

**BITS, PILANI – K. K. BIRLA GOA CAMPUS**

# **Design & Analysis of Algorithms**

## **(CS F364)**

**Lecture No. 9**



# All-Pairs Shortest Path

$G = (V, E)$  be a graph

G has **no negative weight cycles**

Vertices are labeled 1 to n

If  $(i, j)$  is an edge its weight is denoted by  $w_{ij}$

**Optimal Substructure Property**

**Easy to Prove**

# Recursive formulation - 1

**Let**  $l_{ij}^{(m)}$  be the minimum weight of any path from vertex  $i$  to vertex  $j$  that contains atmost  $m$  edges.

Then,

$$l_{ij}^{(0)} = \begin{cases} 0, & \text{if } i = j \\ \infty, & \text{if } i \neq j \end{cases}$$

$$l_{ij}^{(m)} = \min(l_{ij}^{(m-1)}, \min_{1 \leq k \leq n} \{l_{ik}^{(m-1)} + w_{kj}\})$$

**minimum** over all **predecessors**  $k$  of  $j$

# Recursive formulation - 2

Let  $d_{ij}^{(k)}$  be the weight of the shortest path from vertex  $i$  to vertex  $j$ ,

for which all the intermediate vertices are in the set  $\{1, 2, \dots, k\}$

Then

$$d_{ij}^{(k)} = w_{ij} \text{ if } k = 0$$

$$d_{ij}^{(k)} = \min (d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}) \text{ if } k \geq 1$$

We have tacitly used the fact that an optimal path through  $k$  does not visit  $k$  twice.

# Optimal Binary Search Tree

We start with a problem regarding binary search trees in an environment in which the probabilities of accessing elements and gaps between elements is known.

## **Goal:**

We want to find the binary search tree that minimizes the expected number of nodes probed on a search.

# Optimal Binary Search Tree

## Formally

We have  $n$  keys (represented as  $k_1, k_2, \dots, k_n$ ) in sorted order (so that  $k_1 < k_2 < \dots < k_n$ ),

and we wish to build a binary search tree from these keys.

For each  $k_i$ , we have a probability  $p_i$  that a search will be for  $k_i$ .

In contrast of, some searches may be for values not in  $k_i$ , and so we also have  $n+1$  “dummy keys”  $d_0, d_1, \dots, d_n$

In particular,  $d_0$  represents all values less than  $k_1$ , and

$d_n$  represents all values greater than  $k_n$ , and for  $i = 1, 2, \dots, n-1$ , the dummy key  $d_i$  represents all values between  $k_i$  and  $k_{i+1}$ .

For each  $d_i$ , we have a probability  $q_i$  that a search will be for  $d_i$ .

**The dummy keys are leaves (external nodes), and the data keys mean internal nodes.**

# Optimal Binary Search Tree

- Every search is either successful or unsuccessful and so we have

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

Because we have probabilities of searches for each key and each dummy key, we can determine the expected cost of a search in a given binary search tree ***T***.

Let us assume that the actual cost of a search is the number of nodes examined,

i.e., the depth of the node found by the search in ***T***, ***plus 1***.

# Optimal Binary Search Tree

Then the expected cost of a search in  $T$  is

$E[\text{search cost in } T]$

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

$$= 1 + \sum_{i=1}^n \text{depth}_T(k_i)p_i + \sum_{i=0}^n \text{depth}_T(d_i)q_i \quad (1)$$

Where  $\text{depth}_T$  denotes a node's depth in the tree  $T$



Example

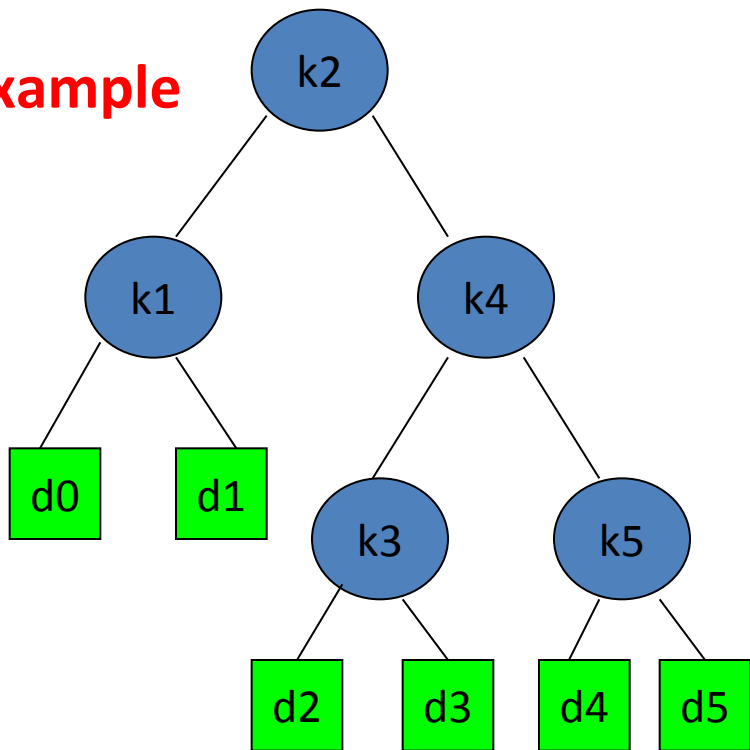


Figure (a)

