

Module 9: Compression

CS 240 - Data Structures and Data Management

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Based on lecture notes by many previous cs240 instructors

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Data Storage and Transmission

The problem: How to store and transmit data?

Source text The original data, string of characters S from the *source alphabet* Σ_S

Coded text The encoded data, string of characters C from the *coded alphabet* Σ_C

Encoding An algorithm mapping source texts to coded texts

Decoding An algorithm mapping coded texts back to their original source text

Note: Source “text” can be any sort of data (not always text!)

Usually the coded alphabet Σ_C is just binary: $\{0, 1\}$.

Judging Encoding Schemes

We can always measure efficiency of encoding/decoding algorithms.

What other goals might there be?

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What other goals might there be?

- Processing speed
- Reliability (e.g. error-correcting codes)
- Security (e.g. encryption)
- Size

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- Processing speed
- Reliability (e.g. error-correcting codes)
- Security (e.g. encryption)
- Size

Encoding schemes that try to minimize $|C|$, the size of the coded text, perform *data compression*. We will measure the *compression ratio*:

$$\frac{|C| \cdot \lg |\Sigma_C|}{|S| \cdot \lg |\Sigma_S|}$$

Types of Data Compression

Logical vs. Physical

- **Logical Compression** uses the meaning of the data and only applies to a certain domain (e.g. sound recordings)
- **Physical Compression** only knows the physical bits in the data, not the meaning behind them

Lossy vs. Lossless

- **Lossy Compression** achieves better compression ratios, but the decoding is approximate; the exact source text S is not recoverable
- **Lossless Compression** always decodes S exactly

For media files, lossy, logical compression is useful (e.g. JPEG, MPEG)

We will concentrate on physical, lossless compression algorithms.

These techniques can safely be used for any application.

Character Encodings

Standard *character encodings* provide a matching from the source alphabet Σ_S (sometimes called a *charset*) to binary strings.

ASCII (American Standard Code for Information Interchange):

- Developed in 1963
- 7 bits to encode 128 possible characters:
“control codes”, spaces, letters, digits, punctuation
- Not well-suited for non-English text:
ISO-8859 extends to 8 bits, handles most Western languages

Other (earlier) codes: Morse code, Baudot code

Decoding Dictionary

To decode ASCII, we look up each 7-bit pattern in a table.

More generally, we might have a *decoding dictionary* D mapping $\Sigma_C^* \rightarrow \Sigma_S^*$:

- Must be **prefix-free** (why?)
- Might be used and stored explicitly (e.g. as a trie), or only implicitly
- Might be agreed in advance (*fixed*),
stored alongside the message (*static*),
or stored implicitly within the message (*adaptive*)

ASCII is called a *fixed-length code* because every key string in D has the same length (7 bits)

Variable-Length Codes

Definition: Different key strings in D have different lengths

The UTF-8 encoding of Unicode provides a simple example:

- Encodes any Unicode character (more than 107,000 characters) using 1-4 bytes
- Every ASCII character is encoded in 1 byte with leading bit 0, followed by the 7 bits for ASCII
- Otherwise, the first byte starts with 1-4 1's indicating the total number of bytes, followed by a 0 and 3-5 bits. The remaining bytes each start with 10 followed by 6 bits.

Char. number range (hexadecimal)	UTF-8 octet sequence (binary)
0000 0000–0000 007F	0xxxxxxx
0000 0080–0000 07FF	110xxxxx 10xxxxxx
0000 0800–0000 FFFF	1110xxxx 10xxxxxx 10xxxxxx
0001 0000–0010 FFFF	11110xxx 10xxxxxx 10xxxxxx 10xxxxxx

Character Frequency

Observation: Some letters in Σ occur more often than others.
So let's use shorter codes for more frequent characters.

