

CS342: Bioinformatics

Lecture 7

Greedy Algorithms

- **Def:** Algorithms that make locally optimal choices using a metric with the hope of finding a globally optimal solution.
- **Example:** Making change with US coins.
- **Optimization Problem:** Given an input, compute a solution, subject to various constraints, that either minimizes cost or maximizes profit.

Coin-Changing: Greedy Algorithm

Cashier's algorithm. At each iteration, add coin of the largest value that does not take us past the amount to be paid.

```
Sort coins denominations by value:  $c_1 < c_2 < \dots < c_n$ .  
  ↙ coins selected  
 $S \leftarrow \phi$   
while ( $x \neq 0$ ) {  
    let  $k$  be largest integer such that  $c_k \leq x$   
    if ( $k = 0$ )  
        return "no solution found"  
     $x \leftarrow x - c_k$   
     $S \leftarrow S \cup \{k\}$   
}  
return  $S$ 
```

Greedy Motif Search

GreedyMotifSearch(DNA, k, t)

BestMotifs \leftarrow empty motif list
BestScore $\leftarrow t * k$

for each k-mer Motif in the first string from DNA

Motif1 \leftarrow Motif

for i = 2 to t

 form Profile from motifs Motif1, ..., Motifi - 1

 Motifi \leftarrow Profile-most probable k-mer in the i-th string in DNA

Motifs \leftarrow (Motif1, ..., Motift)

if Score(Motifs) < BestScore:

 BestMotifs \leftarrow Motifs

 BestScore \leftarrow Score(Motifs)

return BestMotifs

A Serious Scientific Problem ...

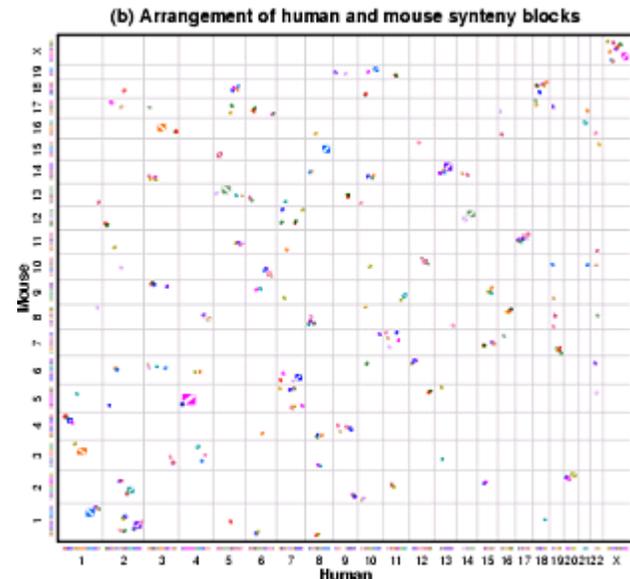
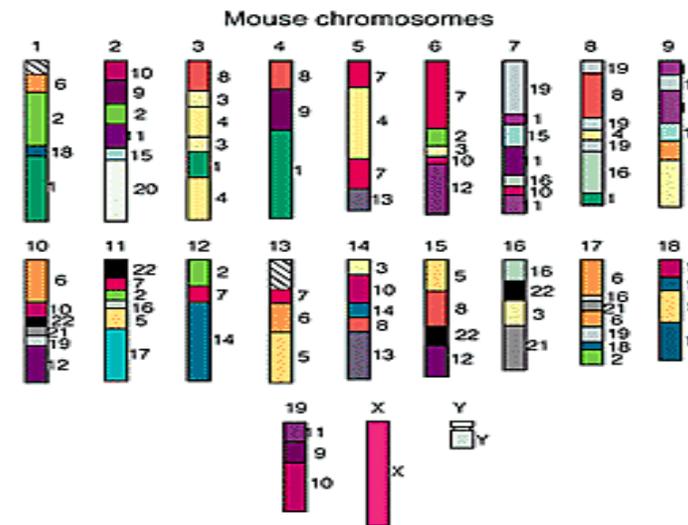
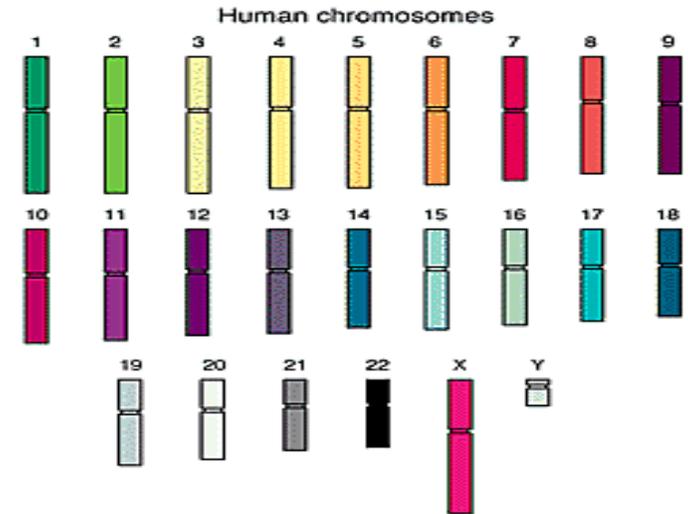
Differences between species?

