

# 11 A: Algorithm Design Techniques

CS1102S: Data Structures and Algorithms

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- 1 NP-Complete Problems
- 2 Greedy Algorithms
- 3 Divide and Conquer
- 4 Dynamic Programming

## Example: Hamiltonian Cycles

Given Input

An undirected connected graph

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A path that starts and ends in the same vertex and contains all other vertices exactly once.

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### Given Input

An undirected connected graph

### Desired Output

A path that starts and ends in the same vertex and contains all other vertices exactly once.

### No efficient algorithm known

We do not know if there is a  $k$  such that the problem can be solved in  $O(N^k)$ .

# Our Last Question

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If we do not have a polynomial algorithm, can we always prove that there is none?

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### Answer

No: we cannot (at this moment) prove that there is no polynomial algorithm for the Hamiltonian cycle problem

# Verifying Solutions

## Example: Hamiltonian cycle problem

If we have a candidate of a solution to the problem, we can easily check that it is correct.



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If we have a candidate of a solution to the problem, we can easily check that it is correct.

Simply check that the last vertex in the cycle is that same as the first, and that every vertex of the graph is contained in the cycle.

# Verifying Solutions

## Example: Hamiltonian cycle problem

If we have a candidate of a solution to the problem, we can easily check that it is correct.

Simply check that the last vertex in the cycle is that same as the first, and that every vertex of the graph is contained in the cycle.

## Definition

We call the class of problems for which solution candidates can be checked in polynomial time *NP*.

# NP-Complete Problems

## Other problems in NP

- Boolean satisfiability problem (SAT)
- Graph coloring problem
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## Reducibility

All these problems can be transformed into each other in polynomial time! Thus if we can solve one, we can solve all.

# NP-Complete Problems

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- Boolean satisfiability problem (SAT)
- Graph coloring problem
- Clique problem

## Reducibility

All these problems can be transformed into each other in polynomial time! Thus if we can solve one, we can solve all.

## NP-Complete Problems

The class of problems that can be transformed into the Hamiltonian path problem in polynomial time is called the class of *NP-complete problems*.

- 1 NP-Complete Problems
- 2 **Greedy Algorithms**
  - Scheduling
  - Huffman Codes
- 3 Divide and Conquer
- 4 Dynamic Programming

# Nonpreemptive Scheduling

## Input

A set of jobs with a running time for each

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## Desired output

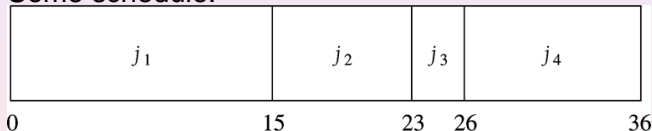
A sequence for the jobs to execute on on single machine, minimizing the average completion time



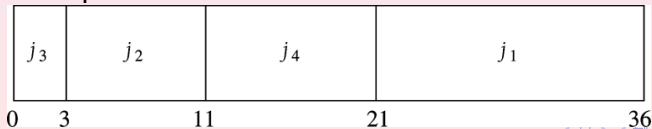
# Example

Job	Time
$j_1$	15
$j_2$	8
$j_3$	3
$j_4$	10

Some schedule:



The optimal schedule:

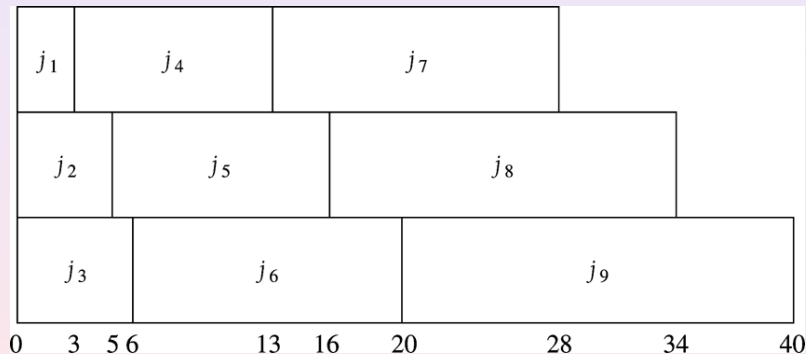


# The Multiprocessor Case

$N$  processors

Now we can run the jobs on  $N$  identical machines. What is a schedule that minimizes the average completion time?

