

## Chapter 5, Dynamic Programming:

- ▶ Introduction (week-before Wednesday)
- ▶ Principles of DP, including sample problems (week-before Friday)
- ▶ DP algorithms, solutions to sample problems (last week Monday)
- ▶ *Review for Test 2 (last week Wednesday)*
- ▶ *Test 2 (last week Thursday, in lab)*
- ▶ *No class (last week Friday)*
- ▶ Optimal BSTs (**Today**)
- ▶ *Begin Chapter 6, Hashtables (Wednesday)*

## Today:

- ▶ The optimal BST definition
- ▶ The optimal-BST-building problem
- ▶ The dynamic programming solution

## Why this problem?

- ▶ It connects dynamic programming with the quest for a better map.
- ▶ Its hardness is in the right places (building the table—hard; reconstructing solution—trivial).
- ▶ It is a representative of a bigger concept: What if we had more information—how would that change the problem.

## Game plan:

- ▶ Understand the problem itself
- ▶ Understand the recursive characterization
- ▶ Understand the table-building algorithm

## The **optimal binary search tree** problem:

- ▶ Assume we know all the keys  $k_0, k_1, \dots, k_{n-1}$  ahead of time.
- ▶ Assume further that we know the probabilities  $p_0, p_1, \dots, p_{n-1}$  of each key's lookup.
- ▶ Assume even further that we know the "miss probabilities"  $q_0, q_1, \dots, q_n$  where  $q_i$  is the probability that an *extraneous* key falling between  $k_{i-1}$  and  $k_i$  will be looked up.
- ▶ We want to build a tree to minimize the *expected cost* of a look up, which is the *total weighted depth* of the tree:

$$\sum_{i=0}^{n-1} p_i \text{ depth}(k_i) + \sum_{i=0}^n q_i \text{ depth}(m_i)$$

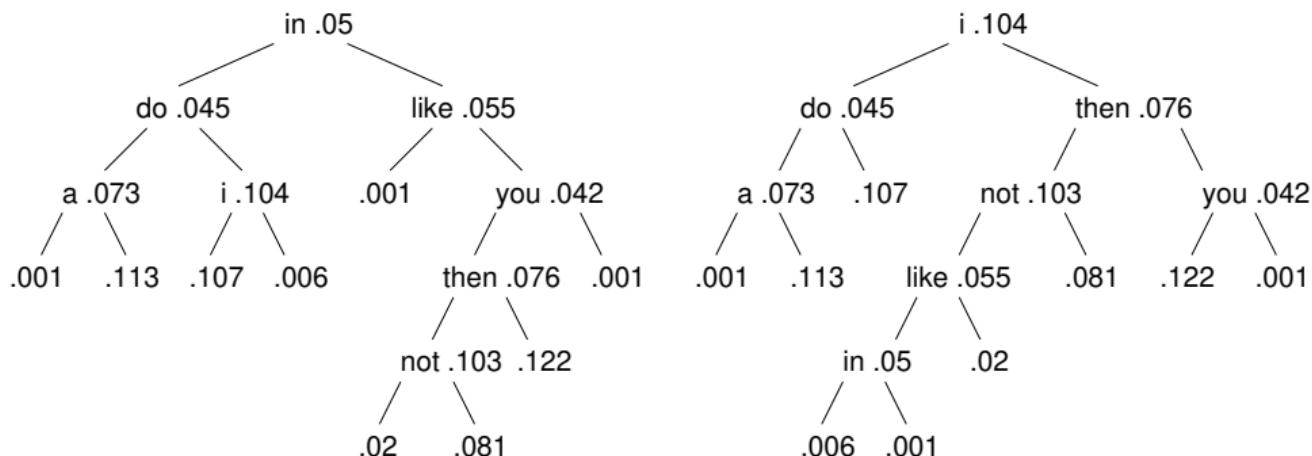
where  $\text{depth}(x)$  is the number of nodes to be inspected on the route from the root to node  $x$ ,  $k_i$  stands for the node containing key  $k_i$  [notational abuse], and  $m_i$  is the dummy node between keys  $k_{i-1}$  and  $k_i$ .

