

Text Indexing

Lecture 04: Text-Compression

Florian Kurpicz



PINGO





https://pingo.scc.kit.edu/651997





Definition: Suffix Array [GBS92; MM93]

Given a text T of length n, the suffix array (SA) is a permutation of [1..n], such that for $i \le j \in [1..n]$

$$T[SA[i]..n] \leq T[SA[j]..n]$$

Definition: Longest Common Prefix Array

Given a text T of length n and its SA, the LCP-array is defined as

$$LCP[i] = \begin{cases} 0 & i = 1\\ \max\{\ell \colon T[SA[i]..SA[i] + \ell) = \\ T[SA[i - 1]..SA[i - 1] + \ell)\} & i \neq 1 \end{cases}$$

	1	2	3	4	5	6	7	8	9	10	11	12	13
Т	а	b	а	b	С	а	b	С	а	b	b	а	\$
SA	13	12	1	9	6	3	11	2	10	7	4	8	5
LCP	0	0	1	2	2	5	0	2	1	1	4	0	3
	\$	a \$	ababcabcabba\$	a b b a \$	a b c a b b a \$	abcabcabba\$	b a \$	babcabcabba\$	b b a \$	bcabba\$	bc a bc a b ba \$	c a b b a \$	cabcabba\$





Types of Compression

- lossy compressionaudio, video, pictures, . . .
- lossless compression
 - audio, text, ...





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- faster data transfer
- cheaper storage costs
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Why Compression



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Types of Text-Compression

- entropy coding o compress characters
- dictionary compression o compress substings

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This Lecture

- measure compressibility
- different compression algorithms
 - both types
- space/time requirements of compression algorithms
- make use of known concepts





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Given a text T of length n over an alphabet $\Sigma = [1, \sigma]$ and its histogram Hist , then

$$H_0(T) = (1/n) \sum_{i=1}^{\sigma} Hist[i] \lg(n/Hist[i])$$





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- T = abbaaacaaba
- *n* = 12
- *Hist*[a] = 7
- Hist[b] = 3Hist[c] = 1
- Hist[\$] = 1





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- $H_0(T) = (1/12)(7 \lg(12/7) + 3 \lg(12/3) + 1 \lg(12/1) + 1 \lg(12/1)) \approx 1.55$





Given a text T over an alphabet Σ and a string $S \in \Sigma^k$, T_S the concatenation of all characters that occur in T after S in text order

- \blacksquare T = abcdabceabcd
- lacksquare S = abc
- lacksquare $T_{\mathcal{S}} = \mathsf{ded}$

Definition: *k*-th Order Empirical Entropy

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k-th Order Empirical Entropy (2/2)



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PINGO can we describe a property of H_k



Example for *k***-th Order Empirical Entropy [Kur20]**

Name	σ	n	H_0	H_1	H_2	<i>H</i> ₃
Commoncrawl	243	196,885,192,752	6.19	4.49	2.52	2.08
DNA	4	218,281,833,486	1.99	1.97	1.96	1.95
Proteins	26	50,143,206,617	4.21	4.20	4.19	4.17
Wikipedia	213	246,327,201,088	5.38	4.15	3.05	2.33
SuffixArrayCC	n	137,438,953,472	$37 (= \lg n)$	0	0	0
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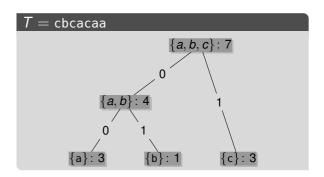
- does not measure repetitions well
- there are other measures



Huffman Coding [Huf52]



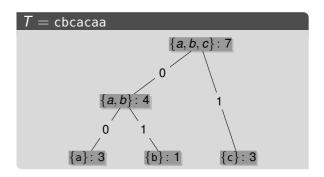
- idea is to create a binary tree
- \blacksquare each character α is a leaf and has weight $\mathit{Hist}[\alpha]$
- create node for two nodes without parent with smallest weight
- give new node total weight of children
- repeat until only one node without parent remains
- label edges:
 - left edge: 0
 - right edge: 1
- path to children gives code for character



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- codes are variable length and prefix-free
- tree/dictionary needed for decoding





- start with Huffman codes, code word 0, and length 1
- to get canonical code for current length, then add 1 to code word
- to update length add 1 and append required amount of zeros to code word



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- PINGO what are some advantages of canonical Huffman codes?

