  $i := 1$  to  $m$

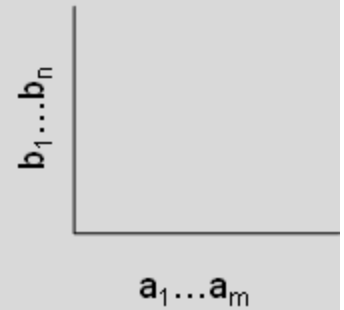
    for  $j := 1$  to  $n$

        if  $A[i] = B[j]$  {  $Opt[i, j] := 1 + Opt[i-1, j-1]$ ;  $Best[i, j] := \text{Diag}$ ; }

        else if  $Opt[i-1, j] \geq Opt[i, j-1]$

            {  $Opt[i, j] := Opt[i-1, j]$ ,  $Best[i, j] := \text{Left}$ ; }

        else     {  $Opt[i, j] := Opt[i, j-1]$ ,  $Best[i, j] := \text{Down}$ ; }







# How good is this algorithm?

- Is it feasible to compute the LCS of two strings of length 300,000 on a standard desktop PC? Why or why not.

# Implementation 1

```
public int ComputeLCS() {
    int n = str1.Length;
    int m = str2.Length;

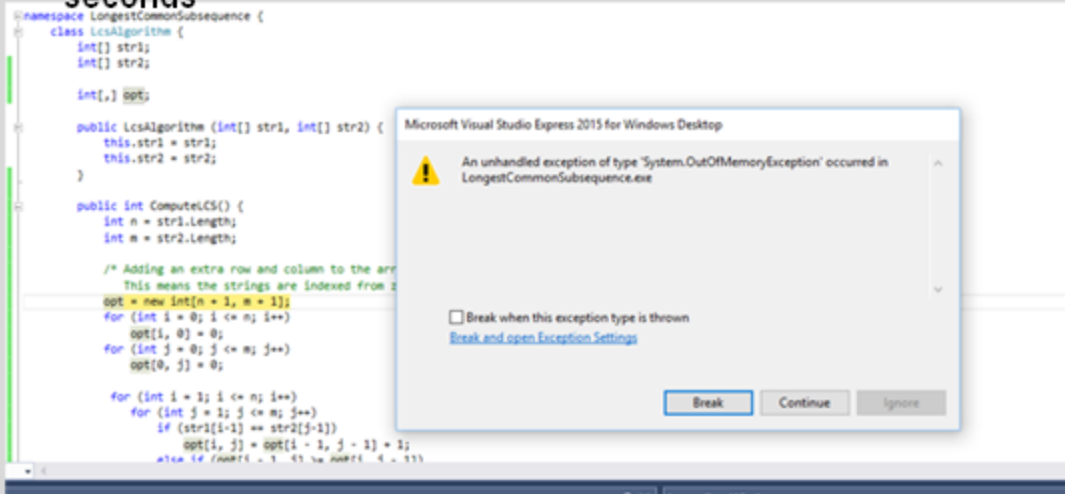
    int[,] opt = new int[n + 1, m + 1];
    for (int i = 0; i <= n; i++)
        opt[i, 0] = 0;
    for (int j = 0; j <= m; j++)
        opt[0, j] = 0;

    for (int i = 1; i <= n; i++)
        for (int j = 1; j <= m; j++)
            if (str1[i-1] == str2[j-1])
                opt[i, j] = opt[i - 1, j - 1] + 1;
            else if (opt[i - 1, j] >= opt[i, j - 1])
                opt[i, j] = opt[i - 1, j];
            else
                opt[i, j] = opt[i, j - 1];

    return opt[n,m];
}
```

# N = 17000

Runtime should be about 5 seconds\*



\* Personal PC, 10 years old

Manufacturer:	Dell
Model:	Optiplex 990
Processor:	Intel(R) Core(TM) i5-2400 CPU @ 3.10GHz 3.10 GHz
Installed memory (RAM):	8.00 GB (7.88 GB usable)
System type:	64-bit Operating System, x64-based processor

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# Implementation 2

```
public int SpaceEfficientLCS() {
    int n = str1.Length;
    int m = str2.Length;
    int[] prevRow = new int[m + 1];
    int[] currRow = new int[m + 1];

    for (int j = 0; j <= m; j++)
        prevRow[j] = 0;

    for (int i = 1; i <= n; i++) {
        currRow[0] = 0;
        for (int j = 1; j <= m; j++) {
            if (str1[i - 1] == str2[j - 1])
                currRow[j] = prevRow[j - 1] + 1;
            else if (prevRow[j] >= currRow[j - 1])
                currRow[j] = prevRow[j];
            else
                currRow[j] = currRow[j - 1];
        }
        for (int j = 1; j <= m; j++)
            prevRow[j] = currRow[j];
    }

    return currRow[m];
}
```

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$$N = 300000$$

N: 10000	Base 2 Length: 8096	Gamma: 0.8096	Runtime:00:00:01.86
N: 20000	Base 2 Length: 16231	Gamma: 0.81155	Runtime:00:00:07.45
N: 30000	Base 2 Length: 24317	Gamma: 0.8105667	Runtime:00:00:16.82
N: 40000	Base 2 Length: 32510	Gamma: 0.81275	Runtime:00:00:29.84
N: 50000	Base 2 Length: 40563	Gamma: 0.81126	Runtime:00:00:46.78
N: 60000	Base 2 Length: 48700	Gamma: 0.8116667	Runtime:00:01:08.06
N: 70000	Base 2 Length: 56824	Gamma: 0.8117715	Runtime:00:01:33.36

N: 300000 Base 2 Length: 243605 Gamma: 0.8120167  
Runtime:00:28:07.32

# Observations about the Algorithm

- The computation can be done in  $O(m+n)$  space if we only need one column of the Opt values or Best Values
- The computation requires  $O(nm)$  space if we store all of the string information