

# An Introduction to Algorithms

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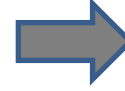
Intro



Complexity



Data Structure



Trees



Hash Functions



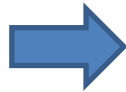
Sorting



Dynamic  
Programming



Greedy Algorithm



Misc Graph/Tree  
Algorithms

**"GREED IS SO DESTRUCTIVE. IT DESTROYS  
EVERYTHING."**

**EARTHA KITT**

© 1991 Litterack Quotes

**I THINK GREED  
SOMETIMES GETS  
THE BEST OF  
EVERYBODY.**

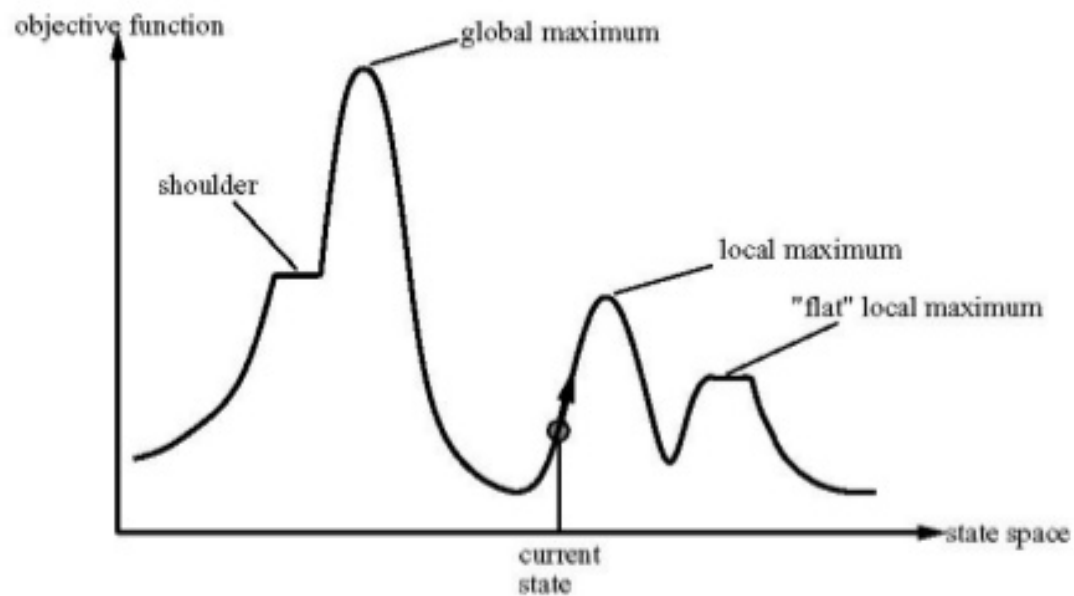
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# Optimization Problems

- For most optimization problems you want to find, not just a solution, but the best solution.
- A greedy algorithm sometimes works well for optimization problems. It works in phases. At each phase:
  - You take the best you can get right now, without regard for future consequences.
  - You hope that by choosing a local optimum at each step, you will end up at a global optimum.

## Hill Climbing – Some Problems



# Example: Counting Money

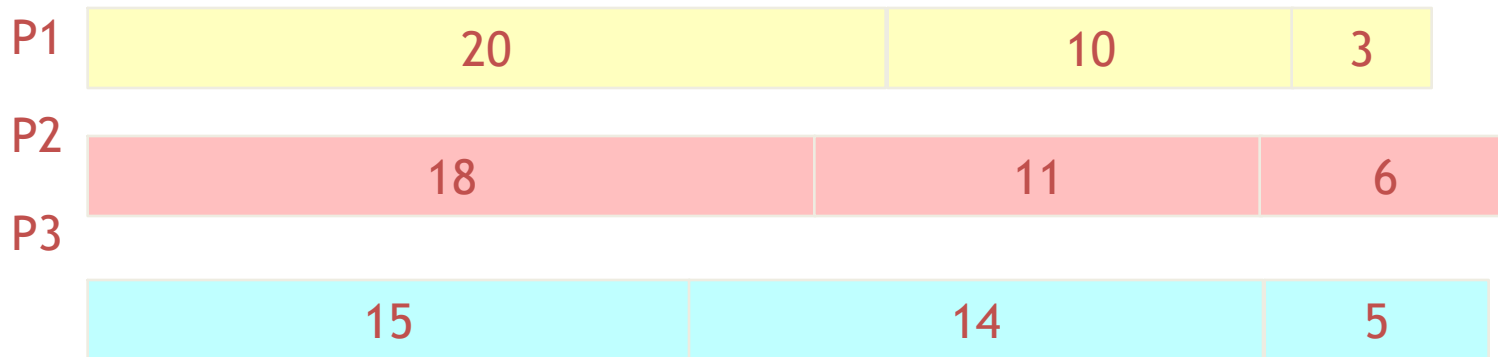
- Suppose you want to count out a certain amount of money, using the fewest possible bills and coins
- A greedy algorithm to do this would be:  
At each step, take the largest possible bill or coin that does not overshoot
  - Example: To make \$6.39, you can choose:
    - a \$5 bill
    - a \$1 bill, to make \$6
    - a 25¢ coin, to make \$6.25
    - A 10¢ coin, to make \$6.35
    - four 1¢ coins, to make \$6.39
- For US money, the greedy algorithm always gives the optimum solution

# Greedy Algorithm Failure

- In some (fictional) monetary system, “krons” come in 1 kron, 7 kron, and 10 kron coins
- Using a greedy algorithm to count out 15 krons, you would get
  - A 10 kron piece
  - Five 1 kron pieces, for a total of 15 krons
  - This requires six coins
- A better solution would be to use two 7 kron pieces and one 1 kron piece
  - This only requires three coins
- The greedy algorithm results in a solution, but not in an optimal solution

# A Scheduling Problem

- You have to run nine jobs, with running times of 3, 5, 6, 10, 11, 14, 15, 18, and 20 minutes.
- You have three processors on which you can run these jobs.
- You decide to do the longest-running jobs first, on whatever processor is available.

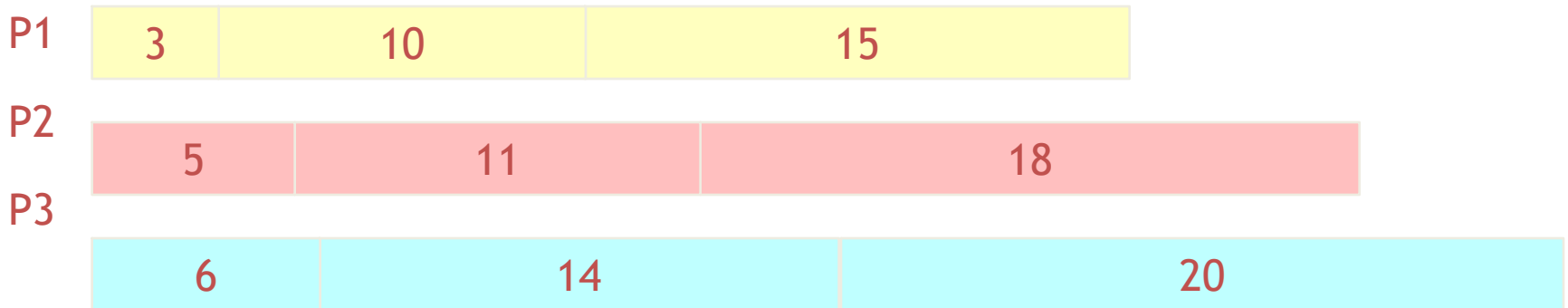


- Time to completion:  $18 + 11 + 6 = 35$  minutes
- This solution isn't bad, but we might be able to do better



# Another Approach

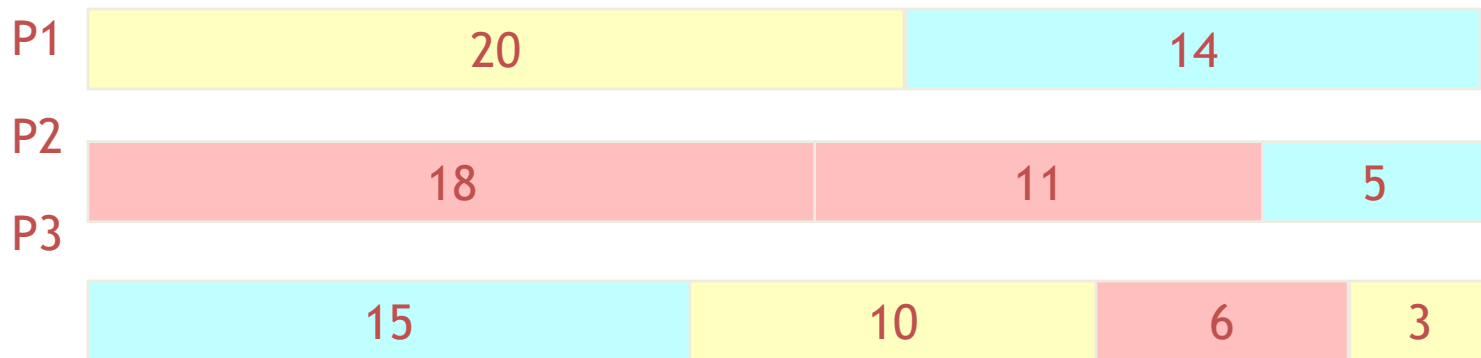
- What would be the result if you ran the shortest job first?
- Again, the running times are 3, 5, 6, 10, 11, 14, 15, 18, and 20 minutes



- That wasn't such a good idea; time to completion is now  $6 + 14 + 20 = 40$  minutes
- Note, however, that the greedy algorithm itself is fast
  - All we had to do at each stage was pick the minimum or maximum

# An Optimum Solution

- Better solutions do exist:



- How do we find such a solution?
  - One way: Try all possible assignments of jobs to processors
  - Unfortunately, this approach can take exponential time

# Compression

- Definition
  - Reduce size of data  
(number of bits needed to represent data)
- Benefits
  - Reduce storage needed
  - Reduce transmission cost / bandwidth

# Sources of Compressibility

- Redundancy
  - Recognize repeating patterns
  - Exploit using
    - Dictionary
    - Variable length encoding
- Human perception
  - Less sensitive to some information
  - Can discard less important data

# Types of Compression

- Lossless
  - Preserves all information
  - Exploits redundancy in data
  - Applied to general data
- Lossy
  - May lose some information
  - Exploits redundancy & human perception
  - Applied to audio, image, video

# Effectiveness of Compression

- Metrics
  - Bits per byte (8 bits)
    - 2 bits / byte  $\Rightarrow$   $\frac{1}{4}$  original size
    - 8 bits / byte  $\Rightarrow$  no compression
  - Percentage
    - 75% compression  $\Rightarrow$   $\frac{1}{4}$  original size

# Effectiveness of Compression

- Depends on data
  - Random data  $\Rightarrow$  hard
    - Example: 1001110100  $\Rightarrow$  ?
  - Organized data  $\Rightarrow$  easy
    - Example: 1111111111  $\Rightarrow 1 \times 10$
- Corollary
  - No universally best compression algorithm

# Effectiveness of Compression

- Lossless Compression is not always possible
  - If compression is always possible (alternative view)
    - Compress file (reduce size by 1 bit)
    - Recompress output
    - Repeat (until we can store data with 0 bits)

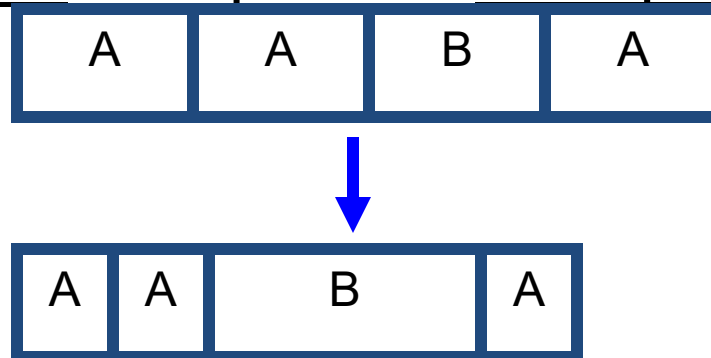


# Lossless Compression Techniques

- LZW (Lempel-Ziv-Welch) compression
  - Build pattern dictionary
  - Replace patterns with index into dictionary
- Run length encoding
  - Find & compress repetitive sequences
- Huffman codes
  - Use variable length codes based on frequency

# Huffman Code

- Approach
  - Variable length encoding of symbols
  - Exploit statistical frequency of symbols
  - Efficient when symbol probabilities vary widely
- Principle
  - Use fewer bits to represent frequent symbols
  - Use more bits to represent infrequent symbols



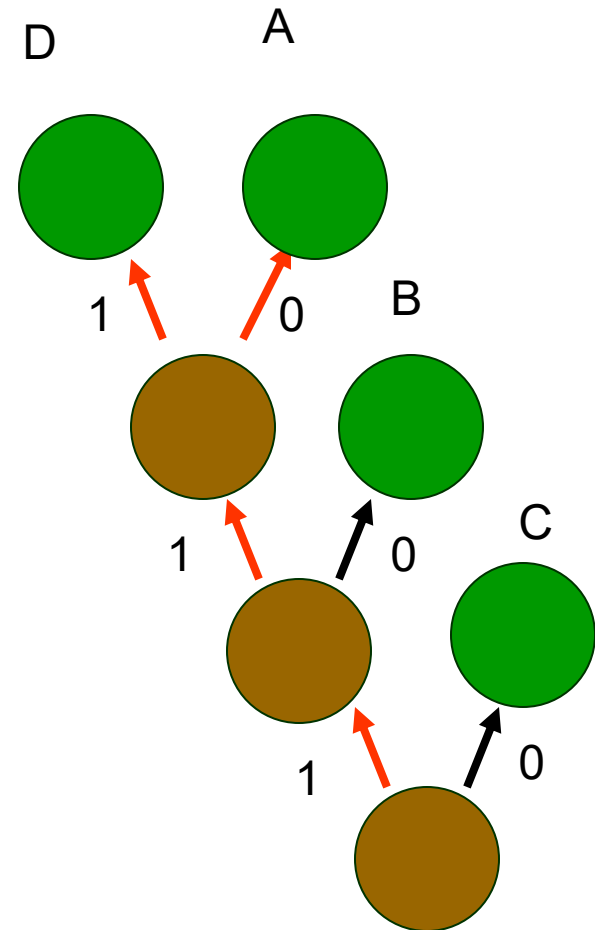
# Huffman Code Example

Symbol	A	B	C	D
Frequency	13%	25%	50%	12%
Original Encoding	00	01	10	11
	2 bits	2 bits	2 bits	2 bits
Huffman Encoding	110	10	0	111
	3 bits	2 bits	1 bit	3 bits

- Expected size
  - Original  $\Rightarrow 1/8 \times 2 + 1/4 \times 2 + 1/2 \times 2 + 1/8 \times 2 = 2$  bits / symbol
  - Huffman  $\Rightarrow 1/8 \times 3 + 1/4 \times 2 + 1/2 \times 1 + 1/8 \times 3 = 1.75$  bits / symbol

# Huffman Code Data Structures

- Binary (Huffman) tree
  - Represents Huffman code
  - Edge  $\Rightarrow$  code (0 or 1)
  - Leaf  $\Rightarrow$  symbol
  - Path to leaf  $\Rightarrow$  encoding
  - Example
    - A = "110", B = "10", C = "0"



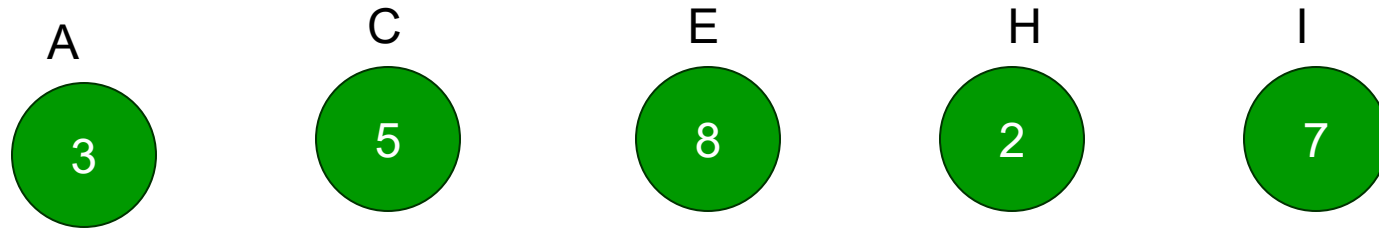
# Huffman Code Algorithm Overview

- Encoding
  - Calculate frequency of symbols in file
  - Create binary tree representing “best” encoding
  - Use binary tree to encode compressed file
    - For each symbol, output path from root to leaf
    - Size of encoding = length of path
  - Save binary tree

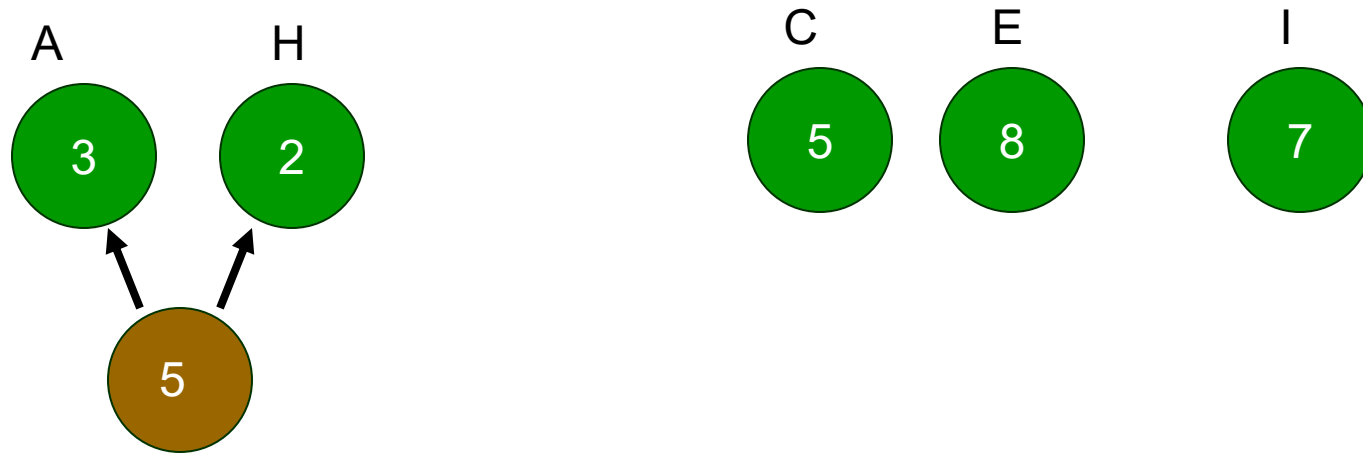
# Huffman Code – Creating Tree

- Algorithm
  - Place each symbol in leaf
    - Weight of leaf = symbol frequency
  - Select two trees L and R (initially leafs)
    - Such that L, R have lowest frequencies in tree
  - Create new (internal) node
    - Left child  $\Rightarrow$  L
    - Right child  $\Rightarrow$  R
    - New frequency  $\Rightarrow$  frequency( L ) + frequency( R )
  - Repeat until all nodes merged into one tree