- ▶ Note that the rules of probability require  $\sum_{i=0}^{n-1} p_i + \sum_{i=0}^n q_i = 1$

i	84	eat	24	ham	10	fox	7	rain	4
not	83	will	21	there	9	on	7	see	4
them	61	sam	19	train	9	tree	6	try	4
a	59	with	19	anywhere	8	say	5	boat	3
like	44	am	16	house	8	so	5	that	3
in	40	could	14	mouse	8	be	4	are	2
do	36	here	11	or	8	goat	4	good	2
you	34	the	11	box	7	let	4	thank	2
would	26	eggs	10	car	7	may	4	they	2
and	24	green	10	dark	7	me	4	if	1

Key or miss event	combined frequency
{ }	0
a	59
{ am and anywhere are be boat box car could dark }	92
do	36
{ eat eggs fox goat good green ham here house }	86
i	84
{ if let }	5
in	40
{ }	0
like	44
{ may me mouse }	16
not	83
{ on or rain same say see so thank that the }	65
then	61
{ there they train tree try will with would }	99
you	34
{ }	0

	0	1	2	3	4	5	6	7
$k_i$	a	do	i	in	like	not	then	you
$p_i$	.073	.045	.104	.05	.055	.103	.076	.042
$q_i$	.001	.113	.107	.006	.001	.02	.081	.122



$$\begin{aligned}1 \cdot .02 + 1 \cdot .081 \\= .101\end{aligned}$$

$$\begin{matrix}1 & & 1 \\ .02 & & .081\end{matrix}$$

$$\begin{aligned}2 \cdot .02 + 2 \cdot .081 \\+ 1 \cdot .103 + 1 \cdot .122 \\= .427\end{aligned}$$

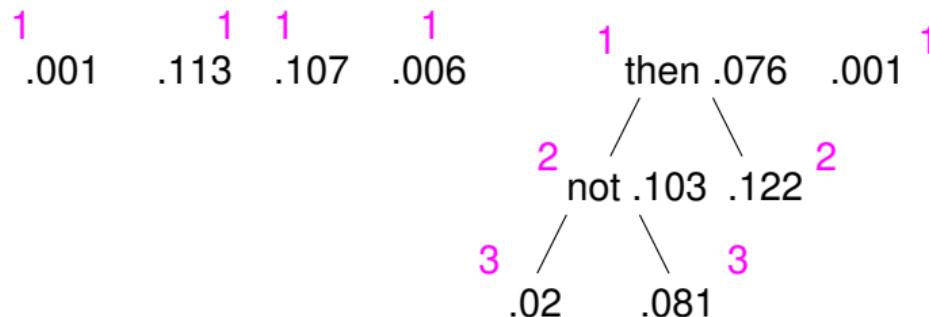
$$\begin{array}{ccccc}1 & & & & 1 \\ & \text{not} & .103 & .122 & \\ 2 & / & & \backslash & 2 \\ .02 & & & .081 & \end{array}$$

$$3 \cdot .02 + 3 \cdot .081$$

$$+ 2 \cdot .103 + 2 \cdot .122$$

$$+ 1 \cdot .001 + 1 \cdot .133 + 1 \cdot .107 + 1 \cdot .006 + 1 \cdot .076 + 1 \cdot .001$$

$$= 1.057$$



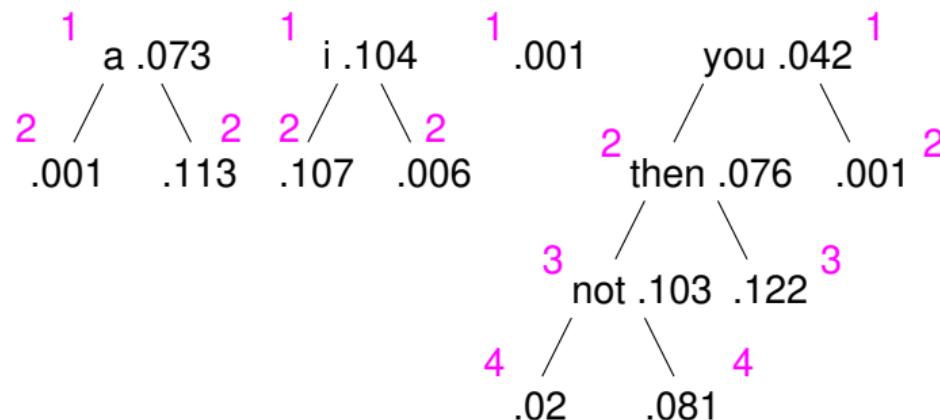
$$4 \cdot .02 + 4 \cdot .081$$

$$+ 3 \cdot .103 + 3 \cdot .122$$

$$+ 2 \cdot .001 + 2 \cdot .133 + 2 \cdot .107 + 2 \cdot .006 + 2 \cdot .076 + 2 \cdot .001$$

