



Comparative Genome Analysis using Sample-Specific String Detection in Accurate Long Reads

Parsoa Khorsand

Luca Denti

HGSVC

Paola Bonizzoni

Rayan Chikhi

Fereydoun Hormozdiari



UCDAVIS



MOTIVATION AND CONTRIBUTION

Long reads improve variant detection:

- SVs in repeated regions of the genome
- Hard to detect variations

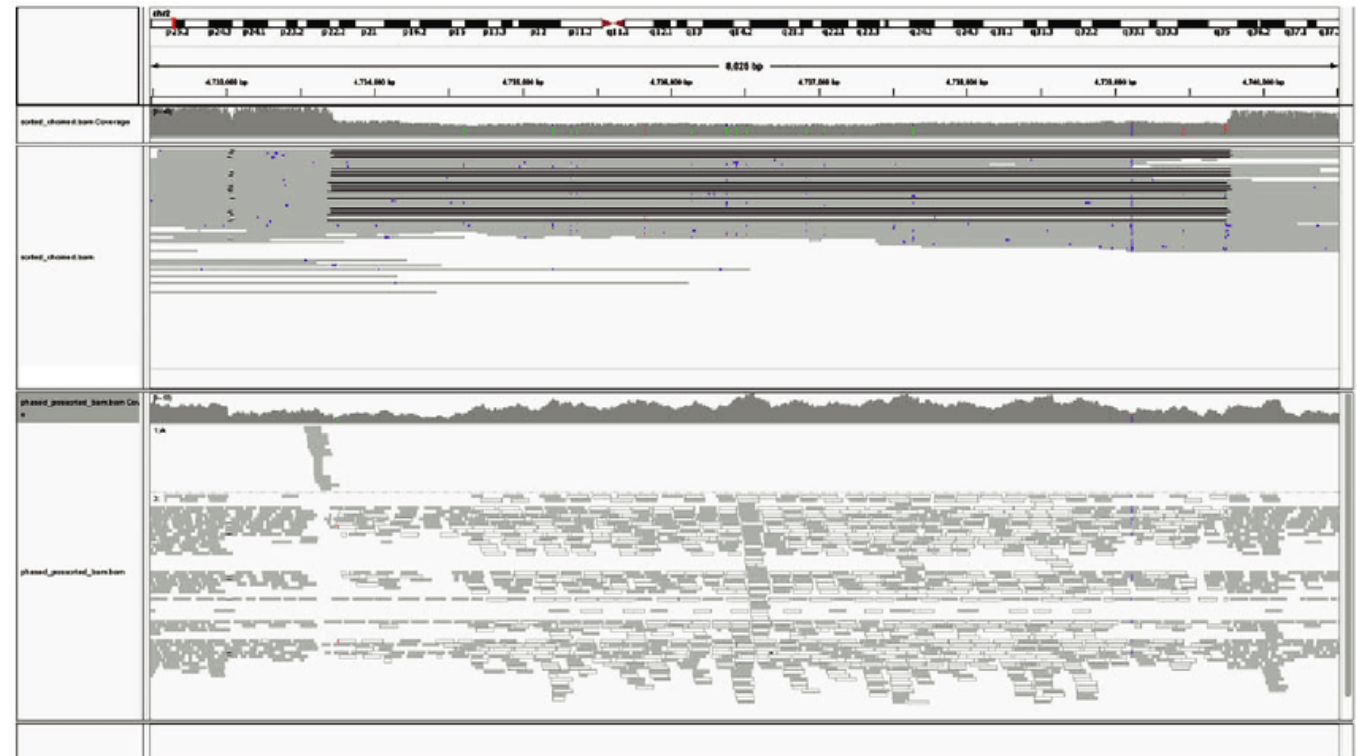


Fig.3 from Pollard et al. **Long reads: their purpose and place.** *Human molecular genetics* (2018)

MOTIVATION AND CONTRIBUTION

Long reads improve variant detection:

- SVs in repeated regions of the genome
- Hard to detect variations

Current approaches:

- Alignment-based (e.g., CuteSV) fails to detect complex SVs (inaccurate alignments)
- Alignment-free (e.g., MALVA, Nebula) are limited by kmer size (short reads)

MOTIVATION AND CONTRIBUTION

Long reads improve variant detection:

- SVs in repeated regions of the genome
- Hard to detect variations

Current approaches:

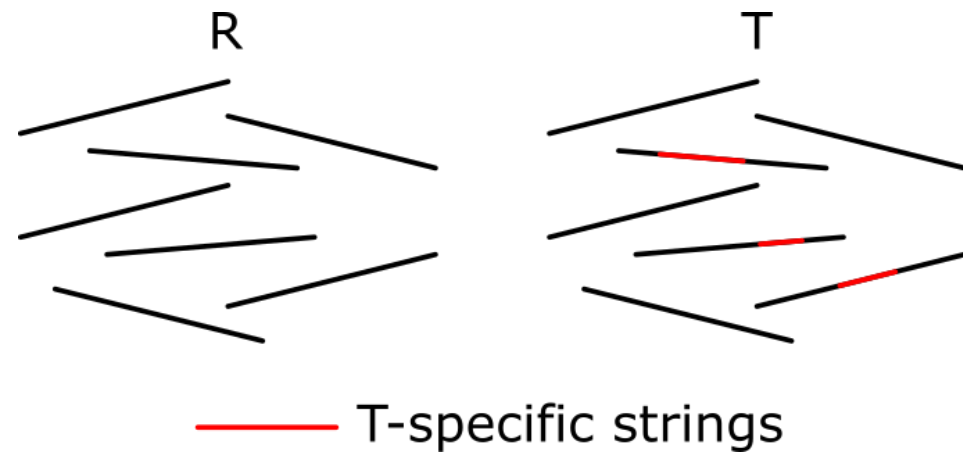
- Alignment-based (e.g., CuteSV) fails to detect complex SVs (inaccurate alignments)
- Alignment-free (e.g., MALVA, Nebula) are limited by kmer size (short reads)

Novel **alignment-free framework** for variant detection from HiFi long reads **not limited by** kmer size