# Huffman Tree Construction 1

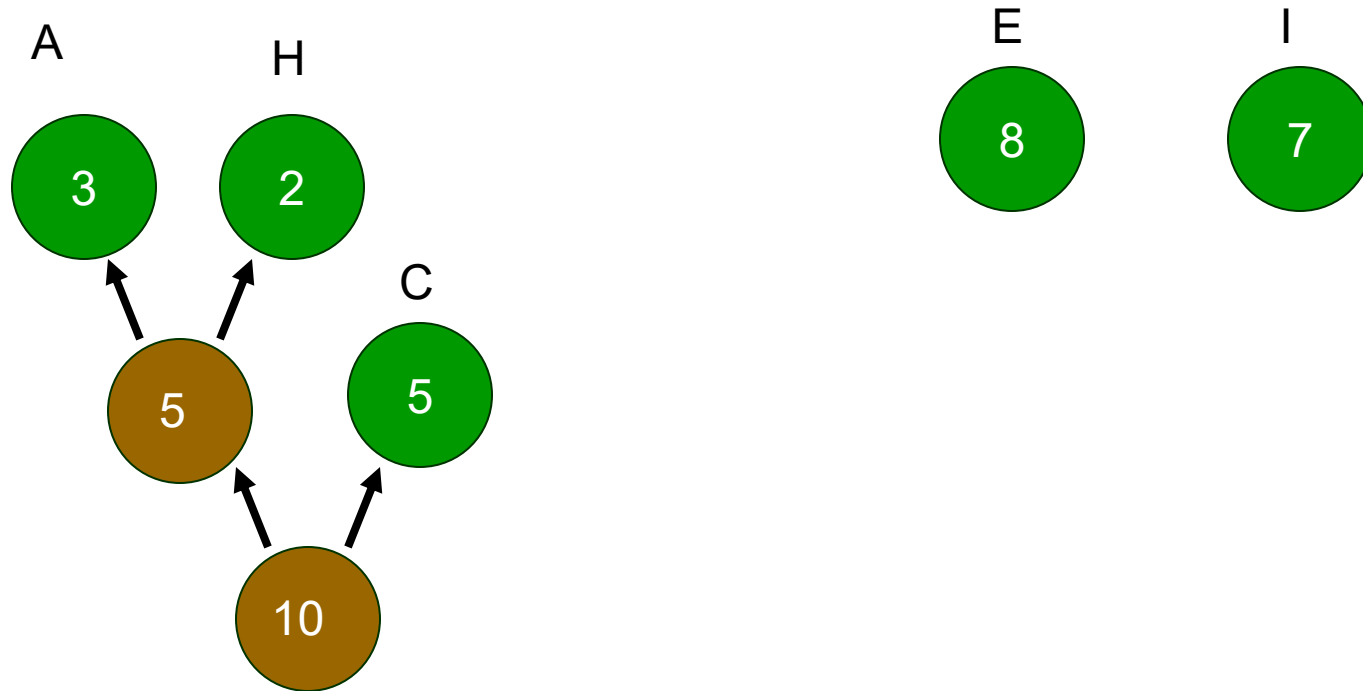


# Huffman Tree Construction 2

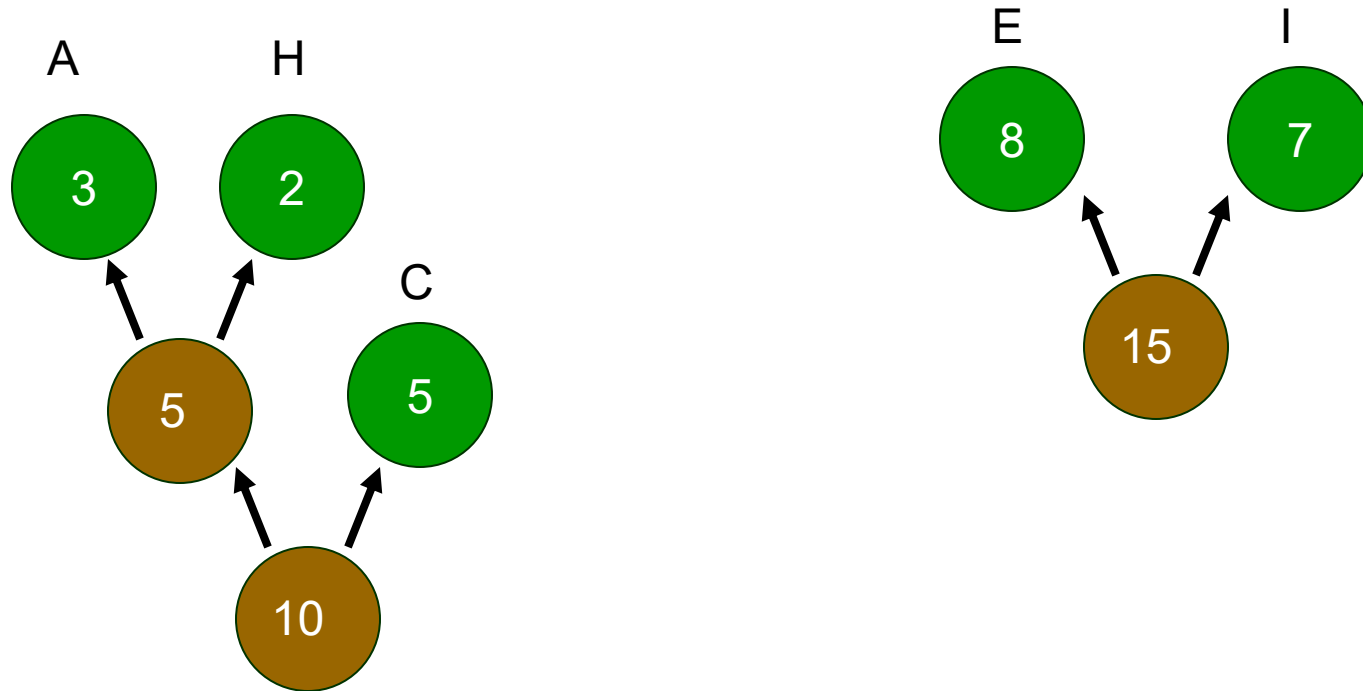




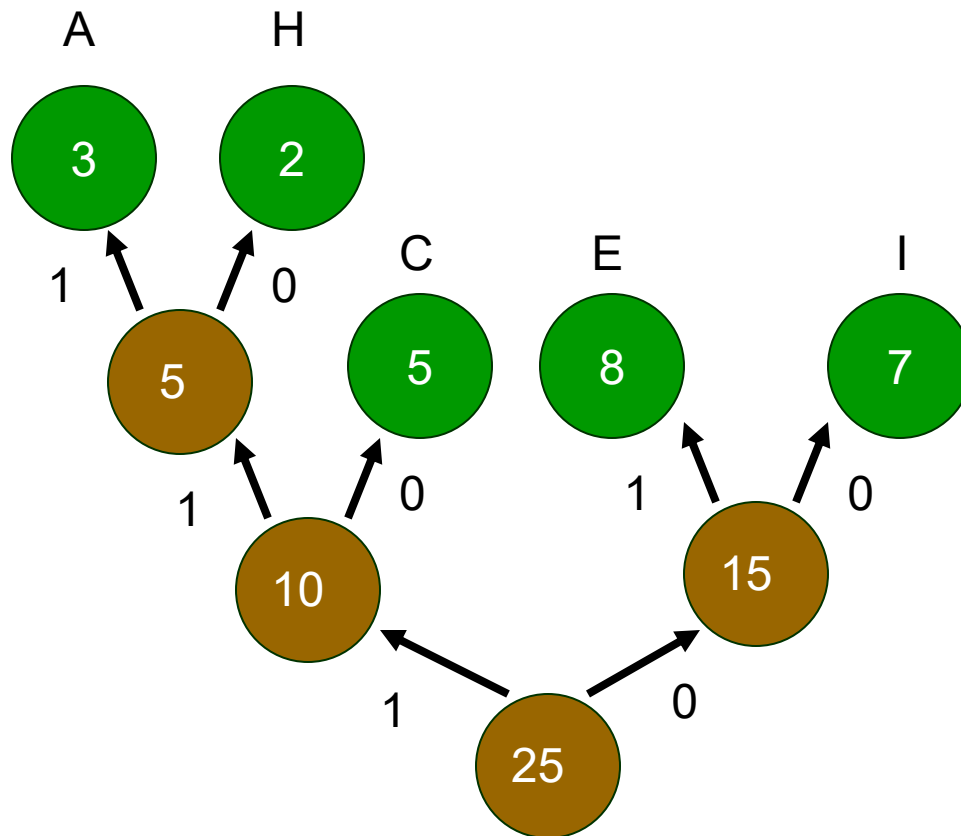
# Huffman Tree Construction 3



# Huffman Tree Construction 4



# Huffman Tree Construction 5



E = 01  
I = 00  
C = 10  
A = 111  
H = 110

# Huffman Coding Example

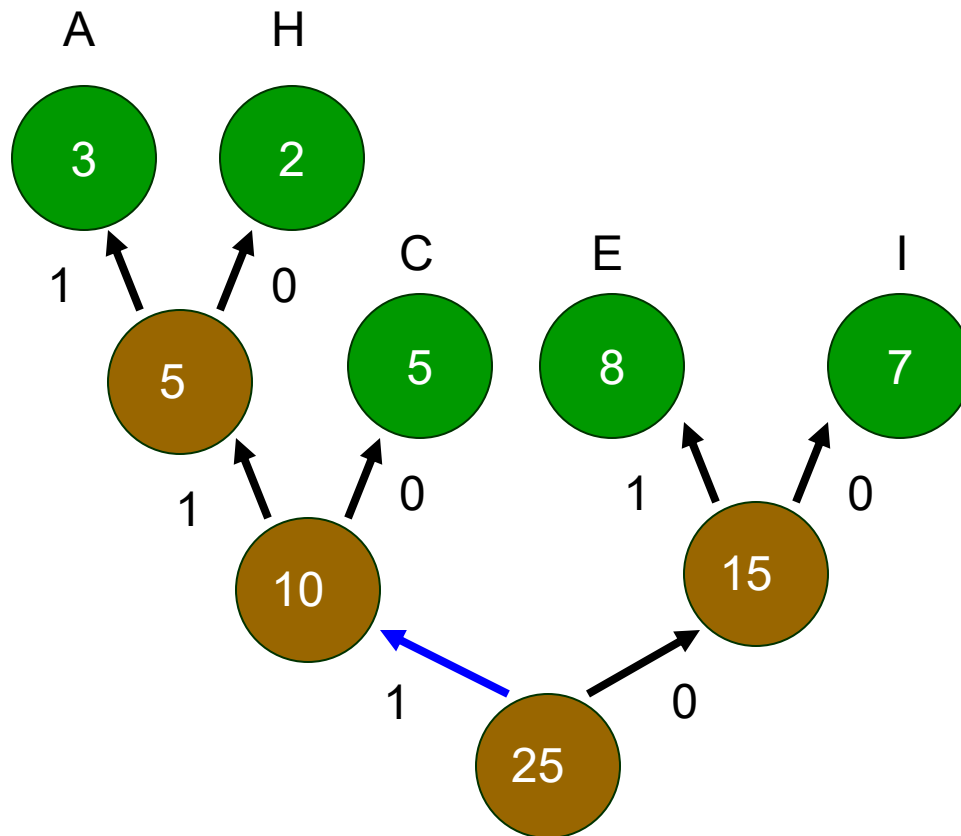
- Huffman code

E	=	01
I	=	00
C	=	10
A	=	111
H	=	110
- Input
  - ACE
- Output
  - (111)(10)(01) = 1111001

# Huffman Code Algorithm Overview

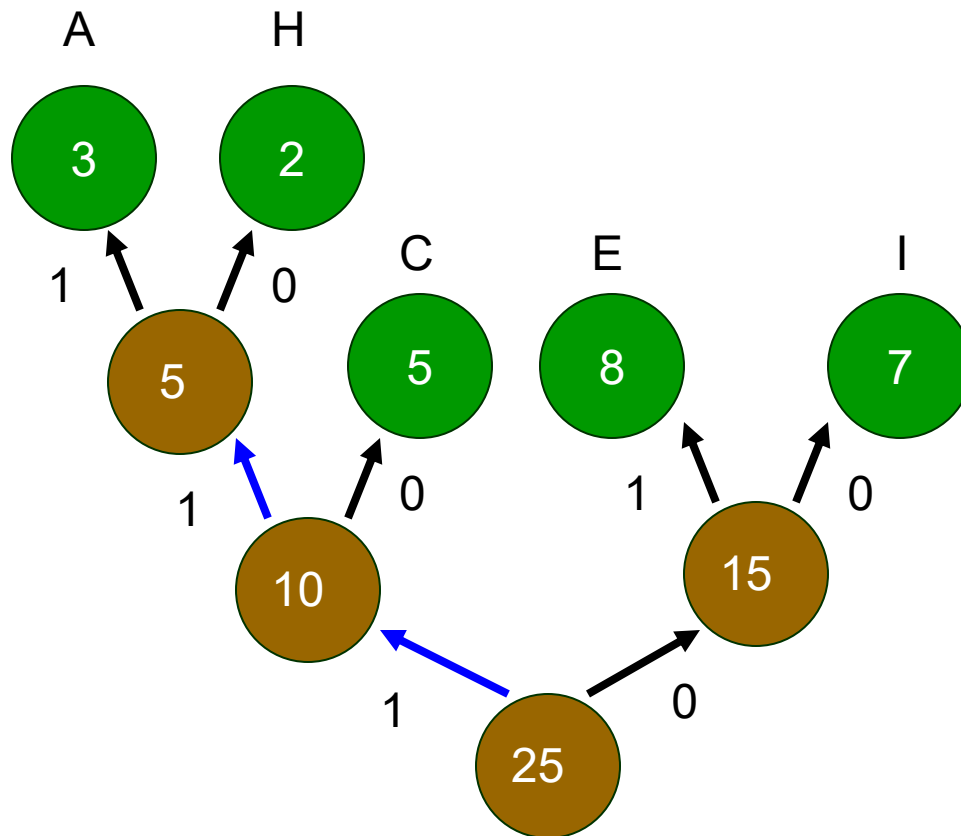
- Decoding
  - Read compressed file & binary tree
  - Use binary tree to decode file
    - Follow path from root to leaf