$$+ 1 \cdot .073 + 1 \cdot .104 + 1 \cdot .001 + 1 \cdot .042$$

$$= 1.907$$



$$5 \cdot .02 + 5 \cdot .081$$

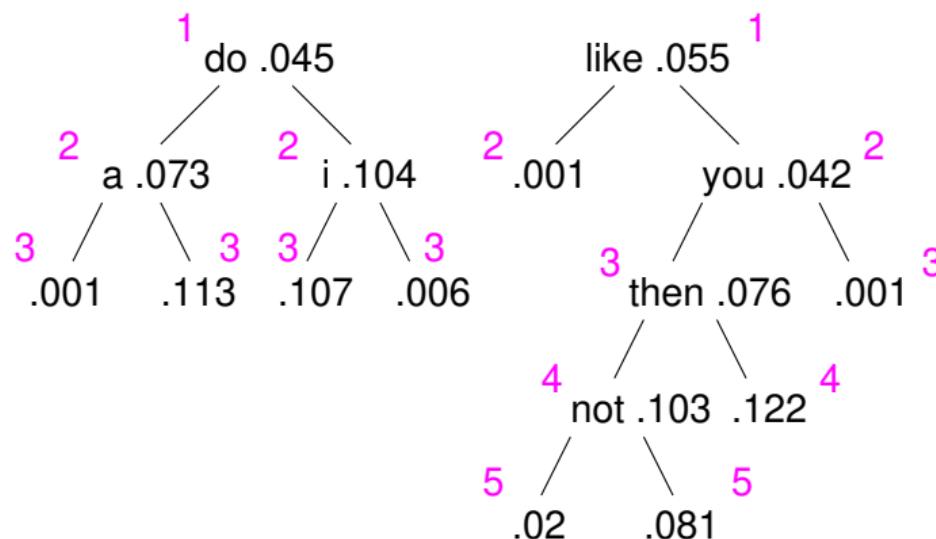
$$+ 4 \cdot .103 + 4 \cdot .122$$

$$+ 3 \cdot .001 + 3 \cdot .133 + 3 \cdot .107 + 3 \cdot .006 + 3 \cdot .076 + 3 \cdot .001$$

