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Christian Baudet, Ulisses Dias, Zanoni Dias. Sorting by weighted inversions considering length and symmetry. BMC Bioinformatics, 2015, 16 (Suppl 19), pp.11. 10.1186/1471-2105-16-S19-S3 . hal-01316998

**HAL Id: hal-01316998**

**<https://inria.hal.science/hal-01316998>**

Submitted on 17 May 2016

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RESEARCH

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# Sorting by weighted inversions considering length and symmetry

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From Brazilian Symposium on Bioinformatics 2014  
Belo Horizonte, Brazil. 28-30 October 2015

## Abstract

Large-scale mutational events that occur when stretches of DNA sequence move throughout genomes are called genome rearrangements. In bacteria, inversions are one of the most frequently observed rearrangements. In some bacterial families, inversions are biased in favor of symmetry as shown by recent research. In addition, several results suggest that short segment inversions are more frequent in the evolution of microbial genomes. Despite the fact that symmetry and length of the reversed segments seem very important, they have not been considered together in any problem in the genome rearrangement field. Here, we define the problem of sorting genomes (or permutations) using inversions whose costs are assigned based on their lengths and asymmetries. We consider two formulations of the same problem depending on whether we know the orientation of the genes. Several procedures are presented and we assess these procedure performances on a large set of more than  $4.4 \times 10^9$  permutations. The ideas presented in this paper provide insights to solve the problem and set the stage for a proper theoretical analysis.

## Background

Among various large-scale rearrangement events that have been proposed to date, inversions were established as the main explanation for the genomic divergence in many organisms [1-3]. An inversion occurs when a chromosome breaks at two locations, and the DNA between those locations is reversed.

In some families of bacteria, an 'X'-pattern is observed when two circular chromosomes are aligned [2,4]. Inversions symmetric to the origin of replication (meaning that the breakpoints are equally distant from the origin of replication) have been proposed as the primary mechanism that explains the pattern [4]. The justification relies on the fact that one single highly asymmetric inversion affecting a large area of the genome could destroy the 'X'-pattern, although short inversions may still preserve it.

Darling, Miklós and Ragan [1] studied eight *Yersinia* genomes and added evidence that symmetric inversions are "over-represented" with respect to other types of

inversions. They also found that inversions are shorter than expected under a neutral model. In many cases, short inversions affect only a single gene, as observed by Lefebvre *et al.* [5] and Sankoff *et al.* [6], which contrasts with the null hypothesis that the two endpoints of an inversion occur by random and independently.

Despite the importance of symmetry and length of the reversed segment, both have been somewhat overlooked in the genome rearrangement field. Indeed, the most important result regarding inversions is a polynomial time algorithm presented by Hannenhalli and Pevzner [3] that considers an unit cost for each inversion no matter its length or symmetry. When gene orientation is not taken into account, finding the minimum number of inversions that transform one genome into the other is a NP-Hard problem [7].

Some results have considered at least one of the concepts. There is a research line that considers the total sum of the inversion lengths as the objective function of a minimization problem. Several results have been presented both when gene orientation is considered [8,9] and when it is not [10-12].

Regarding symmetry, the first results were presented by Ohlebusch *et al.* [13]. Their algorithm uses symmetric

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inversions in a restricted setting to compute an ancestral genome and, therefore, is not a generic algorithm to compute the rearrangement distance using only symmetric inversions. In 2012, Dias *et al.* presented an algorithm that considers only symmetric and almost-symmetric inversions [14]. They later included unitary inversions to the problem and provided a randomized heuristic to compute scenarios between two genomes that uses solely these operations [15].

Here we propose a new genome rearrangement problem that combines the concepts of symmetry and length of the reversed segments. Whereas previous works restricted the set of allowed operations by considering only inversions that satisfy constraints like symmetry or almost-symmetry [14,15], here we allow all possible inversions.

The problem we are proposing aims at finding low-cost scenarios between genomes when gene orientation is taken into account and when it is not, which is useful, among others, for building phylogenetic trees, annotating genomes or correcting already existing annotations. The results obtained are the first steps in exploring this interesting new problem.

## Definitions

Formally, a chromosome is represented as a  $n$ -tuple whose elements represent genes. If we assume no gene duplication, then this  $n$ -tuple is a permutation  $\pi = (\pi_1 \pi_2 \dots \pi_n)$ ,  $1 \leq |\pi_i| \leq n$  and  $|\pi_i| \leftrightarrow |\pi_j| \leftrightarrow i \neq j$ . Because we focus on bacterial chromosomes, we assume permutations to be circular, and  $\pi_1$  is the first gene after the origin of replication.

We consider two cases depending on whether we know the orientation of the genes. If we know the orientation, then  $\pi$  is a signed permutation such that each element  $\pi_i \in \{-n, -(n-1), \dots, -1, +1, +2, \dots, +n\}$ . If we do not know the orientation of the genes, then  $\pi$  is an unsigned permutation such that  $\pi_i \in \{1, 2, \dots, n\}$ .

We treat permutations as functions such that  $\pi(i) = \pi_i$  and  $\pi(-i) = -\pi(i)$ . The inverse of a permutation  $\pi$  is denoted by  $\pi^{-1}$ , for which  $\pi_{\pi_i}^{-1} = i$  for all  $1 \leq i \leq n$ . The composition between two permutations  $\pi$  and  $\sigma$  is similar to function composition in such way that  $\pi \cdot \sigma = (\pi_{\sigma(1)} \pi_{\sigma(2)} \dots \pi_{\sigma(n)})$ .

Let  $\pi = (\pi_1 \pi_2 \dots \pi_n)$  be a signed permutation, an inversion  $\rho(i, j)$ ,  $1 \leq i \leq j \leq n$  is an operation such that  $\pi \cdot \rho(i, j) = (\pi_1 \dots \pi_{i-1} - \pi_j - \pi_{j-1} \dots - \pi_{i+1} \dots - \pi_i \pi_{j+1} \dots \pi_n)$ . Let  $\pi = (\pi_1 \pi_2 \dots \pi_n)$  be an unsigned permutation, an inversion  $\rho(i, j)$ ,  $1 \leq i < j \leq n$  is an operation such that  $\pi \cdot \rho(i, j) = (\pi_1 \dots \pi_{i-1} \pi_j \pi_{j-1} \dots \pi_{i+1} \pi_i \pi_{j+1} \dots \pi_n)$ .

Given two permutations  $\alpha$  and  $\sigma$ , we are interested in finding rearrangement scenarios that link  $\alpha$  to  $\sigma$ . Therefore, our scenarios are sequences of operations  $\rho_1, \rho_2, \dots, \rho_t$  such that  $\alpha \cdot \rho_1 \cdot \rho_2 \dots \rho_t = \sigma$ .

Let  $\iota = (1 \ 2 \ \dots \ n)$  be the identity permutation, sorting a permutation  $\pi = (\pi_1 \ \pi_2 \ \dots \ \pi_n)$  is the process of transforming  $\pi$  into  $\iota$ . Note that  $\sigma \cdot \sigma^{-1} = \sigma^{-1} \cdot \sigma = \iota_n$ . Thus, the scenario that transforms  $\alpha$  into  $\sigma$  can be used to transform a permutation  $\pi$  into a permutation  $\iota$  if we take  $\pi = \sigma^{-1} \cdot \alpha$ . Therefore, we hereafter consider sorting permutations by inversions.

Previous researchers have worked on the inversion distance  $d(\pi)$  of an arbitrary permutation  $\pi$ , which is the minimum number of inversions that transform  $\pi$  into  $\iota$ . Here we consider that each inversion  $\rho(i, j)$  has a cost which is based on the length and the symmetry of endpoints  $i$  and  $j$ .