# Huffman Decoding 1



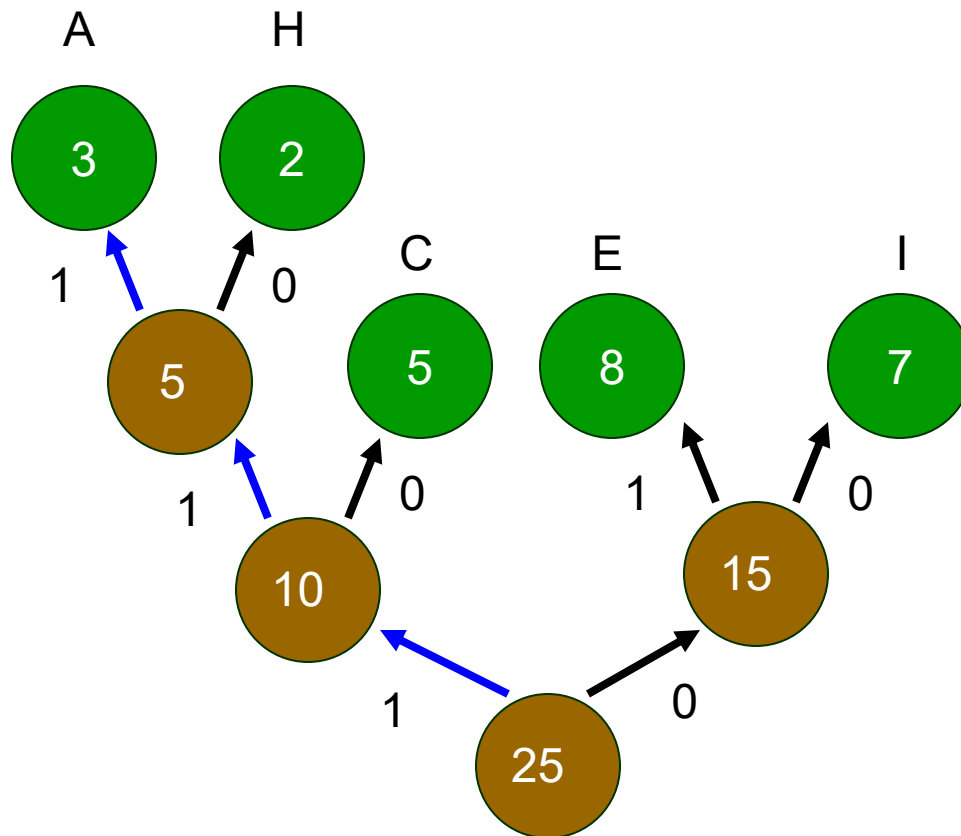
1111001

# Huffman Decoding 2



1111001

# Huffman Decoding 3

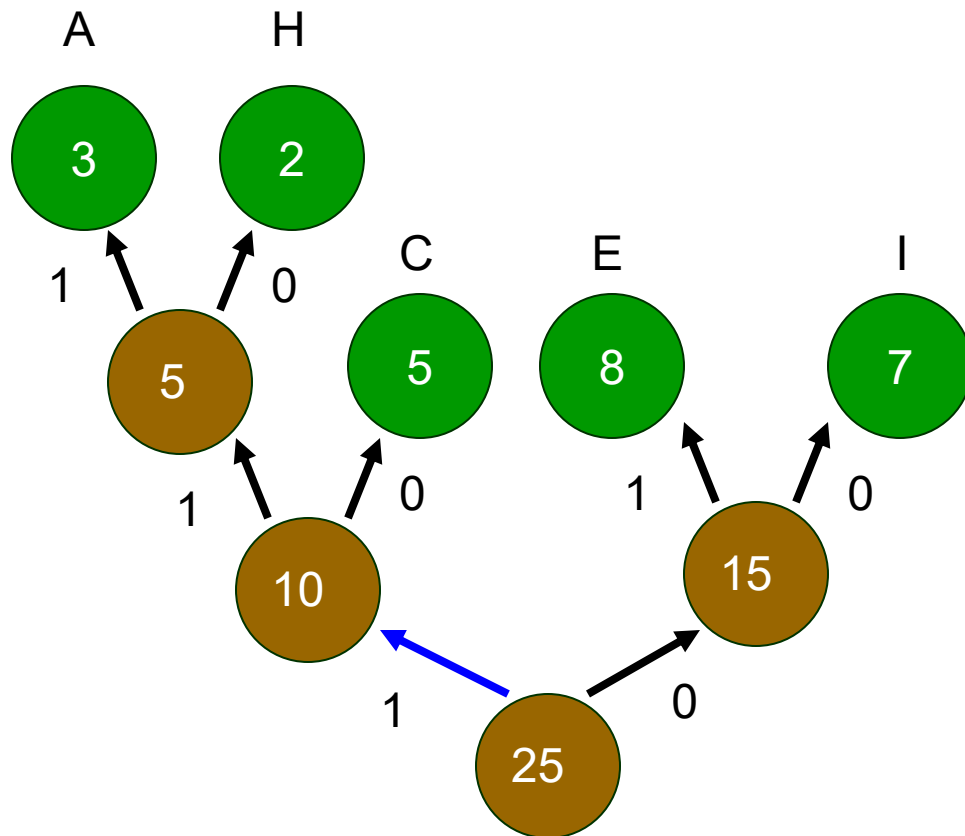


1111001

A



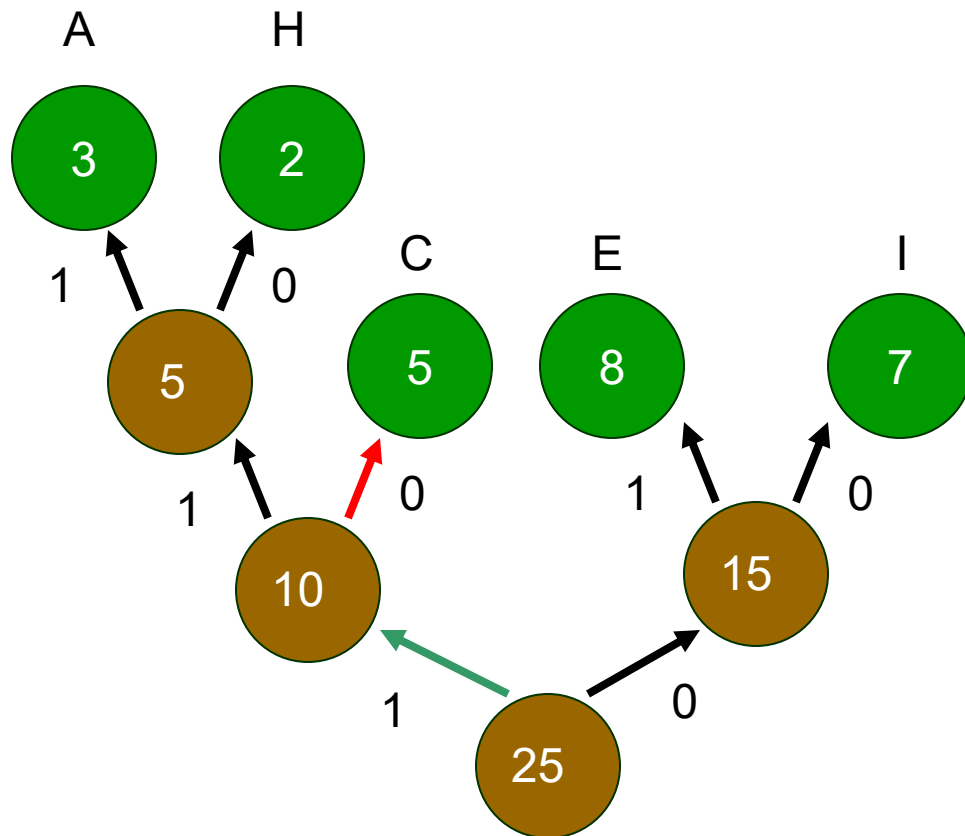
# Huffman Decoding 4



1111001

A

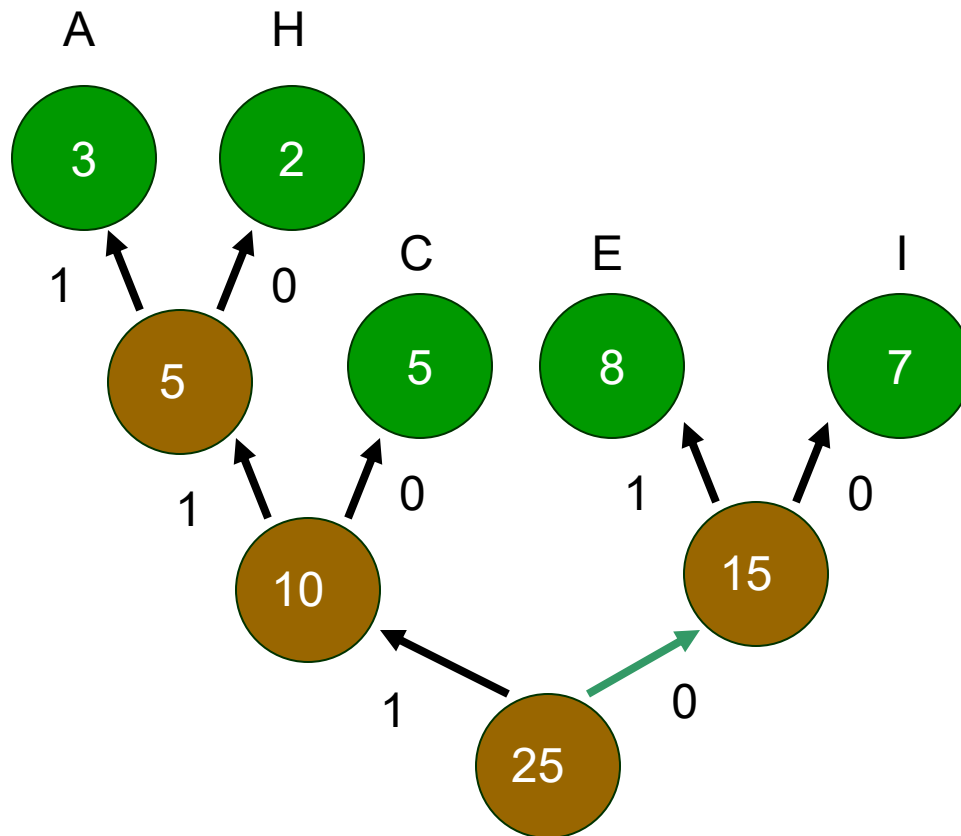
# Huffman Decoding 5



1111001

AC

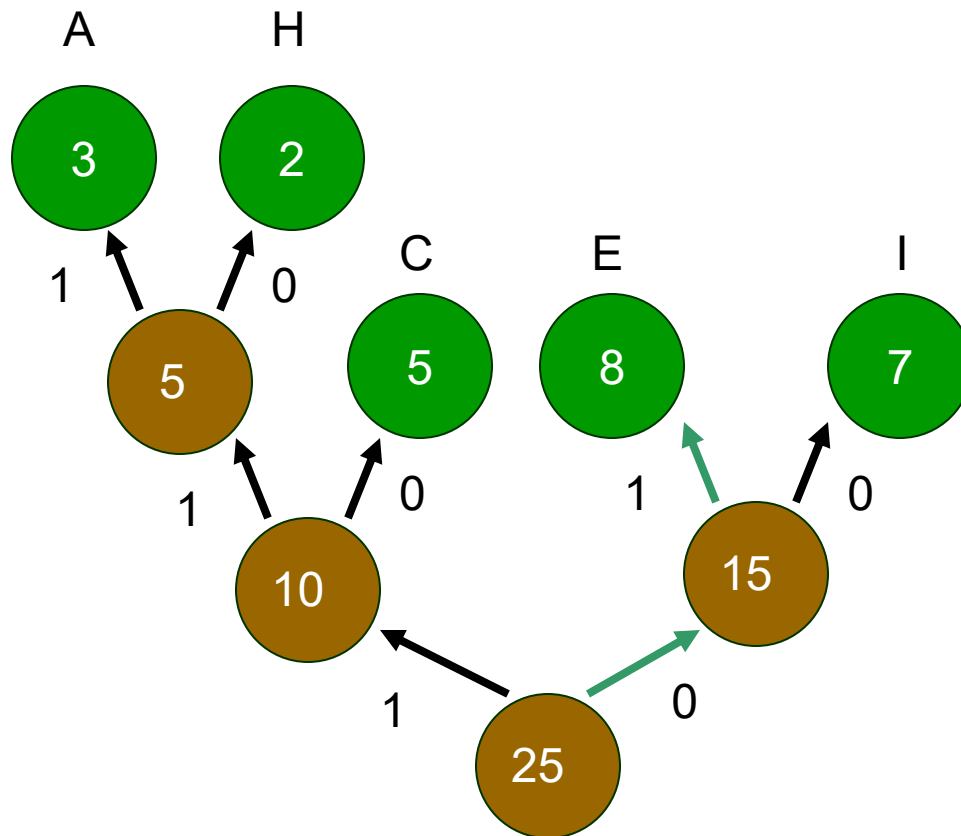
# Huffman Decoding 6



1111001

AC

# Huffman Decoding 7



1111001

ACE

# Huffman Code Properties

- Prefix code
  - No code is a prefix of another code
  - Example
    - $\text{Huffman}(\text{"I"}) \Rightarrow 00$
    - $\text{Huffman}(\text{"X"}) \Rightarrow 001$  // not legal prefix code
  - Can stop as soon as complete code found
  - No need for end-of-code marker
- Nondeterministic
  - Multiple Huffman coding possible for same input
  - If more than two trees with same minimal weight

# Huffman Code Properties

- Greedy algorithm
  - Chooses best local solution at each step
  - Combines 2 trees with lowest frequency
- Still yields overall best solution
  - Optimal prefix code
  - Based on statistical frequency
- Better compression possible (depends on data)
  - Using other approaches (e.g., pattern dictionary)



Quiz



# Huffman Code Construction

- Character count in text.
- Character Encoding?

