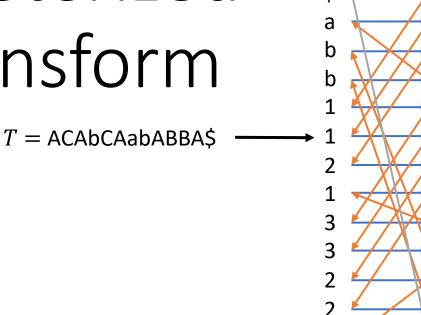
Extending the Parameterized Burrows-Wheeler Transform

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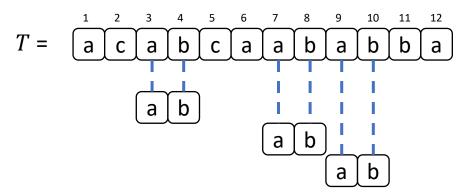
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Pattern Matching

- alphabet Σ
- text $T \in \Sigma^*$, pattern $P \in \Sigma^*$
- occurrence of P in T:
 substring of T that equals P
- PM: count all occurrences of P in T write as T. count(P)
- goal: index T for efficient PM

- $\Sigma = \{a,b,c\}$
- T = acabcaababba
- P = ab
- occurrences of P in T
 at positions 3, 7 and 9
- T.count(P) = 3



Parameterized Strings

- alphabet Σ_s of static symbols (s-symbols)
- alphabet Σ_p of parameterized symbols (p-symbols)
- $\Sigma_s \cap \Sigma_p = \emptyset$
- string over $\Sigma := \Sigma_s \cup \Sigma_p$ is a parameterized string (p-string)
- character in Σ called symbol, $\sigma:=|\Sigma|$ size of alphabet example
- $\Sigma_S = \{a, b\}, \Sigma_p = \{A, B, C\}$
- T = ACAbCAabABBA

Parameterized Matching (p-Matching)

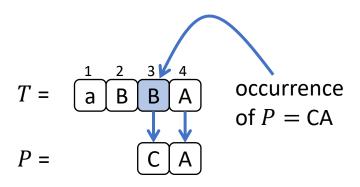
- *U*, *V* p-strings
- U p-matches $V:\Leftrightarrow$ if |U|=|V| and \exists a bijection $\psi:\Sigma_p\to\Sigma_p$ with
 - U[i] = V[i] if $V[i] \in \Sigma_S$
 - $U[i] = \psi(V[i])$ otherwise
- write $U =_p V$ iff U and V p-match example
- U = bBCAaCB
- V = bCABaAC
- $\psi(A) = C$, $\psi(B) = A$, $\psi(C) = B$

 $\Rightarrow U =_{p} V$

Parameterized Pattern Matching (PPM)

[Baker '93]

- *T* : text p-string
- P : pattern p-string
- occurrence of P in T : substring of T that p-matches P
- PPM: count all occurrences of P in T, written as T.count(P)
- goal: index text T for efficient PPM



Indexes for PPM

data structure	time for PPM	reference
suffix tree	$O(m \log \sigma)$	[Baker '93]
suffix array	$O(m + \log n)$	[Deguchi + '08]
position heap	$O(m \log \sigma + m \sigma_p)$	[Diptarama+ '17]
suffix tray	$O(m + \log \sigma)$	[Fujisato+ '21]
DAWG	$O(m \log \sigma)$	[Nakashima+ '22]

- $\sigma := |\Sigma|$ alphabet size
- $\sigma_p := |\Sigma_p|$
- n := |T|, text size
- m: |P|, pattern length

All data structures need $O(n \log n)$ bits

PPM in small memory

- parameterized Burrows-Wheeler transform (pBWT) [Ganguly+'17]
 - $n \lg \sigma + O(n)$ bits
 - computes T.count(P) in $O(m \log \sigma)$ time
- simplified pBWT [Kim, Cho '21]
 - $2n \lg \sigma + O(n)$ bits
 - same time complexities
- both approaches use space linear in the number of bits of the input!

