Chapter 3: Algorithm Strategies

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- Divide-and-conquer
- Backtracking
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Brute-force

Brute-force

- A straightforward approach to solving a problem, usually directly based on the problem statement and definition
- Systematically enumerates all possible candidates for the solution and checks whether each candidate satisfies the problem statement

Selection sort

Based on sequentially finding the smallest elements

Bubble Sort

Based on consecutive swapping adjacent pairs. This causes a slow migration of the smallest elements to the left of the array.

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Based on consecutive swapping adjacent pairs. This causes a slow migration of the smallest elements to the left of the array.

```
1: procedure BUBBLESORT (A[0...n-1])

2: for i \leftarrow 0 to n-2 do

3: for j \leftarrow 0 to n-2-i do

4: if A[j+1] < A[j] then

5: swap A[j+1] and A[j]

6: end if

7: end for

8: end procedure
```

Sequential Searching

```
1: procedure SequentialSearch(A[0 . . . n-1], K) \triangleright K is the search key
       i \leftarrow 0
 2:
       while i < n and A[i] \neq K do
       i \leftarrow i + 1
 4:
    end while
    if i < n then
          return i
       else
 8:
          return -1
 9:
       end if
10:
11: end procedure
```

Brute-Force String Matching: Searching for a pattern, P[0...m-1], in text, T[0...n-1]

```
1: procedure BruteForceStringMatch(T[0...n-1], P[0...m-1])
       for i \leftarrow 0 to n - m do
 2:
          i \leftarrow 0
 3:
          while j < m and P[j] == T[i+j] do
 4:
             j \leftarrow j + 1
 5:
          end while
 6:
          if j == m then
 7:
              return i
 8:
          end if
 9:
       end for
10:
       return -1
11:
12: end procedure
```

The Greedy method

The greedy method

- Builds up a solution piece by piece, always choosing the next piece that looks best at the moment
- The main idea is to make locally optimal choice in the hope that this choice will lead to a globally optimal solution
- Greedy algorithms do not always yield optimal solutions, but for many problems they do

Coding: Assigning binary codewords to (blocks of) source symbols

Huffman coding is a lossless data compression algorithm.

Idea:

- Assign variable-length codes to input characters, based on the frequencies of corresponding characters.
- Instead of using ASCII codes, store the more frequently occurring characters using fewer bits and less frequently occurring characters using more bits.

There are mainly two major parts in Huffman Coding

- 1. Build a **Huffman Tree** from input characters.
- 2. Traverse the Huffman Tree and assign codes to characters.

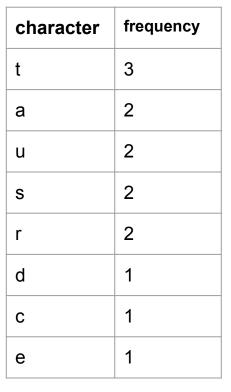
- Organize the entire character set into a row, ordered according to frequency from highest to lowest (or vice versa). Each character is now a node at the leaf level of a tree
- 2. Find two nodes with the smallest combined frequency weights and join them to form a third node, resulting in a simple two-level tree. The weight of the new node is the combined weights of the original two nodes.
- 3. Repeat step 2 until all of the nodes, on every level, are combined into a single tree.

Input character string: "datastructures"

We first build the frequency table

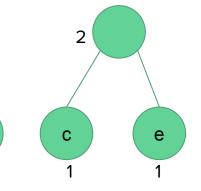
character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1

Build a Huffman tree:



t a u s r d c e 3 2 2 1 1 1 1

Build a Huffman tree:



character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1

t

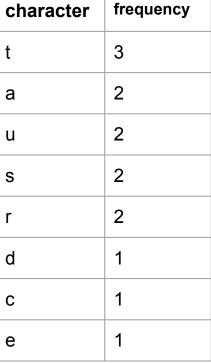
<u>u</u>

2

r

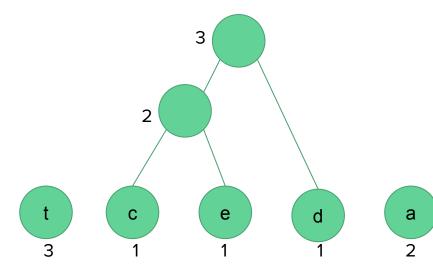
d

Build a Huffman tree:



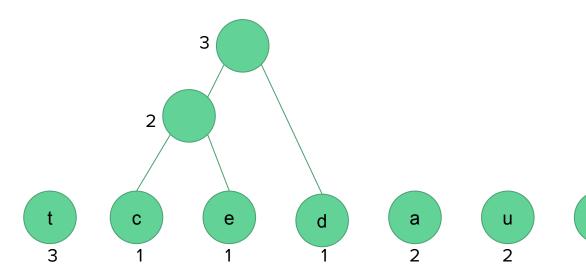
u

Build a Huffman tree:



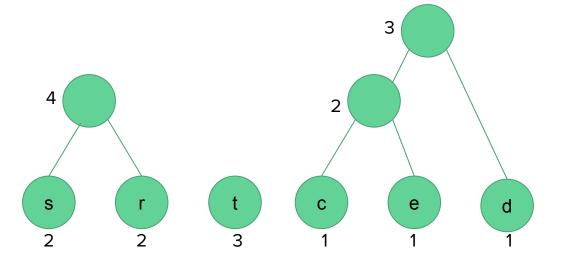
character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1

r 2



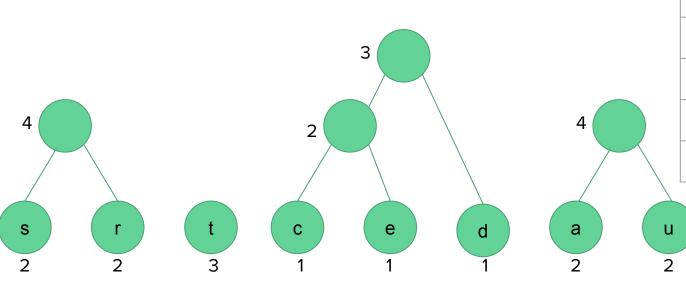
	character	frequency
	t	3
	а	2
	u	2
	S	2
	r	2
	d	1
	С	1
	е	1
7		

Build a Huffman tree:

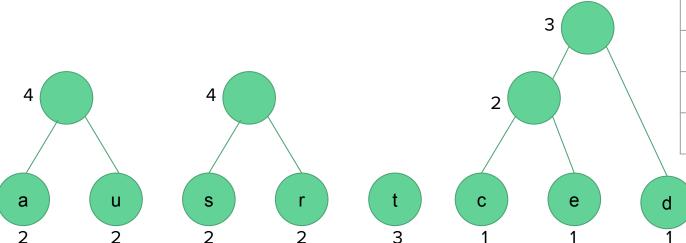


character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1

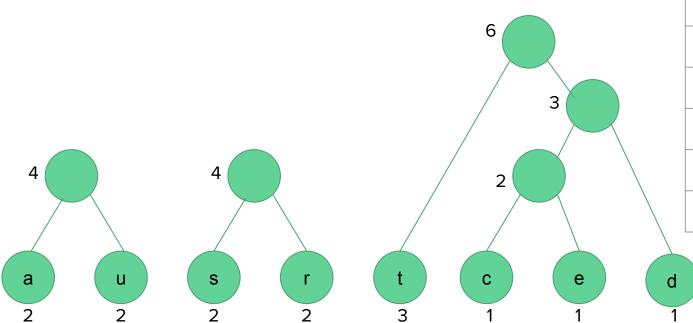
u 2



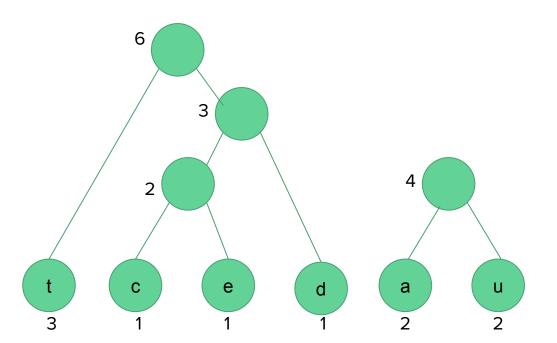
character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1



