

# Text Compression

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- Text Compression
- Huffman Code
- Decompression
- Analysis of Huffman Encoding

## ❖ Text Compression

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Problem: Efficiently encode a given string  $T$  by a smaller string  $E$

Applications:

- Save memory and/or bandwidth

### Huffman's algorithm

- computes frequency  $f(c)$  for each character  $c$
- encodes high-frequency characters with short code word
- no code word is a prefix of another code word (e.g. can't have  $00 + 001$ )
- uses **encoding tree** to determine the code words
- many encodings are possible; aims to find optimal encoding

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### Code ...

- mapping of each character to a binary code word (e.g. ascii)

### Prefix code ...

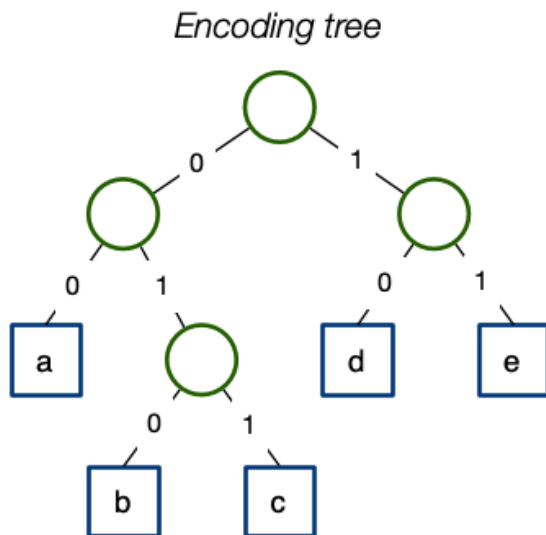
- binary code such that no code word is prefix of another code word

### Encoding tree ...

- represents a prefix code
- each leaf stores a character
- code given by the path from the root to the leaf (0 for left, 1 for right)

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Example:



*Encoding table*

char	a	b	c	d	e
code	00	010	011	10	11

*Sample encoding*

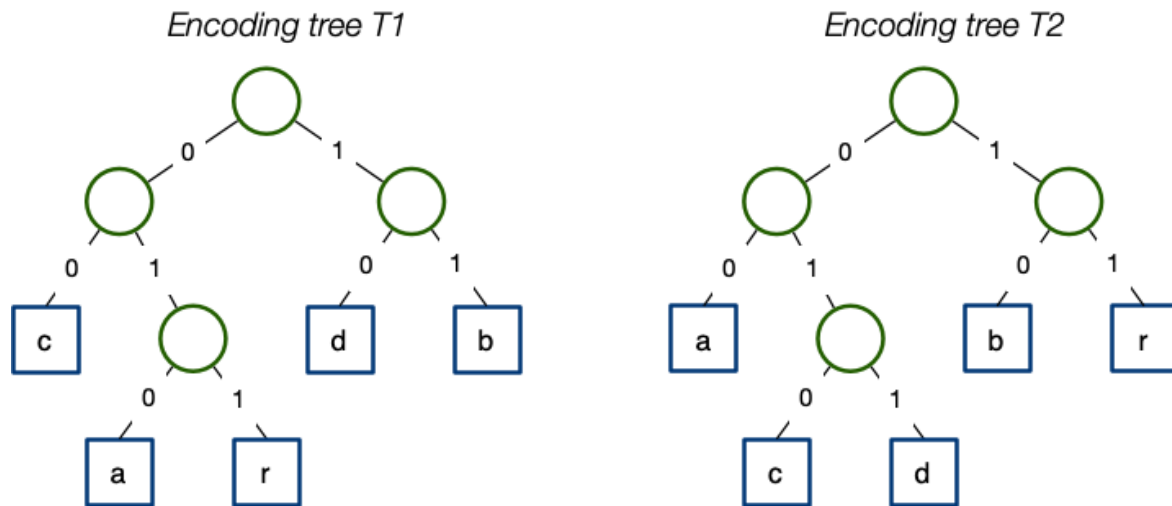
$T = \text{deadcab}$

$E = 1011001001100010$

$T$  as ascii chars would require 56 bits;  $E$  needs only 16 bits

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Example:  $T = \text{abracadabra}$



Encoding with  $T_1$  is **010.11.011.010.00.010.10.010.11.011.010** i.e. 29 bits

Encoding with  $T_2$  is **00.10.11.00.010.00.011.00.10.11.00** i.e. 24 bits

The dots are there only to distinguish characters; they are not part of the encoding.

