



# Introduction to Compression Algorithms

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# Why compression?

Stored information is redundant.



57% of the original size losslessly (PNG)

11% with lossy compression (JPEG)

```
png_struct png_ptr;  
png_info info_ptr;  
png_info_end_info;  
  
if ( (fp=fopen(name, "r")) ) {  
    fread(&header, 1, 8, fp);  
    png_sig_cmp(&header, 0, 8);  
    (png_ptr = png_create_read_struct(PNG_LIBPNG_VER_STRING, NULL, NULL, NULL,  
    return 0;  
  
    if ( (info_ptr = png_create_info_struct( png_ptr )) ) {  
        png_destroy_read_struct( &png_ptr, (png_info_ptr) NULL, (png_info_ptr) NULL );  
        return 0;  
    }  
  
    if ( (end_info = png_create_info_struct( png_ptr )) ) {  
        png_destroy_read_struct( &png_ptr, &info_ptr, (png_info_ptr) NULL );  
        return 0;  
    }  
  
    if ( setjmp(png_jmpbuf(png_ptr)) ) {  
        png_destroy_read_struct( &png_ptr, &info_ptr, &end_info );  
        fclose(fp);  
        return 0;  
    }  
  
    png_init_io(png_ptr, fp);  
    png_set_sig_bytes(png_ptr, 8);  
    if ( row_callback ) png_set_read_status_fn( png_ptr, row_callback );  
    png_read_info( png_ptr, info_ptr );  
  
    png_get_IHDR(png_ptr, info_ptr, (png_uint_32*) &width, (png_uint_32*) &height,  
    &bit_depth, &color_type, &interlace_type,  
    &compression_type, &filter_method);  
  
    if ( xres ) *xres = width;  
    if ( yres ) *yres = height;  
  
    if ( color_type == PNG_COLOR_TYPE_PALETTE )  
        png_set_palette_to_rgb(png_ptr);  
  
    if ( color_type == PNG_COLOR_TYPE_GRAY && bit_depth < 8 )
```

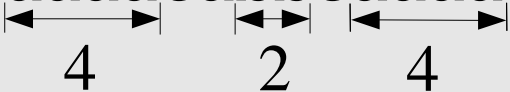
16% losslessly (bzip2)

# Sequential Coders

# RLE Encoding 1

Substitutes a sequence of the same letters with double-code which contains the letter itself and sequence length.

"abaaaacabbcaaaa" 15 B – input message



The diagram shows the input message "abaaaacabbcaaaa" with three double-headed arrows below it. The first arrow spans the four 'a's and is labeled '4'. The second arrow spans the two 'b's and is labeled '2'. The third arrow spans the four 'a's and is labeled '4'.

"a1b1a4c1a1b2c1a4" 16 B – [letter,length] - output

compression ratio:  $16/15 = 1.06$

minimal compression ratio:  $2/15 = 0.13$

maximal compression ratio:  $30/15 = 2.00$

# RLE Encoding 2

Do not encode unique letters.

"abaaaacabbcaaaa" 15 B – input message  
|←4→| |←2→| |←4→|  
4 2 4

"aba4cab2ca4" 11 B – not reconstructable

"abaa4cabb2caa4" 14 B – [letter] or [letter,letter,length]

compression ratio:  $14/15 = 0.93$

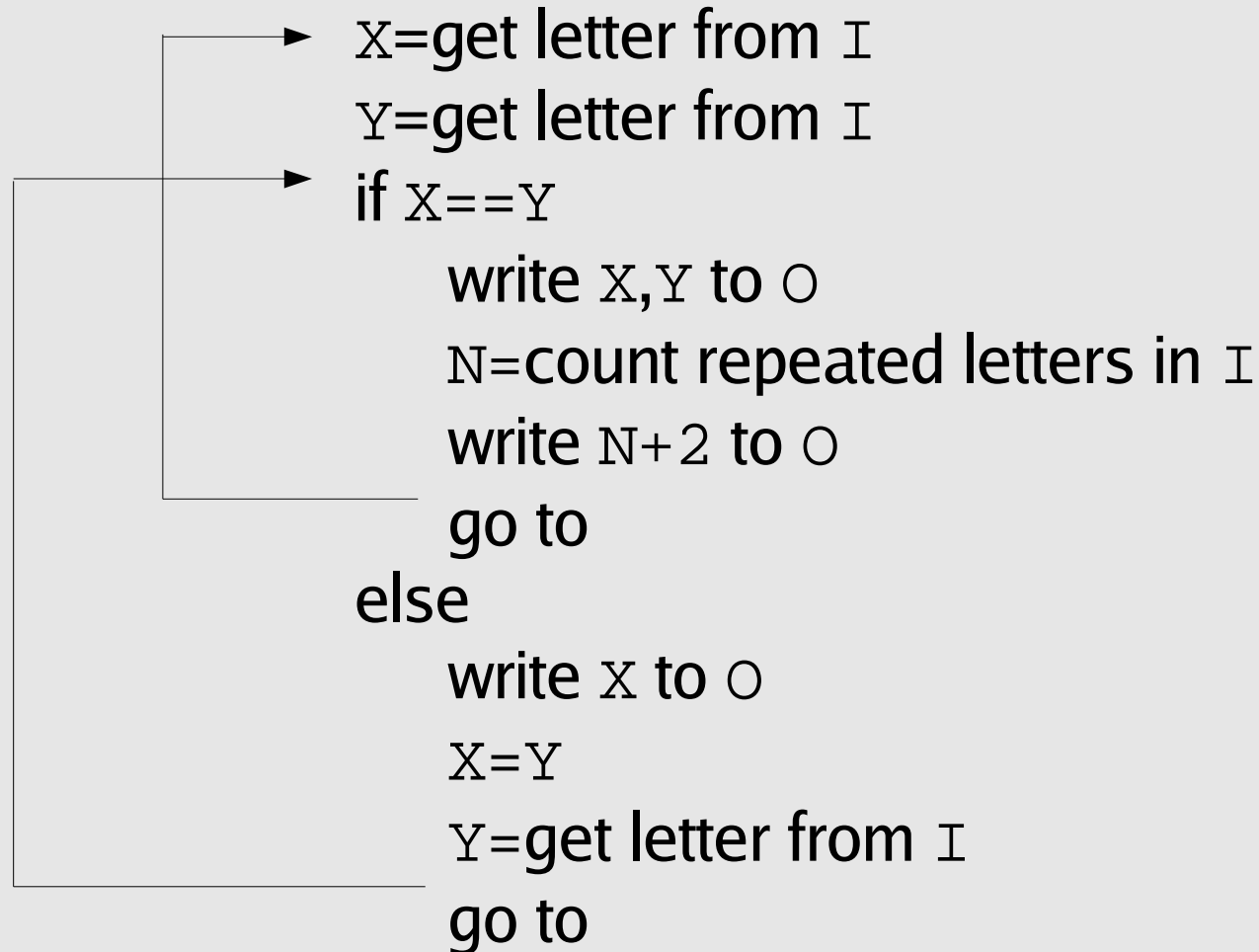
minimal compression ratio:  $3/15 = 0.20$