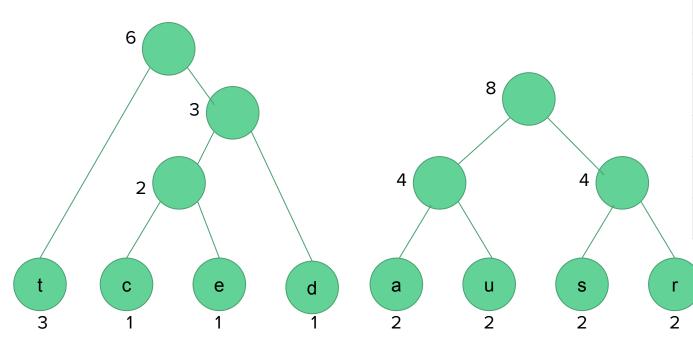
character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1



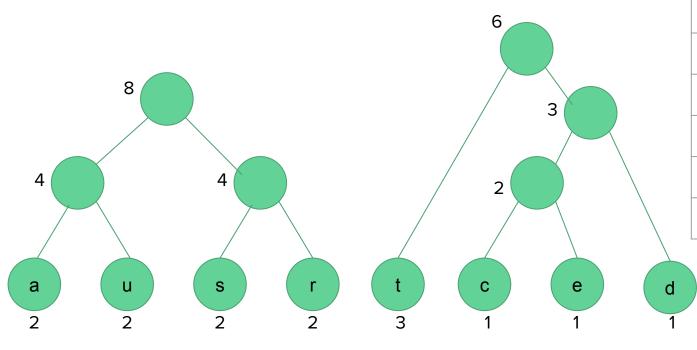
character	frequency
t	3
а	2
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r	2
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С	1
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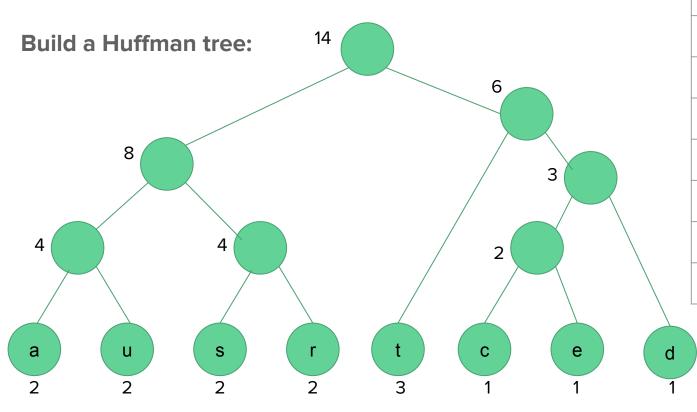
character	frequency
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а	2
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character	frequency
t	3
а	2
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character	frequency
t	3
а	2
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S	2
r	2
d	1
С	1
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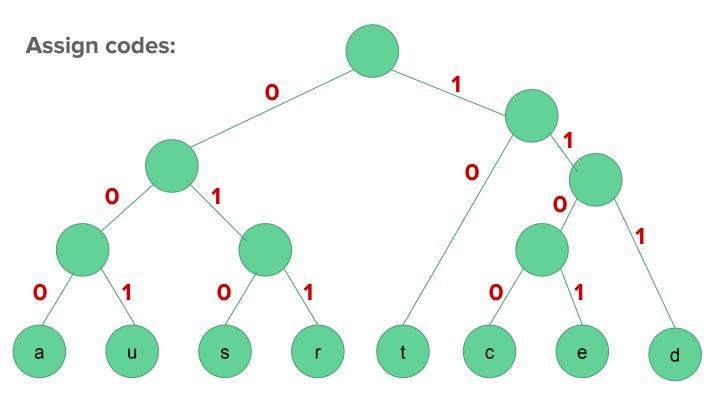


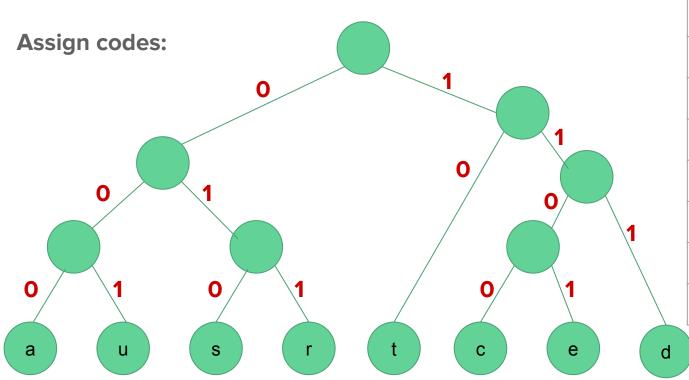
character	frequency
t	3
а	2
u	2
S	2
r	2
d	1
С	1
е	1

