

15-750

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Dynamic Programming An Alg Design Technique

Thousands of Applications

Kai-Fu Lee's Speech Alg

Genome Seg

Clark-Bryant Hardware checking

We will do

1) Classic Ex

2) Not so classic one.

Your solutions should have a special form.

4-Steps

Optimal Binary Search Trees

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Input:

Keys	K_1	K_2	...	K_n
Freq	P_1	P_2		
Prob				P_n

Goal: BST s.t.

Cost Search (K_i) = Depth(K_i) eg Depth of root = 1

$$\text{Expected Cost} = \sum_{i=1}^n P_i \cdot \text{Depth}(K_i)$$

BST with min expected cost.

Naive Alg: Try all possible trees

Side Question: How many trees (Binary) BT.

We will use Dyn Prog to compute # tree

Counting #BT

Step 1 Give a definition of subproblems being solved

eg $T(n) = \# \text{BST with } n \text{ nodes}$

Step 2 Give a recurrence over subproblems
Include base cases!

eg $T(1) = 1$ $T(2) = 2$ or $T(0) = 1$

$$T(n+1) = \sum_{i=0}^n T(i) \cdot T(n-i)$$

Step 3 Prove recurrence correct by induction.

eg Let $S_n = \text{set of all } n \text{ node BTs.}$

Partition S_n by # nodes in left subtree.

$$S_n^0, S_n^1, \dots, S_n^{n-1}$$

Claim $|S_n^i| = T(i) \cdot T(n-i-1)$ by induction

Step 4 Determine run time.

eg Subprobs $T(1), \dots, T(n)$

$C_i \equiv$ cost to compute $T(i)$ from $T(1), \dots, T(i-1)$

$$C_i = O(i)$$

$$\therefore \text{Total Cost} = \sum_{i=1}^n C_i = O(n^2)$$

Finding Opt-BST

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Input: p_1, \dots, p_n

Trick 1 Compute expect-cost (find tree later)

Trick 2 Solve for more than asked.

eg Solve $C_{ij} \equiv$ expect cost for opt for $p_i \dots p_j$ $i \leq j$

(Step 1) Note: $\sum_{i=1}^n p_i < 1$

Step 2 Recurrence relation

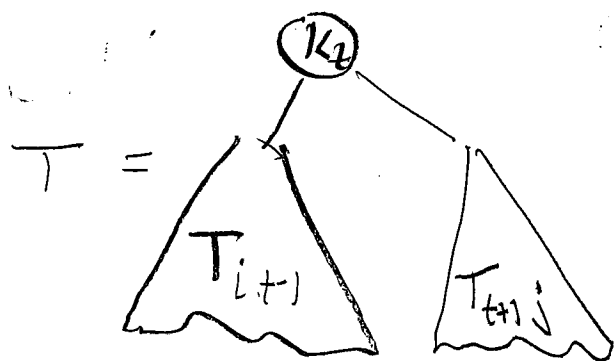
Base Case $C_{ij} = p_i$ if $i=j$ ($C_{ij} = 0$ if $j < i$)

Def $W_{ij} = p_i + \dots + p_j$

$$C_{ij} = \min_{i \leq t \leq j} \{ C_{i,t-1} + C_{t+1,j} \} + W_{ij} \quad (*)$$

Step 3 Correctness

Let $T \equiv$ opt tree subject to condition root k_t .



$$T_L = T_{i+1}$$

$$T_R = T_{t+1:j}$$

$$\text{Cost}(T) = P_t + \sum_{l=i}^{t-1} P_l (\text{Depth}_T(k_l)) + \sum_{l=t+1}^j P_l (\text{Depth}_T(k_l))$$

$$= P_t + \sum_{l \neq t} P_l + \sum_{l \neq t} P_l \text{Depth}_{T_L}(k_l) + \sum_{l=t+1}^j P_l \text{Depth}_{T_R}(k_l)$$

$$= W_{i,j} + \text{Cost}(T_L) + \text{Cost}(T_R)$$

Claim $Opt_{ij} \leq C_{ij}$

We could return a tree with cost = C_{ij}
the root will be k_t for $t \min$ (*)

Claim $C_{ij} \leq Opt_{ij}$

pf induction of $j-i$

Let T_{ij} be opt tree

Suppose root is k_t .

$$\begin{aligned}
C_{ij} &\leq C_{i|t-1} + C_{t+1j} + W_{ij} \\
&\leq Opt_{i|t-1} + Opt_{t+1j} + W_{ij} \\
&= Cost(T_{ij}) = Opt_{ij}
\end{aligned}$$

Recurrence as code (memoization)

Procedure $C(i, j)$

if $j < i$ return 0

elseif $i = j$ return P_j

elseif $\text{hash}(i, j) \neq 0$ return $\text{hash}(i, j)$

else

$$\text{hash}(i, j) = \min_{i \leq t \leq j} \{ C(i, t-1) + C(t, j) \} + W_{ij}$$

return $\text{hash}(i, j)$

Timing Table size = $O(n^2)$

Cost per entry = $O(n)$

Total $O(n^3)$

Table Method

$P = .2, .1, .7$

	1	2	3	4
3	1.4	.9	.7	0
2	.4	.1	0	0
1	.2	0	0	0
0	0	0	0	0

$$.4 = .3 + \min\{.1, .2\}$$

$$.9 = .8 + \min\{.1, .7\}$$

$$1.4 = 1 + \min\{.9+0, .2+.7, 0+.4\}$$

Can we do better? Yes $O(n^2)$ time

Def $r(i, j) \equiv$ index of root of an opt tree

Claim $r(i, j-1) \leq r(i, j) \leq r(i+1, j)$