$$+ 2 \cdot .073 + 2 \cdot .104 + 2 \cdot .001 + 2 \cdot .042$$

$$+ 1 \cdot .045 + 1 \cdot .055$$

$$= 2.857$$



$$6 \cdot .02 + 6 \cdot .081$$

$$+ 5 \cdot .103 + 5 \cdot .122$$

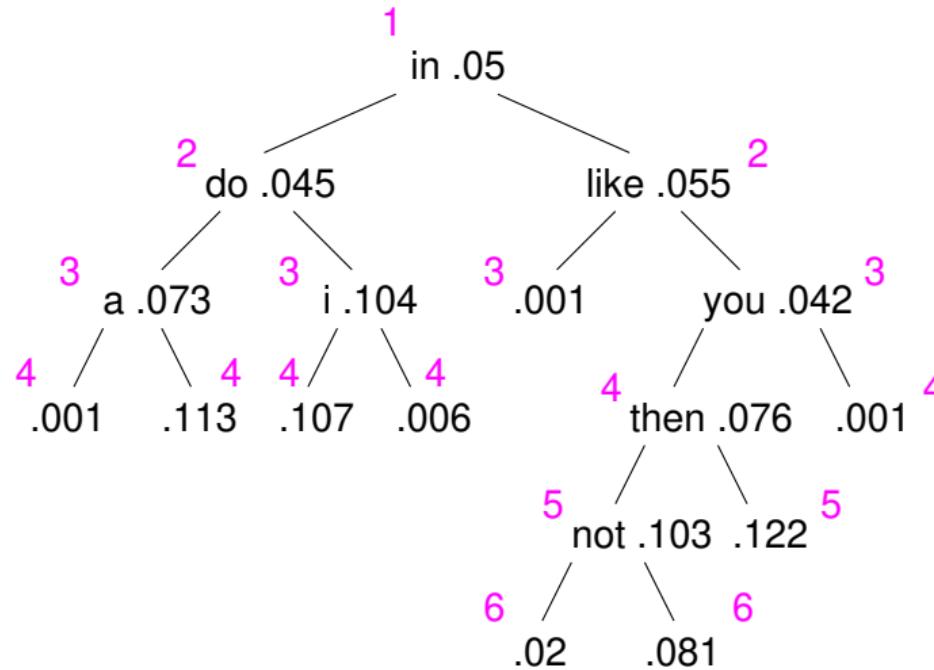
$$+ 4 \cdot .001 + 4 \cdot .133 + 4 \cdot .107 + 4 \cdot .006 + 4 \cdot .076 + 4 \cdot .001$$

$$+ 3 \cdot .073 + 3 \cdot .104 + 3 \cdot .001 + 3 \cdot .042$$

$$+ 2 \cdot .045 + 2 \cdot .055$$

$$+ 1 \cdot .05$$

$$= 3.857$$

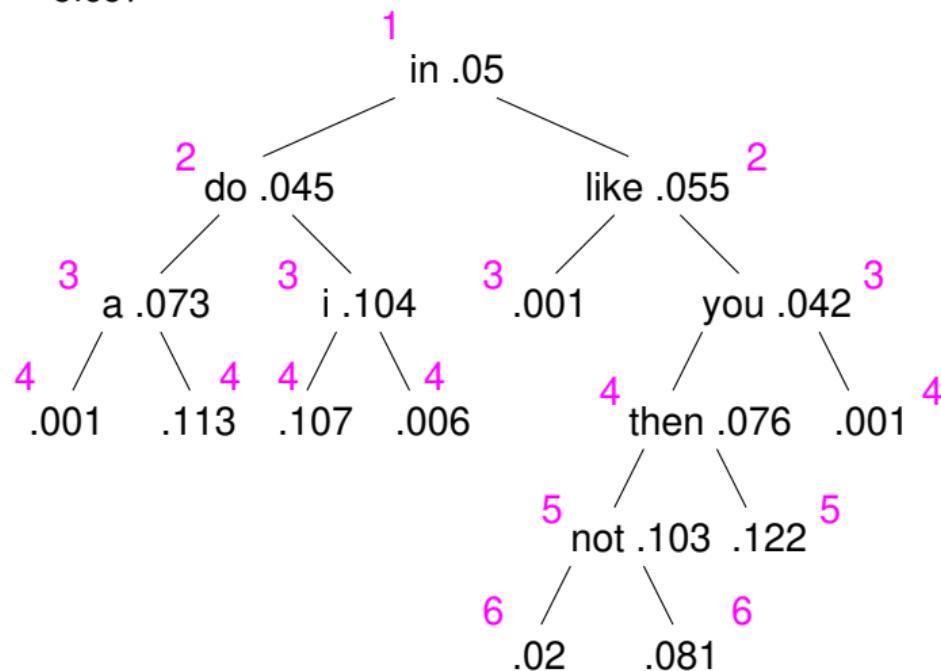


$$4 \cdot .001 + 3 \cdot .073 + 4 \cdot .133 + 2 \cdot .045 + 4 \cdot .107 + 3 \cdot .104 + 4 \cdot .006$$

