# 15 Dynamic Programming

## 15 Dynamic Programming

- Assembly Lines Scheduling
- Steel Rod Cutting
- Matrix-Chain Multiplication (15.2) 矩阵链相乘,或矩阵连乘问题
- Characteristics(Elements) of dynamic programming (15.3)
- Longest Common Subsequence (15.4)
- Optimal binary search trees (15.5)
   最优二叉搜索树

• Given a sequence (chain)  $<A_1, A_2, ..., A_n>$  of n matrices to be multiplied, and we wish to compute the product n 个矩阵相乘,称为'矩阵连乘'(或矩阵链乘法),如何求积?

$$A_1 A_2 A_3 A_4$$
 (15.10)   
  $(A_1 (A_2 (A_3 A_4))), (A_1 ((A_2 A_3) A_4)), ((A_1 A_2) (A_3 A_4)), \dots$ 

• A product of matrices is fully parenthesized if it is either a single matrix or the product of two fully parenthesized matrix products, surrounded by parentheses. Example,

$$A_1$$
,  $(A_1((A_2A_3)A_4))$ ,  $(A_1((A_2A_3)(A_4A_5)))$ .

矩阵连乘全括号: 仅有一个矩阵, 或者两个"矩阵连乘全括号"的乘积且外层包括一个括号, 如:

$$(\underline{A_1}(\underline{(\underline{A_2}\underline{A_3})}\underline{A_4}))$$

这是嵌套的矩阵对,它给出了矩阵连乘的一种求解顺序,也简称"矩阵全括号"。

• We can evaluate (15.10) using the <u>standard algorithm</u> for multiplying pairs of matrices as a subroutine once we have parenthesized it. "矩阵全括号"给出后,就可以用两个矩阵相乘的标准算法作为子程序来计算式 (15.10)。

## Example: Multiplication of two matrices (矩阵相乘)

two  $n \times n$  matrices A and B, Complexity( $C = A \times B$ ) = ?

#### Standard method

$$c_{ij} = \sum_{k=1}^{n} a_{ik} b_{kj}$$

```
MATRIX-MULTIPLY(A, B)

for i \leftarrow 1 to n

for j \leftarrow 1 to n

C[i, j] \leftarrow 0

for k \leftarrow 1 to n

C[i, j] \leftarrow C[i, j] + A[i, k] \cdot B[k, j]

return C
```

Complexity:  $O(n^3)$  multiplications and additions.  $T(n) = O(n^3)$ .

• Given a sequence (chain)  $\langle A_1, A_2, ..., A_n \rangle$  of *n* matrices to be multiplied, and we wish to compute the product

$$A_1 A_2 A_3 A_4$$
 (15.10)

• Matrix multiplication is associative, so all parenthesizations yield the same product. For example, if the chain of matrices is  $<A_1, A_2, A_3, A_4>$ , the product  $A_1A_2A_3A_4$  can be fully parenthesized in five distinct ways: 矩阵连乘满足结合律,因此对所有加括号的方式,矩阵连乘的积相同。例如...

$$(A_1(A_2(A_3A_4)))$$
,  $(A_1((A_2A_3)A_4))$ ,  $((A_1A_2)(A_3A_4))$ ,  $((A_1A_2)(A_3A_4))$ ,  $((A_1A_2)(A_3)A_4)$ .

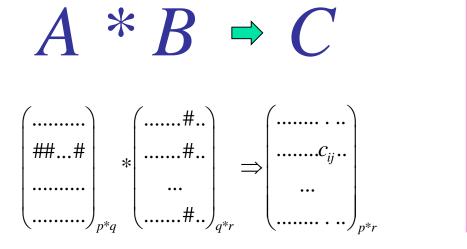
The way we parenthesize a chain of matrices can have a dramatic impact on the cost of evaluating the product.

## 采用不同的加括号方式,可导致差异极大的乘法开销

$$(A_1(A_2(A_3A_4)))$$
,  $(A_1((A_2A_3)A_4))$ ,  $((A_1A_2)(A_3A_4))$ ,  $((A_1(A_2A_3))A_4)$ ,

 $(((A_1A_2)A_3)A_4)$ .

- First, consider the cost of multiplying two matrices.
- Two matrices A and B can be multiplied only if they are compatible: columns of A = rows of B. 仅当矩阵A和B相容时,A和B能相乘
  - If A is  $p \times q$ , B is  $q \times r$ , then C is  $p \times r$ .
  - ◆ The time to compute C is dominated by the number of scalar multiplications in line 7, which is pqr. 两个矩阵相乘,标量乘法的次数是 pqr



```
MATRIX-MULTIPLY(A, B)

1 if columns[A] \neq rows[B]

2 return "error: incompatible dimensions"

3 else for i \leftarrow 1 to rows[A] // p is row[A]

4 for j \leftarrow 1 to columns[B] // r is columns[B]

5 C[i,j] \leftarrow 0

6 for k \leftarrow 1 to columns[A] // q is columns[A]

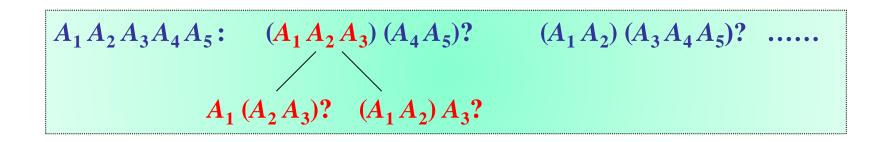
7 C[i,j] \leftarrow C[i,j] + A[i,k] \cdot B[k,j]

8 return C
```

- For  $A_{p\times q}$ ,  $B_{q\times r}$ , C=AB is  $p\times r$ . The # of scalar multiplications is pqr. 标量乘法的次数是 pqr
- 考虑一个简单问题,三个矩阵连乘  $\langle A_1, A_2, A_3 \rangle$ , 设  $A_1$ :  $10 \times 100$ ;  $A_2$ :  $100 \times 5$ ;  $A_3$ :  $5 \times 50$ 
  - $\bullet \quad \stackrel{\omega}{=} \quad A = ((A_1 A_2) A_3) \ ,$ 
    - (a)  $C = A_1 A_2$ , 标量乘法的次数是:  $10 \cdot 100 \cdot 5 = 5,000$ ,  $C_{10 \times 5}$
    - (b)  $A = CA_3$ , 标量乘法的次数是:  $10 \cdot 5 \cdot 50 = 2,500$ ,  $A_{10 \times 50}$

因此,标量乘法的次数是: for a total of 7,500.

- $\stackrel{\text{def}}{=} A = (A_1(A_2A_3))$ ,
- 7,500 < 75,000 (a)  $C_{100\times50} = A_2A_3$ , 标量乘法的次数是:  $100 \cdot 5 \cdot 50 = 25,000$ ,
  - (b)  $A_{10\times50} = A_1C$ , 标量乘法的次数是:  $10\cdot100\cdot50 = 50,000$ ,
  - 因此, 标量乘法的次数是: for a total of 75,000.
- ◆ The first case is 10 times faster than the second. 运算效率十倍之差!



• MCM problem: Given a chain  $\langle A_1, A_2, ..., A_n \rangle$ , i = 1, 2, ..., n, matrix  $A_i$  has dimension  $p_{i-1} \times p_i$ , fully parenthesize the product  $A_1 A_2 ... A_n$  in a way that minimizes the number of scalar multiplications.

n 个矩阵  $< A_1, A_2, ..., A_n >$ , i = 1, 2, ..., n,  $A_i$  的维数为  $p_{i-1} \times p_i$ ,给出矩阵连乘的一种加括号方式,使得标量乘法的次数最小(少)。

#### Counting the number of parenthesizations (有多少种全括号方式)

$$A_1 A_2 A_3 A_4 A_5$$

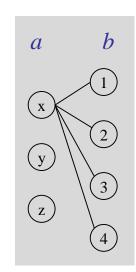
$$(\underbrace{(A_1 A_2 A_3)}_{(A_1 A_2 A_3)} (A_4 A_5))?$$
  $(A_1 A_2) (A_3 A_4 A_5)?$  .....  
 $(A_1 (A_2 A_3))?$   $((A_1 A_2) A_3)?$ 

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

- 暴力穷举(枚举所有的全括号方式)(全括号:矩阵连乘加括号)
- P(n): n 个矩阵连乘,有 P(n) 种全括号方式
  - $n = 1, \frac{1}{2} \frac{1$
  - $n \ge 2$ ,

$$P(n) = \begin{cases} 1 & , & \text{if } n = 1, \\ \sum_{k=1}^{n-1} P(k)P(n-k), & \text{if } n \ge 2. \end{cases}$$
 (15.11)

• 式 (15.11) 的解的计算时间是  $\Omega(2^n)$  ? (guess, then prove), a poor strategy.



#### Counting the number of parenthesizations (有多少种全括号方式)

$$A_1 A_2 A_3 A_4 A_5$$

$$\frac{\left( (A_1 A_2 A_3) (A_4 A_5) \right)?}{(A_1 (A_2 A_3))?} (A_1 A_2) (A_3 A_4 A_5)? \dots$$

$$\frac{\left( (A_1 (A_2 A_3))? ((A_1 A_2) A_3)?}{((A_i (A_{i+1} \dots) (\dots) \dots A_k) (A_{k+1} \dots A_{j-1} A_j))} \right)$$

$$P(n) = \begin{cases} 1 & \text{, if } n = 1, \\ \sum_{k=1}^{n-1} P(k)P(n-k), & \text{if } n \ge 2. \end{cases}$$
 (15.11)

Running time is  $\Omega(2^n)$ 



## **Dynamic Programming to solve MCM:**

**Four Steps** 

#### Step 1: The structure of an optimal parenthesization (最优全括号的结构)

$$\frac{(A_1 A_2 A_3) (A_4 A_5)?}{(A_1 (A_2 A_3))? (A_1 A_2) (A_3 A_4 A_5)?} \dots$$

$$\frac{(A_1 (A_2 A_3))? ((A_1 A_2) A_3)?}{(A_i (A_{i+1} \dots) (\dots) \dots A_k) (A_{k+1} \dots A_{j-1} A_j)}$$

#### 寻找最优子结构...

 $A_{i..j}$  ( $i \le j$ ): 矩阵连乘  $A_i A_{i+1} ... A_k A_{k+1} ... A_j$ 

- i < j, nontrivial, any parenthesization of the product  $A_i A_{i+1} ... A_j$  must split the product between  $A_k$  and  $A_{k+1}$  for some integer k in the range  $i \le k < j$ . 非平凡情况下,全括号 $A_i A_{i+1} ... A_j$ 必定在位置k 把问题分成两个部分,如下
- $\bullet \quad A_{i..k} \cdot A_{k+1..j} = A_{i...j}$

全括号  $A_i A_{i+1} ... A_j$  的计算代价(标量乘法), $cost(A_{i...j})$   $cost(A_{i...j}) = cost(A_{i...k}) + cost(A_{k+1...j})$  + the cost of multiplying  $A_{i...k} \cdot A_{k+1...j}$ 

#### Step 1: The structure of an optimal parenthesization (最优全括号的结构)

#### 最优子结构

◆ 最优全括号 $A_{i...j}$ 在位置 $A_k$ 和 $A_{k+1}$ 处把问题分成两个部分 $A_{i...k}$ 和 $A_{k+1...j}$ 之积  $( (A_i(A_{i+1}...)(...)...A_k)(A_{k+1}...A_{j-1}A_j) )$ 

The parenthesization of the "prefix" subchain  $A_i A_{i+1} ... A_k$  within this optimal parenthesization of  $A_i A_{i+1} ... A_i$  must be an optimal parenthesization of  $A_i A_{i+1} ... A_k$ ?

