

A Greedy Approach to Motif Finding

Selected Topics in Computer Intelligence - 2015

Bioinformatics Programming

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Greedy Algorithms

- Choosing interesting or sensible option and not consider any other
 - Coin changing problem!!! – choosing the largest one first, minimize the number of coins returned
- Sometimes returns incorrect results in some cases
 - Short-sighted notion of “good” option
 - Aka. Suboptimal results
- Takes very little time

Genome Rearrangement



■ Waardenburg Syndrome


- Described in detail by the Dutch ophthalmology in 1951
 - A rare human genetic disorder resulting in **hearing loss** and **pigmentary abnormality**
 - e.g., Two differently colored eyes
 - Gene implication in **human chromosome 2**
- Mutated mouse with pigmentary abnormalities had been studied and found a “**splotch**” **gene** in its chromosomes

Genome Rearrangement



- According to gene mapping:
 - There are groups of genes in mice that appear in the same order as they do in humans
- Human genome is like the mouse genome cut into about 300 large genomic fragments (synteny blocks) that have been pasted together in different order
 - Both sequences are two different shuffling of the ancient mammalian genome
- Synteny
 - The condition of possessing common chromosome sequence
 - The conservation of blocks of order within two sets of chromosomes that are being compared with each other

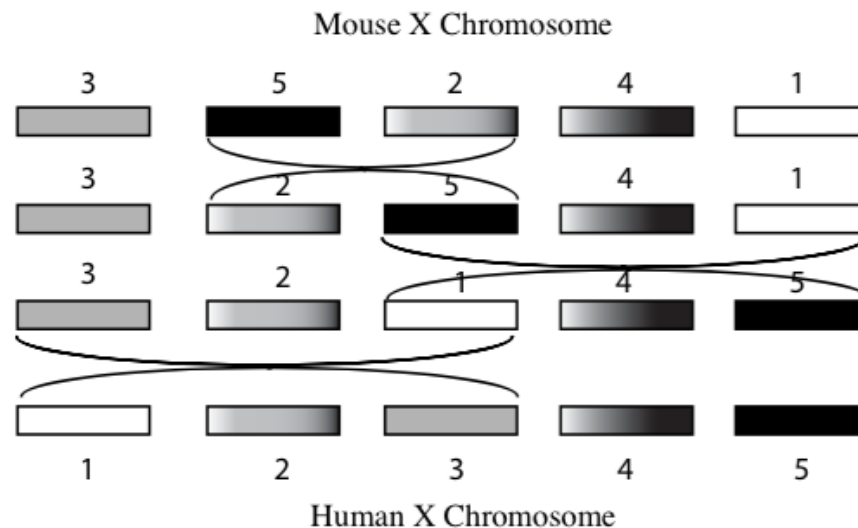
Genome Rearrangement

- Find a gene in mice leads to clues about the location of the related gene in humans 
- Series of rearrangements can alter the genomic architecture of a species
- Analyzing the rearrangement history of genomes is a challenge problem
 - Solving the combinatorial puzzle of finding a series of rearrangements that transform one genome to another

Genome Rearrangement

■ Rearrangement scenario

- The elementary rearrangement event is the **flipping** of genomic segment – a **reversal** or an inversion



- Biologists are interested in the **scenario involving the smallest number of reversals**
 - **Lower bound** of rearrangement that could occur

Sorting by Reversals

- Rearrangement events can be modeled by a series of reversals
- The order of genes (synteny blocks) can be represented by a permutation

$$\pi = \pi_1 \pi_2 \cdots \pi_n$$

- E.g. synteny blocks on chromosome X vs. ordering in mice

$$(1, 2, 3, 4, 5) \text{ vs. } (3, 5, 2, 4, 1)$$

Sorting by Reversals

- A reversal $\rho(i, j)$ transforms

$$\pi = \pi_1 \cdots \pi_{i-1} \underbrace{\pi_i \pi_{i+1} \cdots \pi_{j-1} \pi_j \pi_{j+1} \cdots \pi_n}_{\rightarrow}$$

into

$$\pi \cdot \rho(i, j) = \pi_1 \cdots \pi_{i-1} \underbrace{\pi_j \pi_{j-1} \cdots \pi_{i+1} \pi_i \pi_{j+1} \cdots \pi_n}_{\leftarrow}$$

- e.g.,

$$\pi = 1\,2\,4\,3\,7\,5\,6$$

$$\pi \cdot \rho(3, 6) = 1\,2\,5\,7\,3\,4\,6$$

Sorting by Reversals

■ Biological rearrangement process

Reversal Distance Problem:

Given two permutations, find a shortest series of reversals that transforms one permutation into another.