The following functions help us to define our cost function and can be applied to identify any element  $i$  in the permutation  $\pi$ . **Position:**  $p(\pi, i) = k \Leftrightarrow |\pi_k| = i$ ,  $p(\pi, i) \in \{1, 2, \dots, n\}$ . **Sign:**  $s(\pi, i) = k \Leftrightarrow \pi_p(\pi, i) = k \times i$ ,  $s(\pi, i) \in \{-1, +1\}$ . **Slice:**  $slice(\pi, i) = \min\{p(\pi, i), n - p(\pi, i) + 1\}$ ,  $slice(\pi, i) \in \{1, 2, \dots, \lceil \frac{n}{2} \rceil\}$ .

Figure 1(b) shows the values of these functions for the signed permutation  $\pi = (-5 \ +3 \ +4 \ -2 \ +1)$ . Note that the values for the functions **slice** and **position** would not change if instead we had the unsigned permutation  $\pi = (5 \ 3 \ 4 \ 2 \ 1)$ .

Our cost function is:  $cost(\rho(i, j)) = |slice(\iota, i) - slice(\iota, j)| + 1$ . Its behavior is explained by looking at the two cases that arise. Figure 2 illustrates both cases.

**Case 1:**  $i, j \leq \lceil \frac{n}{2} \rceil$  or  $i, j \geq \lceil \frac{n}{2} \rceil$ .

In this case, the cost function can be simplified to  $cost(\rho(i, j)) = abs(i-j)+1$ , which means that it is proportional to the number of elements in the reversed segment. This cost is what one would expect from a length-weighted inversion distance in such a way that larger inversions cost more than short inversions.

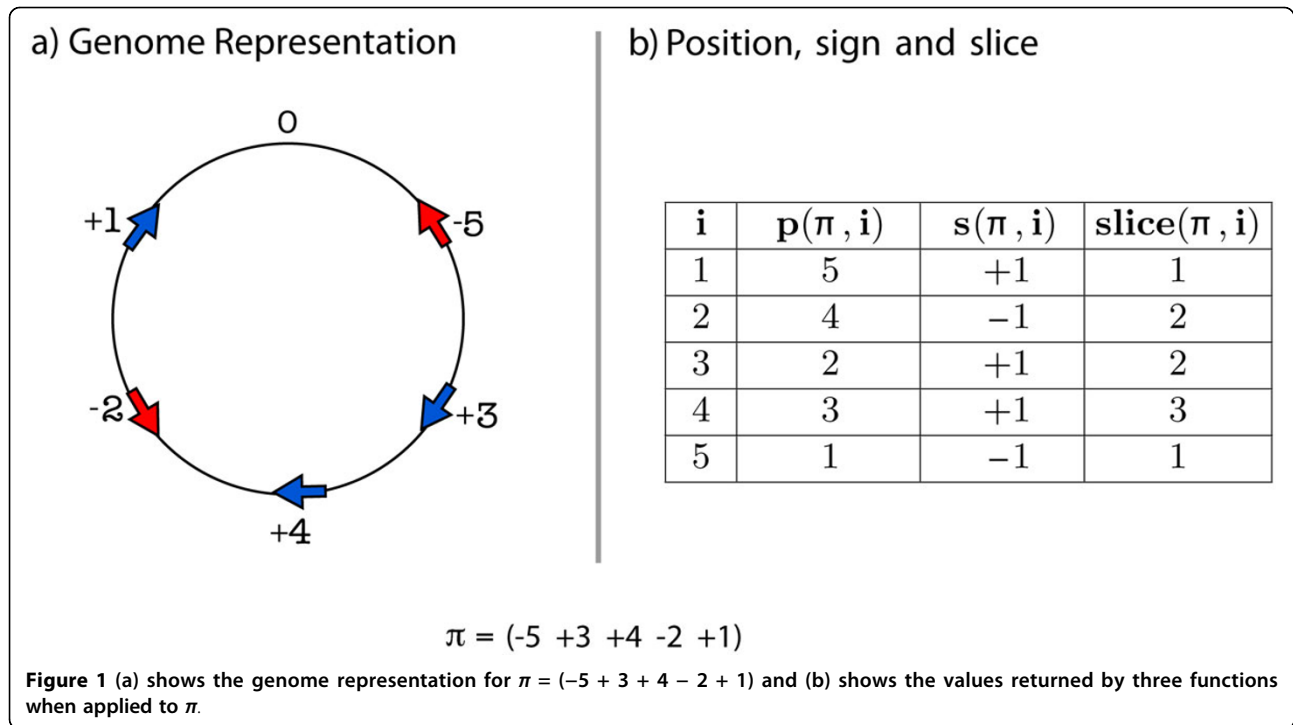
**Case 2:**  $i > \lceil \frac{n}{2} \rceil$  and  $j < \lceil \frac{n}{2} \rceil$ , or  $j > \lceil \frac{n}{2} \rceil$  and  $i < \lceil \frac{n}{2} \rceil$ .

In this case, the cost function is penalizing the asymmetry instead of the number of elements in the reversed segment. In effect, if the inversion  $\rho(i, j)$  is perfectly symmetric (meaning that  $i$  and  $j$  are equally distant from the origin of replication, so  $slice(\iota, i) = slice(\iota, j)$ ), then the cost is given by  $cost(\rho(i, j)) = 1$ .

Therefore, our problem is to find a sequence of operations  $\rho_1, \rho_2, \dots, \rho_t$  such that  $\pi \cdot \rho_1 \cdot \rho_2 \dots \rho_t = \iota$  and  $\sum_{k=1}^t cost(\rho_k)$  is minimum.

## Methods

This section presents several greedy algorithms that take advantage from the characteristics of the cost function. The first greedy approach was named LR and constructs a solution by placing one element each time in the final



position. After we place an element, we guarantee that we will not move it again.

Three other greedy approaches have been established, named NB, SMP and NB+SMP. These approaches rely on greedy functions to estimate how good an inversion might be. For each greedy function  $f$ , the benefit of an inversion  $\rho$  is the difference in the value returned by  $f$  before and after  $\rho$  is applied divided by the cost of  $\rho$ . For instance, let  $\pi$  be an arbitrary permutation, the benefit of  $\rho$  is computed as  $\text{ben}_f(\pi, \rho) = \frac{f(\pi) - f(\pi \cdot \rho)}{\text{cost}(\rho)}$ .

Note that we expect  $f$  to assign smaller values to permutations closer to the identity.

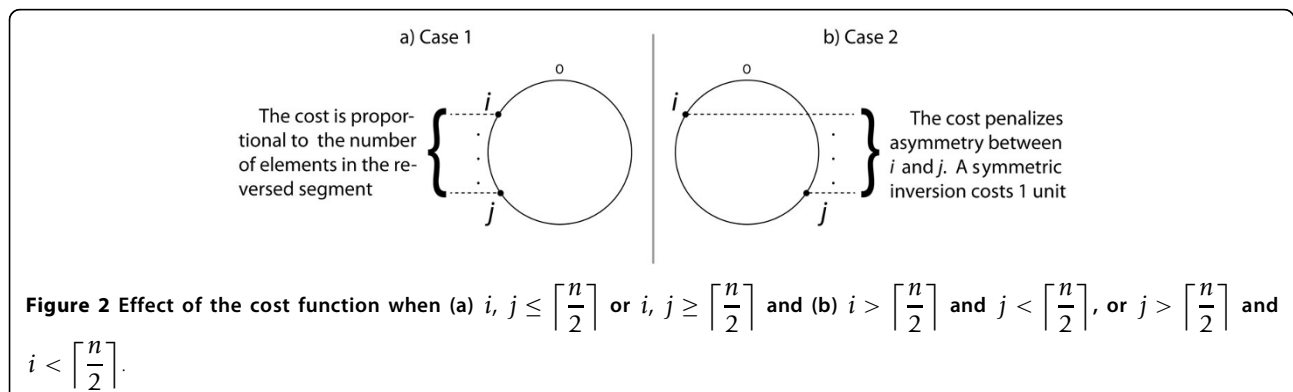
Using the greedy function  $f$ , we construct a sequence of inversions that sort  $\pi$  by iteratively adding an inversion with the best benefit among all possible inversions. The greedy function  $f$  guarantees a (possibly not

optimum) solution if we can always find an inversion  $\rho$  such that  $f(\pi \cdot \rho) < f(\pi)$  for any  $\pi \neq \iota$ . However, our greedy functions presented in the following sections do not always guarantee that. Therefore, we study each case and we developed ways to circumvent this issue.