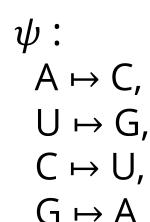
Applications for PPM

many use cases

- software maintenance [Baker '97]
- plagiarism detection
- analyzing genetic data [Shibuya '04]

RNA matching

- matching RNA base pair
- *X* = AUGCAUC
- Y = CGAUCGU
- $\psi(X) = Y$



[Shibuya' 04]

 but some RNA structures are cyclic, so there is a need for cyclic pattern matching ⇒ circular parameterized pattern matching (CPPM)

$$X = \bigcup_{\mathcal{D}} \bigcap_{\mathbf{A}} =_{\mathbf{P}} \supset_{\mathbf{P}} \bigcap_{\mathbf{D}} = Y$$

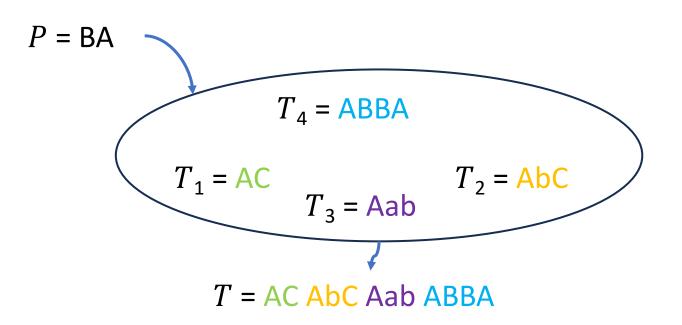
Circular PPM (CPPM)

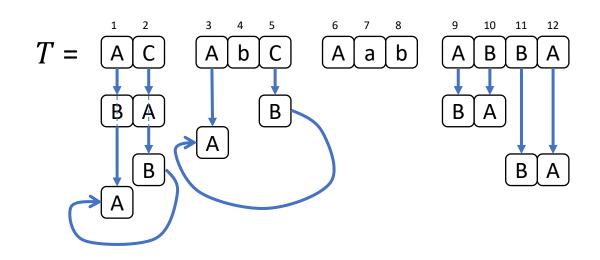
 $T = \begin{bmatrix} 1 & 2 & 3 & 4 \\ A & C & A & B & A \end{bmatrix}$ occurrence of P = CC

- text p-strings $T = \{T_1, ..., T_d\}$
- pattern p-string P
- occurrence of P in T refers to the starting position of a substring of T_1, \ldots, T_d that p-matches P
- all text p-strings are viewed circularly
- CPPM: count all occurrences of P in T
- goal: index texts T_1 , ..., T_d for efficient CCPM

Example: CPPM

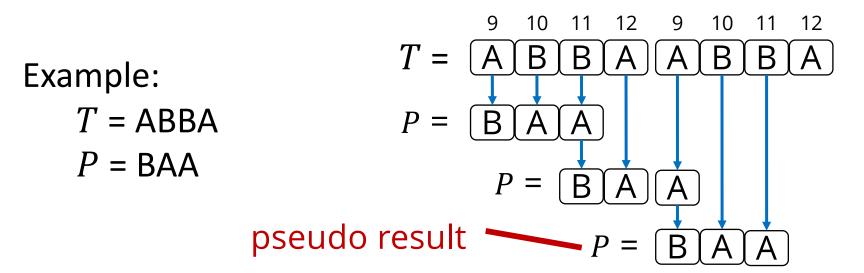
- $T = \{AC, AbC, Aab, ABBA\}$
- P = BA
- occurrences of P in T at positions 1, 2, 4, 9 and 11
- T.count(P) = 5





Simple idea for CPPM

- general naive approach for matching *P* in *T* circularly:
- perform classic matching of P in $T \cdot T$
- may generate pseudo results in the second part
- discard pseudo results in postprocessing



From pBWT to epBWT

define epBWT based on two encodings

- prev-encoding $\langle V \rangle$ [Baker '93]
- Hashimoto-encoding «V», [Hashimoto+ '22] motivation is explained with a review of pBWT

BWT (Burrows-Wheeler transform):

- last character of all cyclic rotations sorted in lexicographic order
- T.count(P) via length of reported range of backward search
- how to use that technique with p-matching?

pBWT

review of the simplified pBWT [Kim, Cho '21] for PPM

Comparing p-Strings

• consider conjugates of V = AABB

- AABB $=_p$ BBAA \neq_p ABBA $=_p$ BAAB, but AABB < ABBA < BAAB < BBAA
- cannot use original p-string to sort or p-match!

prev-Encoding

given p-string V, compute prev-encoding $\langle V \rangle$ of V as follows:

- replace leftmost occurrence of any p-symbol in V by ∞
- replace each other by distance to its previous occurrence

for every p-string
$$U: \langle V \rangle = \langle U \rangle \Leftrightarrow V =_p U$$
 [Baker '93]

$$T$$
 = A C A b C A a b A B B A $\langle T \rangle$ = ∞ ∞ 2 b 3 3 a b 3 ∞ 1 3

but unstable under rotation

$$\langle T \rangle [2..] \langle T \rangle [1] = \infty$$
 2 b 3 3 a b 3 ∞ 1 3 ∞ $T[2..] T[1] = C$ A b C A a b A B B A A $\langle T[2..] T[1] \rangle = \infty$ ∞ b 3 3 a b 3 ∞ 1 3 1

