

More dynamic programming

Tuesday, October 13, 2020 1:31 PM

longest increasing subsequence

e.g. $A = [6, 3, 5, 2, 7, 8, 1]$

subsequence: e.g. $6, 5, 2$

increasing subsequence... e.g. $[6, 7, 8]$

names of subproblems.

$[3, 5, 7, 8]$? \checkmark
 $[2, 7, 8]$?

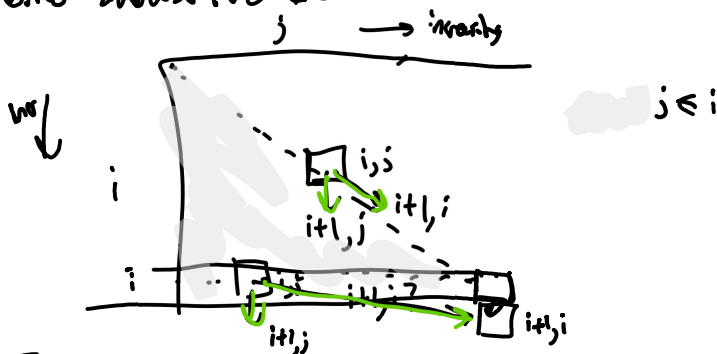
$LIS(i, seq\ so\ far) =$ longest subsequence of $A[i \dots n]$ that is increasing if appended to $seq\ so\ far$.
 0 if $i = n+1$

$\max \left\{ \begin{array}{l} 1 + LIS(i+1, seq\ so\ far + A[i]) // \text{include } A[i] \\ LIS(i+1, seq\ so\ far) // \text{exclude } A[i] \end{array} \right.$
if $A[i] >$ last element of $seq\ so\ far$.

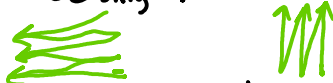
$LIS(i, j) =$ length of longest subsequence of $A[i, \dots, n]$ containing elements larger than $A[j]$
 0 if $i = n+1$

$\max \left\{ \begin{array}{l} 1 + LIS(i+1, j) // \text{include } A[i] \\ LIS(i+1, i) // \text{exclude } A[i] \end{array} \right.$

Memo data structure



Evaluation order: each subproblem i, j depends on $i+1, j$ and $i+1, i$.
 so evaluate in order of: decreasing i , or decreasing j .



row major decreasing in i, increasing in j

col major increasing in i, decreasing in j

Estimate running time: # subproblems: $O(n^2)$

Time for each subproblem: $O(1)$

Running time: $O(n^2)$

~ ~ ~

Edit Distance.

- Distance between strings
- min # of edit operations needed to turn string X into string Y
- edit operations:
 - insertion of 1 character
 - delete 1 character
 - substitute 1 character

ex of operations

sub CAT \rightarrow CAB (subst T for B)

ins CAT \rightarrow CABT

del CAT \rightarrow CA

ex

ALGORITHM
 ALTRUISTIC

.. 1 2 3 4 5 6 7

Subs
Ins?

ALGORITHM
 ALGRITHM - on the top means insert

X = ALGOR-I-THM

AL-T-RUISTIC

1 2 3 4 5 6

- on the bottom means: delete

Go!

See your notes for a proof of the fibbing.

First recurrence:

Dist(X, Y) edit distance X to Y.

Base - Dist(ϵ , Y) = |Y| // |Y| insertions

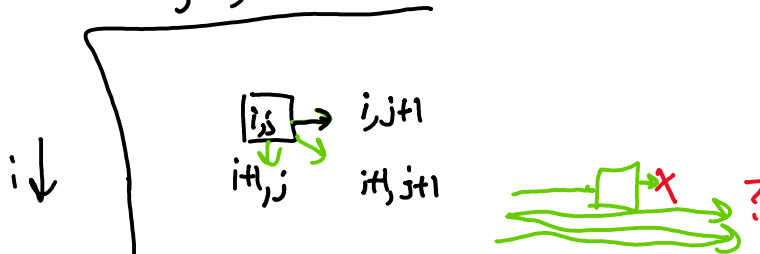
cases — $\text{Dist}(X, \epsilon) = |X|$ // $|X|$ deletions.

Inductive case — $\text{Dist}(ax', by') =$
 $\min \begin{cases} 1 + \text{Dist}(ax', y') // \text{inserting } \frac{b|ax'}{b|y'} \text{ subproblem start goal} \\ 1 + \text{Dist}(x', by') // \text{deleting } \frac{a|x'}{b|y'} \\ (a+b) + \text{Dist}(x', y') // \text{substituting } \frac{a|x'}{b|y'} \end{cases}$
 (if $a \neq b$, otherwise)

$\text{Dist}(i, j)$ (i indexed s.t. def) edit distance from $X[i \dots m]$ to $Y[j \dots n]$ $|X|=m$ $|Y|=n$ $|Y[j \dots n]| = n - j + 1$

$\text{Dist}(m+1, j) = n - j + 1$ // $n - j + 1$ insertions
 $\text{Dist}(i, n+1) = m - i + 1$ // $m - i + 1$ deletions.