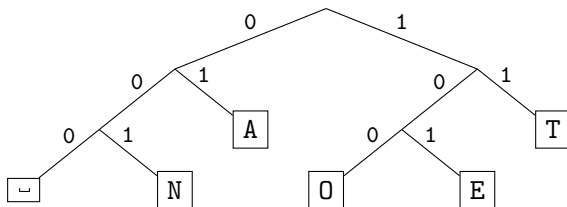
For example, the frequency of letters in typical English text is:

e	12.70%	d	4.25%	p	1.93%
t	9.06%	l	4.03%	b	1.49%
a	8.17%	c	2.78%	v	0.98%
o	7.51%	u	2.76%	k	0.77%
i	6.97%	m	2.41%	j	0.15%
n	6.75%	w	2.36%	x	0.15%
s	6.33%	f	2.23%	q	0.10%
h	6.09%	g	2.02%	z	0.07%
r	5.99%	y	1.97%		

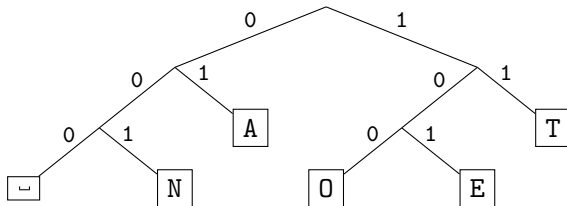
Huffman Coding

- Source alphabet is arbitrary (say Σ), coded alphabet is $\{0,1\}$
- We build a binary trie to store the decoding dictionary D
- Each character of Σ is a leaf of the trie

Example: $\Sigma = \{\text{AENOT}_\perp\}$

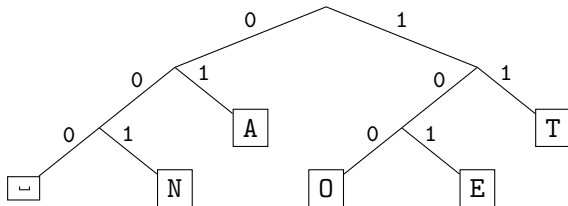


Huffman Encoding/Decoding Example



- Encode AN₀ANT
- Decode 1110000001010111

Huffman Encoding/Decoding Example



- Encode AN ANT \rightarrow 010010000100111
- Decode 111000001010111 \rightarrow TO EAT

Building the best trie

For a given source text S , how to determine the “best” trie which minimizes the length of C ?

- 1 Determine the frequency of each character $c \in \Sigma$ in S
- 2 Make $|\Sigma|$ height-0 tries holding each character $c \in \Sigma$.
Assign a “weight” to each trie: sum of frequencies of all letters in trie (initially, these are just the character frequencies)
- 3 Merge two tries with the least weights, new weight is their sum (corresponds to adding one bit to the encoding of each character)
- 4 Repeat Step 3 until there is only 1 trie left; this is D .

What data structure should we store the tries in to make this efficient?

Building the best trie

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- 4 Repeat Step 3 until there is only 1 trie left; this is D .

What data structure should we store the tries in to make this efficient?

A min-ordered heap! Step 3 is two *delete-mins* and one *insert*

Building trie example

Example text: LOSSLESS

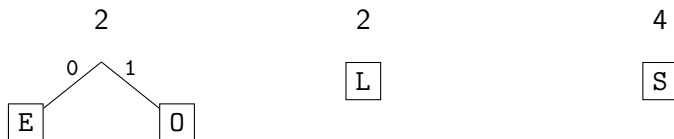
Character frequencies: E : 1, L : 2, O : 1, S : 4

1	2	1	4
E	L	O	S

Building trie example

Example text: LOSSLESS

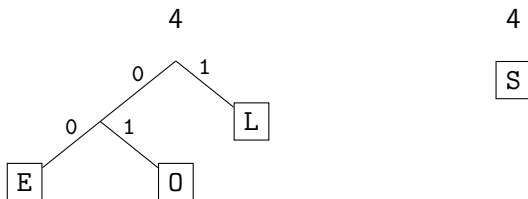
Character frequencies: E : 1, L : 2, O : 1, S : 4



Building trie example

Example text: LOSSLESS

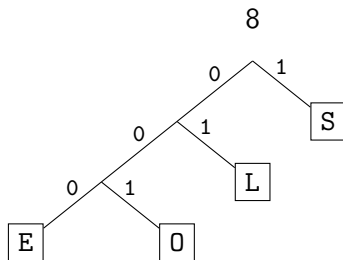
Character frequencies: E : 1, L : 2, O : 1, S : 4



