# Conway-Bromage-Lyndon (CBL): 

## AN EXACT, DYNAMIC REPRESENTATION OF K-MER SETS

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## MOTIVATION OF THIS WORK

Plenty of compact data structures for storing $k$-mers ...but most of them are static


Query time and memory usage of some efficient data structures, taken from [Alanko et al. 22]

## OUR FOCUS FOR THIS TALK

Goal: designing a dynamic index of $k$-mers with fast queries and relatively good compression

- membership
- enumeration
- insertion
- deletion
- set operations $(\cup, \cap, \backslash)$

CTGAAATG...


## Starting from a simple idea: $K$-Mers as a sparse set of integers

## Limitations:

- we can see $k$-mers as integers in $\llbracket 4^{k} \rrbracket$ $\mathrm{A} \rightarrow 00 \quad \mathrm{C} \rightarrow 01 \quad \mathrm{G} \rightarrow 10 \quad \mathrm{~T} \rightarrow 11$
- since they're usually very sparse, we can store them in a sparse bitvector (as in [Conway \& Bromage 11])
- difficult to compress (especially if it's dynamic)
- not cache-efficient
$i d(A T G G C A) \ll i d(T G G C A T)$ (average distance of $4^{k} / 3$ )


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What if we changed our representation of $k$-mers?


## THE NECKLACE TRANSFORMATION

## NECKLACE TRANSFORMATION OF K-MERS



The necklace of $x$ is its smallest cyclic rotation $\langle x\rangle=\min _{0 \leqslant i<k} x^{(i)}$
To make this transformation reversible, keep track of the rotation index

$$
x \longmapsto(\langle x\rangle, \text { rotation index })
$$

## NeCKLACES OF CONSECUTIVE K-MERS SHARE LONG PREFIXES

$k-m e r$ view
GTCGTTCTTCCTAACGTCATCTCTCATTCTG
TCGTTCTTCCTAACGTCATCTCTCATTCTGT
CGTTCTTCCTAACGTCATCTCTCATTCTGTG
GTTCTTCCTAACGTCATCTCTCATTCTGTGA
TTCTTCCTAACGTCATCTCTCATTCTGTGAC
TCTTCCTAACGTCATCTCTCATTCTGTGACA
CTTCCTAACGTCATCTCTCATTCTGTGACAC
TTCCTAACGTCATCTCTCATTCTGTGACACG
TCCTAACGTCATCTCTCATTCTGTGACACGC
CCTAACGTCATCTCTCATTCTGTGACACGCA
CTAACGTCATCTCTCATTCTGTGACACGCAG
TAACGTCATCTCTCATTCTGTGACACGCAGG
AACGTCATCTCTCATTCTGTGACACGCAGGG
ACGTCATCTCTCATTCTGTGACACGCAGGGT

## NeCKLACES OF CONSECUTIVE K-MERS SHARE LONG PREFIXES

> necklace view AACGTCATCTCTCATTCTG GTCGTTCTTCCT AACGTCATCTCTCATTCTGT TCGTTCTTCT AACGTCATCTCTCATTCTGTG CGTTCTTCCT AACGTCATCTCTCATTCTGTGA GTTCTTCCT AACGTCATCTCTCATTCTGTGAC TTCTTCT AACGTCATCTCTCATTCTGTGACA TCTTCCT AACGTCATCTCTCATTCTGTGACAC CTTCCT AACGTCATCTCTCATTCTGTGACACG TTCT AACGTCATCTCTCATTCTGTGACACGC TCCT AACGTCATCTCTCATTCTGTGACACGCA CCT AACGTCATCTCTCATTCTGTGACACGCAG CT AACGTCATCTCTCATTCTGTGACACGCAGG T AACGTCATCTCTCATTCTGTGACACGCAGGG ACACGCAGGGT ACGTCATCTCTCATTCTGTG
k-mer view
GTCGTTCTTCCTAACGTCATCTCTCATTCTG TCGTTCTTCCTAACGTCATCTCTCATTCTGT CGTTCTTCCTAACGTCATCTCTCATTCTGTG GTTCTTCCTAACGTCATCTCTCATTCTGTGA TTCTTCCTAACGTCATCTCTCATTCTGTGAC TCTTCCTAACGTCATCTCTCATTCTGTGACA CTTCCTAACGTCATCTCTCATTCTGTGACAC tTCCTAACGTCATCTCTCATTCTGTGACACG tCCTAACGTCATCTCTCATTCTGTGACACGC CCTAACGTCATCTCTCATTCTGTGACACGCA CTAACGTCATCTCTCATTCTGTGACACGCAG TAACGTCATCTCTCATTCTGTGACACGCAGG AACGTCATCTCTCATTCTGTGACACGCAGGG acGTCATCTCTCATTCTGTGACACGCAGGGT

## NeCKLACES OF CONSECUTIVE K-MERS SHARE LONG PREFIXES

## necklace view

AACGTCATCTCTCATTCTG GTCGTTCTTCCT AACGTCATCTCTCATTCTGT TCGTTCTTCCT AACGTCATCTCTCATTCTGTG CGTTCTTCCT AACGTCATCTCTCATTCTGTGA GTTCTTCCT AACGTCATCTCTCATTCTGTGAC TTCTTCCT AACGTCATCTCTCATTCTGTGACA TCTTCCT AACGTCATCTCTCATTCTGTGACAC CTTCCT AACGTCATCTCTCATTCTGTGACACG TTCCT AACGTCATCTCTCATTCTGTGACACGC TCCT AACGTCATCTCTCATTCTGTGACACGCA CCT
 AACGTCATCTCTCATTCTGTGACACGCAG CT AACGTCATCTCTCATTCTGTGACACGCAGG T AACGTCATCTCTCATTCTGTGACACGCAGGG

## QUICKLY COMPUTING STREAMS OF NECKLACES

Basic approach: compute every cyclic rotation and select the smallest in $\mathcal{O}(k)$.
$\rightarrow \mathcal{O}(n k)$ for $n$ necklaces

Better: amortize the computation for consecutive $k$-mers.