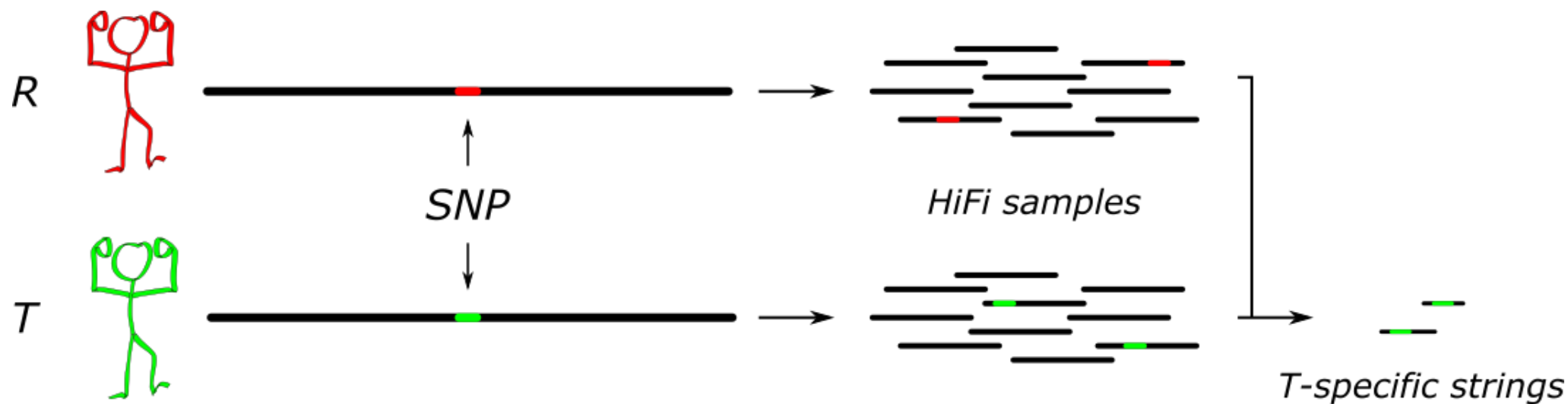
SAMPLE-SPECIFIC STRINGS

Given two set of strings R and T, we define **T-specific** any string:

- occurring in T (*substring*)
- not occurring in R (*substring*)



COMPARING TWO INDIVIDUALS



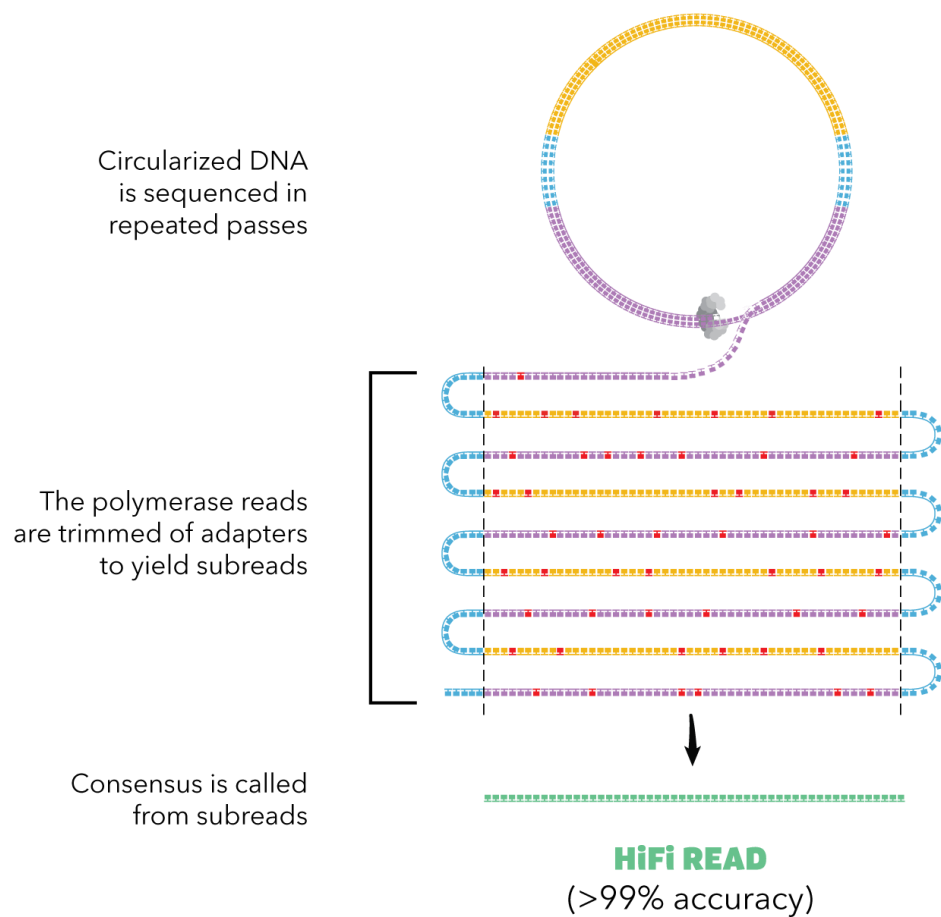
Rationale: any variation specific to T should produce one (or more) T-specific string



PRELIMINARIES



PACBIO SINGLE MOLECULE HIGH-FIDELITY (HiFi reads)



**SHORT
READS**

100%

**HiFi
READS**

Accuracy

**LONG
READS**

80%

0

Read Length (kb)

50

FMD-INDEX¹

FM-Index of a *bidirectional* collection of DNA sequences:

$$R_1 \bar{R}_1 R_2 \bar{R}_2 \dots R_n \bar{R}_n$$

(\bar{R} is the reverse-and-complement of R)

FMD-INDEX¹

FM-Index of a *bidirectional* collection of DNA sequences:

$$R_1 \bar{R}_1 R_2 \bar{R}_2 \dots R_n \bar{R}_n$$

(\bar{R} is the reverse-and-complement of R)

Single index for both forward and reverse strands that allows efficient, $O(1)$, queries:

- backward extensions
- **forward extensions**

FMD-INDEX¹

FM-Index of a *bidirectional* collection of DNA sequences:

$$R_1 \bar{R}_1 R_2 \bar{R}_2 \dots R_n \bar{R}_n$$

(\bar{R} is the reverse-and-complement of R)

Single index for both forward and reverse strands that allows efficient, $O(1)$, queries:

- backward extensions
- **forward extensions**

Indeed, by backward extending pattern P with a , we also forward extend \bar{P} with t .

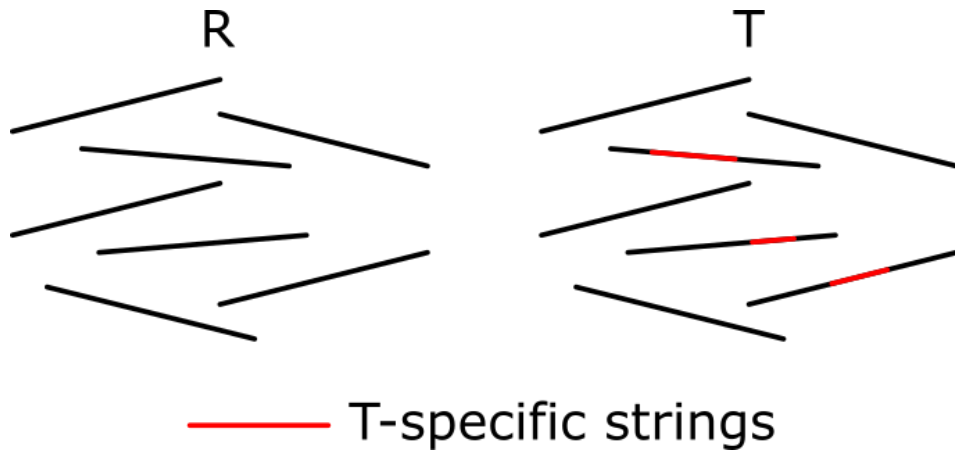


CONTRIBUTION |

T-SPECIFIC STRINGS

Given (R,T) , we define **T-specific** any string:

- occurring in T (substring)
- not occurring in R (substring)