Building trie example

Example text: LOSSLESS

Character frequencies: E : 1, L : 2, O : 1, S : 4

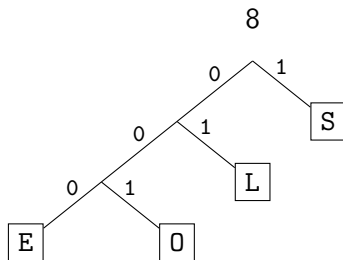


LOSSLESS →

Building trie example

Example text: LOSSLESS

Character frequencies: E : 1, L : 2, O : 1, S : 4



LOSSLESS \rightarrow 01001110100011

Compression ratio: $14/16 \approx 88\%$

Huffman Coding Summary

- Encoder must do lots of work:
 - ① Build decoding trie (one pass through S , cost is $O(|S| + |\Sigma| \log |\Sigma|)$)
 - ② Construct *encoding* dictionary from the trie mapping $\Sigma \rightarrow \{0,1\}^*$
 - ③ Encode $S \rightarrow C$ (second pass through S)
- Note: constructed trie is **not necessarily unique** (why?)
- Decoding trie must be transmitted along with the coded text C
- Decoding is faster; this is an *asymmetric* scheme.
- The constructed trie is an *optimal* one that will give the shortest C (we will not go through the proof)
- Huffman is the best we can do for encoding one character at a time.

Run-Length Encoding

- Variable-length code with a fixed decoding dictionary, but one which is not explicitly stored.
- Not a character-encoding (multiple characters represented by one dictionary-entry)
- The source alphabet and coded alphabet are both binary: $\{0, 1\}$.

Observation: 0's and 1's in S may be repeated many times in a row (called a “run”).

S is encoded as the first bit of S (either 0 or 1), followed by a sequence of integers indicating run lengths. (We don't have to encode the value of each bit since it will alternate.)

Question: How to encode a run length k in binary?

Prefix-free Integer Encoding

The encoding of run-length k must be *prefix-free*, because the decoder has to know when to stop reading k .

We will encode the binary length of k in **unary**, followed by the actual value of k in **binary**.

The binary length of k is $len(k) = \lfloor \lg k \rfloor + 1$.

Since $k \geq 1$, we will encode $len(k) - 1$, which is at least 0.

The prefix-free encoding of the positive integer k is in two parts:

- 1 $\lfloor \lg k \rfloor$ copies of 0, followed by
- 2 The binary representation of k

Examples: $1 \rightarrow 1$, $3 \rightarrow 011$, $5 \rightarrow 00101$, $23 \rightarrow 000010111$

RLE Example

$S = 1111111001000000000000000000000011111111111$

$C = 1$

RLE Example

$S = 1111111001000000000000000000000011111111111$

$k = 7$

$C = 100111$

RLE Example

$S = 1111111001000000000000000000000011111111111$

$k = 2$

$C = 100111010$

RLE Example

$S = 111111100$ **1** $00000000000000000000000011111111111$

$k = 1$

$C = 100111010$ **1**

RLE Example

$S = 11111110010000000000000000000011111111111$

$k = 20$

$C = 1001110101000010100$

RLE Example

$S = 11111110010000000000000000000000$ 1111111111

 $k = 11$

$C = 10011101010000101000001011$

RLE Example

$S = 11111111001000000000000000000000111111111111$

$C = 10011101010000101000001011$

Compression ratio: $26/41 \approx 63\%$

RLE Properties

- Compression ratio could be smaller than 1%
- Usually, we are not that lucky:
 - ▶ No compression until run-length $k \geq 6$
 - ▶ **Expansion** when run-length $k = 2$ or 4
- Method can be adapted to larger alphabet sizes
- Used in some image formats (e.g. TIFF)

Adaptive Dictionaries

ASCII, UTF-8, and RLE use *fixed* dictionaries.

In Huffman, the dictionary is not fixed, but it is *static*: the dictionary is the same for the entire encoding/decoding.

Properties of *adaptive encoding*:

- There is an initial dictionary D_0 . Usually this is fixed.
- For $i \geq 0$, D_i is used to determine the i 'th output character
- After writing the i 'th character to output, both encoder and decoder update D_i to D_{i+1}

Note that both encoder and decoder must have the same information. Usually encoding and decoding algorithms will have the same cost.

Longer Patterns in Input

Huffman and RLE mostly take advantage of frequent or repeated *single characters*.

Observation: Certain *substrings* are much more frequent than others.

Examples:

- English text:
Most frequent digraphs: TH, ER, ON, AN, RE, HE, IN, ED, ND, HA
Most frequent trigraphs: THE, AND, THA, ENT, ION, TIO, FOR, NDE
- HTML: "<a href", "<img src", "
"
- Video: repeated background between frames, shifted sub-image

Lempel-Ziv

Lempel-Ziv is a family of *adaptive* compression algorithms.

