Using necklaces to build a Locality-preserving and dynamic INDEX FOR K-MERS

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## dNA Sequencing \& Tokenization with k-mers

DNA samples $\stackrel{\circ}{\circ} \longrightarrow$

$\longrightarrow$ CTGAAATG ...

We typically index the words of size $k$ ( $k$-mers) instead of the sequence itself.

In practice, we usually consider $k \leqslant 63$ so that each $k$-mer fits inside a machine word.

CTGAA
TGAAA
GAAAT
AAATG

## 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]

## REVISITING A SIMPLE IDEA: $K$-MERS AS A SPARSE SET OF INTEGERS

[Conway \& Bromage 11]

- 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 use a sparse bitvector to store them

Limitations

- it's not really dynamic
- it's not cache-efficient
- index(ATAACGCCA ) $=49,556$
- index ( TAACGCCAT $)=198,227$
$\rightarrow$ average distance of $4^{k} / 3$


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How can we improve this approach?


## WISH LIST FOR AN IDEAL DATA STRUCTURE

- space-efficient: few bits / $k$-mer
- dynamic: support insertion and deletion after construction
- efficient queries:
- membership
- enumeration
- insertion
- deletion
- locality-preserving: reduce cache misses when querying consecutive $k$-mers



## PRESERVING LOCALITY WITH NECKLACES

## A LOCALITY-PRESERVING ENCODING OF K-MERS



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Alternative encoding based on necklaces
The necklace of $x$ is its smallest cyclic rotation $\langle x\rangle=\min _{0 \leqslant i<k} x^{(i)}$

- $x \mapsto(\langle x\rangle$, rotation index $)$ is a reversible transformation
- necklaces of consecutive $k$-mers share long prefixes


## A CLOSER LOOK AT THE LOCALITY OF NECKLACES

AACGTCATCTCTCATTCTGGTCGTTCTTCCT AACGTCATCTCTCATTCTGITCGTTCTTCCT AACGTCATCTCTCATTCTGTGCGTTCTTCCT AACGTCATCTCTCATTCTGTGAGTTCTTCCT AACGTCATCTCTCATTCTGTGACTTCTTCCT AACGTCATCTCTCATTCTGTGACATCTTCCT AACGTCATCTCTCATTCTGTGACACCTTCCT AACGTCATCTCTCATTCTGTGACACGTTCCT AACGTCATCTCTCATTCTGTGACACGCTCCT AACGTCATCTCTCATTCTGTGACACGCACCT AACGTCATCTCTCATTCTGTGACACGCAGCT AACGTCATCTCTCATTCTGTGACACGCAGGT AACGTCATCTCTCATTCTGTGACACGCAGGG ACACGCAGGGTACGTCATCTCTCATTCTGTG

Size of common prefix
between necklaces of successive $k$-mers ( $k=31$ )


## PRACTICAL USE OF NECKLACES

## Overview of our data structure (CBL)

Quotiented
data structure

Query $x$ :

1. compute $\langle x\rangle$
2. split $\langle x\rangle$ as $q \| r$
3. look for $(q, r)$


## Accelerating the computation of consecutive necklaces

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

Better approach: amortize the computation cost for consecutive queries.

## Key observation

Given a fixed $m$, 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.

## Accelerating the computation of consecutive 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 $o(1 / k)$.


$$
\text { By choosing } m=\Theta(\log k) \text {, }
$$

the smallest factor of size $m$ is unique w.h.p.
$\rightarrow \mathcal{O}(n m)=\mathcal{O}(n \log k)$ for $n$ queries (on average)

## DENSIFIYING THE SPACE OF NECKLACES

## 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}
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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

## CAN WE DO better for consecutive necklaces? (I don't know yet)

Ranking in $\mathcal{O}\left(k^{2}\right)$ is generally too expensive for our use case, but it might be faster to rank necklaces of consecutive $k$-mers.

Since most necklaces of consecutive words share the same starting position, they only differ by a single letter. AACGTCATCTCTCATTCTGGTCGTTCTTCCT AACGTCATCTCTCATTCTGITCGTTCTTCCT

## Formulation in the binary case ( $\sigma=2$ )

How does the rank of $\langle x\rangle$ change if we flip its $i$-th bit?

## CONCLUSION

## Take-home messages \& Open questions

Indexing $k$-mers with their necklaces:

- preserves locality
- improves compression
- fits in well with a quotiented data structure
- combines easily with dynamic operations

Future questions:
-What is the average distance between necklaces of consecutive $k$-mers?

- Can we rank necklaces in subquadratic time?
- Can we accelerate ranking for necklaces of consecutive $k$-mers?


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Thank you!


## References I

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