

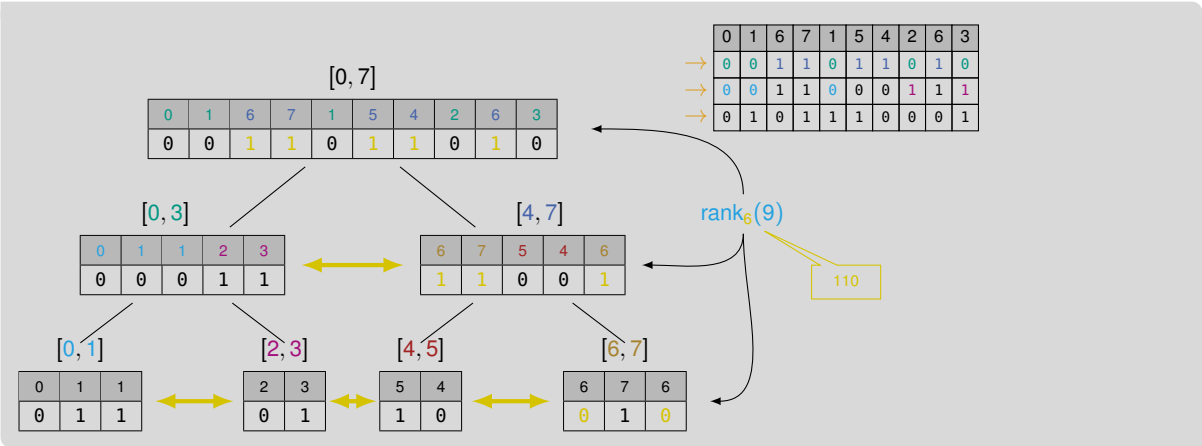
# Text Indexing

## Lecture 08: FM-Index and $r$ -Index

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# Recap: Wavelet Trees



# Recap: Compressed Wavelet Trees

0	1	3	7	1	5	4	2	6	3
0	1	1	0	1	0	0	0	0	1
0	7	5	4	2	6	1	3	1	3
0	1	0	0	0	1	1	0	1	0
0	5	4	2	7	6				
1	0	0	1	0	1				
5	4	0	2						
0	1	1	0						

- intervals are only missing to the right (white space)
- no holes allow for easy querying

- build wavelet tree for compressed text
- compress text using bit-wise negated canonical Huffman-codes
- can a wavelet tree be compressed further?

# Bit Vector Compression (1/2)

- compress (sparse) bit vectors
- bit vector contains  $k$  one bits
- use  $O(k \lg \frac{n}{k}) + o(n)$  bits
- retrieve  $\Theta(\lg n)$  bits at the same time
- similar to *rank* data structure

- split bit vector into (super-)blocks
- blocks of size  $s = \frac{\lg n}{2}$
- super-blocks of size  $s' = s^2$

## Array $C$

- number of ones in  $i$ -th block

## Lookup-Tables $L_i$

- for  $i \in [0, s]$  store lookup-table containing all bit vectors with  $i$  one bits

- use variable-length codes to identify content of block
- concatenate all codes in bit vector  $V$

## Bit Vector $V$

- let  $k_i$  be number of ones in  $i$ -th block
- use  $\lceil \lg \binom{s}{k_i} \rceil$  bits to encode block  $i$  position in lookup-table
- concatenate all codes

# Bit Vector Compression (2/2)

## Array *SBlock*

- for every super-block  $i$ ,  $SBlock[i]$  contains position of encoding of first block in  $i$ -th super-block in  $V$
- $\lceil \lg n \rceil$  bits per entry

## Array *Block*

- for every block  $i$ ,  $Block[i]$  contains position of encoding of  $i$ -th block in  $V$  relative to its super-block
- $O(\lg \lg n)$  bits per entry

## Lemma: Compressed Bit Vectors

A bit vector of size  $n$  containing  $k$  ones can be represented using  $O(k \lg \frac{n}{k}) + o(n)$  bits allowing  $O(1)$  time access to individual bits