- Some are obviously similar...
- Some are obviously different...
- Some are close calls...
- The differences that matter are in the genes!
- And the gene order is important!



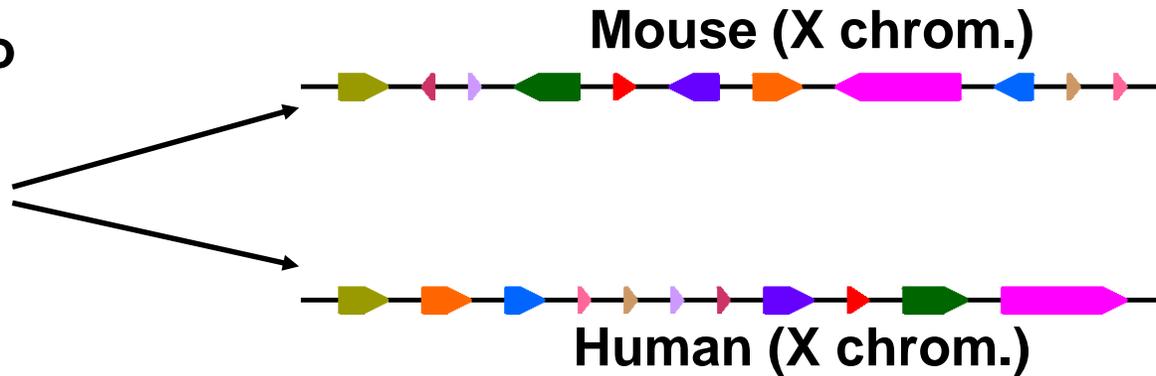
Genome Rearrangements

- Humans and mice have similar genomes, but their genes are ordered differently
- ~245 rearrangements
- ~ 300 large *synteny blocks*