Main Idea: Each character in the coded text C either refers to a single character in Σ_S , or a *substring* of S that both encoder and decoder have already seen.

Variants:

- LZ77** Original version (“sliding window”)
Derivatives: LZSS, LZFG, LZRW, LZW, DEFLATE, ...
DEFLATE used in (pk)zip, gzip, PNG
- LZ78** Second (slightly improved) version
Derivatives: LZW, LZMW, LZAP, LZJ, ...
LZW used in compress, GIF

Lempel-Ziv

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LZW used in compress, GIF
Patent issues!

LZW Overview

- Fixed-width encoding using k bits (e.g. $k = 12$).
Store decoding dictionary with 2^k entries.
- First $|\Sigma_S|$ entries are for single characters,
remaining entries involve multiple characters
- After encoding/decoding a substring y of S , add xc to D ,
where x is previously encoded/decoded substring of S ,
and c is the first character of y
Note: start adding to D after second substring of S is
encoded/decoded

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

$C =$

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	
129	
130	
131	
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

$C =$ ^Y
89

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	
129	
130	
131	
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y O
C = 89 79

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	
130	
131	
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y O !
C = 89 79 33

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	
131	
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y O ! _
C = 89 79 33 32

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y O ! _ YO
C = 89 79 33 32 128

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U
$C = 89$	79	33	32	128	85

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_
$C = 89$	79	33	32	128	85	130

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_	YOU
$C = 89$	79	33	32	128	85	130	132

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_	YOU	R
$C = 89$	79	33	32	128	85	130	132	82

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	YOUR
136	
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_	YOU	R	_Y
$C = 89$	79	33	32	128	85	130	132	82	131

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	YOUR
136	R_
137	
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_	YOU	R	_Y	O
$C = 89$	79	33	32	128	85	130	132	82	131	79

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	YOUR
136	R_
137	_YO
138	
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y	O	!	_	YO	U	!_	YOU	R	_Y	O	YO
$C = 89$	79	33	32	128	85	130	132	82	131	79	128

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	YOUR
136	R_
137	_YO
138	OY
139	

LZW Example

Input: YO!_YOU!_YOUR_YOYO!

Σ_S = ASCII character set (0–127)

Y O ! _ YO U !_ YOU R _Y O YO !
 $C = 89$ 79 33 32 128 85 130 132 82 131 79 128 **33**

$D =$

Code	String
...	
32	_
33	!
...	
79	O
...	
82	R
...	
85	U
...	
89	Y
...	

Code	String
128	YO
129	O!
130	!_
131	_Y
132	YOU
133	U!
134	!_Y
135	YOUR
136	R_
137	_YO
138	OY
139	YO!

LZW compresor

LZW-encode(S)

1. $w \leftarrow \text{NIL}$
2. **while** there is input in S **do**
3. $K \leftarrow$ next symbol from S
4. **if** wK exists in the dictionary
5. $w \leftarrow wK$
6. **else**
7. output $\text{index}(w)$
8. add wK to the dictionary
9. $w \leftarrow K$

LZW decoding

- Same idea: build dictionary while reading string.
- Example:

$D =$

Code #	String
...	
32	~
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67

$D =$

Code #	String
...	
32	~
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65

$D =$

Code #	String
...	
32	~
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65 78

$D =$

Code #	String
...	
32	~
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65 78 32

$D =$

Code #	String
...	
32	␣
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N
32	␣	130	N␣	78,␣

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65 78 32 **66**

$D =$

Code #	String
...	
32	_
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N
32	_	130	N_	78, _
66	B	131	_B	32,B

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65 78 32 66 **129**

$D =$

Code #	String
...	
32	_
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N
32	_	130	N_	78, _
66	B	131	_B	32,B
129	AN	132	BA	66,A

LZW decoding

- Same idea: build dictionary while reading string.
- Example: 67 65 78 32 66 129 **133**

$D =$

Code #	String
...	
32	_
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N
32	_	130	N_	78, _
66	B	131	_B	32,B
129	AN	132	BA	66,A
133	???	133		