Hashimoto-Encoding [Hashimoto+ '22]

- view p-string V circularly and replace each occurrence of a p-symbol in V by the number of distinct p-symbols until its next occurrence
- write «V» for the Hashimoto-encoding of V

$$T=$$
 A C A b C A a b A B B A $\ll T \gg =$ 2 2 2 b 3 1 a b 2 1 3 1

- for every p-string $U: \ll V \gg = \ll U \gg \iff V =_p U$ [Hashimoto+ '22]
- encoding is commutative with rotation!

```
(T)[2..](T)[1] = 2 2 b 3 1 a b 2 1 3 1 2 T[2..]T[1] = C A b C A a b A B B A A T[2..]T[1] = 2 2 b 3 1 a b 2 1 3 1 2
```

Parameterized BWT (pBWT)

- text T = ACAbCAabABBA\$
- $<\!\!< T >\!\!> = 222b31ab2131 <\!\!>$
- pBWT(T) = (F_T , L_T)
- first and last symbols of Hashimotoencoded conjugates sorted by their prev-encodings
- similar entries of both strings are sorted by succeeding context! [Iseri+ '23]

```
L_T
$ 2 2 2 b 3 1 a b 2 1 3 1
a b 2 1 3 1 $ 2 2 2 b 3 1
b 3 1 a b 2 1 3 1 $ 2 2 2
b 2 1 3 1 $ 2 2 2 b 3 1 a
1 $ 2 2 2 b 3 1 a b 2 1 3
1 a b 2 1 3 1 $ 2 2 2 b 3
2 b 3 1 a b 2 1 3 1 $ 2 2
131$222b31ab2
31$222b31ab21
31ab2131$222b
22b31ab2131$2
2131$222b31ab
2 2 2 b 3 1 a b 2 1 3 1 $
```

LF (Mapping) Property

- text T = ACAbCAabABBA\$
- $\langle T \rangle = 2_1 2_2 2_3 b_1 3_1 1_1 a_1 b_2 2_4 1_2 3_2 1_3 \$_1$
- first column $F_T = \$_1 a_1 b_1 b_2 1_3 1_1 2_3 1_2 3_2 3_1 2_2 2_4 2_1$
- last column $L_T = 1_3 1_1 2_3 a_1 3_2 3_1 2_2 2_4 1_2 b_1 2_1 b_2 \$_1$
- define permutation LF_T by mapping from ith occurrence of a symbol $x \in \Sigma_s \cup [1..|\Sigma_p|]$ in L_T to ith occurrence of x in F_T
- LF property: maps x_k of L_T to x_k of F_T !

```
1 $<sub>1</sub> 2 2 2 b 3 1 a b 2 1 3 1<sub>3</sub>
  a_1 b 2 1 3 1 $ 2 2 2 b 3 1_1
     b_1 3 1 a b 2 1 3 1 $ 2 \nearrow 23
  5 1<sub>2</sub> $ 2 2 2 b 3 1 a b
 8 1<sub>2</sub> 3 1 5 2 2 2 b 3 1 a b 2<sub>4</sub>
 9 3_2 1 $ 2 2 2 p_T^3 1 a b 2 1_2
10 3<sub>1</sub> 1 a b 2 1 3 1 $ 2 2 2 b<sub>1</sub>
11 2<sub>2</sub> 2 b 3 1 a b 2 1 3 1 $ 2<sub>1</sub>
12 2<sub>4</sub> 131$222b31ab<sub>2</sub>
13 2<sub>1</sub> 2 2 b 3 1 a b 2 1 3 1 $<sub>1</sub>
```

epBWT

from pBWT to epBWT

ω -Order

idea: use the infinite iteration of a conjugate as key for sorting

- V^{\omega}: infinite iteration of V
- $root(V) := primitive root of V \quad (V = ababab \Rightarrow root(V) = ab)$

- *V* , *U* : finite strings
- $V =_{\omega} U :\Leftrightarrow \text{root}(V) = \text{root}(U)$
- $V \prec_{\omega} U :\Leftrightarrow \exists i \colon V^{\omega}[..i] = U^{\omega}[..i] \land V^{\omega}[i+1] < U^{\omega}[i+1]$

Extending the ω -Order to p-Strings

- *V* , *U* : finite p-strings
- $V =_{\omega} U :\Leftrightarrow root(\langle V \rangle) = root(\langle U \rangle)$
- $V \prec_{\omega} U :\Leftrightarrow \exists i : \langle V^{\omega} \rangle [..i] = \langle U^{\omega} \rangle [..i] \wedge \langle V^{\omega} \rangle [i+1] < \langle U^{\omega} \rangle [i+1]$