 $A_{i..i}$  的最优全括号中的 $A_{i..k}$  的全括号必定是 $A_{i..k}$  的最优全括号

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j))$$

$$X$$

**Proof** 

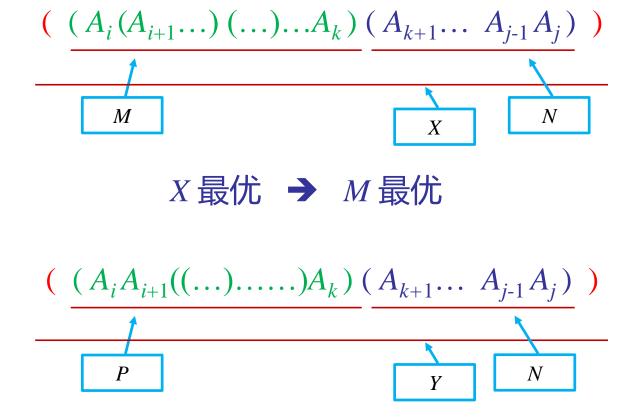
如果 X 最优,则 M 最优

Step 1: The structure of an optimal parenthesization (最优全括号的结构)

**最优子结构:**  $A_{i,j}$  的最优全括号 X 中的  $A_{i,k}$  的全括号M 必定是  $A_{i,k}$  的最优全括号。

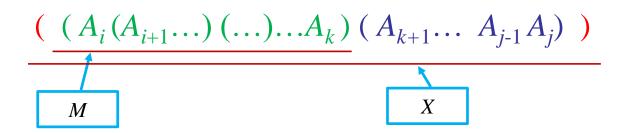
#### **Proof**

对子问题  $A_{i,j}$ , 设存在另一种最优全括号形式 P, 即, P 的标量乘法比 M 还少,显然,Y 对应的全括号所需要的标量乘法次数比X少,跟 X 是最优全括号矛盾。



同理,适用于全括号N

#### Step 1: The structure of an optimal parenthesization (最优全括号的结构)



- 根据最优子结构,可以从用 M 来构造 X (M 是  $A_{i...k}$  的最优解,X 是  $A_{i...j}$  的最优解)
- X 的求解过程: .
  - ◆ 分割: 一个最优解 X 在某个 k 分割  $A_{i...j}$  为  $A_{i...k}$  和  $A_{k+1...j}$
  - ◆ 求子问题  $A_{i..i}$  的最优解 M
  - ◆ 合并: 从 *M* 构造 *X*

Step 1: The structure of an optimal parenthesization (最优全括号的结构)

$$A_{1} A_{2} A_{3} A_{4} A_{5}$$

$$(\underbrace{(A_{1} A_{2} A_{3})}_{(A_{4} A_{5})})? (A_{1} A_{2}) (A_{3} A_{4} A_{5})? \dots$$

$$(A_{1} (A_{2} A_{3}))? ((A_{1} A_{2}) A_{3})?$$

$$(\underbrace{(A_{i} (A_{i+1} \dots) (\dots) \dots A_{k})}_{(A_{k+1} \dots A_{j-1} A_{j})})$$

We must consider all possible places so that we are sure of having examined the optimal one.

需要考虑所有分割位置 k 以确保最优解是其中之一

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

- Define the cost of an optimal solution recursively in terms of the optimal solutions to subproblems.
   根据子问题的最优解可以递归地定义原问题的最优解
- Subproblems  $A_{i..j}$ : determining the minimum cost of a parenthesization of  $A_i A_{i+1} ... A_j$  for  $1 \le i \le j \le n$ .

Not  $A_1A_2...A_j$ , Why? 为什么子问题定义为  $A_{i..j}$  而不是  $A_{1..j}$ 

#### Step 2: A recursive solution (递归解)

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j))$$

$$M$$

m[i, j] = |X|: the minimum # of scalar multiplications to compute  $A_{i..j}$ ; the cost of a cheapest way to compute  $A_{1..n}$  is m[1, n].

 $m[i,j]:A_{i,j}$ 的最优全括号的标量乘法次数。

 $m[1,n]:A_{1,n}$  的最优全括号的标量乘法次数,即,原问题最优值。

- i = j,  $A_{i,i} = A_i$ , **一个矩阵,没有乘法,显然**, m[i, i] = 0, i = 1, 2, ..., n.
- i < j?

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

 $m[i,j]:A_{i..j}$ 的最优全括号的标量乘法次数。

当 i < j,令最优的分割位置是  $A_k$  和  $A_{k+1}$  之间,  $i \le k < j$ . 因此 ,

$$m[i,j] = m[i,k] + m[k+1,j] + p_{i-1}p_kp_j$$

 $A_i$  的维数为  $p_{i-1} \times p_i$ , 因此  $A_{i...k}$  的维数为  $p_{i-1} \times p_k$ ,  $A_{k+1...j}$  的维数为  $p_k \times p_j$ .

#### Step 2: A recursive solution (递归解)

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

$$m[i,j] = m[i,k] + m[k+1,j] + p_{i-1}p_kp_j$$

- 该递归方程基于已知 k,而 k 是未知的变量,取值区间为  $i \le k < j$ .
- 为了解的完备性,需要遍历所有的k,因此有

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1}p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)

m[i,j]:子问题  $A_{i,j}$  的最优全括号的标量乘法次数。

#### Step 2: A recursive solution (递归解)

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1}p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)

m[i,j]:子问题  $A_{i..i}$  的最优全括号的标量乘法次数。

## 构造最优解:

式(15.12)求得的 m[i, j] 是最优值。

定义 s[i,j] 用于存储值 k,表示子问题  $A_{i,j}$  的最优全括号时的分割位置,以便用于构造最优解(找到每个最优加括号的位置)。

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1}p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)

m[1, n] 表示原问题  $A_1A_2...A_n$  的解. 式 (15.12) 的递归算法:

```
RE-MCM(p, i, j)

1 if i == j

2 return 0

3 m[i, j] \leftarrow \infty

4 for k \leftarrow i to j-1

5 q \leftarrow \text{RE-MCM}(p, i, k) + \text{RE-MCM}(p, k+1, j) + p_{i-1}p_kp_j

6 if q < m[i, j]

7 m[i, j] \leftarrow q

8 return m[i, j]
```

**Running time?** 

$$\underbrace{\left( (A_{i}(A_{i+1}...)(...)A_{k})(A_{k+1}...A_{j-1}A_{j}) \right)}_{m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases} (15.12)$$

Recursion, Extremely slow! 直接递归,极慢!

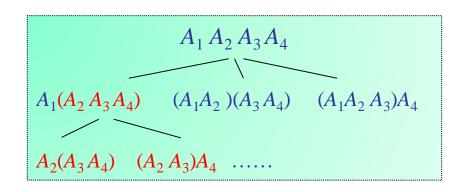
# of subproblems: one problem for each choice of i and j satisfying  $1 \le i \le j \le n$ ?

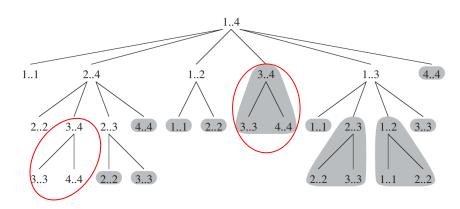
## 矩阵连乘的所有子问题个数为?

$$C_n^2 + C_n^1 = \Theta(n^2)$$
  
 $1 \le i = j \le n, \ C_n^1 = n$   
 $1 \le i < j \le n, \ C_n^2 = n(n-1)/2$ 

$$\underbrace{\left(A_{i}(A_{i+1}...)(...)A_{k}\right)(A_{k+1}...A_{j-1}A_{j})}_{m[i,j] = \begin{cases} 0, & \text{if } i = j, \\ \min_{i \leq k < i} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases} (15.12)$$

- A recursive algorithm may encounter each subproblem many times in different branches of its recursion tree. 递归计算时,递归过程会重复计算相同的子问题
- Overlapping subproblems: the second hallmark of the applicability of dynamic programming. 重叠子问题,动态规划法的第二个重要特点

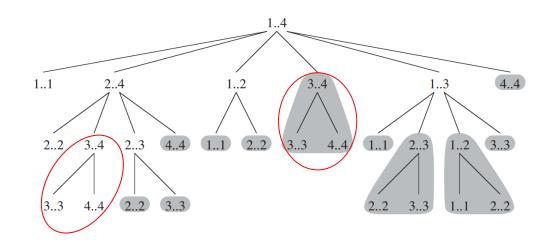




$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_k p_j\}, & \text{if } i < j. \end{cases}$$
(15.12)

- # of subproblems: one problem for each choice of i and j satisfying  $1 \le i \le j \le n$ , or  $C_n^2 + C_n^1 = \frac{n(n-1)}{2} + n = \frac{1}{2}(n^2 + n) = \Theta(n^2)$  in all. 子问题总数为  $\Theta(n^2)$
- Instead of recursive method, computing the optimal cost by using a tabular, bottom-up approach.
   不用递归方法,而采用列表方式、自底向上的方法计算最优解



$$(A_{i}(A_{i+1}...)(...)A_{k})(A_{k+1}...A_{j-1}A_{j}))$$

$$m[i,j] = \begin{cases} 0, & \text{if } i = j, \\ \min_{i \leq k \neq i} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases}$$

$$(15.12)$$

问题: 矩阵 $A_i$ , 维数 $p_{i-1} \times p_i$ 

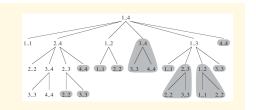
输入:  $p = \langle p_0, p_1, ..., p_n \rangle$ .

程序: 表格(数组) $m_{n,n}$ 存储每一个子问题的最优值 m[i,j];

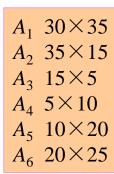
辅助表格  $s_{n,n}$ , 其中 s[i,j] 记录求 m[i,j] 时的最优分割位置 k.

