

# HammerIT: homopolymer-space Hamming clustering for IonTorrent read error correction

Anton Korobeynikov<sup>1,2</sup>, Artem Tarasov<sup>1</sup>

<sup>1</sup> Department of Statistical Modelling, St. Petersburg State University, St. Petersburg, Russia; <sup>2</sup> Algorithmic Biology Laboratory, St. Petersburg Academic University, St. Petersburg, Russia

## Introduction

Error correction of sequenced reads remains a difficult task, especially for data obtained using IonTorrent technology due to its higher error rate. The task is even more challenging in single-cell sequencing projects with extremely non-uniform coverage.

The existing error correction tools assume that the most sequencing errors in the data are mismatches and thus perform poorly on IonTorrent data with its prevailing errors due to homopolymer indels.

We introduce HAMMERIT — a novel error correction tool which is specifically tuned for IonTorrent sequencing errors.

## IonTorrent error profile

Corrected flow signal intensities are available in BAM files produced by versions of Ion Torrent Suite prior to 3.4. Called homopolymer length is obtained as corrected flow signal intensity rounded to the nearest integer.

We have studied flow signal intensity distributions around insertion/deletion sites. File B7-295.bam, downloaded from Ion Community website, contained 4.6M insertions, 5.0M deletions, and 1.5M mismatches.

Overwhelming majority of errors turned out to be insertions/deletions of length 1, occurring when flow signal intensity is approximately halfway between two adjacent integers.