Now we **assign codes** to the tree by **placing a 0 on every left branch and a 1 on every right branch**

A traversal of the tree from root to leaf give the Huffman code for that particular leaf character

These codes are then used to encode the string





character	Huffman code
t	10
а	000
u	001
S	010
r	011
d	111
С	1100
е	1101

Thus "datastructures" turns into

If 8-bit ASCII code had been used instead of Huffman coding, "datastructures" would have been

character	Huffman code	ASCII code
t	10	01110100
а	000	01100001
u	001	01110101
S	010	01110011
r	011	01110010
d	111	01100100
С	1100	01100011
е	1101	01100101

Uncompression:

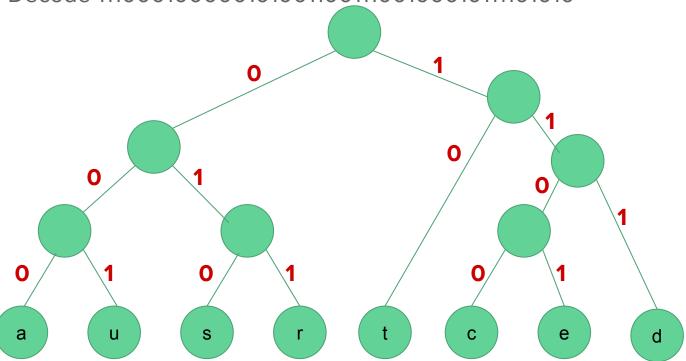
Read the file bit by bit

- 1. Start at the root of the tree
- 2. If a 0 is read, head left
- 3. If a 1 is read, head right
- 4. When a leaf is reached, decode that character and start over again at the root of the tree

Uncompression example:

Decode 111000100001010011001100100010111101010 using the previous Huffman tree

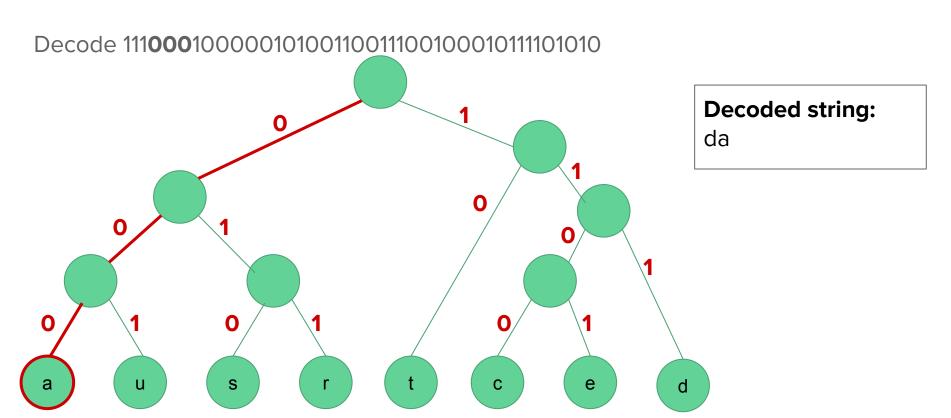
Uncompression example

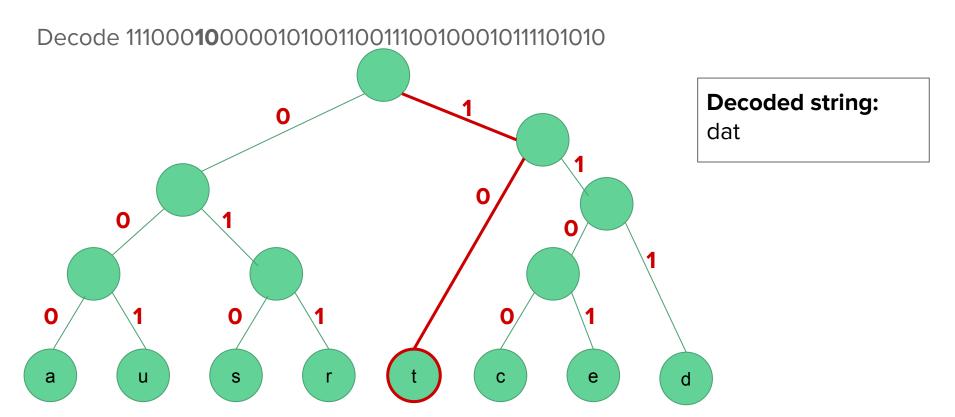


Uncompression example

Decode **111**00010000010100110011100100010111101010 **Decoded string:** d 0 0

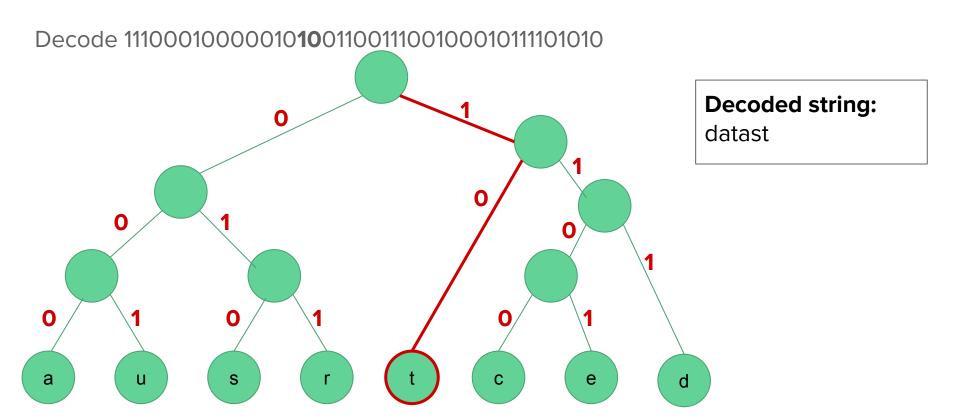
Uncompression example



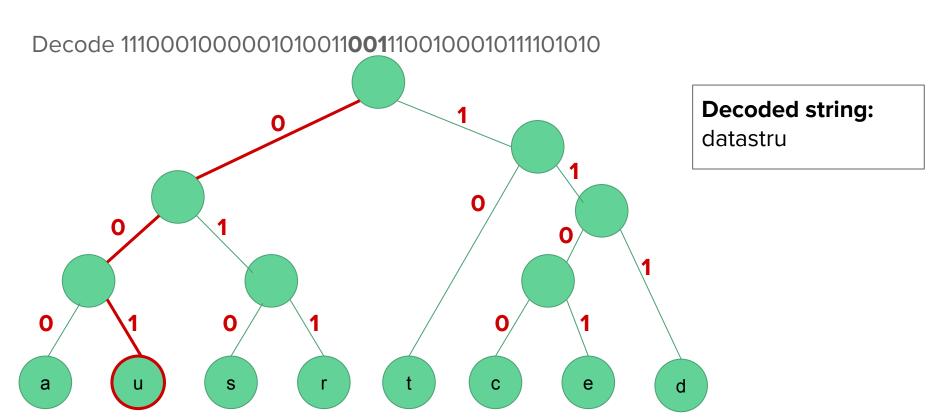


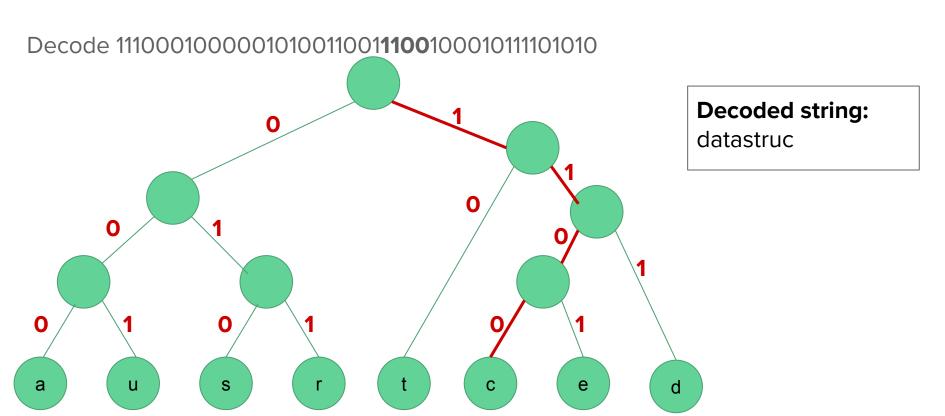
Decode 11100010**000**010100110011100100010111101010 **Decoded string:** data 0