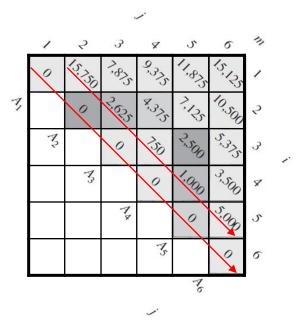
```
MCM-DP(p)
1 n \leftarrow length[p] - 1
2 for i \leftarrow 1 to n
3 	 m[i, i] \leftarrow 0
4 for l \leftarrow 2 to n // l is the chain length.
        for i \leftarrow 1 to n - l + 1
   j \leftarrow i + l - 1
      m[i,j] \leftarrow \infty
            for k \leftarrow i to j - 1
                    q \leftarrow m[i, k] + m[k+1, j] + p_{i-1}p_kp_i
10
                   if q < m[i, j]
                         m[i,j] \leftarrow q
11
                        s[i, j] \leftarrow k
13 return m and s
```

$$\underbrace{\left( A_{i}(A_{i+1}...)(...)...A_{k} \right) \left( A_{k+1}... A_{j-1}A_{j} \right)}_{m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases} (15.12)$$



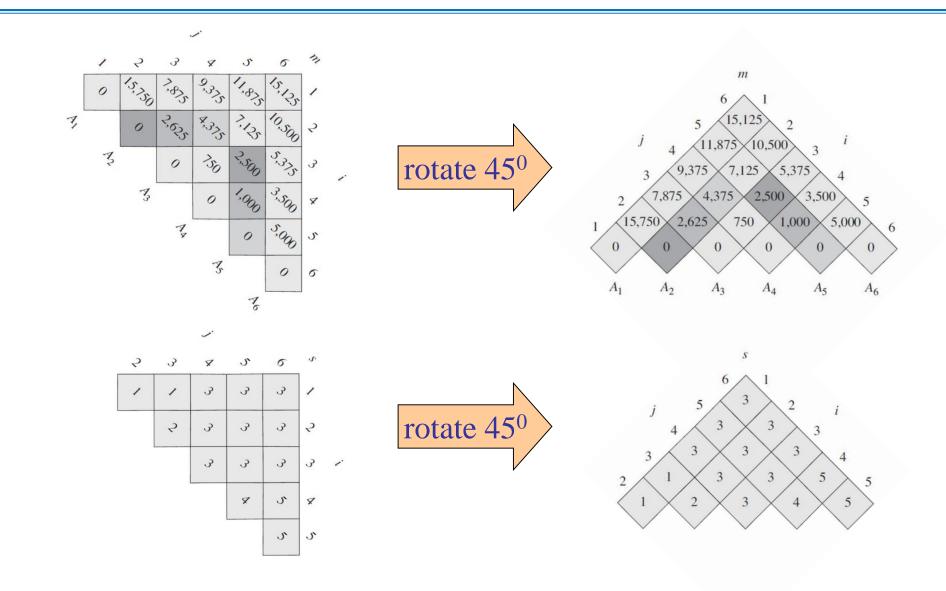
```
MCM-DP(p)
1 n \leftarrow length[p] - 1
2 for i \leftarrow 1 to n
        m[i, i] \leftarrow 0
4 for l \leftarrow 2 to n // l is the chain length.
        for i \leftarrow 1 to n - l + 1
       j \leftarrow i + l - 1
      m[i,j] \leftarrow \infty
             for k \leftarrow i to j - 1
                    q \leftarrow m[i, k] + m[k+1, j] + p_{i-1}p_kp_i
                    if q < m[i, j]
                         m[i,j] \leftarrow q
                       s[i,j] \leftarrow k
13 return m and s
```





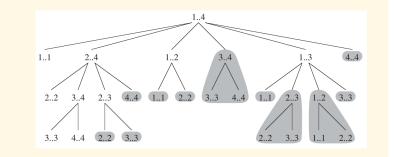
5	3	¥	5	6	S	
1	1	3	3	3	1	
	5	ઝ	3	3	5	
		ۍ	3	3	ۍ	1
			8	5	×	
				5	5	

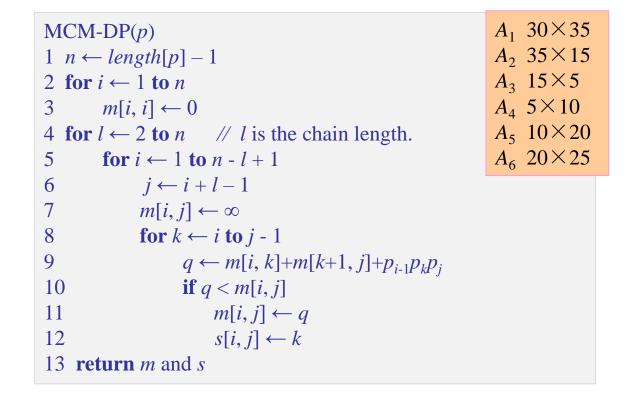
Step 3: Computing the optimal costs (计算最优全括号的乘法次数)

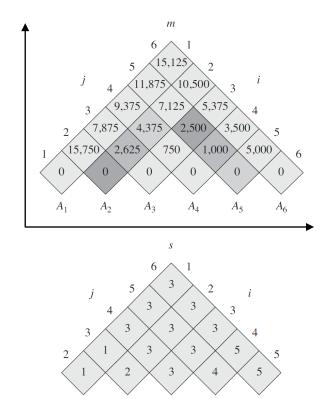


( 
$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$
 )

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1}p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)







```
MCM-DP(p)
1 n \leftarrow length[p] - 1
2 for i \leftarrow 1 to n
         m[i, i] \leftarrow 0
4 for l \leftarrow 2 to n // l is the chain length.
         for i \leftarrow 1 to n - l + 1
               j \leftarrow i + l - 1
              m[i,j] \leftarrow \infty
               for k \leftarrow i to j - 1
9
                      q \leftarrow m[i, k] + m[k+1, j] + p_{i-1}p_kp_j
10
                      if q < m[i, j]
11
                           m[i,j] \leftarrow q
                           s[i,j] \leftarrow k
12
13 return m and s
```

The running time?

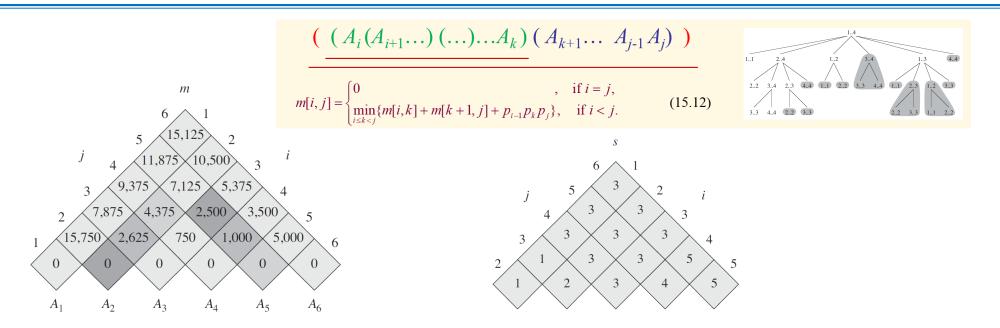
Space requirement?

```
MCM-DP(p)
1 n \leftarrow length[p] - 1
2 for i \leftarrow 1 to n
        m[i, i] \leftarrow 0
4 for l \leftarrow 2 to n // l : n-1 times
        for i \leftarrow 1 to n - l + 1 // i : n - l + 1 times
              j \leftarrow i + l - 1
              m[i,j] \leftarrow \infty
              for k \leftarrow i to j - 1 // k : j-i=l-1 times
                     q \leftarrow m[i, k] + m[k+1, j] + p_{i-1}p_kp_i
10
                     if q < m[i, j]
11
                          m[i,j] \leftarrow q
                          s[i,j] \leftarrow k
12
13 return m and s
```

#### **Exercise:**

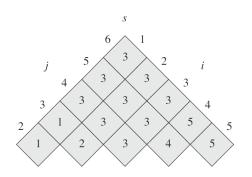
$$T(n) = \sum_{l=2}^{n} (n-l+1)(l-1)$$

The running time?



- MCM-DP determines the optimal number m[i, j], but does not directly show how to multiply the matrices.
  - 算法MCM-DP 给出了如何求最优全括号的乘法次数 m[i, j],但对于按什么顺序来相乘各矩阵,没有给出具体方法 (giving optimal value, no optimal solution)
- Constructing an optimal solution from table s[1.. n-1, 2.. n].
   利用辅助数组 s 来构造最优解

$$\underbrace{\left(\begin{array}{c} \left(A_{i}\left(A_{i+1}...\right)\left(...\right)...A_{k}\right)\left(A_{k+1}...A_{j-1}A_{j}\right)}_{m[i,j] = \begin{cases} 0 & , & \text{if } i = j, \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases} (15.12)$$