Genome Rearrangements

Unknown ancestor
~ 75 million years ago



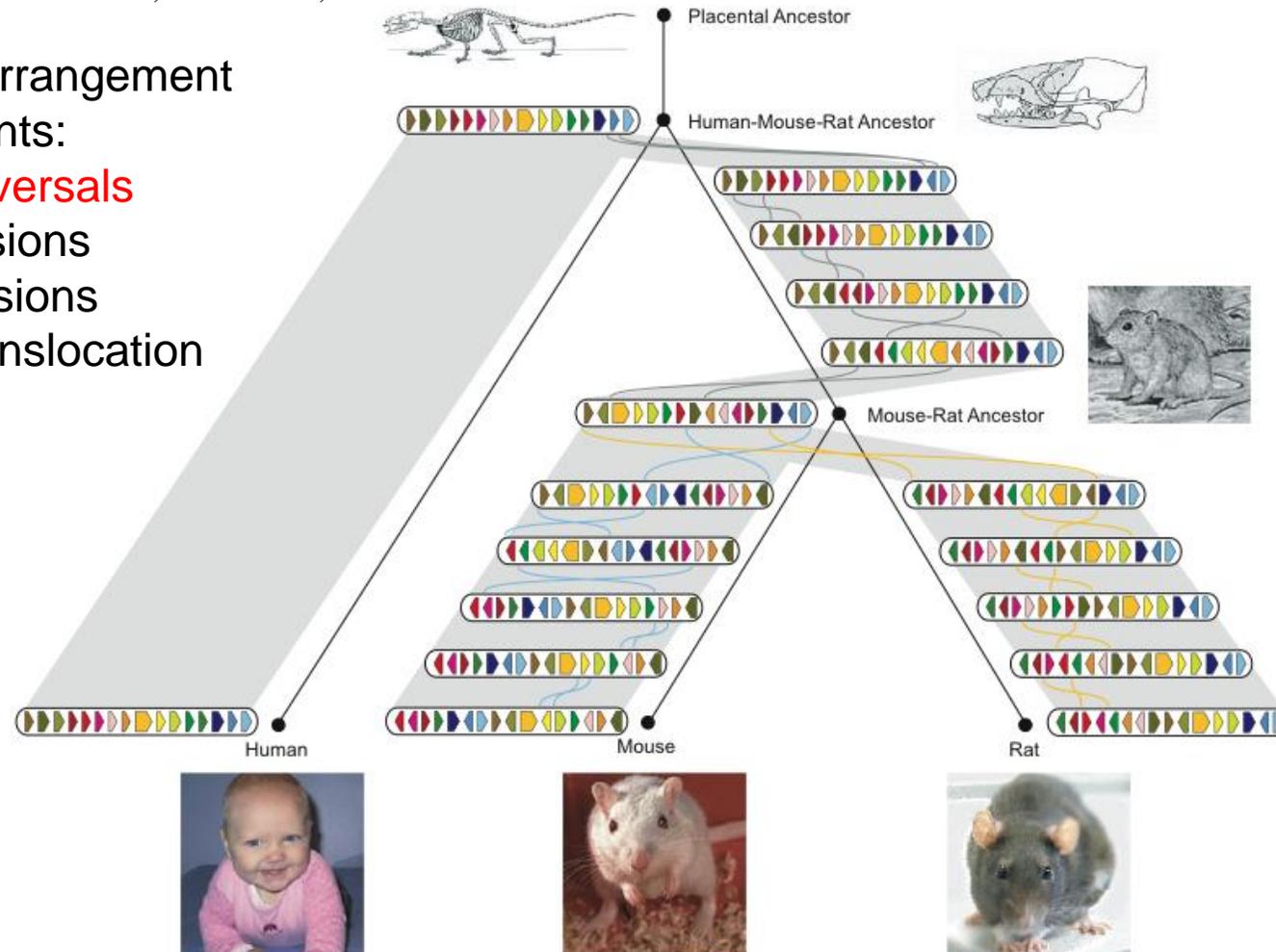
- What are the similarity blocks and how to find them?
- What is the architecture of the ancestral genome?
- What is the evolutionary scenario for transforming one genome into the other?