Char	Freq
E	125
T	93
A	80
O	76
I	73
N	71
S	65
R	61
H	55
L	41
D	40
C	31
U	27



# Huffman Code Construction

Char	Freq
E	125
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S	65
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H	55
L	41
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U	27

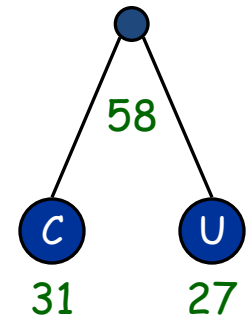
C  
31

U  
27

# Huffman Code Construction

Char	Freq
E	125
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	58
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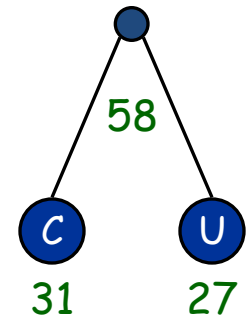
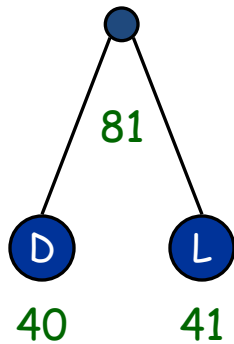
C	31
U	27



# Huffman Code Construction

Char	Freq
E	125
T	93
	81
A	80
O	76
I	73
N	71
S	65
R	61
	58
H	55

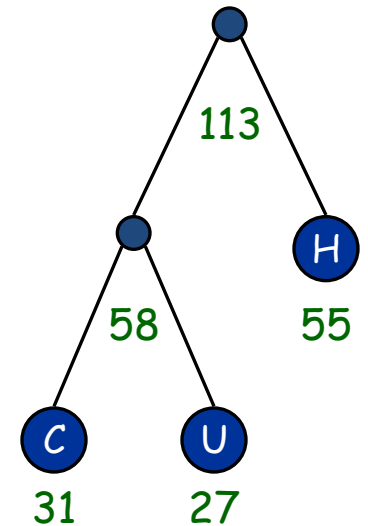
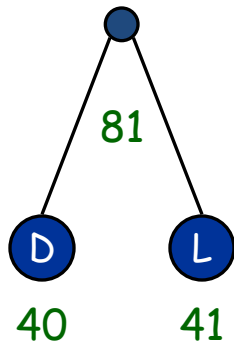
L	41
D	40



# Huffman Code Construction

Char	Freq
E	125
	113
T	93
	81
A	80
O	76
I	73
N	71
S	65
R	61

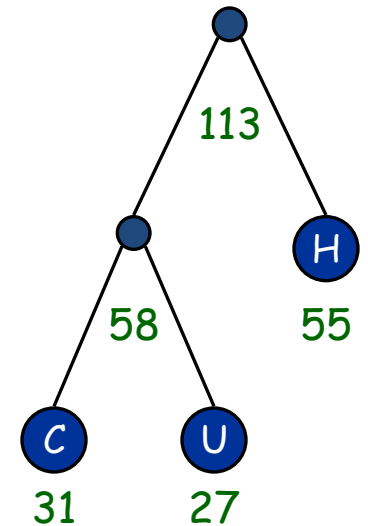
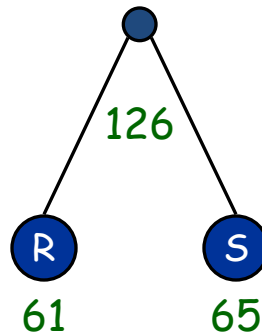
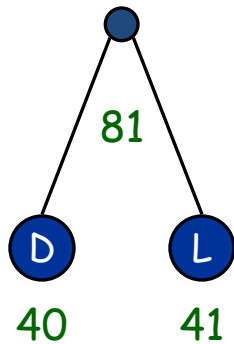
	58
H	55



# Huffman Code Construction

Char	Freq
	126
E	125
	113
T	93
	81
A	80
O	76
I	73
N	71

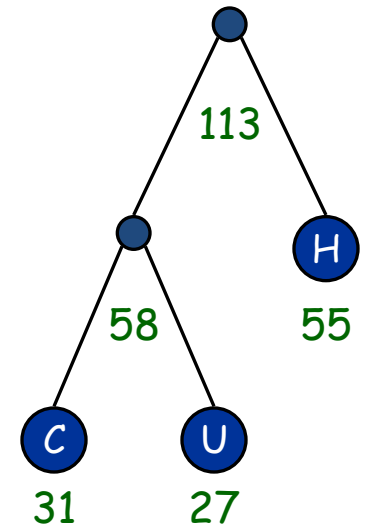
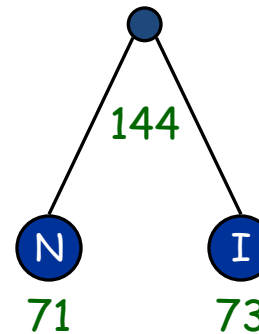
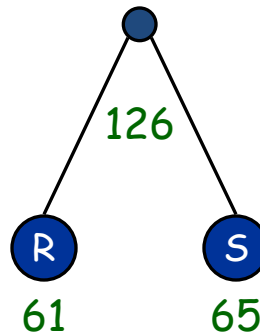
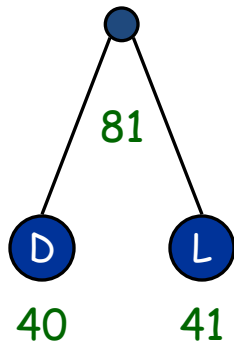
S	65
R	61



