

BINARY CHARACTER CODES

for

LOSSLESS COMPRESSION

- Introduction to data compression

- Khalid Sayood

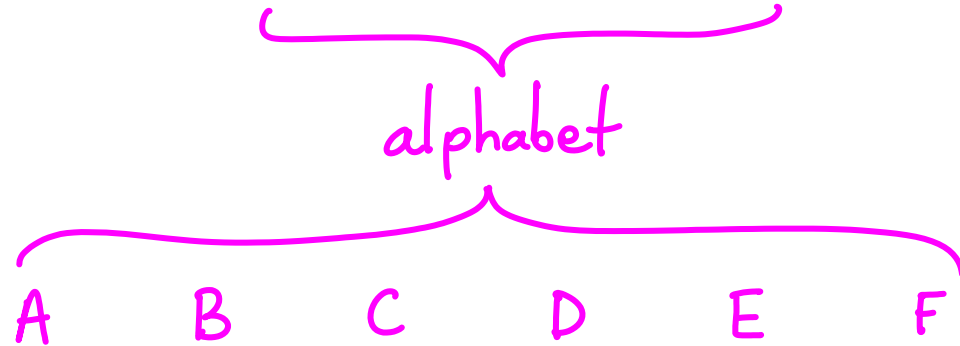
& CLRS

BINARY CHARACTER CODES

How to encode ...ABACBBFAEFDACB... in binary, knowing full string

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How to encode ...**ABACBBFAEFDACB**... in binary, knowing full string

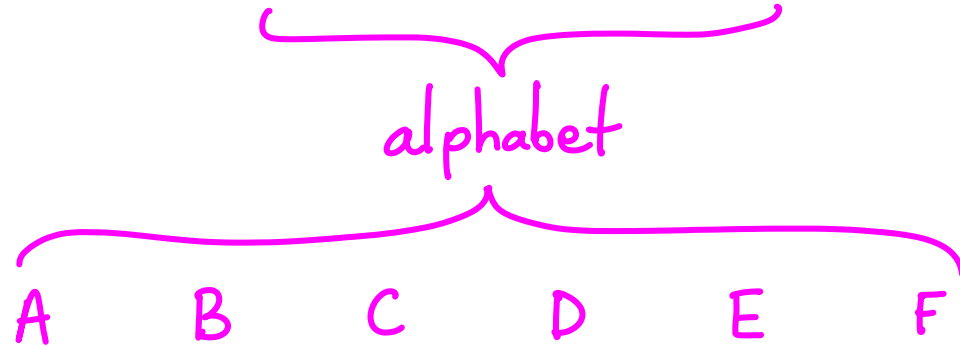


000 001 010 011 100 101 →

3 bits/char.
impossible w/ 2 bits

BINARY CHARACTER CODES

How to encode ...**ABACBBFAEFDACB**... in binary, knowing full string



000 001 010 011 100 101 →

Why not 0 1 01 011 100 101 ?

3 bits/char.
impossible w/ 2 bits

A

B

C

D

0

0

1

10

terrible

A

B

C

D

0

0

1

10

terrible

0

1

00

11

ambiguous

A	B	C	D	
0	0	1	10	terrible
0	1	00	11	ambiguous
0	10	110	111	uniquely decodable

A

B

C

D

0

0

1

10

terrible

0

1

00

11

ambiguous

0

10

110

111

uniquely decodable

0

01

011

0111

?

A	B	C	D	
0	0	1	10	terrible
0	1	00	11	ambiguous
0	10	110	111	uniquely decodable
0	01	011	0111	uniquely decodable but not instantaneous
				01 = "B" or continue?

How do we know if a code is uniquely decodable?

0

01

11

A C C C C ...
0 1 1 1 1 1 1 1 1 1 1 } resolved at end
B C C C C

A

B

C

How do we know if a code is uniquely decodable?

0

01

11

A C C C C ...
0 1 1 1 1 1 1 1 1 1 1 1 1 1 } resolved at end
B C C C C

A

B

C

0

01

10

?

How do we know if a code is uniquely decodable?

0	01	11	$\begin{array}{c} A \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \dots \\ \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \underbrace{1} \text{ } \dots \\ B \text{ } C \text{ } C \text{ } C \text{ } C \end{array}$	} resolved at end
A	B	C		

0	01	10	$\begin{array}{c} A \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \text{ } \overbrace{C} \\ \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \text{ } \underbrace{1} \text{ } \underbrace{0} \\ B \text{ } B \text{ } B \text{ } B \text{ } B \text{ } B \text{ } B \text{ } B \text{ } A \end{array}$	Problem
A	B	C		

How do we know if a code is uniquely decodable?

0	01	11	
A	B	C	

Diagram illustrating a decoding failure:

Top row (Red): A C C C C C C C C

Bottom row (Green): B C C C C C C C ?

Arrows indicate that the top row is resolved at the end, while the bottom row fails to resolve. A pink bracket on the right says "resolved at end". A yellow highlight is under the question mark.

If we encoded ACC... then decoding BCC... fails 😞
(& vice versa)

0	01	10	

Diagram illustrating a decoding problem:

Top row (Red): A C C C C C C C C

Bottom row (Green): B B B B B B B B A

Arrows indicate that the top row is resolved at the end, while the bottom row fails to resolve. The word "Problem" is written in orange.

Whatever we encoded, both decodings work 😞

How do we know if a code is uniquely decodable?

$$A = c_1 c_2 c_3 \dots c_k$$

$$B = c_1 c_2 c_3 \dots c_k d_1 d_2 \dots d_j$$

} A is a prefix of B

How do we know if a code is uniquely decodable?

$$\begin{array}{l} A = c_1 c_2 c_3 \dots c_K \\ B = c_1 c_2 c_3 \dots c_K d_1 d_2 \dots d_j \end{array} \left. \begin{array}{l} \text{Yellow vertical lines connect } c_i \text{ in } A \text{ to } c_i \text{ in } B. \\ \text{Red bracket under } d_1 d_2 \dots d_j \text{ in } B. \end{array} \right\} \begin{array}{l} A \text{ is a prefix of } B \\ \text{"dangling suffix"} \end{array}$$

How do we know if a code is uniquely decodable?

Not good:

$A = c_1 c_2 c_3 \dots c_k$

$B = c_1 c_2 c_3 \dots c_k d_1 d_2 \dots d_j$

} A is a prefix of B...
and

"dangling suffix"...is another codeword

A	B	C
0	01	1

How do we know if a code is uniquely decodable?

Not good: $A = c_1 c_2 c_3 \dots c_k$
(but also not a sufficient test) $B = c_1 c_2 c_3 \dots c_k d_1 d_2 \dots d_j$ } A is a prefix of B...
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0	01	11	☺
A	B	C	
0	01	10	☹
	also ok ?!		

How do we know if a code is uniquely decodable?

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0	01	11	☺
A	B	C	

Extended test: if $\exists$ dangling suffix

0	01	10	☹
---	----	----	---

0 01 11

0 01 10

How do we know if a code is uniquely decodable?

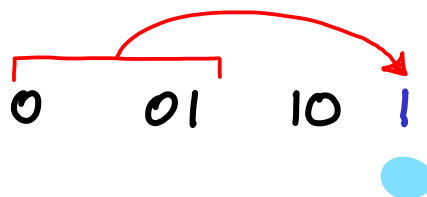
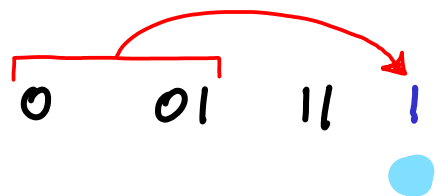
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0 01 11 ☺

A B C

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Extended test: if $\exists$ dangling suffix
add it to a list (set)



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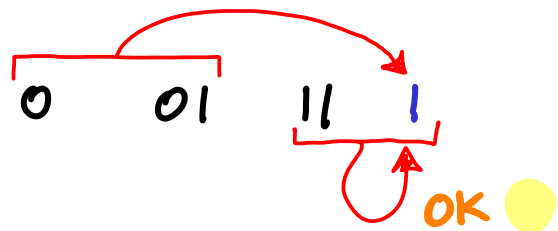
0 01 11 ☺

A B C

0 01 10 ☹

Extended test: if $\exists$ dangling suffix
add it to a list (set)

& repeat original test
on combined set
until no new listwords
get generated.



0 01 10 1

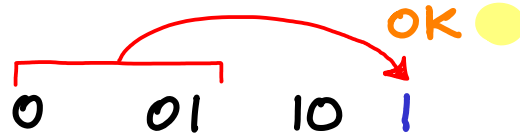
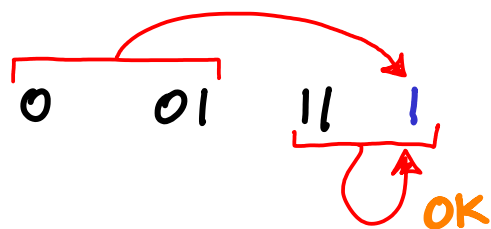
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$\left. \begin{array}{l} A \text{ is a prefix of } B \dots \\ \text{and} \end{array} \right\}$
"dangling suffix"...is another codeword

0	01	11	☺
A	B	C	
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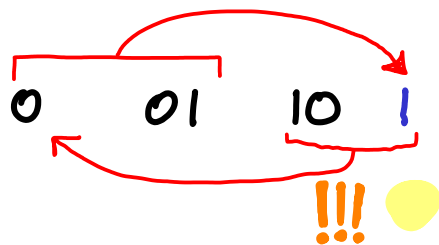
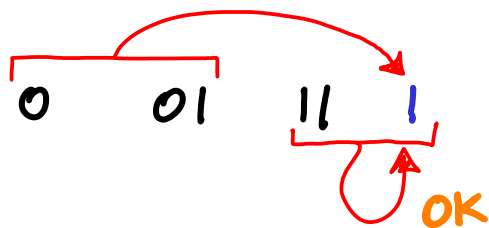


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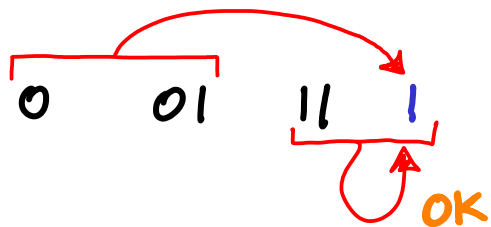
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"dangling suffix"...is another codeword

0	01	11	STILL OK 😊
A	B	C	Extended test:
0	01	10	

PROOF OMITTED

if $\exists$ dangling suffix
add it to a list (set)
& repeat original test
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get generated.



BINARY CHARACTER CODES

How to encode ...ABACBBFAEFDACB... in binary, knowing full string

alphabet

A B C D E F

45 13 12 16 9 5 : % frequencies

000 001 010 011 100 101 →

3 bits/char.

impossible w/ 2 bits

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0 101 100 111 1101 1100 : Variable length code

BINARY CHARACTER CODES

How to encode ...ABACBBFAEFDACB... in binary, knowing full string

alphabet

A B C D E F

45 13 12 16 9 5 : % frequencies

000 001 010 011 100 101 → average 3 bits/char.

impossible w/ 2 bits

0 101 100 111 1101 1100 : Variable length code

$\Sigma(\text{freq} \cdot \text{bits})$ $.45 \cdot 1 + (.13 + .12 + .16) \cdot 3 + (.09 + .05) \cdot 4$

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$\Sigma(\text{freq} \cdot \text{bits})$.45·1 + (.13+.12+.16)·3 + (.09+.05)·4 → average 2.24 bits/char.

BINARY CHARACTER CODES

How to encode ...ABACBBFAEFDACB... in binary, knowing full string

alphabet

A B C D E F

New example → 20 13 17 16 19 15 : % frequencies

000 001 010 011 100 101 → average 3 bits/char.

impossible w/ 2 bits

10 000 010 011 11 001 : Variable length code

$.2 \cdot 2 + (.13 + .17 + .16) \cdot 3 + .19 \cdot 2 + .15 \cdot 3 \rightarrow$ average 2.46 bits/char.

BINARY CHARACTER CODES

How to encode ...ABACBBFAEFDACB... in binary, knowing full string

alphabet

A B C D E F

3rd example → 95 1 1 1 1 1 : % frequencies

000 001 010 011 100 101 → average 3 bits/char.

for fixed length: impossible w/ 2 bits

1 000 001 0100 0101 011 : Variable length code

$.95 \cdot 1 + (.01 + .01) \cdot 3 + (.01 + .01) \cdot 4 + .01 \cdot 3 \rightarrow$ average 1.12 bits/char.

How to encode A A A C B A D A E F D A C B...

A	B	C	D	E	F	:	% frequencies
45	13	12	16	9	5		

How to encode A A A C B A D A E F D A C B...

000 • 000 • 000 • 010 • 001 • 000 • 011 • 000 • 100 • 101 • 011 • 000 • 010 • 001 • ...

0 • 0 • 0 • 100 • 101 • 0 • 111 • 0 • 1101 • 1100 • 111 • 0 • 100 • 101

A A A C B A D A E F D A C B...

A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies
000	001	010	011	100	101	: fixed length code
0	101	100	111	1101	1100	: Variable length code

How to encode A A A C B A D A E F D A C B...

000 • 000 • 000 • 010 • 001 • 000 • 011 • 000 • 100 • 101 • 011 • 000 • 010 • 001 • ...

Fixed

00000000000100010000110001001010110000010001

Variable

00010010101110110111001110100101

0 • 0 • 0 • 100 • 101 • 0 • 111 • 0 • 1101 • 1100 • 111 • 0 • 100 • 101

A A A C B A D A E F D A C B...

why ok?

A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies
000	001	010	011	100	101	: fixed length code
0	101	100	111	1101	1100	: Variable length code

A A A C B A D A E F D A C B...

000 • 000 • 000 • 010 • 001 • 000 • 011 • 000 • 100 • 101 • 011 • 000 • 010 • 001 • ...

Fixed

0000000000010001000011000100101011000010001

why ok?

Variable

00010010101110110111001110100101

0 • 0 • 0 • 100 • 101 • 0 • 111 • 0 • 1101 • 1100 • 111 • 0 • 100 • 101

A A A C B A D A E F D A C B...

Prefix codes: no code is a prefix of another e.g., only A starts with 0

unlike, say, A=01, B=011, C=1 → 011 = B or AC?

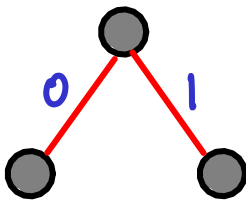
We still need to know when each code ends
^{word}

How do we know there is no 0001 ?