#### Left or right heuristic

We use the term LR to refer to this approach as an acronym for Left or Right. We first divide the elements in the permutation in two groups. The first group refers to the elements that are in slices classified as sorted and the second group comprises those elements that are in unsorted slices. A slice  $s$  is in the sorted group if  $p(\pi, s) = s$ ,  $p(\pi, n - s + 1) = n - s + 1$ , and the slices  $\{1, 2, \dots, s - 1\}$  are also in the sorted group. Otherwise,  $s$  is in the unsorted group.

First, the *Left or Right* heuristic selects the least slice in the unsorted group. Then, we determine the element



that should be moved first to that slice: the left or the right. The left (right) side is composed of the elements which are in positions that have indices bigger (lower) than the middle position.

To make this choice, we compute the total cost to put a given element in its final place. For signed permutations, we also consider the cost to place the element with a positive sign. After computing a cost for placing the left and the right elements, we choose the side that has the minimum cost. In case of tie, we move the right element.

If the slice has only one element that does not belong to it, we find the element that should be in that slice and perform an inversion to place it in the final position. After placing the element, we might have to change its sign on the signed version of our problem.

### Slice-misplaced pairs heuristic

We use the term SMP to refer to this heuristic as an acronym for Slice-Misplaced Pairs.

**Definition 1** We say that a pair  $\{\pi_i, \pi_j\}$ ,  $1 \leq i < j \leq n$ , is slice-misplaced in  $\pi$  if  $slice(\pi, \pi_i) > slice(\pi, \pi_j)$  and  $slice(\iota, \pi_i) < slice(\iota, \pi_j)$ . We use  $\pi_i \sim \pi_j$  to represent that  $\pi_i$  and  $\pi_j$  are slice-misplaced.

Let  $SMP(\pi)$  be the number of slice-misplaced pairs in  $\pi$  and  $\Delta_{SMP}(\pi, \rho) = SMP(\pi \cdot \rho) - SMP(\pi)$  be the variation in the number of slice-misplaced pairs caused by an inversion  $\rho$ , then the benefit of an inversion is given by

$$ben_{SMP} = -\frac{\Delta_{SMP}(\pi, \rho)}{cost(\rho)}.$$

For some permutations  $\pi \neq \iota$ , there is no inversion  $\rho$  such that  $SMP(\pi \cdot \rho) < SMP(\pi)$ . The following lemmas give us a better understanding from the properties of these permutations. We start by stating in Lemma 1 when we can be sure that at least one slice-misplaced pair can be removed.

**Lemma 1** Let  $\pi_i, \pi_j$  be two elements in  $\pi$  such that  $|i - j| = 1$  and  $\pi_i \sim \pi_j$ . Then there is at least one inversion  $\rho$  such that  $\Delta_{SMP}(\pi, \rho) < 0$ .

*Proof* Let us assume, without loss of generality, that  $slice(\pi, \pi_i) < slice(\pi, \pi_j)$ . Therefore,  $slice(\iota, \pi_i) > slice(\iota, \pi_j)$  since  $\pi_i \sim \pi_j$ . The inversion  $\rho = \rho(i, j)$  if  $i < j$  or  $\rho = \rho(j, i)$  if  $i > j$  will remove the slice-misplaced pair  $\{\pi_i, \pi_j\}$  when creating the permutation  $\sigma = \pi \cdot \rho$ . Let  $\pi_k$  and  $\pi_l$  be the elements in the same slice as  $\pi_i$  and  $\pi_j$  in  $\pi$ , respectively, the following statements suffice to conclude that  $\Delta_{SMP}(\pi, \rho) < 0$ . Note that it may occur that  $\pi_j$  is the only element in the slice, therefore it suffices to prove the first statement.

- If  $\pi_j \neq \pi_k$  in  $\pi$ , then  $\pi_i \neq \pi_k$  in  $\sigma$ . We know that  $slice(\iota, \pi_j) > slice(\iota, \pi_k)$  since in  $\pi$  we have  $\pi_j \neq \pi_k$  and  $slice(\pi, \pi_j) > slice(\pi, \pi_k)$ . Therefore,  $\pi_i \neq \pi_k$  in  $\sigma$  because  $slice(\iota, \pi_k) < slice(\iota, \pi_j) < slice(\iota, \pi_i)$  and  $slice(\sigma, \pi_k) < slice(\sigma, \pi_i)$ .

- If  $\pi_i \neq \pi_l$  in  $\pi$ , then  $\pi_j \neq \pi_l$  in  $\sigma$ . We know that  $slice(\iota, \pi_i) < slice(\iota, \pi_l)$  since in  $\pi$  we have  $\pi_i \neq \pi_l$  and  $slice(\pi, \pi_i) < slice(\pi, \pi_l)$ . Therefore,  $\pi_j \neq \pi_l$  in  $\sigma$  because  $slice(\iota, \pi_l) > slice(\iota, \pi_i) > slice(\iota, \pi_j)$  and  $slice(\sigma, \pi_l) > slice(\sigma, \pi_j)$ .

Observe that we do not need to consider cases where  $\pi_i \sim \pi_l$  or  $\pi_j \sim \pi_k$  in  $\pi$ , because in the worst scenario these slice-misplaced pairs will not be removed.

**Lemma 2** Let  $\pi_i$  be an element in  $\pi$  such that  $slice(\iota, \pi_i) = \left\lceil \frac{n}{2} \right\rceil$ . If  $slice(\pi, \pi_i) \neq \left\lceil \frac{n}{2} \right\rceil$ , then there is at least one inversion  $\rho$  such that  $\Delta_{SMP}(\pi, \rho) < 0$ .  $\square$

*Proof* If  $\pi_i$  is the only element having  $slice(\iota, \pi_i) = \left\lceil \frac{n}{2} \right\rceil$  or if it occurs that  $slice(\pi, \pi_i) \geq slice(\pi, \pi_j)$  for  $\pi_i, \pi_j$  such that  $slice(\iota, \pi_i) = slice(\iota, \pi_j) = \left\lceil \frac{n}{2} \right\rceil$ , then  $\pi_i$  will form a slice-misplaced pair with all possible element  $\pi_k$  such that  $slice(\pi, \pi_k) > slice(\pi, \pi_i)$ . Therefore, it is straightforward from Lemma 1 that at least one inversion  $\rho$  such that  $\Delta_{SMP}(\pi, \rho) < 0$  exists as long as  $slice(\pi, \pi_i) \neq \left\lceil \frac{n}{2} \right\rceil$ .