LZW decoding: the catch

- In this example: Want to decode 133, but not yet in dictionary!
- Generally: decoder is “one step behind” in creating dictionary
- So problem occurs if we want to use a code that we’re about to build.
- But then we actually know what’s going on!
 - ▶ Say we’re seeing code-number k at the time when we’re assigning k .
 - ▶ Encoder assigned $k \rightarrow s_{prev} + s[0]$
 - ▶ Decoder knows s_{prev} (decoded in previous step)
 - ▶ Decoder wants s , i.e., what k encodes, i.e., $s_{prev} + s[0]$
 - ▶ Therefore $s[0] = \text{first character of } s_{prev}$
 - ▶ Therefore $s = s_{prev} + s_{prev}[0]$

LZW decoding pseudocode

LZW-decode(S)

1. $D \leftarrow$ dictionary that maps $\{0, \dots, 127\}$ to ASCII
2. $idx \leftarrow 128$
3. $code \leftarrow$ first code from S
4. $s \leftarrow D(code)$; output s
5. **while** there are more codes in S **do**
6. $s_{prev} \leftarrow s$
7. $code \leftarrow$ next code of S
8. **if** $code == idx$ **do**
9. $s \leftarrow s_{prev} + s_{prev}[0]$
10. **else**
11. $s \leftarrow D(code)$
12. $D.insert(idx, s_{prev} + s[0])$
13. $idx \leftarrow idx + 1$

LZW decoding example revisited

- Example: 67 65 78 32 66 129 133 83

$D =$

Code #	String
...	
32	␣
...	
...	
65	A
66	B
67	C
...	
78	N
...	
83	S
...	

input	decodes to	Code #	String (human)	String (computer)
67	C			
65	A	128	CA	67,A
78	N	129	AN	65,N
32	␣	130	N␣	78,␣
66	B	131	␣B	32,B
129	AN	132	BA	66,A
133	ANA	133	ANA	129,A
83	S	134	ANAS	133,S

Compression summary

Huffman	run-length encoding	Lempel-Ziv-Welch
variable-length	variable-length	fixed-length
single-character	multi-character	multi-character
2-pass	1-pass	1-pass
60% compression (English)	bad on text	45% compression (English)
optimal 01-prefix-code	good if long runs (pictures)	good on English text
must send dictionary	can be worse than ASCII	can be worse than ASCII
rarely used directly part of pkzip, JPEG, MP3	rarely used directly fax machines, old picture-formats	frequently used GIF, some variants of PDF

Text transformations

For efficient compression, we need

- frequently repeating characters and/or
- frequently repeating substrings

Idea: Modify the text first so that these are likely.

Move-to-Front

Recall the MTF heuristic for self-organizing search:

- Dictionary is stored as an unsorted linked list
- After an element is accessed, move it to the front of the list.

How can we use this idea for text transformations?

Move-to-Front

Recall the MTF heuristic for self-organizing search:

- Dictionary is stored as an unsorted linked list
- After an element is accessed, move it to the front of the list.

How can we use this idea for text transformations?

Take advantage of *locality* in the data.

If we see a character now, we'll probably see it again soon.

Specifics: MTF is an *adaptive* compression algorithm.

If the source alphabet is Σ_S with size $|\Sigma_S| = m$,
then the coded alphabet will be $\Sigma_C = \{0, 1, \dots, m - 1\}$.

Move-to-Front Encoding/Decoding

MTF-encode(S)

1. $L \leftarrow$ linked list with Σ_S in some pre-agreed, fixed order
2. **while** S has more characters **do**
3. $c \leftarrow$ next character of S
4. **output** index i such that $L[i] = c$
5. Move c to position $L[0]$

Decoding works in *exactly* the same way:

MTF-decode(C)

1. $L \leftarrow$ linked list with Σ_S in some pre-agreed, fixed order
2. **while** C has more characters **do**
3. $i \leftarrow$ next integer from C
4. **output** $L[i]$
5. Move $L[i]$ to position $L[0]$

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INEFFICIENCIES

$C =$

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S = \text{INEFFICIENCIES}$

$C = 8$

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
I	A	B	C	D	E	F	G	H	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S = \text{INEFFICIENCIES}$

$C = 8 \ 13$

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
N	I	A	B	C	D	E	F	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S = \text{INEFFICIENCIES}$

$C = 8\ 13\ 6$

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
E	N	I	A	B	C	D	F	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INE**F**FICIENCIES

$C =$ 8 13 6 **7**

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
F	E	N	I	A	B	C	D	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INEFICIENCIES