# Example



## A “Slight” Variant

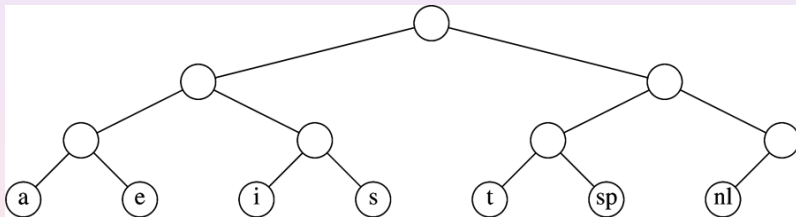
Minimizing *final* completion time

If we want to minimize the *final* completion time (completion time of the last task), the problem becomes NP-complete!

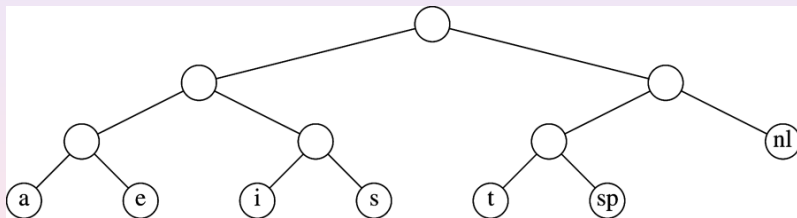
# Standard Coding Scheme

Character	Code	Frequency	Total Bits
<i>a</i>	000	10	30
<i>e</i>	001	15	45
<i>i</i>	010	12	36
<i>s</i>	011	3	9
<i>t</i>	100	4	12
<i>space</i>	101	13	39
<i>newline</i>	110	1	3
Total			174

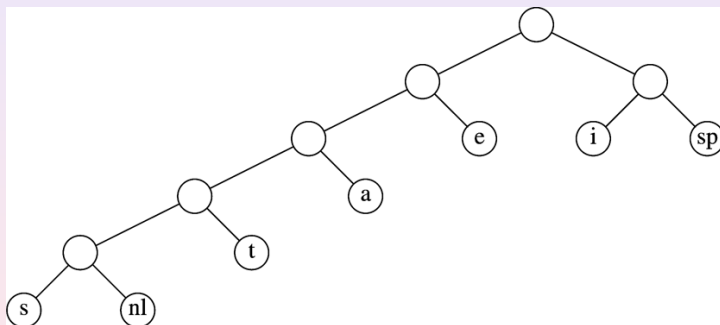
# Representation in a Tree



# A Slightly Better Representation



# Optimal Prefix Code in Tree Form





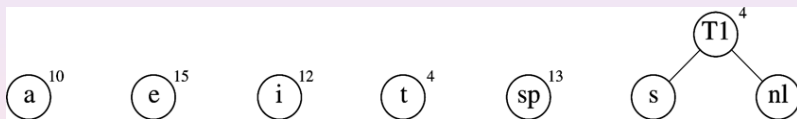
## Optimal Prefix Code in Table Form

Character	Code	Frequency	Total Bits
<i>a</i>	001	10	30
<i>e</i>	01	15	30
<i>i</i>	10	12	24
<i>s</i>	00000	3	15
<i>t</i>	0001	4	16
<i>space</i>	11	13	26
<i>newline</i>	00001	1	5
Total			146