History of Chromosome X

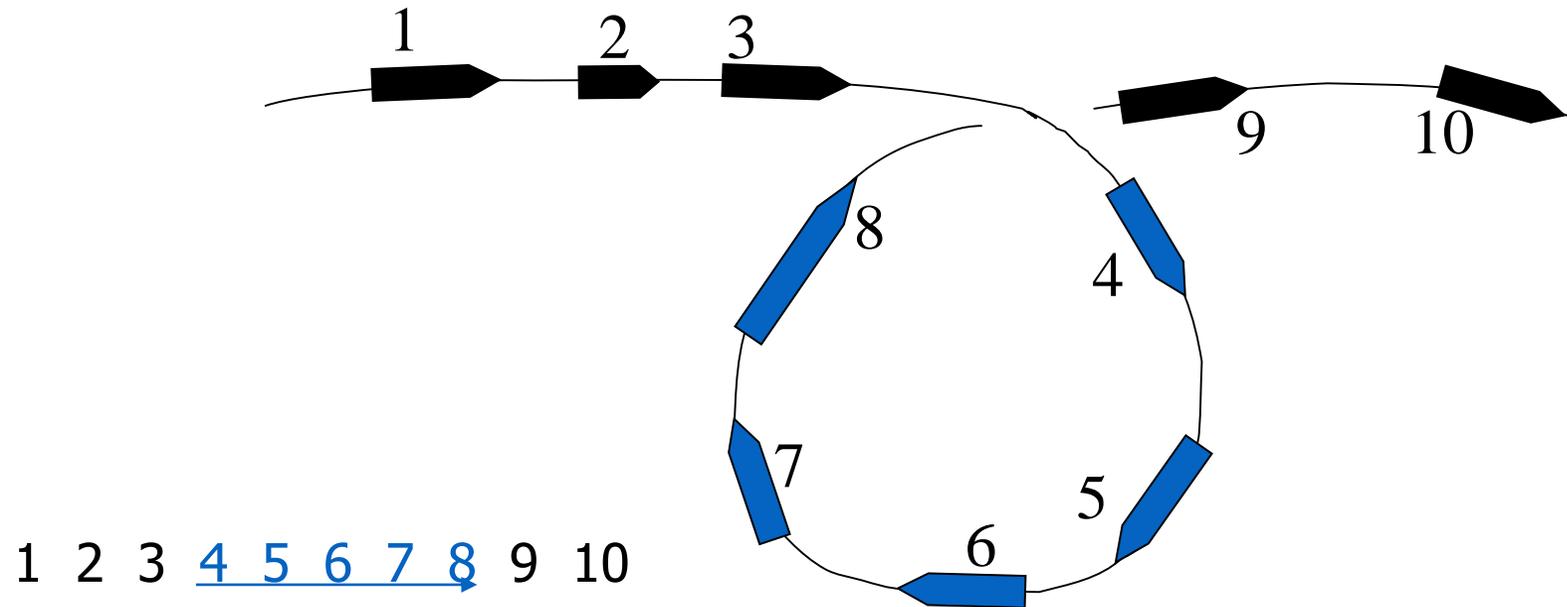
Rat Consortium, *Nature*, 2004

Rearrangement
Events:

