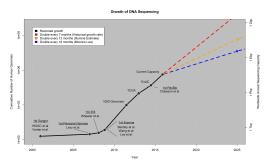
HowDeSBT and **Simka** wedding:

What has been done and what we plan to do?

<u>Téo Lemane</u>, Paul Medvedev, Rayan Chikhi, Pierre Peterlongo

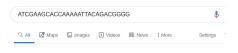
DSB meeting 4 February 2020

High production of sequencing data



Stephens et al.

- ► Non-indexed datasets
- Query only the metadata, not the sequences content



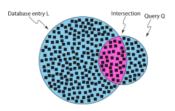
Your search - ATCGAAGCACCAAAAATTACAGACGGGG - did not match any documents.

Query a sequence of interest?

Given experiment sets, and a sequence of interest, which dataset contains this sequence?

In terms of k-mers:

A query Q matches an experiment L if at least a fraction θ of Q's k-mers are present in L.



k-mers based indexing: general processing

- ► k-mers counting for each dataset
 - dealing with sequencing errors
- ▶ indexing each set of k-mers (using AMQ for example)
 - with or without the counting information, depending on the data structure used
- ► merging each index in one
 - ▶ with possible adjustments to optimize storage and query

Example: SBT, HowDeSBT, BIGSI ...

Common to all methods: k-mers counting

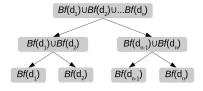
Dealing with sequencing errors

 Usually, k-mer counters use simple error correction (throw out k-mers whose abundance are below a hard threshold)

- ▶ low abundance k-mer: not necessarily an error (under-represented sequence)
- ▶ metagenomics, transcriptomics, cancer genomics (Griffith et al.)

HowDeSBT¹: main idea

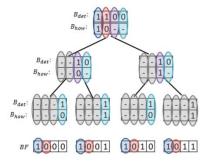
- ► Indexing datasets using their k-mers content
- ▶ Bloom filters as a basic structure
- ► Final index based on Sequence Bloom Tree (Solomon and Kingsford)



- ► One leaf = one experiment.
- ► Internal node represents k-mers in its subtree.
- ► Root node is the union of each BF in the tree.

¹ Harris and Medvedev

HowDeSBT: data structure

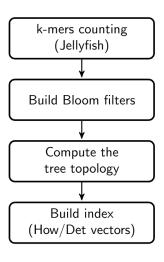


Medvedev, DSB 2019

Two binary vectors by node:

- ▶ B_{det} : Bit is **det**ermined in the subtree?
- ▶ B_{how} : If **det**, **how** is it determined? (0/1)

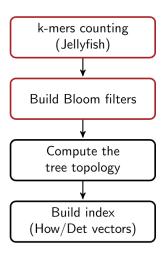
HowDeSBT: pipeline



What we want to do?

- ► Include particular error handling
- ► At the same time: can computation times and memory footprint be improved?

HowDeSBT: pipeline



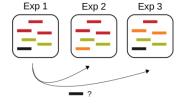
What we want to do?

- ► Include particular error handling
- ► At the same time: can computation times and memory footprint be improved?

► Focus on the two first steps

How to improve errors handling?

- ► Leverage information across samples
- ▶ e.g. For a k-mer seen less than N times, check the count in the other datasets.



▶ Verification only in datasets with compatible metadata

How? Help from Simka (Benoit et al.)

- ► *de novo* comparative metagenomics tool.
- ► Compute ecological distances between samples using k-mers decomposition.

	Α	В	С	D
Α	0	0.2	0.1	0.4
В	0.2	0	0.6	0.3
С	0.1	0.6	0	0.5
D	0.4	0.3	0.5	0

k-mers matrix

▶ After modifications, we can obtain a count table with datasets in columns and k-mers in lines.

	Α	В	<u>C</u>	D
ACGT	12	16	7	1
ACTG	1	0	0	0
CTGA	8	1	8	0
GATA	21	10	21	20
TCGA	12	0	0	4
TCGT	0	1	0	1

k-mers matrix

- ► Leverage information across samples:
 - ► easily applicable with this kind of matrix
- ► Add this new errors handling feature in Simka
- ► Many ways to do it