Input: Permutations π and σ .

Output: A series of reversals $\rho_1, \rho_2, \dots, \rho_t$ transforming π into σ (i.e., $\pi \cdot \rho_1 \cdot \rho_2 \cdots \rho_t = \sigma$), such that t is minimum.

$d(\pi, \sigma)$ – reversal distance between π and σ

in practice, we use σ to be the identity permutation $1, 2, \dots, n$

Sorting by Reversals

Sorting by Reversals Problem:

Given a permutation, find a shortest series of reversals that transforms it into the identity permutation.

Input: Permutation π .

Output: A series of reversals $\rho_1, \rho_2, \dots, \rho_t$ transforming π into the identity permutation such that t is minimum.

$d(\pi)$ – reversal distance between π and the *identity permutation*

■ We do not need to move the *already-sorted elements*

$$\pi = 1\,2\,3\,6\,4\,5.$$

$$1\,2\,3\,\underline{6}\,\underline{4}\,5 \rightarrow 1\,2\,3\,4\,\underline{6}\,5 \rightarrow 1\,2\,3\,4\,5\,6.$$

$$\text{prefix}(\pi) = 123$$

Sorting by Reversals

- Moving its i^{th} element to the j^{th} position

SIMPLEREVERALSORT(π)

```
1  for  $i \leftarrow 1$  to  $n - 1$ 
2       $j \leftarrow$  position of element  $i$  in  $\pi$  (i.e.,  $\pi_j = i$ )
3      if  $j \neq i$ 
4           $\pi \leftarrow \pi \cdot \rho(i, j)$ 
5          output  $\pi$ 
6      if  $\pi$  is the identity permutation
7          return
```

- A greedy algorithm that chooses the “best” reversals at every step but increasing $prefix(\pi)$ does NOT guarantee the optimal

612345 \rightarrow 162345 \rightarrow 126345 \rightarrow 123645 \rightarrow 123465 \rightarrow 123456

612345 \rightarrow 543216 \rightarrow 123456 *optimal!!!*

Sorting by Reversals

- **SIMPLEREVERALSORT** is not a correct algorithm!
 - Takes $n - 1$ steps to sort $\pi = n\ 1\ 2\ \dots\ (n - 1)$
 - Even though the **minimum reversal distance**, $d(\pi) = 2$
- **Pancake Flipping Problem:**
 - Finding $d(\pi)$ of the form $\rho(1, i)$ sorting π
 - Rearranging pancake
 - **Goal:** Smaller on top of the larger one
 - Reversing **several from the top**
 - Repeating **as many times as necessary**
 - **What is the maximum number of flips?**
 - Given that there are n pancakes



(123645 \rightarrow 632145 \rightarrow 541236 \rightarrow 321456 \rightarrow 123456)

Approximation Algorithms

- Computer scientists often find a compromise in approximation algorithms – **approx. solution**
 - Better than an optimal one (with trade-off)

- Approx. ratio** of algorithm A on input π is defined as

$$\frac{\mathcal{A}(\pi)}{OPT(\pi)}$$

Soln. produced by algorithm A
The optimal soln. of the problem

- Performance guarantee of algorithm A** (approx. ratio)

$$\max_{|\pi|=n} \frac{\mathcal{A}(\pi)}{OPT(\pi)}$$

A is minimization algorithm

$$\min_{|\pi|=n} \frac{\mathcal{A}(\pi)}{OPT(\pi)}$$

A is maximization algorithm

*Worst-case
scenario*

Approximation Algorithms

- SIMPLEREVERSALSORT has approx. ratio = $(n - 1) / 2$
 - If n is 1001, the algorithm could have reversals of 500 times (of the optimal)
- Goal: design approx. algorithms with better performance guarantee

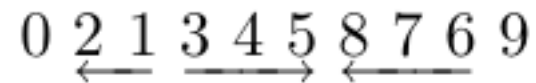
Breakpoints

- **SIMPLEREVERSALSORT** is a naive measure of progress toward the identity permutation
- Better algorithm can be derived based on breakpoints rather than $\text{prefix}(\pi)$
- Extending the permutation π by adding $\pi_0 = 0$ and $\pi_{n+1} = n+1$ on both ends
 - We do NOT move π_0 and π_{n+1}
- Adjacency – if π_i and π_{i+1} are consecutive numbers
- Breakpoint – if π_i and π_{i+1} are NOT consecutive numbers