0001001010111011011001110100101

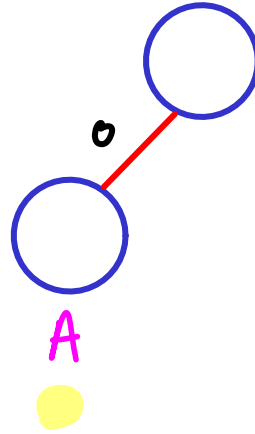
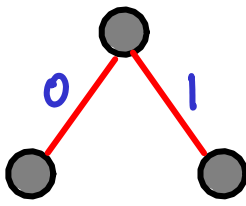
A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies
0	101	100	111	1101	1100	: Variable length (prefix) code

We still need to know when each code ends → use binary tree



A	B	C	D	E	F	
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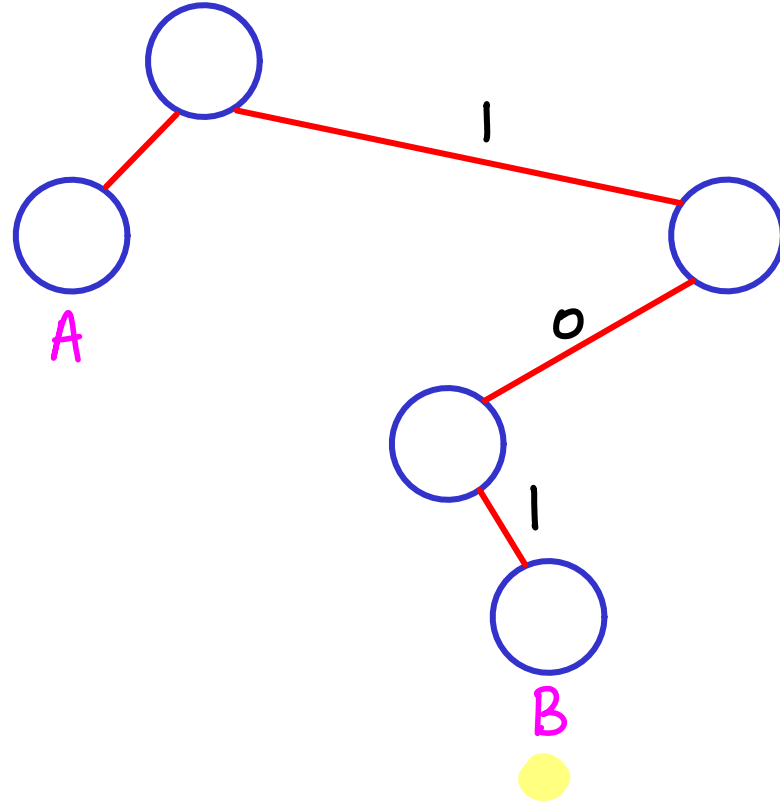
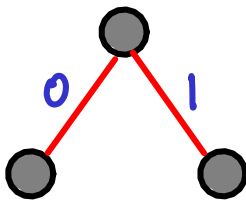
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●
A B C D E F : % frequencies
45 13 12 16 9 5

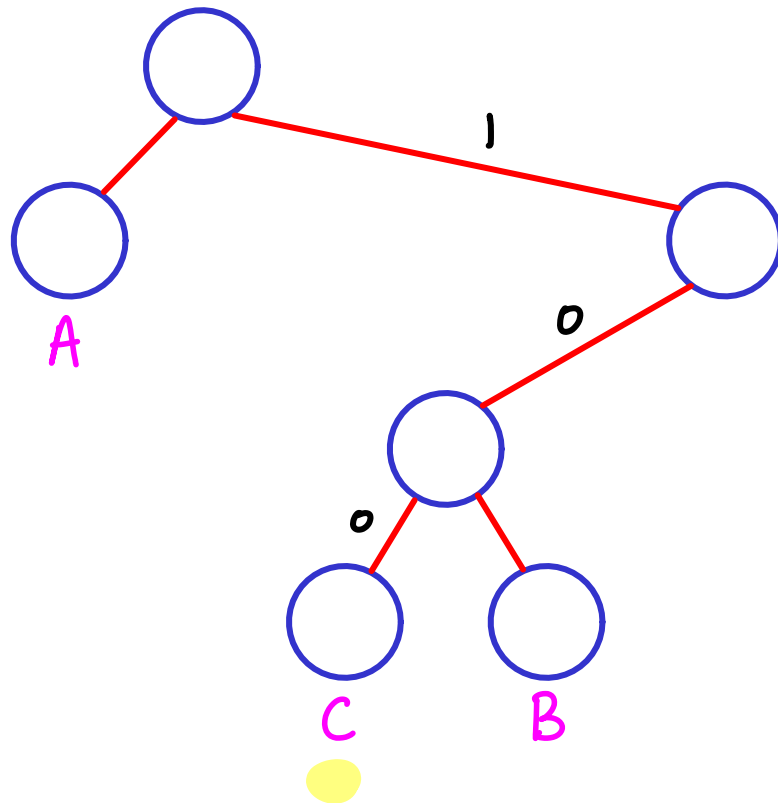
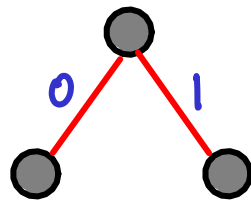
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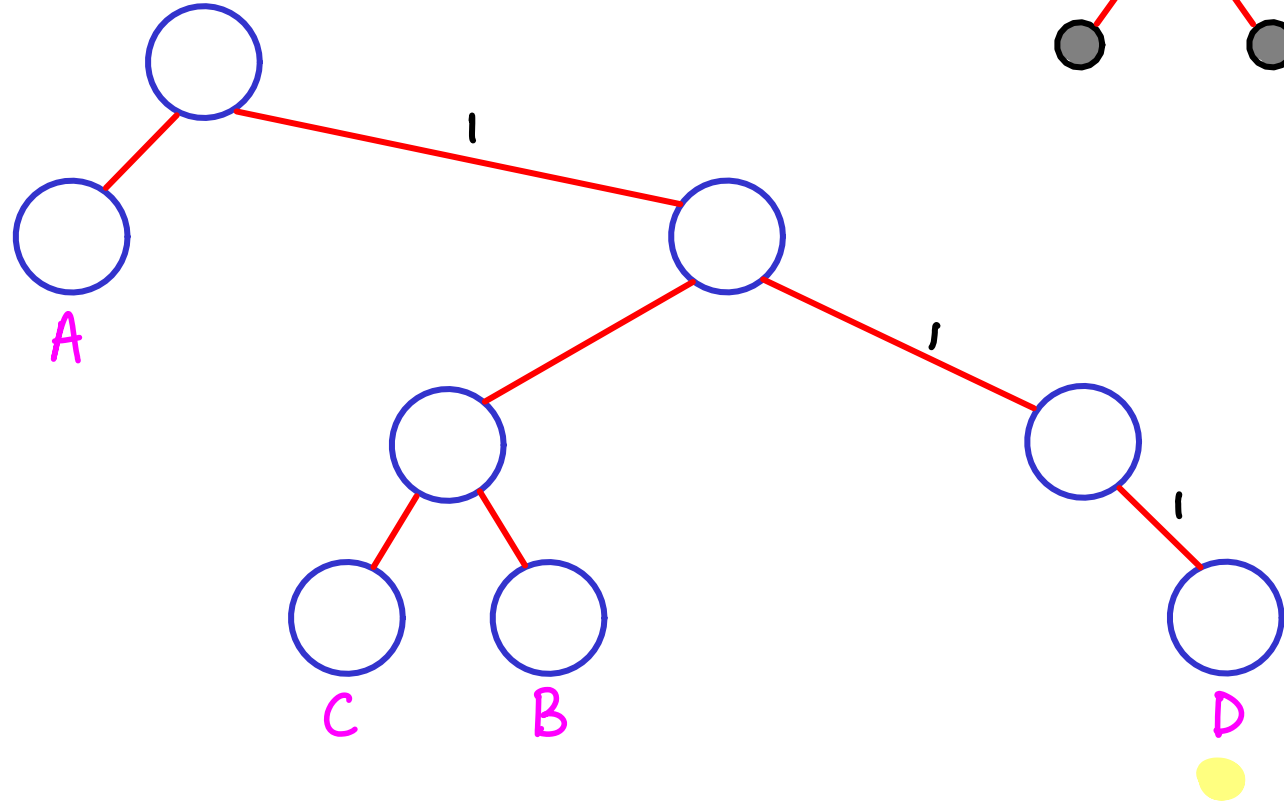
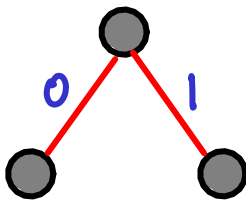
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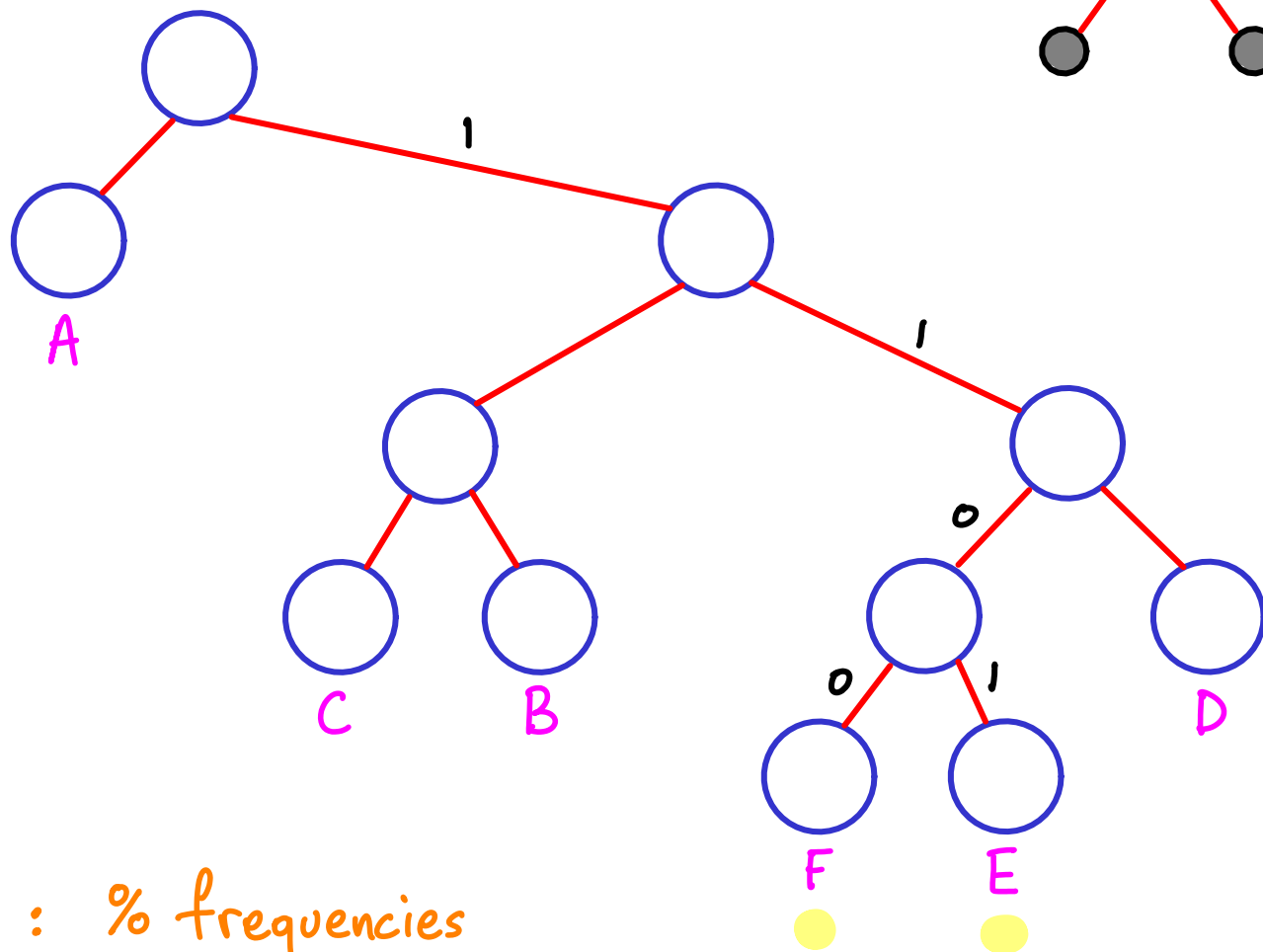
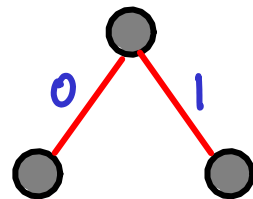
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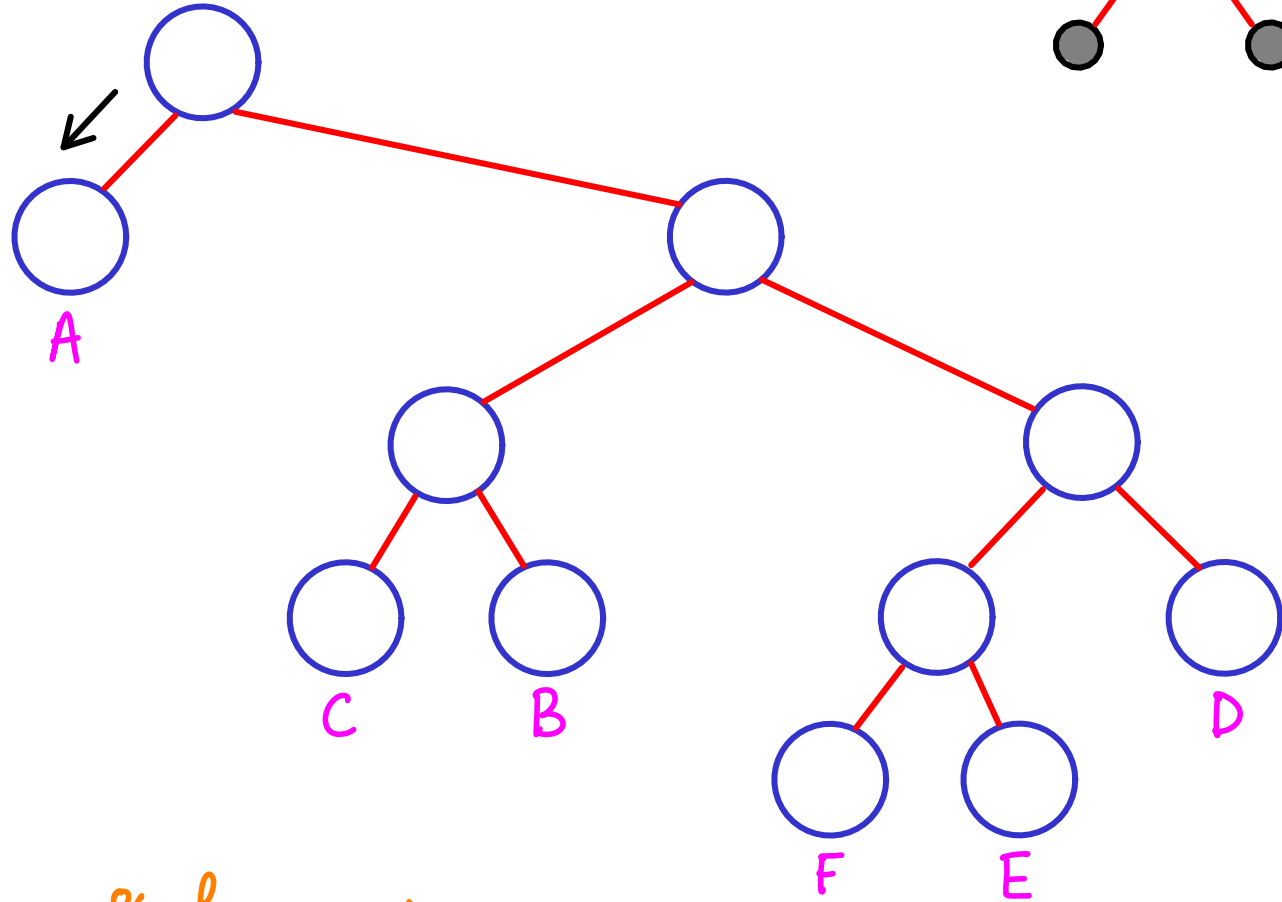
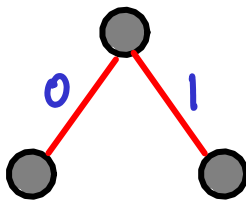
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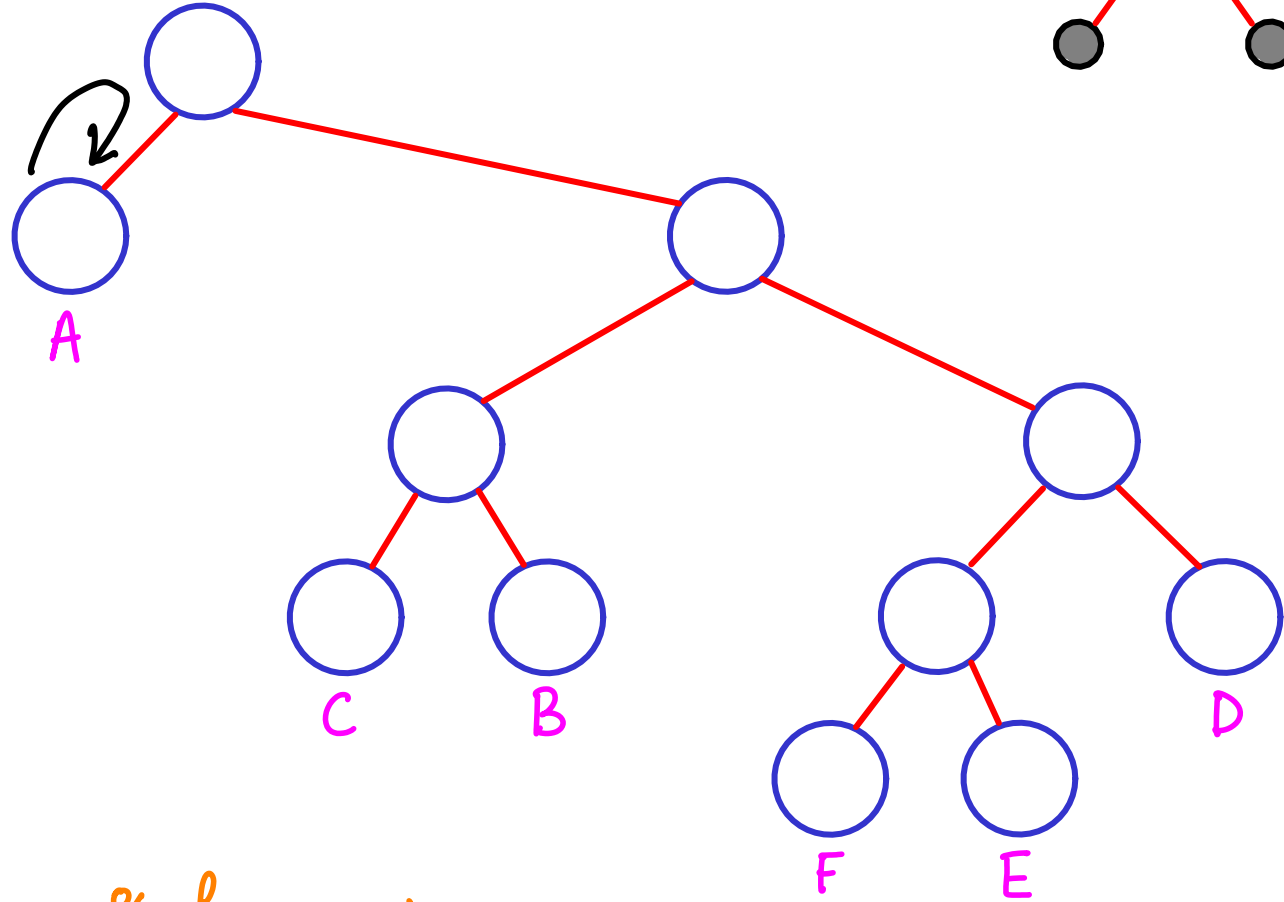
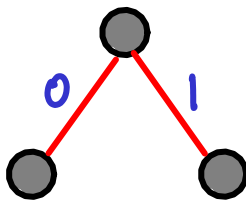