$$+ 1 \cdot .05$$

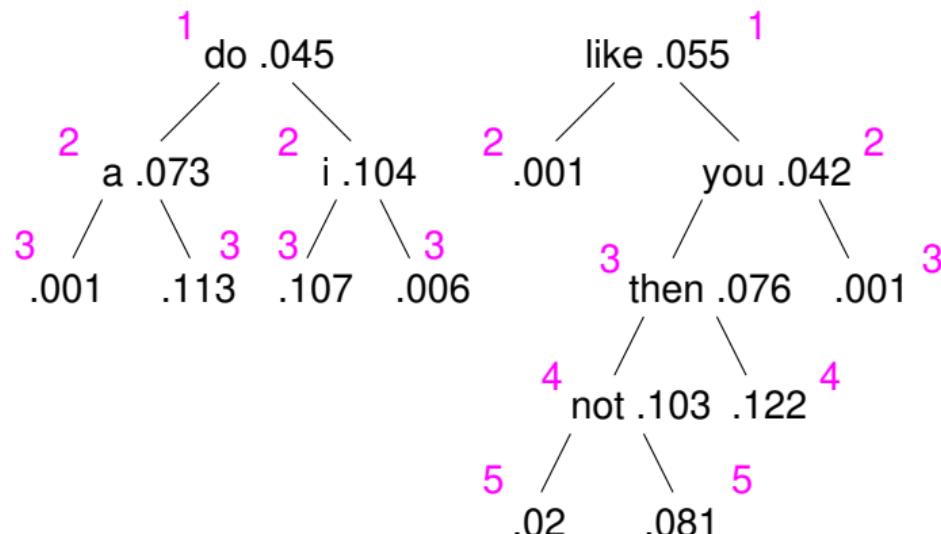
$$+ 3 \cdot .001 + 2 \cdot .055 + 6 \cdot .02 + 6 \cdot .081 + 4 \cdot .076 + 5 \cdot .122 + 3 \cdot .042 + 4 \cdot .001$$

$$= 3.857$$



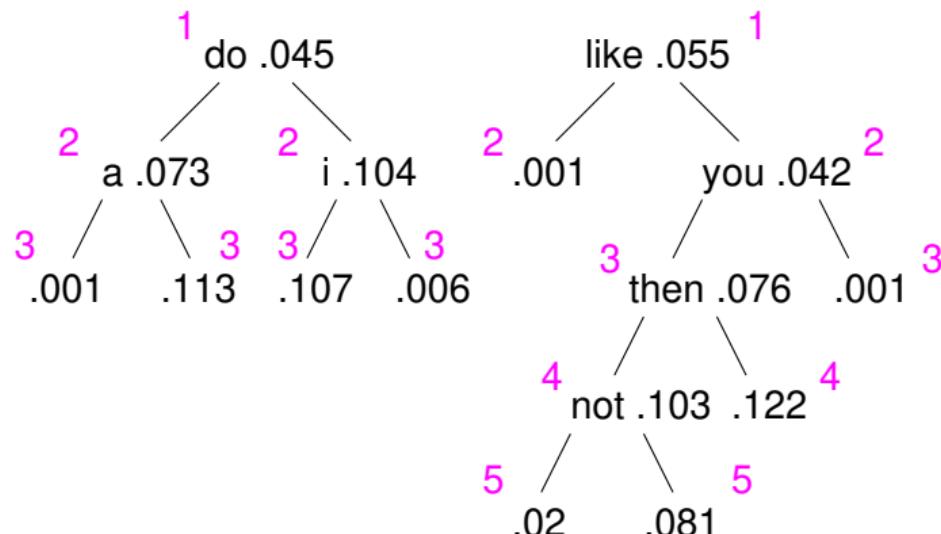
$$\begin{aligned}
 & 3 \cdot .001 + 2 \cdot .073 + 3 \cdot .133 + 1 \cdot .045 + 3 \cdot .107 + 2 \cdot .104 + 3 \cdot .006 \\
 & + .001 + .073 + .133 + .045 + .107 + .104 + .006 \\
 & + .05 \\
 & + 2 \cdot .001 + 1 \cdot .055 + 5 \cdot .02 + 5 \cdot .081 + 3 \cdot .076 + 4 \cdot .122 + 2 \cdot .042 + 3 \cdot .001 \\
 & + .001 + .055 + .02 + .081 + .076 + .122 + .042 + .001 \\
 & = 3.857
 \end{aligned}$$

