

CSE 421

Algorithms

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Lecture 18

Dynamic Programming

Announcements

- Nov 11, No class (holiday)
- HW 7 is available
 - 5 dynamic programming problems

One dimensional dynamic programming: Interval scheduling

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



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 $\{w_1, \dots, w_k\}$ that sums to exactly K , false otherwise
- $\text{Sum}[i, K] = \text{Sum}[i - 1, K] \text{ OR } \text{Sum}[i - 1, K - w_i]$
- $\text{Sum}[0, 0] = 1$; $\text{Sum}[i, 0] = 0$ for $i \neq 0$

- To count the number of solutions
 - $\text{Count}[i, K] = \text{Count}[i - 1, K] + \text{Count}[i - 1, K - w_i]$
 - $\text{Count}[0, 0] = 1$; $\text{Count}[i, 0] = 0$ for $i \neq 0$

- To allow for negative numbers, we need to fill in the array between K_{min} and K_{max}

Dynamic Programming Examples

- Examples
 - Optimal Billboard Placement
 - Text, Solved Exercise, Pg 307
 - Linebreaking with hyphenation
 - Compare with HW problem 6, Pg 317
 - String approximation
 - Text, Solved Exercise, Page 309

Billboard Placement

- Maximize income in placing billboards
 - $b_i = (p_i, v_i)$, v_i : value of placing billboard at position p_i
- Constraint:
 - At most one billboard every five miles
- Example
 - $\{(6,5), (8,6), (12, 5), (14, 1)\}$

Design a Dynamic Programming Algorithm for Billboard Placement

- Compute $\text{Opt}[1], \text{Opt}[2], \dots, \text{Opt}[n]$
- What is $\text{Opt}[k]$?

Input b_1, \dots, b_n , where $b_i = (p_i, v_i)$, position and value of billboard i

Solution

```
j = 0;           // j is five miles behind the current position
                 // the last valid location for a billboard, if one placed at P[k]
for k := 1 to n
    while (P[ j ] < P[ k ] - 5)
        j := j + 1;
    j := j - 1;
    Opt[ k ] = Max(Opt[ k-1 ] , V[ k ] + Opt[ j ]);
```

String approximation

- Given a string S , and a library of strings $B = \{b_1, \dots, b_m\}$, construct an approximation of the string S by using copies of strings in B .

$B = \{abab, bbbaaa, ccbb, ccaacc\}$

$S = abaccbbbaabbccbbccaabab$

Formal Model

- Strings from B assigned to non-overlapping positions of S
- Strings from B may be used multiple times
- Cost of δ for unmatched character in S
- Cost of γ for mismatched character in S
 - $\text{MisMatch}(i, j)$ – number of mismatched characters of b_j , when aligned starting with position i in s .

Design a Dynamic Programming Algorithm for String Approximation

- Compute $\text{Opt}[1], \text{Opt}[2], \dots, \text{Opt}[n]$
- What is $\text{Opt}[k]$?

Target string $S = s_1s_2\dots s_n$

Library of strings $B = \{b_1, \dots, b_m\}$

$\text{MisMatch}(i,j)$ = number of mismatched characters with b_j when aligned starting at position i of S .

$$\text{Opt}[k] = \text{fun}(\text{Opt}[0], \dots, \text{Opt}[k-1])$$

- How is the solution determined from sub problems?

Target string $S = s_1s_2\dots s_n$

Library of strings $B = \{b_1, \dots, b_m\}$

$\text{MisMatch}(i,j)$ = number of mismatched characters with b_j when aligned starting at position i of S .

Solution

for $i := 1$ to n

$\text{Opt}[k] = \text{Opt}[k-1] + \delta;$

 for $j := 1$ to $|B|$

$p = i - \text{len}(b_j);$

$\text{Opt}[k] = \min(\text{Opt}[k], \text{Opt}[p-1] + \gamma \text{Mismatch}(p, j));$

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

ocurranec

attacggct

occurrence

tacgacca

Determine the LCS of the following strings

BARTHOLEMEWSIMPSON

KRUSTYTHECLOWN

String Alignment Problem

- Align sequences with gaps

CAT TGA AT

CAGAT AGGA

- Charge δ_x if character x is unmatched
- Charge γ_{xy} if character x is matched to character y

Note: the problem is often expressed as a minimization problem,
with $\gamma_{xx} = 0$ and $\delta_x > 0$

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

Give the Optimization Recurrence for the String Alignment Problem

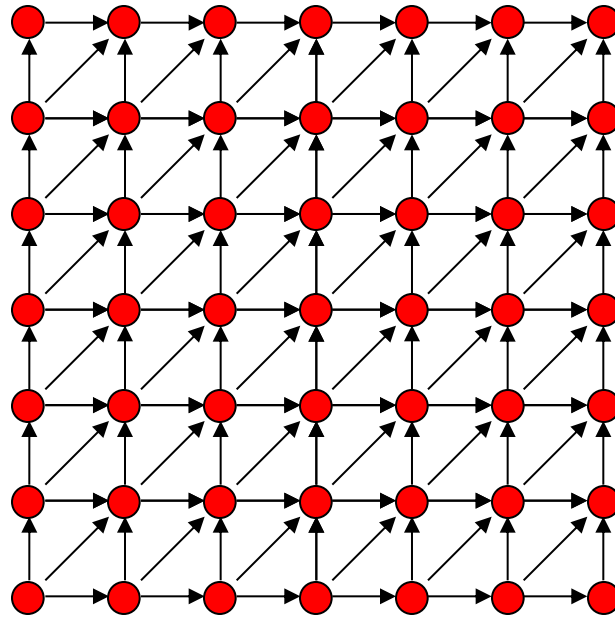
- Charge δ_x if character x is unmatched
- Charge γ_{xy} if character x is matched to character y

Opt[j , k] =

Let $a_j = x$ and $b_k = y$

Express as minimization

Dynamic Programming Computation



Code to compute $\text{Opt}[j,k]$

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

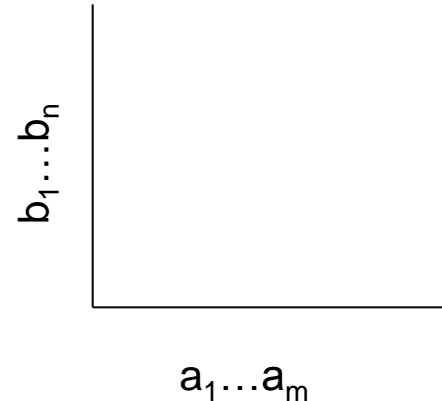
 for $j := 1$ to n

 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.