maximal compression ratio:  $22/15 = 0.68$

## The Algorithm – Encoder

I = "abaaaacabbcaaaa" 15 B – input message

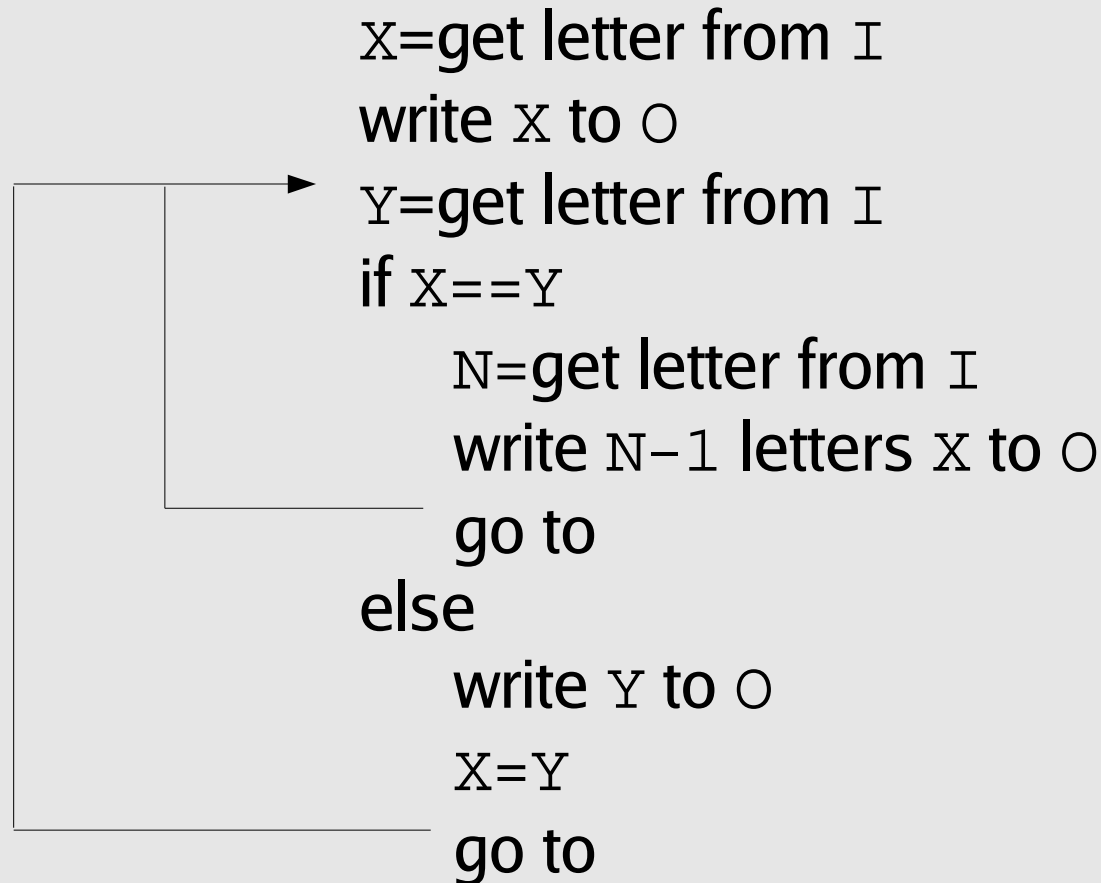
O = "abaa4cabb2caa4" 14 B – output message



## The Algorithm – Decoder

I = "abaa4cabb2caa4" 14 B – input message

O = "abaaaacabbcaaaa" 15 B – output message



# RLE Encoding 3

Enhancement: “skipping RLE”

“abaaaacabbcaaaa” 15 B – input message  
          |←→| |←→| |←→|  
          4   2   4

“2ab4a5cabbc4a” 13 B – [length]

compression ratio:  $13/15 = 0.86$

minimal compression ratio:  $3/15 = 0.20$

maximal compression ratio:  $16/15 = 1.06$



# RLE Algorithms Conclusion

- ... used mostly as a first pass method
- ... it's fast,  $O(n)$
- ... ineffective

# Burrows-Wheeler Transform

(block-sorting compression)

# BWT - Coder

1. generate rotated sequence of the input message
2. sort sequences lexicographically
3. output is the last letters from the message + starting index, where  $I_f=1$

$I_f$	rotated $M$	$I$	$I_f$	rotated, sorted $M$	output $M_{\text{BWT}}$
0	abaca	0	4	aabac	c
1	bacaa	1	0	abaca	a
2	acaab	2	2	acaab	b
3	caaba	3	1	bacaa	a
4	aabac	4	3	caaba	a

cabaa, 3

# BWT - Decoder

The input to the decoder is:

cabaa , 3

$I$		$M_{\text{BWT}}$		$I_b$	sorted $M_{\text{BWT}}$
0	$\xRightarrow{4}$	c	$\xRightarrow{4}$	1	a
1	$\xRightarrow{5}$	a	$\xRightarrow{5}$	3	a
2	$\xRightarrow{2}$	b	$\xRightarrow{2}$	4	a
3	$\xRightarrow{1}$	a	$\xRightarrow{1}$	2	b
4	$\xRightarrow{3}$	a	$\xRightarrow{3}$	0	c

- 1  $i = I_1$
- 2  $M_{\text{BWT}}(i) \rightarrow$  output
- 3  $i = I_b(i)$
- 4 if  $i \neq I_0$  go to 2