00010010101101101100110100101
A

A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies

A	B	C	D	E	F	
0	101	100	111	1101	1100	: Variable length (prefix) code

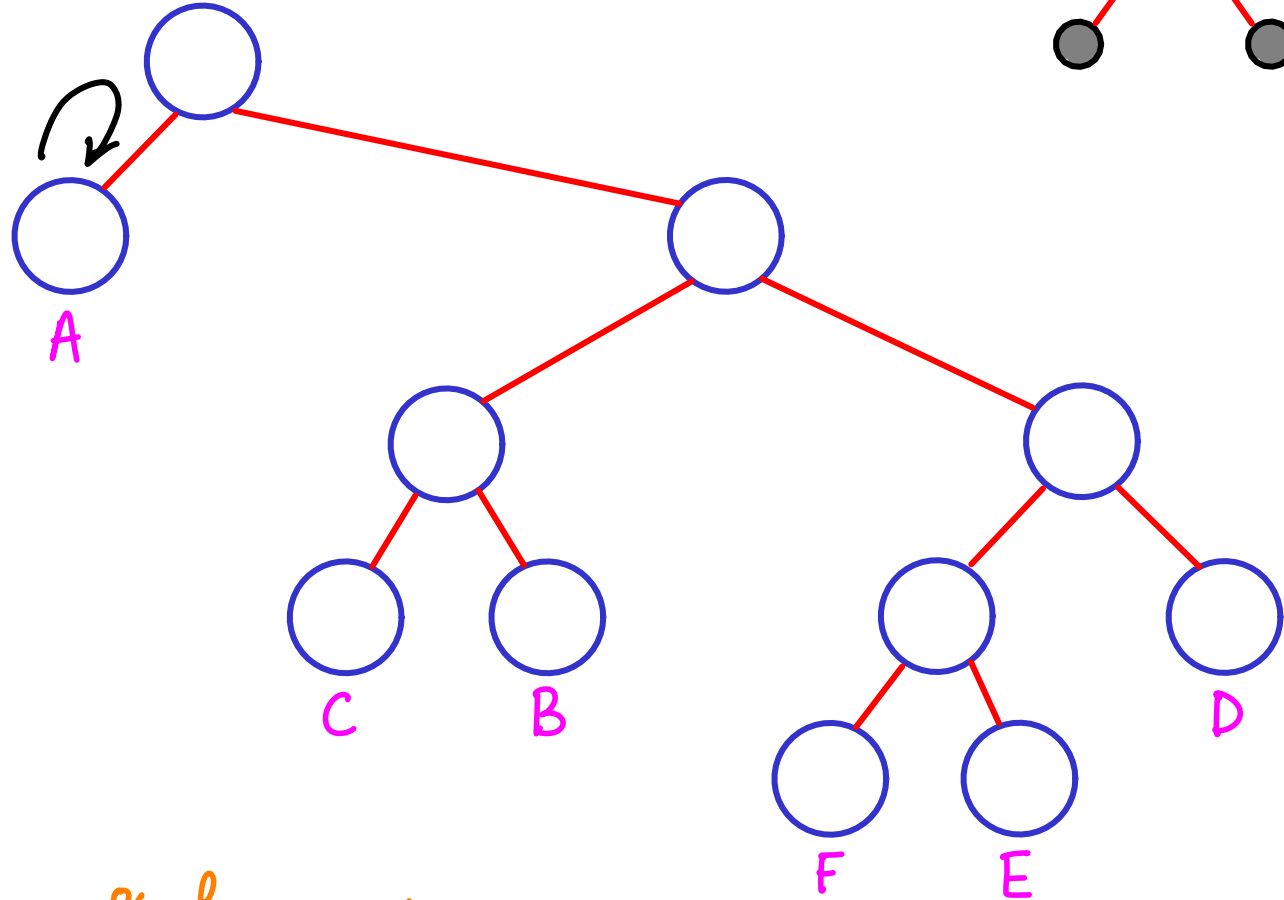
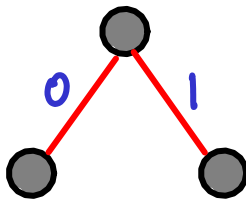
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00010010101101101100110100101
AA

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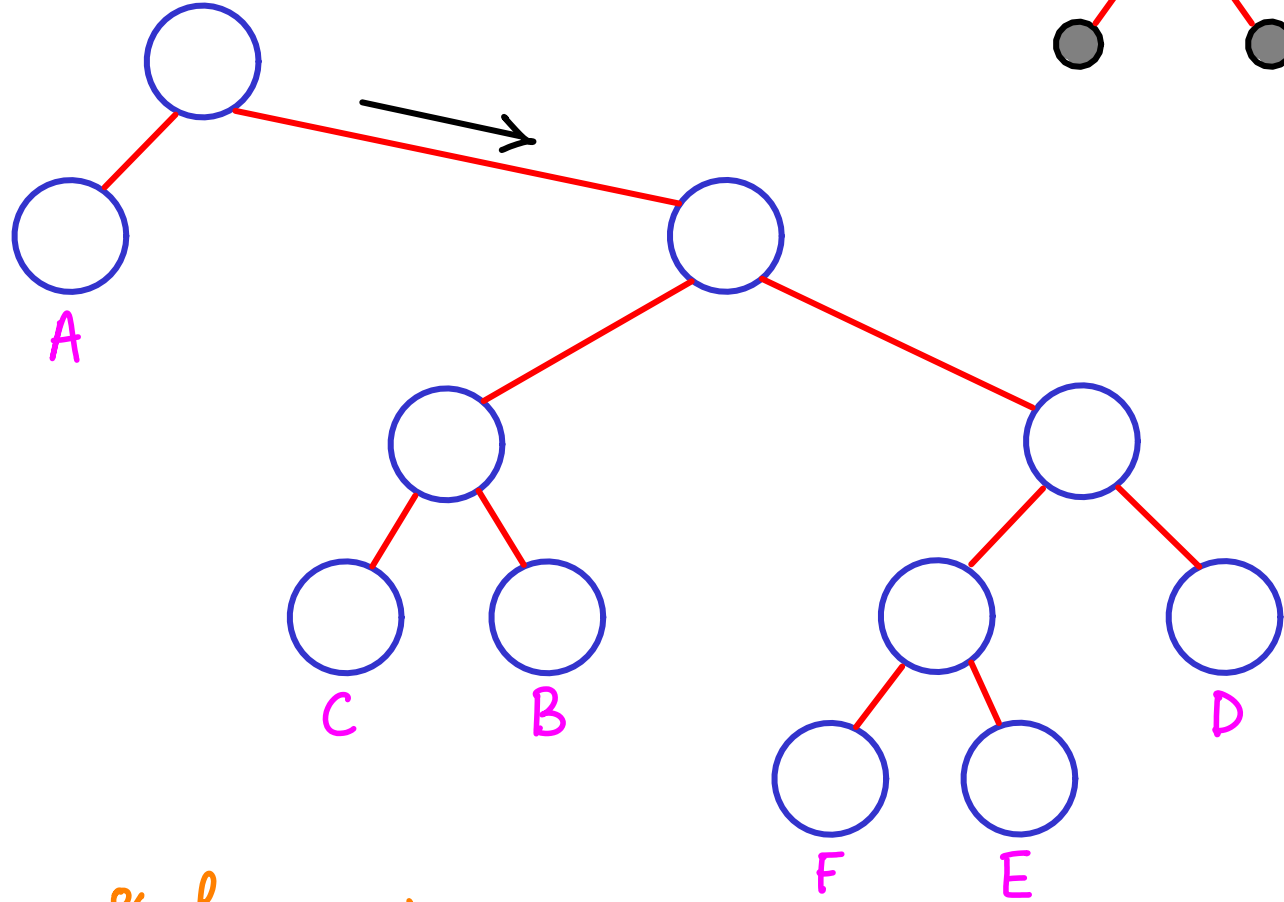
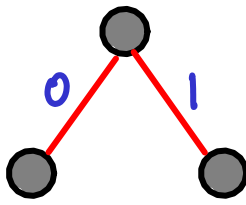
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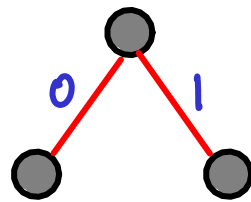
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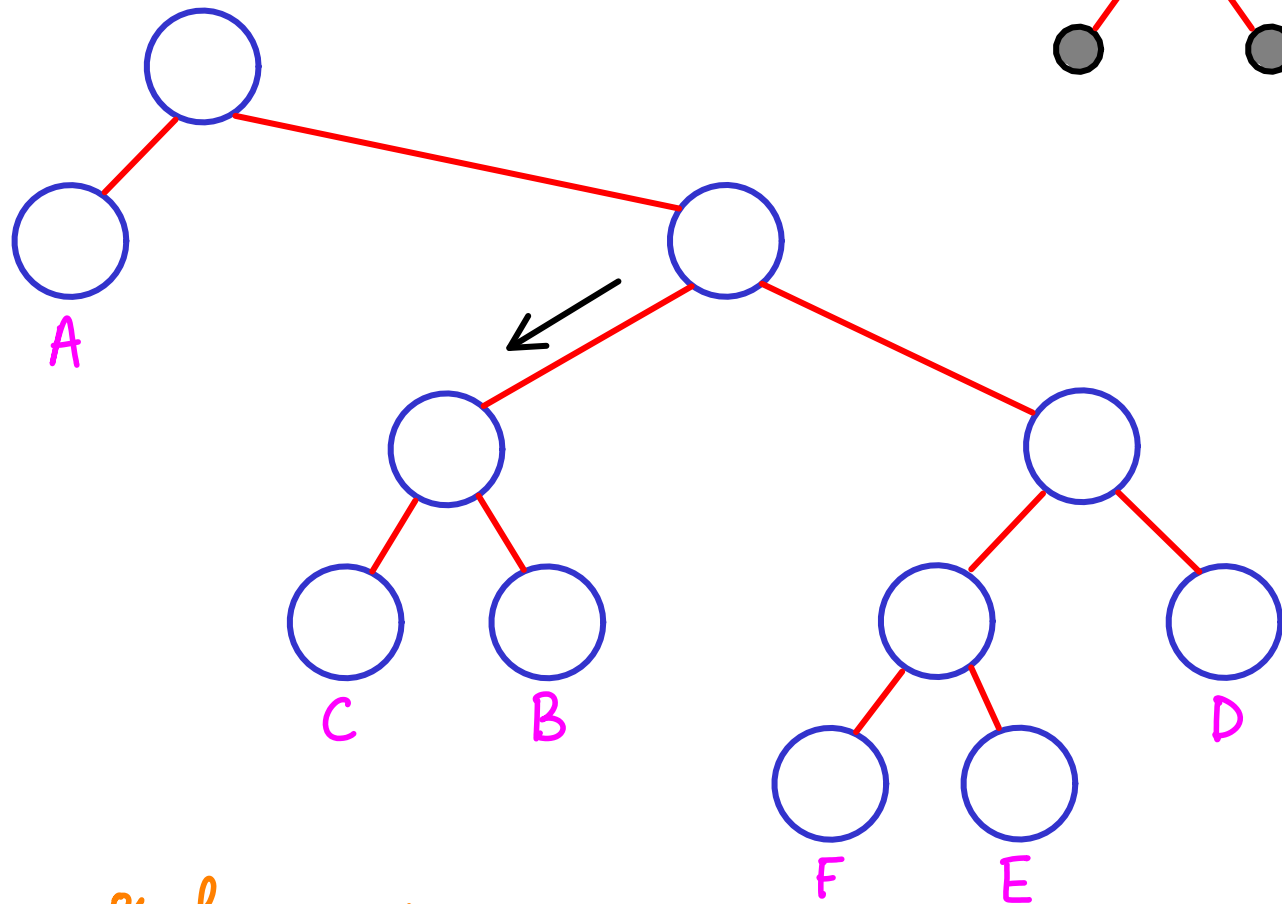
00010010101101101100110100101
AAA

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We still need to know when each code ends → use binary tree



