Analysis and Design of Algorithms

Greedy Algorithms (Part 2):

Counting Money and Huffman Compression

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I THINK GREED Sometimes gets The best of Everybody.

QUOTEHD,COM

Alan Haft

Optimization Problems

- For most <u>optimization problems</u> you want to find, <u>not</u> just *a* <u>solution</u>, but the <u>best solution</u>.
- A <u>greedy algorithm</u> 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 <u>local optimum</u> at each step, you will end up at a <u>global optimum</u>.



Example: Counting Money

- Suppose you want to <u>count out</u> a certain amount of money, using the <u>fewest</u> possible bills and coins
- A greedy algorithm to do this would be: At each step, <u>take the largest possible bill or coin</u> 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 <u>US money</u>, the greedy algorithm <u>always</u> gives the <u>optimum</u> solution

Greedy Algorithm Failure

- In some (fictional) monetary system, "krons"
- come in <u>1</u> kron, <u>7</u> kron, and <u>10</u> kron coins
- Using a greedy algorithm to <u>count out 15</u> krons, you would get
 - A <u>10</u> kron piece
 - Five 1 kron pieces, for a total of 15 krons
 - This requires six coins
- A better solution would be to use two <u>7</u> 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 <u>nine jobs</u>, with <u>running times</u> of 3, 5, 6, 10, 11, 14, 15, 18, and 20 minutes.
- You have <u>three processors</u> on which you can run these jobs.
- You decide to do the <u>longest-running jobs first</u>, 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 <u>size</u> of data
 (number of <u>bits</u> needed to <u>represent</u> data)
- Benefits
 - <u>Reduce storage</u> needed
 - <u>Reduce transmission</u> cost / bandwidth

Sources of Compressibility

- <u>Redundancy</u>
 - Recognize <u>repeating</u> patterns
 - Exploit using
 - <u>Dictionary</u>
 - 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
 - <u>May lose</u> some information
 - Exploits redundancy & human perception
 - Applied to audio, image, video

Effectiveness of Compression

- Metrics
 - Bits per byte (8 bits)
 - 2 bits / byte \Rightarrow ¼ original size
 - 8 bits / byte \Rightarrow no compression
 - Percentage
 - 75% compression \Rightarrow ¼ original size

Effectiveness of Compression

- Depends on data
 - -<u>Random</u> data \Rightarrow hard
 - Example: $1001110100 \Rightarrow ?$
 - -<u>Organized</u> data \Rightarrow easy
 - Example: 1111111111 \Rightarrow 1×10
- Corollary
 - <u>No</u> universally <u>best</u> 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 <u>index</u> into dictionary
- Run length encoding
 - Find & compress <u>repetitive</u> sequences
- Huffman codes

Use variable length codes based on <u>frequency</u>

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	Α	B	С	D
Frequency	13%	25%	50%	12%
Original	00	01	10	11
Encoding	2 bits	2 bits	2 bits	2 bits
Huffman	110	10	0	111
Encoding	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 <u>binary tree</u> 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
 - <u>Weight</u> of leaf = symbol <u>frequency</u>
 - 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 <u>frequency</u> ⇒ frequency(L) + frequency(R)

- <u>Repeat</u> until all nodes <u>merged</u> into one tree

















Huffman Coding Example

Huffman code

- Input
 - -ACE
- Output

-(111)(10)(01) = 1111001

Huffman Code Algorithm Overview

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





















Huffman Code Properties

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

Huffman Code Properties

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



- •Character count in text.
- •Character Encoding?

Char	Freq
E	125
Т	93
A	80
0	76
I	73
N	71
S	65
R	61
Н	55
L	41
D	40
С	31
U	27

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Char	Freq
	126
Е	125
	113
Т	93
	81
A	80
0	76
Ι	73
N	71

S	65
R	61







Char	Freq
	144
	126
Е	125
	113
Т	93
	81
A	80
0	76

I	73
N	71





Ι



Char	Freq
	156
	144
	126
Е	125
	113
Т	93
	81





27

С

Char	Freq
	174
	156
	144
	126
E	125
	113







Char	Freq
	238
	174
	156
	144
	126





Char	Freq
	270
	238
	174
	156





Char	Freq
	330
	270
	238





27

Η

55

С

Char	Freq
	508
	330