# Huffman Code Construction

Char	Freq
	144
	126
E	125
	113
T	93
	81
A	80
O	76

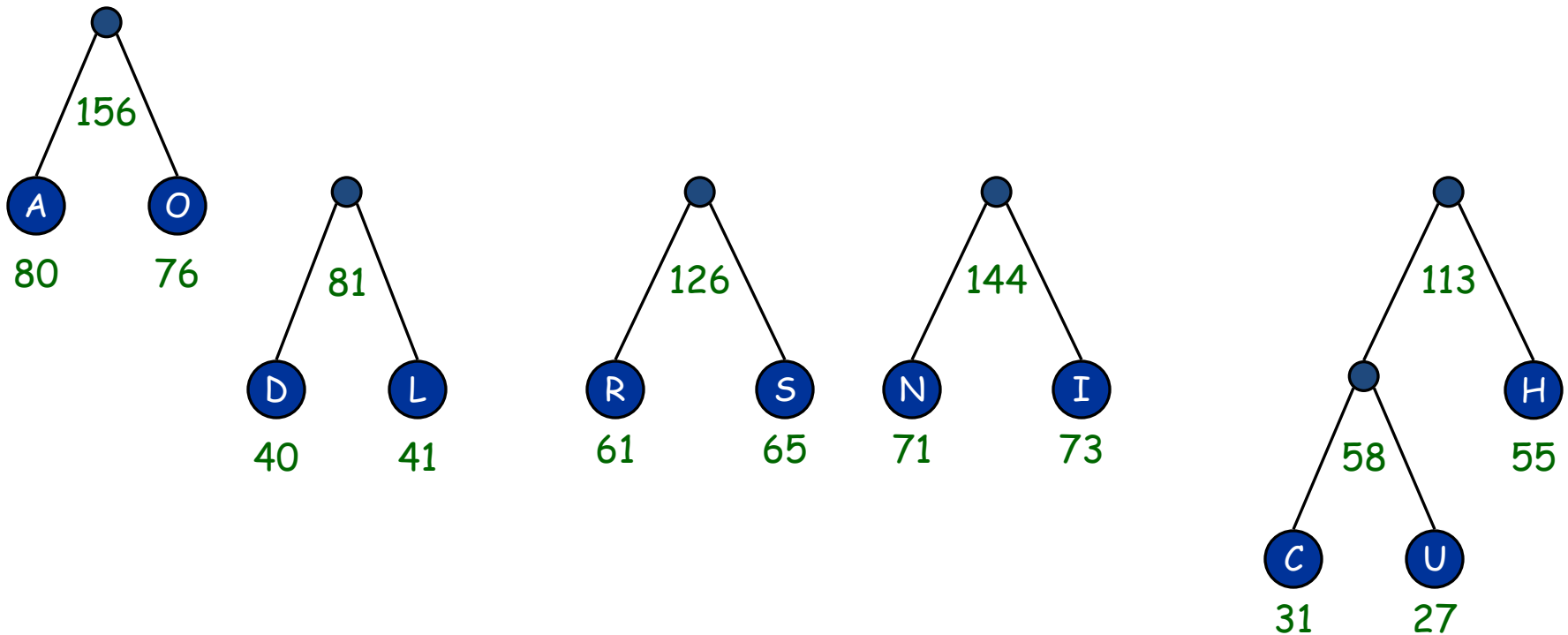
I	73
N	71



# Huffman Code Construction

Char	Freq
	156
	144
	126
E	125
	113
T	93
	81

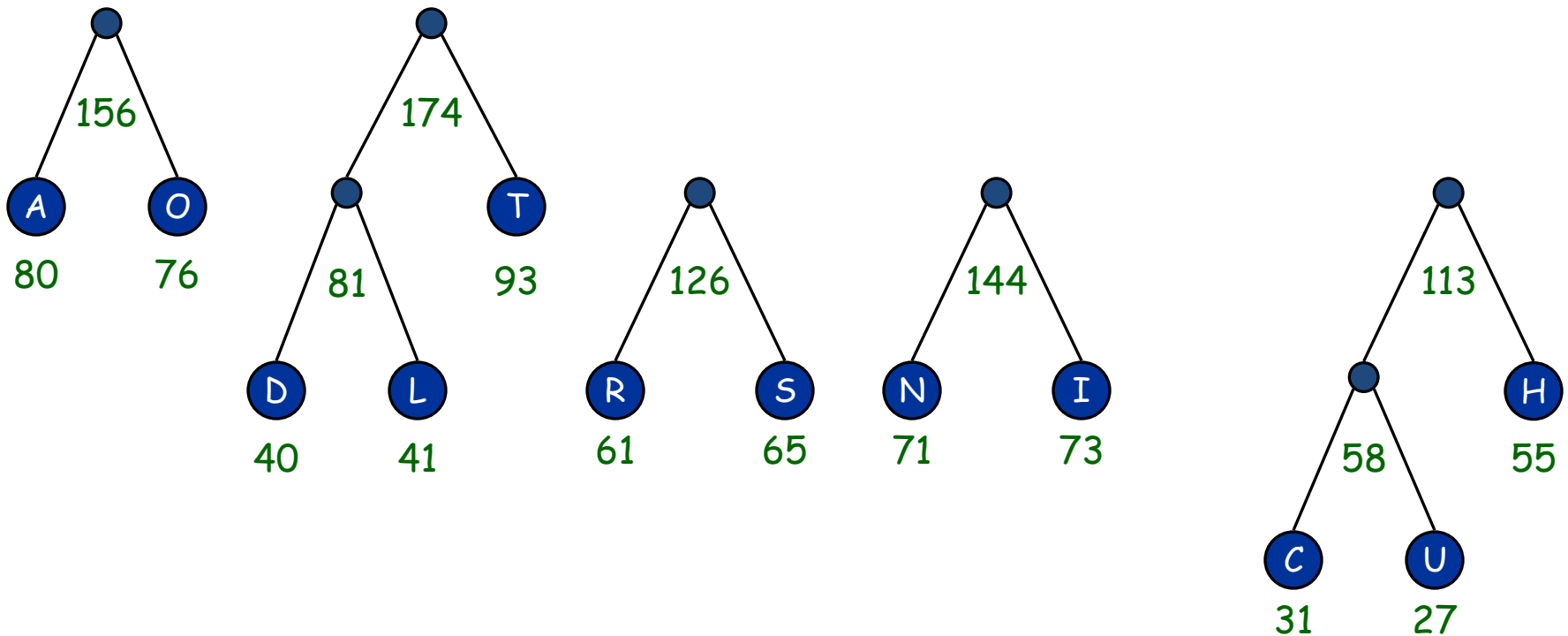
A	80
O	76



# Huffman Code Construction

Char	Freq
	174
