



Information and Compression

File, 100K characters

Each character has 8 possible types a, b, c, d, e, f, g, h

↓
Can be encoded with 3 bits

File: 300K bits

→ you know in addition the frequency of each character

Huffman Coding

Def (Prefix-Freeness)

A variable length encoding is prefix free if the encoding of any character is not a prefix of encoding of any other.

01 a

010 b

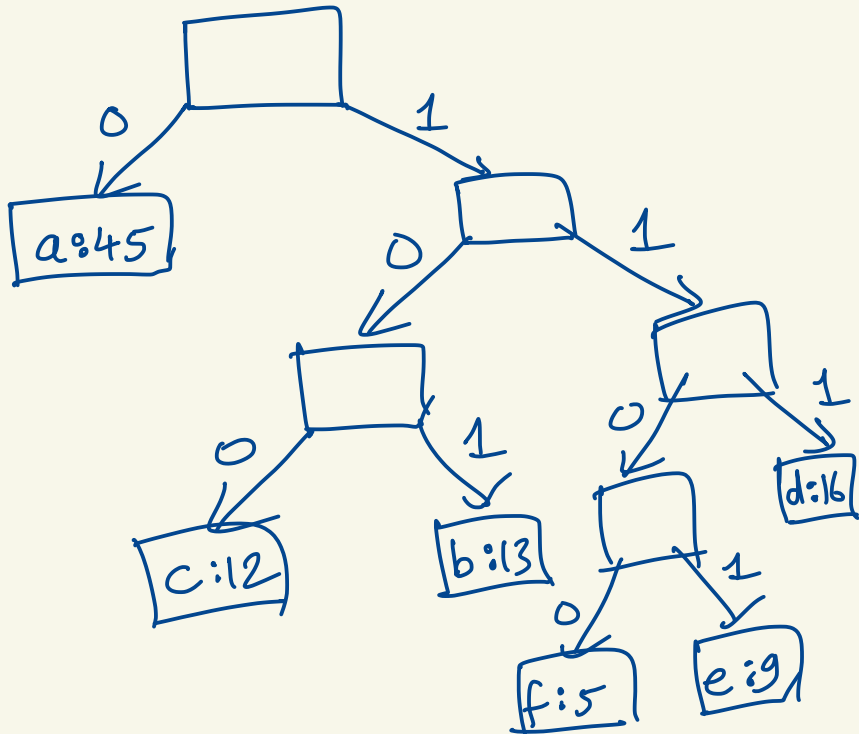
Huffman Coding Algorithm: Greedy Algorithm

A variable length encoding can be described by a tree.

a	b	c	d	e	f	g	h
45K	13K	12K	16K	9K	5K	0K	0K

V.L.E: 0 101 100 111 1101 1100 - -

decode:



- 1) decoding $\equiv$ tree
 - 2) every leaf is a symbol.
 - 3) if you branch at a node, then both branches lead to proper symbols
-

algo: Huffman($X, f(\cdot)$)

$n = |X|$, # of chars.

for all $x \in X$, enqueue $((x, f(x)), Q)$

for $i = 1$ to n ,

allocate a tree node z

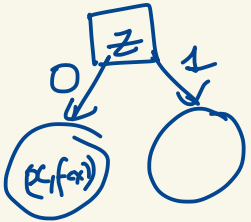
left-child = delete-min(Q)

right-child = delete-min(Q)

$f(z) = f(\text{left-child}) + f(\text{right-child})$

Make left-child & right-child the children of z

enqueue $((z, f(z)), Q)$.



Runtime: $O(n \cdot \log n)$

Theorem: Alg above produces a tree T s.t.
the cost of encoding with T is minimum
possible among all prefix free encodings

Pf: $|X| = 2$ ✓

$|X| > 2$: x, y : lowest freq elements

$$\bar{X} = X - \{x, y\} \cup \{z\}$$

$$|\bar{X}| = |X| - 1$$

(Inductively assume that the algo builds
the $\bar{T}$ tree for $\bar{X}$.)

