

## CSE 421 Algorithms

Autumn 2019, Lecture 20  
Memory Efficient Dynamic Programming

### Longest Common Subsequence

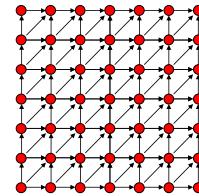
- $C=c_1\dots c_g$  is a subsequence of  $A=a_1\dots a_m$  if  $C$  can be obtained by removing elements from  $A$  (but retaining order)
- $\text{LCS}(A, B)$ : A maximum length sequence that is a subsequence of both  $A$  and  $B$

$\text{LCS}(\text{BARTHOLEMEWSIMPSON}, \text{KRUSTYTHECLOWN})$   
= RTHOWN

### LCS Optimization

- $A = a_1a_2\dots a_m$ ,  $B = b_1b_2\dots b_n$
- $\text{Opt}[j, k]$  is the length of  $\text{LCS}(a_1a_2\dots a_j, b_1b_2\dots b_k)$
- Optimization Recurrence:
  - If  $a_j = b_k$ ,  $\text{Opt}[j, k] = 1 + \text{Opt}[j-1, k-1]$
  - If  $a_j \neq b_k$ ,  $\text{Opt}[j, k] = \max(\text{Opt}[j-1, k], \text{Opt}[j, k-1])$

### Dynamic Programming Computation

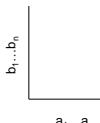


### Code to compute $\text{Opt}[n, m]$

```
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] := Diag; }
        else if Opt[i-1, j] >= Opt[i, j-1]
            { Opt[i, j] := Opt[i-1, j], Best[i,j] := Left; }
            else { Opt[i, j] := Opt[i, j-1], Best[i,j] := Down; }
```



## Reconstructing Path from Distances

## Implementation 1

```

public int ComputeLCSubstr() {
    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];
}

```

## Implementation 2

```

public int SpaceEfficientCS() {
    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];
}

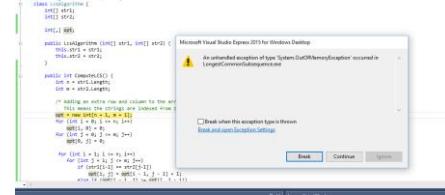
```

## 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.

$$N = 17000$$

Runtime should be about 5 seconds\*



\* Personal PC, 6 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

$$N = 300000$$

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

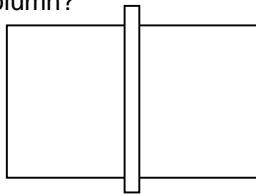
- The computation can be done in  $O(m+n)$  space if we only need one column of the Opt values or Best Values
- The algorithm can be run from either end of the strings

Computing LCS in  $O(nm)$  time and  $O(n+m)$  space

- Divide and conquer algorithm
- Recomputing values used to save space

## Divide and Conquer Algorithm

- Where does the best path cross the middle column?



- For a fixed  $i$ , and for each  $j$ , compute the LCS that has  $a_i$  matched with  $b_j$

## Constrained LCS

- $LCS_{i,j}(A,B)$ : The LCS such that
  - $a_1, \dots, a_i$  paired with elements of  $b_1, \dots, b_j$
  - $a_{i+1}, \dots, a_m$  paired with elements of  $b_{j+1}, \dots, b_n$
- $LCS_{4,3}(abbacbb, cbbaa)$

A = RRSSRTTRTS  
B=RTSRRSTST

Compute  $LCS_{5,0}(A,B), LCS_{5,1}(A,B), \dots, LCS_{5,9}(A,B)$

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B=RTSRRSTST

Compute  $LCS_{5,0}(A,B), LCS_{5,1}(A,B), \dots, LCS_{5,9}(A,B)$

j	left	right
0	0	4
1	1	4
2	1	3
3	2	3
4	3	3
5	3	2
6	3	2
7	3	1
8	4	1
9	4	0

## Computing the middle column

- From the left, compute  $\text{LCS}(a_1 \dots a_{m/2}, b_1 \dots b_j)$
- From the right, compute  $\text{LCS}(a_{m/2+1} \dots a_m, b_{j+1} \dots b_n)$
- Add values for corresponding  $j$ 's



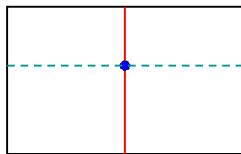
- Note – this is space efficient

## Divide and Conquer

- $A = a_1, \dots, a_m$        $B = b_1, \dots, b_n$
- Find  $j$  such that
  - $\text{LCS}(a_1 \dots a_{m/2}, b_1 \dots b_j)$  and
  - $\text{LCS}(a_{m/2+1} \dots a_m, b_{j+1} \dots b_n)$  yield optimal solution
- Recurse

## Algorithm Analysis

- $T(m,n) = T(m/2, j) + T(m/2, n-j) + cnm$



Prove by induction that  
 $T(m,n) \leq 2cmn$

$$T(m,n) = T(m/2, j) + T(m/2, n-j) + cnm$$

## Memory Efficient LCS Summary

- We can afford  $O(nm)$  time, but we can't afford  $O(nm)$  space
- If we only want to compute the length of the LCS, we can easily reduce space to  $O(n+m)$
- Avoid storing the value by recomputing values
  - Divide and conquer used to reduce problem sizes