- **Reversals**
- Fusions
- Fissions
- Translocation

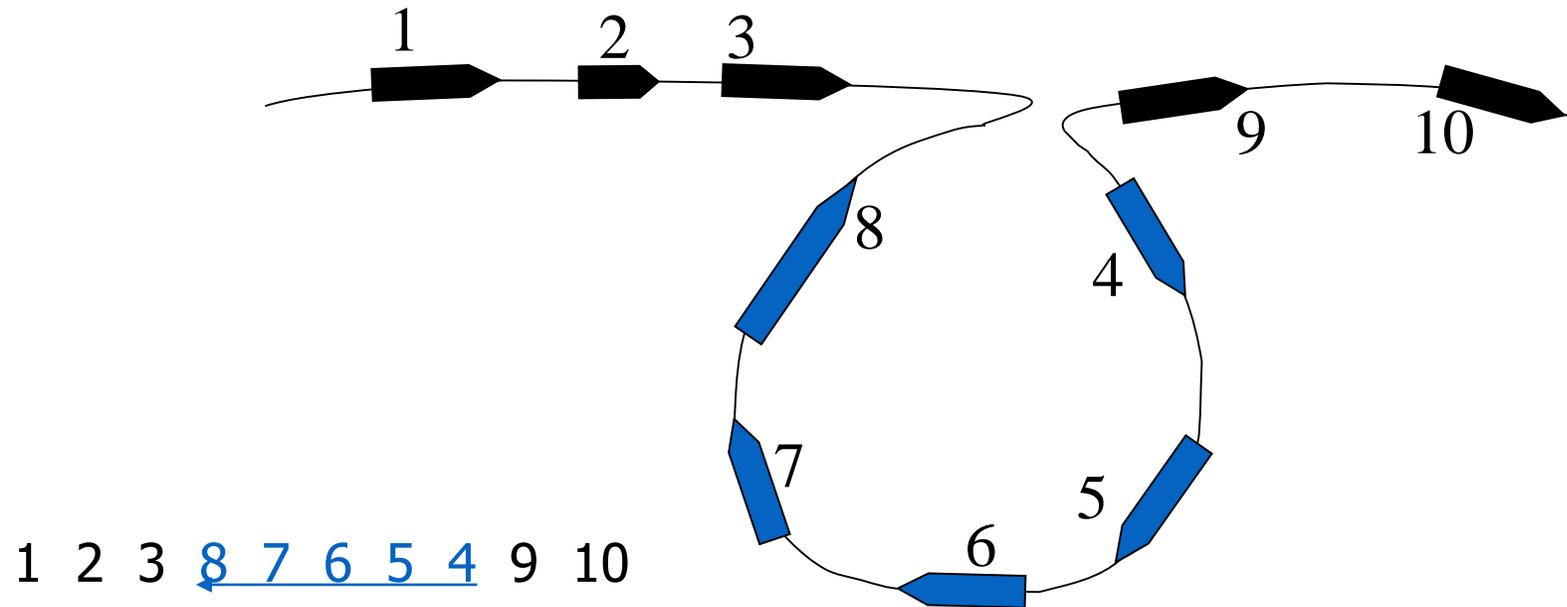


Reversals



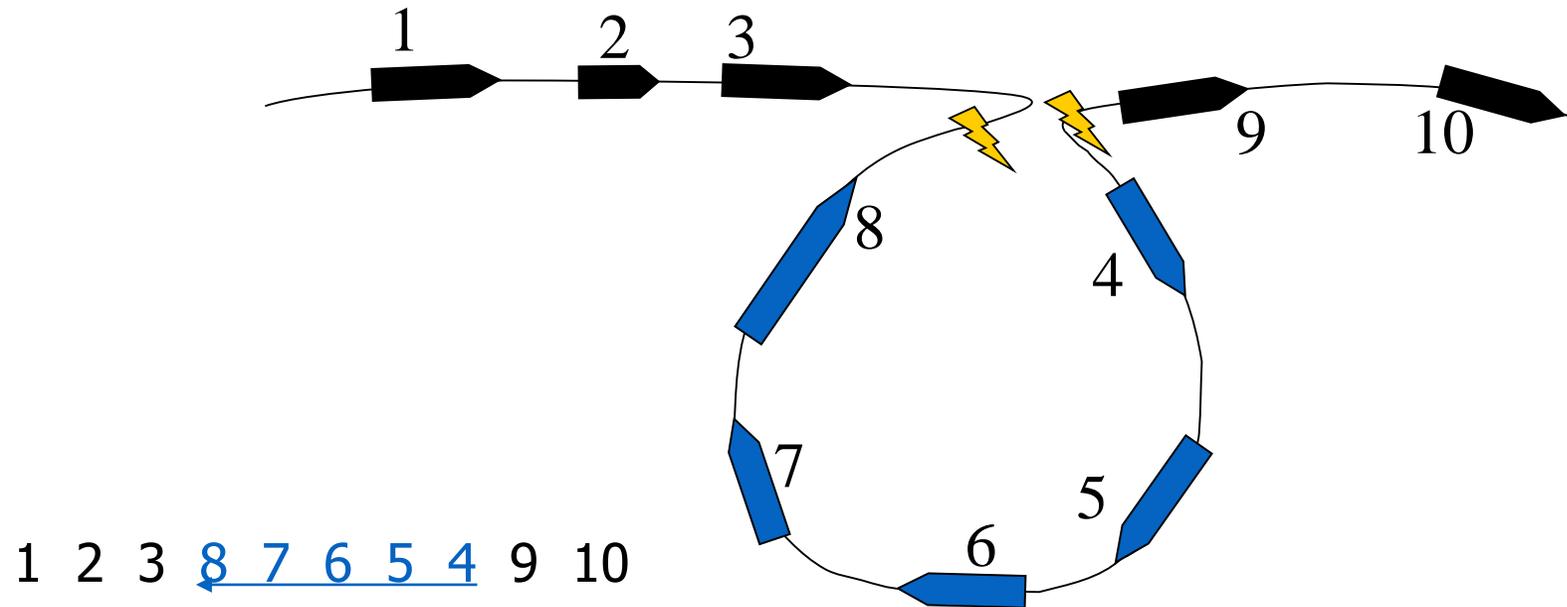
- Blocks represent conserved genes.
- Reversals, or *inversions*, are particularly relevant to speciation. Recombinations cannot occur between reversed and normally ordered segments.