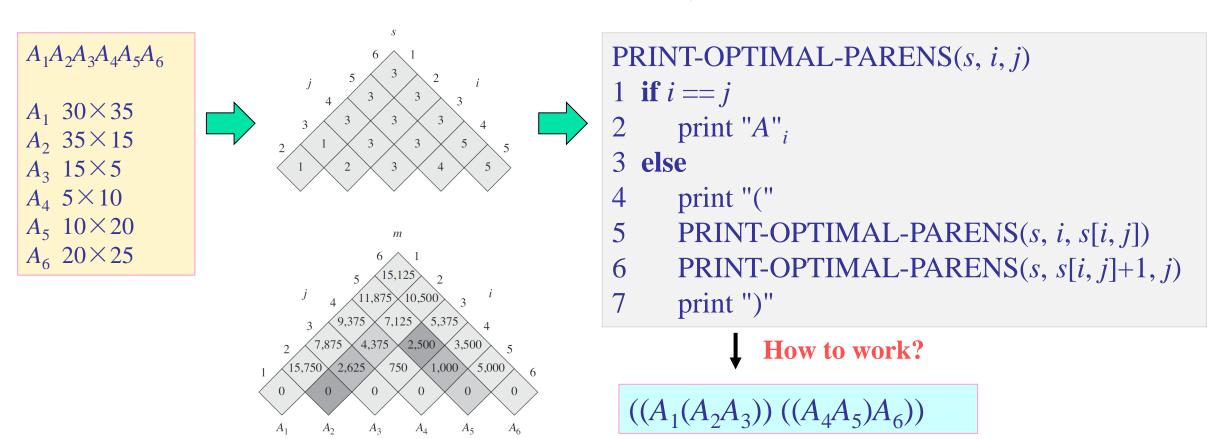


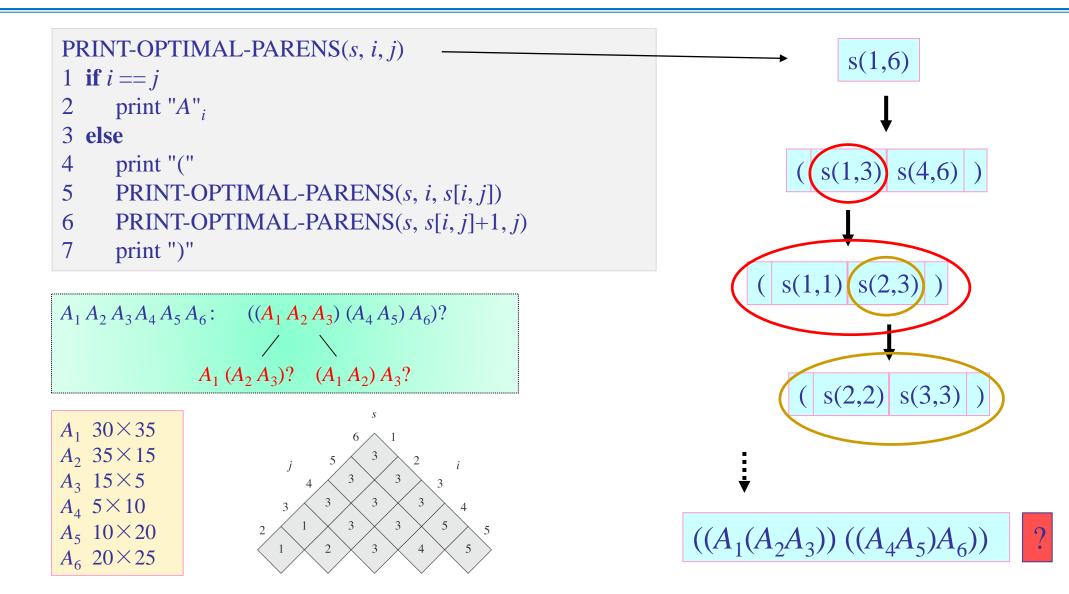
- Each entry s[i,j] records the value of k such that the optimal parenthesization of  $A_iA_{i+1}\cdots A_j$  splits the product between  $A_k$  and  $A_{k+1}$ . Thus, the final matrix multiplication in computing  $A_{1..n}$  optimally is  $A_{1..s[1,n]}A_{s[1,n]+1..n}$ . s[i,j] 记录值 k,表示在矩阵连乘  $A_iA_{i+1}\cdots A_j$  的最优全括号中,分割点位于  $A_k$  和  $A_{k+1}$  之间。因此,矩阵连乘  $A_{1..n}$  的最优分割方式为  $(A_1A_2...A_{s[1,n]})(A_{s[1,n]+1}...A_n)$ .
  - ◆ 矩阵连乘的括号可以递归计算,

s[1, s[1, n]] 计算 splits  $A_{1...s[1, n]}$  的分割位置 s[s[1, n] + 1, n] 计算 splits  $A_{s[1, n] + 1...n}$  的分割位置

• PRINT-OPTIMAL-PARENS(s, i, j)

PRINT-OPTIMAL-PARENS(s, i, j) printing an optimal parenthesization of  $< A_i, A_{i+1}, ..., A_j >$  recursively, given the s table. The initial call i=1, j=n. 求得辅助矩阵 s 后,如下递归算法输出最优加括号





## 15 Dynamic Programming

- Assembly Lines Scheduling
- Steel Rod Cutting
- Matrix-Chain Multiplication (15.2)
   矩阵链相乘,或矩阵连乘问题
- Characteristics(Elements) of dynamic programming (15.3)
- Longest Common Subsequence (15.4)
- Optimal binary search trees (15.5)
   最优二叉搜索树

输入法的词库选择

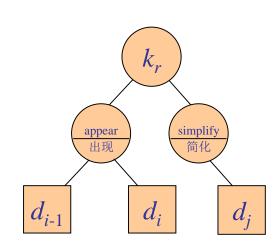


Design a program to translate text from English to Chinese (翻译软件的词库的字典顺序如何设计)



#### Design a program to translate text from English to Chinese



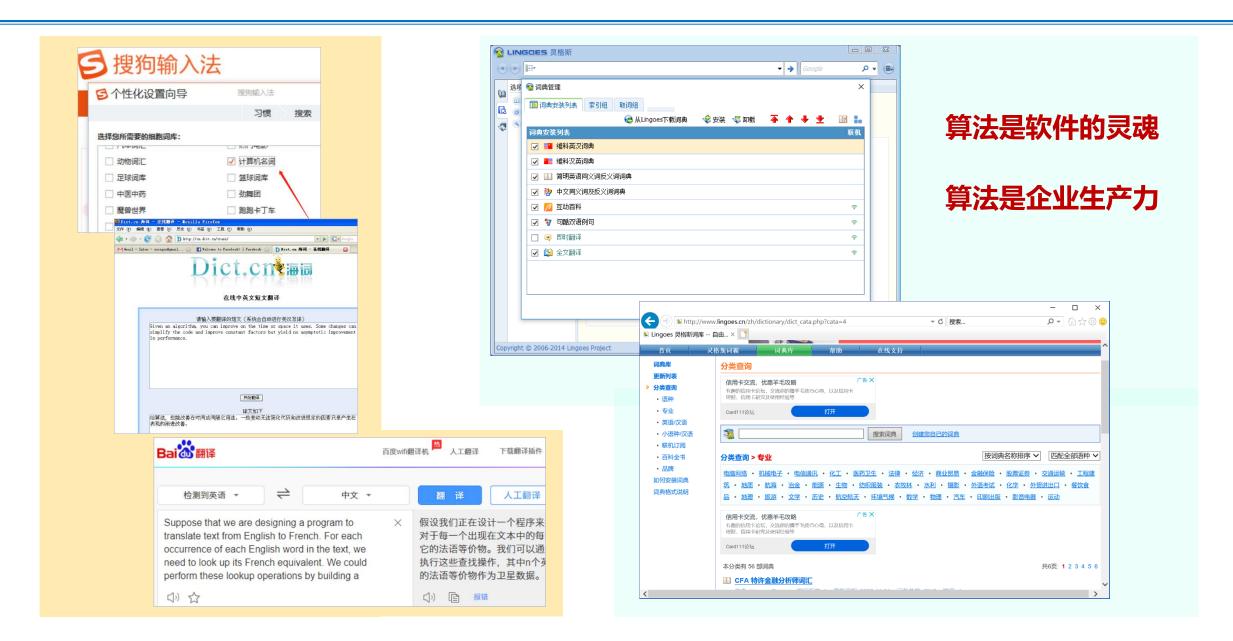


生词表	
字段1	字段2
aggregate	综合,总体
amortized	分摊,平摊
arbitrary	任意的,武断的
auxiliary	辅助的
binomial	二项的,二项式的
bog	沼泽,陷于泥沼
design	设计

# an O(n) search time per occurrence by using any linear table operation 设计为线性表,每个单词查找的效率为 O(n)



生词表	
字段1	字段2
aggregate	综合,总体
amortized	分摊,平摊
arbitrary	任意的,武断的
auxiliary	辅助的
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design	设计



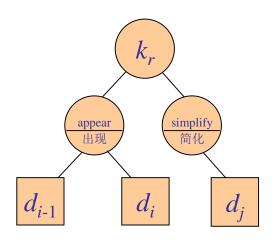
- lookup operations: build a binary search tress (BST) with
  - *n* English words as keys
  - Chinses equivalents as satellite data

为了高效查找,需要构建一棵二叉搜索树,树的 每一个节点是一个单词,包括关键字(比如英语) 及其从属数据(比如中文)。

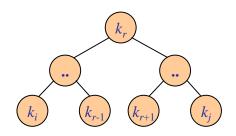
二叉搜索树,二叉排序树:二叉树,一个节点大于其左子树的所有节点,小于其右子树的所有节点。



- Because we will search the tree for each individual word in the text, we want the total time spent searching to be as low as possible.
  - 对于文本中出现的每个单词,都需要搜索该二叉树,如何设计搜索树,使得总的搜索次数最少?
- an  $O(\lg n)$  search time per occurrence by using any balanced BST. 对于任何一个单词的搜索,使用平衡二分搜索法的时间为 $O(\lg n)$ .



A balanced BST...
平衡二叉搜索树



生词表	
	N. 40
字段1	字段2
aggregate	综合,总体
amortized	分摊, 平摊
arbitrary	任意的,武断的
auxiliary	辅助的
binomial	二项的,二项式的
bog	沼泽, 陷于泥沼

However, Words appear with different frequencies...?

每个单词在文本中出现的频率(频次)不同,该如何设计二叉搜索树?

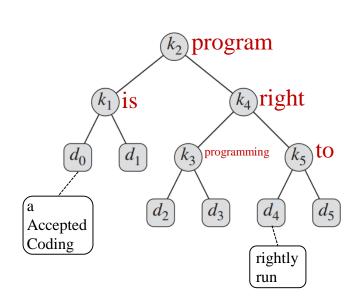
## An example

给定一颗 BSTT, 对文本中的所有单词在树中进行搜索,访问的节点总数称为期望的搜索代价 (expected cost of a search in T)

$$E[\text{search cost in } T] = \sum_{i=1}^{n} (\text{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\text{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \text{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \text{depth}_{T}(d_{i}) \cdot q_{i} , \qquad (15.16)$$

文本示例: Accepted Coding programming is rightly to program a right program to run a right program



node	word	depth	times 单词出现次数	probability 单词出现频次 <i>p<sub>i</sub></i>	contribution	
	a		2			
$d_0$	Accepted	2	1	4/15	12/15	
	Coding		1	]		
$k_1$	is:是	1	1	1/15	2/15	
$k_2$	program : 编程	0	3	3/15	3/15	
$k_3$	programming : 编程	2	1	1/15	3/15	
$k_4$	right:正确	1	2	2/15	4/15	
1	rightly	3	1	2/15	0/15	
$d_4$	run	3	1	2/15	8/15	
$k_5$	to: 去	2	2	2/15	6/15	
$d_{ m other}$			0			
total					38/15	

对示例问题,构建二叉搜索树(字典) BSTT,其中红色的单词为字典中能查到的,黑色单词为字典中查不到的,则用搜索树T对文本进行全文搜索的搜索代价为 38/15。树不一样,搜索代价不同。

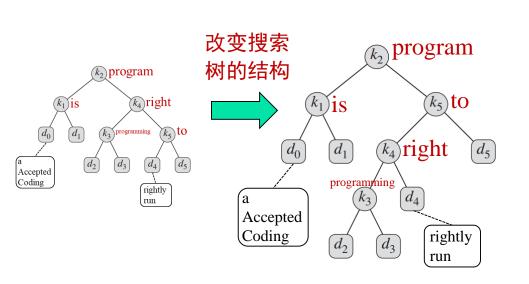
## An example

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$$= 1 + \sum_{i=1}^{n} \text{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \text{depth}_{T}(d_{i}) \cdot q_{i} , \qquad (15.16)$$

文本示例: Accepted Coding programming is rightly to program a right program to run a right program



node	word	depth	times 单词出现次数	probability 单词出现频次 <i>p<sub>i</sub></i>	contribution	
	a		2			
$d_0$	Accepted	2	1	4/15	12/15	
	Coding		1			
$k_1$	is:是	1	1	1/15	2/15	
$k_2$	program:编程	0	3	3/15	3/15	
$k_3$	programming : 编程	3	1	1/15	4/15	
$k_4$	right:正确	2	2	2/15	6/15	
ı	rightly	3	1	2/15	8/15	
$d_4$	run	3	1	2/13	8/13	
$k_5$	to: 去	1	2	2/15	4/15	
$d_{ m other}$			0			
total					39/15	