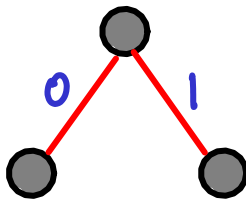
00010010101101101100110100101
AAA



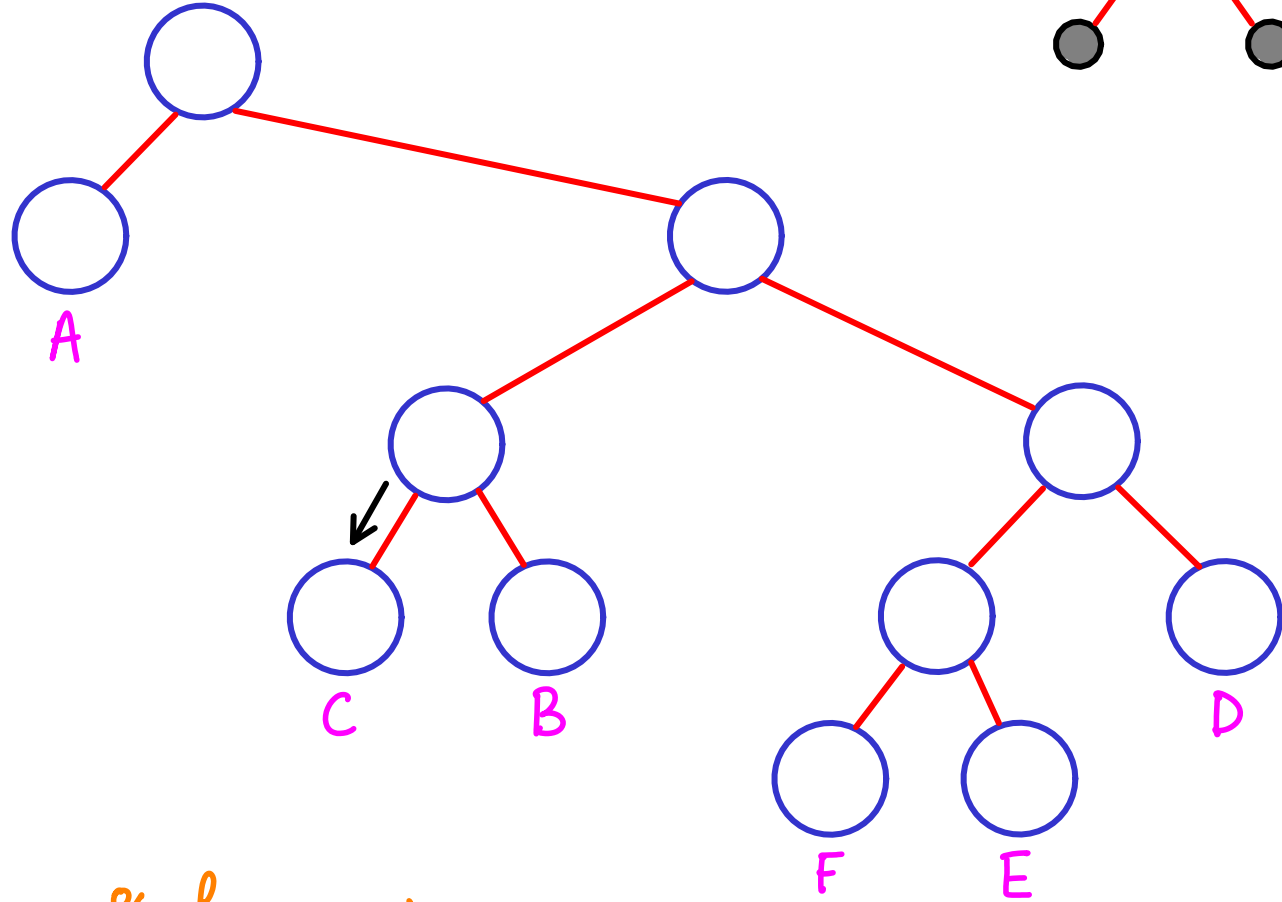
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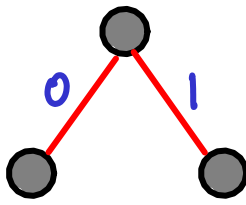
00010010101101101100110100101
AAA C



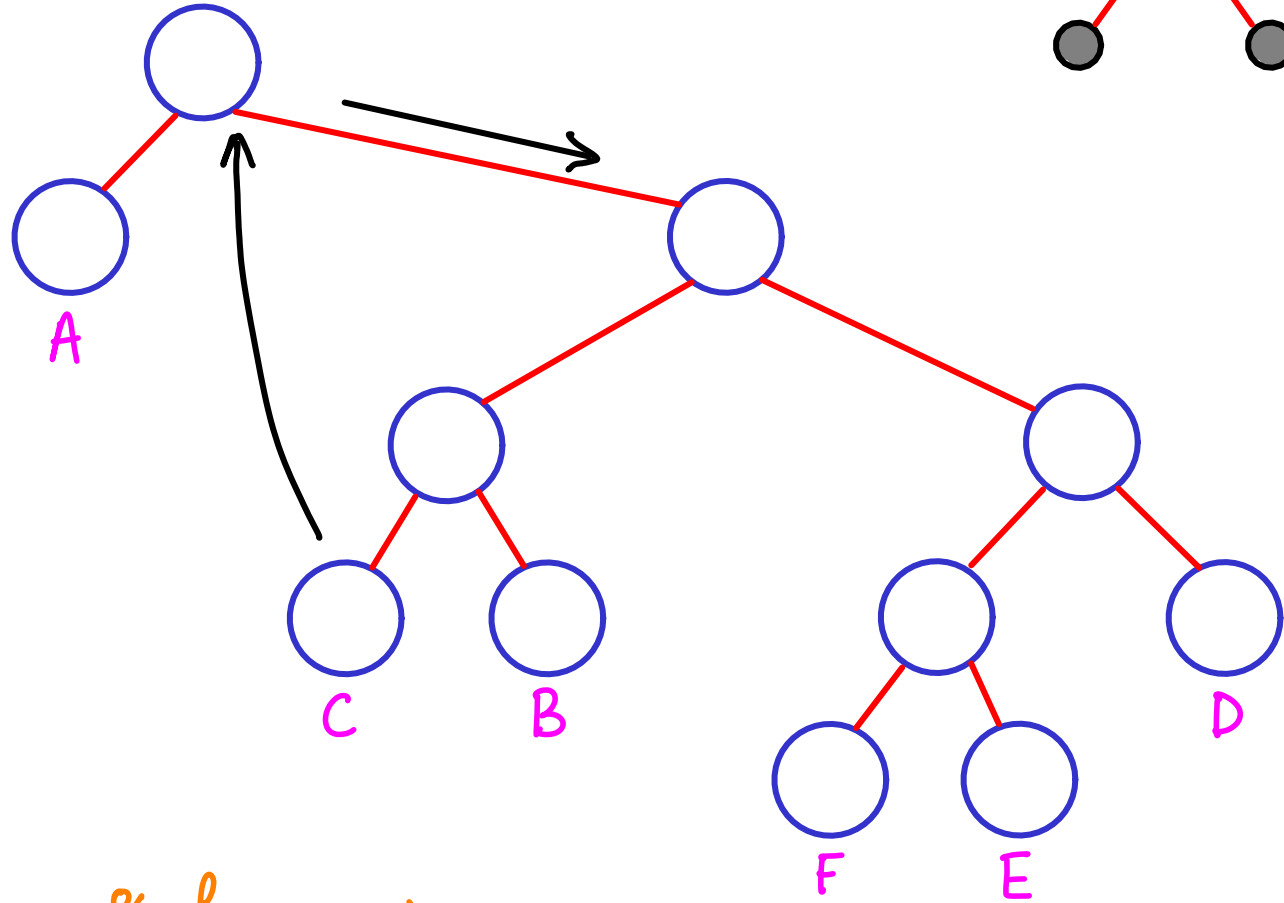
A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies

0	101	100	111	1101	1100	
						: Variable length (prefix) code

We still need to know when each code ends → use binary tree



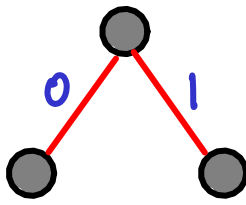
00010010101101101100110100101
AAA C



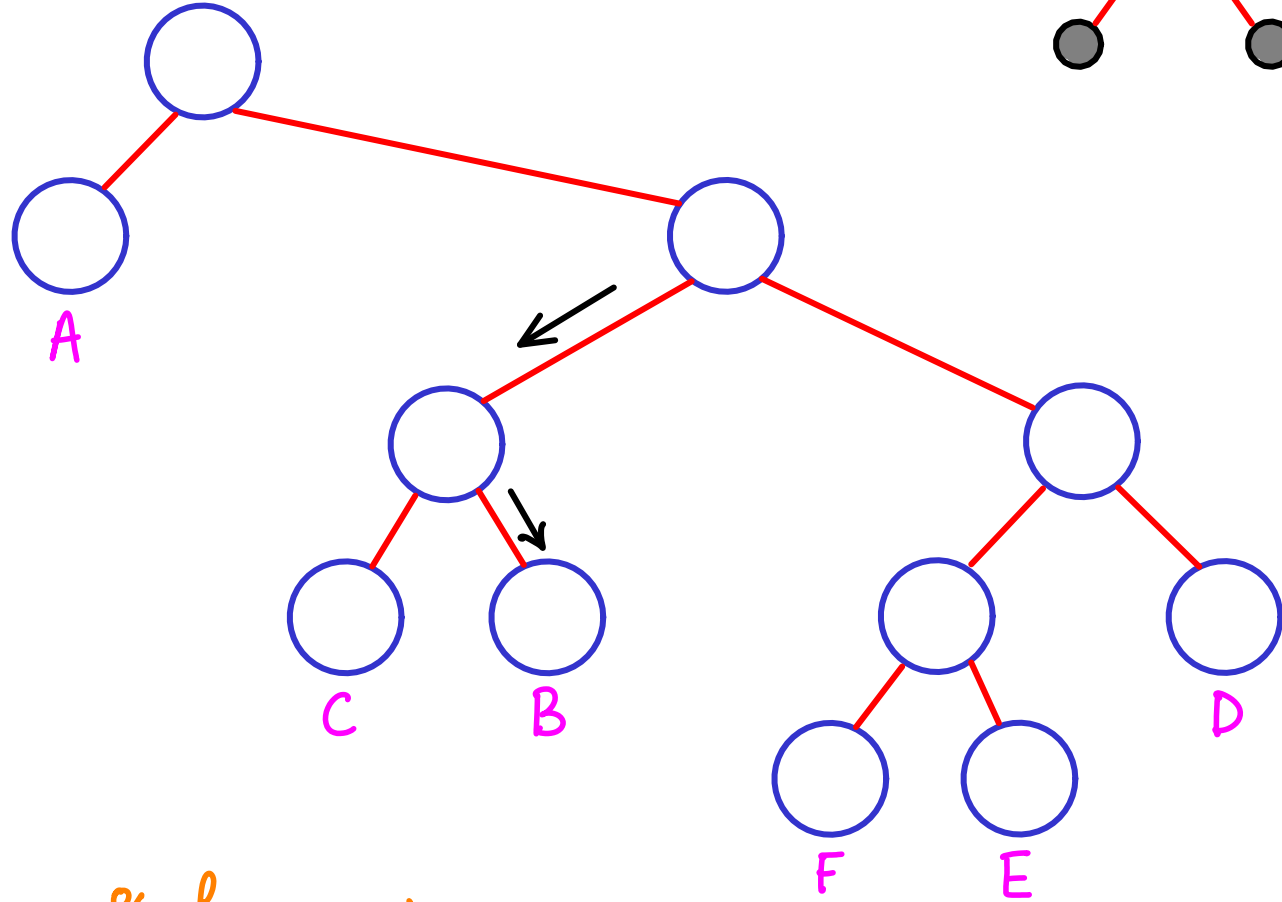
A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies

0	101	100	111	1101	1100	
						: Variable length (prefix) code

We still need to know when each code ends → use binary tree

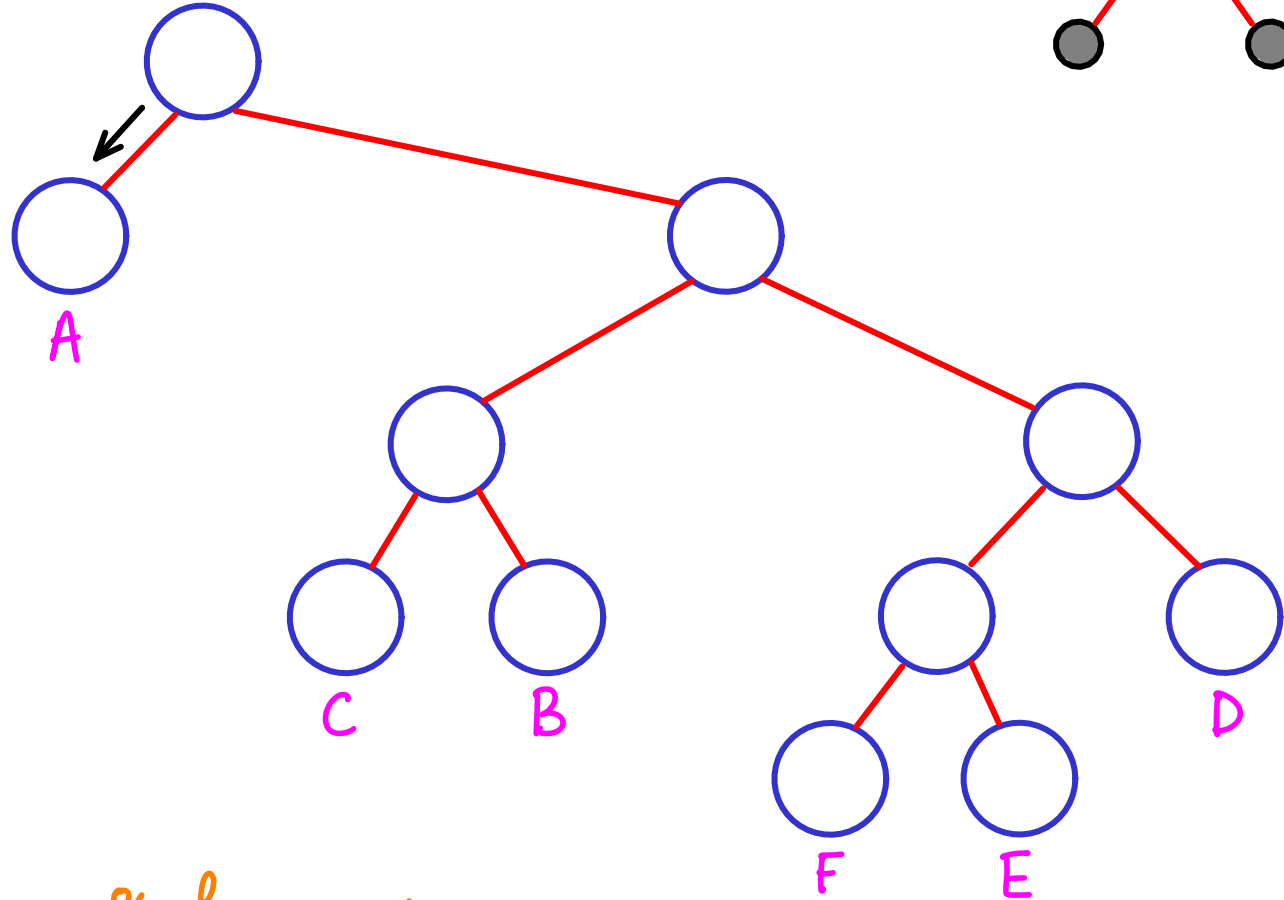
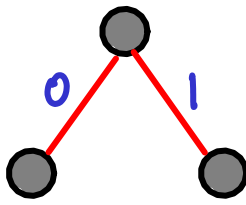


00010010101101101100110100101
AAA C B



A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies
0	101	100	111	1101	1100	: Variable length (prefix) code

