



### Algorithms Theory

14 – Dynamic Programming (3)

Optimal binary search trees

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#### Method of dynamic programming



**Recursive approach:** Solve a problem by solving several smaller analogous subproblems of the same type. Then combine these solutions to generate a solution to the original problem.

**Drawback:** Repeated computation of solutions

**Dynamic-programming method:** Once a subproblem has been solved, store its solution in a table so that it can be retrieved later by simple table lookup.

#### Optimal substructure

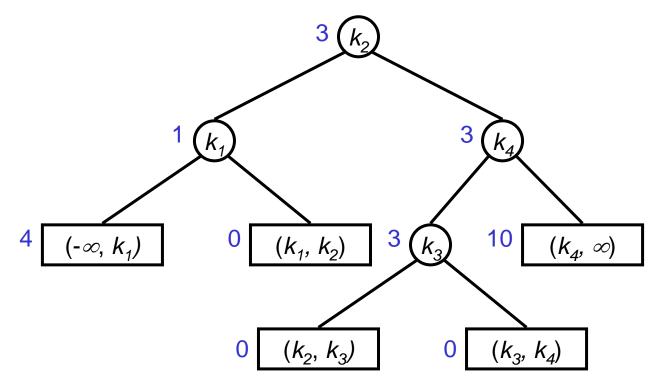


Dynamic programming is typically applied to *optimization problems*.

An optimal solution to the original problem contains *optimal* solutions to smaller subproblems.



$$(-\infty, k_1) k_1 (k_1, k_2) k_2 (k_2, k_3) k_3 (k_3, k_4) k_4 (k_4, \infty)$$
4 1 0 3 0 3 0 3 10



weighted path length:

$$3 \cdot 1 + 2 \cdot (1 + 3) + 3 \cdot 3 + 2 \cdot (4 + 10)$$



**Given**: set S of keys

$$S = \{k_1, \dots, k_n\}$$
  $-\infty = k_0 < k_1 < \dots < k_n < k_{n+1} = \infty$ 

 $a_i$ : (absolute) frequency of requests to key  $k_i$ 

 $b_{j}$ : (absolute) frequency of requests to  $x \in (k_{j}, k_{j+1})$ 

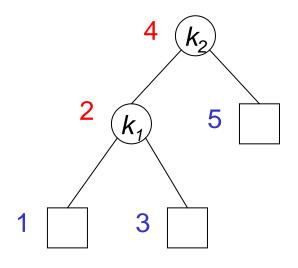
Weighted path length P(T) of a binary search tree T for S:

$$P(T) = \sum_{i=1}^{n} (depth(k_i) + 1)a_i + \sum_{j=0}^{n} depth((k_j, k_{j+1}))b_j$$

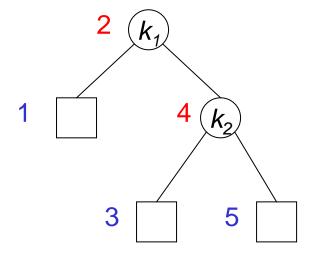
**Goal**: Binary search tree with minimum weighted path length P for S.





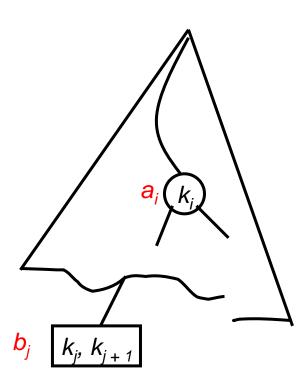


$$P(T_1) = 21$$



$$P(T_2) = 27$$

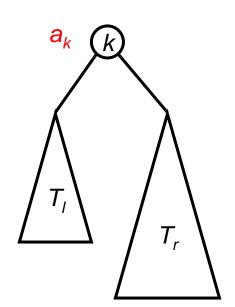




An optimal binary search tree is a binary search tree with minimum weighted path length.



T



$$P(T) = P(T_i) + W(T_i) + P(T_r) + W(T_r) + a_{root}$$
  
=  $P(T_i) + P(T_r) + W(T) mit$ 

W(T) := total weight of all nodes in T

If T is a tree with minimum weighted path length for S, then subtrees  $T_i$  and  $T_r$  are trees with minimum weighted path length for subsets of S.