# BWT – Why is it used?

Burrows-Wheeler Transform doesn't compress the input message, so why would we bother using it?

```
\chapter{Wavelets}\label{cha:wav}.\section{Why wavelets?}.\initial{U}p to now we have discussed only transforms calculating decomposition of the entire signal. to a linear combination of non-localized basis vectors. Thus in general, a change of one coefficient in transform domain affects all samples in reconstructed signal. The right opposite is the case of wavelets. The most apparent comparison can be made if we consider FFT of a time domain signal. By FFT of a signal we obtain an information about its frequency components precisely, but we can say absolutely nothing about its frequency content in a particular time interval shorter.
```

```
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```

# Lempel-Ziv Algorithms

The first dictionary based compression method was created by Abraham Lempel and Jacob Ziv – Lempel-Ziv-77 - LZ77.

The output message contains symbols of three types:

1. letter = uncompressed letter
2. code word = contains [position,length]
3. flag = a bit telling encoder to expect either a letter or a code word

# Lempel-Ziv 77 Algorithm

"abaaaacabbcaaaa"

0123456789ABCDE

"abaaacabbcaaa"

0000101011

"abaa2c0b52"

2 2 23

Assuming:

- 4 bits for position (0..15)
- 4 bits for length (1..16)
- 8 bits for a letter (0..255)

... and the compression ratio is:

$$(6 \cdot 8 + 10 + 8 \cdot 4) / 14 \cdot 8 = 0.80$$

# Lempel-Ziv 77 – Finite Window

Finite window encoding principle:

- position of a match is set relatively from encoding position
- size of a window where a match is searched is limited by bits to express the match position

Let's see: 2-bit position:

3210

"abaaaacabbcaaaa"

or later in encoding process:

3210

"abaaaacabbcaaaa"



# Lempel-Ziv 77 – Finite Window

input:           "abaa**a**ac**ab**bbca**aaa**"

output:           0  