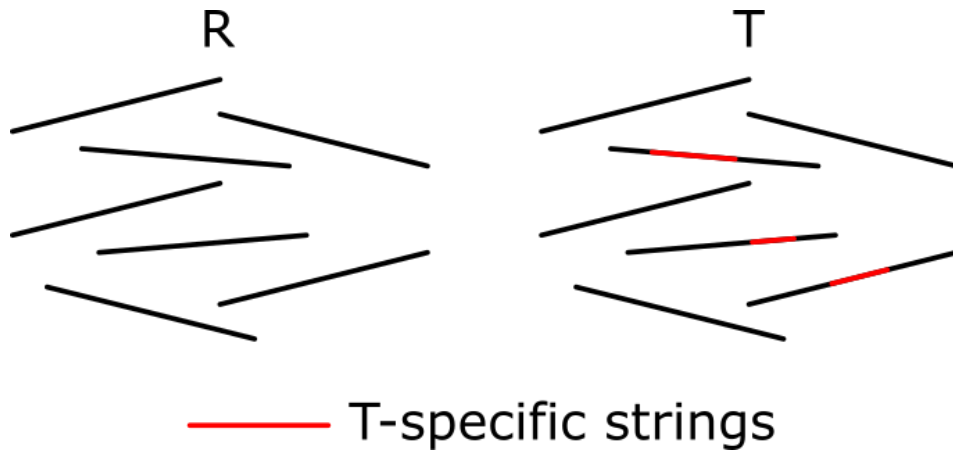


T-SPECIFIC STRINGS

Given (R,T), we define **T-specific** any string:

- occurring in T (substring)
- not occurring in R (substring)

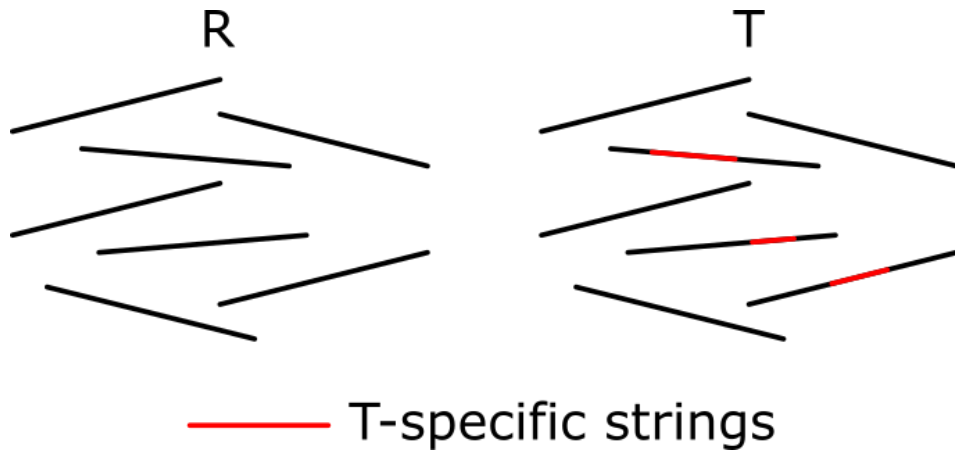
$R = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{T} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$
 $T = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{A} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$



T-SPECIFIC STRINGS

Given (R, T) , we define **T-specific** any string:

- occurring in T (substring)
- not occurring in R (substring)



$$R = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{T} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$$
$$T = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{A} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$$

ACAAGAG
ACAAGA
CAAGAG
AAGAG
ACAAG
AAG
...

$O(|T|^2)$ T-specific strings

T-SPECIFIC STRINGS

Given (R, T) , we define **T-specific** any string:

- occurring in T (substring)
- not occurring in R (substring)

- s.t. no other T-specific strings are contained in it
e.g., it's the shortest (*substring-free property*)

$R = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{T} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$
 $T = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{A} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$

ACAAGAG
ACAAGA
CAAGAG
AAGAG
ACAAG
AAG
...

$O(|T|^2)$ T-specific strings

T-SPECIFIC STRINGS

Given (R, T) , we define **T-specific** any string:

- occurring in T (substring)
- not occurring in R (substring)
- s.t. no other T-specific strings are contained in it
e.g., it's the shortest (*substring-free property*)

$R = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{T} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$
 $T = \{ \overset{0}{A} \overset{1}{C} \overset{2}{A} \overset{3}{A} \overset{4}{G} \overset{5}{A} \overset{6}{G} \}$

ACAAGAG
ACAAGA
CAAGAG
AAGAG
ACAAG
AAG
...

$O(|T|^2)$ T-specific strings

AGA
AA

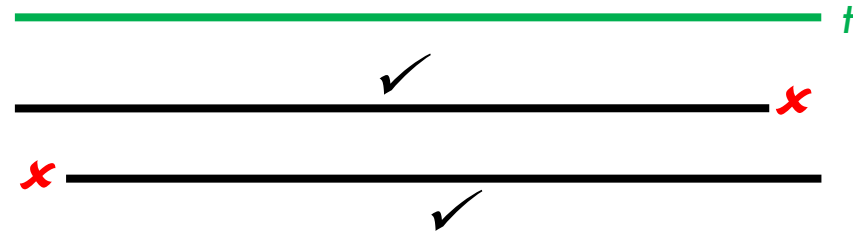


Substring-free
T-specific strings

PROPERTY OF T-SPECIFIC STRINGS

Let t be a n -long T-specific string. By definition:

- $t[0:n]$ **isn't** found in R
- $t[0:n-1]$ **is** found in R
- $t[1:n]$ **is** found in R

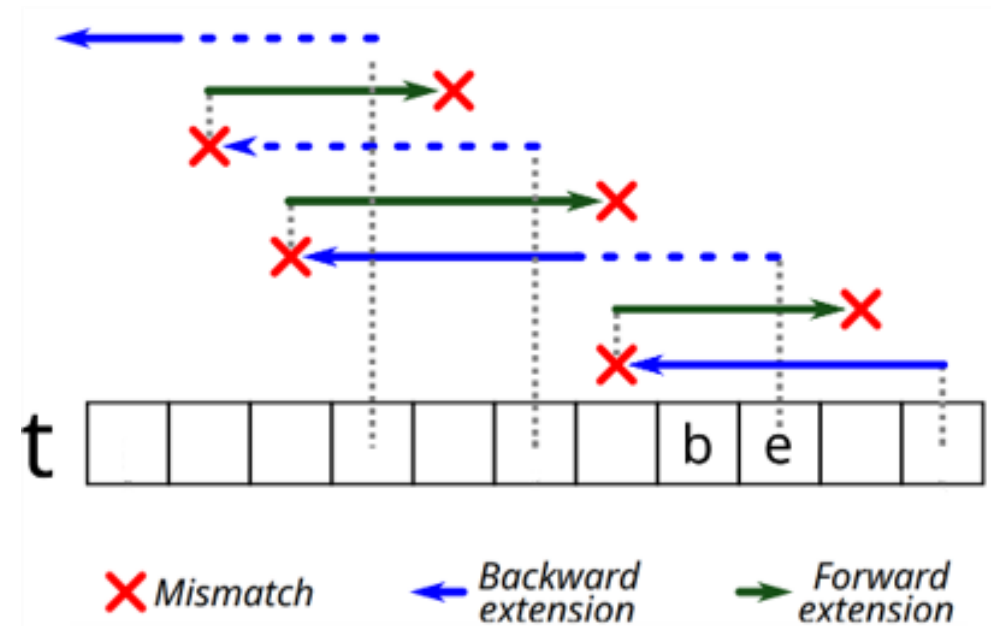


The first and last base of t are **mismatches**.