in .05



$$\begin{aligned}
 & 3 \cdot .001 + 2 \cdot .073 + 3 \cdot .133 + 1 \cdot .045 + 3 \cdot .107 + 2 \cdot .104 + 3 \cdot .006 \\
 & + 2 \cdot .001 + 1 \cdot .055 + 5 \cdot .02 + 5 \cdot .081 + 3 \cdot .076 + 4 \cdot .122 + 2 \cdot .042 + 3 \cdot .001 \\
 & + .001 + .073 + .133 + .045 + .107 + .104 + .006 \\
 & + .05 \\
 & + .001 + .055 + .02 + .081 + .076 + .122 + .042 + .001 \\
 & = 3.857
 \end{aligned}$$

in .05



Total weighted depth for a given tree (expected lookup cost):

$$\underbrace{\sum_{i=0}^{n-1} p_i \text{depth}(k_i)}_{\text{keys}} + \underbrace{\sum_{i=0}^n q_i \text{depth}(m_i)}_{\text{misses}}$$

Let  $\text{depth}_{k_a}(k_i)$  be the depth of the node with  $k_i$  in the subtree rooted at node with  $k_a$ . For example, if  $k_r$  is the root of the entire tree and  $k_a$  is a child of the root, then

$$\text{depth}_{k_r}(k_i) = \text{depth}_{k_a}(k_i) + 1$$

Then we can rewrite the total weighted depth as

$$\underbrace{\sum_{i=0}^{r-1} p_i \text{depth}_{k_r}(k_i) + \sum_{i=0}^r q_i \text{depth}_{k_r}(m_i)}_{\text{left subtree total weighted depth (absolute)}}$$

$$\underbrace{p_r + \sum_{i=r+1}^{n-1} p_i \text{depth}_{k_r}(k_i) + \sum_{i=r+1}^n q_i \text{depth}_{k_r}(m_i)}_{\text{right subtree total weighted depth (absolute)}}$$

Again, let  $k_r$  be the root of the entire tree and  $k_a$  and  $k_b$  be the root's children. Then

$$\underbrace{\sum_{i=0}^{r-1} p_i(depth_{k_a}(k_i) + 1) + \sum_{i=0}^r q_i(depth_{k_a}(m_i) + 1) + p_r}_{\text{left subtree total weighted depth (absolute)}} + \underbrace{p_r + \sum_{i=r+1}^{n-1} p_i(depth_{k_b}(k_i) + 1) + \sum_{i=r+1}^n q_i(depth_{k_r}(m_i) + 1)}_{\text{right subtree total weighted depth (absolute)}}$$

Convert to “relative depth”:

$$\underbrace{\sum_{i=0}^{n-1} p_i + \sum_{i=0}^n q_i}_{\text{total probability}} + \underbrace{\sum_{i=0}^{r-1} p_i \ depth_{k_a}(k_i) + \sum_{i=0}^r q_i \ depth_{k_a}(m_i)}_{\text{left subtree total weighted depth (relative)}} + \underbrace{\sum_{i=r+1}^{n-1} p_i \ depth_{k_b}(k_i) + \sum_{i=r+1}^n q_i \ depth_{k_r}(m_i)}_{\text{right subtree total weighted depth (relative)}}$$

Let  $TWD(k)$  be the total weighted depth of the tree rooted at  $k$  (relative to  $k$ ) and  $TP(k)$  be the total probability of the tree rooted at  $k$ . Then

$$TWD(k_r) = TP(k_r) + TWD(k_a) + TWD(k_b)$$

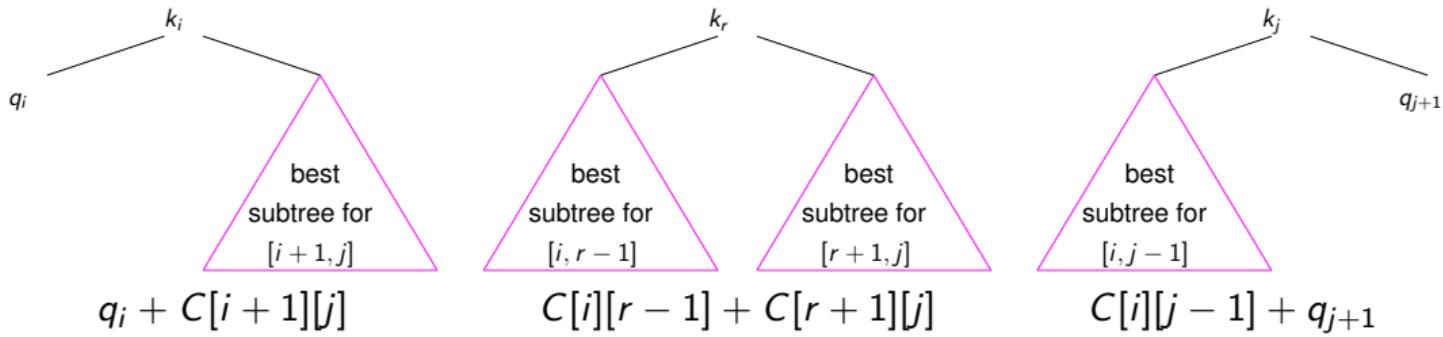
Let  $P[i][j]$  be the total probabilities of the keys and misses in the range  $[i, j]$ :