Breakpoints

■ Example:

0 2 1 3 4 5 8 7 6 9



■ Extended by 0 and 9 on the ends

■ Decreasing strip (\leftarrow)

■ Increasing strip (\rightarrow)

■ Adjacencies: 21, 34, 45, 87, 76

■ Breakpoints: 02, 13, 58, 69

■ Maximum # breakpoints = $(n + 1)$

0 6 1 3 5 7 2 4 8

Breakpoints

- The **identity permutation** is the only one without breakpoints



$0\ 1\ 2\ 3\ 4\ 5\ \dots\ n\ (n+1)$

- A **breakpoint** must be separated in process of transforming π to the identity permutation
- Every reversal can eliminate at most two breakpoints

$$d(\pi) \geq \frac{b(\pi)}{2}, \quad b(\pi) \text{ is the number of breakpoints}$$

Breakpoints

Algorithm with breakpoint



BREAKPOINTREVERSALSORT(π)

```
1  while  $b(\pi) > 0$ 
2      Among all reversals, choose reversal  $\rho$  minimizing  $b(\pi \cdot \rho)$ 
3       $\pi \leftarrow \pi \cdot \rho$ 
4      output  $\pi$ 
5  return
```

- Why is this algorithm better than SimpleReversalSort?
- Does removing breakpoints does not introduce other?
- **Strip** – interval between two consecutive breakpoints
 - *Increasing* and *decreasing* strips
 - Single-element \rightarrow *decreasing* strip
 - Element 0 and $(n+1) \rightarrow$ *increasing* strip

Approximation Algorithms

■ Theorem 5.1

- If a permutation π contains a decreasing strip
- There is a reversal ρ that decreases # breakpoints in π

$$b(\pi \cdot \rho) < b(\pi)$$

■ Example:

0 1 2 7 6 5 8 4 3 9

→ ← ← ← →

0 1 2 7 6 5 8 4 3 9 → 0 1 2 3 4 8 5 6 7 9

→ ← → →

- If a permutation π does NOT contain a decreasing strip
- Find any increasing strip (exclude π_0 and π_{n+1}) and flip it

Breakpoints

■ Improved Breakpoint algorithm

IMPROVEDBREAKPOINTREVERALSORT(π)

```
1  while  $b(\pi) > 0$ 
2      if  $\pi$  has a decreasing strip
3          Among all reversals, choose reversal  $\rho$  minimizing  $b(\pi \cdot \rho)$ 
4      else
5          Choose a reversal  $\rho$  that flips an increasing strip in  $\pi$ 
6       $\pi \leftarrow \pi \cdot \rho$ 
7      output  $\pi$ 
8  return
```

Breakpoints

■ Theorem 5.2

- The algorithm is an approx. algorithm with a **performance guarantee** of at most 4
- The algorithm will eliminate at least one breakpoint in every two steps
 - Firstly flip the increasing strip, then remove breakpoints
 - **Worst-case** requires $2b(\pi)$ steps
 - The **approx. ratio** = $2b(\pi)/d(\pi)$, where $d(\pi) \geq \frac{b(\pi)}{2}$
 - The **performance guarantee** bounded above by:

$$\frac{2b(\pi)}{d(\pi)} \leq \frac{2b(\pi)}{\frac{b(\pi)}{2}} = 4.$$

A Greedy Motif Finding

■ Based on Gary Stormo and Gerald Hertz

```
GREEDYMOTIFSEARCH( $DNA, t, n, l$ )
1  bestMotif  $\leftarrow (1, 1, \dots, 1)$ 
2   $s \leftarrow (1, 1, \dots, 1)$ 
3  for  $s_1 \leftarrow 1$  to  $n - l + 1$ 
4      for  $s_2 \leftarrow 1$  to  $n - l + 1$ 
5          if  $Score(s, 2, DNA) > Score(\text{bestMotif}, 2, DNA)$ 
6               $BestMotif_1 \leftarrow s_1$ 
7               $BestMotif_2 \leftarrow s_2$ 
8   $s_1 \leftarrow BestMotif_1$ 
9   $s_2 \leftarrow BestMotif_2$ 
10 for  $i \leftarrow 3$  to  $t$ 
11     for  $s_i \leftarrow 1$  to  $n - l + 1$ 
12         if  $Score(s, i, DNA) > Score(\text{bestMotif}, i, DNA)$ 
13              $bestMotif_i \leftarrow s_i$ 
14      $s_i \leftarrow bestMotif_i$ 
15 return bestMotif
```