- extended ω -order already used when defining the pBWT!
- coincides with prev-order for p-strings of the same length

Example: ω -Order on p-Strings

- $T_1 = AB$, $T_2 = ABA$, $T_3 = ABAB$
- $\langle T_1 \rangle < \langle T_2 \rangle < \langle T_3 \rangle$
- $T_1^{\omega}[...8] = ABABABABAB$
- $T_2^{\omega}[...8] = ABAABAAB$
- $T_3^{\omega}[...8] = A B A B A B A B$
- $<\!\!< T_1 >\!\!> = 22$
- $\langle T_3 \rangle = 2 \ 2 \ 2 \ 2$

•
$$\langle T_1 \rangle = \infty \infty$$

•
$$\langle T_2 \rangle = \infty \infty 2$$

•
$$\langle T_3 \rangle = \infty \infty 2 2$$

•
$$\langle T_1^{\omega} \rangle [...8] = \infty \infty 2 2 2 2 2 2$$

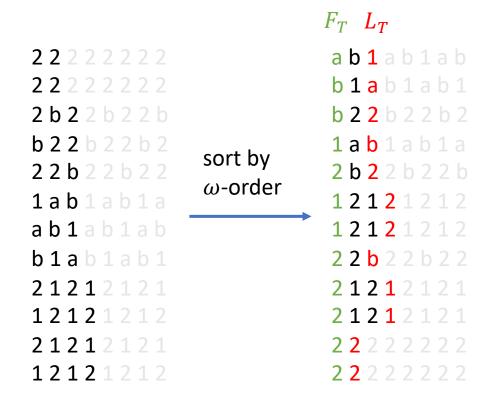
•
$$\langle T_2^{\omega} \rangle [...8] = \infty \infty 2 1 3 2 1 3$$

•
$$\langle T_3^{\omega} \rangle [...8] = \infty \infty 2 2 2 2 2 2$$

•
$$T_2 \prec_{\omega} T_1 =_{\omega} T_3$$

Extended pBWT (epBWT)

- sort conjugates by ω -order tie-break:
 - first by index of text string,
 - second by text position
- $T = \{AC, AbC, Aab, ABBA\}$
- epBWT(T) = (F_T , L_T)
- first and last symbols of Hashimoto-encoded conjugates sorted by their prev-encodings in ω -order



LF (Mapping) Property

- $T = \{AC, AbC, Aab, ABBA\}$
- $\{ \langle T_1 \rangle, \langle T_2 \rangle, \langle T_3 \rangle, \langle T_4 \rangle \} = \{ 2_1 2_2, 2_3 b_1 2_4, 1_1 a_1 b_2, 2_5 1_2 2_6 1_3 \}$
- first column $F_T = a_1b_2b_11_12_31_21_32_42_52_62_12_2$
- last column $L_T = 1_1 a_1 2_3 b_2 2_4 2_5 2_6 b_1 1_3 1_2 2_2 2_1$
- define permutation LF_T by mapping from ith occurrence of a symbol $x \in \Sigma_s \cup [1..|\Sigma_p|]$ in L_T to ith occurrence of x in F_T
- only maps x_k of L_T to x_k of F_T if Hashimoto-encoded texts are primitive! («AC» and «ABBA» are not primitive!)
- remedy: build epBWT on the Hashimoto-encoded roots!

```
1 a_1 b_2 1_1
  a_{1} b_{2} 1_{1} a_{1}
        b_1 2_4 2_3
       1_1 a_1 b_2
  5 \quad 2_3 \quad b_1 \quad 2_4
 9 2_5 \perp_2 2_6 1_3
10 2<sub>6</sub> 1<sub>3</sub> 2<sub>5</sub> 1<sub>2</sub>
11 \quad 2_1 \quad 2_2
12 2<sub>2</sub> 2<sub>1</sub>
```

Summary

epBWT is a CPPM index for a set of p-strings

- builds upon pBWT of [Kim, Cho '21] and eBWT of [Mantaci+ '07]
- uses $2n \lg \sigma + O(n)$ bits of space partially in the paper (full version will follow):
- T.count(P) in $O(m \lg \sigma)$ time for CPPM
- reconstruction of input up to p-matching equivalence
- construction of index in $O\left(n\frac{\lg^2 n}{\lg\lg n}\right)$ time with $O(n\lg n)$ bits of space
- applications to other matchings such as Cartesian-Tree matching