                  "a"

coding position:    0  
bits per position:  3   (0..7)  
bits per length:   1   (1..2)  
bits per letter:    8   (0..255)

# Lempel-Ziv 77 – Finite Window

0

input: "abaaacabbcaaaa"

output: 00

"ab"

coding position:	1	
bits per position:	3	(0..7)
bits per length:	1	(1..2)
bits per letter:	8	(0..255)

# Lempel-Ziv 77 – Finite Window

10

input: "abaaacabbcaaaa"

output: 001  
"ab1"  
1

coding position: 2  
bits per position: 3 (0..7)  
bits per length: 1 (1..2)  
bits per letter: 8 (0..255)

# Lempel-Ziv 77 – Finite Window

210

input:        "abaaaacabbcaaaa"

output:       0011  
              "ab10"  
              11

coding position:     3  
bits per position:   3   (0..7)  
bits per length:     1   (1..2)  
bits per letter:     8   (0..255)

# Lempel-Ziv 77 – Finite Window

3210

input:            "abaaaacab**bb**caaaa"

output:           00111  
                  "ab101"  
                      112

coding position:    4  
bits per position:  3   (0..7)  
bits per length:   1   (1..2)  
bits per letter:    8   (0..255)

# Lempel-Ziv 77 – Finite Window

543210

input:            "abaaaacabbcaaaa"

output:           001110  
                  "ab101c"  
                      112

coding position:    6  
bits per position:  3  (0..7)  
bits per length:    1  (1..2)  
bits per letter:    8  (0..255)

# Lempel-Ziv 77 – Finite Window

6543210

input:            "abaaaacabbcaaaa"

output:           0011101  
                  "ab101c6"  
                  112 2

coding position:    7  
bits per position:  3  (0..7)  
bits per length:   1  (1..2)  
bits per letter:    8  (0..255)

# Lempel-Ziv 77 – Finite Window

76543210

input: "abaaaacabbcaaaa"

output: 00111011  
"ab101c60"  
112 21

coding position: 9  
bits per position: 3 (0..7)  
bits per length: 1 (1..2)  
bits per letter: 8 (0..255)



# Lempel-Ziv 77 – Finite Window

76543210

input: "abaaacabbcaaaa"