ALGORITHM: PING-PONG

Input: two sets of strings (T,R)

Output: T-specific strings w.r.t. R

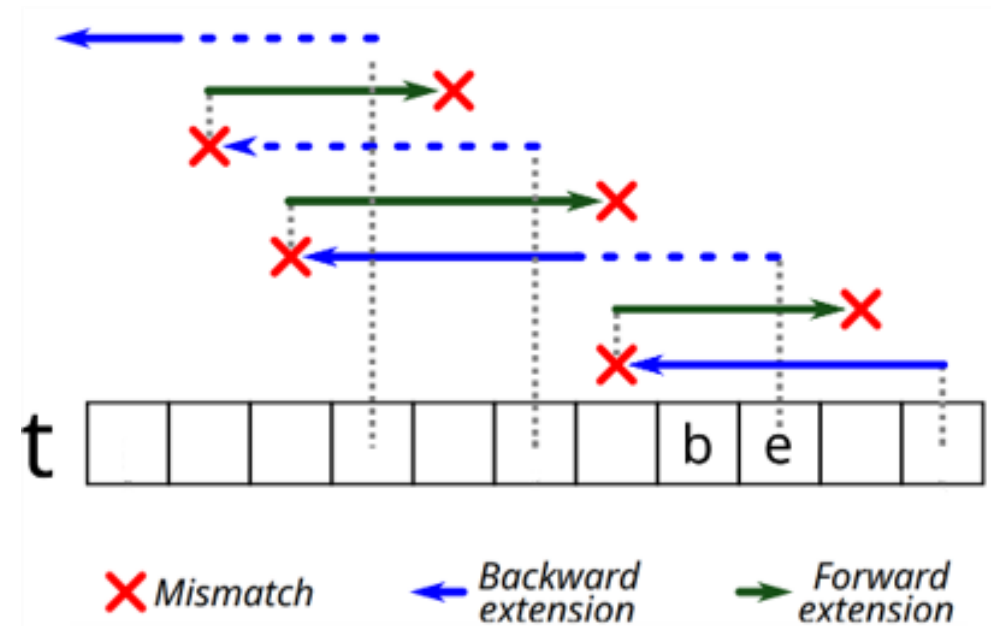


ALGORITHM: PING-PONG

Input: two sets of strings (T,R)

Output: T-specific strings w.r.t. R

1. Build the FMD-Index of R

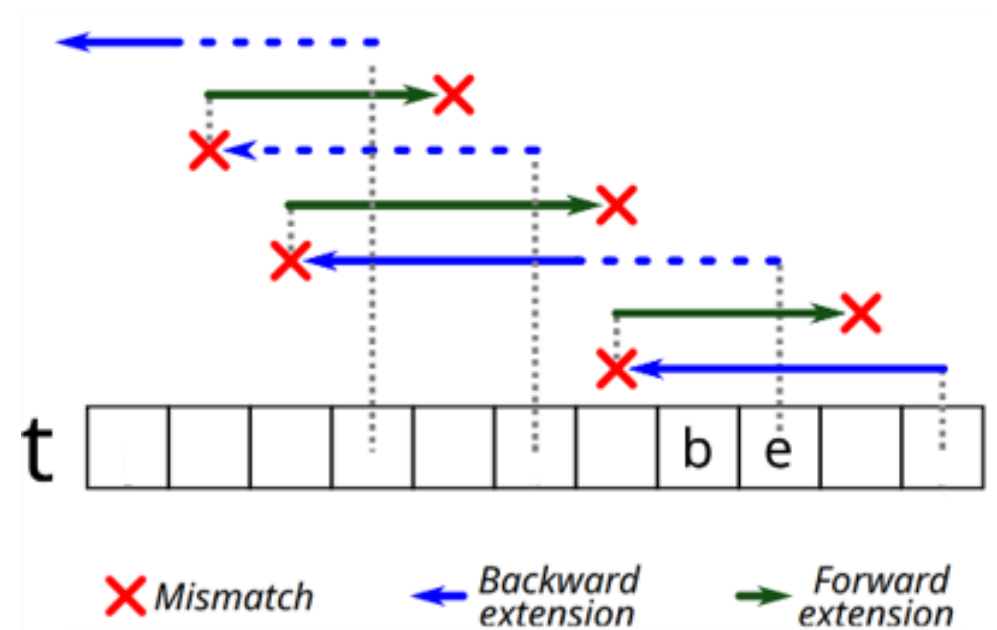


ALGORITHM: PING-PONG

Input: two sets of strings (T,R)

Output: T-specific strings w.r.t. R

1. Build the FMD-Index of R
2. For each string t in T:
 1. Traverse backward the index until a mismatch
 2. Traverse forward the index until a mismatch
 3. Return the string between the mismatches (included)
 4. Reiterate

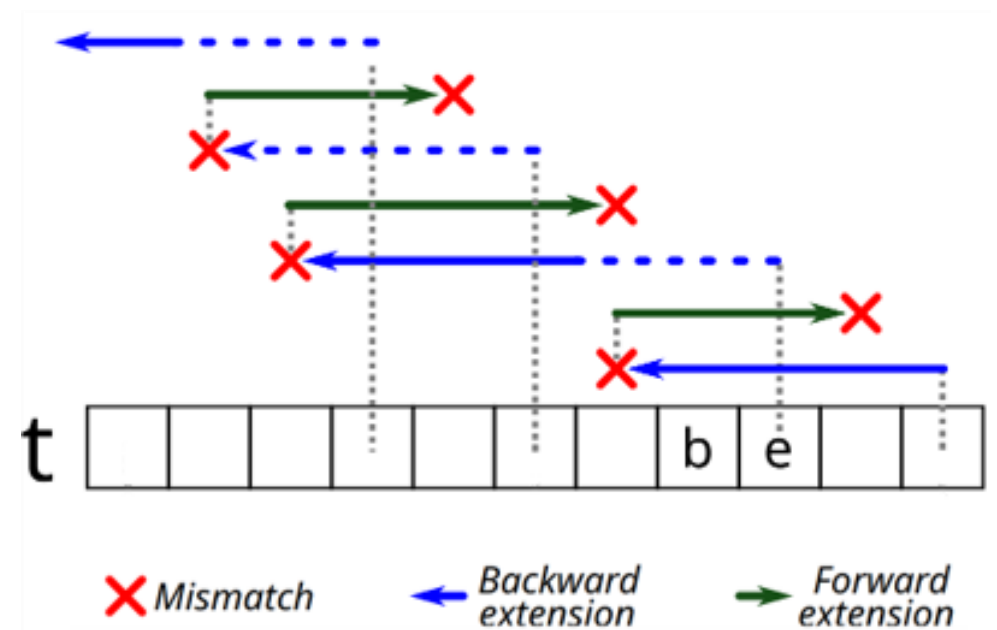


ALGORITHM: PING-PONG

Input: two sets of strings (T,R)

Output: T-specific strings w.r.t. R

1. Build the FMD-Index of R
2. For each string t in T:
 1. Traverse backward the index until a mismatch
 2. Traverse forward the index until a mismatch
 3. Return the string between the mismatches (included)
 4. Reiterate



Q1 to audience: is this notion related to the one of SMEMs?