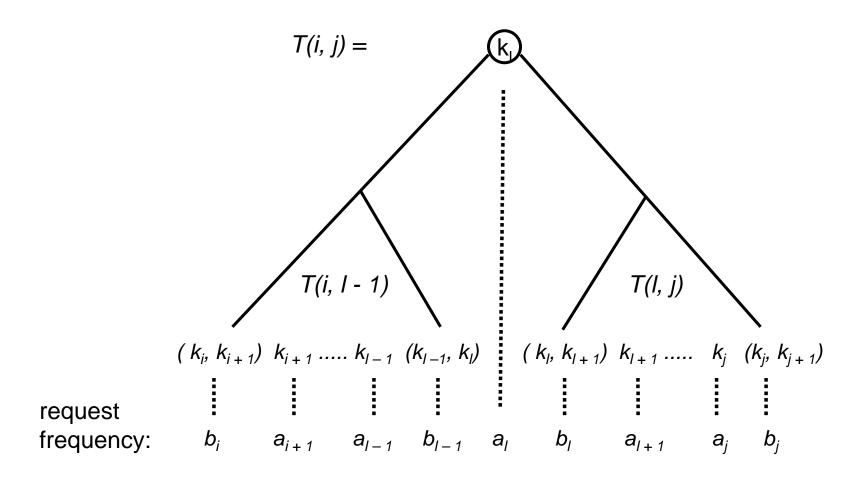
#### Let

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T(i, j): optimal binary search tree for (k_i, k_{i+1}) k_{i+1} \dots k_j (k_j, k_{j+1}),
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W(i, j): weight of T(i, j), i.e. W(i, j) =  $b_i + a_{i+1} + ... + a_j + b_j$ ,

P(i, j): weighted path length of T(i, j).







$$\begin{aligned} W(i, i) &= b_i &, \text{ for } 0 \leq i \leq n \\ W(i, j) &= W(i, j - 1) + a_j + b_j &, \text{ for } 0 \leq i < j \leq n \end{aligned}$$
 
$$P(i, i) &= 0 &, \text{ for } 0 \leq i \leq n$$
 
$$P(i, j) &= W(i, j) + \min \left\{ P(i, l - 1) + P(l, j) \right\}, \text{ for } 0 \leq i < j \leq n$$
 
$$(*)$$

r(i, j) = the index / for which the minimum is achieved in (\*)



#### **Base cases**

Case 1: 
$$h = j - i = 0$$

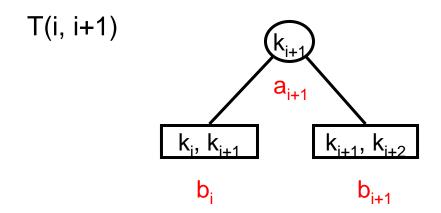
$$T(i, i) = (k_i, k_{i+1})$$

$$W(i, i) = b_i$$

$$P(i, i) = 0, r(i, i) \text{ not defined}$$



Case 2: h = j - i = 1



$$W(i, i+1) = b_i + a_{i+1} + b_{i+1} = W(i, i) + a_{i+1} + W(i+1, i+1)$$
  
 $P(i, i+1) = W(i, i+1)$   
 $r(i, i+1) = i+1$ 

# Computating the minimum weighted path length using dynamic programming



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Case 3: h = j - i > 1

for h = 2 to n do

for i = 0 to (n - h) do

\{ j = i + h;

determine (greatest) I, i < l \le j, s.t. P(i, l - 1) + P(l, j) is minimal P(i, j) = P(i, l - 1) + P(l, j) + W(i, j);

r(i, j) = l;

}
```



#### **Define:**

$$P(i, j) := \min$$
 weighted path length for  $b_i a_{i+1} b_{i+1} \dots a_j b_j$   $W(i, j) := \sup$  of

#### Then:

$$W(i,j) = \begin{cases} b_i & \text{if } i = j \\ W(i,j-1) + a_j + W(j,j) & \text{otherwise} \end{cases}$$

$$P(i,j) = \begin{cases} 0 & \text{if } i = j \\ W(i,j) + \min_{i < l \le j} \{P(i,l-1) + P(l,j)\} & \text{otherwise} \end{cases}$$

→ Computing the solution P(0,n) takes  $O(n^3)$  time. and requires  $O(n^2)$  space.

#### **Theorem**

An optimal binary search tree for n keys and n + 1 intervals with known request frequencies can be constructed in  $O(n^3)$  time.