output: 001110111  
"ab101c603"  
112 212

coding position: 10  
bits per position: 3 (0..7)  
bits per length: 1 (1..2)  
bits per letter: 8 (0..255)

# Lempel-Ziv 77 – Finite Window

76543210

input: "abaaaacabbcaaaa"

output: 0011101111  
"ab101c6037"  
112 2122

coding position: 12  
bits per position: 3 (0..7)  
bits per length: 1 (1..2)  
bits per letter: 8 (0..255)

# Lempel-Ziv 77 – Finite Window

76543210

input:           "abaaacabbcaaaa"

output:           00111011111  
                  "ab101c60370"  
                      112 21221

coding position:     14  
bits per position:   3   (0..7)  
bits per length:     1   (1..2)  
bits per letter:     8   (0..255)

# Lempel-Ziv 77 – Results

output:

00111011111  
"ab101c60370"  
112 21221

output in binary:

0 01100001 0 01100010 1 001 0 1 000 0 1 001 1  
0 01100011 1 110 1 1 000 0 1 011 1 1 111 1  
1 000 0

Bits of the original message:  $15 \cdot 8 = 120$

Bits of the compressed message:  $3 \cdot 8 + 8 \cdot (3+1) + 12 \cdot 1 = 64$

Compression ratio: 0.53

We losslessly compressed the message to 53% of its length.

# Lempel-Ziv-Welch Algorithm

Terry Welch invented a modification of LZ77 in 1984 called LZW. It is the first fully dictionary based method.

Extends letter alphabet of a certain number of bits.

For instance:

each 8-bit letter (0..255) within a message extends to 9 bits (0..511), where:

(0..255)      are considered an original letter,  
(256..511)    are considered links to a created dictionary.

# Statistical Coders

(aka Entropy Coders)

# What's Entropy?

Let's have an alphabet  $L=\{l_0, l_1, \dots, l_{N-1}\}$ , of  $N$  letters.

When the probability distribution  $P=\{p_0, p_1, \dots, p_{N-1}\}$  is known, then the amount of information needed to encode a letter is  $H(l_n)$ , the entropy of the alphabet is  $H(L)$ :

$$H(l_n) = -\log_2 p_n, \quad H(L) = \sum_{n=0}^{N-1} p_n H(l_n),$$