COMPLEXITY

Property:

$$S_T = \bigcup_{t \in T} S_t$$

COMPLEXITY

Property:

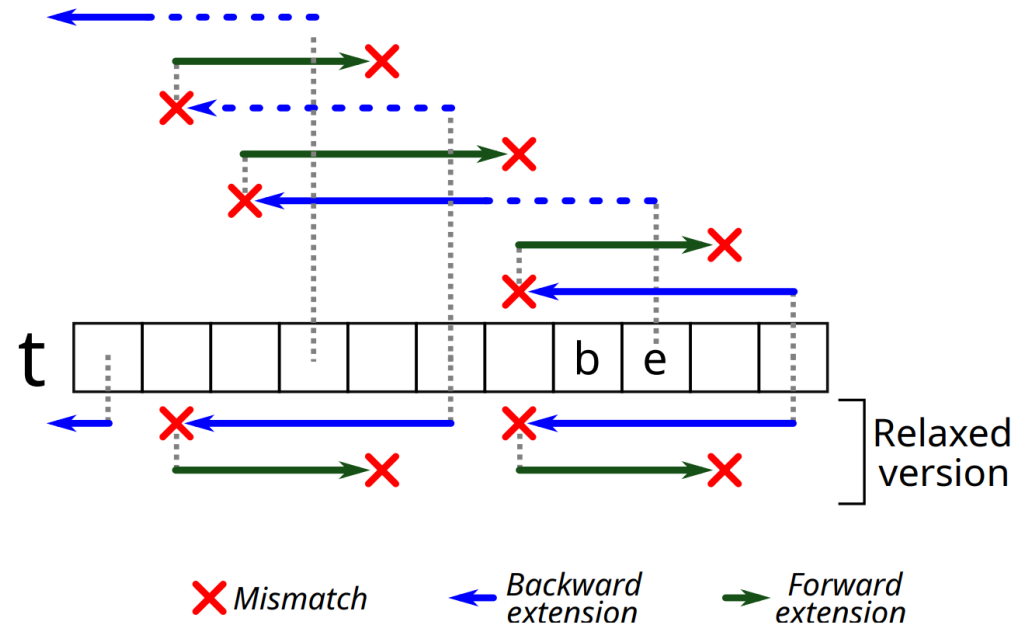
$$S_T = \bigcup_{t \in T} S_t$$

Theorem:

Ping-Pong algorithm retrieves all T -specific strings in $O(n^2)$, where n is the total length of T .

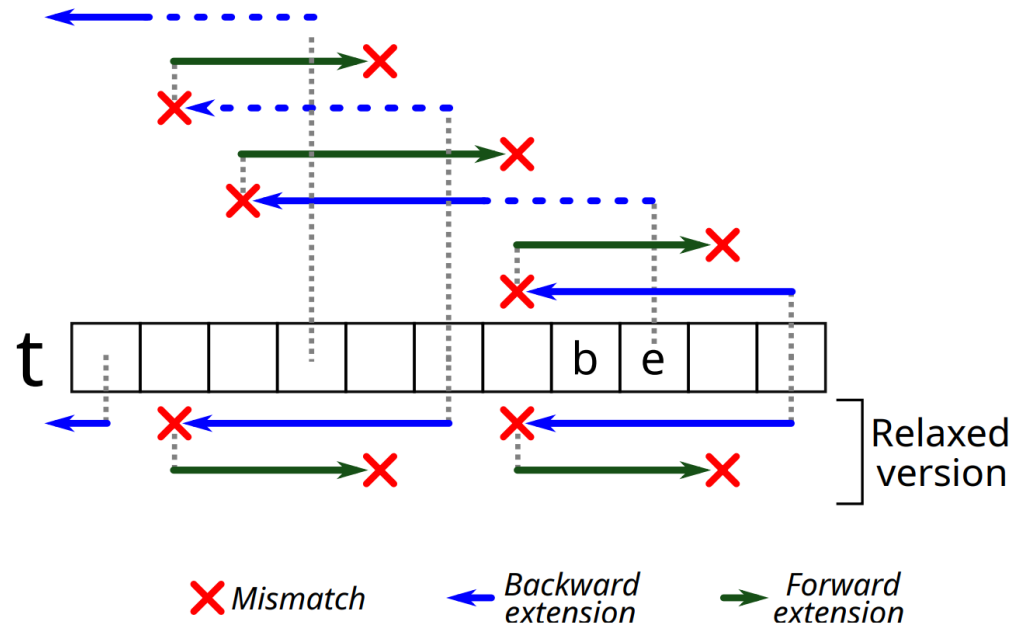
PING-PONG (RELAXED)

Instead of retrieving all t-specific strings, we can retrieve all strings “non-overlapping” on t.



PING-PONG (RELAXED)

Instead of retrieving all t -specific strings, we can retrieve all strings “non-overlapping” on t .



Theorem:

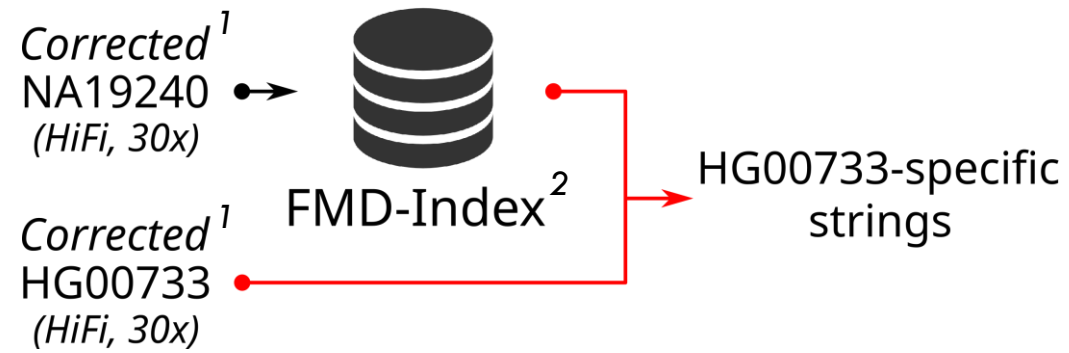
Relaxed Ping-Pong algorithm retrieves a subset of T -specific strings in $O(n)$.