$$P[i][j] = \sum_{k=i}^j p_k + \sum_{k=i}^{j+1} q_k$$

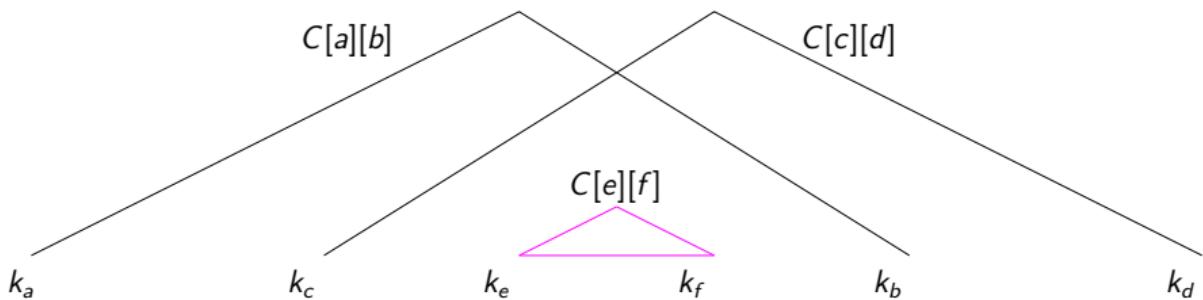
Let  $C[i][j]$  be the least total weighted depth of any BST composed from keys in the range  $[i, j]$ . The recursive characterization is

$$C[i][j] = \begin{cases} 2q_i + p_i + 2q_{i+1} & \text{if } i = j \\ P[i][j] + \min \left\{ \begin{array}{l} q_i + C[i+1][j] \\ C[i][r-1] + C[r+1][j] \text{ for } r \in (i, j) \\ C[i][j-1] + q_{j+1} \end{array} \right\} & \text{if } i < j \end{cases}$$

$$C[i][j] = \begin{cases} 2q_i + p_i + 2q_{i+1} & \text{if } i = j \\ P[i][j] + \min \left\{ \begin{array}{l} q_i + C[i+1][j] \\ C[i][r-1] + C[r+1][j] \text{ for } r \in (i, j) \\ C[i][j-1] + q_{j+1} \end{array} \right\} & \text{if } i < j \end{cases}$$



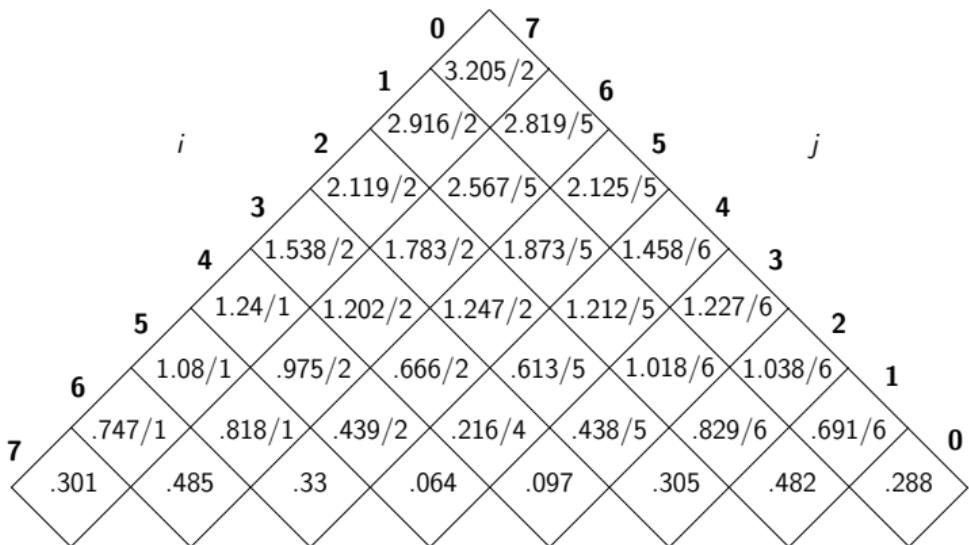
$$C[i][j] = \begin{cases} 2q_i + p_i + 2q_{i+1} & \text{if } i = j \\ P[i][j] + \min \left\{ \begin{array}{l} q_i + C[i+1][j] \\ C[i][r-1] + C[r+1][j] \text{ for } r \in (i, j) \\ C[i][j-1] + q_{j+1} \end{array} \right\} & \text{if } i < j \end{cases}$$



