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.

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Alan Haft

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.

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:
		- \cdot 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 jobsto 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 % original size
		- 8 bits / byte \Rightarrow no compression
	- Percentage
		- 75% compression \Rightarrow % original size

Effectiveness of Compression

- Depends on data
	- $-$ Random data \Rightarrow hard
		- Example: $1001110100 \Rightarrow$?
	- $-$ Organized data \Rightarrow easy
		- Example: $1111111111 \Rightarrow 1\times10$
- 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

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

Huffman Code Data Structures

- Binary (Huffman) tree
	- Represents Huffman code
	- $-$ Edge \Rightarrow code (0 or 1)
	- $-\text{Leaf} \implies$ 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 Coding Example

• Huffman code

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

- 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 Code Properties

- Prefix code
	- $-$ No code is a prefix 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
- 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)

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

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