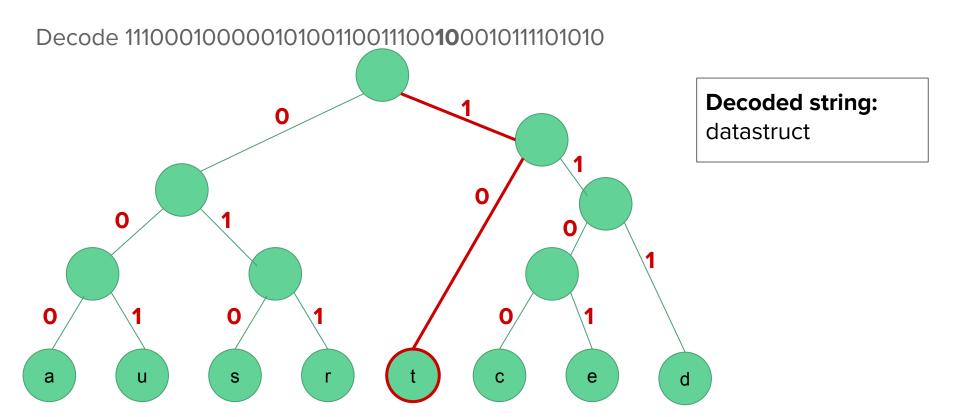
Decode 11100010000**010**100110011100100010111101010 **Decoded string:** datas 0

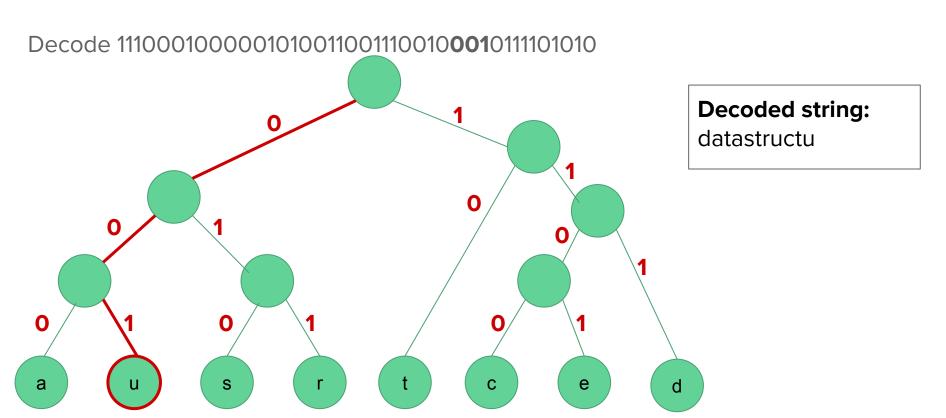


Decoded string: datastr 0 0



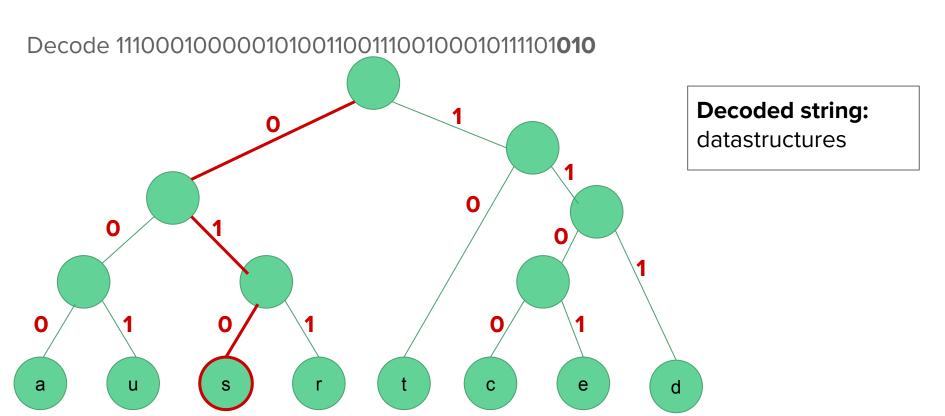






Decode 1110001000001010011001110010001**011**1101010 **Decoded string:** datastructur 0 0

Decode 1110001000001010011001110010001011**1101**010 **Decoded string:** datastructure 0 0



Activity-Selection Problem

The problem of scheduling several competing activities that require exclusive use of a common resource, with a goal of selecting a maximum-size set of mutually compatible activities.