If  $\pi_i = \left\lceil \frac{n}{2} \right\rceil$  and  $|i - j| \neq 1$ , it is straightforward from Lemma 1 that we can move  $\pi_j$  toward the slice. When  $|i - j| = 1$  we apply the inversion  $\rho(i, k)$  if  $i < k$  or  $\rho(k, i)$  if  $k < i$ , where  $k = n - j + 1$ .  $\square$

**Lemma 3** Let  $\pi = (\pi_1 \pi_2 \dots \pi_n)$  be a permutation such that Lemmas 1 and 2 find no inversion that decreases the number of slice-misplaced pairs, then  $\pi$  has the form:

$$\text{For } n \text{ odd} \begin{cases} slice(i, \pi_1) \leq slice(i, \pi_2) \leq \dots \leq slice\left(i, \frac{\pi_{n-1}}{2}\right) < slice\left(i, \frac{\pi_{n+1}}{2}\right) \\ slice\left(i, \frac{\pi_{n+1}}{2}\right) > slice\left(i, \frac{\pi_{n+1}}{2+1}\right) \geq \dots \geq slice(i, \pi_{n-1}) \geq slice(i, \pi_n) \end{cases}$$

$$\text{For } n \text{ even} \begin{cases} slice(i, \pi_1) \leq slice(i, \pi_2) \leq \dots \leq slice\left(i, \frac{\pi_n}{2-1}\right) < slice\left(i, \frac{\pi_n}{2}\right) \\ slice\left(i, \frac{\pi_n}{2}\right) = slice\left(i, \frac{\pi_{n+1}}{2+1}\right) \\ slice\left(i, \frac{\pi_n}{2+1}\right) > slice\left(i, \frac{\pi_n}{2+2}\right) \geq \dots \geq slice(i, \pi_{n-1}) \geq slice(i, \pi_n) \end{cases}$$

*Proof* Lemma 2 implies that if  $n$  is odd, then  $\frac{\pi_{n+1}}{2}$  is in the highest slice in  $\iota$ . The same occurs with the elements  $\frac{\pi_n}{2}$  and  $\frac{\pi_n}{2+1}$  that should be in the highest slice in  $\iota$  if  $n$  is even. We know that  $\pi_i \neq \pi_{i+1}$ , otherwise one could find  $\rho$  such that  $\Delta_{SMP}(\pi, \rho) < 0$  using the Lemma 1. That leads to the elements being ordered according to their slices in  $\iota$  as shown.  $\square$

**Lemma 4**  $\Delta_{SMP}(\pi, \rho) = 0$  for any perfectly symmetric inversion  $\rho(i, j)$  such that  $j = n - i + 1$ .

*Proof* Inversions  $\rho(i, j)$  such that  $j = n - i + 1$  have no effect in the slice of any element in  $\pi$ . That said, no slice-misplaced pair will be created or removed whatsoever.

**Lemma 5** Let  $\pi$  be a permutation such that  $SMP(\pi) = 0$ , then  $slice(\pi, \pi_i) = slice(\iota, \pi_i)$  for any  $1 \leq i \leq n$ . In this case, perfectly symmetric inversions can sort  $\pi$  if it is unsigned and perfectly symmetric plus unitary inversions should be enough to sort  $\pi$  if it is signed.

*Proof* Let  $\pi_i$  be an element in  $\pi$  such that  $slice(\pi, \pi_i) \neq slice(\iota, \pi_i)$ . There must exist an element  $\pi_j$  such that  $slice(\pi, \pi_j) = slice(\iota, \pi_i) \neq slice(\iota, \pi_j)$ . Two cases are possible: (i)  $\pi_i \sim \pi_j$ , therefore  $SMP(\pi) > 0$  or (ii)  $\pi_i \nmid \pi_j$ , thus there must exist an element  $\pi_k$  such that  $slice(\pi, \pi_k) = slice(\iota, \pi_i) \neq slice(\iota, \pi_k)$ . In this second case, we can restart the entire process by calling  $\pi_j$  and  $\pi_k$  as  $\pi_i$  and  $\pi_j$ , respectively. Since we have a finite number of elements in the permutation, we know that the first case will be reached eventually.

If  $slice(\pi, \pi_i) = slice(\iota, \pi_i)$  for any  $1 \leq i \leq n$ , we can reach the identity permutation by first performing perfectly symmetric inversions  $\rho(i, j)$ ,  $slice(\iota, i) = slice(\iota, j)$ , if  $|\pi_i| = j$  and  $|\pi_j| = i$ . It requires at most  $\left\lceil \frac{n}{2} \right\rceil$  inversions and each one will cost one unit. It should be enough for unsigned permutations, but for signed permutations we still need to perform unitary inversions  $\rho(i, i)$  if  $s(\pi_i) = -1$ .

**Lemma 6** Let  $\pi = (\pi_1 \pi_2 \dots \pi_n)$  be a permutation such that Lemmas 1 and 2 find no inversion that decreases the number of slice-misplaced pairs and  $SMP(\pi) > 0$ , then we can apply two inversions  $\rho_1$  and  $\rho_2$  such that  $SMP(\pi) > SMP(\pi \cdot \rho_1 \rho_2)$ .

*Proof* Assuming no inversion as described by Lemmas 1 and 2 is possible, we know that  $\pi$  has the form of Lemma 3. Moreover, there is at least one pair of elements  $\pi_i, \pi_{i+1}$  such that  $slice(\iota, \pi_i) = slice(\iota, \pi_{i+1})$  and  $slice(\iota, \pi_i) \neq \left\lceil \frac{n}{2} \right\rceil$ . We hereafter consider, without loss of generality, that (i)  $slice(\pi, \pi_i) = i$  and  $slice(\pi, \pi_{i+1}) = i+1$ ; (ii)  $slice(\pi, \pi_{i+1}) \geq slice(\pi, \pi_x)$  and  $slice(\pi, \pi_{i+1}) \geq slice(\pi, \pi_{x+1})$  for every pair  $\pi_x, \pi_{x+1}$  such that  $slice(\iota, \pi_x) = slice(\iota, \pi_{x+1})$ ; (iii)  $\pi_j$  and  $\pi_{j+1}$  are elements that share the same slice with  $\pi_{i+1}$  and  $\pi_i$ , respectively.

The first inversion we apply is  $\rho_1(i+1, j)$ . This inversion is perfectly symmetric and hence  $\Delta_{SMP}(\pi, \rho_1) = 0$  by Lemma 4. We assert that the inversion  $\rho_2$  applied after  $\rho_1$  will have  $\Delta_{SMP}(\pi \cdot \rho_1, \rho_2) < 0$  in order to prove the lemma. That said, we start by compiling attributes of  $\sigma = \pi \cdot \rho_1$  and we will use these attributes to find  $\rho_2$ .

- $a = slice(\iota, \pi_i) = slice(\iota, \sigma_i)$
- $a = slice(\iota, \pi_{i+1}) = slice(\iota, \sigma_j)$

- $b = slice(\iota, \pi_j) = slice(\iota, \sigma_{i+1})$
- $c = slice(\iota, \pi_{j+1}) = slice(\iota, \sigma_{j+1})$

We know that  $b \geq c$  because  $\pi$  has the form described by Lemma 3, and we know that  $a \neq b$  and  $a \neq c$  because at most two elements can have  $a$  as slice in the identity permutation. Five different situations are possible:

1  $a > b > c$ : we have  $\sigma_i \sim \sigma_{i+1}$  in  $\sigma$  because  $a > b$ . Therefore, the inversion  $\rho_2(i, i+1)$  has  $\Delta_{SMP}(\pi \cdot \rho_1, \rho_2) < 0$  according to Lemma 1.

2  $b > c > a$ : we have  $\sigma_j \sim \sigma_{j+1}$  in  $\sigma$  because  $c > a$ . Therefore, the inversion  $\rho_2(i, i+1)$  has  $\Delta_{SMP}(\pi \cdot \rho_1, \rho_2) < 0$  according to Lemma 1.

3  $b > a > c$ : we show that this case is not possible. We have assumed that  $slice(\pi, \pi_{i+1}) \geq slice(\pi, \pi_x)$  and  $slice(\pi, \pi_{i+1}) \geq slice(\pi, \pi_{x+1})$  for every pair  $\pi_x, \pi_{x+1}$  such that  $slice(\iota, \pi_x) = slice(\iota, \pi_{x+1})$ . Therefore, the element  $\pi_{j-1}$  cannot have  $slice(\iota, \pi_{j-1}) = b$ , which forces us to conclude that there is one element  $\pi_z$  such that  $slice(\iota, \pi_z) = b$  and  $z = i+2$ . However, since for each pair of elements  $\pi_x, \pi_y$  such that  $slice(\pi, \pi_x) = slice(\pi, \pi_y) > b$  cannot happen  $\pi_x$  and  $\pi_y$  on the same side, we must have more elements in one side of the permutation than in the other, which is impossible.

4  $b = c > a$ : we have  $\sigma_j \sim \sigma_{j+1}$  in  $\sigma$  because  $c > a$ . Therefore, the inversion  $\rho_2(i, i+1)$  has  $\Delta_{SMP}(\pi \cdot \rho_1, \rho_2) < 0$  according to Lemma 1.