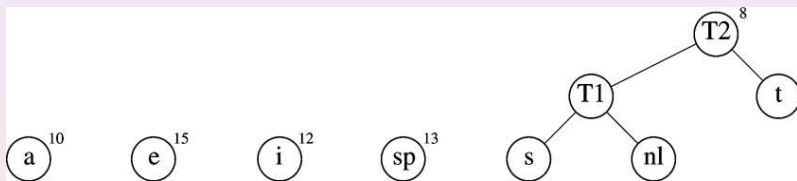
# Huffman's Algorithm: An Example



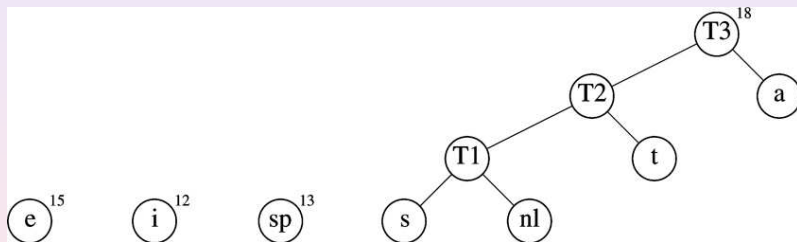
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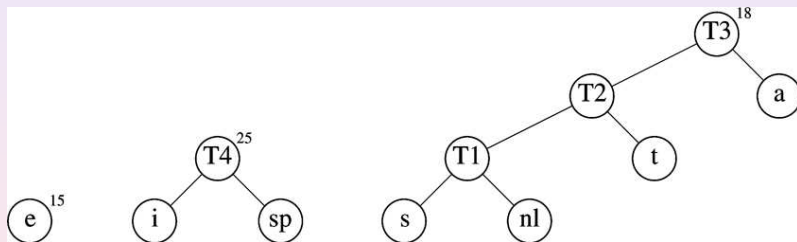
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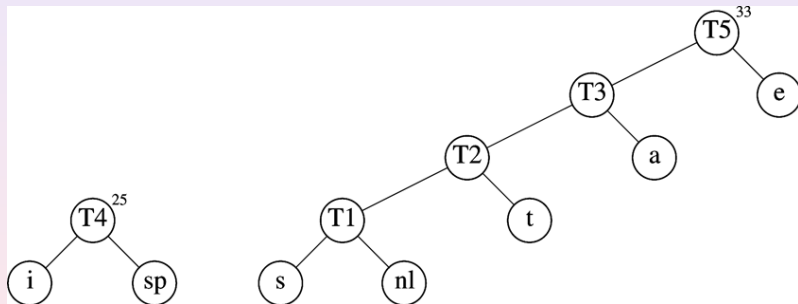
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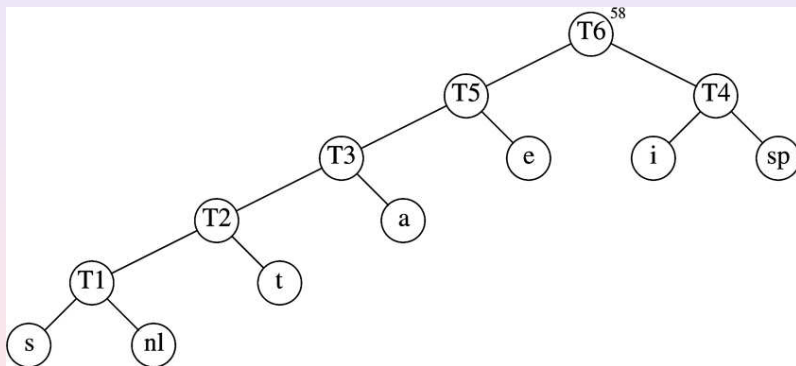
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# Huffman's Algorithm: An Example





- 1 NP-Complete Problems
- 2 Greedy Algorithms
- 3 Divide and Conquer**
  - Sorting
  - Running Time
  - Closest-Points Problem
- 4 Dynamic Programming

# Sorting using Divide and Conquer

## Idea of Mergesort

Split array in two (trivial); sort the two (recursively); merge (linear)

# Sorting using Divide and Conquer

## Idea of Mergesort

Split array in two (trivial); sort the two (recursively); merge (linear)

## Idea of Quicksort

Split array in two, using pivot (linear); sort the two (recursively); merge (trivial)

# Running Time of Divide and Conquer Algorithms

Merge Sort:  $T(N) = 2T(N/2) + O(N)$

$O(N \log N)$

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Merge Sort:  $T(N) = 2T(N/2) + O(N)$