## Proof (Sketch space requirements)


- $|C| = O(\frac{n}{s} \lg s) = o(n)$  bits
- $|SBlock| = O(\frac{n}{s'} \lg n) = o(n)$  bits
- $|Block| = O(\frac{n}{s} \lg s) = o(n)$  bits
- $\sum_{k=0}^s |L_k| \leq (s+1)2^s s = o(n)$  bits
- $|V| = \sum_{i=1}^{\lceil n/s \rceil} \lceil \lg \binom{s}{k_i} \rceil \leq \lg \binom{n}{k} + n/s \leq \lg((n/k)^k) + n/s = k \lg \frac{n}{k} + O(\frac{n}{\lg n})$  bits

# Recap: Backwards Search in the BWT

**Function** *BackwardsSearch*( $P[1..n]$ ,  $C$ ,  $rank$ ):

```

1  |  $s = 1, e = n$ 
2  | for  $i = m, \dots, 1$  do
3  |   |  $s = C[P[i]] + rank_{P[i]}(s - 1) + 1$ 
4  |   |  $e = C[P[i]] + rank_{P[i]}(e)$ 
5  |   | if  $s > e$  then
6  |   |   | return  $\emptyset$ 
7  | return  $[s, e]$ 
  
```

- no access to text or SA required
- no binary search
- existential queries are easy
- counting queries are easy
- reporting queries require additional information
- example on the board 

# The FM-Index [FM00]

## Building Blocks of FM-Index

- wavelet tree on BWT providing *rank*-function
- *C*-array
- sampled suffix array with sample rate  $s$
- bit vector marking sampled suffix array positions

## Lemma: FM-Index

Given a text  $T$  of length  $n$  over an alphabet of size  $\sigma$ , the FM-index requires  $O(n \lg \sigma)$  bits of space and can answer counting queries in  $O(m \lg \sigma)$  time and reporting queries in  $O(\text{occ} + m \lg \sigma)$  time

## Space Requirements

- wavelet tree:  $n \lceil \lg \sigma \rceil (1 + o(1))$  bits
  - *C*-array:  $\sigma \lceil \lg n \rceil$  bits  $\ominus n(1 + o(1))$  bits if  $\sigma \geq \frac{n}{\lg n}$
  - sampled suffix array:  $\frac{n}{s} \lceil \lg n \rceil$  bits
  - bit vector:  $n(1 + o(1))$  bits
- space and time bounds can be achieved with  $s = \lg_{\sigma} n$


# Conclusion FM-Index

- FM-index is easy to compress
  - wavelet tree on *BWT* can be compressed
  - bit vector can be compressed
- 
- very small in comparison with suffix tree or suffix array
  - compression does not make use of structure of *BWT* ⓘ wavelet trees are compressed using Huffman-codes

## Definition: Run (simplified, recap)

Given a text  $T$  of length  $n$ , we call its substring  $T[i..j]$  a **run**, if

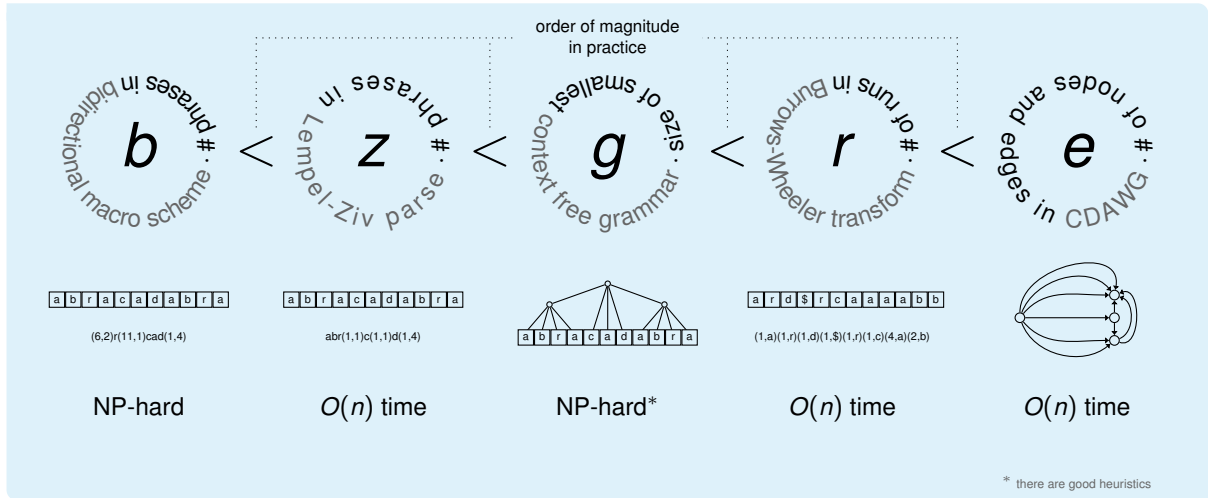
- $T[k] = T[l]$  for all  $k, l \in [i, j]$  and
- $T[i - 1] \neq T[i]$  and  $T[j + 1] \neq T[j]$