EXPERIMENTAL EVALUATION



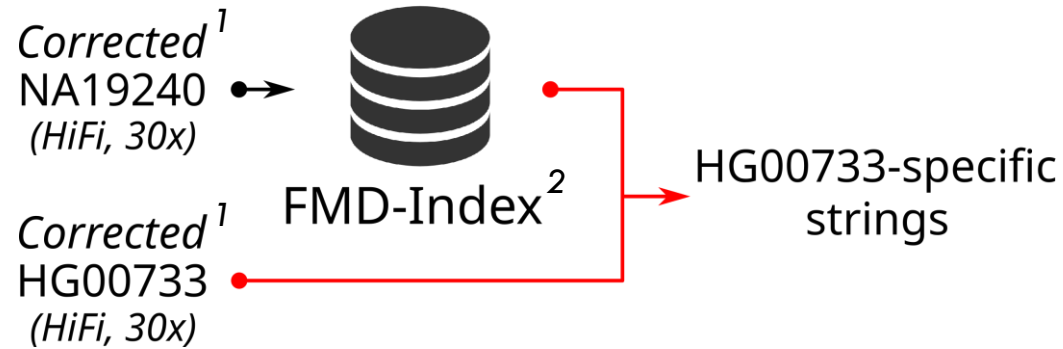
REAL HIFI EXPERIMENTS



¹ Warren et al. *ntEdit: scalable genome sequence polishing*. *Bioinformatics* (2019)

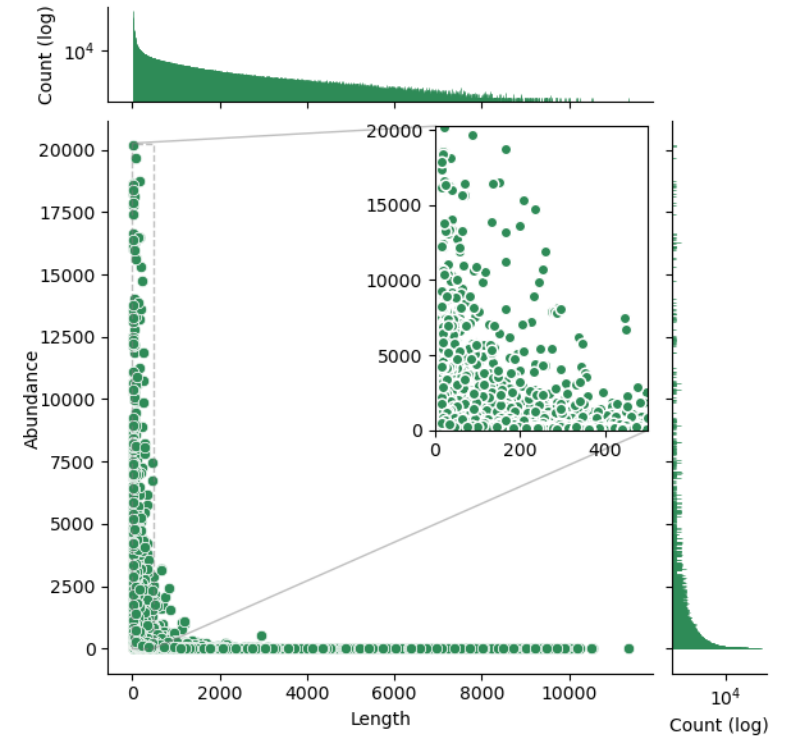
² Heng Li. *Fast construction of FM-index for long sequence reads*. *Bioinformatics* (2014)

REAL HIFI EXPERIMENTS



34 219 149 HG00733-specific strings (≥ 5 abundance)

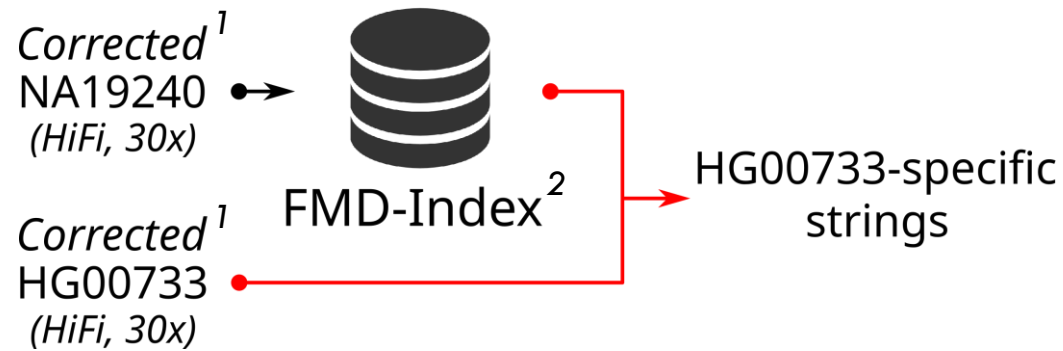
(7 125 436 relaxed)



¹ Warren et al. *ntEdit: scalable genome sequence polishing*. *Bioinformatics* (2019)

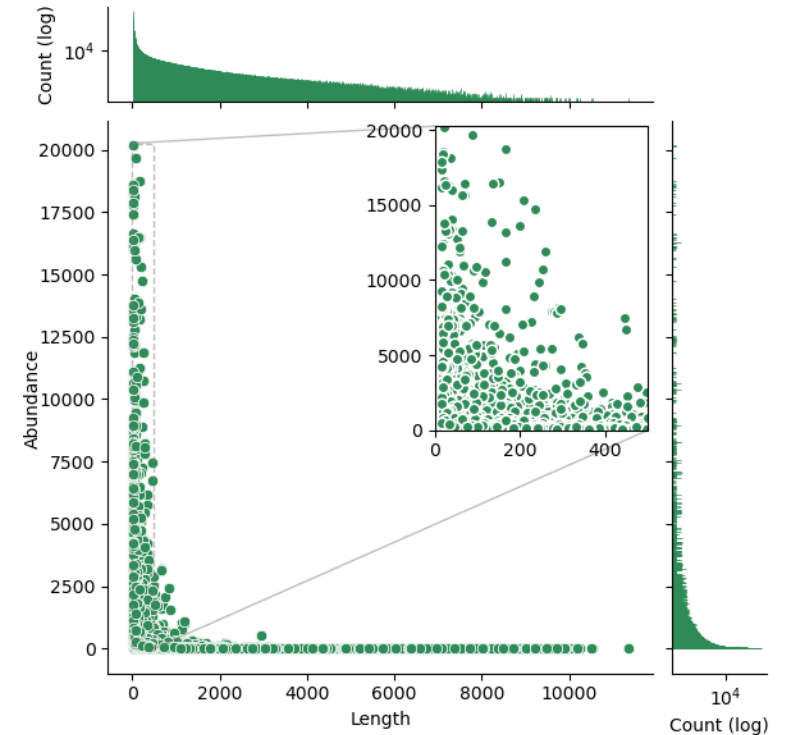
² Heng Li. *Fast construction of FM-index for long sequence reads*. *Bioinformatics* (2014)

REAL HIFI EXPERIMENTS



34 219 149 HG00733-specific strings (≥ 5 abundance)

(7 125 436 relaxed)