Count table

	Α	В	<u>C</u>	D
ACGT	12	16	7	1
ACTG	1	0	0	0
CTGA	8	1	8	0
GATA	21	10	21	20
TCGA	12	0	0	4
TCGT	0	1	0	1

Binary matrix

	Α	В	<u>C</u>	D
ACGT	1	1	1	1
CTGA	1	1	1	0
GATA	1	1	1	1
TCGA	1	0	0	1
TCGT	0	1	0	1

Simka: How it works? - Reminder

► Minimizers & super-k-mers (KMC 2, Deorowicz *et al.*)

```
Read CTCATGCACGTTC
k-mers CTCATG

TCATGC

CATGCA

ATGCAC

TGCACG

GCACGT

CACGTT

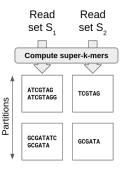
ACGTTC
```

▶ super-k-mers: merged overlapping k-mers that share the same minimizer.

```
2 minimizers, 2 super-k-mers with \emph{k}=6 and \emph{m}=3 CTCATCGAC, TGCACGTTC
```

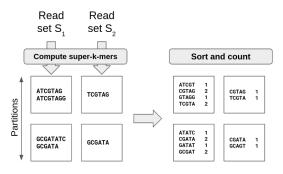
How it works: Counting (GATB-DSK)

► Step 1: Compute super-k-mers from each dataset and store them into partitions according their minimizers.



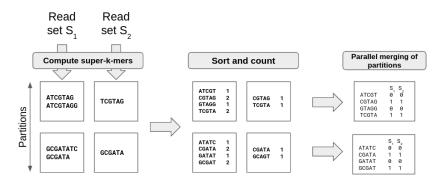
How it works: Counting (GATB-DSK)

► Step 2: In each partition, split super-k-mers into k-mers and sort them. The count is given by identical consecutive k-mers.



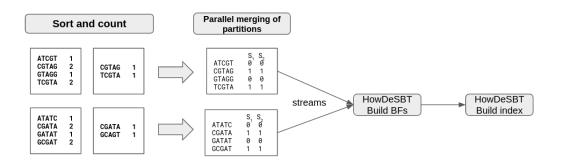
How it works: Counting (GATB-DSK)

► Step 3: Merge equivalent partitions between datasets to obtain sub-matrices.



Simka-HowDeSBT: Current implementation

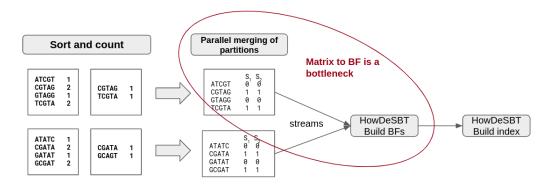
► Merge between datasets and stream each lines to build Bloom filters with HowDeSBT.



▶ https://github.com/tlemane/Simka-HowDeSBT

Simka-HowDeSBT: Current implementation

► Merge partitions and stream each lines to build Bloom filters with HowDeSBT.



► https://github.com/tlemane/Simka-HowDeSBT

Matrix to Bloom filters

- ► Currently two possibilities:
 - ► Stream each line to HowDeSBT
 - ► Binary matrix on disk

In both cases: requires to have as many open filters in memory as there are data sets



Solutions?

- ► Store matrix on disk:
 - ► Multiple passes
 - ► Transposition

- ► Build Bloom filters on files:
 - too many files open at the same time with large number of datasets

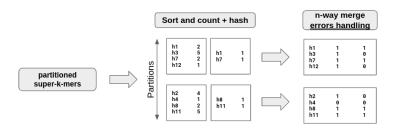
Can we do better?

► Currently, Simka gives textual k-mers counted to HowDeSBT which builds Bloom filters

- ► Modify early steps of Simka to build directly Bloom filters.
 - ► One possibility: hash instead of k-mers

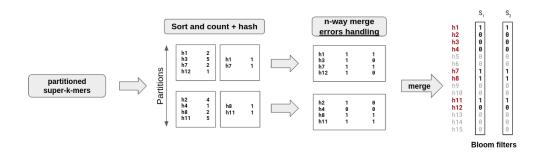
Hash instead of k-mers

lacktriangle hash k-mers at step 2 ightarrow obtain sorted hash values in partitions

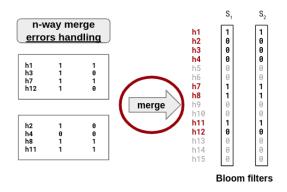


Merging sub-matrices

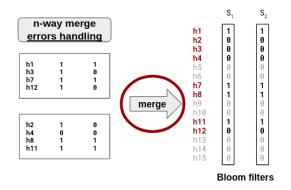
Set missing hash values to $0 \rightarrow Bloom$ filters



► Obtain directly Bloom filters but with one more merge



► Obtain directly Bloom filters but with one more merge

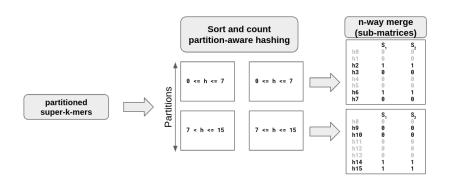


► Can we avoid this second merge?