ⓘ (To be more precise, this is a definition for a run of a periodic substring with smallest period 1, but this is not important for this lecture )

	1	2	3	4	5	6	7	8	9	0	11	12	13
L	a	b	\$	c	c	b	b	a	a	a	a	b	b



# Measures of Repetitiveness (Excerpt)



## Motivation: $r$ -Index

### Measure for Compressibility

- $k$ -th order empirical entropy  $H_k$
- number of LZ factors  $z$
- number of *BWT* runs  $r$

- $z$  and  $r$  not blind to repetitions
- how do they relate?

### Lemma: *BWT* runs and LZ factors [KK20]

Given a text  $T$  of length  $n$ . Let  $z$  be the number of LZ77 factors and  $r$  the number of runs in  $T$ 's *BWT*, then

$$r \in O(z \lg^2 n)$$

- more details in next lecture

# Main Part of Backwards-Search


**Function** *BackwardsSearch*( $P[1..n]$ ,  $C$ ,  $rank$ ):

```
1 |  $s = 1, e = n$ 
2 | for  $i = m, \dots, 1$  do
3 | |  $s = C[P[i]] + rank_{P[i]}(s - 1) + 1$ 
4 | |  $e = C[P[i]] + rank_{P[i]}(e)$ 
5 | | if  $s > e$  then
6 | | | return  $\emptyset$ 
7 | return  $[s, e]$ 
```

## Goals

- simulate *BWT* and *rank* on *BWT* in
- $O(r \lg n)$  bits of space

# The $r$ -Index [GNP20] (1/3)

Given a text  $T$  of length  $n$  over an alphabet  $\Sigma$  and its  $BWT$ , the  $r$ -index of this text consists of the following data structures 

## Array $I$

- $I[i]$  stores position of  $i$ -th run in  $BWT$

## Array $L'$

- $L'[i]$  stores character of  $i$ -th run in  $BWT$
- build wavelet tree for  $L'$

## Array $R$

- lengths of  $BWT$  runs stably sorted by runs' characters
- accumulate for each character by performing exclusive prefix sum over run lengths'

## Array $C'$

- $C'[\alpha]$  stores the start of the run lengths in  $R$  for each character  $\alpha \in \Sigma$  starting at 0

## Bit Vector $B$

- compressed bit vector of length  $n$  containing ones at positions where  $BWT$  runs start and rank-support

## The $r$ -Index (2/3)

### $rank_{\alpha}(BWT, i)$ with $r$ -Index

- compute number  $j$  of run ( $j = rank_1(B, i)$ )
- compute position  $k$  in  $R$  ( $k = C'[\alpha]$ )
- compute number  $\ell$  of  $\alpha$  runs before the  $j$ -th run ( $\ell = rank_{\alpha}(L', j - 1)$ )
- compute number  $k$  of  $\alpha$ s before the  $j$ -th run ( $k = R[k + \ell]$ )
- compute character  $\beta$  of run ( $\beta = L'[j]$ )
- if  $\alpha \neq \beta$  return  $k$  ⓘ  $i$  is not in the run
- else return  $k + i - l[j] + 1$  ⓘ  $i$  is in the run

## The $r$ -Index (3/3)

### Lemma: Space Requirements $r$ -Index

Given a text  $T$  of length  $n$  over an alphabet of size  $\sigma$  that has  $r$   $BWT$  runs, then its  $r$ -index requires

$$O(r \lg n) \text{ bits}$$

and can answer *rank*-queries on the  $BWT$  in  $O(\lg \sigma)$ .  
Given a pattern of length  $m$ , the  $r$ -index can answer pattern matching queries in time


$$O(m \lg \sigma)$$

- what about reporting queries?