Input: A set of activities that we wish to use a resource (such as classroom) which can serve only one activity at a time. Each activity a_i in the set $S = \{a_1, a_2, ..., a_n\}$ has a start time s_i and a finish time f_i , where $0 \le s_i < f_i < \infty$.

If selected, activity a takes place during the time interval [s, f)

Output: A maximum-size subset of mutually compatible activities.

Two activities are compatible if and only if their intervals do not overlap.

Example

Consider the following set S of activities

i	1	2	3	4	5	6	7	8	9	10	11
s_i	1	3	0	5	3	5	6	8	8	2	12
f_i	4	5	6	7	9	9	10	11	12	14	16

Greedy approach

• Choose an activity that leaves the resource available for as many other activities, i.e.

Choose an activity with the earliest finish time

Greedy algorithm:

We assume that n input activities are already ordered by monotonically increasing finish time:

$$f_1 \le f_2, \le f_3 \le ... \le f_{n-1} \le f_n$$

- 1. Select the activity with the earliest finish time
- 2. Eliminate the activities that could not be scheduled / incompatible activities
- 3. Repeat

Input: start times s, finish times f, the index k that defines the subproblem S_k it is to solve, and the size n of the original problem

Input: start times s, finish times f, the index k that defines the subproblem S_k it is to solve, and the size n of the original problem

```
RECURSIVE-ACTIVITY-SELECTOR (s, f, k, n)

1 m = k + 1

2 while m \le n and s[m] < f[k] // find the first activity in S_k to finish

3 m = m + 1

4 if m \le n

5 return \{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)

6 else return \emptyset
```

We start with k = 0 and a fictitious activity a0 with f0 = 0

```
RECURSIVE-ACTIVITY-SELECTOR (s, f, k, n)

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4 if m \le n

5 return \{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)

6 else return \emptyset
```

```
\begin{array}{c|cccc} k & s_k & f_k \\ \hline 0 & - & 0 & & \underline{a_0} \end{array}
```

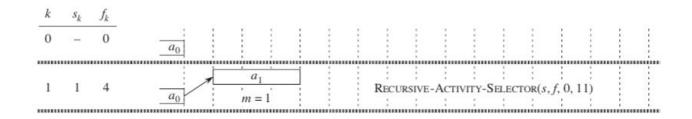
```
RECURSIVE-ACTIVITY-SELECTOR (s, f, k, n)

1 m = k + 1

2 while m \le n and s[m] < f[k] // find the first activity in S_k to finish m = m + 1
```

4 **if** $m \le n$ 5 **return** $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$