$$O(N \log N)$$

Generalization:  $T(N) = aT(N/b) + \Theta(N^k)$

$$T(N) = O(N^{\log_b a}) \quad \text{if } a > b^k$$

$$T(N) = O(N^k \log N) \quad \text{if } a = b^k$$

$$T(N) = O(N^k) \quad \text{if } a < b^k$$

# Closest-Points Problem

## Input

Set of points in a plane

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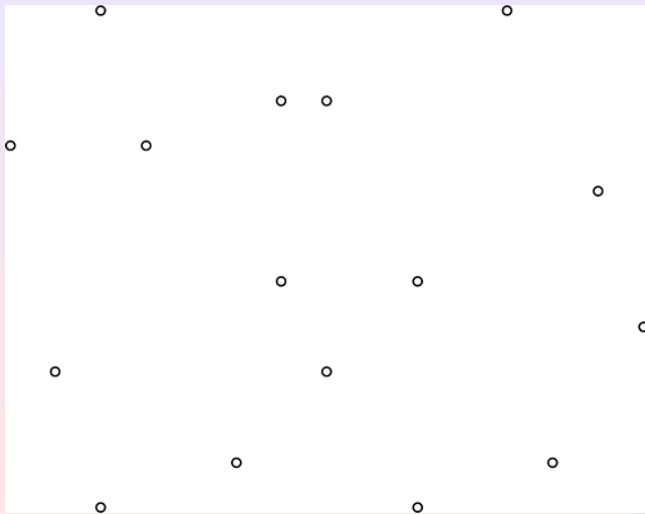
$$[(x_1 - x_2)^2 + (y_1 - y_2)^2]^{1/2}$$

## Required output

Find the closest pair of points



# Example



# Naive Algorithm

## Exhaustive search

Compute the distance between each two points and keep the smallest

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Compute the distance between each two points and keep the smallest

## Run time

There are  $N^2$  pairs to check, thus  $O(N^2)$

# Idea

## Preparation

Sort points by x coordinate;

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## Divide and Conquer

Split point set into two halves,  $P_L$  and  $P_R$ .  
Recursively find the smallest distance in each half.

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## Divide and Conquer

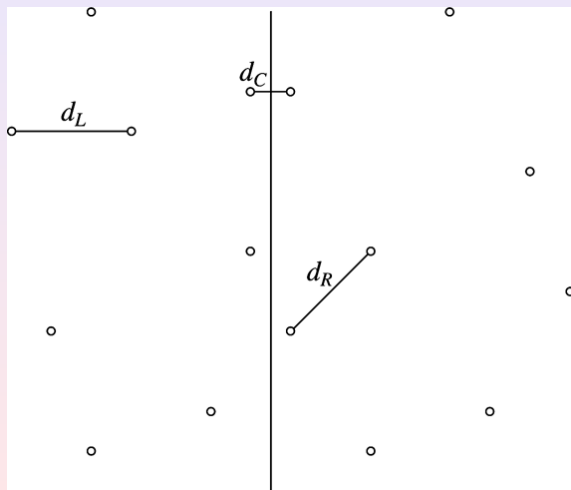
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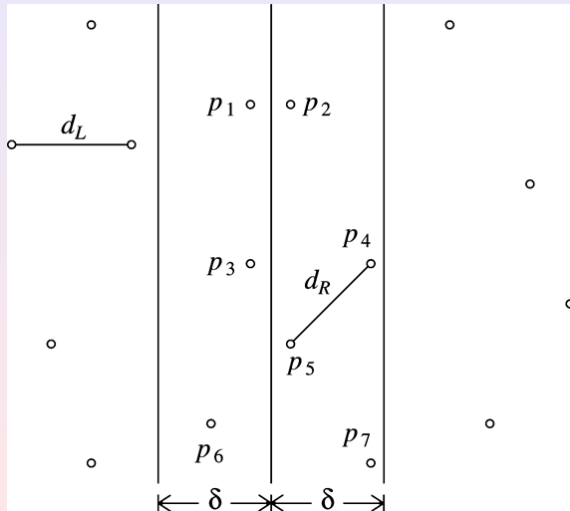
Find the smallest distance of pairs that *cross* the separation line.



# Partitioning with Shortest Distances Shown



# Two-lane Strip



# Brute Force Calculation of $\min(\delta, d_C)$

```
// Points are all in the strip  
  
for( i = 0; i < numPointsInStrip; i++ )  
    for( j = i + 1; j < numPointsInStrip; j++ )  
        if( dist( $p_i$ ,  $p_j$ ) <  $\delta$  )  
             $\delta = \text{dist}(p_i, p_j)$ ;
```

## Better Idea

Sort points by y coordinate

This allows a *scan* of the strip.