i	0	1	2	3	4	5
pi		0.15	0.10	0.05	0.10	0.20
qi	0.05	0.10	0.05	0.05	0.05	0.10

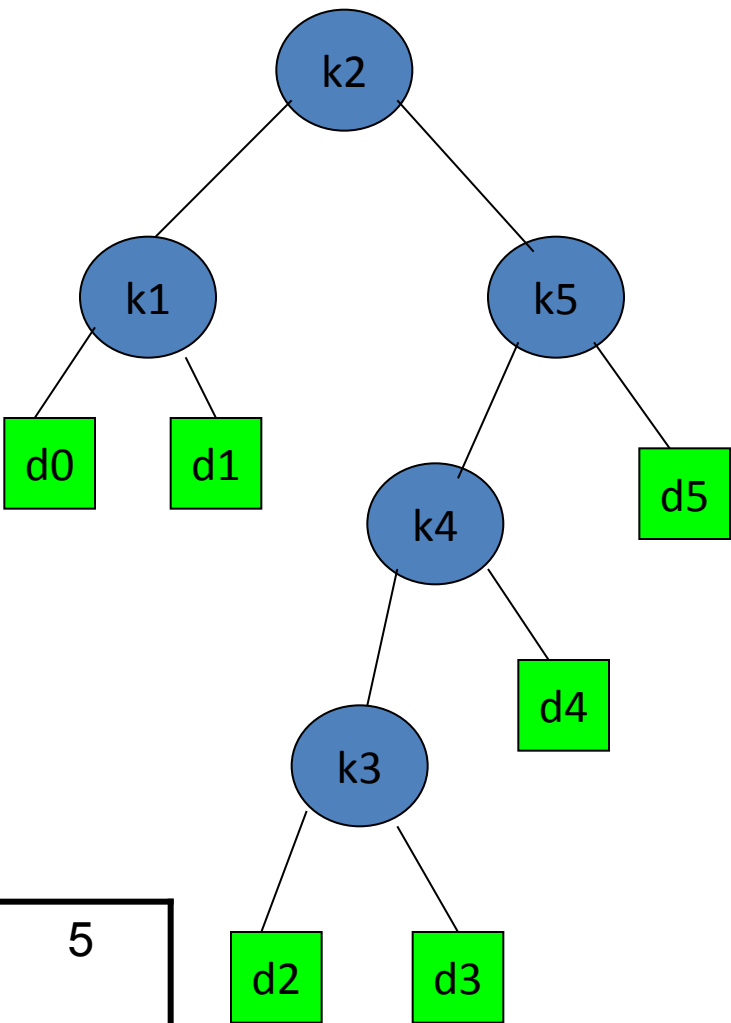


Figure (b)

# Optimal Binary Search Tree

## Observe:

Figure (a) costs 2.80 ,on another,  
the Figure (b) costs 2.75,  
and that tree is really optimal.

We can see the height of (b) is more than (a)

**So optimal BST is not necessarily a tree whose height is smallest.**

# Optimal Binary Search Tree

## Brute Force Approach:

Exhaustive checking of all possibilities.

## Question

What is the number of binary search trees on  $n$  keys?

## Answer

$$t(n) = \sum_{i=1}^n t(i-1) t(n-i).$$

The number of BST is atleast  $\Omega(2^n)$

# Dynamic Programming Solution

## Step1 : Optimal Substructure Property

### Exercise

## Step2 : Recursive Formulation

We pick our subproblem domain as finding an Optimal BST containing the *keys*  $k_i, \dots, k_j$ , where  $i \geq 1, j \leq n$ , and  $j \geq i-1$ . (It is when  $j = i-1$  that there are no actual keys; we have just the dummy key  $d_{i-1}$ .)

Let us define  $e[i, j]$  as the expected cost of searching an Optimal BST containing the keys  $k_i, \dots, k_j$

Ultimately, we wish to compute  $e[1, n]$ .

# Optimal Binary Search Tree

When  $j = i-1$

Then we have just the dummy key  $d_{i-1}$

The expected search cost is  $e[i, i-1] = q_{i-1}$

When  $j \geq i$ , we need to select a root  $k_r$  from among  $k_i, \dots, k_j$

and then make an Optimal BST with keys  $k_i, \dots, k_{r-1}$  as its left subtree

and an Optimal BST with keys  $k_{r+1}, \dots, k_j$  as its right subtree.

# Optimal Binary Search Tree

What happens to the expected search cost of a subtree when it becomes a subtree of a node?

The depth of each node in the subtree increases by 1.

So, by equation (1) the expected search cost of this subtree increases by the sum of all the probabilities in the subtree.

For a subtree with keys  $k_i, \dots, k_j$  let us denote this sum of probabilities as

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

# Optimal Binary Search Tree

Thus, if  $k_r$  is the root of an optimal subtree containing keys  $k_i, \dots, k_j$ , we have

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

Noting that  $w(i, j) = w(i, r-1) + p_r + w(r+1, j)$

We rewrite  $e[i, j]$  as

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

The recursive equation as above assumes that we know which node  $k_r$  to use as the root.

We choose the root that gives the lowest expected search cost

# Optimal Binary Search Tree

Final recursive formulation:

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

The  **$e[i, j]$**  values give the expected search costs in Optimal BST.

To help us keep track of the structure of Optimal BST, we define  **$root[i, j]$** , for  **$1 \leq i \leq j \leq n$** , to be the index  **$r$**  for which  **$k_r$**  is

the root of an Optimal BST containing keys  **$k_i, \dots, k_j$**