Partition-aware hashing

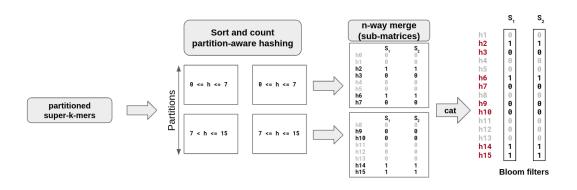
Define hash space according to the partition

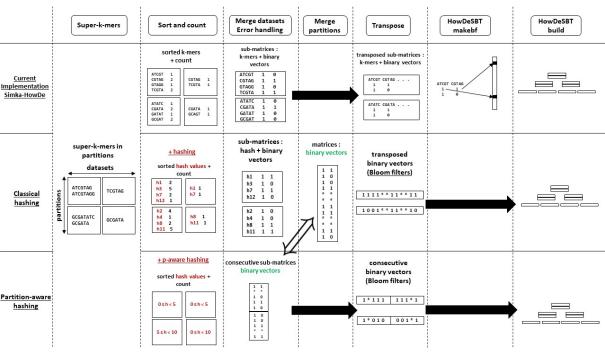
Requires to compute minimizers at query to select the right hash space

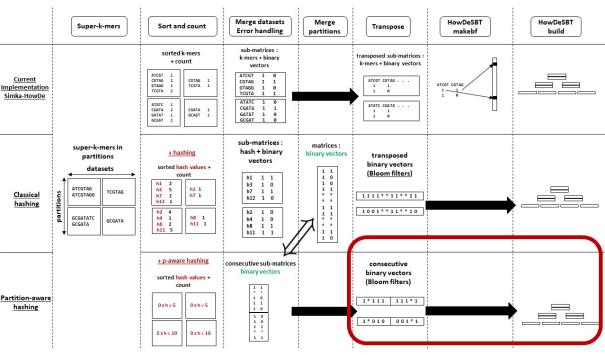


Partition-aware hashing

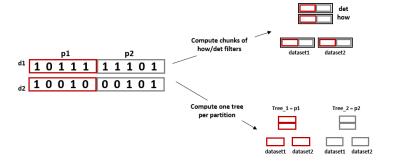
► Merging step becomes concatenation

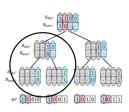






Can the tree computation be parallelized?





- ► Build one global tree vs one tree per partition?
 - ► Are there any issues with building one tree per partition?
 - ► Storage?
 - ► Query efficiency?
 - ▶ In the case of tree per partition: classical hashing ⇔ partition-aware hashing

- ► Build one global tree vs one tree per partition?
 - ▶ Are there any issues with building one tree per partition?
 - Storage?
 - Query efficiency?
 - ▶ In the case of tree per partition: classical hashing ⇔ partition-aware hashing
- ► Build one global tree?
 - ▶ Partition-aware seems to be better than classical hashing
 - ► How to define efficient hash spaces?
 - ► Consideration relative to the tree topology computation

- ► Build one global tree vs one tree per partition?
 - ▶ Are there any issues with building one tree per partition?
 - Storage?
 - ► Query efficiency?
 - ▶ In the case of tree per partition: classical hashing ⇔ partition-aware hashing
- ► Build one global tree?
 - ▶ Partition-aware seems to be better than classical hashing
 - ► How to define efficient hash spaces?
 - ► Consideration relative to the tree topology computation
- What about collisions?

- ▶ Build one global tree vs one tree per partition?
 - ▶ Are there any issues with building one tree per partition?
 - ► Storage?
 - ► Query efficiency?
 - ▶ In the case of tree per partition: classical hashing ⇔ partition-aware hashing
- ► Build one global tree?
 - ▶ Partition-aware seems to be better than classical hashing
 - ► How to define efficient hash spaces?
 - ► Consideration relative to the tree topology computation
- ► What about collisions?
- **▶** Implementation

S1

index ⇔ hash value

р0	p1	p2	
n k-mers	n k-mers	n k-mers	
[0, 33[[33, 66[[66, 100[
100-bits Bloom filter			

index ⇔ hash value

ρυ	bτ	pz		
x k-mers	y k-mers	z k-mers		
[0, 40[[40, 50[[50, 100[
100-bits Bloom filter				

Solution: using oversized hash function