Reversals



- Blocks represent conserved genes.
- In the course of evolution or in a clinical context, blocks 1 ... 10 could be reordered as 1 2 3 8 7 6 5 4 9 10.

Reversals and Breakpoints



The inversion introduced two *breakpoints* ⚡ (disruptions in order).

Reversals and Gene Orders

- Gene order can be represented by a permutation π :

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

$$\begin{array}{c} \rho(i,j) \\ \downarrow \\ \pi_1 \cdots \pi_{i-1} \pi_j \pi_{j-1} \cdots \pi_{i+1} \pi_i \pi_{j+1} \cdots \pi_n \end{array}$$

- Reversal $\rho(i, j)$ reverses (flips) the elements from i to j in π

Reversals: Example

$\pi = 1\ 2\ 3\ 4\ 5\ 6\ 7\ 8$

$\rho(3,5)$ ↓

1 2 5 4 3 6 7 8

$\rho(5,6)$ ↓

1 2 5 4 6 3 7 8

“Reversal Distance” Problem

- Goal: Given two permutations over n elements, find the shortest series of reversals that transforms one into another
- Input: Permutations π and σ
- Output: A series of reversals ρ_1, \dots, ρ_t transforming π into σ , such that t is minimum
- t - reversal distance between π and σ (*# of reversals*)
- $d(\pi, \sigma)$ - smallest possible value of t , given π and σ

“Sorting By Reversals” Problem

A simplified restatement of the same problem....

- Goal: Given a permutation, find a shortest series of reversals that transforms it into the identity permutation $(1\ 2\ \dots\ n)$
- Input: Permutation π
- Output: A series of reversals ρ_1, \dots, ρ_t transforming π into the identity permutation such that t is minimum
- **$t = d(\pi)$** - reversal distance of π

Sorting By Reversals: Example

$\pi =$ 3 4 2 1 5 6 7 10 9 8
4 3 2 1 5 6 7 10 9 8
4 3 2 1 5 6 7 8 9 10
1 2 3 4 5 6 7 8 9 10

$$d(\pi) = 3$$

Sorting by Reversals: 4 flips

Step 0: 2 4 3 5 8 7 6 1

Step 1: 2 3 4 5 8 7 6 1

Step 2: 2 3 4 5 6 7 8 1

Step 3: 8 7 6 5 4 3 2 1

Step 4: 1 2 3 4 5 6 7 8

What is the reversal distance for this permutation?

Can it be sorted in 3 flips?

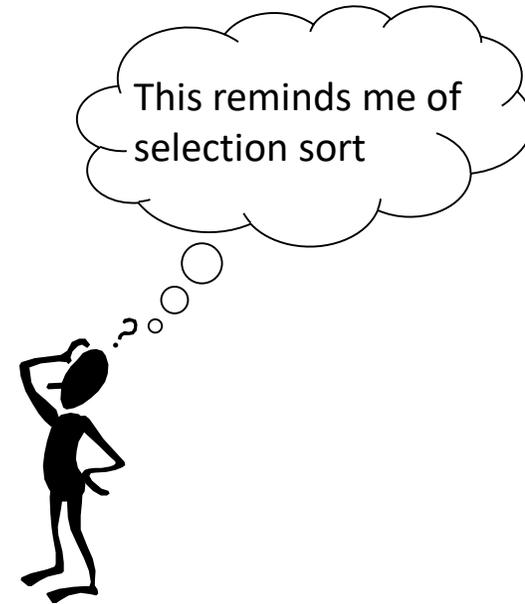
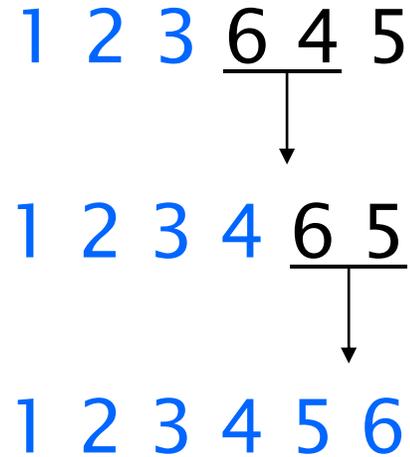
How can we know?