Figure 1: Flow signal intensities at insertion sites

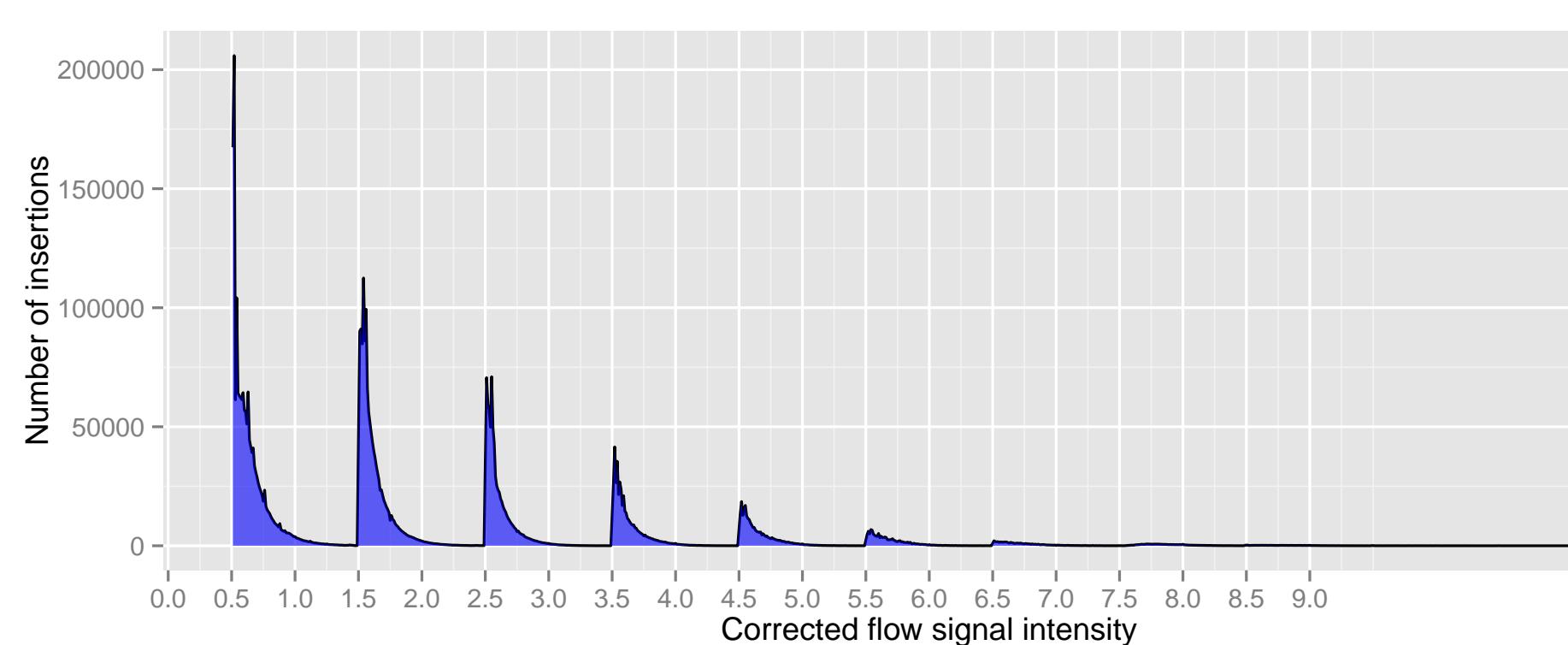
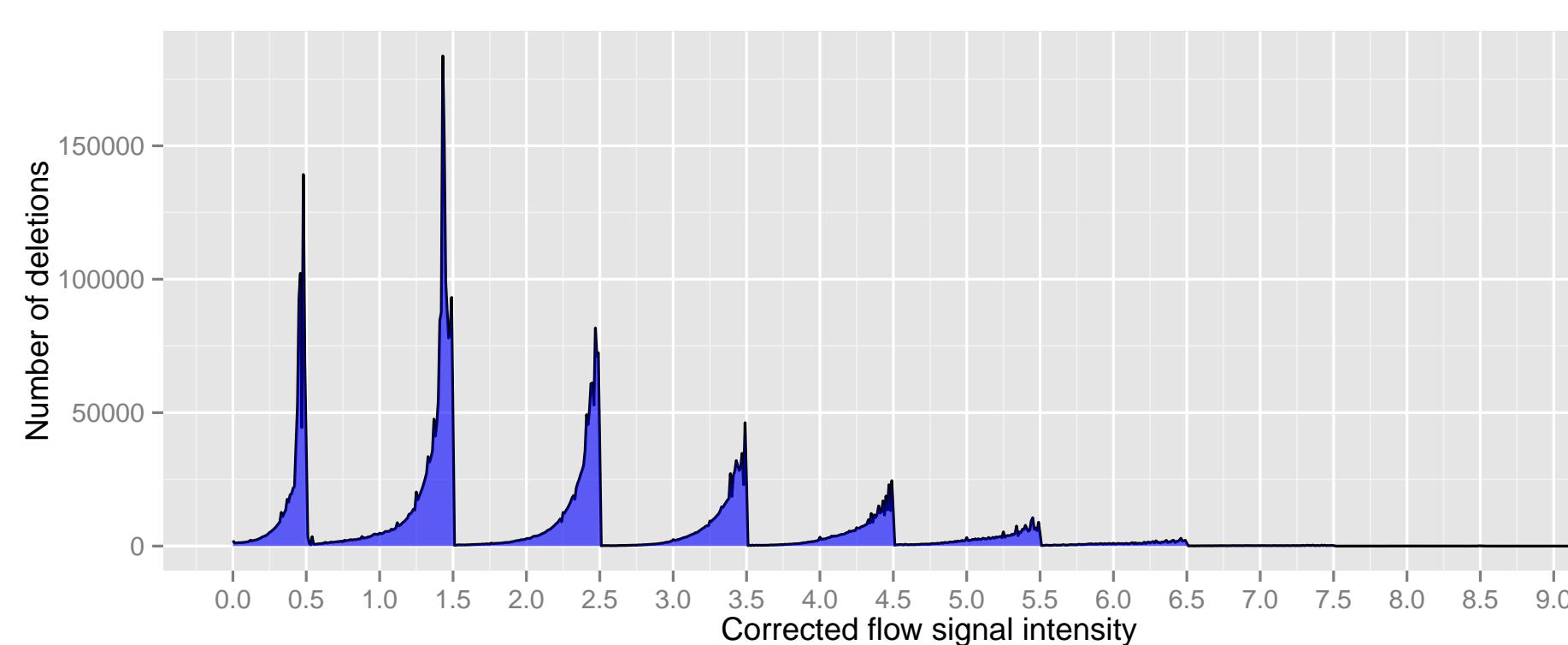