5  $a > b = c$ : we have  $\sigma_i \sim \sigma_{i+1}$  in  $\sigma$  because  $a > b$ . Therefore, the inversion  $\rho_2(i, i+1)$  has  $\Delta_{SMP}(\pi \cdot \rho_1, \rho_2) < 0$  according to Lemma 1.

Lemmas 1 and 2 show cases such that at least one inversion will have positive benefit. No positive benefit inversion is guaranteed on permutations in the form described by Lemma 3, but in those cases we can use Lemmas 5 and 6 to assure that our greedy approach will eventually reach the identity permutation.

### Number of breakpoints heuristic

We use the term NB to refer to this heuristic as an acronym for Number of Break-points. Consider the extended permutation that can be obtained from  $\pi$  by inserting two new elements:  $\pi_0 = 0$  and  $\pi_{n+1} = n+1$ . The extended permutation is still denoted as  $\pi$ . Below, we present two definitions of breakpoint depending on whether we are dealing with signed or unsigned permutations.

**Definition 2** A pair of elements  $\pi_i, \pi_{i+1}$  in a signed permutation  $\pi$ , with  $0 \leq i \leq n$ , is a breakpoint if  $\pi_{i+1} - \pi_i \neq 1$ .

**Definition 3** A pair of elements  $\pi_i, \pi_{i+1}$  in an unsigned permutation  $\pi$ , with  $0 \leq i \leq n$ , is a breakpoint if  $|\pi_{i+1} - \pi_i| \neq 1$ .

We use  $NB(\pi)$  to represent the number of breakpoints in a permutation and  $\Delta_{NB}(\pi, \rho) = NB(\pi \cdot \rho) - NB(\pi)$  to

represent the variation in the number of break-points caused by an inversion  $\rho$ .

The identity permutation  $\iota$  is the only permutation with no breakpoints. Therefore, an inversion that decreases the number of breakpoints indirectly leads to the identity permutation. Since an inversion can affect only two breakpoints, we know that  $\Delta_{NB}(\pi, \rho) \in \{-2, -1, 0, 1, 2\}$ .

The benefit of an arbitrary inversion  $\rho$  can be computed as  $ben_{NB}(\pi, \rho) = -\frac{\Delta_{NB}(\pi, \rho)}{cost(\rho)}$ . We compute the benefit of all possible inversion and we choose one that maximizes the benefit.

Since breakpoint is a very common concept in the genome rearrangement field, previous publications have clearly stated which kind of permutations will not allow any inversion to decrease the number of breakpoints.

**Definition 4** Let  $\pi$  be an unsigned permutation, a strip of  $\pi$  is an interval  $[\pi_i, \dots, \pi_j]$  with no breakpoint such that  $(\pi_{i-1}, \pi_i)$  and  $(\pi_j, \pi_{j+1})$  are breakpoints.

Strips can be either increasing ( $\pi_i < \pi_{i+1} < \dots < \pi_j$ ) or decreasing ( $\pi_i > \pi_{i+1} > \dots > \pi_j$ ). Strips with only one element are considered decreasing strips. Kececiloglu and Sankoff [16] proved that every unsigned permutation with a decreasing strip has at least one inversion that removes at least one breakpoint. Therefore, those inversions will have positive (non-zero) benefit. The same idea holds for signed permutations.

When we find a permutation  $\pi$  with no decreasing strips, we have no positive benefit. Therefore, we can use one of the following strategies to create decreasing strips as a contingency plan.

1. Use the Left or Right Heuristic known as LR.
2. The Left or Right Heuristic could break one strip because only a single element will be moved. Therefore, another approach is to compute the cost of placing the strip that contains the left element in the left side and the strip that contains the right element in the right side and hence use the less costly option.
3. Revert the entire unsorted group with one inversion.
4. Find the increasing strips and compute the cost of all possible inversions that reverse one or more of them. Note that we do not consider inversions that split any strip. After that, use the less costly inversion. We named this approach as the Best Strip strategy.

The Best Strip strategy leads to the best results in our experiments. Therefore, we use it in our comparative analysis.

#### Number of breakpoints plus slice-misplaced pairs heuristic

We use the term NB+SMP to refer to this approach since it uses concepts retrieved from NB and SMP. We

decided to favor breakpoints reduction in our greedy function:  $\Delta_{NB+SMP}(\pi, \rho) = \Delta_{NB}(\pi, \rho) + \frac{\Delta_{SMP}(\pi, \rho)}{n^2}$ . We use

the benefit computed as  $ben_{NB+SMP}(\pi, \rho) = -\frac{\Delta_{NB+SMP}(\pi, \rho)}{cost(\rho)}$ .

For unsigned permutations, we use the inversion  $\rho$  that maximizes the benefit if, and only if,  $ben_{NB+SMP}(\pi, \rho) > 0$ . Otherwise, we use the Best\_Strip strategy in order to guarantee that the input permutation will be sorted.

For signed permutations, we compute the unsigned permutation  $\pi'$  by removing the signs from  $\pi$  and we apply in  $\pi$  the inversion  $\rho'$  that would be used in  $\pi'$  according to the method previously described for unsigned permutations. In the end, if  $\pi \neq \iota$ , we simply change the signs of negative elements with unitary inversions.

## Results and discussion

We implemented the algorithms using C++ Programming Language and experiments were executed at the cluster provided by the IN2P3 Computing Center (<http://cc.in2p3.fr/>).

Our source code is freely available at <https://github.com/chrbaudet/SWI-LS>.

We performed two batches of experiments. We first show experiments using small permutations and then we use considerably longer sequences up to size 100.

### Small permutations

We generated a dataset with all possible unsigned permutations up to size 12, which accounts for

$$\sum_{n=2}^{12} n! = 522,956,312 \text{ instances. We did the same}$$

for all possible signed permutations up to size 10, which accounts for  $\sum_{n=2}^{10} 2^n n! = 3,912,703,160$  instances.