We still need to know when each code ends → use binary tree

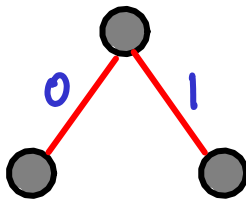


0001001010111011011001110100101
AAA C BA

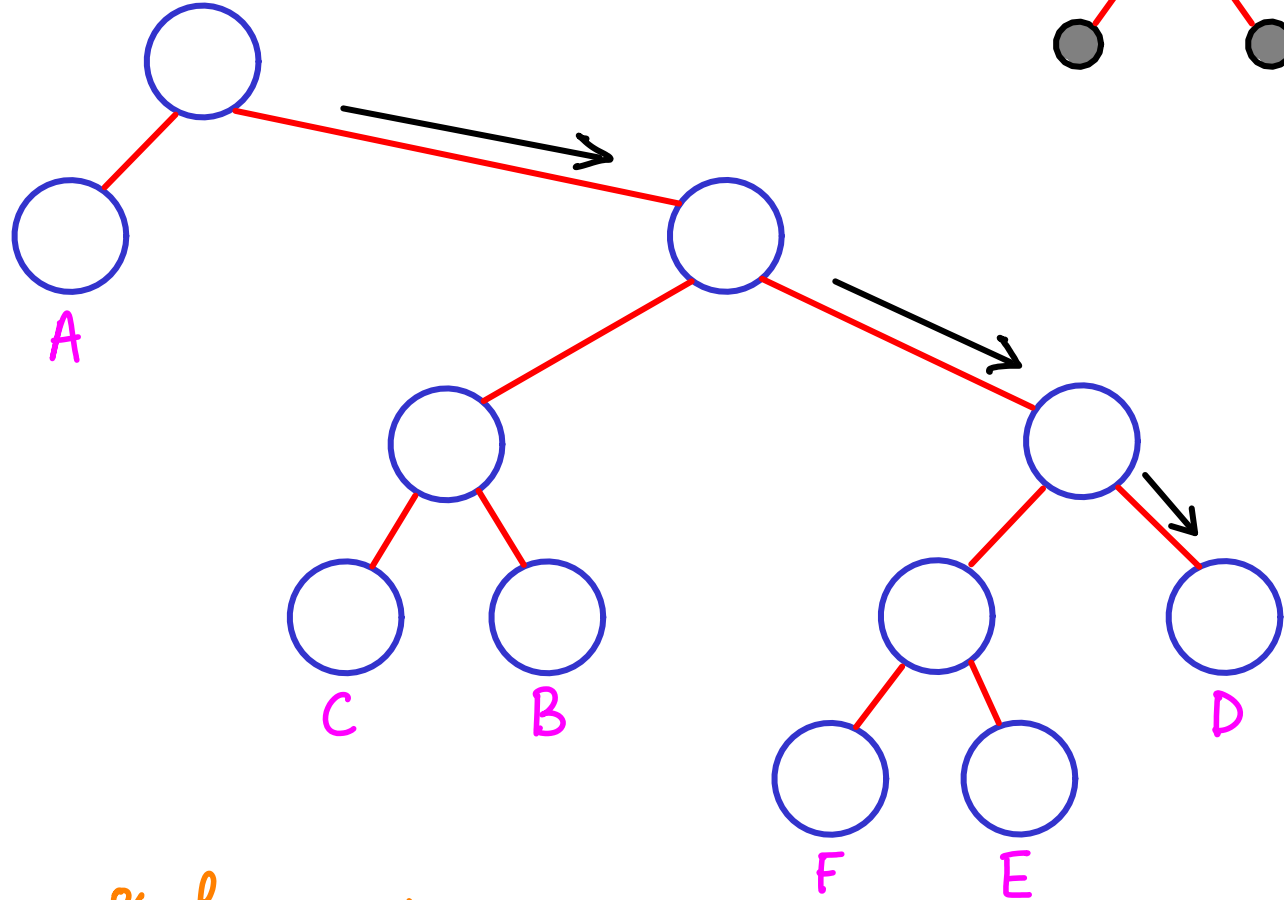
A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies

0	101	100	111	1101	1100	: Variable length (prefix) code
---	-----	-----	-----	------	------	---------------------------------

We still need to know when each code ends → use binary tree



0001001010||01101100||0100101
AAA C BA D etc



A	B	C	D	E	F	
45	13	12	16	9	5	: % frequencies

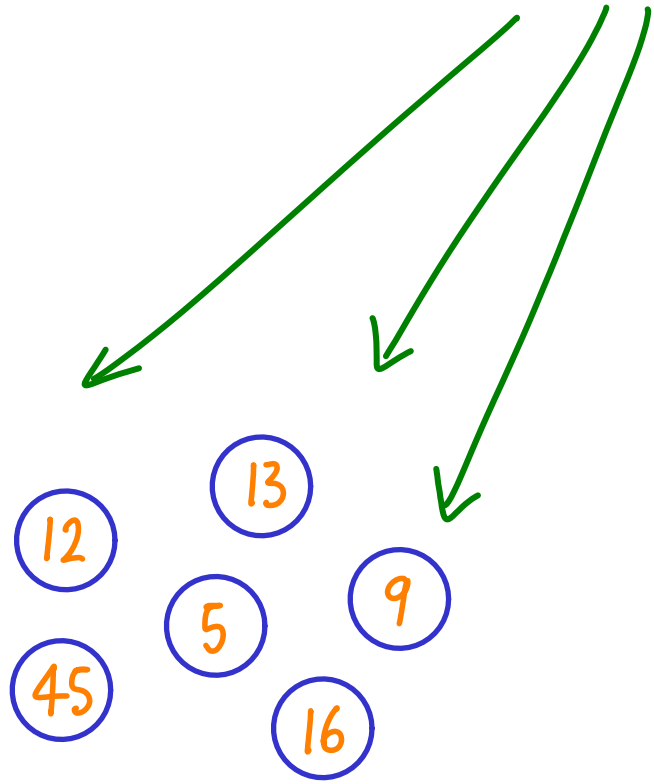
A	B	C	D	E	F	
0	101	100	111	1101	1100	: Variable length (prefix) code

Huffman code

developed in 1951 as a solution to a class assignment

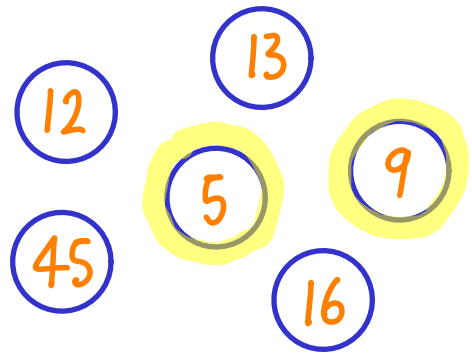
typical compression 20-90%

Huffman code: Make n trees of size 1: one tree per character.



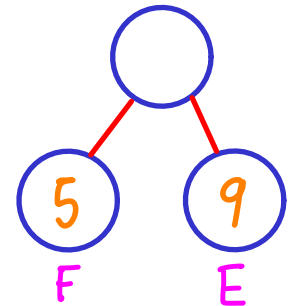
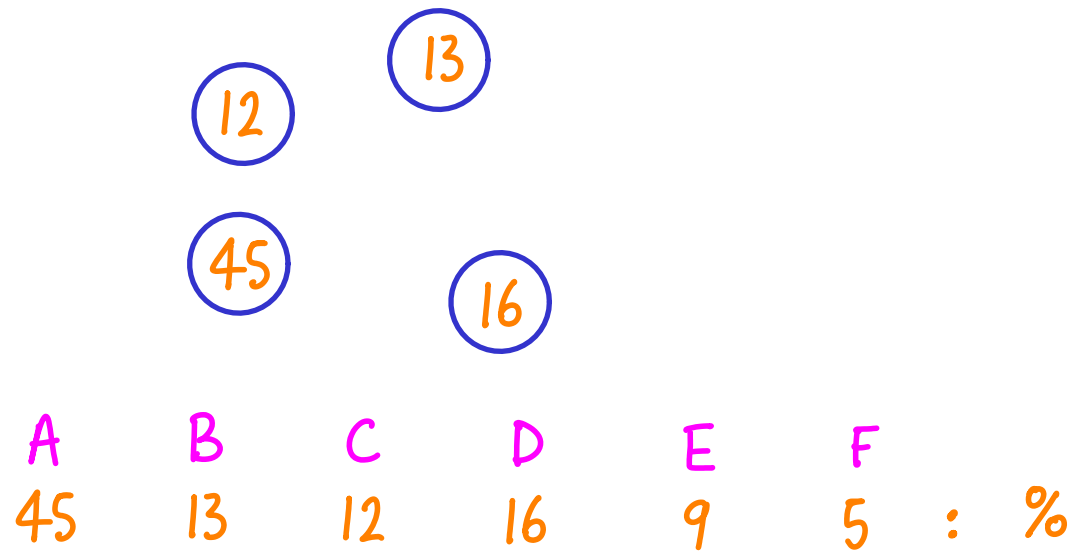
A	B	C	D	E	F	:	%
45	13	12	16	9	5		

Huffman code: Make n trees of size 1: one tree per character.
● The 2 roots with lowest frequencies become siblings.

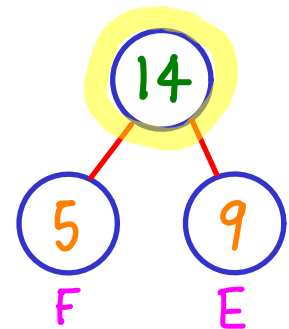
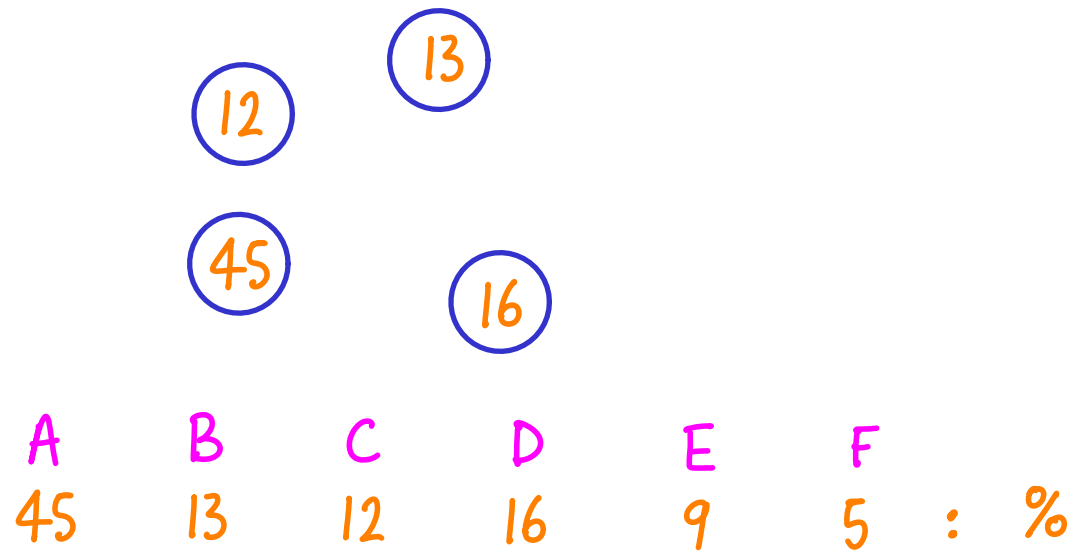


A	B	C	D	E	F	:	%
45	13	12	16	<u>9</u>	<u>5</u>		

Huffman code: Make n trees of size 1: one tree per character.
● The 2 roots with lowest frequencies become siblings.

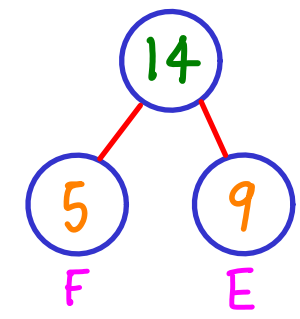
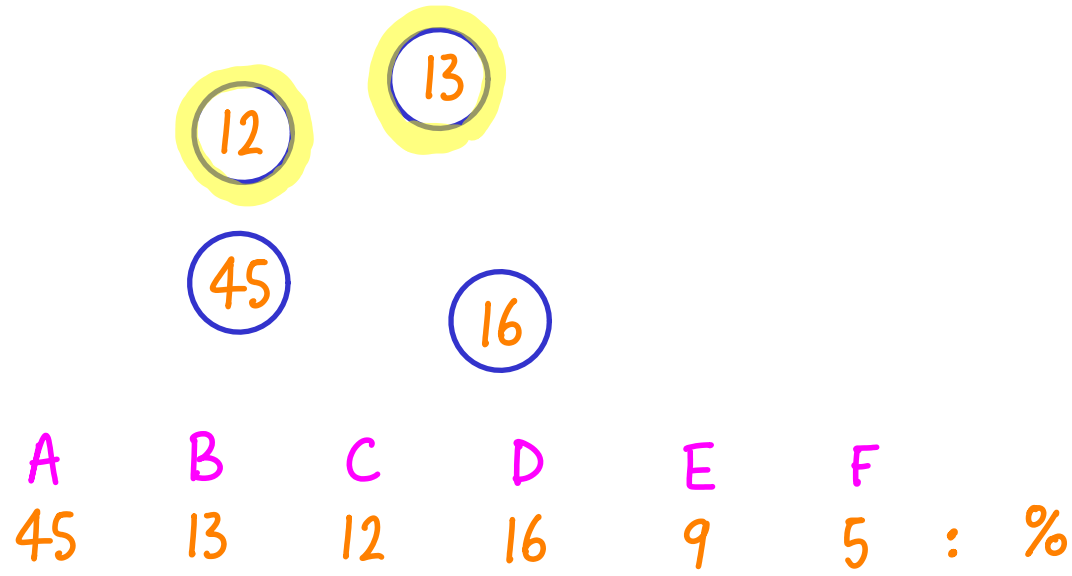


Huffman code: Make n trees of size 1: one tree per character.
The 2 roots with lowest frequencies become siblings.
● Their new parent gets the sum of their frequencies.



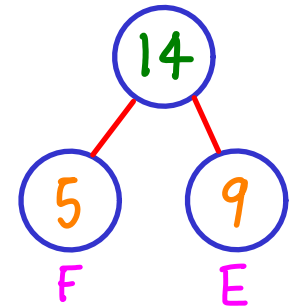
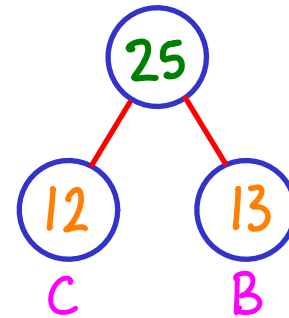
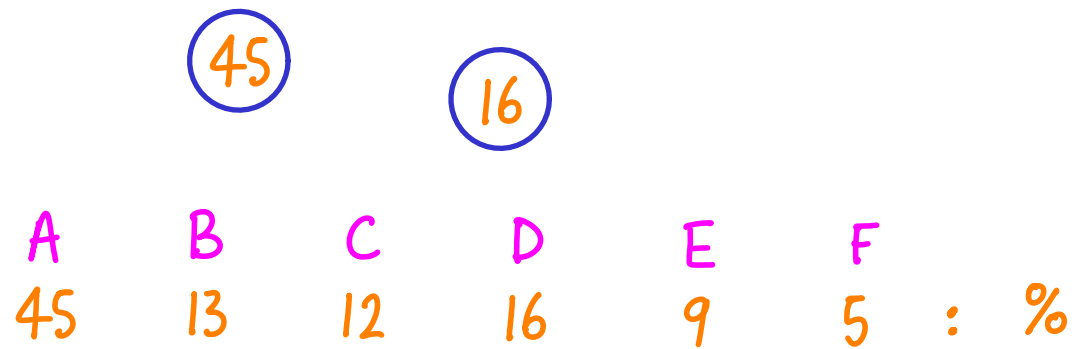
Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.
Their new parent gets the sum of their frequencies.
Repeat



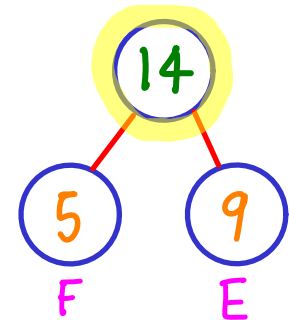
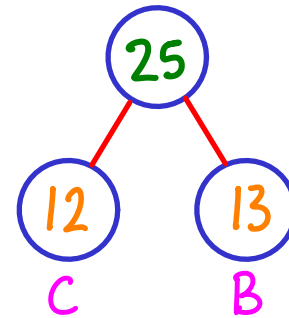
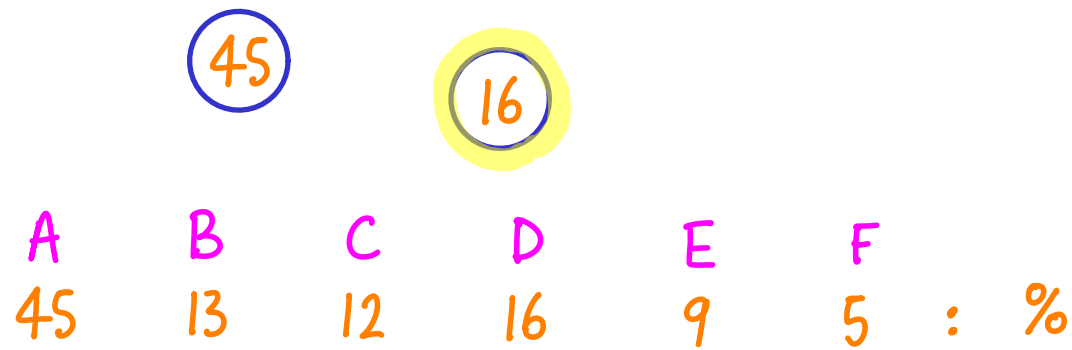
Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.
Their new parent gets the sum of their frequencies.
Repeat



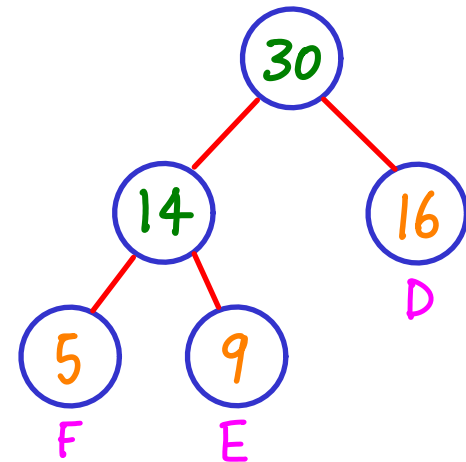
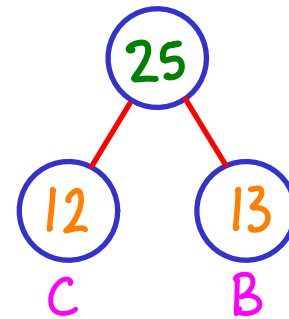
Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.
Their new parent gets the sum of their frequencies.
Repeat



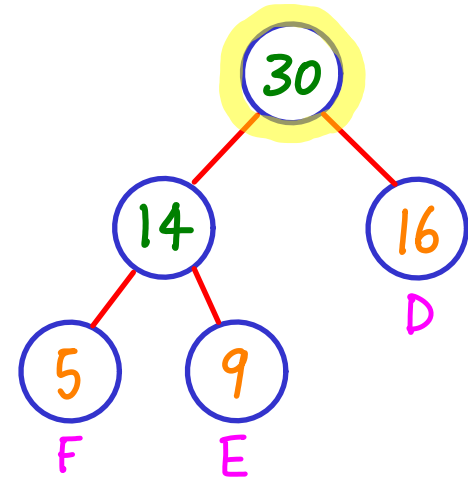
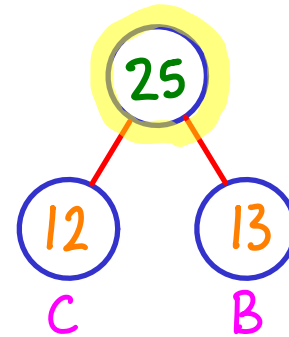
Huffman code: Make n trees of size 1: one tree per character.