$\text{Dist}(i, j) = \min \begin{cases} 1 + \text{Dist}(i, j+1) // \text{insertion} \\ 1 + \text{Dist}(i+1, j) // \text{deletion} \\ (X[i] \neq Y[j]) + \text{Dist}(i+1, j+1) // \text{subst} \end{cases}$
 $j \rightarrow$



Evaluation Order: decreasing i and decreasing j .

e.g. row major w/ decreasing i and decreasing j .

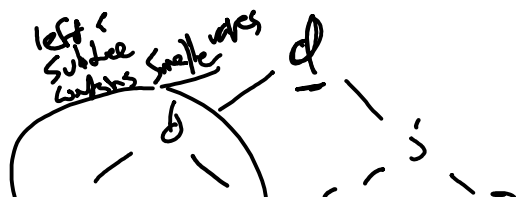
of subproblems: $O(nm)$

Cost of each subproblem: $O(1)$

Overall: $O(nm)$ running time.

Pseudocode: ... leave to later if time.

Optimal Binary Search Tree





In balanced tree: depth $O(\log_2 n)$ worst case.

Search(X, T):

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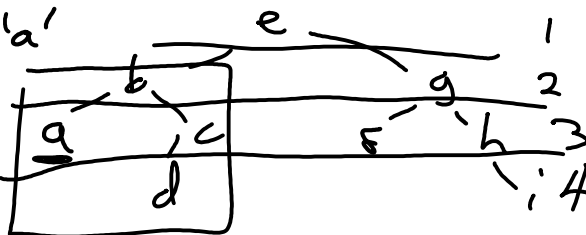
if J = e then False
if T.val = X then True
if X < T.val then Search(X, T.left)
else Search(X, T.right)
    
```

What about weighted average case search time?

Input: $A = [a, b, c, d, e, f, g, h, i]$ (sorted)

Frequency = $[1, 5, 1, 1, 1, 1, 2, 1, 1, 1]$

Avg Cost:
 half of all queries are for 'a'
 $1 \cdot (1 \cdot 0.1)$
 $+ 2 \cdot (1 \cdot 0.1)$
 $+ 3 \cdot (1 \cdot 0.1)$
 $+ \dots$



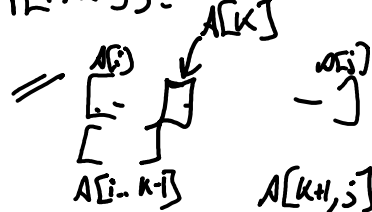
What's the tree that gives best avg cost?

optBST(i, j)

if $i=j$: $1 \cdot F[i]$

else:
 $\min_{i \leq k \leq j} \left\{ \begin{aligned} &\sum F[i \dots j] \\ &+ \text{optBST}(i, k-1) \\ &+ \text{optBST}(k+1, j) \end{aligned} \right.$

Cost of best tree for interval $A[i \dots j]$.



optBST(A, F):

Let Opt be a $(n+1) \times (n+1)$ array.

Let Root be a $(n+1) \times (n+1)$ array. // base cases.

For $i = n$ down to 1:

For $j = i$ up to n

Best := 0
 Bestk := 0

What value of k give me the best score?

for $k = i$ upto j :
 $x = \sum \text{freq}[i \dots j] + \text{opt}[i, k-1] + \text{opt}[k+1, j]$
 if $x < \text{Best}$:
 $\text{BestK} := k$
 $\text{Best} := x$
 $\text{Opt}[i, j] := \text{Best}$
 $\text{Root}[i, j] := \text{BestK} \leftarrow$
 return $\text{opt}[1, n]$

To read back the structure:

1. - Store the value of k that gave the best
2. - Recursive function to read it back:

$\text{BST}(i, j)$: // note use of opt_{BST} & Root
 if $i = j$ then return $T(\text{arr}[i], \epsilon, \epsilon)$ \wedge
 else return $T(\text{arr}[\text{Root}[i, j]], \text{BST}(i, \text{Root}[i, j]-1), \text{BST}(\text{Root}[i, j]+1, j))$

1/2 A BST is defined inductively as:

- ϵ is the empty tree
- $T(\text{val}, t_L, t_R)$ is a node with value val , left subtree t_L , right subtree t_R