## Key observation

If $\langle x\rangle$ does not start at one of the $m-1$ last positions of $x$, its prefix of size $m$ is the smallest factor of size $m$ in $x$.

Good news: we can keep track of the smallest factors of size $m$ in $\mathcal{O}(1)$ amortized time using a monotone queue.

## QUICKLY COMPUTING STREAMS OF NECKLACES

## Faster necklace computation

Only consider the cyclic rotations that start:

- at one of the smallest factors of size $m$
- at one of the $m-1$ last positions


## Useful property [Zheng et al. 20]

Assuming $m=\Omega(\log k)$, the probability that a $k$-mer contains duplicate $m$-mers is $o(1 / k)$.


By choosing $m=\Theta(\log k)$,
the smallest factor of size $m$ is unique w.h.p.
$\rightarrow \mathcal{O}(n m)=\mathcal{O}(n \log k)$ for $n$ necklaces (on avg)

## DESIGN OF THE DATA STRUCTURE

## Quotienting the prefixes of necklaces

Quotienting:

- avoids redundancy
- groups consecutive necklaces

Under the hood:

- store the prefixes in a dynamic bitvector supporting rank/select [Marchini \& Vigna 20, Pibiri \& Kanda 21]
- associate suffix buckets using a tiered vector (for fast dynamic insertions) [Bille et al. 17]


## SCALING THE SUFFIX BUCKETS

The structure of the buckets changes dynamically as we add/remove $k$-mers

- for small buckets: packed vector, linear search
- for large buckets: trie, logarithmic search



## LAYOUT OF CBL'S DATA STRUCTURE

1. compute $\langle x\rangle$
2. split $\langle x\rangle$ as $q \| r$
3. query $r$ in the bucket of $q$
$\rightarrow$ faster for
consecutive $k$-mers

plain necklace vector


CBL's data-structure

## COMPARISON TO SOME EXISTING TOOLS

| category | data structure | membership | insert | delete | $\cup \cap \backslash$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| BWT | FM-index | $\checkmark$ | $\times$ | $\times$ | $\times$ |
| - | SBWT | $\checkmark$ | $\times$ | $\times$ | $\times$ |
| - | dynamic BOSS | $\checkmark$ | $\checkmark$ | $\checkmark$ | $\times$ |
| hashing | SSHash | $\checkmark$ | $\times$ | $\times$ | $\times$ |
| - | Bifrost | $\checkmark$ | $\checkmark$ | $\times$ | $\times$ |
| - | Brisk | $\checkmark$ | $\checkmark$ | $\times$ | $\times$ |
| - | Bloom filter | approx | $\checkmark$ | $\times$ | union |
| - | Quotient filter | approx* | $\checkmark$ | $\times$ | union |
| other | Conway-Bromage | $\checkmark$ | $\checkmark$ | $\checkmark$ | $\checkmark$ |
| - | CBL | $\checkmark$ | $\checkmark$ | $\checkmark$ | $\checkmark$ |

[^0]
## Time/memory usage for collections of bacterial genomes from Refseq

| $\bullet$ CBL | $\square$ | SBWT | SBW |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $\star$ | SSHash | Bifrost |  |


| $\bullet$ | CBL | $\square$ | SBWT |
| :--- | :--- | :--- | :--- |
| $\star$ | SSHash | $\Downarrow$ | Bifrost |$\quad$ BufBOSS $\triangle$ HashSet




TLDR: almost as fast as a hash table, more memory-efficient

## Merging collections of bacterial genomes from RefSeq




TLDR: $4 \times$ faster and $3 \times$ smaller than a hash table when merging a billion 31 -mers

WHAT'S NEXT?

## FUTURE STEPS

Improvements of the current structure:

- handle streams of $k$-mers
- improve buckets' memory usage
(some ideas: smaller single buckets, adaptive radix trie)
- use SIMD for core operations

Extending the structure:

- concurrent version of CBL (distribute suffix buckets between threads)
- associate data (e.g. count) to each $k$-mer ( $\rightarrow$ map structure)
[Your suggestion here]: let's discuss!


## TAKE-HOME MESSAGES

- new dynamic structure based on necklaces
- available as a CLI and a Rust library
- very fast queries, cache efficient
- limited memory usage
- scales for large collections
- versatile operations


## Thank you!

```
Try it here:
github.com/imartayan/CBL
```



Preprint (accepted to ISMB)


## ReFERENCES

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## Densifiying the space of necklaces by ranking

The number of necklaces of size $k$ on an alphabet with $\sigma$ letters is

$$
N(k)=\frac{1}{k} \sum_{d \mid k} \varphi\left(\frac{k}{d}\right) \sigma^{d} \sim \frac{\sigma^{k}}{k}
$$

so only a fraction $\frac{1}{k}$ of the universe is actually used


Ranking: given a necklace $\langle x\rangle$, find $i$ s.t. $\langle x\rangle$ is the $i$-th smallest necklace of size $k$
We can compute the rank in $\mathcal{O}\left(k^{2}\right)$ time [Sawada \& Williams 17]
Tradeoff: better locality + compression vs $\mathcal{O}\left(k^{2}\right)$ queries


[^0]:    *exact if a PHF is used