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Repeat



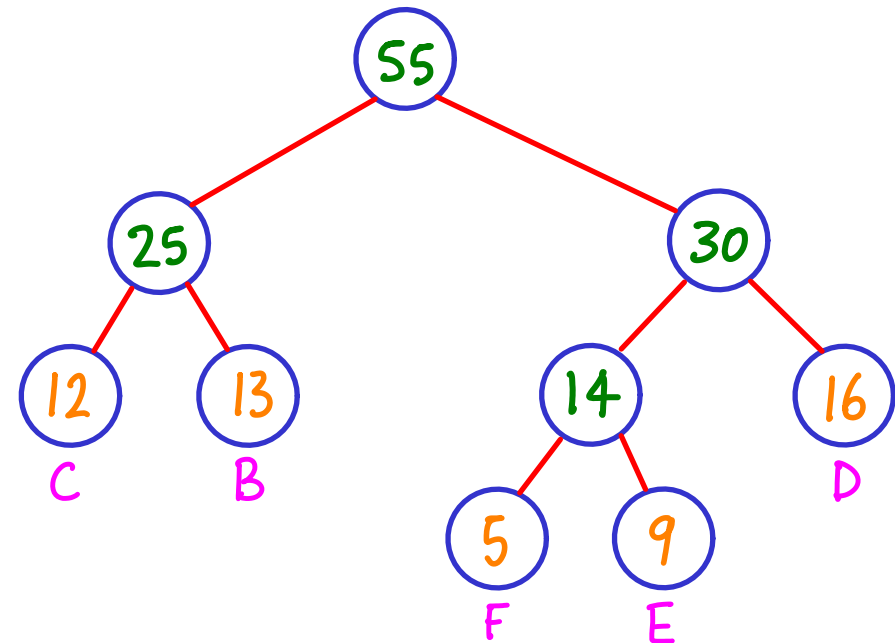
Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.
Their new parent gets the sum of their frequencies.
Repeat



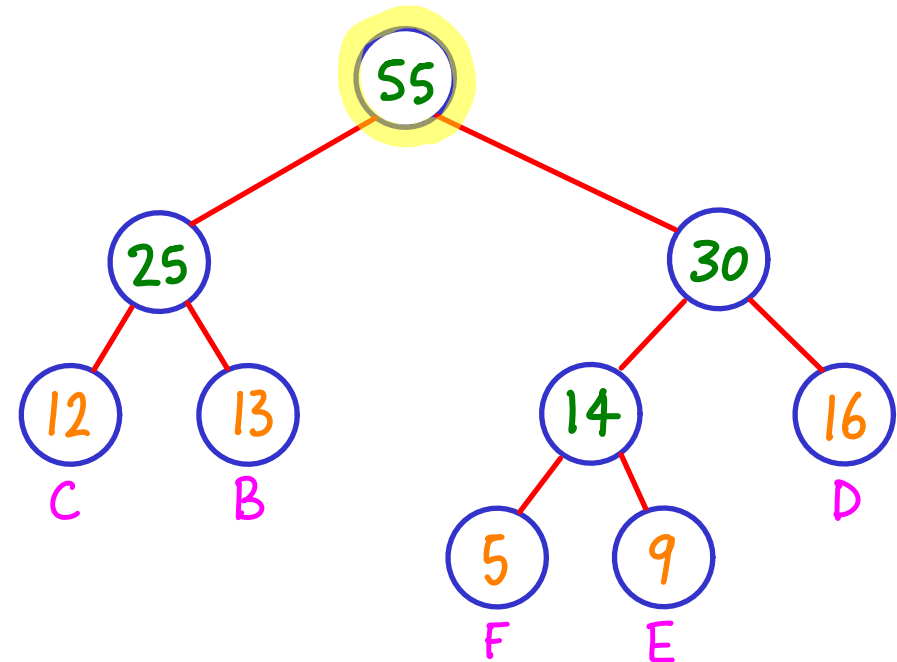
Huffman code: Make n trees of size 1: one tree per character.

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Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.
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Repeat



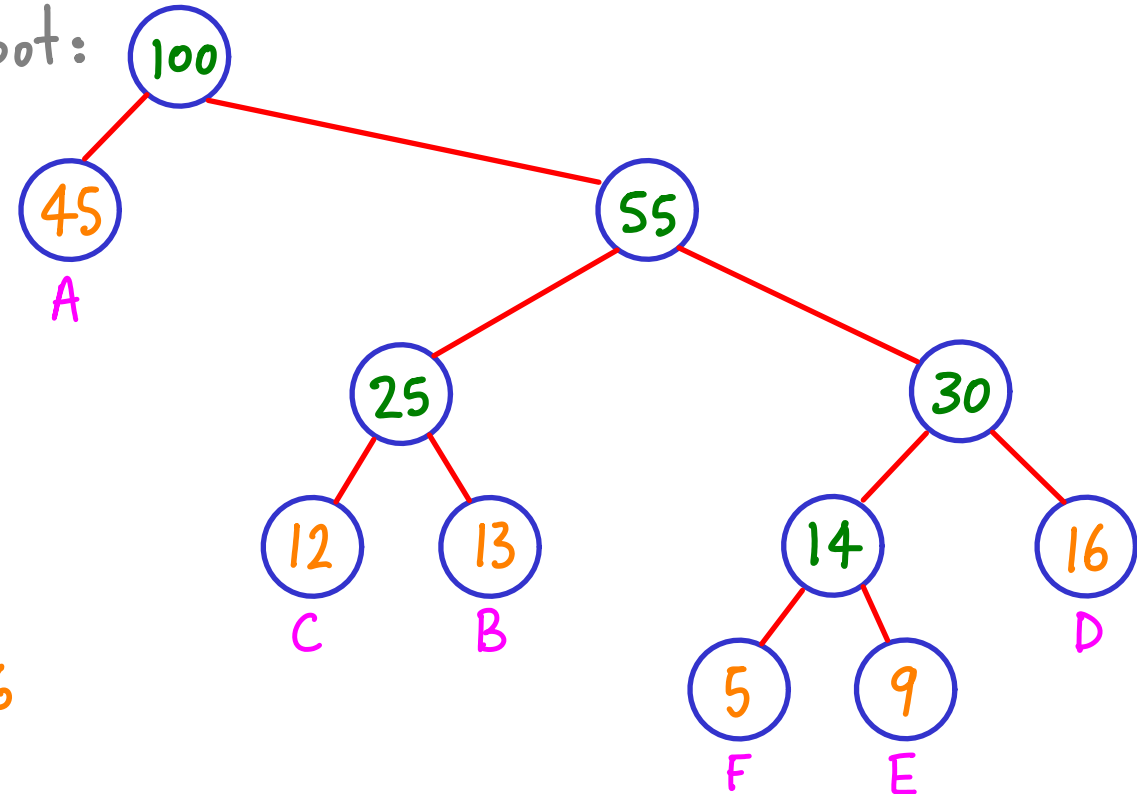
Huffman code: Make n trees of size 1: one tree per character.

→ The 2 roots with lowest frequencies become siblings.

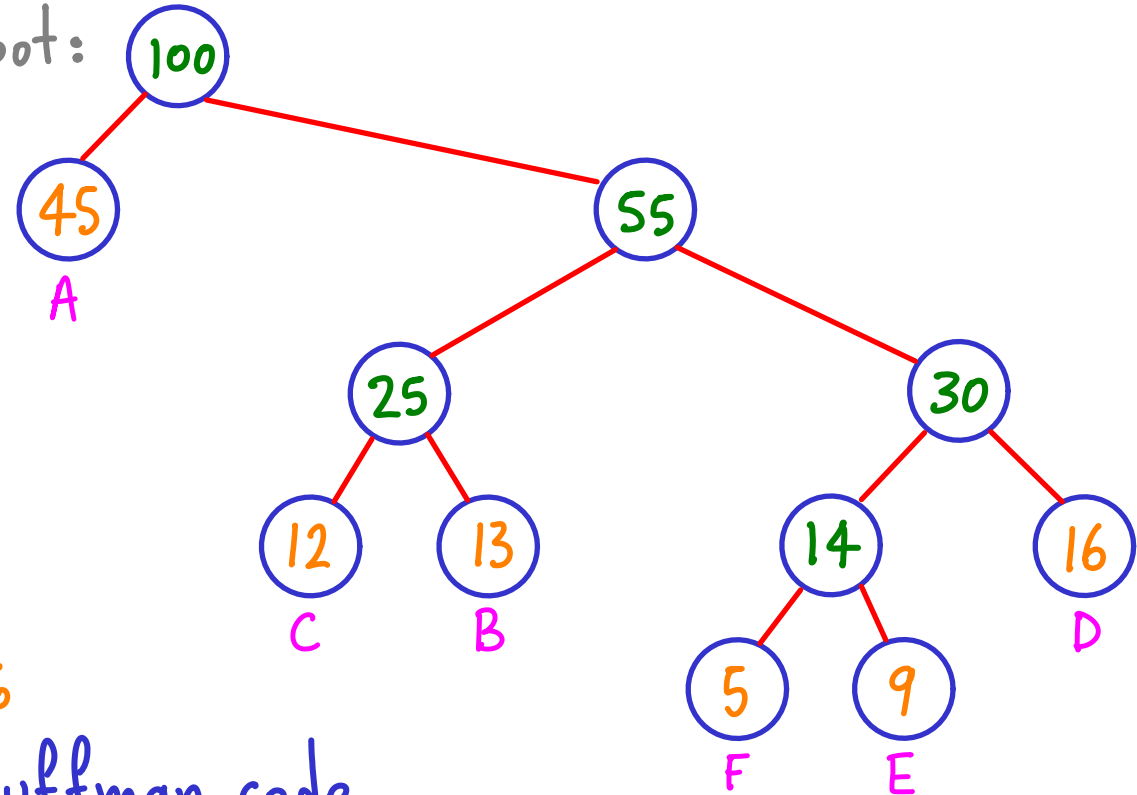
Their new parent gets the sum of their frequencies.

Repeat until 1 root:

A	B	C	D	E	F	:	%
45	13	12	16	9	5		



Huffman code: Make n trees of size 1: one tree per character.
 (greedy algo) → The 2 roots with lowest frequencies become siblings.
 Their new parent gets the sum of their frequencies.
 Repeat until 1 root:



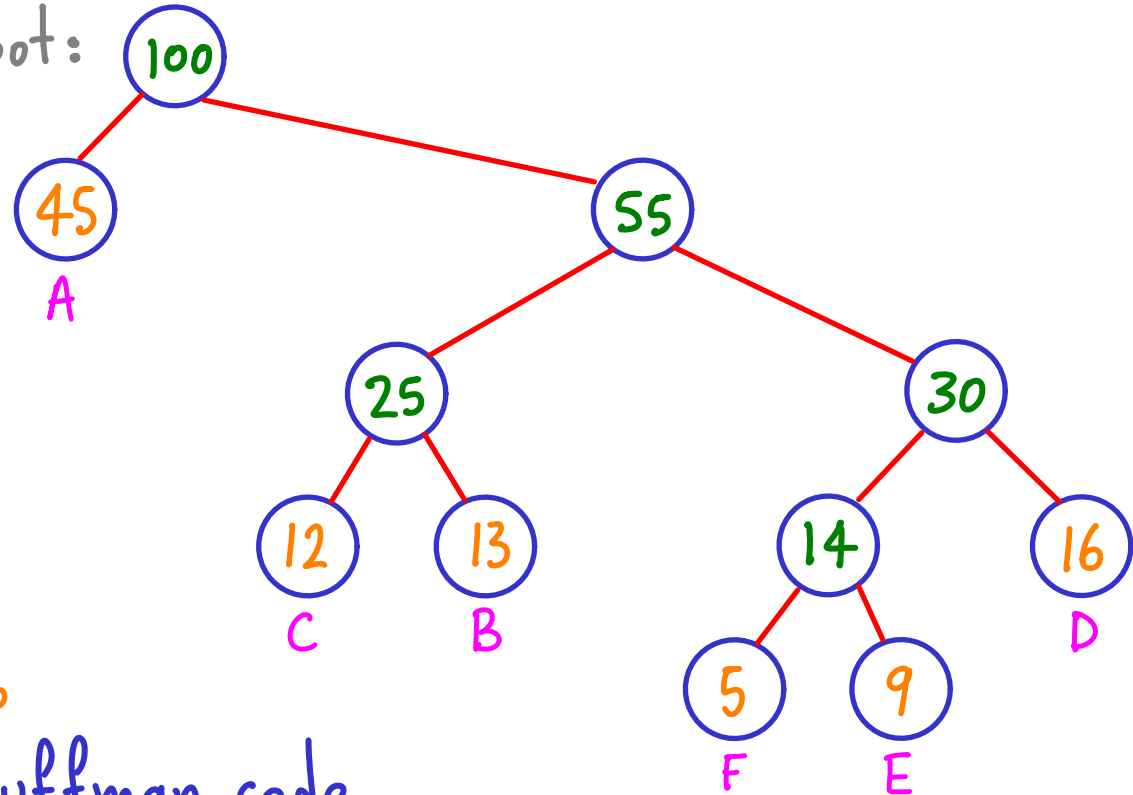
A	B	C	D	E	F	:	%
45	13	12	16	9	5	:	%
0	101	100	111	1101	1100	:	Huffman code

Huffman code: Make n trees of size 1: one tree per character.

The 2 roots with lowest frequencies become siblings.
 Their new parent gets the sum of their frequencies.

Repeat until 1 root:

- Optimal! to show
- Not unique
 - arbitrary sibling order
 - weights could be equal



A	B	C	D	E	F	
45	13	12	16	9	5	: %
0	101	100	111	1101	1100	: Huffman code

Huffman code: time to build?