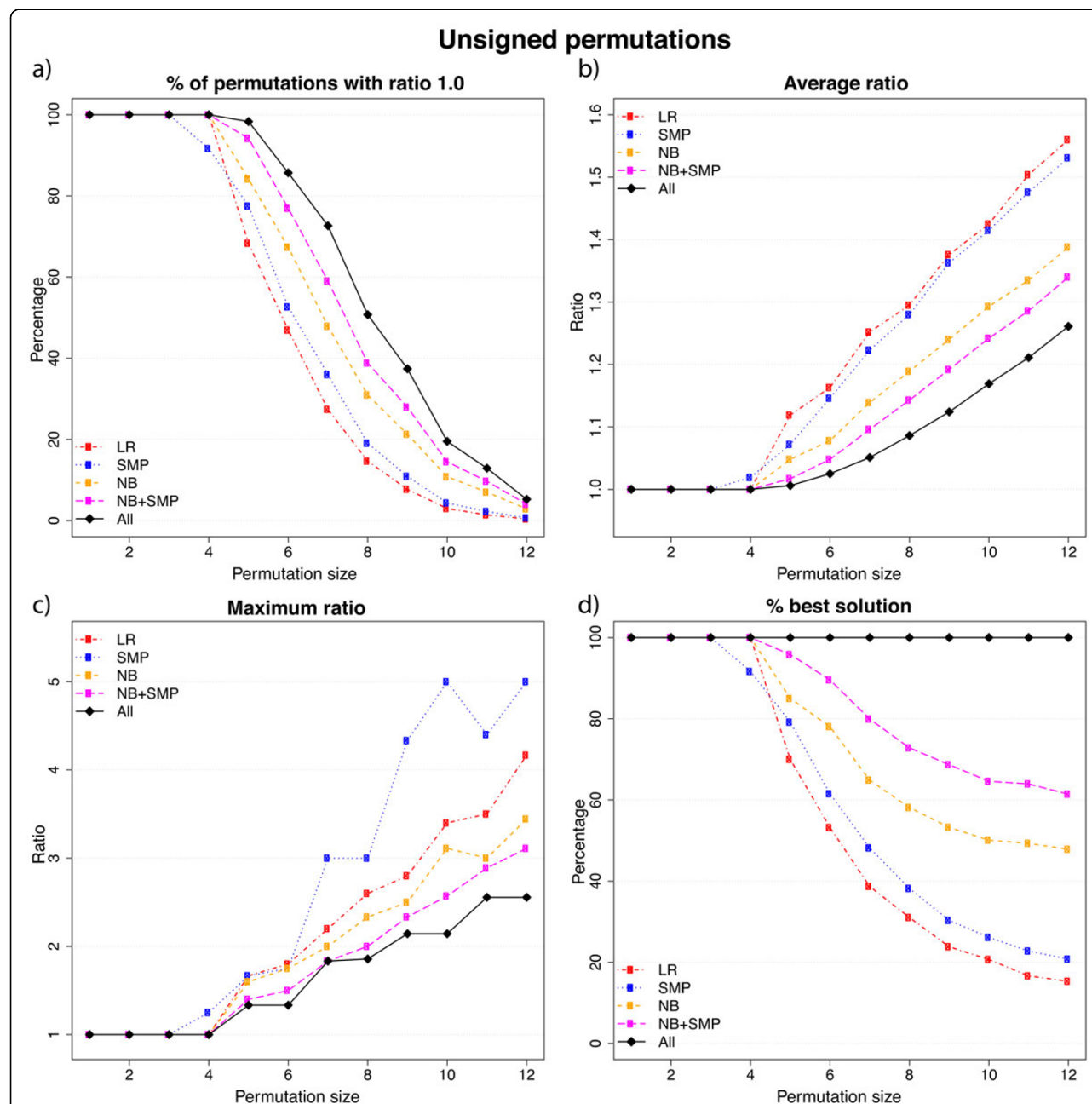
Therefore, we have used a large dataset having more than  $4.4 \times 10^9$  permutations.

For every permutation in the dataset, we were able to compute a minimum cost solution for comparison purposes. The minimum cost solution was calculated using a graph structure  $G_n$ , for  $n \in \{1, 2, \dots, 12\}$ . We define  $G_n$  as follows. A permutation  $\pi$  is a vertex in  $G_n$  if, and only if,  $\pi$  has  $n$  elements. Let  $\pi$  and  $\sigma$  be two vertices in  $G_n$ , we build an edge from  $\pi$  to  $\sigma$  if, and only if, there is an inversion  $\rho$  that transforms  $\pi$  into  $\sigma$ . The weight assigned to this edge is  $cost(\rho)$ . Finally, we calculate the shortest path from  $\iota$  to each vertex in  $G_n$  using a variant of Dijkstra's algorithm for the single-source shortest-paths problem. This variant gives us the minimum cost to sort permutations in  $G_n$ , as well as an optimum scenario of inversions.

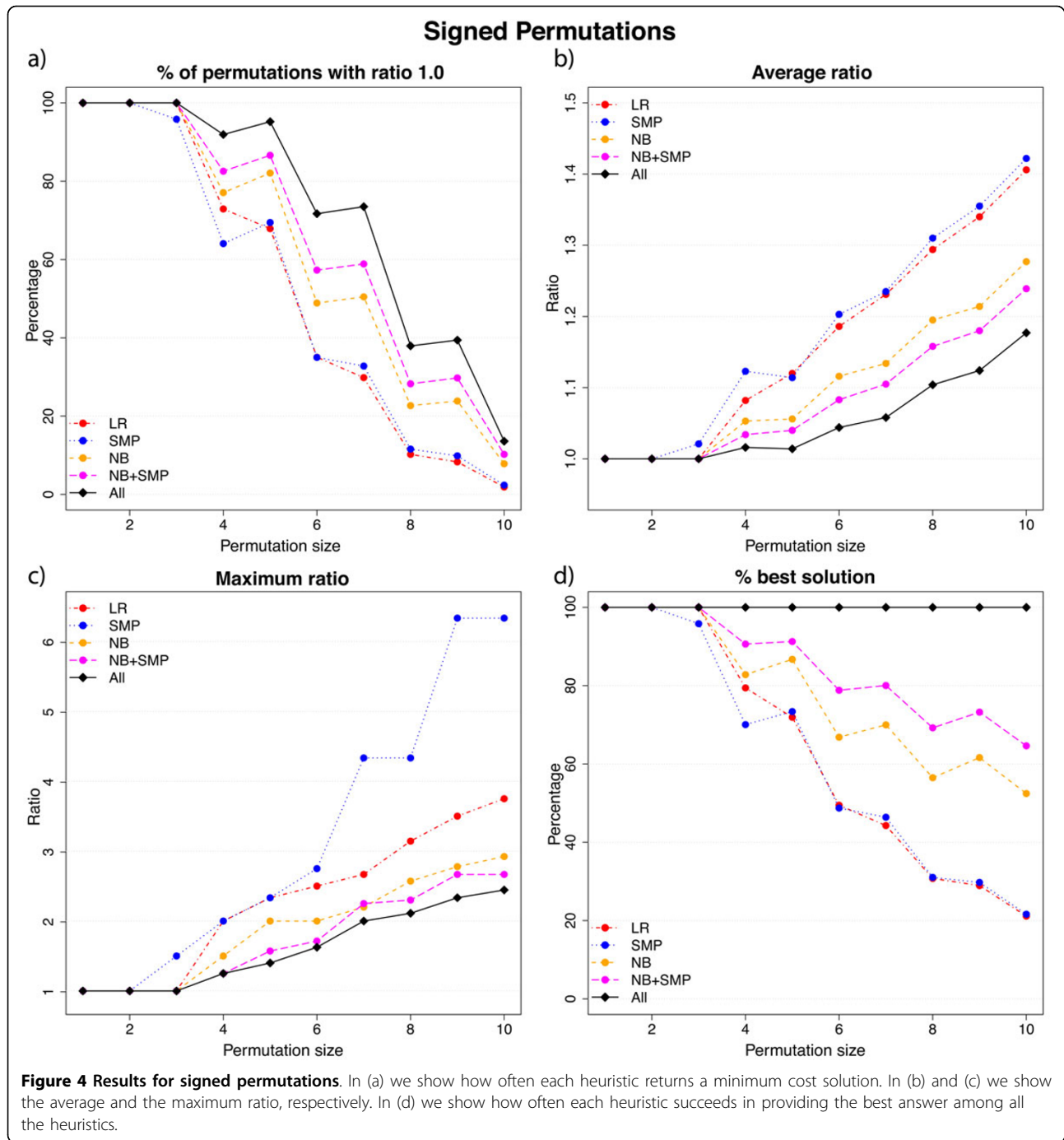
Let  $heu\_cost$  be the cost for sorting a permutation using some of our heuristics and  $opt\_cost$  be the optimum cost, we can compute the approximation ratio as  $\frac{heu\_cost}{opt\_cost}$ .

Figures 3 and 4 summarize our results for unsigned and signed permutations, respectively. The graphs in Figures 3(a) and 4(a) show how often each heuristic

returns the optimum cost. The graphs in Figures 3(b) and 4(b) show the average ratio considering all permutations of a given size, while the graphs in Figures 3(c) and 4(c) exhibit the maximum ratios for the same group of permutations. These graphs are maximum or average values and they may not answer the question: for a single instance  $\pi$ , is there any algorithm that is likely to provide the best answer? The graphs in Figures 3(d) and



**Figure 3 Results for unsigned permutations.** In (a) we show how often each heuristic returns a minimum cost solution. In (b) and (c) we show the average and the maximum ratio, respectively. In (d) we show how often each heuristic succeeds in providing the best answer among all the heuristics.



4(d) discuss this question by assessing the number of times each algorithm provides the least costly sequence. The values we present do not add up to 100% because of ties.