Figure 2: Flow signal intensities at deletion sites



## Pairwise distance calculation

We use 5-base lookahead to compute distance between  $k$ -mers in homopolymer-space. Helper table stores precomputed values of

$$H_k : \mathcal{N}^k \times \mathcal{N}^k \rightarrow \{\text{Insertion, Deletion, Mismatch}\}, k = 1, 2, 3, 4, 5.$$

The chosen value of 5 is a trade-off between accuracy and speed.

(Mapped reads from B7-295.bam dataset were used for training.)

G A G T A C A C T G T C G T C G  
G T G T A C A T G T C G A T G C

$H_5(\text{AGTAC}, \text{TGTAC}) = \text{Mismatch}$   
 $H_5(\text{CTGTC}, \text{TGTCA}) = \text{Deletion}$   
 $H_3(\text{TCG}, \text{ATG}) = \text{Mismatch}$   
 $H_2(\text{CG}, \text{TG}) = \text{Mismatch}$

## Algorithm

**Input:**  $x, y \in \mathcal{N}^+$  — homopolymer-space  $k$ -mers in nucleotide alphabet

**Output:**  $\text{dist}$  — approximate distance between  $x$  and  $y$

$\text{pos}.x \leftarrow 0; \text{pos}.y \leftarrow 0; \text{dist} \leftarrow 0;$

**while**  $\text{pos}.x < |x|$  and  $\text{pos}.y < |y|$  **do**

**if**  $x[\text{pos}.x] = y[\text{pos}.y]$  **then**

$\text{pos}.x \leftarrow \text{pos}.x + 1; \text{pos}.y \leftarrow \text{pos}.y + 1;$

**else**

$k \leftarrow \min(5, |x| - \text{pos}.x, |y| - \text{pos}.y);$

        adjust  $\text{pos}.x$  and  $\text{pos}.y$  according to

$$H_k(x[\text{pos}.x .. \text{pos}.x + k - 1], y[\text{pos}.y .. \text{pos}.y + k - 1]);$$

$\text{dist} \leftarrow \text{dist} + 1;$

## Notation

Let  $\mathcal{N}$  denote the nucleotide alphabet {A, C, G, T}.