Huffman code: easy to implement in $O(n \log n)$ time & $\Theta(n)$ space

→ priority queue to identify 2 lowest frequencies

initialize with n chars : $\Theta(n)$ → extract twice, insert new parent node : $O(\log n)$ $n-1$ rounds

Huffman code: easy to implement in $O(n \log n)$ time & $\Theta(n)$ space

→ priority queue to identify 2 lowest frequencies

initialize with n chars : $\Theta(n)$ → extract twice, insert new parent node : $O(\log n)$ $n-1$ rounds

$O(n \log \log n)$ possible, using a van Emde Boas tree
CLRS ch. 20

Huffman code: Not unique

Must store representation, like BST, otherwise decoder can't work.

↳ costs extra space ::

OPTIMALITY

$$\text{minimize } \sum_{\text{all chars}} (\text{freq.} \cdot \text{bit length}) = \sum (\text{freq.} \cdot \text{node depth})$$

- Not optimal number of bits in general (more later)
- We get OPT assuming one codeword per character

OPTIMALITY] minimize $\sum_{\text{all chars}} (\text{freq.} \cdot \text{bit length}) = \sum (\text{freq.} \cdot \text{node depth})$

Claim: there is always a prefix code that achieves OPT

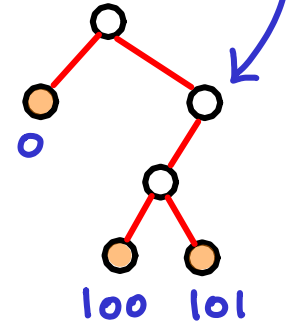
- CLRS skips the proof
- See Sayood 2.4.3 : 3-page proof

OPTIMALITY

$$\text{minimize } \sum_{\text{all chars}} (\text{freq.} \cdot \text{bit length}) = \sum (\text{freq.} \cdot \text{node depth})$$

Claim 2:

OPT tree can't have an internal node w/ 1 child



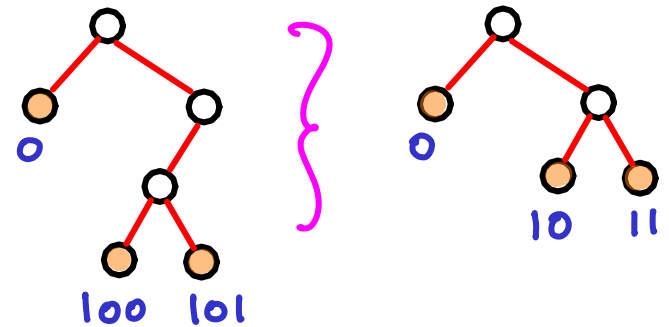
OPTIMALITY

$$\text{minimize } \sum_{\text{all chars}} (\text{freq.} \cdot \text{bit length}) = \sum (\text{freq.} \cdot \text{node depth})$$

Claim 2:

OPT tree can't have an internal node w/ 1 child

↪ contract, improve Σ , still a prefix code = contradiction



OPTIMALITY] minimize $\sum (\text{freq.} \cdot \text{node depth})$

✓ OPT tree can't have an internal node w/ 1 child

Claim 3:

$\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level

OPTIMALITY

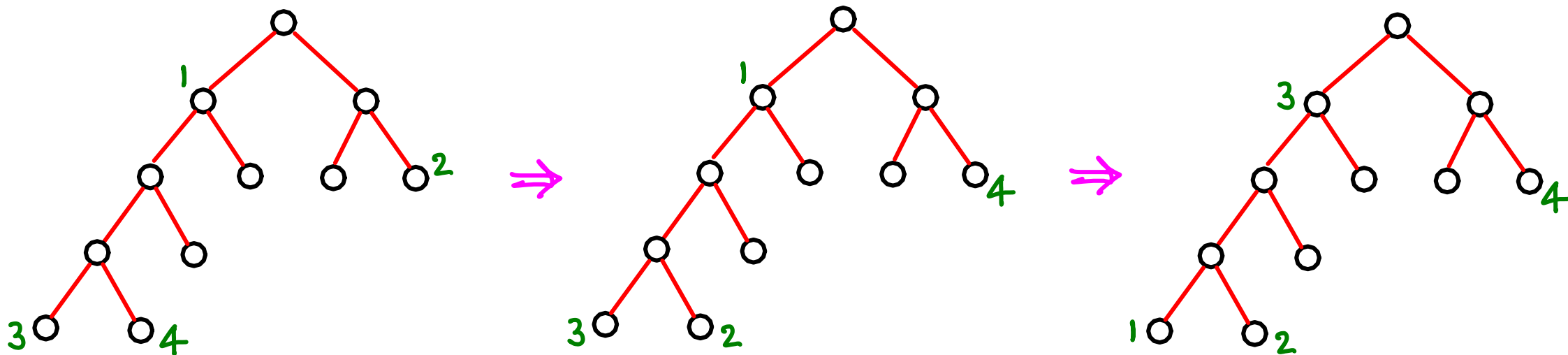
minimize $\sum (\text{freq.} \cdot \text{node depth})$

✓ OPT tree can't have an internal node w/ 1 child

Claim 3:

$\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level

↪ by contradiction: take "OPT", swap nodes, improve Σ



OPTIMALITY] minimize $\sum (\text{freq.} \cdot \text{node depth})$

✓ OPT tree can't have an internal node w/ 1 child

✓ $\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level

OPTIMALITY

minimize $\sum (\text{freq.} \cdot \text{node depth})$

- ✓ OPT tree can't have an internal node w/ 1 child
- ✓ $\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level

Last claim:

if we merge 2 lowest frequencies x, y in tree T
and new tree T' is OPT'
then T was OPT.

OPTIMALITY] minimize $\sum (\text{freq.} \cdot \text{node depth})$

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Last claim:

if we merge 2 lowest frequencies x, y in tree T
and new tree T' is OPT' (node z : $x.\text{freq} + y.\text{freq}$)

then T was OPT.

Weight diff for $T, T' = 1 \cdot (x.\text{freq} + y.\text{freq})$ // all other nodes unchanged

OPTIMALITY

minimize $\sum (\text{freq.} \cdot \text{node depth})$

✓ OPT tree can't have an internal node w/ 1 child

✓ $\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level *

Last claim:

if we merge 2 lowest frequencies x, y in tree T
and new tree T' is OPT' (node $z : x.\text{freq} + y.\text{freq}$)

then T was OPT.

Weight diff for $T, T' = 1 \cdot (x.\text{freq} + y.\text{freq})$ // all other nodes unchanged

Same for diff of any OPT tree that beats T : x, y must be siblings* so
we can contract them & get a tree that beats T' : contradiction

OPTIMALITY] minimize $\sum (\text{freq.} \cdot \text{node depth})$

- ✓ OPT tree can't have an internal node w/ 1 child
- ✓ $\exists$ OPT tree w/ 2 lowest frequencies as siblings at lowest level
- ✓ if we merge 2 lowest frequencies x, y in tree T and new tree T' is optimal then T was OPT.
- ↳ merge 2 smallest, apply induction, uncontract $\rightarrow$ get OPT
- ↳ incremental construction is equivalent



A few words about extensions & other ideas

Huffman code with presorted frequencies

- use 2 queues instead of 1 priority queue

Huffman code with presorted frequencies

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 - place frequencies (single nodes) in queue A
 - queue B holds merged nodes

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 - merge two smallest frequencies & insert in B, repeat
(min two elements are always at top 2+2 positions)

Huffman code with presorted frequencies

$O(n)$ time

- use 2 queues instead of 1 priority queue
 - place frequencies (single nodes) in queue A
 - queue B holds merged nodes
 - merge two smallest frequencies & insert in B, repeat
(min two elements are always at top 2+2 positions)
- proof of correctness is straightforward

Huffman code with minimum codeword length (variance)

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- prioritize merging shorter trees
- sort all frequencies, place in queue A
- use queue B as for presorted data

Huffman code with minimum codeword length (variance)

- prioritize merging shorter trees
 - sort all frequencies, place in queue A
 - use queue B as for presorted data
 - when dequeuing min two elements, prioritize first queue if tied

Decompressing options

(affects storage / transmission)

- store frequencies (could be really bad)
- store tree
- use "canonical tree"
-

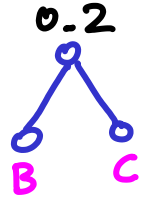
adaptive Huffman coding

A B C

0.8 0.02 0.18

A B C

0.8 0.02 0.18



A

B

C

0

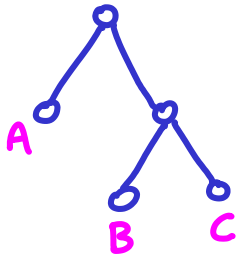
11

10

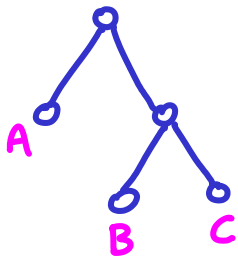
0.8

0.02

0.18



A	B	C
0	11	10
0.8	0.02	0.18



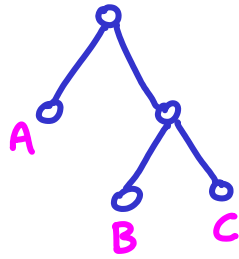
$$0.8 + 0.04 + 0.36 = \underline{1.2}$$

Improve this by encoding blocks

A B C

0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

AA 0.64

AB 0.016

AC 0.144

BA 0.016

BB 0.0004

BC 0.0036

CA 0.144

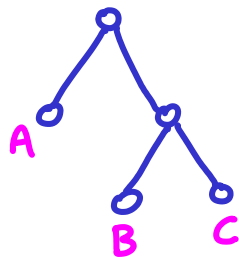
CB 0.0036

CC 0.0324

A B C

0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

AA 0.64

AB 0.016

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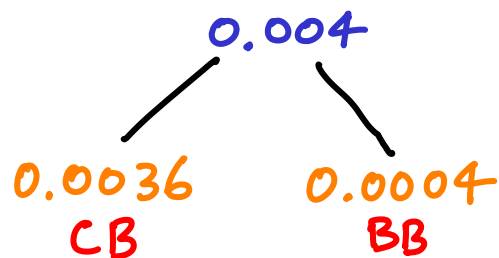
BB 0.0004 •

BC 0.0036

CA 0.144

CB 0.0036 •

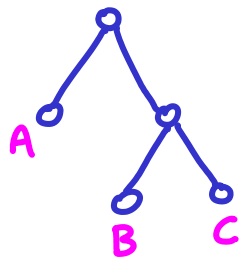
CC 0.0324



A B C

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0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

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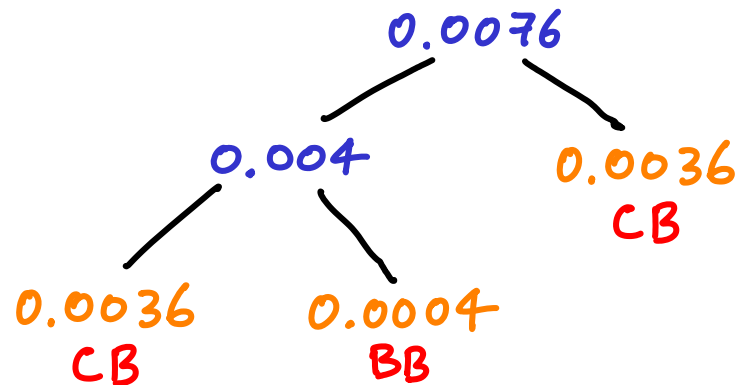
BB 0.0004 •

BC 0.0036 •

CA 0.144

CB 0.0036 ⊙

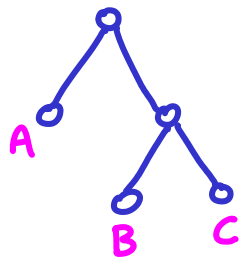
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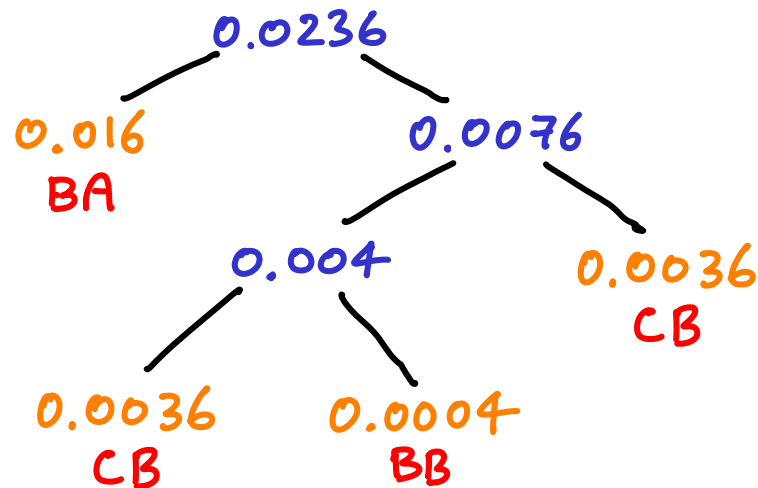
BB 0.0004 •

BC 0.0036 •

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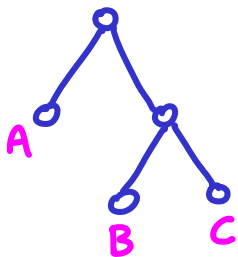
CC 0.0324



A B C

0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

AA 0.64

AB 0.016 ⊖

AC 0.144

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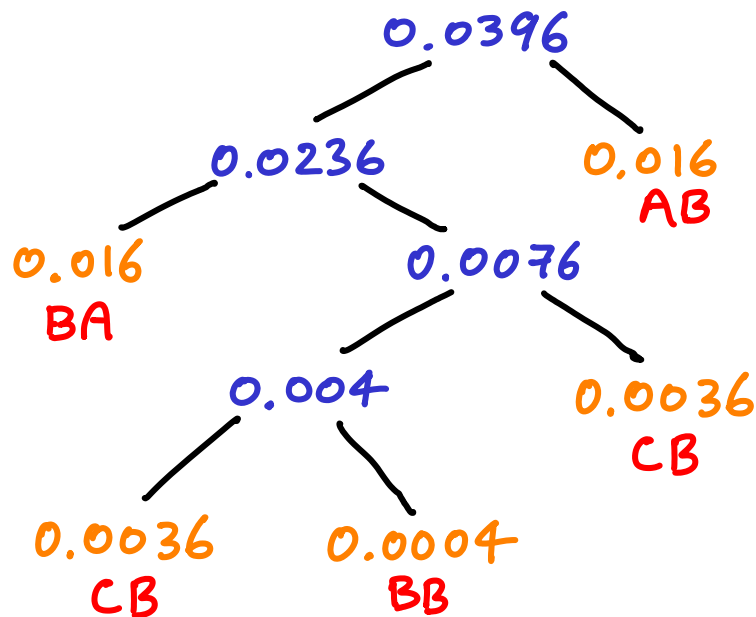
BB 0.0004 •

BC 0.0036 •

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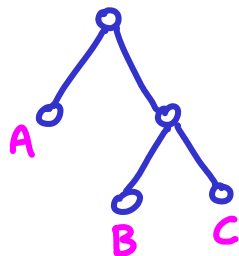
CC 0.0324



A B C

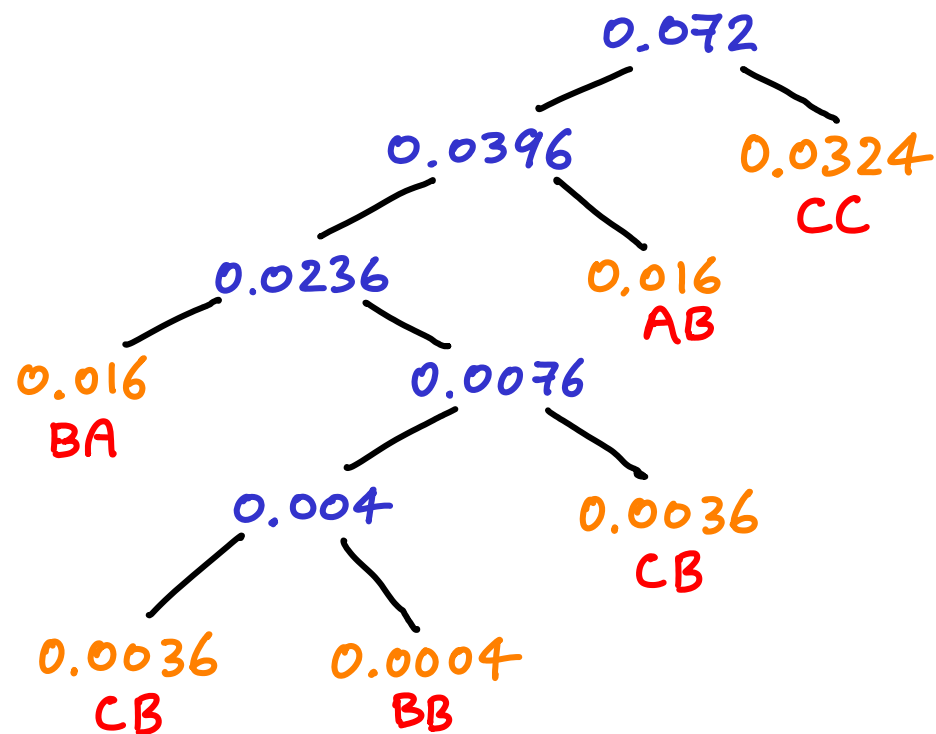
0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