We observe that NB+SMP leads to the best results and it is consistently better than using just the number of breakpoints (NB) or just the variation in the number of slice-misplaced pairs (SMP) in every aspects we plot in Figures 3 and 4.

Individually, NB and SMP may be worse than NB+SMP, but NB returns results that are much closer to the optimum solution than SMP. That is true both on signed and unsigned permutations as we can see in Figures 3(b) and 4(b). The other graphs also corroborate this fact, which supports our decision of favoring the number of breakpoints when we compute  $\Delta_{NB+SMP}$ .

The simplistic approach LR leads to inferior results as reasonable. However, when we consider only the

maximum ratio aspect, we observe that LR outdoes SMP. Indeed, the SMP approach is not consistent with respect to this aspect, which indicates that particular permutations are hard to sort using only the slice-misplaced pairs.

A final test checks if any profit is gained from running all possible heuristic. We added a new curve labeled as All, which selects for each instance the less costly result between those produced by our heuristics. As we can see, running all possible heuristics and keeping the best result accomplishes very good results. Indeed, it is consistently better than using solely the NB+SMP heuristic.

### Large permutations

We ran our algorithm on a set of arbitrarily large permutations. This set is composed of 190,000 random signed permutations and 190,000 random unsigned permutations. In both cases, the permutation size ranges from 10 to 1000 in intervals of 5, with 10,000 permutations of each size. Here, we do not have an exact solutions for these permutations. Therefore, we use the average cost instead of the approximation ratio to base our analysis.

The analysis on random permutations reinforces the notion that NB+SMP leads to the best results. We observe in Figures 5 and 6 that the difference between NB and NB+SMP is very small on average. However, NB+SMP returns the less costly scenario in more cases.

Using random permutations allows us to draw information about the running time of each heuristic. In Table 1 we observe that LR is the fastest heuristic and

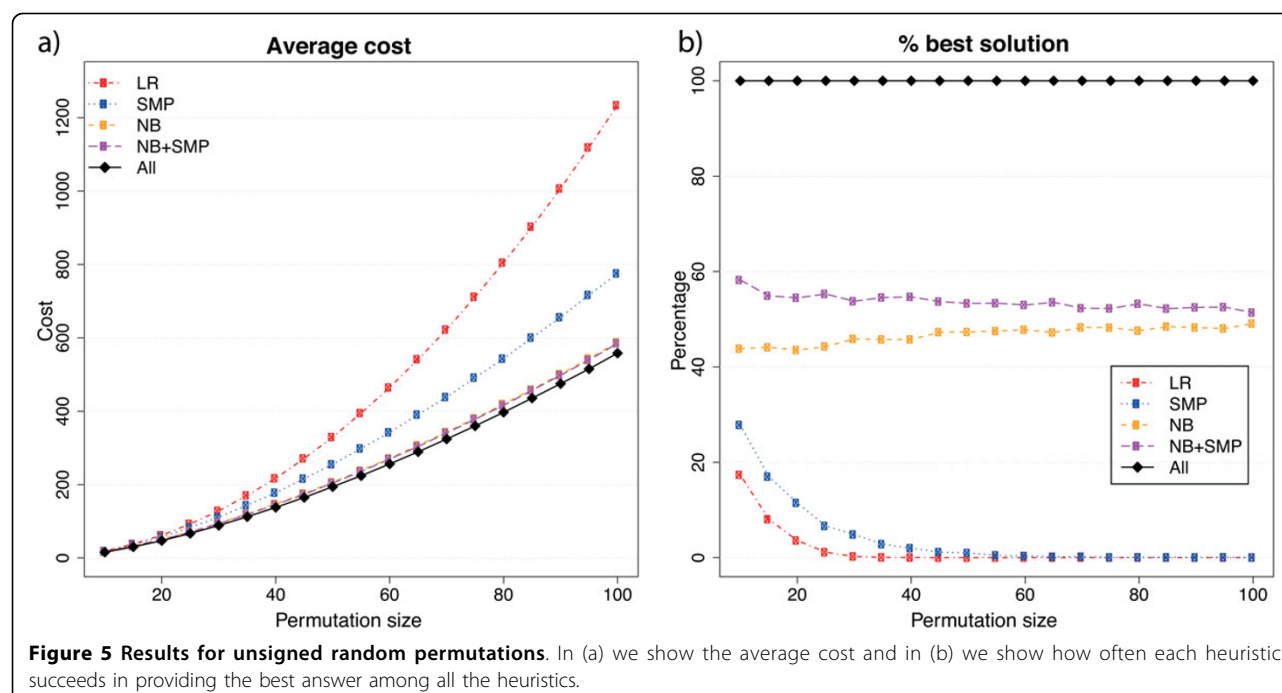
the sorting scenario can be obtained almost instantly (less than 1 milliseconds), which is reasonable since this heuristic is very simplistic. The heuristic SMP and NB+SMP are the ones that take more time to finish, which can be explained, in part, by the fact that both need to compute slice-misplaced pairs. Since SMP returns scenarios with more inversions than NB+SMP, the former requires about twice the time used by the latter.

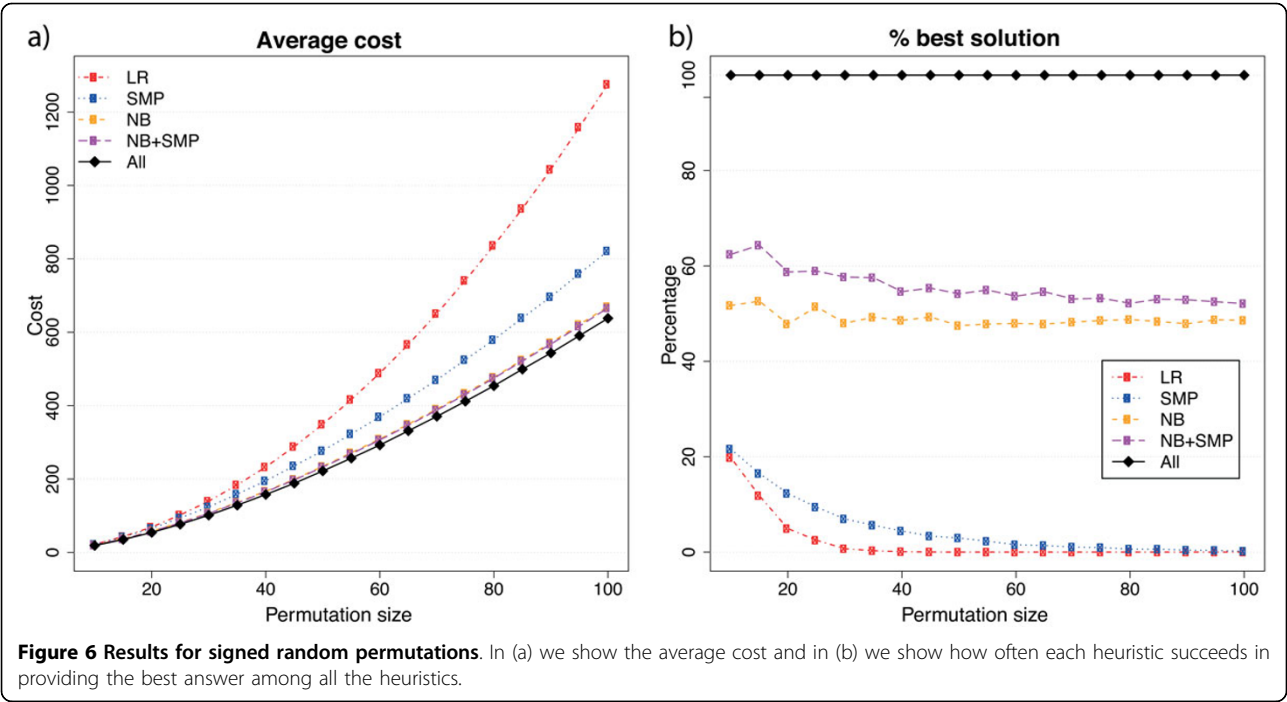
### Conclusions

We have defined a new genome rearrangement problem based on the concepts of symmetry and length of the reversed segments in order to assign a cost for each inversion. The problem we are proposing aims at finding low-cost scenarios between genomes. We considered the cases when gene orientations is taken into account and when it is not. We have provided the first steps in exploring this problem.