## ❖ ... Text Compression

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### Text compression problem

Given a text  $T$ ...

- find a *prefix code* that yields the *shortest encoding* of  $T$

Some obvious strategies ...

- shorter codewords for frequent characters
- longer code words for rare characters

But how to ensure *optimal* encoding?

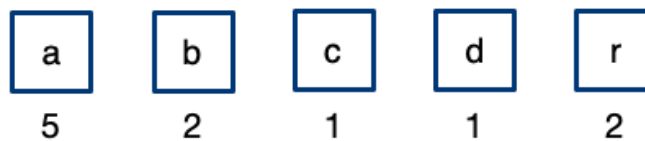
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### Huffman's algorithm

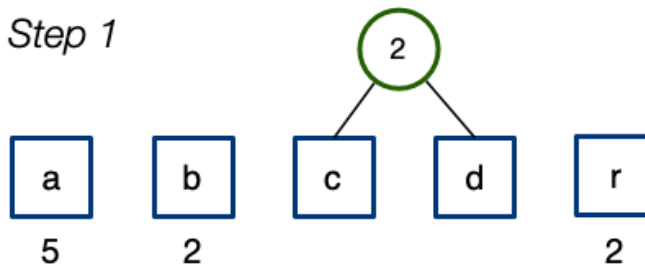
- computes frequency  $f(c)$  for each character
- successively combines pairs of lowest-frequency characters
- builds encoding tree "bottom-up"