$C =$ 8 13 6 7 0

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
F	E	N	I	A	B	C	D	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INEFFICIENCIES

$C =$ 8 13 6 7 0 3

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
I	F	E	N	A	B	C	D	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INEFFI**C**IENCIES

$C =$ 8 13 6 7 0 3 **6**

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
C	I	F	E	N	A	B	D	G	H	J	K	L	M	O	P	Q	R	S	T	U	V	W	X	Y	Z

$S =$ INEFFICIENCIES

$C =$ 8 13 6 7 0 3 6 1

MTF Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
S	E	I	C	N	F	A	B	D	G	H	J	K	L	M	O	P	Q	R	T	U	V	W	X	Y	Z

$S = \text{INEFFICIENCIES}$

$C = 8\ 13\ 6\ 7\ 0\ 3\ 6\ 1\ 3\ 4\ 3\ 3\ 3\ 18$

- What does a run in S encode to in C ?
- What does a run in C mean about the source S ?

MTF Compression Ratio

So far, MTF does not provide any compression *on its own* (why?)

We need to encode the integer sequence.

Two possible approaches:

MTF Compression Ratio

So far, MTF does not provide any compression *on its own* (why?)

We need to encode the integer sequence.

Two possible approaches:

- Prefix-free integer encoding (like in RLE)
- Huffman coding

Burrows-Wheeler Transform

The Burrows-Wheeler Transform is a sophisticated compression technique

- Transforms source text to a coded text with the same letters, just in a different order
- *The coded text will be more easily compressible with MTF*
- Compression algorithm does not make just a few “passes” over S . BWT is a *block* compression method.
- (As we will see) decoding is more efficient than encoding, so BWT is an *asymmetric* scheme.

BWT (followed by MTF, RLE, and Huffman)
is the algorithm used by the bzip2 program.

It achieves the best compression of any algorithm we have seen
(at least on English text).

BWT Encoding

A *cyclic shift* of a string X of length n is the concatenation of $X[i + 1..n - 1]$ and $X[0..i]$, for $0 \leq i < n$.

For Burrows-Wheeler, we assume the source text S ends with a special *end-of-word character* $\$$ that occurs nowhere else in S .

The Burrows-Wheeler Transform proceeds in three steps:

- 1 Place all *cyclic shifts* of S in a list L
- 2 Sort the strings in L lexicographically
- 3 C is the list of trailing characters of each string in L

BWT Example

$S = \text{alf_eats_alfalfa\$}$

① Write all cyclic shifts

```
alf_eats_alfalfa$  
lf_eats_alfalfa$a  
f_eats_alfalfa$al  
_eats_alfalfa$alf  
eats_alfalfa$alf_  
ats_alfalfa$alf_e  
ts_alfalfa$alf_ea  
s_alfalfa$alf_eat  
_alfalfa$alf_eats  
alfalfa$alf_eats_  
lfalfa$alf_eats_a  
falfa$alf_eats_al  
alfa$alf_eats_alf  
lfa$alf_eats_alfa  
fa$alf_eats_alfal  
a$alf_eats_alfalf  
$alf_eats_alfalfa
```

BWT Example

$S = \text{alf_eats_alfalfa\$}$

- 1 Write all cyclic shifts
- 2 Sort cyclic shifts

```
$alf_eats_alfalfa
_alfalfa$alf_eats
_eats_alfalfa$alf
a$alf_eats_alfalf
alf_eats_alfalfa$
alfa$alf_eats_alf
alfalfa$alf_eats_
ats_alfalfa$alf_e
eats_alfalfa$alf_
f_eats_alfalfa$al
fa$alf_eats_alfal
falfa$alf_eats_al
lf_eats_alfalfa$a
lfa$alf_eats_alfa
lfalfa$alf_eats_a
s_alfalfa$alf_eat
ts_alfalfa$alf_ea
```

BWT Example

$$S = \text{alf_eats_alfalfa\$}$$

- 1 Write all cyclic shifts
- 2 Sort cyclic shifts
- 3 Extract last characters from sorted shifts

C = asff\$f_ellaaata

\$alf_eats_alfalfa
_alfalfa\$alf_eats
_eats_alfalfa\$alf
a\$alf_eats_alfalf
alf_eats_alfalfa\$
alfa\$alf_eats_alf
alfalfa\$alf_eats_
ats_alfalfa\$alf_e
eats_alfalfa\$alf_
f_eats_alfalfa\$alf
fa\$alf_eats_alfalf
falfa\$alf_eats_alf
lf_eats_alfalfa\$a
lfa\$alf_eats_alfal
lfalfa\$alf_eats_alf
s_alfalfa\$alf_eat
ts_alfalfa\$alf_ea

BWT Decoding

Idea: Given C , We can generate the *first column* of the array by sorting. This tells us which character comes after each character in S .