搜索树的结构改变,对文本的搜索代价变化。

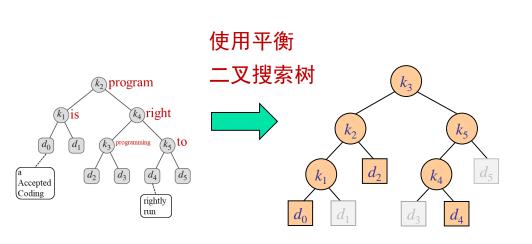
## An example

给定一颗 BST T,对文本中的所有单词在树中进行搜索,访问的节点总数称为期望的搜索代价 (expected cost of a search in T)

$$E[\text{search cost in } T] = \sum_{i=1}^{n} (\text{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\text{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \text{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \text{depth}_{T}(d_{i}) \cdot q_{i} , \qquad (15.16)$$

文本示例: Accepted Coding programming is rightly to program a right program to run a right program



node	word	depth	times 单词出现次数	probability 单词出现频次 <i>p<sub>i</sub></i>	contribution
	a		2		
$d_0$	Accepted	3	1	4/15	16/15
	Coding		1		
$k_1$	is : 是	2	1	1/15	3/15
$k_2$	program : 编程	1	3	3/15	6/15
$k_3$	programming : 编程	0	1	1/15	1/15
$k_4$	right:正确	2	2	2/15	6/15
1	rightly	3	1	2/15	0/15
$d_4$	run	3	1	2/15	8/15
$k_5$	to:去	1	2	2/15	4/15
$d_{ m other}$			0		
total					44/15

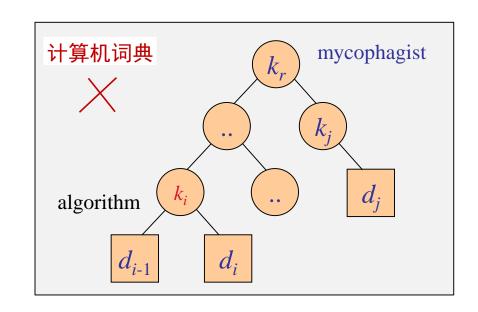
使用平衡二叉树搜索树,搜索代价为 44/15 怎样构建搜索树,使得搜索代价最小?

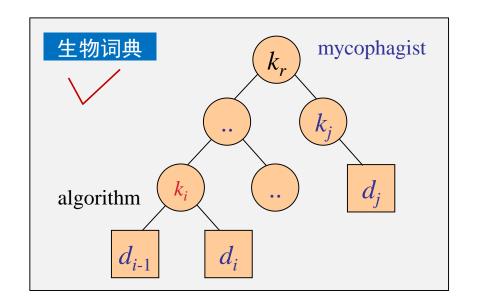
However, Words appear with different frequencies...?

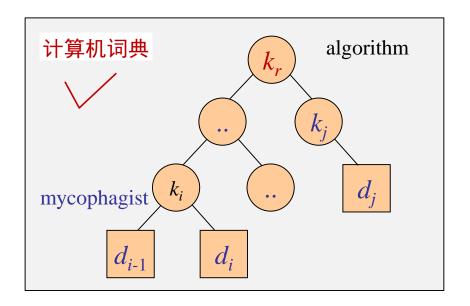
每个单词在文本中出现的频率(频次)不同,该如何设计二叉搜索树?

- It may be:
  - "algorithm" (frequently used) appears far from the root; "mycophagist" (rarely used, 食菌者) appears near the root.
- Such an organization would slow down the translation, since # of nodes visited when searching for a key in a BST is 1+depth.

在翻译时,一个单词(节点)在搜索树BST中每次被访问时,需要访问 1+depth 次。对一个计算机类的词典,如果把出现频率高的单词,如"algorithm",放在远离搜索树树根的节点处,把出现频率低的单词,如"mycophagist",放在靠近树根的节点处,这种设计会让翻译效率很低(翻译速度很慢)。





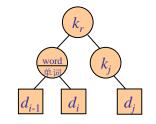


- Words appear with different frequencies
- It may be: "algorithm" (frequently used) appears far from the root; "mycophagist" (rarely used, 食菌者) appears near the root.
- Such an organization would slow down the translation, since # of nodes visited when searching for a key in a BST is 1+depth.
- We want words that occur frequently in the text to be placed nearer the root.

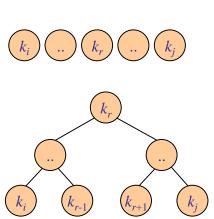
## 文本中出现频率高的单词,应该放在BST中靠近树根处

• Moreover, there may be words in the text for which there is no Chinese translation, and such words might not appear in the BST at all.

文本中有些英语单词没有对应的汉语译文,即这些英语单词不出现在二叉搜索树BST的"词典"中(置于BST的树叶)



• How do we organize a BST so as to minimize the number of nodes visited in all searches, given that we know how often each word occurs? 设已知每个单词出现的概率,如何设计一颗二叉搜索树BST,使得在对文本中每个单词的所有搜索中,被访问的节点的总数最少?



**BST**: Given a sequence  $K = \langle k_1, k_2, ..., k_n \rangle$  of n distinct keys in sorted order  $(k_1 \langle k_2 \rangle \cdots \langle k_n)$ , how to build a BST?

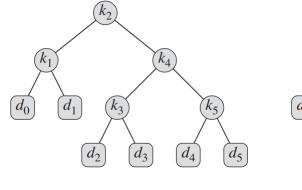
已知 n 个关键字, 其值的集合为  $K = \langle k_1, k_2, ..., k_n \rangle$ ,  $(k_1 < k_2 < \cdots < k_n)$ , 如何构建一颗二叉搜索树

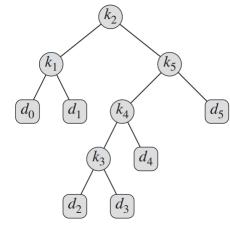
- 关键字  $k_i$  在文本中的搜索概率(频率)为  $p_i$ ,(文本总共有 M 个单词,  $k_i$  出现了 I 次,则  $p_i = I/M$ )
- 文本中有些单词在 K 中不存在,称其为虚关键字,有 n+1 类这样的单词,记为  $< d_0, d_1, ..., d_n >$  ,其出现频率分别为  $q_0, q_1, ..., q_n$  ,其中, $d_0$  对应字典序小于  $k_1$  的一类单词;  $d_n$  对应字典序大于  $k_n$  的一类单词; 对  $1 \le i \le n-1$ , $d_i : k_i < d_i < k_{i+1}$
- BST如图所示,每个关键字  $k_i$  是内节点,虚关键字  $d_i$  是树叶

$(k_1)$	<u></u>	$(k_i)$	<u></u>	$(k_n)$
$\binom{\mathbf{k}_1}{\mathbf{k}_1}$	$\Box$	$\binom{\kappa_i}{k}$	· ·	$\binom{\kappa_n}{n}$

i	0	1	2	3	4	5
$p_i$		0.15	0.10	0.05	0.10	0.20
$q_{i}$	0.05	0.10	0.05	0.05	0.05	0.10

	p	1	p	2		•	p	n	
G	$l_0$	q	11	9	$l_2$	•	••	G	$I_n$

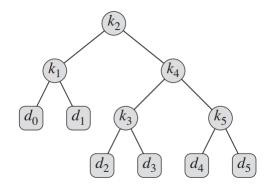




在BST中查找文本的每个单词,有两种情况:能找到(内节点),不能找到(树叶节点)

$$\sum_{i=1}^{n} p_i + \sum_{i=0}^{n} q_i = 1 \tag{15.15}$$

• 给定一颗 BSTT, 对文本中的所有单词在树中进行搜索,访问的节点总数称为期望的搜索代价 (expected cost of a search in T)



 $depth_T$  表示一个节点在树 T 中的 高度(深度),树根高度为0

$$E[\text{search cost in } T] = \sum_{i=1}^{n} (\text{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\text{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \text{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \text{depth}_{T}(d_{i}) \cdot q_{i} , \qquad (15.16)$$

对 BSTT,

期望的搜索代价

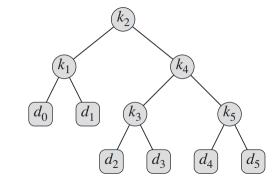
E[search cost in *T*]

$$= \sum_{i=1}^{n} (\operatorname{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\operatorname{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \operatorname{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \operatorname{depth}_{T}(d_{i}) \cdot q_{i}$$

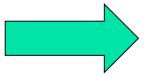


i	0	1	2	3	4	5
$p_{i}$		0.15	0.10	0.05	0.10	0.20
$q_{i}$	0.05	0.10	0.05	0.05	0.05	0.10

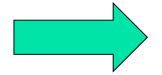


node	depth	probability	contribution
$k_1$	1	0.15	0.30
$k_2$	0	0.10	0.10
$k_3$	2	0.05	0.15
$k_4$	1	0.10	0.20
$k_5$	2	0.20	0.60
$d_0$	2	0.05	0.15
$d_1$	2	0.10	0.30
$d_2$	3	0.05	0.20
$d_3$	3	0.05	0.20
$d_4$	3	0.05	0.20
$d_5$	3	0.10	0.40
Total			2.80

问题



BST T



搜索代价 E[T]

对 BSTT,

期望的搜索代价

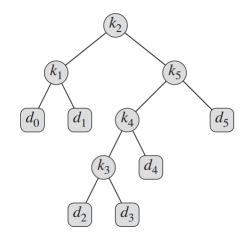
E[search cost in *T*]

$$= \sum_{i=1}^{n} (\operatorname{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\operatorname{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \operatorname{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \operatorname{depth}_{T}(d_{i}) \cdot q_{i}$$

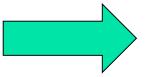


i	0	1	2	3	4	5
$p_{i}$		0.15	0.10	0.05	0.10	0.20
$q_{i}$	0.05	0.10	0.05	0.05	0.05	0.10

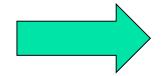


node	depth	probability	contribution
$k_1$	1	0.15	0.30
$k_2$	0	0.10	0.10
$k_3$	3	0.05	0.20
$k_4$	2	0.10	0.30
$k_5$	1	0.20	0.40
$d_0$	2	0.05	0.15
$d_1$	2	0.10	0.30
$d_2$	4	0.05	0.25
$d_3$	4	0.05	0.25
$d_4$	3	0.05	0.20
$d_5$	2	0.10	0.30
Total			2.75

问题

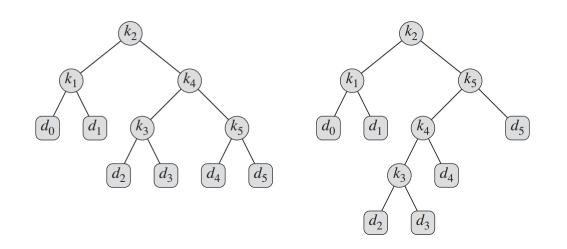


BST T



搜索代价 E[T]

• Optimal BST 最优二叉搜索树:输入各个关键字的概率集,期望搜索代价最小的树

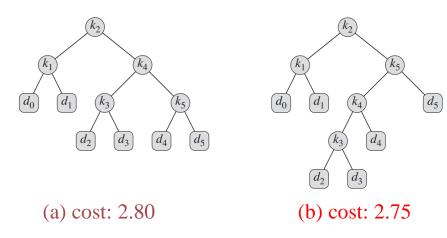


i	0	1	2	3	4	5
$p_i$		0.15	0.10	0.05	0.10	0.20
$q_{i}$	0.05	0.10	0.05	0.05	0.05	0.10

• 如何构造最优二叉搜索树 OBST?