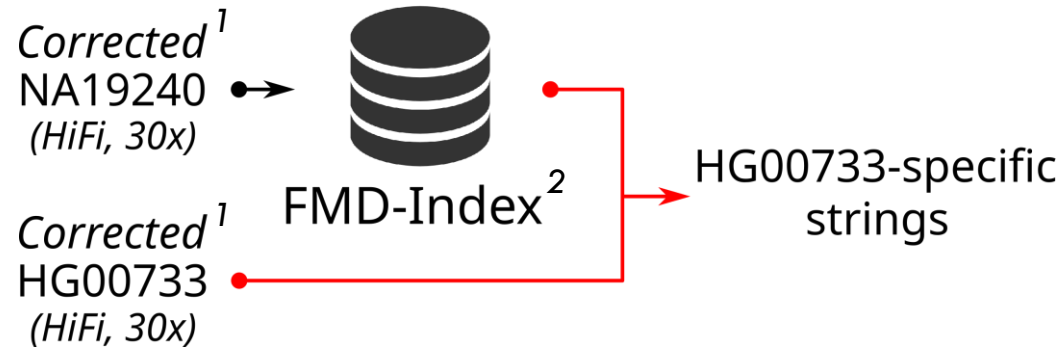


	ntEdit	Indexing	Retrieval	Total/Peak
Time (hh:mm)	05:03	20:30 😞	11:59	37:32
RAM (GB)	36	25	242 🤖	242

¹ Warren et al. *ntEdit: scalable genome sequence polishing*. *Bioinformatics* (2019)

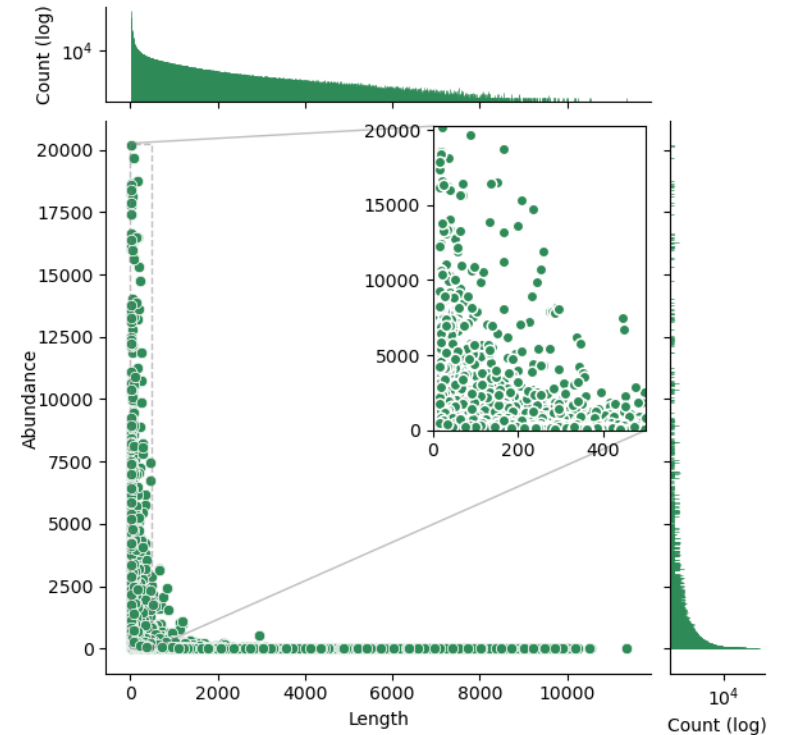
² Heng Li. *Fast construction of FM-index for long sequence reads*. *Bioinformatics* (2014)

REAL HIFI EXPERIMENTS



34 219 149 HG00733-specific strings (≥ 5 abundance)

(7 125 436 relaxed)



	ntEdit	Indexing	Retrieval	Total/Peak
Time (hh:mm)	05:03	20:30 😞	11:59	37:32
RAM (GB)	36	25	242 🤖	242

¹ Warren et al. *ntEdit: scalable genome sequence polishing*. *Bioinformatics* (2019)

² Heng Li. *Fast construction of FM-index for long sequence reads*. *Bioinformatics* (2014)

Q2 to audience: can we make indexing faster?

VALIDATION

1. **Contigs-based:** *check if our algorithm is correct*
2. **Haplotype-based:** *check if HG00733-specific strings cover variations specific to
HG00733 (w.r.t. NA19240)*

Comparison: specific 31-mers and 101-mers (KMC)

VALIDATION USING CONTIGS

Goal: *check if HG00733-specific strings are effectively specific (algorithm correctness)*

How:

1. get the HG00733 and NA19240 contigs
2. align specific strings/kmers to the contigs
3. evaluate alignment quality

[Porubsky et al. Nat. Comm. 2020]

(bbmap2/bwa/minimap2)

Metric: C-Precision *(fraction of strings aligned perfectly only to HG00733 contigs)*

VALIDATION USING CONTIGS

Goal: *check if HG00733-specific strings are effectively specific (algorithm correctness)*

Metric	Method	Missed	Total	%
C-precision	Exact Ping-Pong	7 024 433	34 219 149	79.47
	Relaxed Ping-Pong	669 324	7 125 436	90.61
	31-mers	23 170 031	97 975 734	76.35
	101-mers	84 211 940	387 221 925	78.25

Metric: C-Precision (*fraction of strings aligned perfectly only to HG00733 contigs*)

VALIDATION USING HAPLOTYPES

Goal: *check if HG00733-specific strings cover variations specific to HG00733*

How:

1. get the known variations (VCF) *[Porubsky et al. Nat. Comm. 2020]*
2. build HG00733 haplotypes *(BCFtools)*
3. align specific strings/kmers to the haplotypes *(bbmap2/bwa/minimap2)*
4. Intersect alignments and specific variations *(BEDtools)*