Decoding Algorithm:

View the coded text C as an array of characters.

- 1 Make array of A of tuples $(C[i], i)$
- 2 Sort A by the characters, record integers in array N
(Note: $C[N[i]]$ follows $C[i]$ in S , for all $0 \leq i < n$)
- 3 Set j to index of \$ in C and S to empty string
- 4 Set $j \leftarrow N[j]$ and append $C[j]$ to S
- 5 Repeat Step 4 until $C[j] = \$$

BWT Decoding Example

$C = \text{ard\$rcaaaabb}$

$S =$

A	
a,	0
r,	1
d,	2
\$,	3
r,	4
c,	5
a,	6
a,	7
a,	8
a,	9
b,	10
b,	11

BWT Decoding Example

$C = \text{ard\$rcaaaabb}$

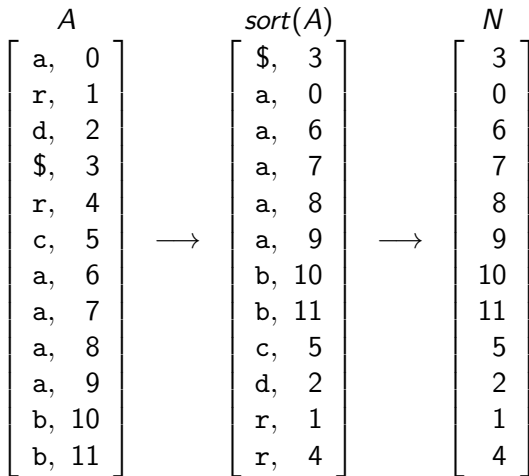
$S =$

A		\rightarrow	$\text{sort}(A)$
a,	0		\$, 3
r,	1		a, 0
d,	2		a, 6
\$,	3		a, 7
r,	4		a, 8
c,	5		a, 9
a,	6		b, 10
a,	7		b, 11
a,	8		c, 5
a,	9		d, 2
b,	10		r, 1
b,	11		r, 4