Intuitively, the overall height is smallest; the key with the greatest probability at the root. 直观上看,树的高度应最小;出现概率(频率)最大的关键字应作为树根。

Optimal BST 最优二叉搜索树:输入各个关键字的概率集,期望搜索代价最小的树



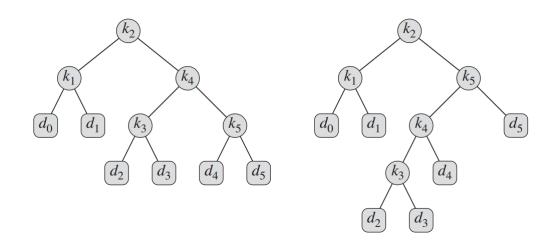
i	0	1	2	3	4	5
$p_i$		0.15	0.10	0.05	0.10	0.20
$q_i$	0.05	0.10	0.05	0.05	0.05	0.10

- 图(b)是一颗OBST, 其搜索代价为 2.75
  - ◆ An Optimal BST is not necessarily a tree whose overall height is smallest. 不一定要求树的高度最小
  - Nor can we necessarily construct an Optimal BST by always putting the key with the greatest probability at the root. (The lowest expected cost of any BST with  $k_5$  (the greatest probability) at the root is 2.85.)

不一定将概率最大的 key 放在树根,如...

- Exhaustive checking of all possibilities fails to yield an efficient algorithm.
  - ALS, RodCut, MCM

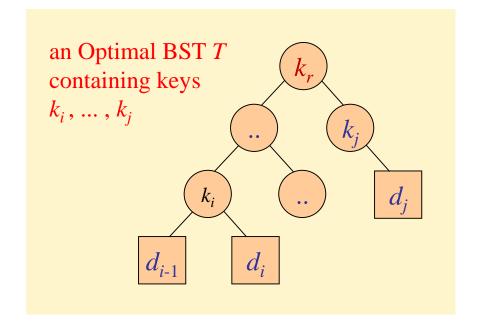
• The # of BST with n nodes is  $\Omega(4^n/n^{3/2})$  (Problem 12-4).



• Not surprisingly, we will solve this problem with dynamic programming.

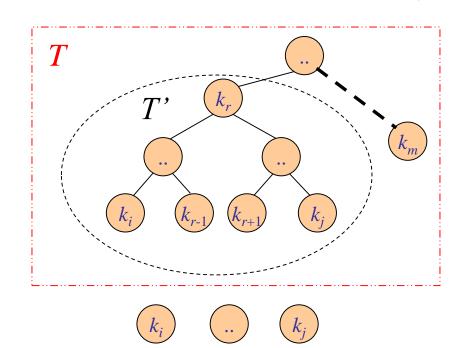
- Start with an observation about subtrees.
- Consider any subtree of a BST
  - It must contain keys in a contiguous range  $k_i$ , ...,  $k_j$ , for some  $1 \le i \le j \le n$ .
  - In addition, the subtree must also have as its leaves the dummy keys  $d_{i-1}$ , ...,  $d_i$ .
- Optimal substructure?

考虑包括节点  $k_i$ , ...,  $k_j$ ,  $1 \le i \le j \le n$ , 的一颗 BST T, 是否有最优子结构?



Optimal substructure: If an Optimal BST T has a subtree T' containing keys  $k_i$ , ...,  $k_j$ , then this subtree T' must be optimal as well for the subproblem with keys  $k_i$ , ...,  $k_j$  and dummy keys  $d_{i-1}$ , ...,  $d_j$ .

最优子结构:设 T'为 OBST T 的一个子树,T'包含keys  $k_i$ , ...,  $k_j$ , 那么 T'是子问题〔关于keys  $k_i$ , ...,  $k_j$ 和dummy keys  $d_{i-1}$ , ...,  $d_j$ 〕的 OBST



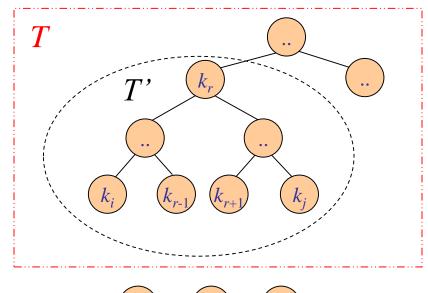
T: search tree of  $k_i$ , ...,  $k_m$ 

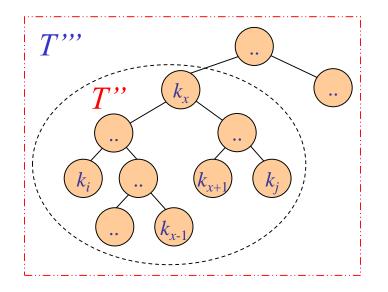
T': search tree of  $k_i$ , ...,  $k_j$ 

T is OBST  $\rightarrow$  T' is OBST

Idea of Proof: Cut-and-paste. 证明思想: 剪切粘贴法

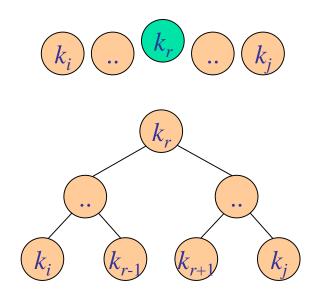
设 BST T 的搜索代价最小,T' 是 T 的搜索子树。如果 T 的一颗搜索子树 T'' 的搜索代价比 T' 的搜索代价更小,在搜索树 T 中把 T' 换成 T'' ,得到一颗新的搜索树 T''',其搜索代价比 T 更小,与假设矛盾。



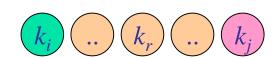


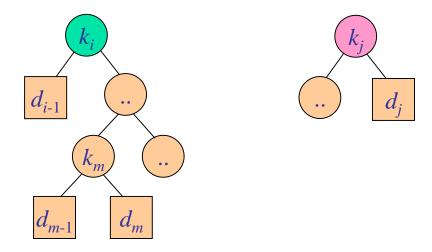
$$E[cost(T)] = 1 + \sum_{i=1}^{n} depth_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} depth_{T}(d_{i}) \cdot q_{i}$$

- 通过子问题的最优解构造原问题的最优解
- 给定 keys  $k_i$ , ...,  $k_j$ , 设某个  $k_r$  ( $i \le r \le j$ ) 是最优搜索子树的根,子问题包括由  $k_i$ , ...,  $k_{r-1}$  构成的搜索子树,和由  $k_{r+1}$ , ...,  $k_i$  构成的子树
- 需要检验所有的  $k_r$ ,  $i \le r \le j$ , 并求相应子树的 OBST, 然后求出原问题的 OBST



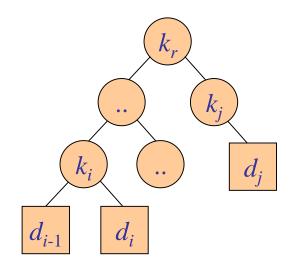
- 特例, 空子树: 不包括任何 key
- 给定 keys  $k_i, ..., k_j$ ,
  - ◆ 如果  $k_i$  是搜索子树的根,其左子树没有 key,此时,值小于  $k_i$  的值的所有 dummy keys 记为  $d_{i-1}$  ,表示左子树。
  - ◆  $k_j$  是搜索子树的根,同理,其右子树记为  $d_i$ ,其值大于  $k_i$  的值。

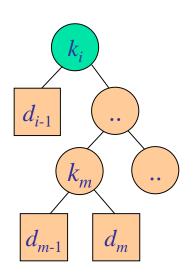




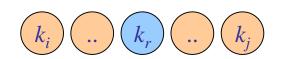
子问题: 给定 keys  $k_i$ , ...,  $k_j$ ,  $i \ge 1$ ,  $j \le n$ , and  $j \ge i$ -1,设求一颗 OBST(当 j = i-1,OBST 只有一个节点,即 dummy key  $d_{i-1}$ .)

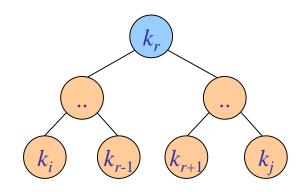
- e[i,j]: 一颗 OBST 的搜索代价,最优值
- 原问题为 *e*[1, *n*]
- 当 j = i-1, OBST 只有一个节点  $d_{i-1}$ ,  $e[i, i-1] = q_{i-1}$
- $rightharpoonup j \ge i$ ?





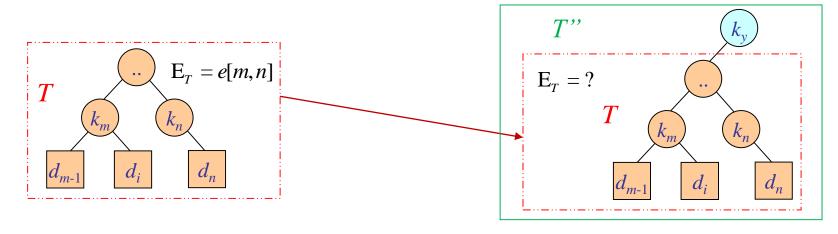
当  $j \ge i$ ,选树根节点  $k_r$  子,求子问题:包括 keys  $k_i$  , ...,  $k_{r-1}$ 的OBST,作为其左子树;包括 keys  $k_{r-1}$  , ...,  $k_j$  的OBST,作为其右子树。





## 当一个搜索子树 T 成为另一个节点 $k_v$ 的子树时(如图),对 T 的搜索代价如何变化?

◆ 子树T的每一个节点在T"中深度增加1,其搜索代价增加值为其所有节点的概率和,如式:



#### 所有节点的概率和:

$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l$$
 (15.17)

$$E_{T} = \sum_{x=m}^{n} (\operatorname{depth}(k_{x}) + 1 + 1) \cdot p_{i} + \sum_{x=m-1}^{n} (\operatorname{depth}(d_{x}) + 1 + 1) \cdot q_{x}$$

$$= \sum_{x=m}^{n} (\operatorname{depth}(k_{x}) + 1) \cdot p_{i} + \sum_{x=m-1}^{n} (\operatorname{depth}(d_{x}) + 1) \cdot q_{x} + \sum_{x=m}^{n} p_{i} + \sum_{x=m-1}^{n} q_{x}$$

$$= e[m, n] + w[m, n]$$

增量为 w[m, n]

OBST T与 OBS-subTree T'的关系:

如果  $k_r$  是 keys  $k_i$ , ...,  $k_j$  的一颗OBST 的树根,则

$$e[i,j] = p_r + (e[i,r-1] + w[i,r-1]) + (e[r+1,j] + w[r+1,j])?$$

因为  $w[i,j] = w[i,r-1] + p_r + w[r+1,j]$ 

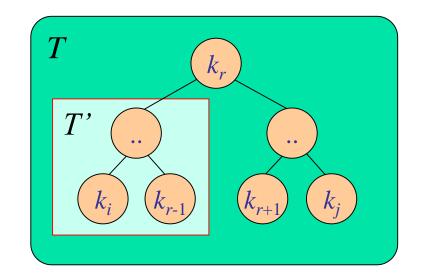
$$\left(w[i,r-1] = \sum_{l=i}^{r-1} p_l + \sum_{l=i-1}^{r-1} q_l \quad , \quad w[r+1,j] = \sum_{l=r+1}^{j} p_l + \sum_{l=r}^{j} q_l\right)$$

所以

$$e[i,j] = e[i,r-1] + e[r+1,j] + w[i,j]$$
 (15.18)

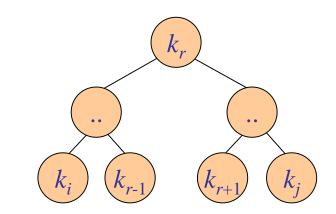
e[i,j]: 问题的最优值; e[i,r-1], e[r+1,j], 子问题的最优值。

递归方程 (15.18) 假定已知节点  $k_r$ ,但事实上节点未知,怎么办?