6 else return Ø



```
RECURSIVE-ACTIVITY-SELECTOR (s, f, k, n)

1 m = k + 1

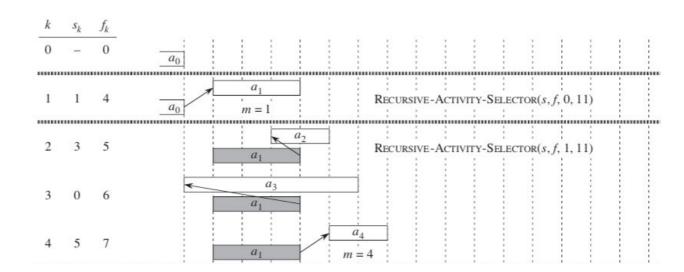
2 while m \le n and s[m] < f[k] // find the first activity in S_k to finish

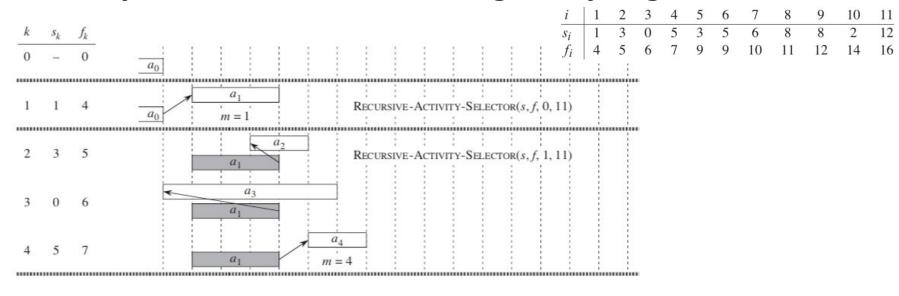
3 m = m + 1

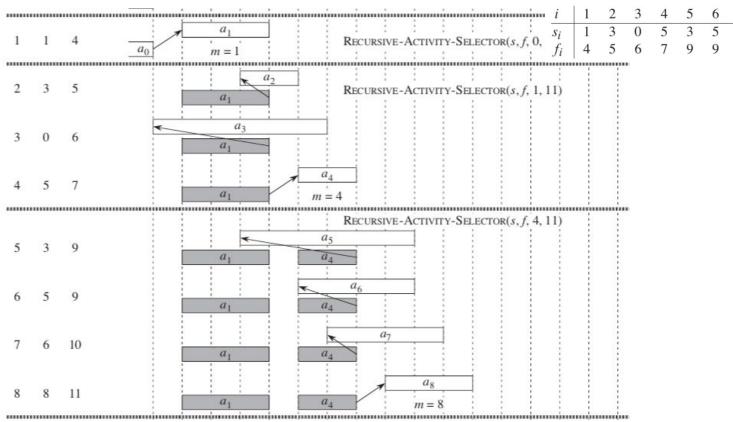
4 if m \le n

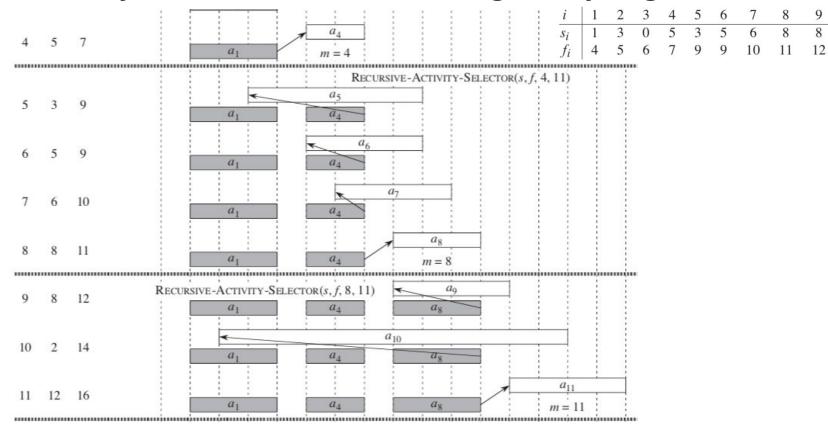
5 return \{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)

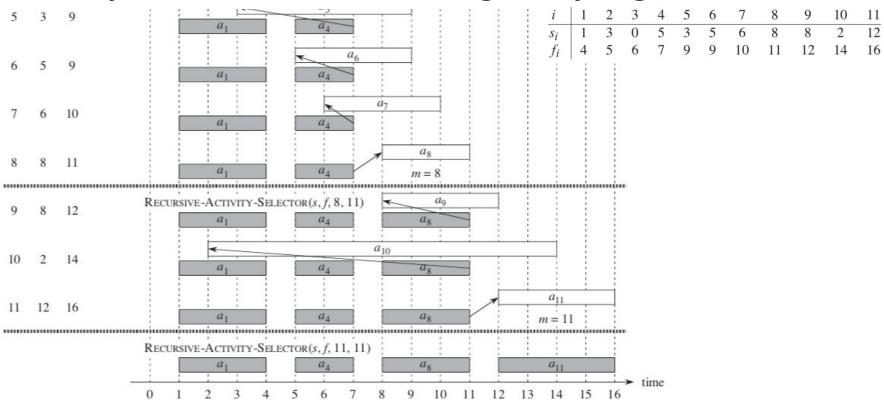
6 else return \emptyset
```











```
GREEDY-ACTIVITY-SELECTOR (s, f)
1 n = s.length
A = \{a_1\}
3 k = 1
  for m = 2 to n
      if s[m] \geq f[k]
  A = A \cup \{a_m\}
          k = m
   return A
```

Optimal substructure

A problem is said to have optimal substructure if an optimal solution can be constructed from optimal solutions of its subproblems.