BWT Decoding Example

$C = \text{ard\$rcaaaabb}$

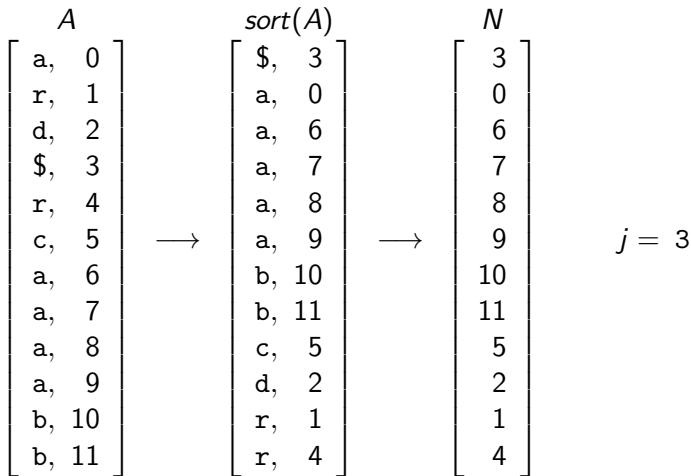
$S =$



BWT Decoding Example

$C = \text{ard\$rcaaaabb}$

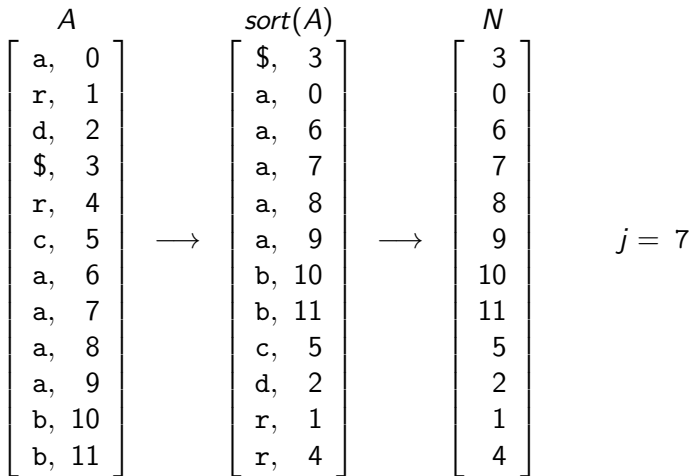
$S =$



BWT Decoding Example

$C = \text{ard\$rca}\textcolor{red}{a}\text{aabb}$

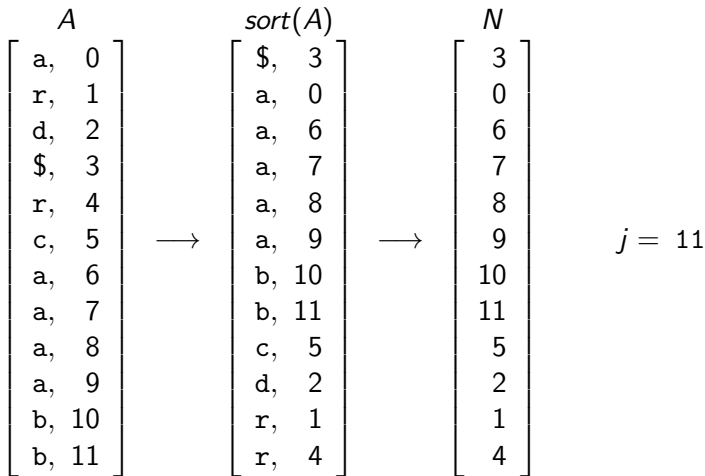
$S = \textcolor{red}{a}$



BWT Decoding Example

$C = \text{ard\$rcaaaab}$ **b**

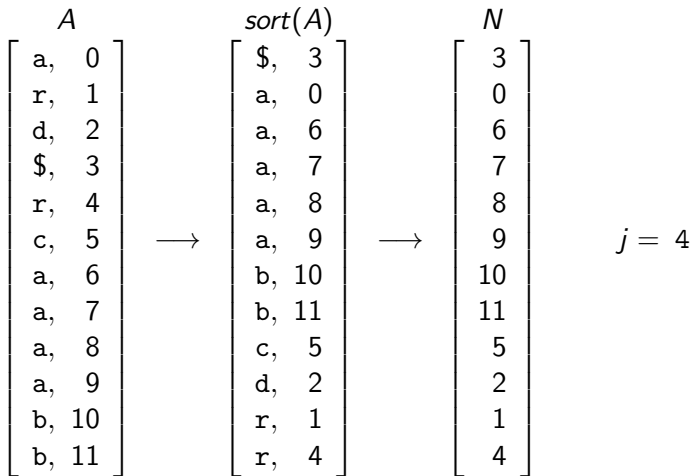
$S = \text{a}$ **b**



BWT Decoding Example

$C = \text{ard}\$ \text{rcaaaabb}$

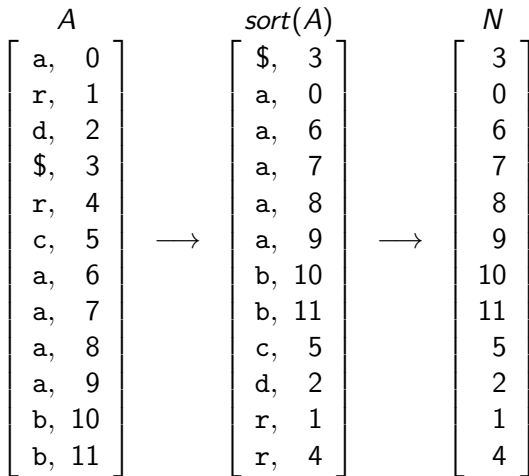
$S = \text{ab}\text{r}$



BWT Decoding Example

$C = \text{ard\$rcaaaabb}$

$S = \text{abracadabra\$}$



BWT Overview

Encoding cost: $O(n^2)$ (using radix sort)

- Sorting cyclic shifts is equivalent to sorting suffixes
- This can be done by traversing suffix trie
- Possible in $O(n)$ time

Decoding cost: $O(n)$ (faster than encoding)

Encoding and decoding both use $O(n)$ space.

Tends to be slower than other methods but (combined with MTF and Huffman) give better compression.