	156
	144
	126
E	125
	113

T	93
	81

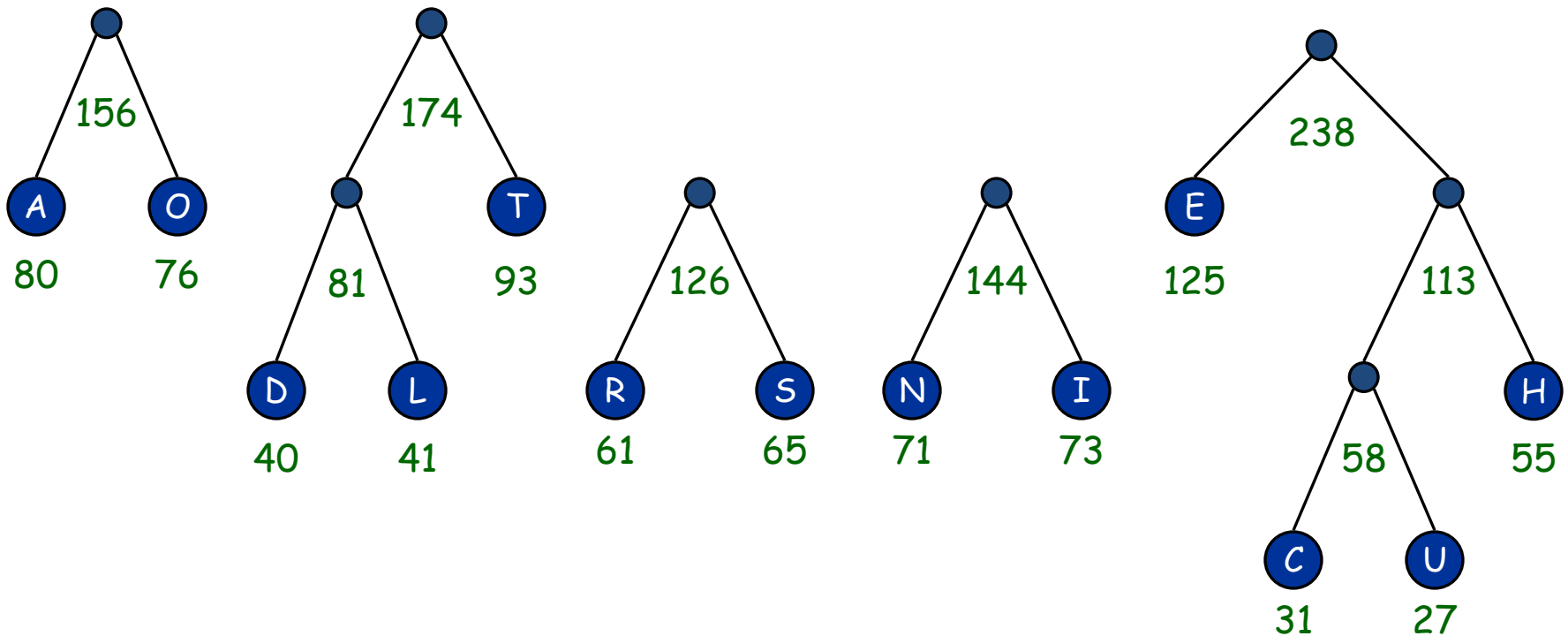




# Huffman Code Construction

Char	Freq
	238
	174
	156
	144
	126

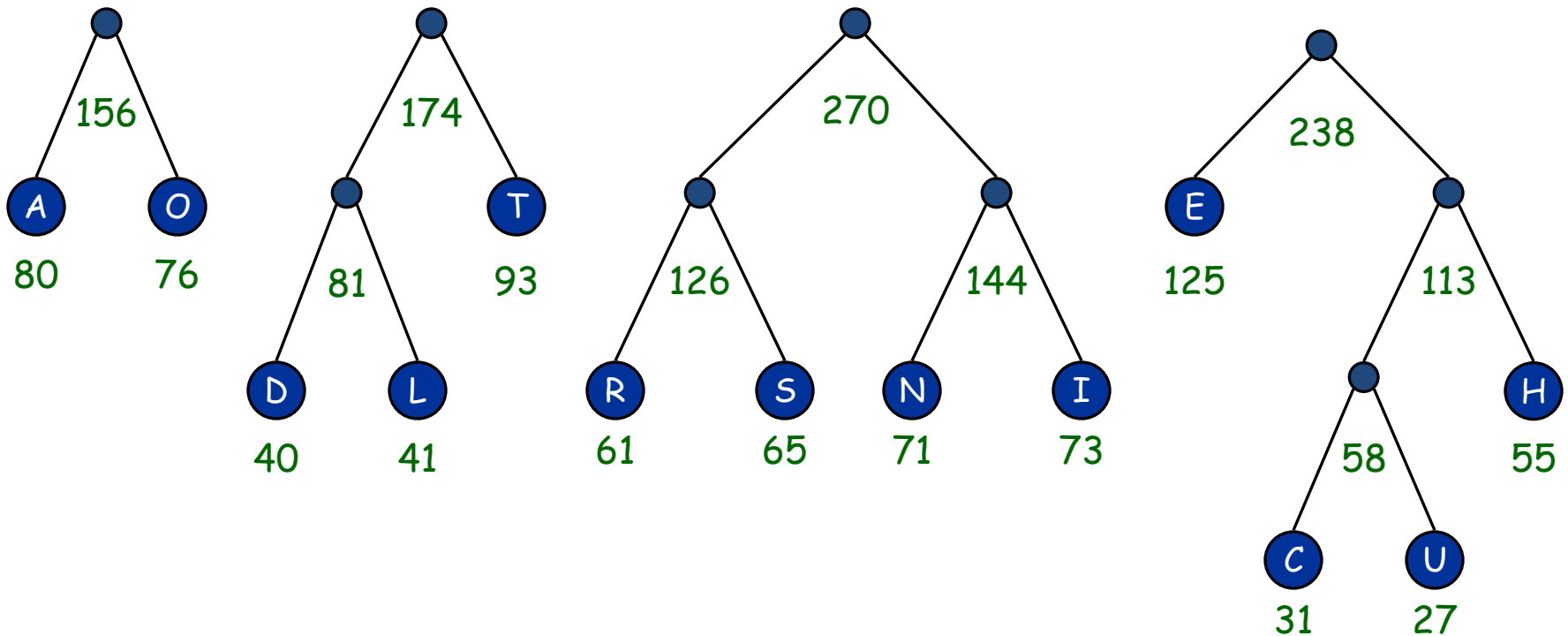
E	125
	113



# Huffman Code Construction

Char	Freq
	270
	238
	174
	156

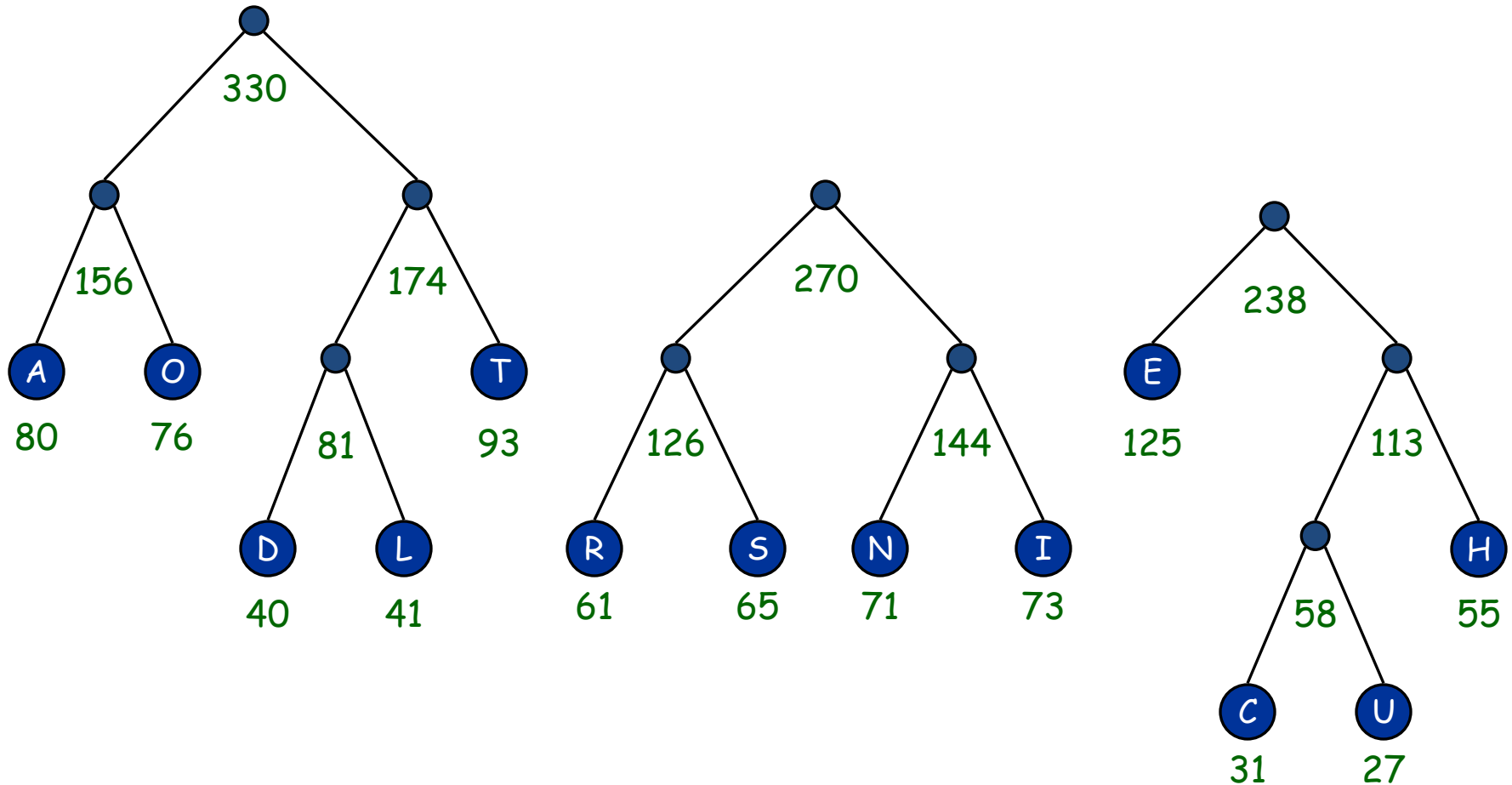
	144
