XBWTing Readsets

Taking advantage of read alignment to obtain better compression

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The FM-Index

On Strings

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TTAGATACAT\$GA

Compresses and indexes the input.		F L
compresses and mackes the input.	0	\$GATTAGATACAT
Applied in short read aligners such as Bowtie and BWA .	1	ACAT\$GATTAGAT
∇	2	AGATACAT\$GATT
Recently improved to support in O(r) space, with r the	3	AT\$GATTAGATAC
number of runs in the BWT :	4	ATACAT\$GATTAG
 count(P): counting the occ. of a pattern P in O(P) time. 	5	ATTAGATACAT\$G
	6	CAT\$GATTAGATA
 - locate(P): locate the occ. Of P in O(P +occ) time. 	7	GATACAT\$GATTA
[Gagie, Navarro, and Prezza., Fully functional suffix trees and optimal text searching	8	GATTAGATACAT\$
in bwt-runs bounded space]	9	T\$GATTAGATACA
Example:	10	TACAT\$GATTAGA
S = GATTAGATACAT	11	TAGATACAT\$GAT

BWT(S) = TTTCGGAA $\Rightarrow 8 runs.$

Can we have the **same number of runs** for reads extracted from S?

Concatenate the reads with a separator "\$" and build the FM-index of the entire string.

Example:

S = GATTA\$TTAGA\$TAGATA\$GATAC\$ATACAT\$

BWT(S) = CATAATGTTTTTCGG\$GAAAA\$\$AATAAT\$A\$ => 20 runs.

Issues:

- Computing the BWT for such a long string is challenging (and the context before the dollars is not relevant).
- The \$ break some runs as in CGG\$G in the example.



Extended BWT: Permute the characters in the strings into the lexicographic order of the suffixes that immediately follow them, considering each string to be cyclic.

Example:

S = {GATTA\$, TTAGA\$, TAGATA\$, GATAC\$, ATACAT\$}
EBWT(S) = TCAAATTGTTTTCGG\$GAAAA\$\$ATAAAT\$A\$ => 19 runs.

Easier to build and update than the naive approach.

Issues:

• Still has more than double the number of run than BWT (GATTAGATACAT).

The FM-Index for Readsets

Heuristics for read ordering for the EBWT

RLO: Reorganizing the reads in reversed lexicographic order (co-lexicographic order)

Example:

RLO(S) = {GATTA\$, TTAGA\$, TAGATA\$, GATAC\$, ATACAT\$}
EBWT(RLO(S)) = AAACTGTTTTTTCGG\$GAAAA\$\$AATAAT\$A\$ => 19 runs

SPRING: attempts to reorder reads according to their position in the genome. EBWT(SPRING(S)) = ACATATTGTTTTCGG\$GAAAA\$\$ATAAAT\$A\$ => 22 runs

The FM-Index for Readsets



eXtended BWT: Generalization of the BWT for labeled trees where the characters get sorted according to the label of there outgoing node.

Can also be seen as a sub-case of a Wheeler graph.

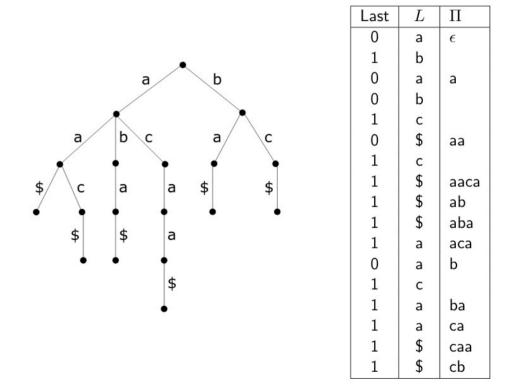


Figure: XBWT Tricks, Giovanni Manzini

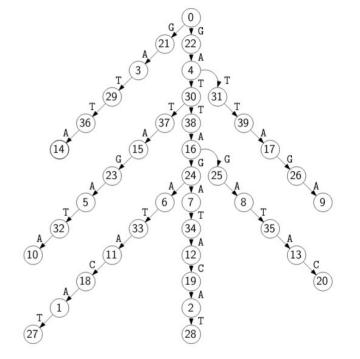


If the reads correspond to a known assembled genome, use the genome as additional context for better compression.

Example:

TTAGA TAGATA GATTAGATACAT GATTA ATACAT GATAC







Theorem 1

If we create such a tree, where the **reads are errorlessly sampled and aligned to the reference**, then the XBWT of the tree has the **same number of runs** as the BWT of the reverse of the reference.

Theorem 2

If we create a labeled tree T as explained, let: r be the number of runs in the XBWT, t be the number of reads, Then in O(r+t) words of space, We can compute Locate(P) in O((|P|+occ)log logn)

In reality the reads we have to index are not perfectly matching the assembled genome. How does this approach to compression work in practice ?