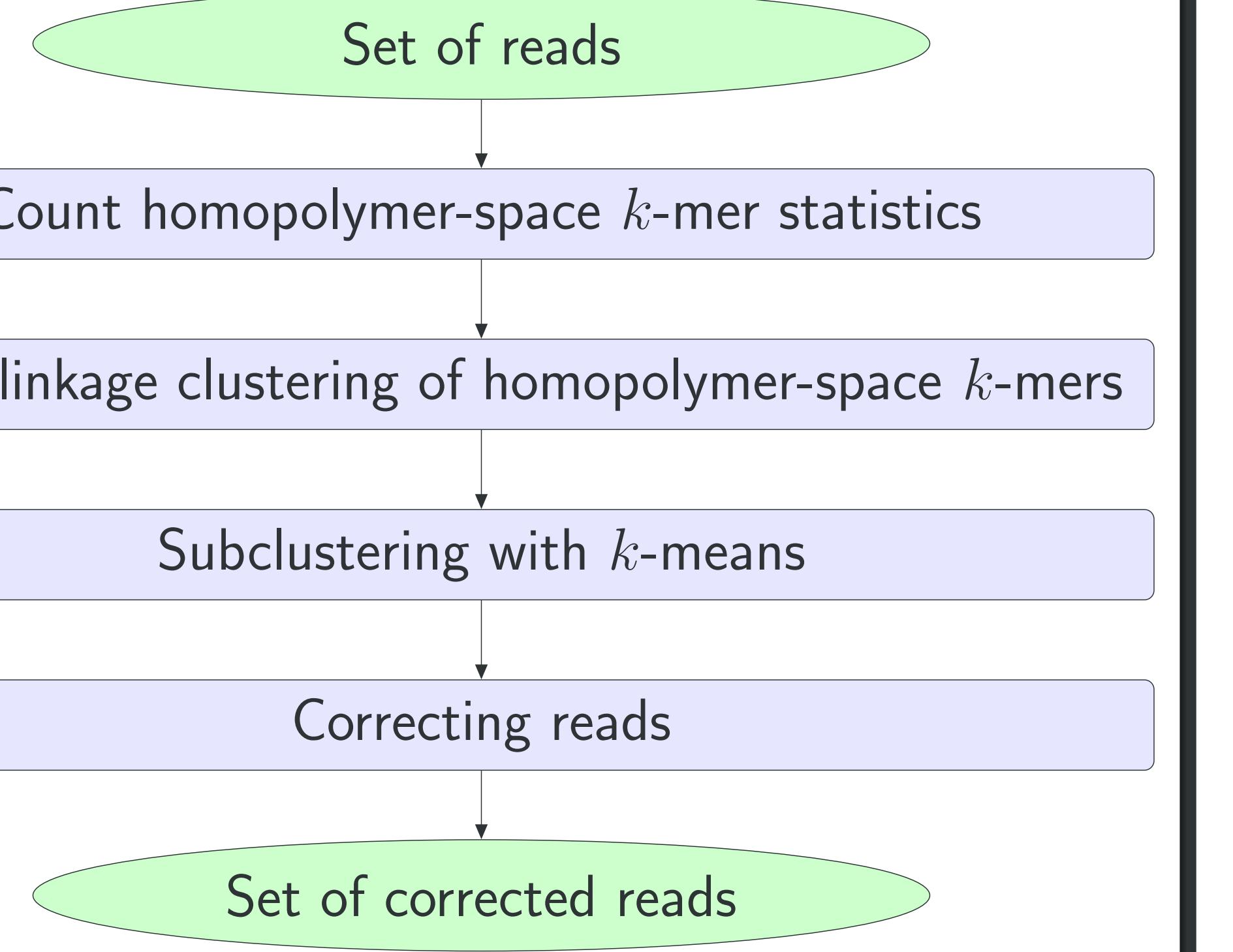
By definition, put  $\mathcal{H} = \mathcal{N} \times \mathcal{N}^+$ .

We call an element of the alphabet  $\mathcal{H}$  a *homopolymer run*, and an element of  $\mathcal{H}^k$  a *homopolymer-space  $k$ -mer*.

We use  $x[k]$  to denote  $k$ -th element of a sequence, using zero-based indexing;  $x[k .. l]$  to denote a subsequence  $x[k]x[k + 1] \dots x[l]$ ; finally, length of  $x$  is denoted by  $|x|$ .

Distance between  $i, j \in \mathcal{H}^k$  is defined as the minimum number of 1-base insertions/deletions/mismatches needed to align common part of  $i$  and  $j$ . It is denoted by  $\text{dist}(i, j)$ .

## HammerIT workflow



## Homopolymer-space $k$ -mer clustering

The starting point is single-linkage clustering of homopolymer-space  $k$ -mers. Two homopolymer-space  $k$ -mers belong to the same cluster if the distance between their nucleotide representations does not exceed one.