$$H(L) = -\sum_{n=0}^{N-1} p_n \log_2 p_n$$

# Shannon-Fano Coder

input message: "2ab4a5cabbc4a"

order	letter	count	p(letter)	H(letter)
1	a	4	0.30	1.70
2	b	3	0.23	2.11
3	c	2	0.15	2.70
4	4	2	0.15	2.70
5	2	1	0.07	3.70
6	5	1	0.07	3.70



# Shannon-Fano Coder

"2ab4a5cabbc4a"

"ab"		"c425"
7		6

letter	count
a	4
b	3
c	2
4	2
2	1
5	1

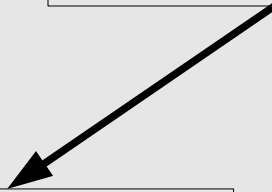
# Shannon-Fano Coder

"2ab4a5cabbc4a"

letter	count
a	4
b	3
c	2
4	2
2	1
5	1

"ab" | "c425"  
7                      6

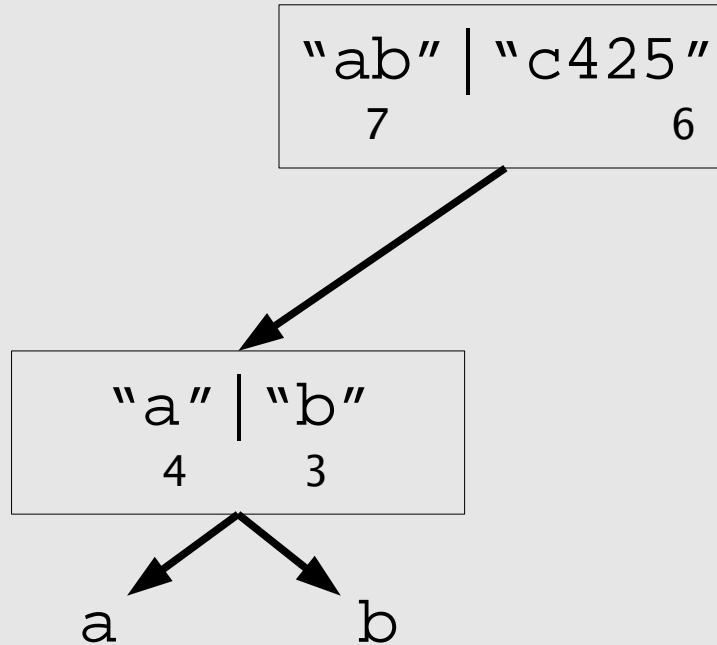
"a" | "b"  
4                      3



# Shannon-Fano Coder

"2ab4a5cabbc4a"

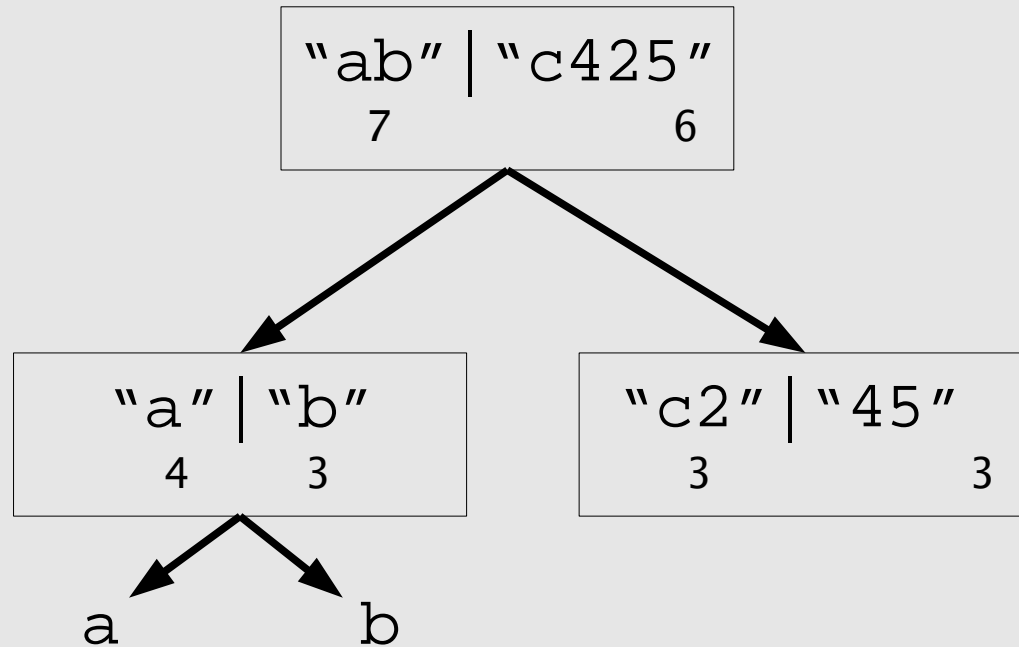
letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder

"2ab4a5cabbc4a"

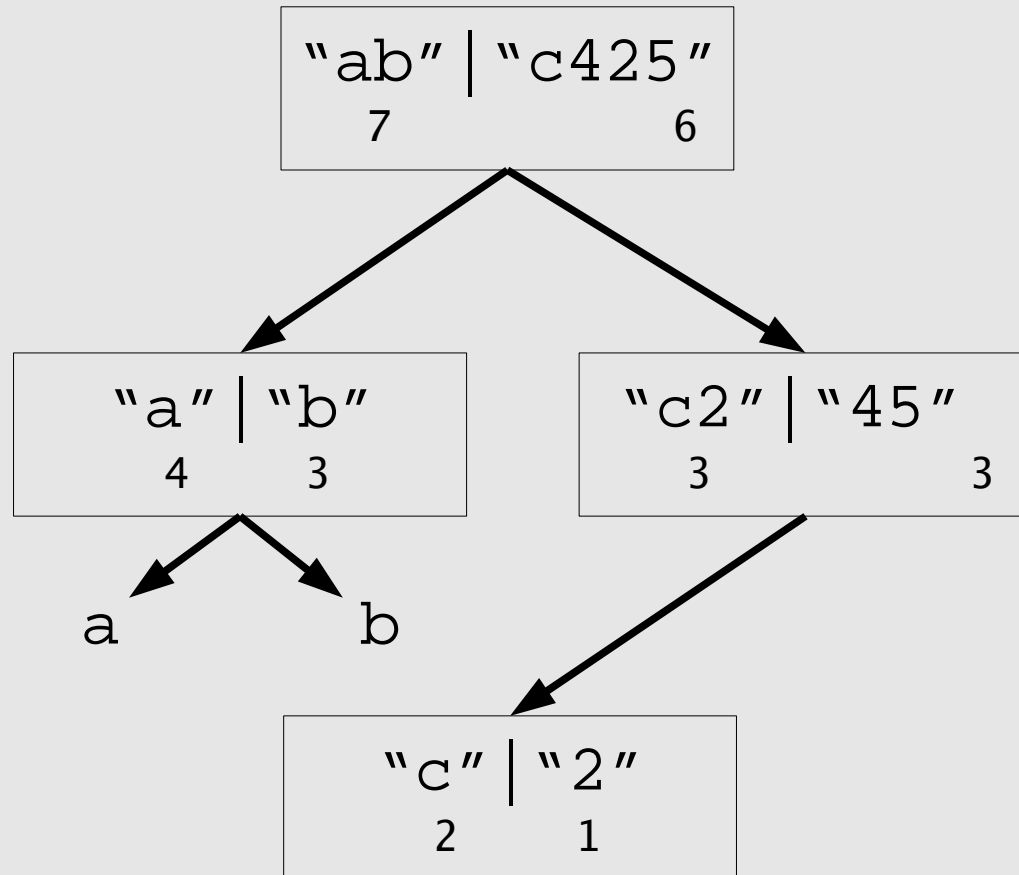
letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder

"2ab4a5cabbc4a"

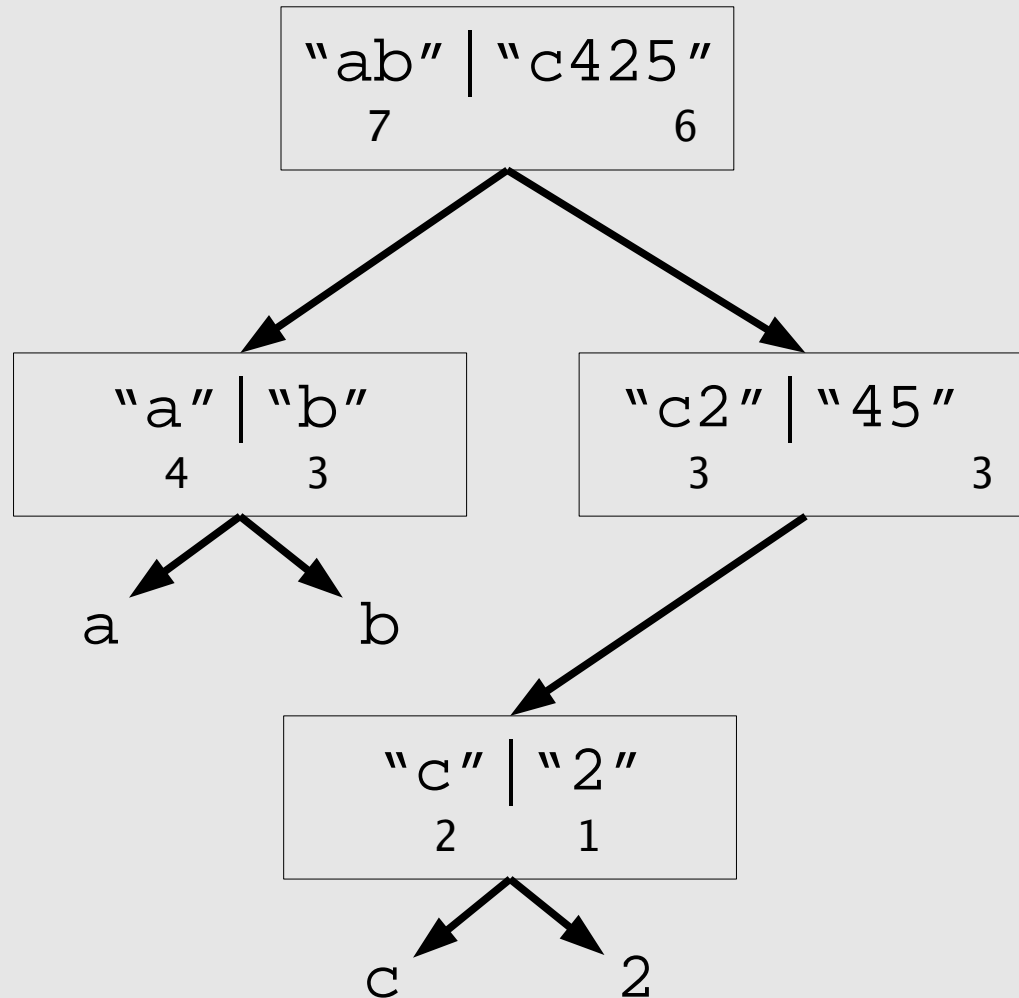
letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder

"2ab4a5cabbc4a"

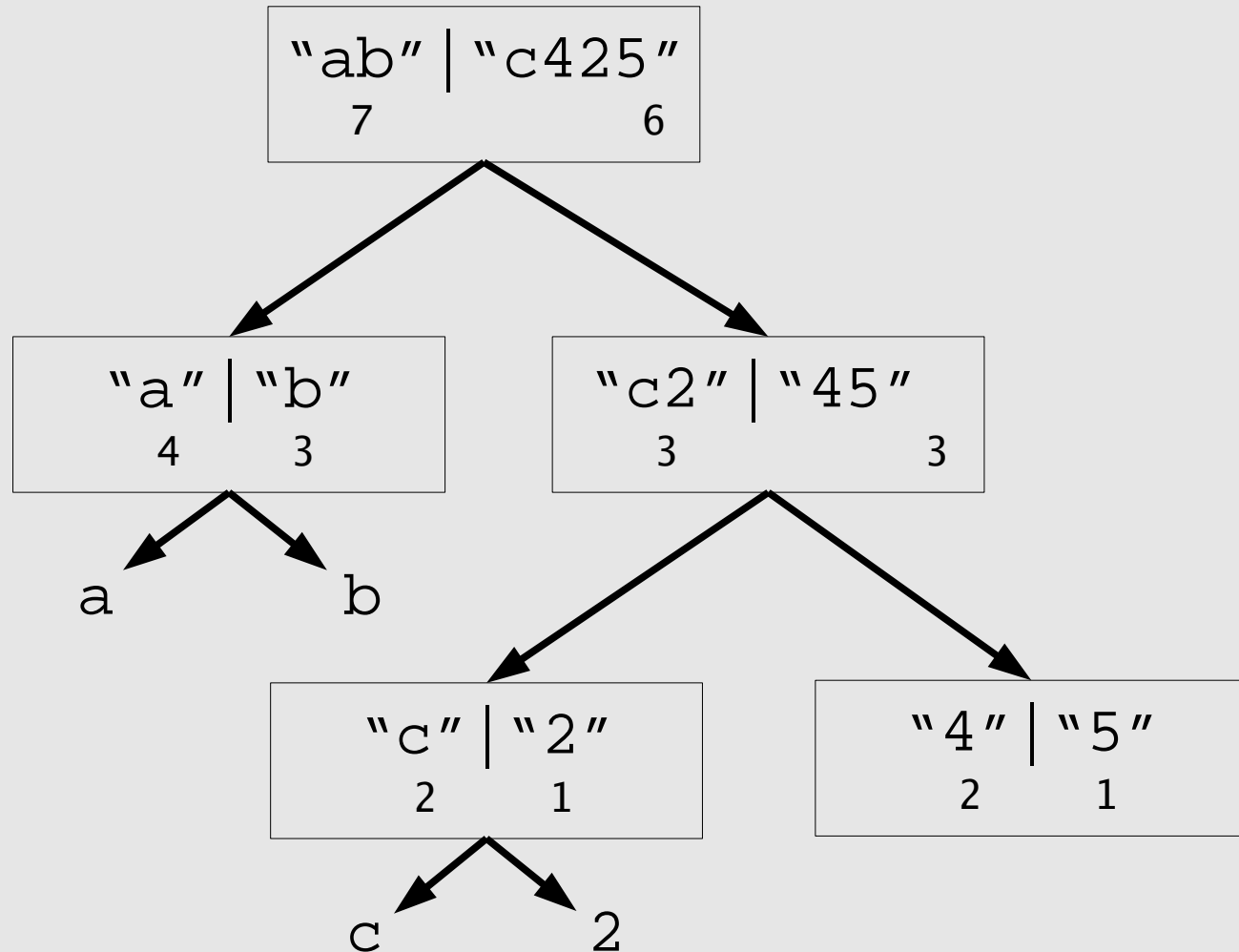
letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder

"2ab4a5cabbc4a"

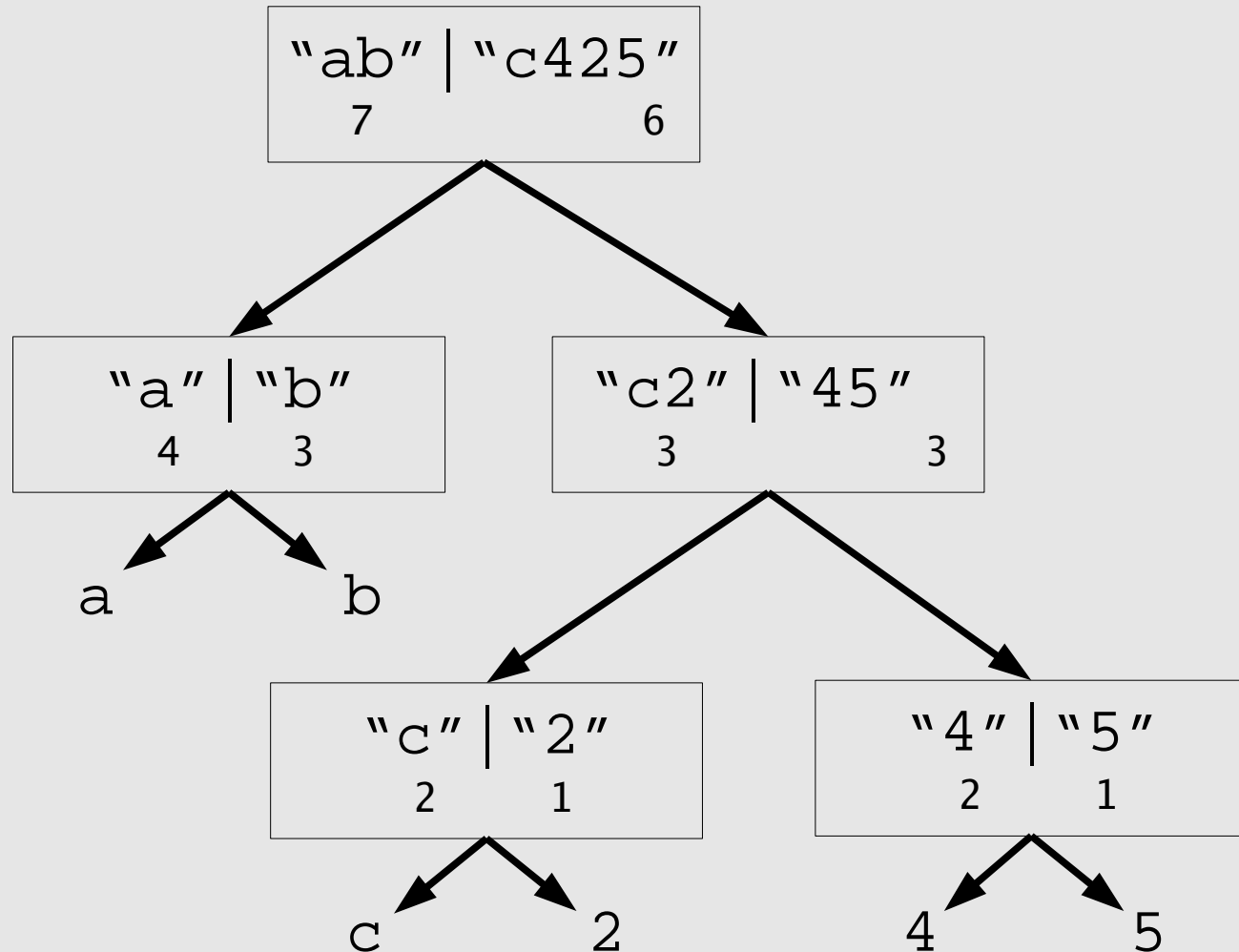
letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder

"2ab4a5cabbc4a"

letter	count
a	4
b	3
c	2
4	2
2	1
5	1

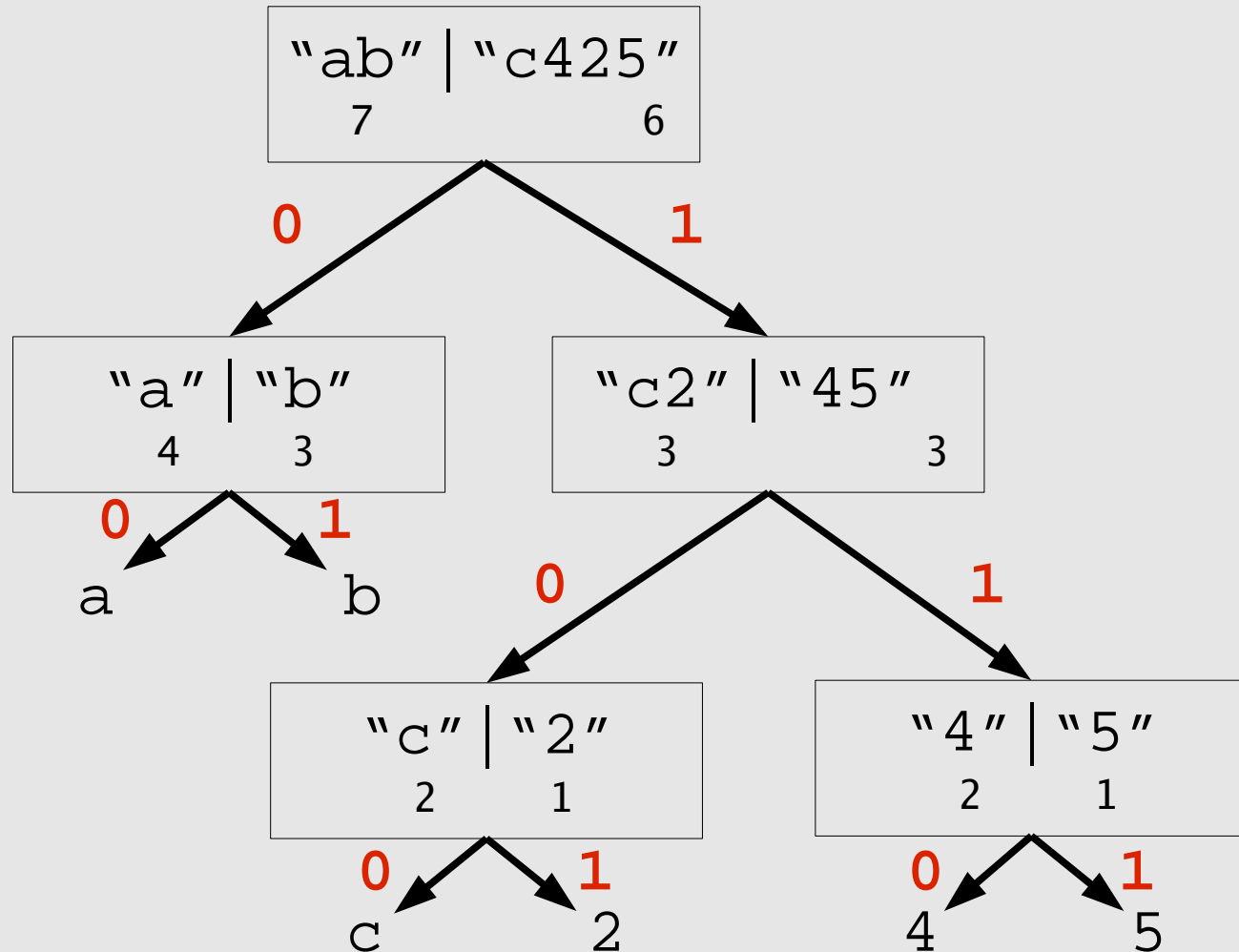




# Shannon-Fano Coder

"2ab4a5cabbc4a"

letter	count
a	4
b	3
c	2
4	2
2	1
5	1



# Shannon-Fano Coder – Results

input message: "2ab4a5cabbc4a"

order	l	count	p(l)	H(l)	code(l)	bits(code(l))	$\Sigma$ bits
1	a	4	0.30	1.70	00	2	8
2	b	3	0.23	2.11	01	2	6
3	c	2	0.15	2.70	100	3	6
4	4	2	0.15	2.70	110	3	6
5	2	1	0.07	3.70	101	3	3
6	5	1	0.07	3.70	111	3	3
				H=2.76			32

# Shannon-Fano Coder – Results

input message: "2ab4a5cabbc4a"

l	a	b	c	4	2	5
kód(l)	00	01	100	110	101	111

2 a b 4 a 5 c a b b c 4 a

101 00 01 110 00 111 100 00 01 01 100 110 00

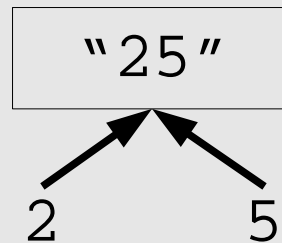
10100011100011110000010110011000