Example:  $T = \mathbf{a b r a c a d a b r a}$

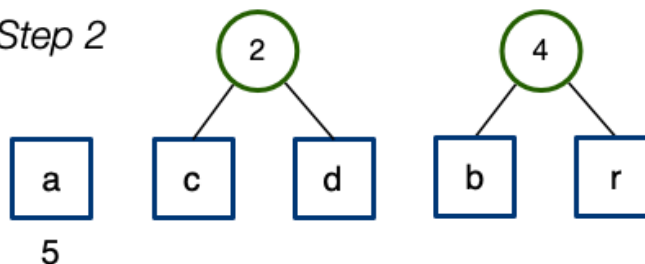
*Initially*



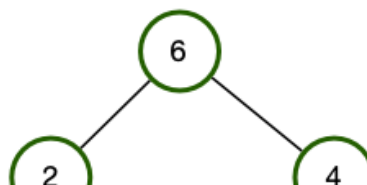
*Step 1*

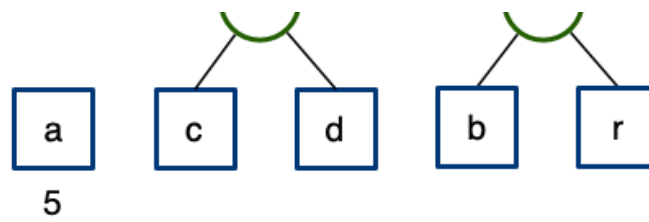


*Step 2*

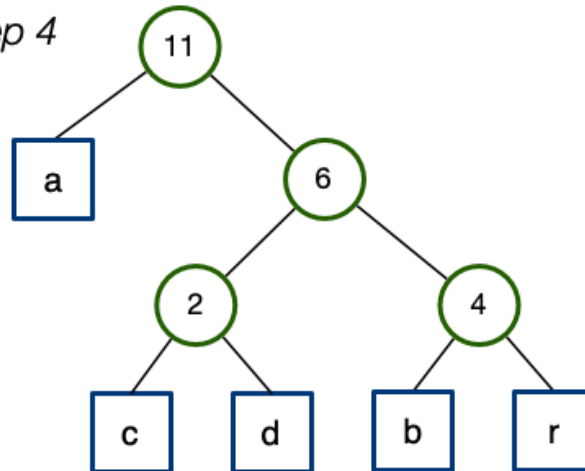


*Step 3*





Step 4



Encoding table

char	a	b	c	d	r
code	0	110	100	101	111



## ❖ Huffman Code

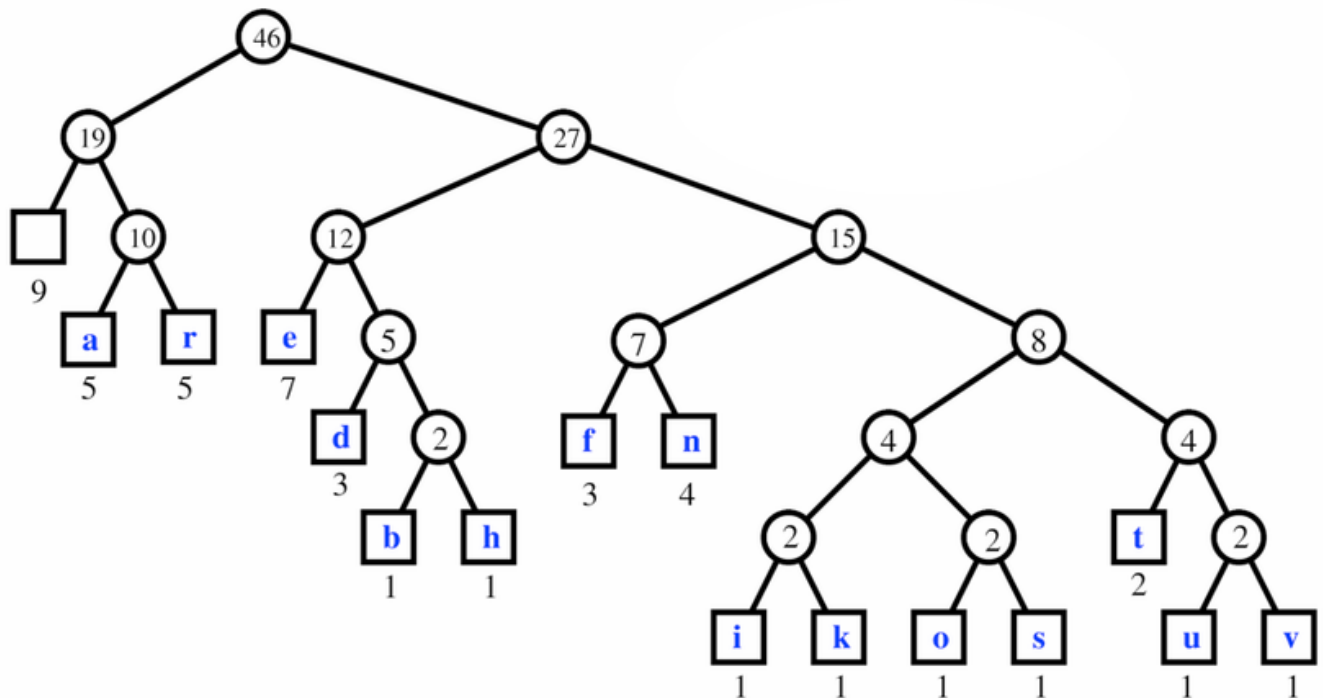
Huffman's algorithm using a **priority queue**:

```
HuffmanCode(T):  
  Input  string T of size n  
  Output optimal encoding tree for T  
  
  compute frequency array  
  Q = new priority queue (ordered low key to high key)  
  for all characters c do  
    T = new single-node tree storing c  
    join(Q,T) with frequency(c) as key  
  end for  
  while |Q| ≥ 2 do  
    f1 = Q.minKey(), T1 = leave(Q)  
    f2 = Q.minKey(), T2 = leave(Q)  
    T = new tree node with subtrees T1 and T2  
    join(Q,T) with f1+f2 as key  
  end while  
  return leave(Q)
```

## ❖ ... Huffman Code

Larger example:  $T =$  a fast runner need never be afraid of the dark

Character		a	b	d	e	f	h	i	k	n	o	r	s	t	u	v
Frequency	9	5	1	3	7	3	1	1	1	4	1	5	1	2	1	1



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## ❖ Decompression

Decompression involves repeated traversal of paths in tree ...

```
decompress(B, T):  
  | Input bit-string B, encoding tree T  
  | Output original string  
  
  | start from root of Tree  
  | for each b in Bits do  
  |   if b = 1 then  
  |     go right in Tree  
  |   else  
  |     go left in Tree  
  |   end if  
  |   if reached leaf then  
  |     print char in leaf  
  |     return to root of Tree  
  |   end if  
  | end for
```

## ❖ Analysis of Huffman Encoding

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Analysis of Huffman's algorithm:

- assume  $length(T) = m$ ,  $vocab(T) = v$
- build frequency table: scan entire input ( $m$ )
- build the tree: use frequency table ( $v$ ) via priority queue ( $\log_2 v$ )

Gives complexity:  $O(m + v \log v)$  time

Many variations exist to improve compression (e.g. gzip, bzip2, xz)

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