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- each character $\alpha \in \Sigma$ receives a code of length $\ell_{\alpha} = \lceil \lg \frac{n}{Hist[\alpha]} \rceil$
- show that there always exists such a code
- assume a complete binary tree of depth $\ell_{\mathsf{max}} = \mathsf{max}_{\alpha \in \Sigma} \, \ell_{\alpha}$ with all free nodes
- left edges labeled 0, right edges labeled 1
- characters ordered by frequency $(\ell_1 > \ell_2 > \cdots > \ell_{\sigma})$
- assign characters the leftmost free node
- mark all nodes above and below as non-free <a>I

Shannon-Fano Coding [Fan49; Sha48]



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- a code ℓ_{α} marks $2^{\ell_{\text{max}}-\ell_{\alpha}}$ nodes
- total number of marked leafs is

$$\begin{split} \sum_{\alpha \in \Sigma} 2^{\ell_{\mathsf{max}} - \ell_{\alpha}} &= 2^{\ell_{\mathsf{max}}} \sum_{\alpha \in \Sigma} 2^{-\ell_{\alpha}} \\ &= 2^{\ell_{\mathsf{max}}} \sum_{\alpha \in \Sigma} 2^{-\lceil \lg \frac{n}{\mathsf{Hist}[\alpha]} \rceil} \\ &\leq 2^{\ell_{\mathsf{max}}} \sum_{\alpha \in \Sigma} 2^{-\lg \frac{n}{\mathsf{Hist}[\alpha]}} \\ &= 2^{\ell_{\mathsf{max}}} \sum_{\alpha \in \Sigma} \frac{\mathsf{Hist}[\alpha]}{n} \\ &= 2^{\ell_{\mathsf{max}}} \end{split}$$

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- H₀ gives average number of bits needed to encode character
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Proof (Sketch)

- let T be a text of length n over an alphabet Σ with histogram Hist
- let T_{SF} be the Shannon-Fano encoded text
- average length of encoded character is

$$(1/n)|T_{SF}| = (1/n) \sum_{\alpha \in \Sigma} Hist[\alpha] \lceil \lg \frac{n}{Hist[\alpha]} \rceil$$

$$\leq \sum_{\alpha \in \Sigma} \frac{Hist[\alpha]}{n} (\lg \frac{n}{Hist[\alpha]} + 1)$$

$$= \sum_{\alpha \in \Sigma} \frac{Hist[\alpha]}{n} \lg \frac{n}{Hist[\alpha]} + \sum_{\alpha \in \Sigma} \frac{Hist[\alpha]}{n}$$

$$= H_0(T) + 1$$



Problem with the Previous Approaches

- does not work well with repetitions
- better encode 605 × a





Definition: LZ77 Factorization

Given a text T of length n over an alphabet Σ , the **LZ77 factorization** is

- a set of z factors $f_1, f_2, \ldots, f_z \in \Sigma^+$, such that
- $T = f_1 f_2 \dots f_z$ and for all $i \in [1, z]$ f_i is
- single character not occurring in $f_1 \dots f_{i-1}$ or
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T = abababbbbabas \bullet $f_A = bbb$

- $f_2 = b$
- \bullet $f_5 = aba$

 \bullet $f_3 = abab$



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T = abababbbbaba\$

 $f_1 = a$

 \bullet $f_4 = bbb$

 $f_2 = b$

• $f_5 = aba$

$$T = \underbrace{\mathsf{aaa} \dots \mathsf{aa}}_{n-1 \text{ times}} \$$$

- $f_1 = a$
- $f_2 = \underbrace{\mathsf{aaa} \dots \mathsf{aa}}_{n-2 \text{ times}}$
- $f_3 = \$$



factors can be represented as tuple

$$(\ell_i, p_i)$$

- $\ell_i = 0$
 - factor is a single character
 - encode character in p_i
- $\ell_i > 0$
 - factor is a length- ℓ_i substring
 - $\bullet f_i = T[p_i..p_i + \ell_i)$



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- $f_1 = a = (0, a)$
- $f_2 = b = (0, b)$
- $f_3 = abab = (4, 1)$
- $f_4 = bbb = (3,6)$
- $f_5 = aba = (3,1) = (3,3)$
- $f_6 = \$ = (0,\$)$



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$$f_6 = \$ = (0,\$)$$

finding the right-most reference is hard





Definition: Previous and Next Smaller Value Arrays

Let A[1..n] be an integer array, then

- $PSV[i] = \max\{j \in [1, i) : A[j] < A[i]\}$
- $NSV[i] = min\{j \in (i, n]: A[j] < A[i]\}$

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In the Context of SA

- close to the suffix in SA
- longest possible common prefix
- before the suffix in text order

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Previous and Next Smaller Values (1/2)



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SA	13	12	1	9	6	3	11	2	10	7	4	8	5
PSV	0	0	0	3	3	3	6	3	8	8	8	11	11
NSV	2	3	∞	5	6	8	8	∞	10	11	∞	13	∞
LCP	0	0	1	2	2	5	0	2	1	1	4	0	3

PINGO how fast can we compute NSV/PSV?





