# **Compression & Huffman Codes**

# Compression

#### Definition

Reduce size of data

(number of bits needed to represent data)

#### Benefits

- Reduce storage needed
- Reduce transmission cost / latency / 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 ⇒ no compression
  - Percentage
    - **75% compression**  $\Rightarrow$  <sup>1</sup>/<sub>4</sub> original size

## **Effectiveness of Compression**

#### Depends on data

- **Random data**  $\Rightarrow$  hard
  - **Example:**  $1001110100 \Rightarrow ?$
- Organized data ⇒ easy
  - **Example:** 111111111  $\Rightarrow$  1×10

#### 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	Α	В	С	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  $\Rightarrow$  1/8×3 + 1/4×2 + 1/2×1 + 1/8×3 = 1.75 bits / symbol

## **Huffman Code Data Structures**





# **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 ⇒ frequency(L) + frequency(R)

Repeat until all nodes merged into one tree

















# **Huffman Coding Example**



#### 

### 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

1111001





### **11**11001











# **Huffman Code Properties**

#### Prefix code

- No code is a prefix of another code
- Example
  - Huffman("I") ⇒ 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)