• 遍历所有的  $k_r$ 

$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}$$
(15.19)



• 定义辅助数组 root[i, j] = r,用于记录一颗 OBST 的树根(记录OBST的结构),此时  $k_r$  是树根。

$$A_{i} ... A_{k} A_{k+1} ... A_{j}$$

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases}$$

$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \leq r \leq j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \leq j. \end{cases}$$

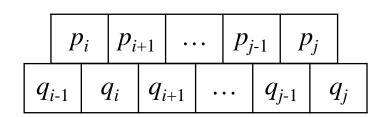
$$(15.12)$$

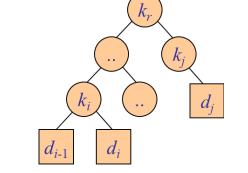
- Similarity: OBST and matrix-chain multiplication.
- A direct, recursive implementation would be as inefficient?
- Store the e[i, j] values in a table e[1...n+1, 0...n].
  - The first index runs to n+1, in order to have a subtree containing only  $d_n$ , need to compute and store e[n+1, n]. The second index starts from 0, in order to have a subtree containing only  $d_0$ , need to compute and store e[1, 0]. 数组的索引范围略有区别
- root[i, j], recording the root of the subtree containing keys  $k_i$ , ...,  $k_j$ .

#### 继续优化

$$e[i,j] = e[i,r-1] + e[r+1,j] + w[i,j]$$
 (15.18)

• 输入: 每个 key 出现的频率

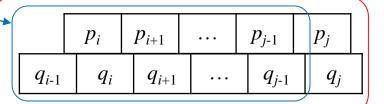




- 在式(15.18)中,每次计算 e[i,j] 时,无需 O(n) 遍历计算 w[i,j] , …
- 基本情况,只有一个假节点, $w[i, i-1] = q_{i-1}$  for  $1 \le i \le n$ .
- $rightharpoonup j \geq i$ ,

$$w[i, j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i, j-1] + p_j + q_j$$
 (15.20)

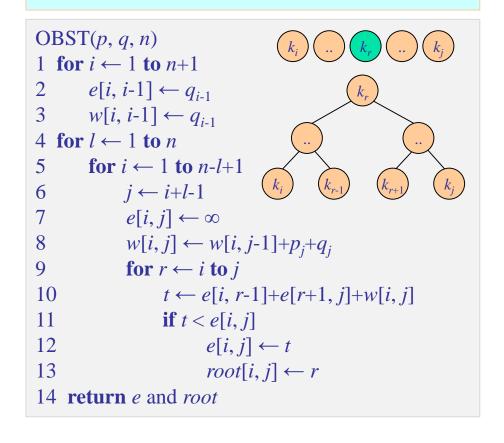
• 计算每个w[i,j],从O(n) 从优化为O(1)



$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}$$

$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$

$$(15.20)$$



#### $((A_i(A_{i+1}...)(...)...\underline{A_k})(A_{k+1}...A_{j-1}A_j))$

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1} p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)

VS

```
OBST(p, q, n)
1 for i \leftarrow 1 to n+1
e[i, i-1] \leftarrow q_{i-1}
3 \quad w[i, i-1] \leftarrow q_{i-1}
4 for l ← 1 to n // 求 l 个元素的 Opti-BST
       for i \leftarrow 1 to n-l+1 // ?1
   j \leftarrow i+l-1 // ?2
   e[i,j] \leftarrow \infty
   w[i,j] \leftarrow w[i,j-1] + p_i + q_i
      for r \leftarrow i to j
10
                 t \leftarrow e[i, r-1] + e[r+1, j] + w[i, j]
11
                 if t < e[i, j]
                       e[i,j] \leftarrow t
13
                       root[i, j] \leftarrow r
14 return e and root
```

```
e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}  (15.19)
```

```
w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j  (15.20)
```

```
?1
e[i, j]:
l 个元素的 Opti-BST 的 cost i = 1, j = l, i = 2, j = l+1, ...
i = x, j = n, n-x+1 = l = > x = n-l+1
?2
j-i+1 = l = = > j = i+l-1
```

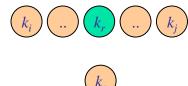
$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w(i,j)\} & \text{if } i \le j. \end{cases}$$

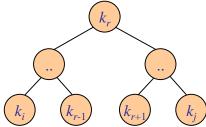
```
OBST(p, q, n)
1 for i \leftarrow 1 to n+1
    e[i, i-1] \leftarrow q_{i-1}
   w[i, i-1] \leftarrow q_{i-1}
4 for l ← 1 to n // 求 l 个元素的Opti-BST
        for i \leftarrow 1 to n-l+1
            j \leftarrow i+l-1
       e[i,j] \leftarrow \infty
       w[i,j] \leftarrow w[i,j-1] + p_i + q_i
       for r \leftarrow i to j
10
                   t \leftarrow e[i, r-1] + e[r+1, j] + w[i, j]
                   if t < e[i, j]
                          e[i,j] \leftarrow t
                          root[i, j] \leftarrow r
14 return e and root
```

$$w(i, j) = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w(i, j-1) + p_j + q_j$$

Innermost for loop, in lines 9–13, tries each candidate index r to determine which key  $k_r$  to use as the root of an OBST containing keys  $k_i$ , ...,  $k_j$ .

对包含  $k_i$ , ...,  $k_j$  的最优 BST, 遍历每一个  $k_r$  作 为树根, ...





#### Step 3: Computing the expected search cost (最优值)

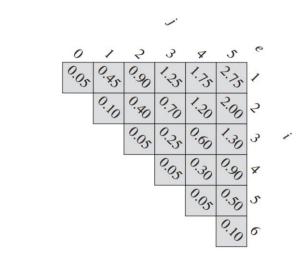
$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}$$

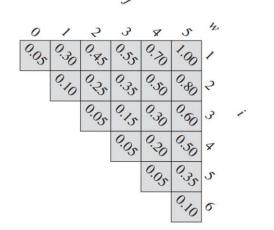
$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$

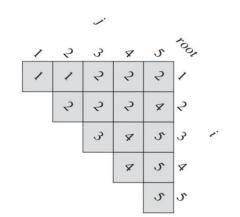
$$(15.20)$$

$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$
 (15.20)

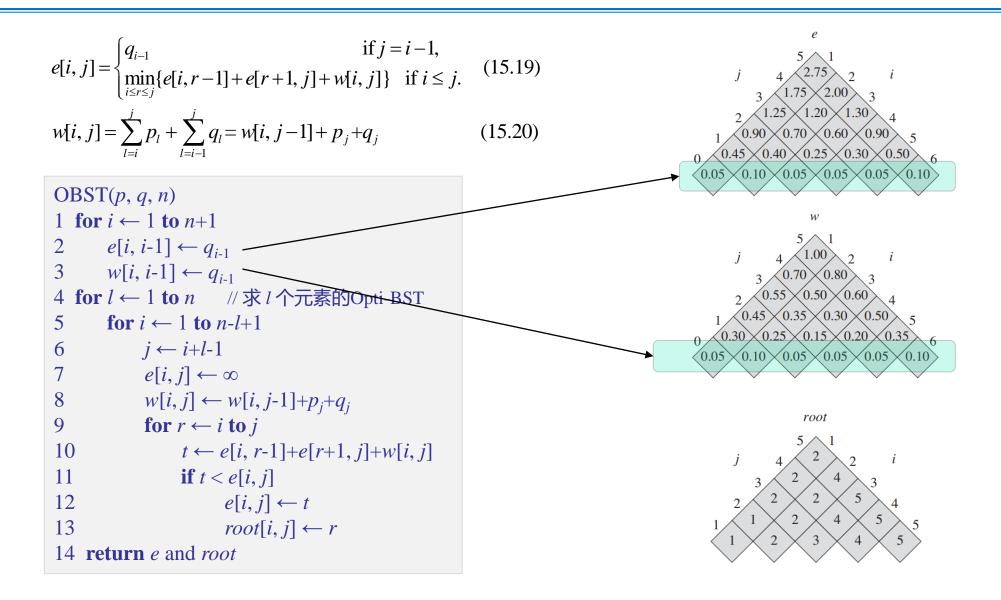
```
OBST(p, q, n)
1 for i \leftarrow 1 to n+1
        e[i, i-1] \leftarrow q_{i-1}
        w[i, i-1] \leftarrow q_{i-1}
4 for l ← 1 to n // 求 l 个元素的Opti-BST
        for i \leftarrow 1 to n-l+1
        j \leftarrow i+l-1
          e[i,j] \leftarrow \infty
         w[i,j] \leftarrow w[i,j-1] + p_i + q_i
             for r \leftarrow i to j
                   t \leftarrow e[i, r-1] + e[r+1, j] + w[i, j]
                   if t < e[i, j]
                          e[i,j] \leftarrow t
                         root[i, j] \leftarrow r
14 return e and root
```