- both arrays can be computed in linear time
- consider the PSV arrayNSV works analogously
- prepend $-\infty$ at index 0

```
Function ComputePSV(SA with -\infty):

1 | for i = 1, ..., n do
2 | j = i - 1
3 | while j \ge 1 and SA[i] < SA[j] do
4 | j = PSV[j]
5 | PSV[i] = j
6 | return PSV
```





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        while j \ge 1 and SA[i] < SA[j] do
         i = PSV[i]
        PSV[i] = i
     return PSV
```

- follow already computed values
- nothing in between can be PSV
- compare each element at most twice
- compute PSV and NSV in O(n) time
- example on the board <a>=

NSV, PSV, and RMQ



Recap: Range Minimum Queries

- for a range $[\ell..r]$, return position of smallest entry in an array in that range
- query time: O(1) using O(n) space
- can be used to compute the *lcp*-value of any two suffixes using the *LCP*-array
- use all arrays to find lexicographically closest suffixes
- that occur before current suffix in text order <a>=

	1	2	3	4	5	6	7	8	9	10	11	12	13
T	а	b	а	b	С	а	b	С	а	b	b	а	\$
SA	13	12	1	9	6	3	11	2	10	7	4	8	5
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LCP	0	0	1	2	2	5	0	2	1	1	4	0	3



LZ77 Factorization using SA, ISA, LCP, NSV, PSV, and RMQs

```
Function LZ77(SA, ISA, LCP, RMQ, PSV, NSV):
     pos = 1
     while pos < n do
         psv = SA[PSV[ISA[pos]]]
3
         nsv = SA[NSV[ISA[pos]]]
         if lcp(pos, psv + 1) > lcp(pos + 1, nsv) then
            \ell = lcp(pos, psv + 1) and p = psv
         else
            \ell = lcp(pos + 1, nsv) and p = nsv
         if \ell = 0 then p = pos
         new factor (\ell, T[pos])
10
         pos = pos + max{\ell, 1}
```

bring your own example





Lemma: LZ77 Running Time

The LZ77 factorization of a text of length n can be computed in O(n) time

Proof (Sketch)

- SA, LCP, PSV, NSV, RMQ_{LCP} can be computed in O(n) time
- for each text position only O(1) time





Definition: LZ78 Factorization

Given a text T of length n over an alphabet Σ , the LZ78 factorization is

- a set of z factors $f_1, f_2, \ldots, f_z \in \Sigma^+$, such that
- $T = f_1 f_2 \dots f_z, f_0 = \epsilon$ and for all $i \in [1, z]$
- if $f_1 ldots f_{i-1} = T[1..j-1]$, then f_i is the longest prefix of T[j..n], such that

$$\exists k \in [0, i), \alpha \in \Sigma \cup \{\$\} : f_k = f_i \alpha$$



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$$T = abababbbbaba$$
\$



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- $f_1 = a$



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- $f_3 = ab$
- \bullet $f_{A} = abb$

 \bullet $f_5 = bb$



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T=abababbbbaba\$\$• $f_1=a$ • $f_5=bb$ • $f_2=b$ • $f_3=ab$ • $f_4=abb$



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- $f_2 = b$
- $f_4 = abb$

- $f_5 = bb$
- \bullet $f_6 = aba$
- $f_7 = $$



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- \bullet $f_5 = bb$
- $f_3 = ab$
- \bullet $f_{A} = abb$

- \bullet $f_6 = aba$ $f_7 =$ \$
- T = abababbbbaba





- use dynamic trie to hold computed factors
- our fastest easy to use dynamic trie is?





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- using arrays of fixed size 却





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T = abababbbbaba\$	
■ <i>f</i> ₁ = a	■ <i>f</i> ₅ = bb
$f_2 = b$ $f_3 = ab$	• $f_6 = aba$
• $f_4 = abb$	■ f ₇ = \$





Lemma:

The LZ78 factorization of a text of length n can be computed in O(n) time





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The LZ78 factorization of a text of length n can be computed in O(n) time

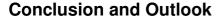
Proof (Sketch)

- search each character of the text at most once (in the trie)
- insert each character of the text at most once (in the trie)





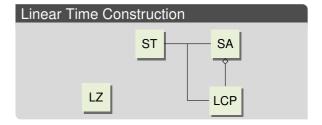
- memory usage of the LZ78 factorization very high ousing arrays of fixed size does not help
- consider only a sliding window of the text
- only factors in the window are found
- space/compression rate trade-off
- used in practice





This Lecture

- different compression methods for texts
- entropy coding
- dictionary compression

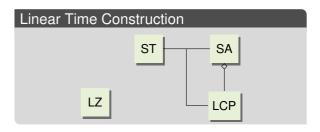


Conclusion and Outlook



This Lecture

- different compression methods for texts
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- LZ77 and LZ78 have been generalize, improved, and combined: a lot!
- LZ77
 - LZSS, LZB, LZR, LZH, . . .
- LZ78
 - LZC, LZY, LZW, LZFG, LZMW, LZJ, . . .

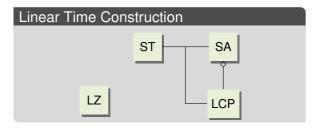


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Next Lecture

easy to compress index

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