We presented several heuristics and we assessed their performances on a large set of more than  $4.4 \times 10^9$  permutations. The ideas we used to develop these heuristics together with the experimental results set the stage for a proper theoretical analysis.

As in other problems in the genome rearrangement field, we would like to know the complexity of determining the distance between any two genomes using the operations we defined. That seems to be a difficult problem that we intend to keep studying. We plan to design approximation algorithms and more effective heuristics.





**Table 1 Average running time for each permutation in milliseconds**

| size | Unsigned Permutations |        |     |        | Signed Permutations |        |     |        |
|------|-----------------------|--------|-----|--------|---------------------|--------|-----|--------|
|      | LR                    | SMP    | NB  | NB+SMP | LR                  | SMP    | NB  | NB+SMP |
| 10   | 0                     | 0      | 0   | 0      | 0                   | 0      | 0   | 0      |
| 20   | 0                     | 3      | 0   | 2      | 0                   | 4      | 0   | 2      |
| 30   | 0                     | 24     | 2   | 15     | 0                   | 34     | 3   | 17     |
| 40   | 0                     | 110    | 9   | 63     | 0                   | 149    | 10  | 69     |
| 50   | 0                     | 356    | 22  | 188    | 0                   | 468    | 23  | 204    |
| 60   | 0                     | 938    | 50  | 466    | 0                   | 1,210  | 51  | 497    |
| 70   | 0                     | 2,110  | 88  | 994    | 0                   | 2,681  | 89  | 1,048  |
| 80   | 0                     | 4,259  | 148 | 1,922  | 0                   | 5,464  | 150 | 2,068  |
| 90   | 0                     | 7,980  | 231 | 3,457  | 0                   | 9,959  | 230 | 3,624  |
| 100  | 0                     | 13,876 | 349 | 5,843  | 0                   | 17,155 | 345 | 6,106  |

**Competing interests**  
The authors declare that they have no competing interests.

**Authors' contributions**  
ZD came up with the idea of using the cost function that uses both symmetry and length of the affected segment. CB, ZD and UD designed the algorithms and the batches of experiments. UD and CB programmed an early prototype in python. After a series of iterative improvements, CB developed the final C++ code. UD drafted the manuscript. All authors read and approved the final manuscript.

**Acknowledgements**  
This work was supported by a Postdoctoral Fellowship from FAPESP to UD (number 2012/01584-3), by project fundings from CNPq (numbers 477692/2012-5 and 483370/2013-4), FAPESP (number 2014/19401-8) and CAPES/COFECUB (number 831/15) to ZD, and by French Project ANR MIRI BLAN08-1335497 and the ERC Advanced Grant SISYPHE to CB. The authors also thank the Center for Computational Engineering and Sciences at Unicamp for financial support through the FAPESP/CEPID Grant 2013/08293-7.

Experiments were executed at the cluster provided by the IN2P3 Computing Center (<http://cc.in2p3.fr/>).

This article has been published as part of BMC Bioinformatics Volume 16 Supplement 19, 2015: Brazilian Symposium on Bioinformatics 2014. The full contents of the supplement are available online at <http://www.biomedcentral.com/bmcbioinformatics/supplements/16/S19>

**Declarations**  
This work was supported by CNPq grants (numbers 477692/2012-5 and 483370/2013-4), FAPESP grants (numbers 2012/01584-3, 2013/08293-7 and 2014/19401-8) and CAPES/COFECUB (number 831/15), by French Project ANR MIRI BLAN08-1335497 and the ERC Advanced Grant SISYPHE (grant number [247073]10). The publication charges of this article were funded by FAPESP/CEPID grant 2013/08293-7.  
This article has been published as part of BMC Bioinformatics Volume 16 Supplement 19, 2015: Brazilian Symposium on Bioinformatics 2014. The full contents of the supplement are available online at <http://www.biomedcentral.com/bmcbioinformatics/supplements/16/S19>

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Published: 16 December 2015

# References

1. Darling AE, Miklós I, Ragan MA: **Dynamics of genome rearrangement in bacterial populations.** *PLoS Genetics* 2008, **4**(7):1000128.
2. Dias U, Dias Z, Setubal JC: **A simulation tool for the study of symmetric inversions in bacterial genomes.** *Comparative Genomics. Lecture Notes in Computer Science Springer* 2011, **6398**:240-251.
3. Hannenhalli S, Pevzner PA: **Transforming cabbage into turnip: polynomial algorithm for sorting signed permutations by reversals.** *Journal of the ACM* 1999, **46**(1):1-27.
4. Eisen JA, Heidelberg JF, White O, Salzberg SL: **Evidence for symmetric chromosomal inversions around the replication origin in bacteria.** *Genome Biology* 2000, **1**(6):0011-100119.
5. Lefebvre JF, El-Mabrouk N, Tillier E, Sankoff D: **Detection and validation of single gene inversions.** *Bioinformatics* 2003, **19**(Suppl 1):190-196.
6. Sankoff D, Lefebvre JF, Tillier E, El-Mabrouk N: **The distribution of inversion lengths in bacteria.** *Comparative Genomics. Lecture Notes in Computer Science, Springer* 2005, **3388**:97-108.
7. Caprara A: **Sorting permutations by reversals and Eulerian cycle decompositions.** *SIAM Journal on Discrete Mathematics* 1999, **12**(1):91-110.
8. Swidan F, Bender M, Ge D, He S, Hu H, Pinter R: **Sorting by length-weighted reversals: Dealing with signs and circularity.** *Combinatorial Pattern Matching. Lecture Notes in Computer Science, Springer* 2004, **3109**:32-46.
9. Arruda TS, Dias U, Dias Z: **Heuristics for the sorting by length-weighted inversions problem on signed permutations.** *Algorithms for Computational Biology. Lecture Notes in Computer Science* 2014, **8542**:59-70.
10. Pinter RY, Skiena S: **Genomic sorting with length-weighted reversals.** *Genome Informatics* 2002, **13**:2002.
11. Bender MA, Ge D, He S, Hu H, Pinter RY, Skiena S, Swidan F: **Improved bounds on sorting by length-weighted reversals.** *Journal of Computer and System Sciences* 2008, **74**(5):744-774.
12. Arruda TS, Dias U, Dias Z: **Heuristics for the sorting by length-weighted inversion problem.** *Proceedings of the International Conference on Bioinformatics, Computational Biology and Biomedical Informatics* 2013, 498-507.
13. Ohlebusch E, Abouelhoda MI, Hockel K, Stallkamp J: **The median problem for the reversal distance in circular bacterial genomes.** *Proceedings of Combinatorial Pattern Matching* 2005, 116-127.
14. Dias Z, Dias U, Setubal JC, Heath LS: **Sorting genomes using almost-symmetric inversions.** *Proceedings of the 27th Symposium On Applied Computing (SAC'2012), Riva del Garda, Italy* 2012, 1-7.
15. Dias U, Baudet C, Dias Z: **Greedy randomized search procedure to sort genomes using symmetric, almost-symmetric and unitary inversions.** *Proceedings of the International Conference on Bioinformatics, Computational Biology and Biomedical Informatics. BCB'13 ACM, New York, NY, USA;* 2013, 181-190.
16. Kececioğlu JD, Ravi R: **Of Mice and Men: Algorithms for Evolutionary Distances Between Genomes with Translocation.** *Proceedings of the 6th Annual Symposium on Discrete Algorithms* 1995, 604-613.

doi:10.1186/1471-2105-16-S19-S3

**Cite this article as:** Baudet et al.: Sorting by weighted inversions considering length and symmetry. *BMC Bioinformatics* 2015 **16**(Suppl 19): S3.

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