# Locating Occurrences (Sketch)

- modify backwards-search that it maintains  $SA[e]$
- after backwards-search output  $SA[e], SA[e - 1], \dots, SA[s]$
- in  $O(r \lg n)$  bits and  $O(occ \cdot \lg \lg r)$  time

## Maintaining $SA[e]$

- sample  $SA$  positions at ends of runs
- if next character is  $BWT[e]$ , then next  $SA[e']$  is  $SA[e] - 1$
- otherwise locate end of run and extract sample 

## Output Result

- following  $LF$  not possible  unbounded
- deduce  $SA[i - 1]$  from  $SA[i]$
- character in  $L$  and  $F$  in same order
- only beginning of runs complicated
- for every character build predecessor data structure over sampled  $SA$ -values at end of runs
- associate with  $\langle i, SA[i] \rangle$

# The Move Data Structure [NT21]

## Definition: Disjoint Interval Sequence

Let  $I = (p_1, q_1), (p_2, q_2), \dots, (p_k, q_k)$  be a sequence of  $k$  pairs of integers. We introduce a permutation  $\pi$  of  $[1, k]$  and sequence  $d_1, d_2, \dots, d_k$  for  $I$ .  $\pi$  satisfies  $q_{\pi[1]} \leq q_{\pi[2]} \leq \dots \leq q_{\pi[k]}$ , and  $d_i = p_{i+1} - p_i$  for  $i \in [1, k]$ , where  $p_{k+1} = n + 1$ . We call the sequence  $I$  a disjoint interval sequence if it satisfies the following three conditions:

- $p_1 = 1 < p_2 < \dots < p_k \leq n$
- $q_{\pi[1]} = 1$ ,
- $q_{\pi[i]} = q_{\pi[i-1]} + d_{\pi[i-1]}$  for each  $i \in [2, k]$ .

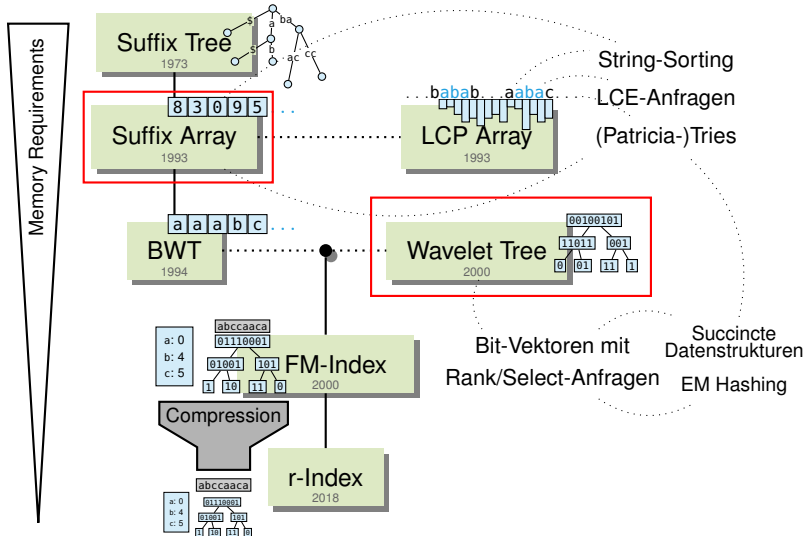
## Move Query

$$\text{move}(i, x) = (i', x')$$

- $i$  position in input interval
- $x$  input interval
- $i'$  position in output interval
- $x'$  input interval covering  $i'$



# From the Suffix Tree to the *r*-Index—Questions?



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