Metric: Recall and H-Precision

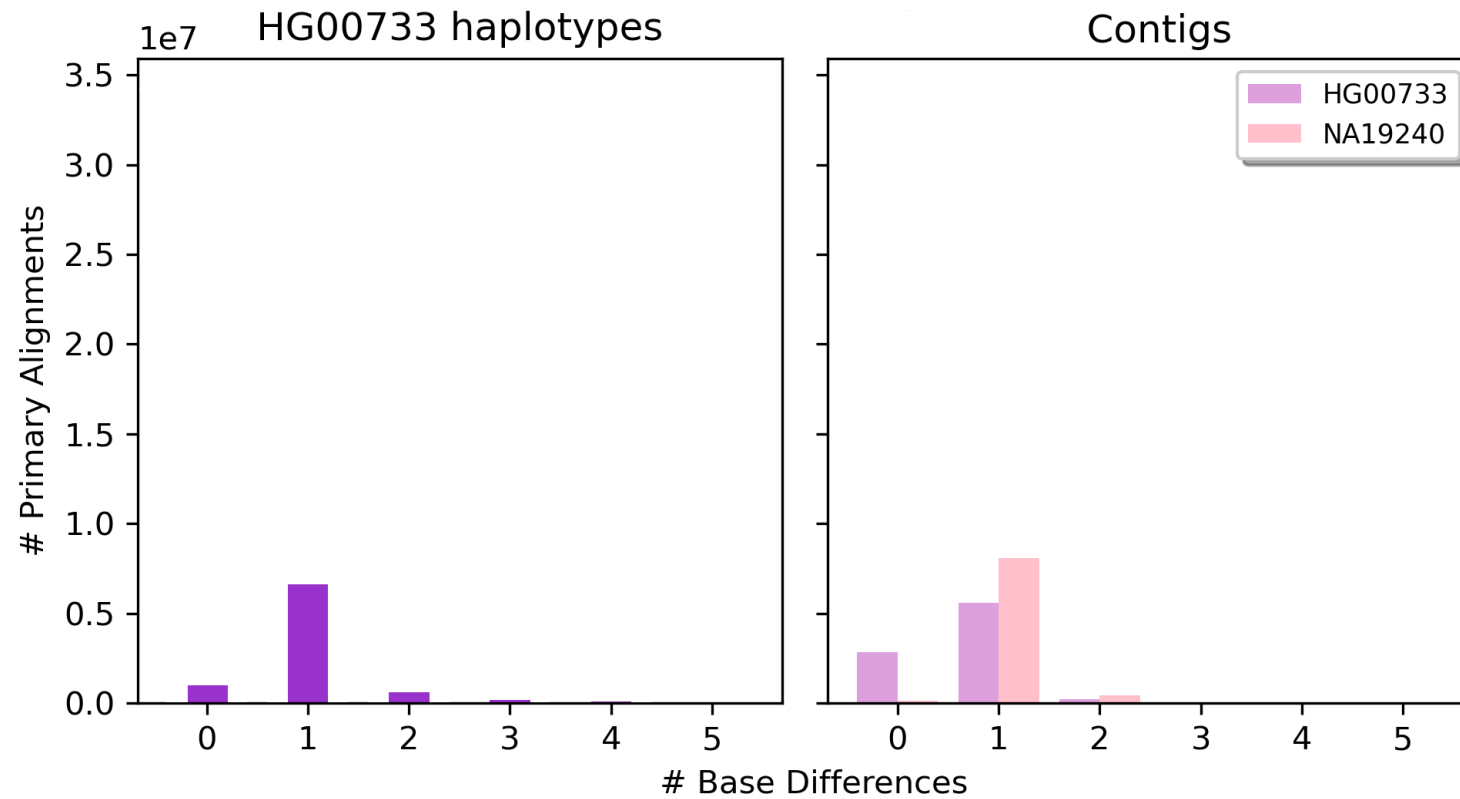
VALIDATION USING HAPLOTYPES

Goal: check if HG00733-specific strings cover variations specific to HG00733

Metric	Method	Missed	Total	%
Recall	Exact Ping-Pong	72 301	3 884 411	98.14
	Relaxed Ping-Pong	236 201		93.92
	31-mers	379 866		90.22
	101-mers	81 065		97.91
H-precision	Exact Ping-Pong	9 093 407	34 219 149	73.43
	Relaxed Ping-Pong	1 583 684	7 125 436	77.77
	31-mers	28 561 768	97 975 734	70.85
	101-mers	120 109 600	387 221 925	68.98

Metric: Recall and H-Precision

VALIDATION USING HAPLOTYPES



Conjecture: *the considered VCF is partially incomplete*

CONCLUSIONS



<https://github.com/Parsoa/PingPong>



Sample-specific strings and Ping-Pong search (*FMD-Index based*)

Take-home messages:

- Specific strings effectively cover variations (*even better than specific kmers*)
- They may replace kmers in some scenarios - thanks to their variable-length nature

CONCLUSIONS



<https://github.com/Parsoa/PingPong>



Sample-specific strings and Ping-Pong search (*FMD-Index based*)

Take-home messages:

- Specific strings effectively cover variations (*even better than specific kmers*)
- They may replace kmers in some scenarios - thanks to their variable-length nature

Open questions:

- Is our notion of sample-specific strings related to the one of SMEMs?
- Can we make the indexing faster?

CONCLUSIONS



<https://github.com/Parsoa/PingPong>



Sample-specific strings and Ping-Pong search (*FMD-Index based*)

Take-home messages:

- Specific strings effectively cover variations (*even better than specific kmers*)
- They may replace kmers in some scenarios - thanks to their variable-length nature

Open questions:

- Is our notion of sample-specific strings related to the one of SMEMs?
- Can we make the indexing faster?

Future steps: variation discovery via sample-specific strings detection



THANK YOU!

Questions?

PSEUDOCODE

Algorithm 1: Computing t -specific strings from FMD-index I_R

```
1 Function PingPongSearch ( $t, I_R$ )
2    $b \leftarrow |t| - 1$ 
3    $[i, j, l] \leftarrow \text{init}(I_R, t[b])$            // init function initializes a
   FMD-Index bi-interval representing a single character
4   while  $b \geq 0$  do
5     while  $l \neq 0 \wedge b > 0$  do           // Step 1 - Backward extension
6        $b \leftarrow b - 1$ 
7        $[i, j, l] \leftarrow \text{backwardExtension}(I_R, [i, j, l], t[b])$ 
8     if  $l \neq 0 \wedge b = 0$  then return
9      $e \leftarrow b$ 
10     $[i, j, l] \leftarrow \text{init}(I_R, t[e])$ 
11    while  $l \neq 0$  do                       // Step 2 - Forward extension
12       $[i_b, j_b, l_b] \leftarrow [i, j, l]$ 
13       $e \leftarrow e + 1$ 
14       $[i, j, l] \leftarrow \text{forwardExtension}(I_R, [i, j, l], t[e])$ 
15    Output  $t[b : e]$ 
16     $[i, j, l] \leftarrow [i_b, j_b, l_b]$ 
```

FULL RESULTS

Metric	Method	Missed	Total	Hits (%)
H-precision	Alg. 1 (exact)	9 093 407	34 219 149	73.43
	Alg. 1 (relaxed)	1 583 684	7 125 436	77.77
	31-mers	28 561 768	97 975 734	70.85
	101-mers	120 109 600	387 221 925	68.98
	31-tigs	2 764 640	5 839 695	52.66
	101-tigs	2 167 395	5 281 605	58.96
C-precision	Alg. 1 (exact)	7 024 433	34 219 149	79.47
	Alg. 1 (relaxed)	669 324	7 125 436	90.61
	31-mers	23 170 031	97 975 734	76.35
	101-mers	84 211 940	387 221 925	78.25
	31-tigs	2 563 021	5 839 695	56.11
	101-tigs	-	-	-

Metric	Method	Missed	Total	Hits (%)
SNPs	Alg. 1 (exact)	39 354	3 147 410	98.75
	Alg. 1 (relaxed)	112 940		96.41
	31-mers	243 363		92.27
	101-mers	46 143		98.53
	31-tigs	267 078		91.51
	101-tigs	55 627		98.23
indels	Alg. 1 (exact)	31 426	716 226	95.61
	Alg. 1 (relaxed)	120 313		83.20
	31-mers	131 591		81.63
	101-mers	32 944		95.40
	31-tigs	175 892		75.44
	101-tigs	31 705		95.57
SVs	Alg. 1 (exact)	1 521	20 775	92.68
	Alg. 1 (relaxed)	2 948		85.81
	31-mers	4 912		76.36
	101-mers	1 978		90.48
	31-tigs	6 383		69.27
	101-tigs	2 698		87.01
All	Alg. 1 (exact)	72 301	3 884 411	98.14
	Alg. 1 (relaxed)	236 201		93.92
	31-mers	379 866		90.22
	101-mers	81 065		97.91
	31-tigs	449 353		88.43
	101-tigs	90 030		97.68