	126



# Huffman Code Construction

Char	Freq
	330
	270
	238

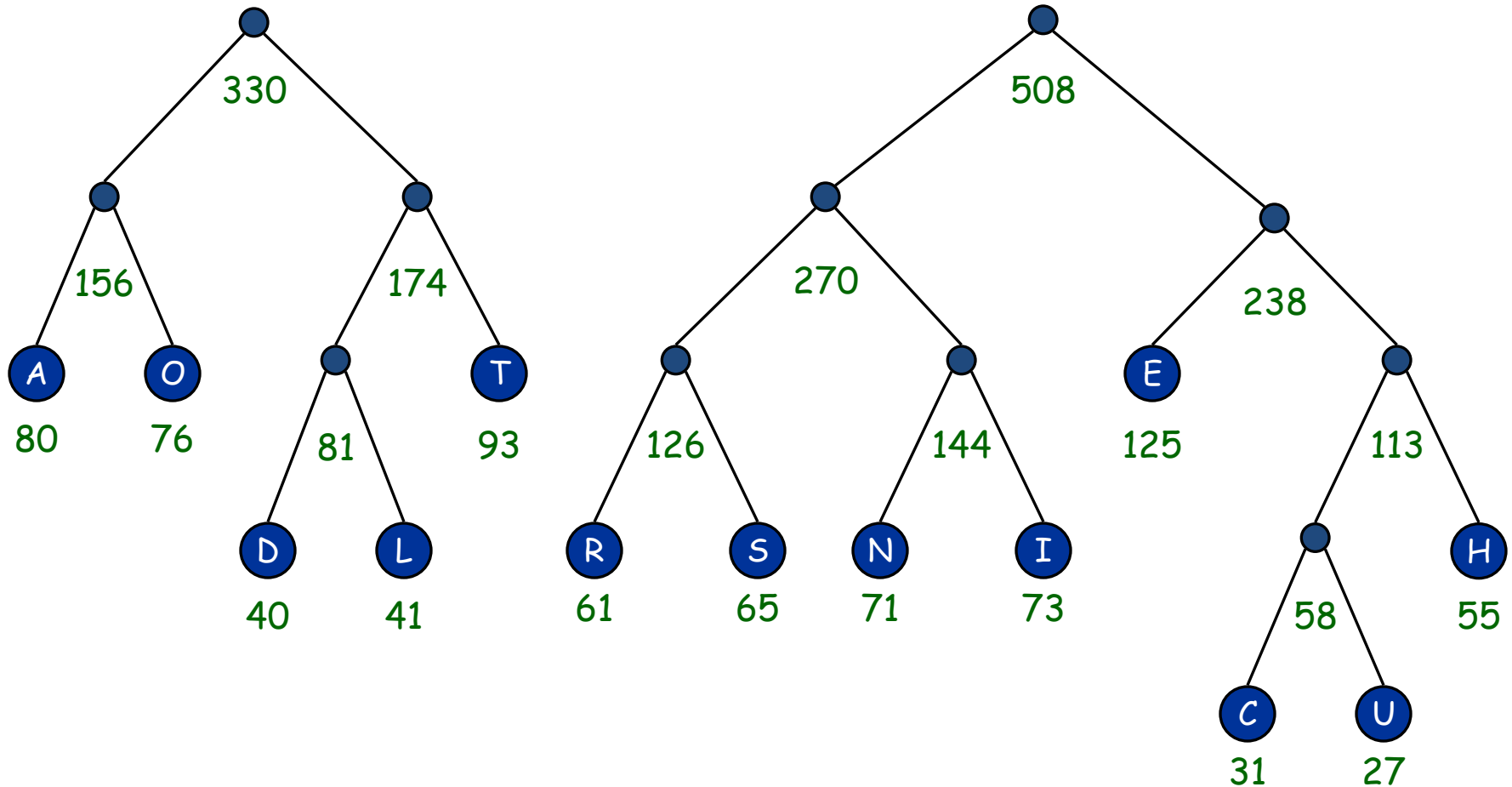
	174
	156



# Huffman Code Construction

Char	Freq
	508
	330

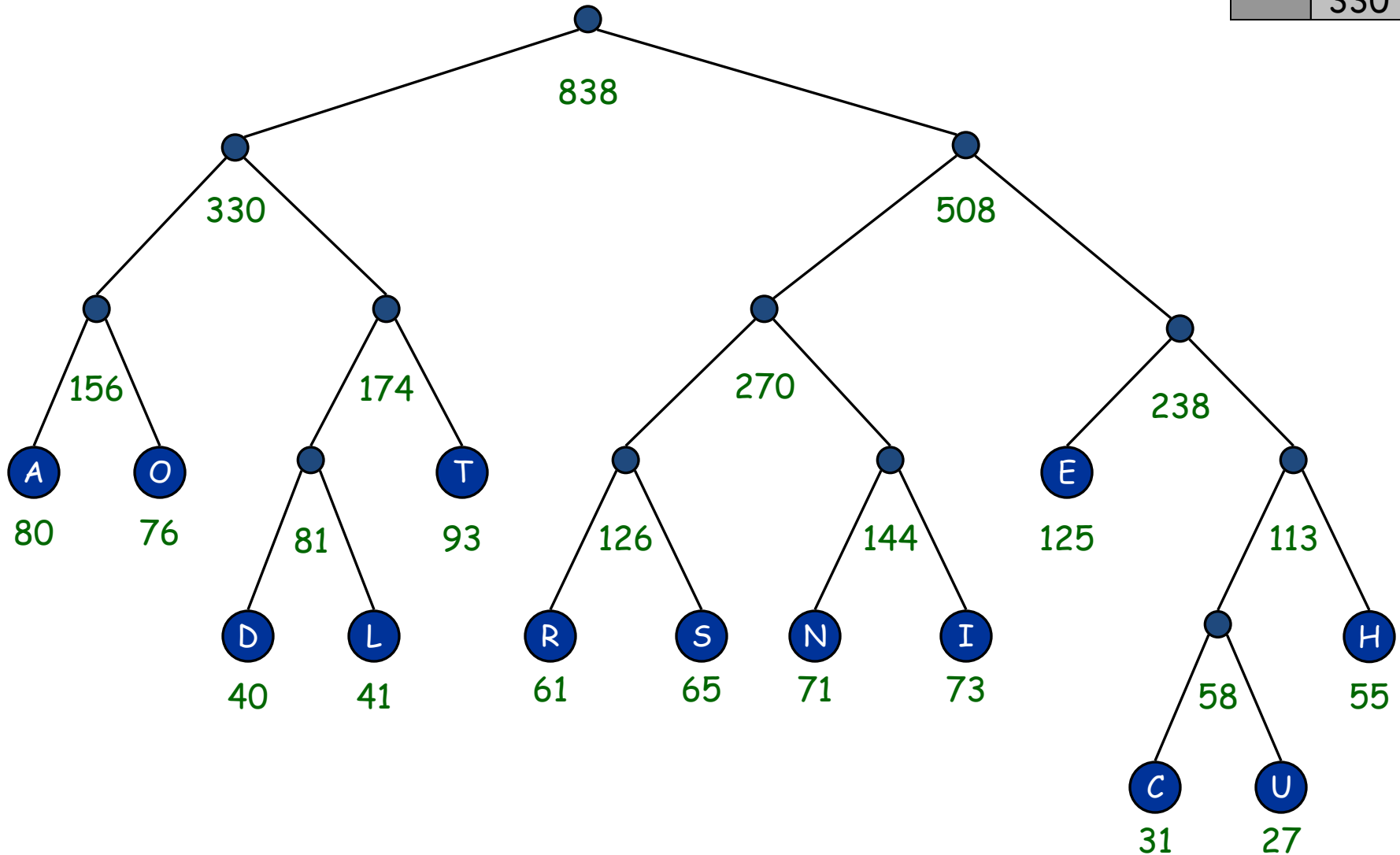
	270
	238



Char	Freq
	838

# Huffman Code Construction

	508
	330



# Huffman Code Construction