We reduce quadratic time requirements of the naive algorithm by noticing that if distance between two  $k$ -mers is less or equal to one, they share a common substring of length at least  $\lfloor k/2 \rfloor$ . This allows us to group  $k$ -mers into smaller blocks sharing a substring, and then use the quadratic algorithm for each block. In order to detect all pairs of connected  $k$ -mers with such grouping, ranges  $(0 .. \lfloor k/2 \rfloor - 1), (1 .. \lfloor k/2 \rfloor), \dots, (\lfloor (k+1)/2 \rfloor .. k - 1)$  are to be examined. Despite of this, we use only the first and the last of the ranges, to speed up the clustering step. This increases the number of *singletons* — clusters consisting of only one  $k$ -mer, usually erroneous — but has little impact on correction performance because of adjacent read  $k$ -mers making much larger contribution into the consensus scores (the number of singleton occurrences is usually low).

$$\text{partition}(\mathcal{K}, i, j) = \{B_s : \bigcup_{s \in \mathcal{H}^{j-i+1}} B_s = \mathcal{K}, \forall x \in B_s x[i .. j] = s\}$$

**Algorithm**

$\text{kmers} \leftarrow \text{homopolymer-space } k\text{-mers seen in the data}$

$\text{components} \leftarrow \{k\} : k \in \text{kmers}$

$\text{blocksL} \leftarrow \text{partition}(\text{kmers}, 0, \lfloor k/2 \rfloor - 1)$

$\text{blocksR} \leftarrow \text{partition}(\text{kmers}, \lfloor (k+1)/2 \rfloor, k - 1)$

**for each**  $\text{block}$  **in**  $\text{blocksL}$ ,  $\text{blocksR}$  **do**

**for each**  $i \in \text{block}$  **do**

**for each**  $j \in \text{block}$  **do**

**if**  $\text{dist}(i, j) \leq 1$  **then**

                    join the components to which  $i$  and  $j$  belong

## Error reduction results

We evaluated HammerIT on 6 publicly available datasets, using the same pipeline as the authors of the recently published article (Jünemann et al, Nat. Biotech., 2013; vol. 31, p.294–296). In that article, error rate in four Ion Torrent datasets has been assessed. We used the same data plus two extra datasets from 314v2 chip, which recently became available on Ion Community Portal.

Indel/mismatch error rates were calculated for uniquely mapped reads before and after correction. For each dataset, correction was done in two ways. In the first setup, trimming was done for read ends that couldn't be corrected due to lack of good  $k$ -mers, while in the second one such read ends were preserved in the output. Relative change in read coverage after correction stayed within 0.4% in all cases.