PFP construction of the XBWT

Scan

- Scan: Slide a window, compute the KR-fingerprint of the window, cutting the text into prefix free phrases
- Scan the reference creating a parse and a dictionnary
- For each read, extend the read, parse the extended read

XBWT of the parse

• Doubling algorithm to determine the XBWT of the parse

Final XBWT

- SA and LCP of all suffixes of the words in the dictionary
- For each suffix, add the correct character depending on the cases



Experimental results

Only on the number of runs for now.

We compared to:

- **EBWT** (using the ropebwt2 implementation), with and without \$.
- **SPRING + EBWT,** with and without \$.
- **RLO + EBWT,** with and without \$.

EBWT > SPRING + EBWT > RLO + EBWT

So in the graphs we focus on RLO+EBWT.

Experimental results

On the following datasets:

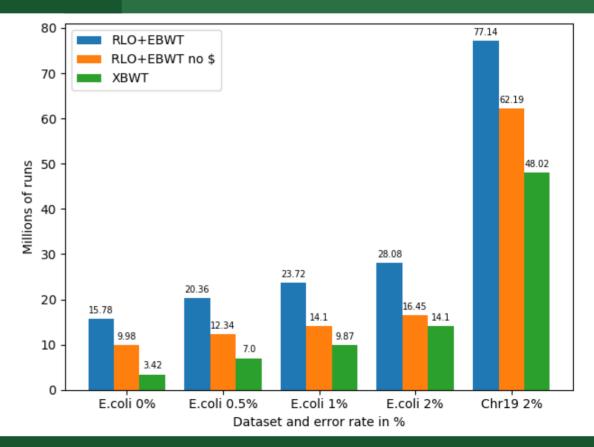
- synthetic: generated reads with controllable error rates
- Real world: only reads matched to the genome, not the reverse complement
 - E.coli: from the single cell dataset
 - Stapphylococcus aureus: from the single cell dataset
 - Hydrophilia: from the Gage-b dataset

Experimental results

Synthetic data

Generated by wgsim:

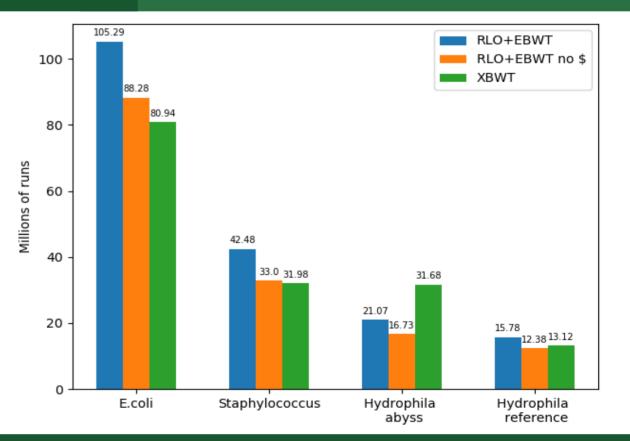
- Error rate from 0 to 2%
- 20 Millions reads of length 100 bp
- 15% of polymorphisms are INDELs with their lengths drawn from a geometric distribution
- The XBWT is always more compact, even for 2% >> 0.1%.



Experimental results

Real data

- Good or similar for E.coli and Staphylococcus.
- Surprising result for hydrophila?
- For E. coli readset, RLO+EBWT without \$ has 16% less runs than with \$ and the XBWT 8.3% less run than RLO+EBWT no \$.



Future work

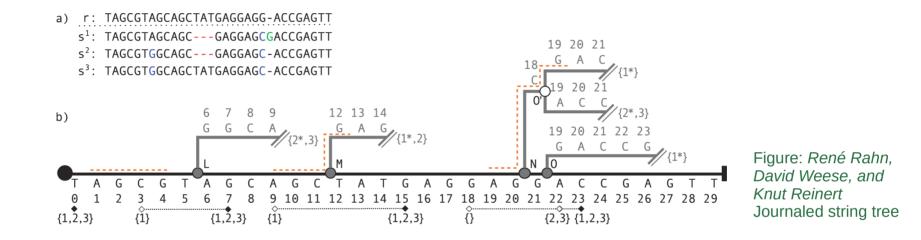
Experiments

To do:

- Evaluation on the human Genome
- FM-index, implementation and time analysis
- Time comparison of PFP construction of the XBWT compared to other construction [BWT-tunneling by Uwe Baier, Wheeler sort by Jarno Alanko]

Future work

Hybrid index improvement?



Storing the XBWT of a JST could be more efficient than storing the BWT of a string kernel.

Thank you for your attention !

Take away message:

- For compression removing the dollars is important.
- Taking advantage of the genome for additional context brings a better compression!

Questions?

Pre-publication: https://arxiv.org/pdf/1809.07320.pdf