... each letter is represented by an integer bit count

# Huffman Coder

"2ab4a5cabbc4a"

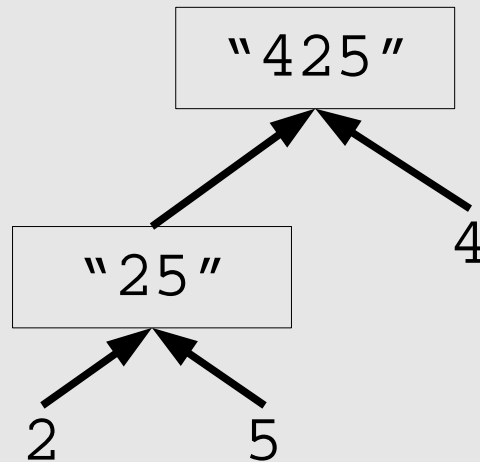
L	n	L	n
a	4	a	4
b	3	b	3
c	2	c	2
4	2	4	2
2	1	25	2
5	1		



# Huffman Coder

"2ab4a5cabbc4a"

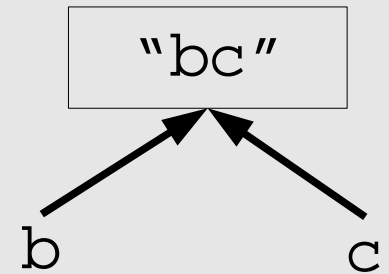
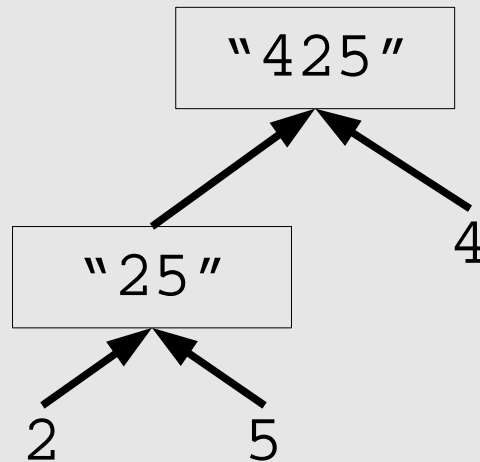
L	n	L	n
a	4	a	4
b	3	425	4
c	2	b	3
4	2	c	2
25	2		



# Huffman Coder

"2ab4a5cabbc4a"

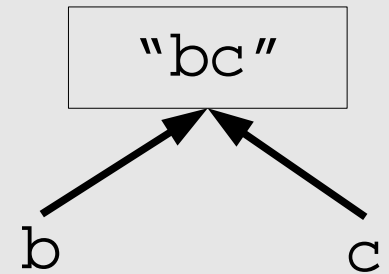
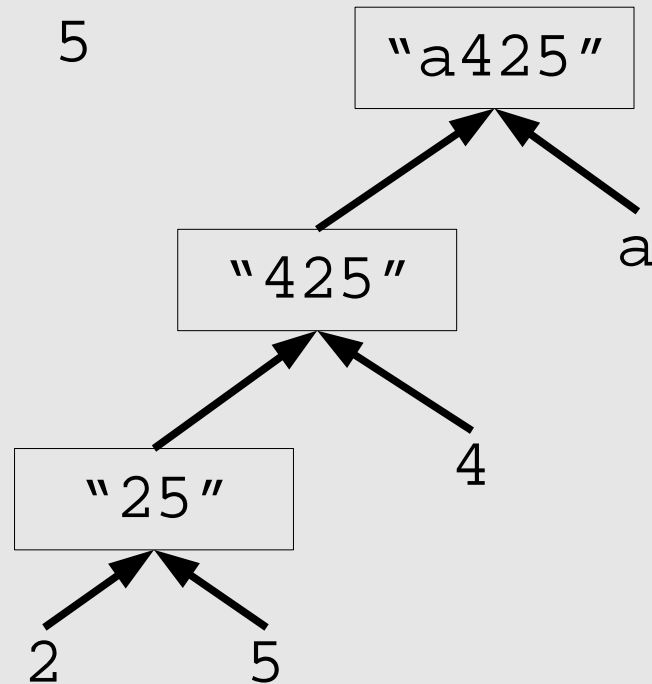
L	n	L	n
a	4	bc	5
425	4	a	4
b	3	425	4
c	2		



# Huffman Coder

"2ab4a5cabbc4a"

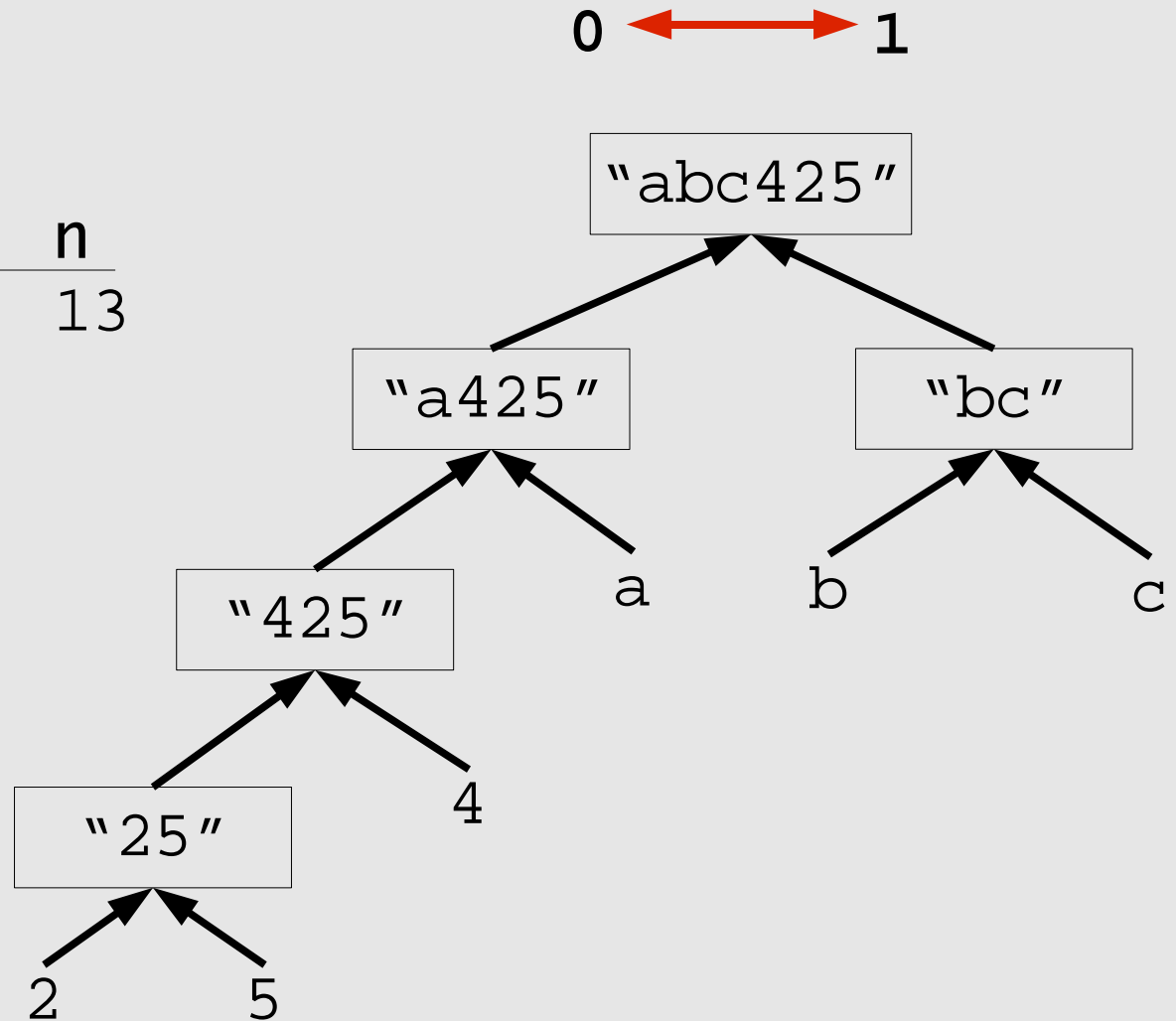
L	n	L	n
bc	5	a425	8
a	4	bc	5
425	4		



# Huffman Coder

"2ab4a5cabbc4a"

L	n	L	n
a425	8	abc425	13
bc	5		





# Huffman Coder – Results

input message: "2ab4a5cabbc4a"

pořadí	l	četnost	$p(l)$	$H(l)$	kód(l)	bitů(kód(l))	$\Sigma$ bitů
1	a	4	0.30	1.70	01	2	8
2	b	3	0.23	2.11	10	2	6
3	c	2	0.15	2.70	11	2	4
4	4	2	0.15	2.70	001	3	6
5	2	1	0.07	3.70	0000	4	4
6	5	1	0.07	3.70	0001	4	4
				H=2.76			32

# Huffman Coder – Results

input message: "2ab4a5cabbc4a"

l	a	b	c	4	2	5
code(l)	01	10	11	001	0000	0001

2      a    b    4      a    5      c    a    b    b    c    4      a

0000 01 10 001 01 0001 11 01 10 10 11 001 01

00000110001010001110110101100101

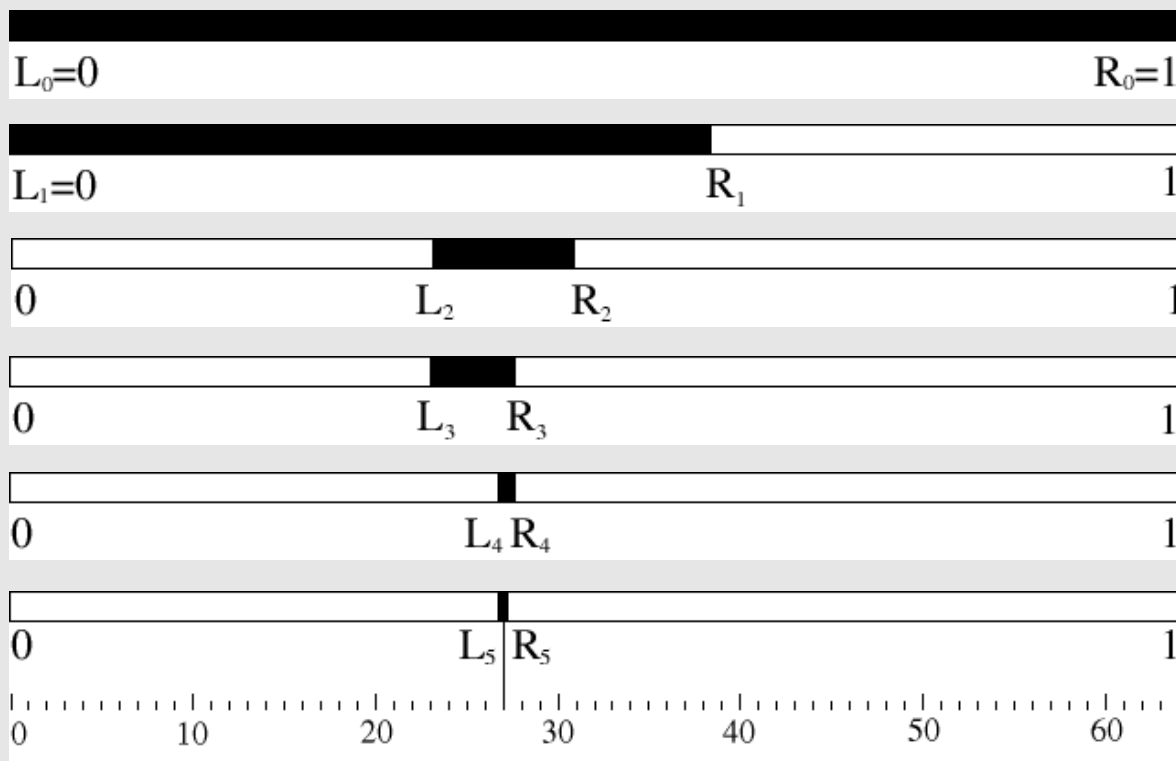
... each letter has to have an integer bit count

... quick,  $O(N \log N)$

# Arithmetic Coder

input message: "abaca"

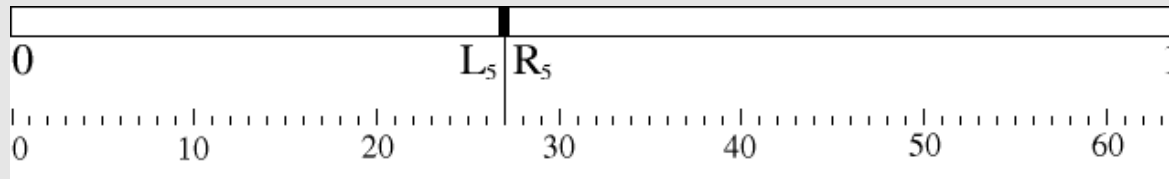
$$\mathbf{a} = \langle 0, \frac{3}{5} \rangle, \mathbf{b} = \langle \frac{3}{5}, \frac{4}{5} \rangle, \mathbf{c} = \langle \frac{4}{5}, 1 \rangle.$$



# Arithmetic Coder

input message: "abaca"

$$\mathbf{a} = \langle 0, \frac{3}{5} \rangle, \mathbf{b} = \langle \frac{3}{5}, \frac{4}{5} \rangle, \mathbf{c} = \langle \frac{4}{5}, 1 \rangle.$$



$i$	$z_i$	$L_i$	$R_i$	$l = R_i - L_i$	$L(z_i)$	$R(z_i)$	$L_{i+1} = L_i + lL(z_i)$	$R_{i+1} = L_i + lR(z_i)$
0	a	0	1	1	0	$\frac{3}{5}$	0	$\frac{3}{5}$