Figure 3: Error rates before and after correction

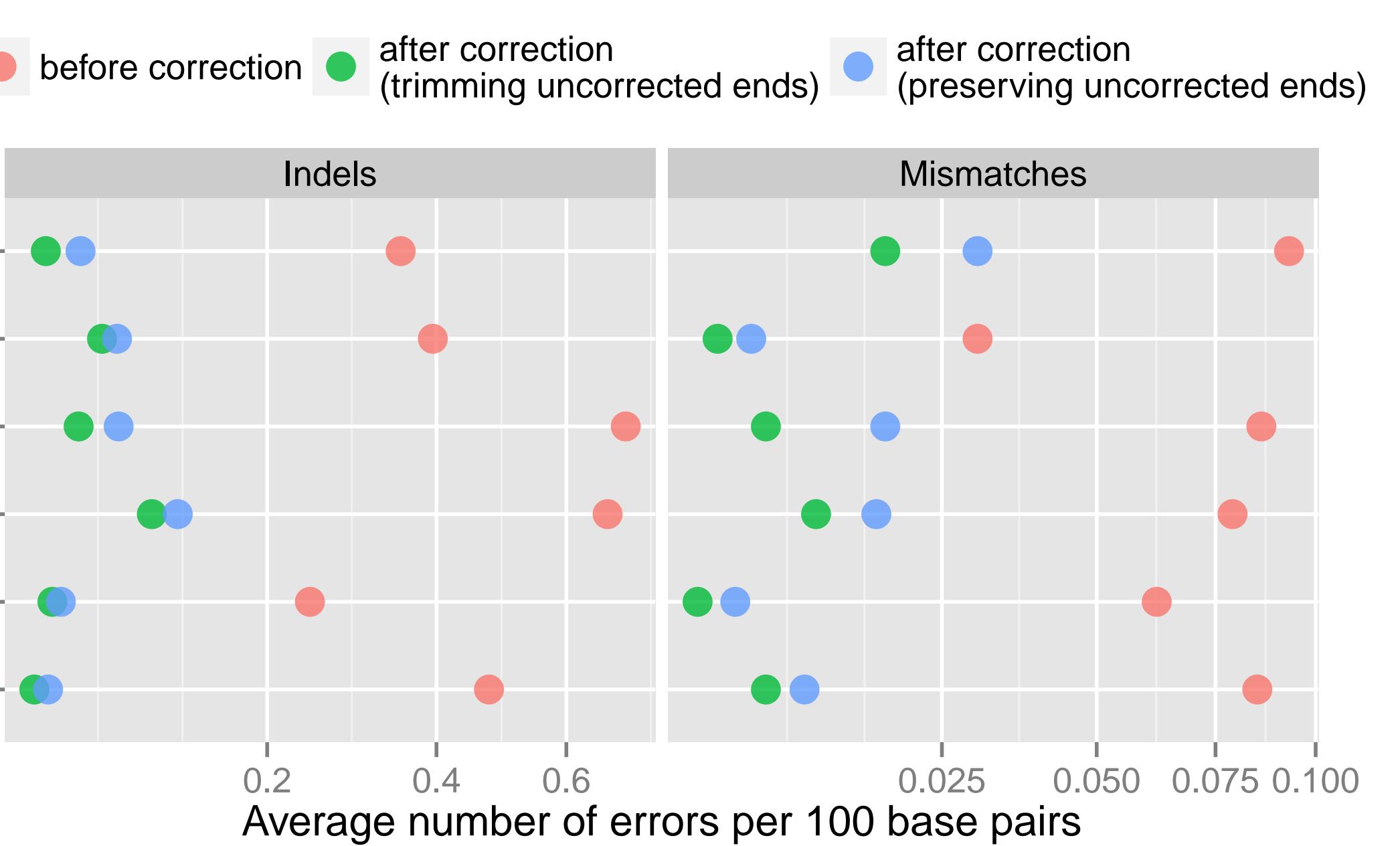
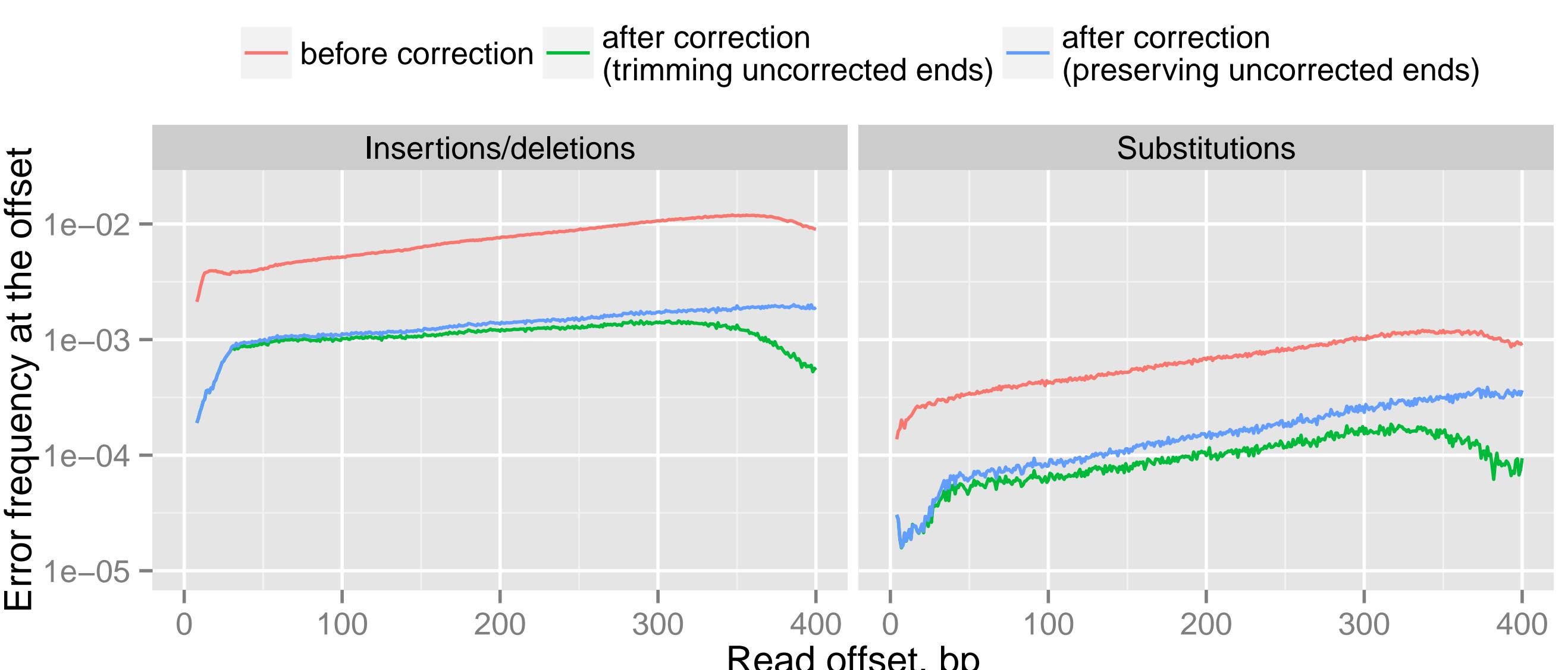