Sorting By Reversals: A Greedy Algorithm

- If sorting permutation $\pi = 1\ 2\ 3\ 6\ 4\ 5$, the first three elements are already in order so it does not make any sense to break them apart.
- The length of the already sorted prefix of π is denoted $prefix(\pi)$
 - $prefix(\pi) = 3$
- This results in an idea for a greedy algorithm: **increase $prefix(\pi)$ at every step**

Sort by Reversals: An Example

- Doing so, π can be sorted



- Number of steps to sort permutation of length n is at most $(n - 1)$

Greedy Algorithm

SimpleReversalSort(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      $\pi \leftarrow \pi \rho(i, j)$ 
5   output  $\pi$ 
6   if  $\pi$  is the identity permutation
7   return
```

Analyzing SimpleReversalSort

- SimpleReversalSort does not guarantee the smallest number of reversals and takes five steps on $\pi = \underline{6} 1 2 3 4 5$:

Flip 1: 1 6 2 3 4 5

Flip 2: 1 2 6 3 4 5

Flip 3: 1 2 3 6 4 5

Flip 4: 1 2 3 4 6 5

Flip 5: 1 2 3 4 5 6

Analyzing SimpleReversalSort

- But it can be sorted in two flips:

$$\pi = \underline{6\ 1\ 2\ 3\ 4\ 5}$$

$$\text{Flip 1: } \underline{5\ 4\ 3\ 2\ 1}\ 6$$

$$\text{Flip 2: } 1\ 2\ 3\ 4\ 5\ 6$$

- So, SimpleReversalSort(π) is not optimal
- Optimal algorithms are unknown for many problems; approximation algorithms are used

Approximation Algorithms

- Find *approximate* solutions rather than *optimal* solutions
- The **approximation ratio** of an algorithm \mathcal{A} on input π is:

$$\mathcal{A}(\pi) / \text{OPT}(\pi)$$

where

$\mathcal{A}(\pi)$ - solution produced by algorithm \mathcal{A}

$\text{OPT}(\pi)$ - optimal solution of the problem

Approximation Ratio/Performance Guarantee

- **Approximation ratio (performance guarantee)** of algorithm \mathcal{A} : max approximation ratio over all inputs of size n
 - For a minimizing algorithm \mathcal{A} (like ours):
 - Approx Ratio = $\max_{|\pi| = n} \mathcal{A}(\pi) / \text{OPT}(\pi) \geq 1.0$
 - For maximization algorithms:
 - Approx Ratio = $\min_{|\pi| = n} \mathcal{A}(\pi) / \text{OPT}(\pi) \leq 1.0$

Approximation Ratio

SimpleReversalSort(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      $\pi \leftarrow \pi \rho(i, j)$ 
5   output  $\pi$ 
6 if  $\pi$  is the identity permutation
7   return
```

Step 0: 6 1 2 3 4 5

Step 1: 1 6 2 3 4 5

Step 2: 1 2 6 3 4 5

Step 3: 1 2 3 6 4 5

Step 4: 1 2 3 4 6 5

Step 5: 1 2 3 4 5 6

Step 0: 6 1 2 3 4 5

Step 1: 5 4 3 2 1 6

Step 2: 1 2 3 4 5 6