Goal: by adding x, y as left & right
child of z to $\bar{T}$ is also optimal for X .

$$\begin{aligned} \text{Cost}(T) &= \text{Cost}(\bar{T}) - f(z) \cdot d(z) + f(x) \cdot d(x) + f(y) \cdot d(y) \\ &= \text{Cost}(\bar{T}) - (f(x) + f(y))(d(x) - 1) + \text{"+"} \end{aligned}$$

$$= \text{Cost}(\bar{T}) + f(x) + f(y)$$

T' : optimal tree.

$$\text{Cost}(T') < \text{Cost}(T) \quad (\text{FSOC})$$

Assume further that x & y are
Siblings in T' .

Exercise: Work out the case when x &
 y are not siblings (HINT: reduce to
the above case)

Goal: prove lower bounds on encoding size of File F given frequencies f . $X = \{0,1\}^n$

Thm 1: $C: \{0,1\}^n \rightarrow \{0,1\}^*$ lossless encoding (one-one-map). Then, there is a file $F \in \{0,1\}^n$ s.t. $|C(F)| \geq n$.

Proof: Suppose not.

$$\begin{aligned}\# \text{ of possible files} &\leq 2^1 + 2^2 + \dots + 2^{n-1} \\ &= 2^n - 2 < 2^n\end{aligned}$$

So contradiction.

Thm 2: $F \in \{0,1\}^n$ be a sequence of n u.a.v. bits. Then,

$$\Pr[|C(F)| \leq n-t] \leq \frac{1}{2^{t-1}}$$

Proof:

$$= \frac{|\{F: |C(F)| \leq n-t\}|}{2^n}$$

$$= \frac{(2^1 + 2^2 + \dots + 2^{n-t})}{2^n} < \frac{2^{n-t+1}}{2^n}$$

$$= \frac{1}{2^{t-1}}$$

Thm 3: $C: \rightarrow$ prefix free encoding -

$F: n$ -random bits.

Then $\mathbb{E}[|C(F)|] \geq n$.

Thm 4:

$G \subseteq \{0,1\}^n$ be a set of possible files. $C: \text{injective function } \{0,1\}^n \rightarrow \{0,1\}^*$.

Then:

1) there's a file s.t. $|C(F)| \geq \log_2 |G| - 1$
 $F \in G$

2) $\Pr[|C(F)| \leq \log_2 |G| - t] \leq \frac{1}{2^{t-1}}$

3) - $\forall$ prefix-free C , $F: n$ -random

$$\mathbb{E}_F[|C(F)|] \geq \log_2 |C(F)|$$

Given: frequencies of characters.

We will find a set of files G
"with typical frequencies"

File contains n characters.

C diff chars possible

Char i has frequency $\frac{f(i)}{n} = p(i)$

Lemma: the number of distinct

files of length n and frequencies

$f(1), f(2), \dots, f(C)$ is

Proof: If there are 2 chars
(i.e. $c=2$), you know this:

$$\binom{n}{f(c)} = \frac{n!}{f(c)!(n-f(c))!}$$

Similar formula for $c > 2$:

$$|G| = \frac{n!}{f(c_1)! f(c_2)! \dots f(c_c)!}$$

By Thm 4, encoding length

$$\sim \log_2 |G|.$$

$$= \log_2(n!) - \sum_{i=1}^c \log(f(i)!) \quad *$$

Use Stirling's approx.

$$M! \sim \left(\frac{M}{e}\right)^M \cdot \sqrt{M}$$

$$\log(M!) \sim M \cdot \log M - \text{"small"}$$

$$* = n \log n - \sum_{i=1}^c f(i) \cdot \log f(i)$$

$$= n \cdot \log n - \sum_{i=1}^c p(i) \cdot n \cdot \log(p(i) \cdot n)$$

$$= n \cdot \log n - \sum_{i=1}^c n \cdot p(i) \cdot \log(p(i)) \\ - \sum_{i=1}^c n \cdot p(i) \cdot \log(n)$$

Since $\sum_{i=1}^C p(i) = 1$, first & last terms
cancel

$$= - \sum_{i=1}^C n \cdot p(i) \cdot \log(p(i))$$

(Now, $-\log p(i) = \log \frac{1}{p(i)}$.)

$$= n \sum_{i=1}^C p(i) \cdot \log \frac{1}{p(i)}$$

Entropy of freq
distr

= minimum possible compressed

size of file with given freq

$\propto$ "information content" in the
file.