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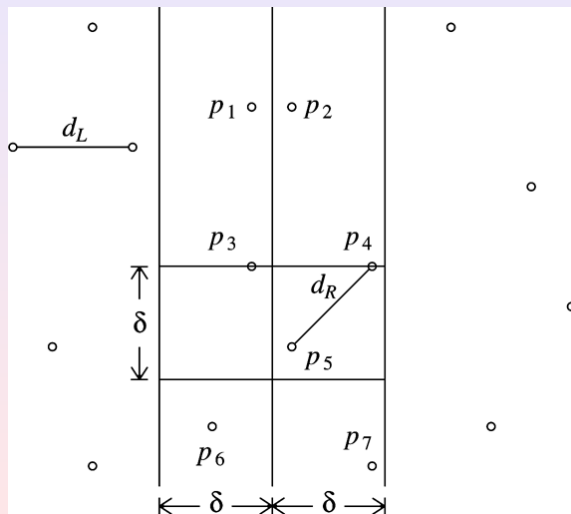
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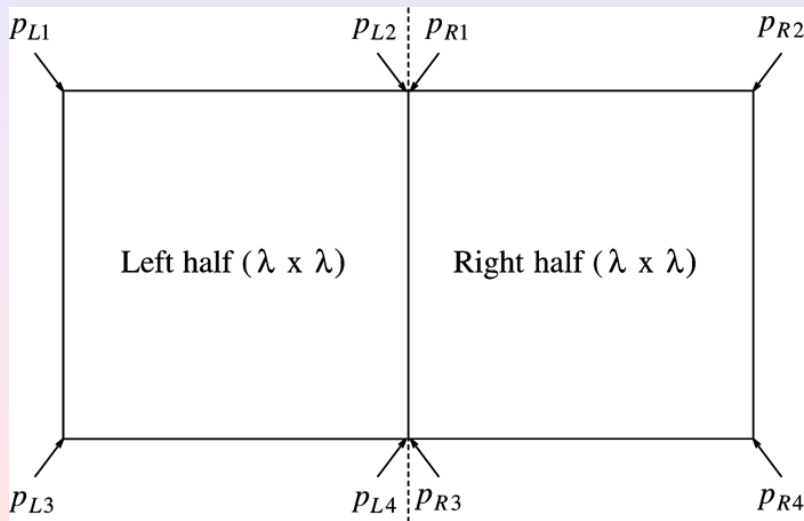
# Only $p_4$ and $p_5$ Need To Be Considered



Refined Calculation of  $\min(\delta, d_C)$ 

```
// Points are all in the strip and sorted by y-coordinate  
for( i = 0; i < numPointsInStrip; i++ )  
    for( j = i + 1; j < numPointsInStrip; j++ )  
        if(  $p_i$  and  $p_j$ 's y-coordinates differ by more than  $\delta$  )  
            break;           // Go to next  $p_i$ .  
        else  
            if(  $\text{dist}(p_i, p_j) < \delta$  )  
                 $\delta = \text{dist}(p_i, p_j)$ ;
```

# At Most Eight Points Fit in Rectangle



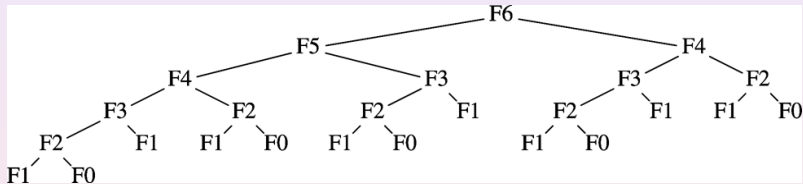


- 1 NP-Complete Problems
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  - Fibonacci Numbers
  - Optimal Binary Search Tree
  - All-pairs Shortest Path

# Inefficient Algorithm

```
public static int fib(int n) {  
    if (n <= 1)  
        return 1;  
    else  
        return fib(n - 1) + fib(n - 2);  
}
```

# Trace of Recursion



# Memoization

```
int[] fibs = new int[100];  
public static int fib(int n) {  
    if (fibs[n]!=0) return fibs[n];  
    if (n <= 1) return 1;  
    int new_fib = fib(n - 1) + fib(n - 2);  
    fibs[n] = new_fib;  
    return new_fib;  
}
```

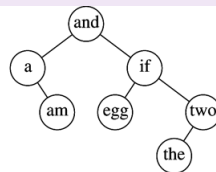
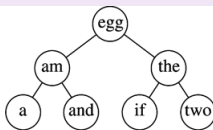
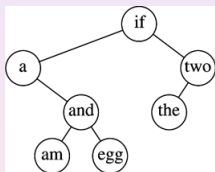
# A Simple Loop for Fibonacci Numbers

```
public static int fib(int n) {  
    if (n <= 1) return 1;  
    int last = 1, nextToLast = 1; answer = 1;  
    for (int i = 2; i <= n; i++) {  
        answer = last + nextToLast;  
        nextToLast = last;  
        last = answer;  
    }  
    return answer;  
}
```