1	b	0	$\frac{3}{5}$	$\frac{3}{5}$	$\frac{3}{5}$	$\frac{4}{5}$	$\frac{9}{25}$	$\frac{12}{25}$
2	a	$\frac{9}{25}$	$\frac{12}{25}$	$\frac{3}{25}$	0	$\frac{3}{5}$	$\frac{9}{25}$	$\frac{54}{125}$
3	c	$\frac{9}{25}$	$\frac{54}{125}$	$\frac{9}{125}$	$\frac{4}{5}$	1	$\frac{261}{625}$	$\frac{54}{125}$
4	a	$\frac{261}{625}$	$\frac{54}{125}$	$\frac{9}{625}$	0	$\frac{3}{5}$	$\frac{261}{625}$	$\frac{1332}{3125}$

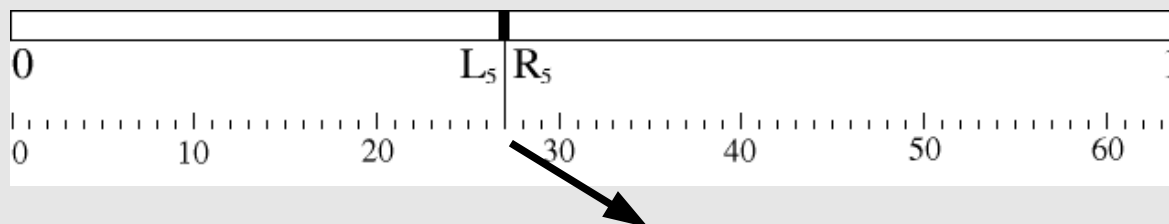
$$p_a p_b p_a p_c p_a = \frac{3}{5} \frac{1}{5} \frac{3}{5} \frac{1}{5} \frac{3}{5} = \frac{1332}{3125} - \frac{261}{625} = \frac{27}{3125}.$$

$$\langle \frac{261}{625}, \frac{1332}{3125} \rangle$$

# Arithmetic Coder

input message: "abaca"

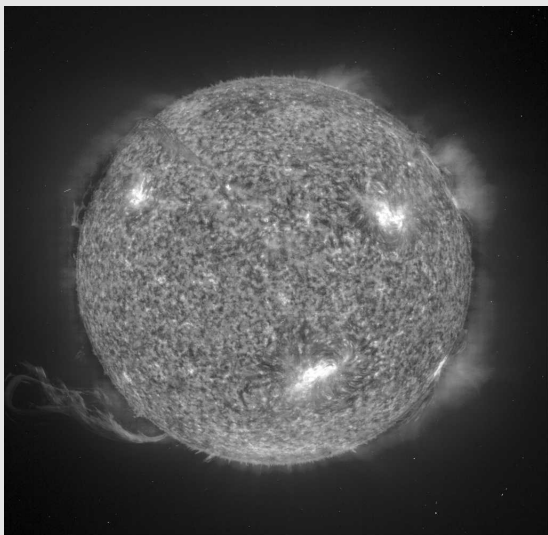
$$a = \langle 0, \frac{3}{5} \rangle, b = \langle \frac{3}{5}, \frac{4}{5} \rangle, c = \langle \frac{4}{5}, 1 \rangle.$$



- ... output of the arithmetic coder is 27 (**11011**)
- ... this only number defines message "abaca" unambiguously
- ... it is necessary to store number of encoded letters as well as a histogram
- ... it doesn't use integer counts for single letter

Adaptive scheme?

# Arithmetic Coder – Finite Context



počet pixelů: 1 048 576

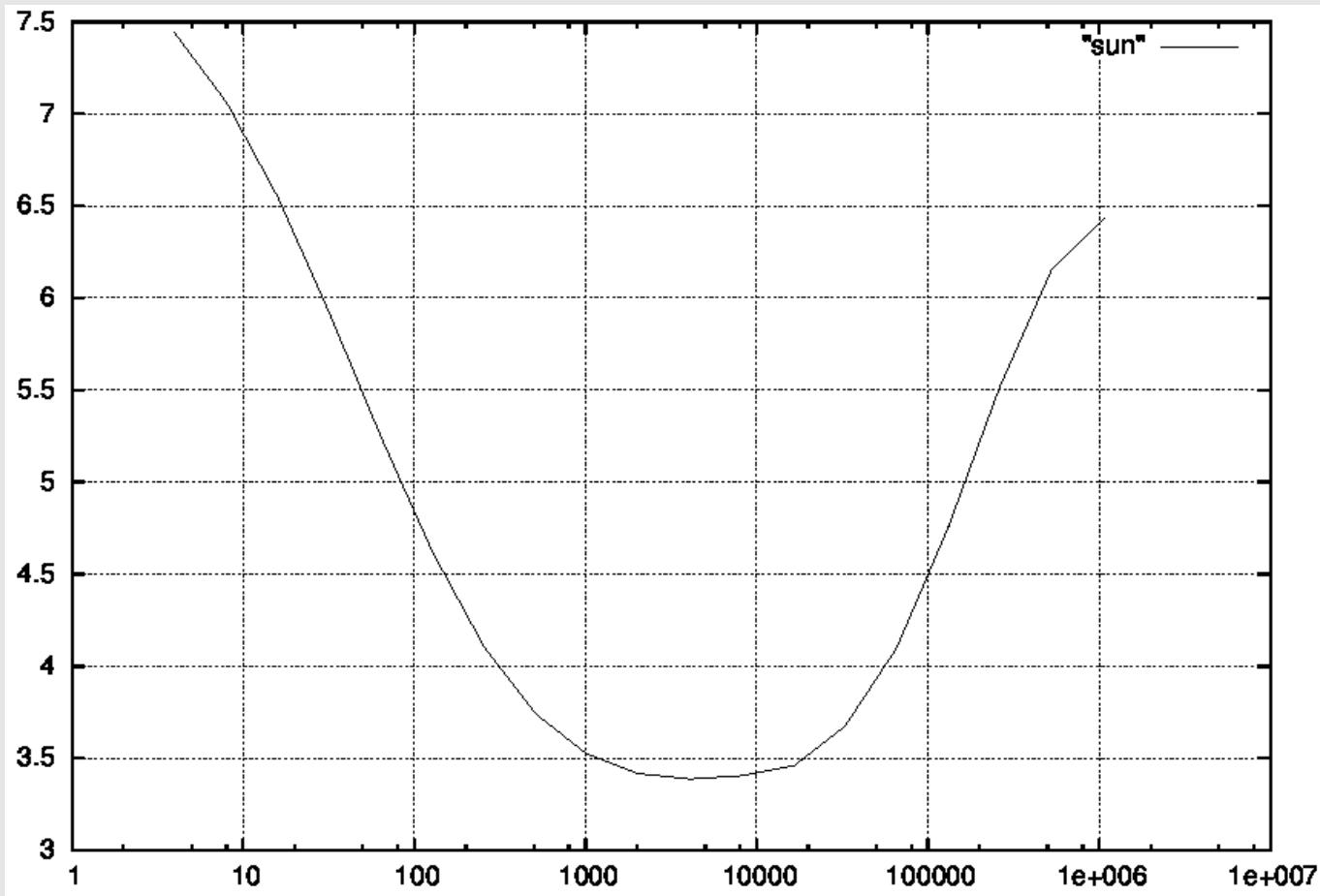
literálů: 135

klasicky: 6.45 b/l

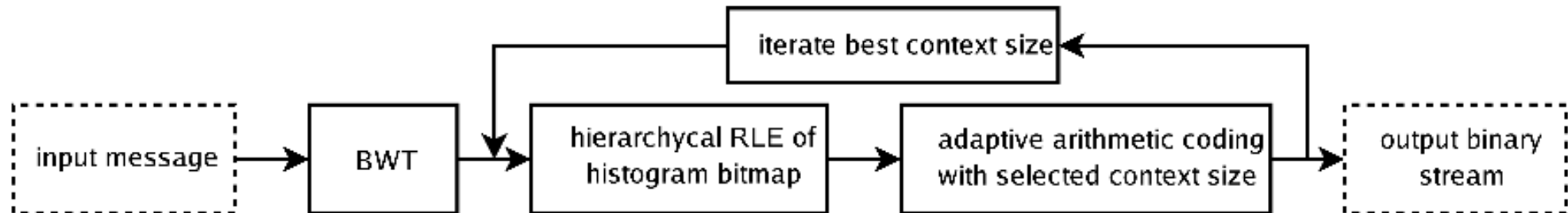
velikost: 845 571 B

kontextově: 3.38 b/l

velikost: 444 014 B



# Combined Methods



Name	File size [B]	Entropy [b]	Algorithm
GIF	700 206	5.342	LZW
TIFF	621 772	4.743	LZ77, Huffman
GNU zip v1.3.3	608 016	4.638	LZ77, Huffman
UNIX compress v4.2.4	604 225	4.609	LZW
RAR v3.3 beta 1	566 518	4.322	unknown
PNG	542 779	4.141	LZ77, Huffman
bzip2 v1.0.2	470 925	3.592	BWT, Huffman
proposed algorithm	444 014	3.387	BWT, Arit.

# Compression Methods Discussed

## Sequential Coders

RLE coding

BWT

Lempel-Ziv coding

LZ77

LZW

## Statistical Coders

Entropy

Shannon-Fano Coding

Huffman Coding

Arithmetic Coding

