#### TwoPaCo

An efficient algorithm to build the compacted de Bruijn graph from many complete genomes

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# Parte I

## Introduction



# A pan-genomic algorithm



# De Bruijn graph



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# Compacted de Bruijn graph



# The problem



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#### **Junctions**



# Parte II

# The algorithm



# Naive algorithm

- Store all (k+1)-mers in a hash table
- For each k-mers query the possible edge
- If multiple matching, mark as a junction



TwoPaCo

# The memory issue

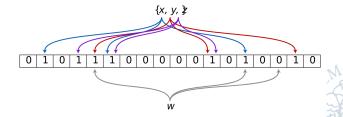


TwoPaCo

#### Bloom filter

A space-efficient probabilistic hash table

Bitmap 
$$V$$
 of size  $b$ ,  $h$  hash functions  $f_0, f_1, ..., f_{h-1}: U \to [0, b-1]$  insertion( $x$ )  $\to V[f_i(x)] = 1$ ,  $\forall \ 0 \le i < h$  contains( $x$ )  $\to$  probabily yes if  $V[f_i(x)] == 1$ ,  $\forall \ 0 \le i < h$ 



Probability of false positive, after n insertion:  $p_{FP} \simeq (1 - e^{-hn/m})^h$ 

11/24

4 0 > 4 70 > 4 75 > 4

#### Two Pass version



TwoPaCo

# The memory issue $^{2}$



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# Multiple rounds: dealing with memory restrictions



#### Parallelization scheme



# Parte III

# Results



#### Source code & Dataset

Original source code by medvedev group available on:

https://github.com/medvedevgroup/TwoPaCo

Personal implementation available on:

https://github.com/GaspareG/TwoPaCo

Dataset for experiments:

5 humans (from human reference genome) 8 primates (from ...) 62 Escherichia coli (from ...) 100 simulated human (from ...)



# Memory complexity

# The memory complexity is the maximum among the first and the second pass of TwoPaCo

- ullet First pass: insert all k-mers in a bloom filter of size b
- Second pass: store all junction candidates in a hash table

How many junction candidates?

- Real junction, J
- False positive induced from the bloom filter, FP

Result:  $O(\max(b, (J + FP)k))$ 



# Time complexity

# The memory complexity is the sum between the first and the second pass of TwoPaCo

- ullet First pass: insert all k-mers in a bloom filter of size b using h hash functions
- Second pass: store all junction candidates in a hash table

How many 
$$k\text{-mers?}$$
  $O(m)\text{, where }m=\Sigma_{s\in S}|s|$  is the total input size

How many junction candidates?

- Real junction, J
- False positive induced from the bloom filter, FP

Result: 
$$O(mh + (|G_c| + FP)k)$$



# Performance comparison

Previous approach to the problem:

- Sibelia (St. Petersburg University, RU, 2012)
- SplitMEM (Stony Brook University, USA, 2014)
- bwt-based (Harbin Institute of Technology, CN, 2016)
- TwoPaCo (The Pennsylvania State University, USA, 2016)

| Algorithm | Time complexity         | Memory complexity     |
|-----------|-------------------------|-----------------------|
| Sibelia   | O(m)                    | O(m)                  |
| SplitMEM  | $O(m \log g)$           | $O(m +  G_c )$        |
| bwt-based | O(m)                    | O(m)                  |
| TwoPaCo   | $O(mh + ( G_c  + FP)k)$ | $O(\max(b, (J+FP)k))$ |

- $m = \sum_{s \in S} |s|$ , total input size
- $g = \max_{s \in S} |s|$ , size of the biggest genoma
- J and  $G_c$ , number of vertex and edge in the de Bruijn Graph
- b and h, size of the bloom filter table and number of hash functions
- FP, number of false positives in first pass

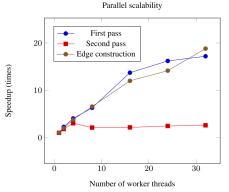
# Fixed memory

How many round do we need to compress the graph without exceeding a given memory threshold?

| Threshold | Used memory | Bloom filter size         | Running time | Rounds |
|-----------|-------------|---------------------------|--------------|--------|
| 10GB      | 8.62GB      | 8.59GB (2 <sup>33</sup> ) | 4h           | 1      |
| 8GB       | 6.73GB      | 4.29GB (2 <sup>32</sup> ) | 7h           | 3      |
| 6GB       | 5.98GB      | 4.29GB (2 <sup>32</sup> ) | 9h           | 4      |
| 4GB       | 3.51GB      | $2.14GB(2^{31})$          | 11h          | MA 61  |

## Parallel scalability

How does the performance of TwoPaCo improve to the increasing of working threads?



- First pass: great improvement thanks to concurrent bloom filter
- Second pass: slight improvement due to race-conditions on the hash table
- Edge construction: great improvement thanks to k-mers independency 3

4 0 > 4 70 > 4 3 >

## Bloom Filter false positive

Is the Bloom Filter really useful in reducing junction candidates?

| Dataset    | k   | Initial           | First pass     | Second pass      | %Fp    |
|------------|-----|-------------------|----------------|------------------|--------|
| 62 E.coli  | 25  | $3.101*10^{8}$    | $2.464 * 10^7$ | $2.457 * 10^7$   | 0.31%  |
| 62 E.coli  | 100 | $3.101 * 10^{8}$  | $2.284 * 10^7$ | $9.492 * 10^6$   | 58.45% |
| 7 humans   | 25  | $2.120*10^{10}$   | $3.489 * 10^9$ | $2.974 * 10^9$   | 14.78% |
| 7 humans   | 100 | $2.120*10^{10}$   | $1.374 * 10^9$ | $1.882 * 10^{8}$ | 86.30% |
| 8 primates | 25  | $2.454 * 10^{10}$ | $5.423 * 10^9$ | $5.401*10^9$     | 0.39%  |
| 8 primates | 100 | $2.454 * 10^{10}$ | $1.174*10^9$   | $5.024*10^{8}$   | 57.20% |

- Initial: total number of k-mers in dataset
- First pass: number of junction candidates (using a bloom filter)
- Second pass: real number of junction (using an hash table)
- False positive: 100\*(First pass Second pass)/(First pass)

#### Conclusion

# Thanks for your attention!