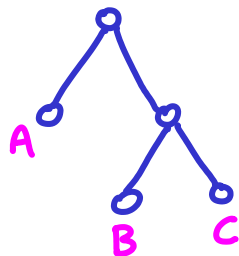
AA	0.64	
AB	0.016	•
AC	0.144	
BA	0.016	•
BB	0.0004	•
BC	0.0036	•
CA	0.144	
CB	0.0036	•
CC	0.0324	⊙



A B C

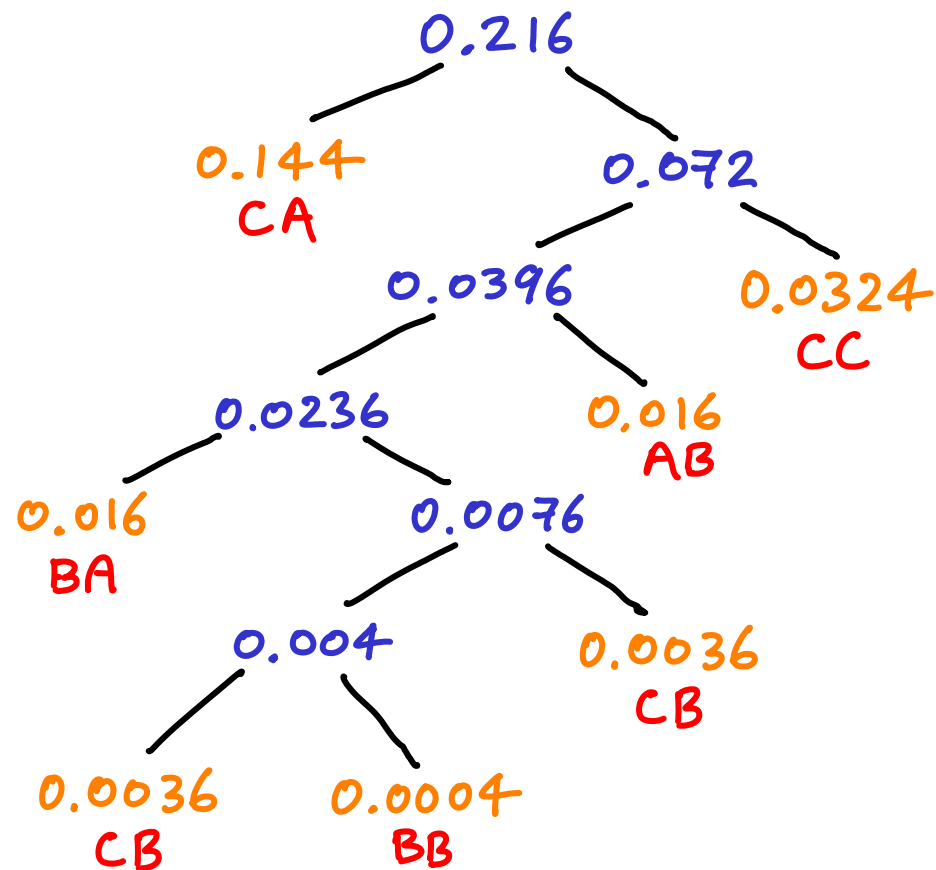
0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

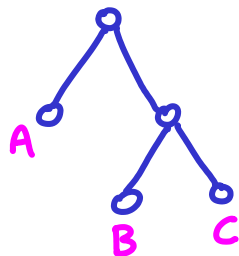
AA	0.64	
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BB	0.0004	•
BC	0.0036	•
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$$0.8 + 0.04 + 0.36 = 1.2$$

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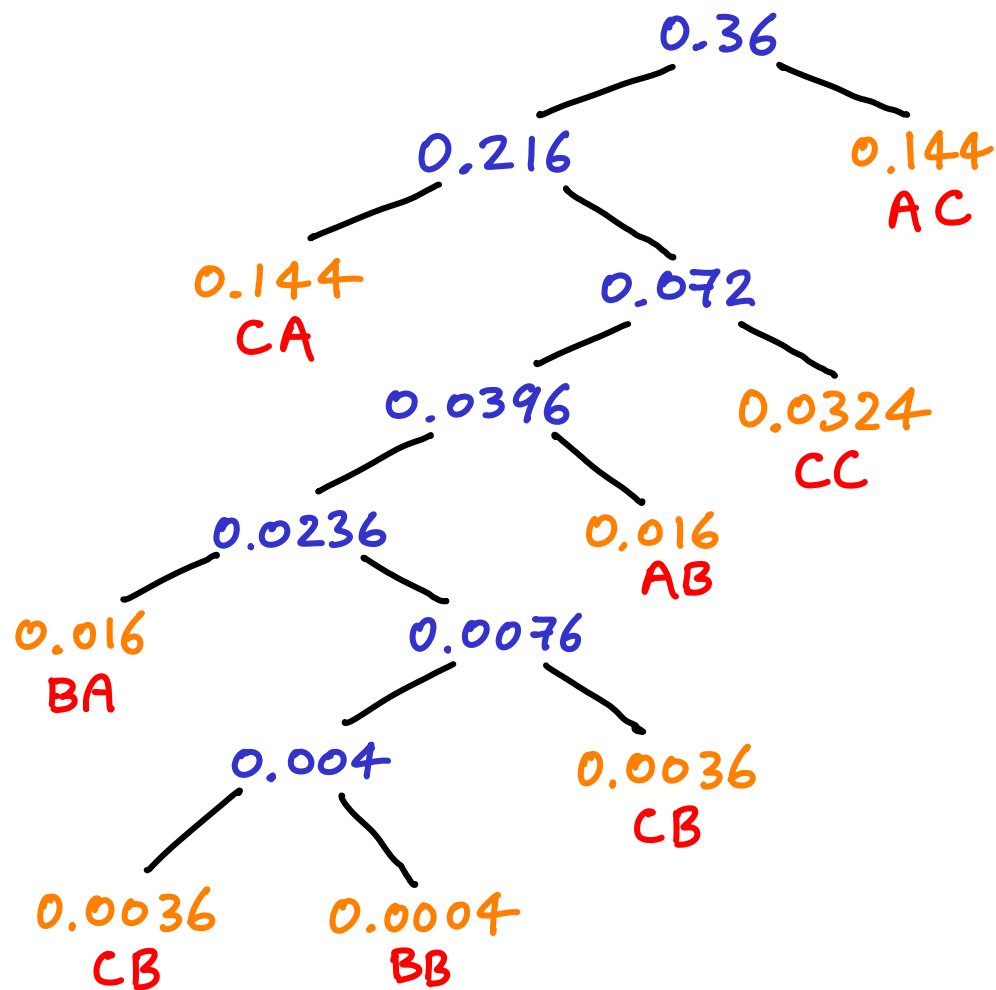
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CA 0.144 .

CB 0.0036 .

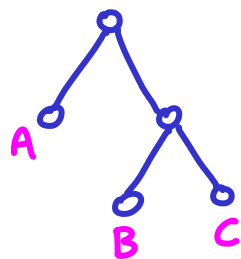
CC 0.0324 .



A B C

0 11 10

0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

AA 0.64 ⊙

AB 0.016 •

AC 0.144 •

BA 0.016 •

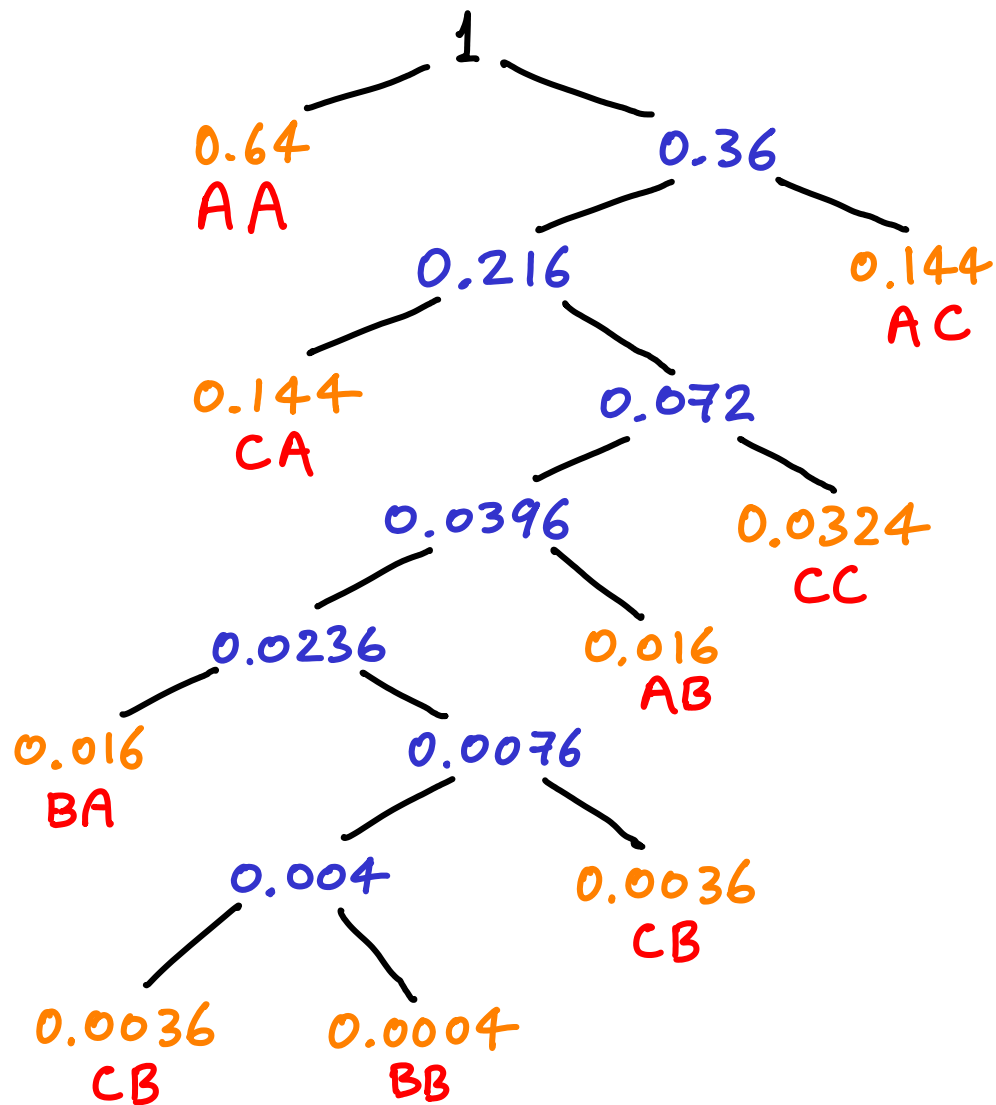
BB 0.0004 •

BC 0.0036 •

CA 0.144 •

CB 0.0036 •

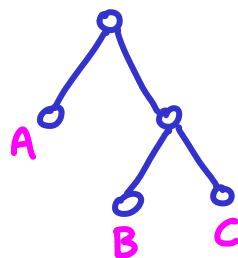
CC 0.0324 •



A B C

0 11 10

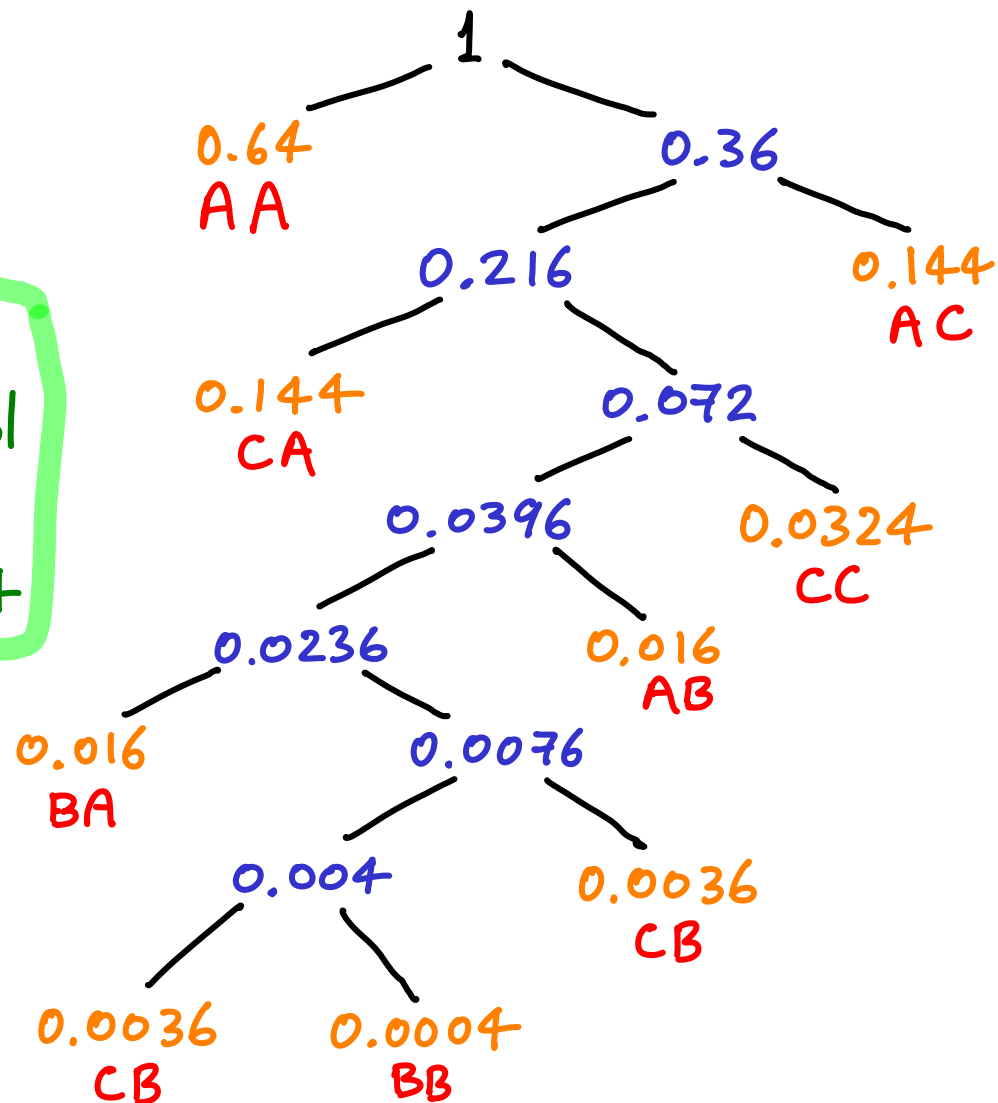
0.8 0.02 0.18



$$0.8 + 0.04 + 0.36 = 1.2$$

AA	0.64	0
AB	0.016	1 0 1 0 1
AC	0.144	1 1
BA	0.016	1 0 1 0 0 0
BB	0.0004	1 0 1 0 0 1 0 1
BC	0.0036	1 0 1 0 0 1 1
CA	0.144	1 0 0
CB	0.0036	1 0 1 0 0 1 0 0
CC	0.0324	1 0 1 1

1.7228
bits / symbol
↓
/2 = 0.8614



AAA B AAB AAB AAB AAA
00 11 01 01 01 00

AAA	00
AAB	01
AB	10
B	11

} 2-bit Tunstall code
(fixed length)

errors do not propagate (Huffman: one incorrect bit can ruin everything)