$$C[i][j] = \begin{cases} 2q_i + p_i + 2q_{i+1} & \text{if } i = j \\ P[i][j] + \min \left\{ \begin{array}{l} q_i + C[i+1][j] \\ C[i][r-1] + C[r+1][j] \text{ for } r \in (i, j) \\ C[i][j-1] + q_{j+1} \end{array} \right\} & \text{if } i < j \end{cases}$$

$$P[i][j] = \begin{cases} q_i + p_i + q_{i+1} & \text{if } i = j \\ \left\{ \begin{array}{l} q_i + p_i + P[i+1][j] \\ \text{or } P[i][r-1] + p_r + P[r+1][j] \text{ for } r \in (i, j) \\ \text{or } P[i][j-1] + p_j + q_{j+1} \end{array} \right\} & \text{if } i < j \end{cases}$$

	0	1	2	3	4	5	6	7
$k_i$	a	do	i	in	like	not	then	you
$p_i$	.073	.045	.104	.05	.055	.103	.076	.042
$q_i$	.001	.113	.107	.006	.001	.02	.081	.122



```

# For each candidate root r between i and j exclusive
for r in range(i+1,j):
    # The cost of making key r the root
    current_subtree_cost = (total_weighted_depths[i][r-1] +
                           total_weighted_depths[r+1][j])
    # If its cost is better than best so far, it's the new best so far
    if current_subtree_cost < least_subtree_cost :
        least_subtree_cost = current_subtree_cost
        best_root = r

    # The cost of making key j the root
    current_subtree_cost = total_weighted_depths[i][j-1] + miss_probs[j+1]
    # If its cost is better than best-so-far, it's the new best-so-far
    if current_subtree_cost < least_subtree_cost :
        least_subtree_cost = current_subtree_cost
        best_root = j

    # Record the best option and corresponding cost in the tables
    total_weighted_depths[i][j] = total_probs[i][j] + least_subtree_cost
    decisions[i][j] = best_root

```

From its similarity to the algorithm for optimal matrix multiplication, we recognize the running time for building the tables as  $\Theta(n^3)$ . See Exercise 6.47 for details.

The value  $C[0][n - 1]$  in `total_weighted_depths[0][n-1]` gives us the cost of the best tree for the given keys with their probabilities. As with other dynamic programming problems, a more useful result is the tree itself. Exercise 6.48 asks you to write a function that reconstructs the optimal binary search tree using a populated decision table, but for Project 6.2 we have an alternate strategy. Instead of reconstructing the tree after building the table, we build the actual optimal subtrees along with the table. Instead of a table of decisions as in the algorithm above, we maintain a table such that in position  $(i, j)$  we store the root of the best subtree for keys  $k_i$  through  $k_j$ .

## Coming up:

**Do Optimal BST project (Due Mon, Nov 24)**

**Due Mon, Nov 17**

*Read Section 6.5*

*(No quiz on Section 6.5)*

**Due Wed, Nov 19**

*Read Sections 7.(1 & 2)*

*Take quiz (actually due Thurs, Nov 20)*

**Due Fri, Nov 21**

*Read Section 7.3*

*Do Exercises 7.(4,5,7,8)*

*Take quiz*