#### Step 3: Computing the expected search cost (最优值)



#### Step 3: Computing the expected search cost (最优值)

$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}$$

$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$

$$(15.20)$$

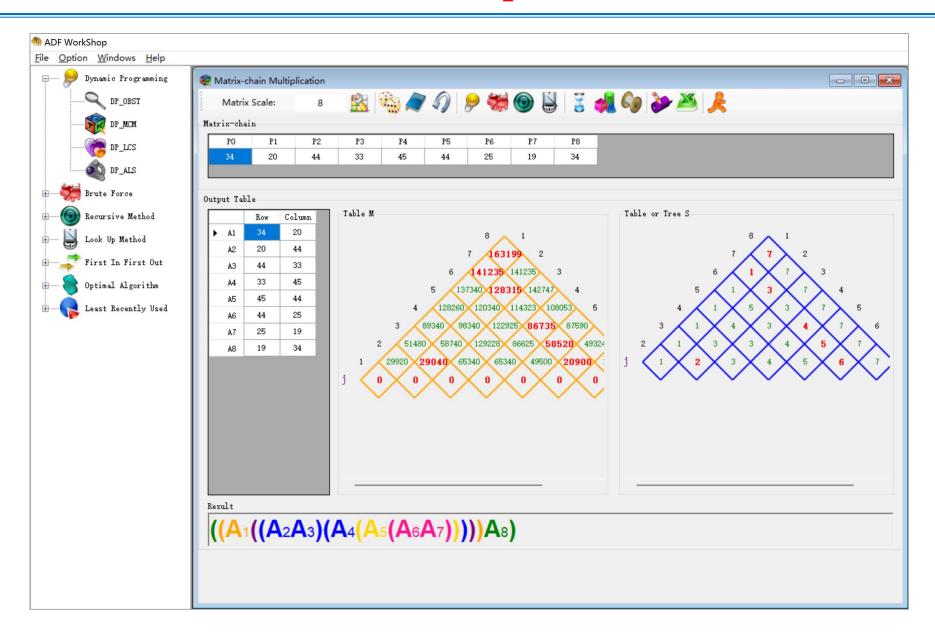
$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$
 (15.20)

```
OBST(p, q, n)
1 for i \leftarrow 1 to n+1
       e[i, i-1] \leftarrow q_{i-1}
   w[i, i-1] \leftarrow q_{i-1}
4 for l ← 1 to n // 求 l 个元素的Opti-BST
       for i \leftarrow 1 to n-l+1
    j \leftarrow i+l-1
      e[i,j] \leftarrow \infty
   w[i,j] \leftarrow w[i,j-1] + p_i + q_i
       for r \leftarrow i to j
10
     t \leftarrow e[i, r-1] + e[r+1, j] + w[i, j]
11 if t < e[i, j]
                      e[i,j] \leftarrow t
                     root[i, j] \leftarrow r
14 return e and root
```

## **Running times?**

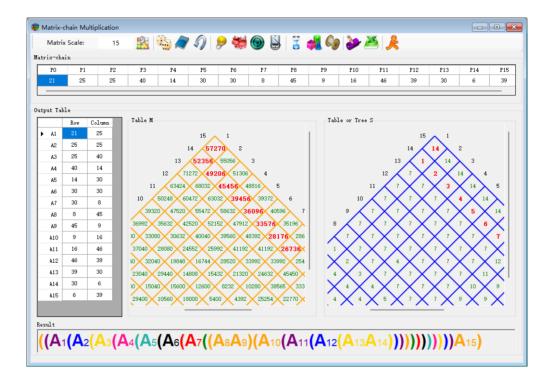
### Exercises

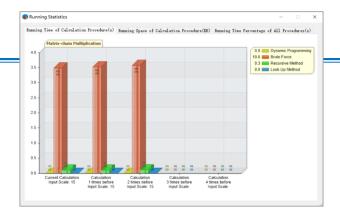
## ADF WorkShop (video-demo)

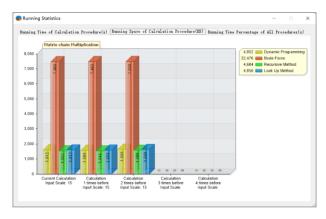


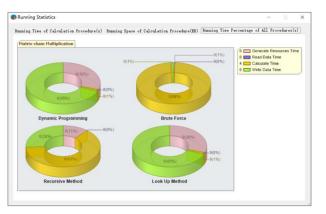
## **Exercises**

# ADF\_WorkShop



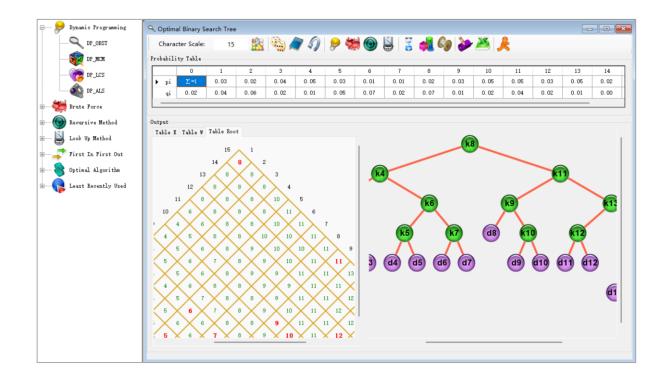


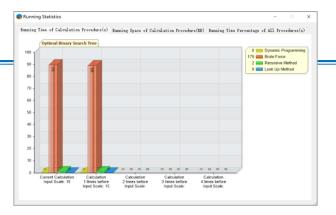


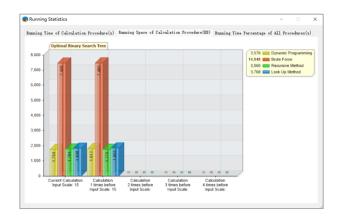


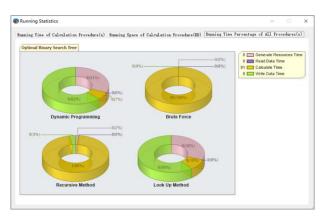
## **Exercises**

# ADF\_WorkShop









$$(A_{i}(A_{i+1}...)(...)A_{k})(A_{k+1}...A_{j-1}A_{j}))$$

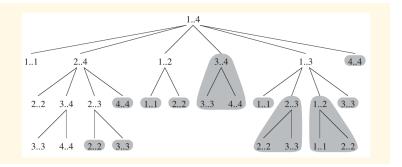
$$m[i,j] = \begin{cases} 0 & \text{if } i = j, \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_{k}p_{j}\}, & \text{if } i < j. \end{cases}$$

$$(15.12)$$

# of subproblems: one problem for each choice of i and j satisfying  $1 \le i \le j \le n$ ? (所有子问题总数为?)

$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{ m[i,k] + m[k+1,j] + p_{i-1}p_k p_j \}, & \text{if } i < j. \end{cases}$$
(15.12)



```
RE-MCM(p, i, j)

1 if i == j

2 return 0

3 m[i, j] \leftarrow \infty

4 for k \leftarrow i to j-1

5 q \leftarrow \text{RE-MCM}(p, i, k) + \text{RE-MCM}(p, k+1, j) + p_{i-1}p_kp_j

6 if q < m[i, j]

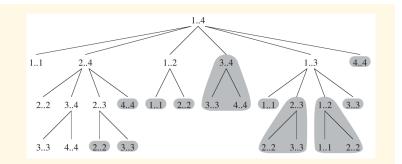
7 m[i, j] \leftarrow q

8 return m[i, j]
```

Running time?

( 
$$(A_i(A_{i+1}...)(...)A_k)(A_{k+1}...A_{j-1}A_j)$$
 )

$$m[i,j] = \begin{cases} 0 &, & \text{if } i = j, \\ \min_{i \le k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_k p_j\}, & \text{if } i < j. \end{cases}$$
(15.12)



#### **Dynamic Programming:**

top-down with memoization?

采用自顶向下递归的方式进行计

算(带备忘录),算法如何写?

$$e[i,j] = \begin{cases} q_{i-1} & \text{if } j = i-1, \\ \min_{i \le r \le j} \{e[i,r-1] + e[r+1,j] + w[i,j]\} & \text{if } i \le j. \end{cases}$$
(15.19)

$$w[i,j] = \sum_{l=i}^{j} p_l + \sum_{l=i-1}^{j} q_l = w[i,j-1] + p_j + q_j$$
 (15.20)

对OBST问题,基于式(15.19)和(15.20),

直接的递归算法如何写?

带备忘录(Memoization)的递归算法如何写?

各自的计算时间(Running time)是什么?



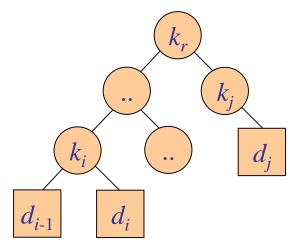
n	0	1	2	3	
$p_n$		:	:	:	:
$q_n$					

$$\sum_{i=1}^{n} p_i + \sum_{i=0}^{n} q_i = 1$$

E[search cost in *T*]

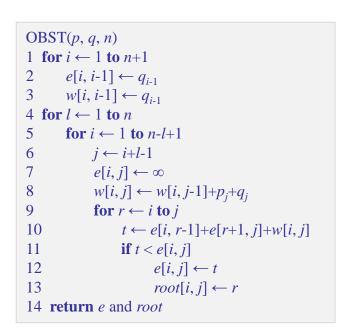
$$= \sum_{i=1}^{n} (\operatorname{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\operatorname{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

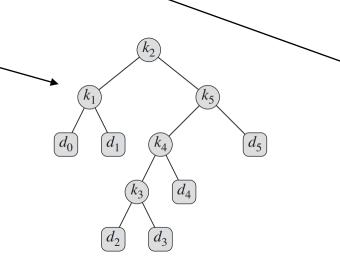
$$= 1 + \sum_{i=1}^{n} \operatorname{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \operatorname{depth}_{T}(d_{i}) \cdot q_{i}$$

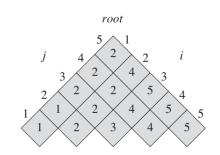


15.5-1 根据所求出的root,写成一个构造OBST 结构的伪代码CONSTRUCT-OBST(root),以文本形式输出这种结构。

## 对应的OBST为







k2 is the root
k1 is the left child of k2
d0 is the left child of k1
d1 is the right child of k1
k5 is the right child of k2
k4 is the left child of k5
k3 is the left child of k4
d2 is the left child of k3
d3 is the right child of k3
d4 is the right child of k4
d5 is the right child of k5

$$A_{1} A_{2} A_{3} A_{4} A_{5} : (A_{1} A_{2} A_{3}) (A_{4} A_{5})? (A_{1} A_{2}) (A_{3} A_{4} A_{5})? \dots \dots$$

$$A_{1} (A_{2} A_{3})? (A_{1} A_{2}) A_{3}?$$

$$((A_{i} (A_{i+1} \dots) (\dots) \dots A_{k}) (A_{k+1} \dots A_{j-1} A_{j}))$$

- Brute force: exhaustively checking all possible parenthesizations.
   P(n): the # of alternative parenthesizations of n matrices.
   矩阵连乘时,暴力穷举所有可能的加括号方式
   令P(n)表示n个矩阵连乘时所有可能的全括号方式的个数
- What is the solution of P(n)?

## Project

根据一本专业书籍(如《算法导论》),建设一个翻译软件中计算机类词库(字典)的OBST。说明:只考虑英语单词作为关键字。

#### 求解思路:

- 1. 统计书籍里有多少个单词 M
- 2. 按字母序, 第 i  $(1 \le i \le n)$  个单词在书中出现了  $k_i$  次  $(k_1 + k_2 + ... + k_n = M)$  , 其词频  $p_i = k_i / M$
- 3. 根据词频表,构建OBST

思考:对比一般的平衡二叉搜索树 BBST (用中间点作为树根), 比较 BBST 与 OBST 的搜索代价。