# Sample Input

Word	Probability
a	0.22
am	0.18
and	0.20
egg	0.05
if	0.25
the	0.02
two	0.08

# Three Possible Binary Search Trees

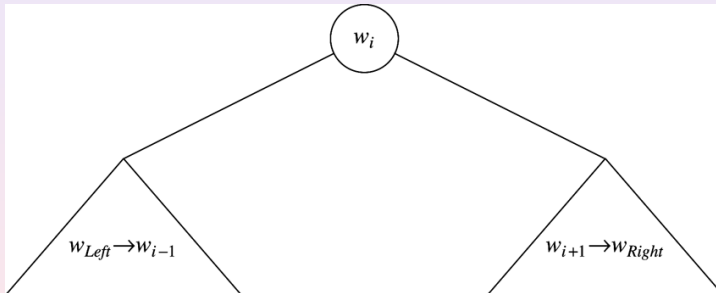


# Comparison of the Three Trees

Input		Tree #1		Tree #2		Tree #3	
Word $w_i$	Probability $p_i$	Access Cost Once	Access Cost Sequence	Access Cost Once	Access Cost Sequence	Access Cost Once	Access Cost Sequence
a	0.22	2	0.44	3	0.66	2	0.44
am	0.18	4	0.72	2	0.36	3	0.54
and	0.20	3	0.60	3	0.60	1	0.20
egg	0.05	4	0.20	1	0.05	3	0.15
if	0.25	1	0.25	3	0.75	2	0.50
the	0.02	3	0.06	2	0.04	4	0.08
two	0.08	2	0.16	3	0.24	3	0.24
Totals	1.00		2.43		2.70		2.15



# Structure of Optimal Binary Search Tree



# Idea

Proceed in order of growing tree size

For each range of words, compute optimal tree

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Memoization

For each range, store optimal tree for later retrieval

# Computation of Optimal Binary Search Tree

	Left=1	Left=2	Left=3	Left=4	Left=5	Left=6	Left=7
Iteration=1	a..a .22   a	am..am .18   am	and..and .20   and	egg..egg .05   egg	if..if .25   if	the..the .02   the	two..two .08   two
Iteration=2	a..am .58   a	am..and .56   and	and..egg .30   and	egg..if .35   if	if..the .29   if	the..two .12   two	
Iteration=3	a..and 1.02   am	am..egg .66   and	and..if .80   if	egg..the .39   if	if..two .47   if		
Iteration=4	a..egg 1.17   am	am..if 1.21   and	and..the .84   if	egg..two .57   if			
Iteration=5	a..if 1.83   and	am..the 1.27   and	and..two 1.02   if				
Iteration=6	a..the 1.89   and	am..two 1.53   and					
Iteration=7	a..two 2.15   and						

# Run Time

For each cell of table

Consider all possible roots

# Run Time

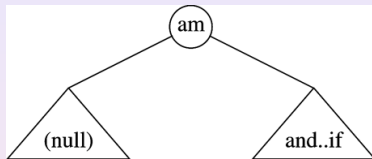
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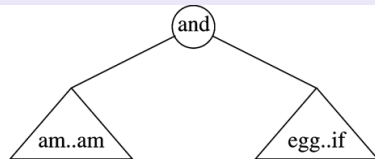
Overall runtime

$$O(N^3)$$

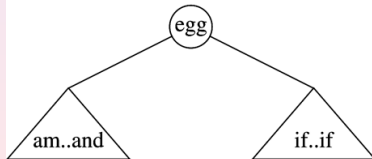
# Example



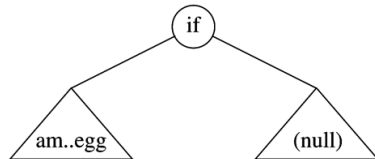
$$0 + 0.80 + 0.68 = 1.48$$



$$0.18 + 0.35 + 0.68 = 1.21$$



$$0.56 + 0.25 + 0.68 = 1.49$$



$$0.66 + 0 + 0.68 = 1.34$$