

Greedy Algorithms: Huffman Coding

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Programming, Data Structures and Algorithms using Python

Week 7

Efficient communication

- Send messages in English or Hindi or Tamil or ...

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 - Optimize data transfer

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 - *aa*

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 - *aa* , *et**et* , *a**e**t* , *e**t**a*?
- Use pauses between letters
 - Like adding a third symbol for encoding

Prefix codes

- Encoding of x , $E(x)$, is not a **prefix** of $E(y)$ for any x, y
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- Average number of bits per letter in encoding

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- Suppose we have the following frequencies for our earlier example

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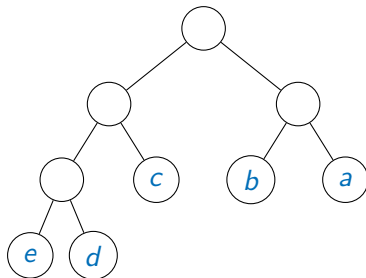
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 - $(0.32 \cdot 2) + (0.25 \cdot 2) + (0.20 \cdot 2) + (0.18 \cdot 3) + (0.05 \cdot 3)$
- Given a set of letters A and frequencies $f(x)$ for each x , produce the most efficient prefix code possible
 - Minimize $ABL(A)$ — Average Bits per Letter

Codes as trees

- Encoding can be viewed as a binary tree

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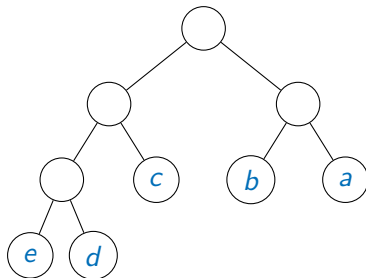


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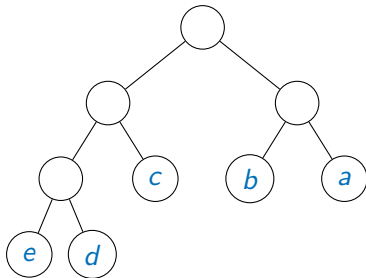


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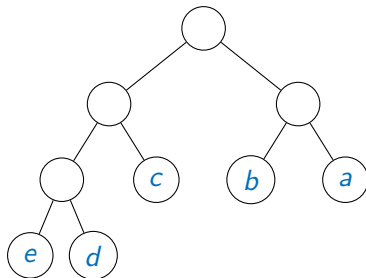


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- Prefix code — no internal nodes encode letters

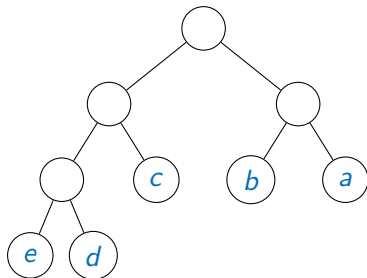


Codes as trees

Claim 1

Any optimal prefix code produces a full tree

- **Full tree** Each node has 0 or 2 children

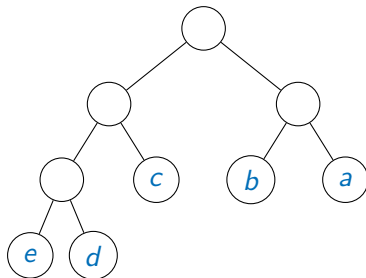


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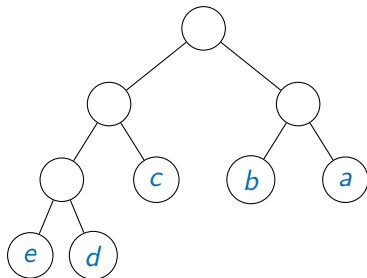
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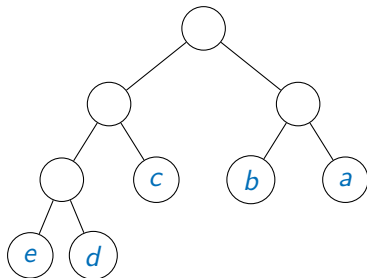
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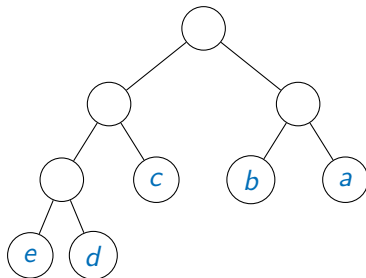
- If $f(y) > f(x)$, exchange labels, improve tree



Codes as trees

Claim 3

In an optimal tree, for any leaf at maximum depth, its sibling is also a leaf x

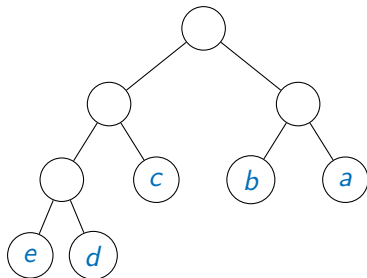


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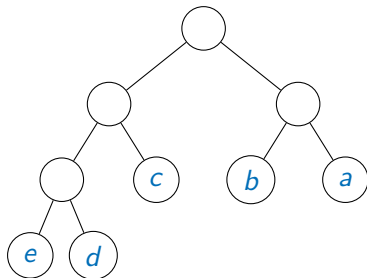


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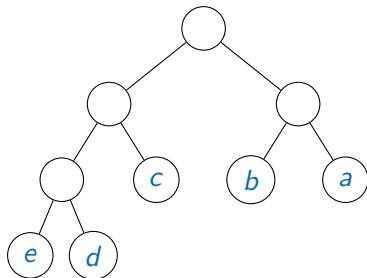


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- If not, the sibling of this leaf has children
- There is a leaf at lower depth
- The leaf we started with is not a maximum depth!



A recursive algorithm

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- From Claim 3, leaves at maximum depth occur in pairs

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- “Combine” x, y into new letter xy with $f(xy) = f(x) + f(y)$

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- **Huffman coding** — David E Huffman

Huffman's algorithm

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- Combine d , e as de

x	a	b	c	de
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Huffman's algorithm

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x	a	b	c	de
$f(x)$	0.32	0.25	0.20	0.23

x	a	b	cde
$f(x)$	0.32	0.25	0.43

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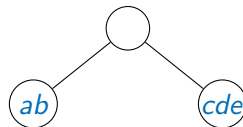
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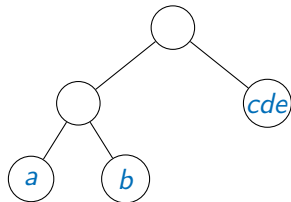
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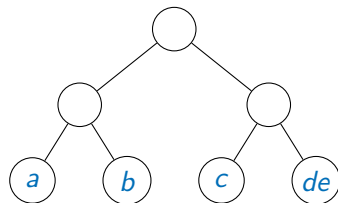
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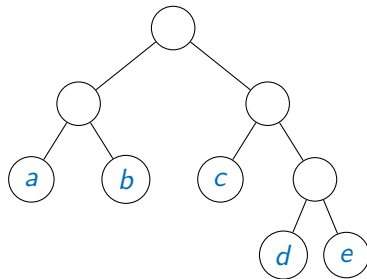
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- Hence $ABL(T) \leq ABL(S)$, a contradiction

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- Instead, maintain frequencies in a heap
 - Extracting two minimum frequency letters and adding back compound letter are both $O(\log k)$
 - Complexity drops to $O(k \log k)$

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