Figure 4: Error reduction by read position for Sakai 400bp reads



## Assembly results

We have assembled *E.coli O157 H7 Sakai* 400bp reads before and after correction with SPAdes 2.4.0. Read coverage is ~200x, length of the reference genome is 5498450 bp.

Table 1: Assembly results. Contigs of length  $\geq 500$  are used.

|                             | uncorrected | corrected |
|-----------------------------|-------------|-----------|
| # contigs                   | 266         | 242       |
| Largest contig              | 316904      | 374932    |
| Total length                | 5322223     | 5320566   |
| NG50                        | 106242      | 146551    |
| NG75                        | 41186       | 44318     |
| # misassemblies             | 0           | 2         |
| # local misassemblies       | 8           | 11        |
| Misassembled contigs length | 0           | 32563     |
| Genome fraction (%)         | 93.880      | 93.966    |
| # mismatches per 100 kbp    | 3.58        | 4.70      |
| # indels per 100 kbp        | 6.21        | 5.88      |

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## Error correction algorithm

```

for each read from the dataset do
    (Producing contiguous corrected parts of read)
    for each homopolymer-space kmer from the read do
        center ← center of the cluster to which kmer belongs;
        if center quality is more than user-specified threshold and
            center bases agree with the previous "good" center then
                include center into consensus score calculation;
        else
            yield new corrected part from current consensus;
            trim homopolymer runs with low consensus score from ends;
            reset consensus table and start a new part;
    (Combining corrected parts)
    while there are two or more parts do
        curr ← first part; next ← second part;
        align last 8 homopolymer runs of curr against the read;
        align first 8 homopolymer runs of next against the read;
        if there is a gap on the read between the two parts then
            copy read homopolymer runs as is;
        else
            select homopolymer runs with higher consensus score
            from the intersection of the two parts;
            replace curr and next with the combined part;
    (Optionally, attaching uncorrected end)
    align last 8 runs of the last chunk against the read sequence;
    append read homopolymer runs after the last aligned run.

```