

# Greedy Algorithms

# US Change Problem

## United States Change Problem:

Convert some amount of money into the fewest number of coins.

**Input:** An amount of money,  $M$ , in cents.

**Output:** The smallest number of quarters  $q$ , dimes  $d$ , nickels  $n$ , and pennies  $p$  whose values add to  $M$  (i.e.,  $25q + 10d + 5n + p = M$  and  $q + d + n + p$  is as small as possible).

It is a **greedy algorithm**:

*At every step of iteration, a greedy algorithm tries to find the best optimal solution (e.g., used the most the coin with the biggest value)*

```
BETTERCHANGE( $M, \mathbf{c}, d$ )
1    $r \leftarrow M$ 
2   for  $k \leftarrow 1$  to  $d$ 
3      $i_k \leftarrow r/c_k$ 
4      $r \leftarrow r - c_k \cdot i_k$ 
5   return  $(i_1, i_2, \dots, i_d)$ 
```

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## Why greedy?

"greedy" means having excessive desire for something without considering the effect or damage done.

# US Change Problem

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6   return  $(i_1, i_2, \dots, i_d)$ 
```

Does it always find a correct solution?

When  $c_1 = 25, c_2 = 20, c_3 = 10, c_4 = 5, c_5 = 1$ ,

if  $M = 40$ , BetterChange returns  $i_1 = 1, i_3 = 1, i_4 = 1$

*We would solve the problem with  $i_2 = 2$ ...*

# US Change Problem

```
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4        $r \leftarrow r - c_k \cdot i_k$ 
5   return  $(i_1, i_2, \dots, i_d)$ 
```

We can ask ourselves: how close are we from the optimal solution?

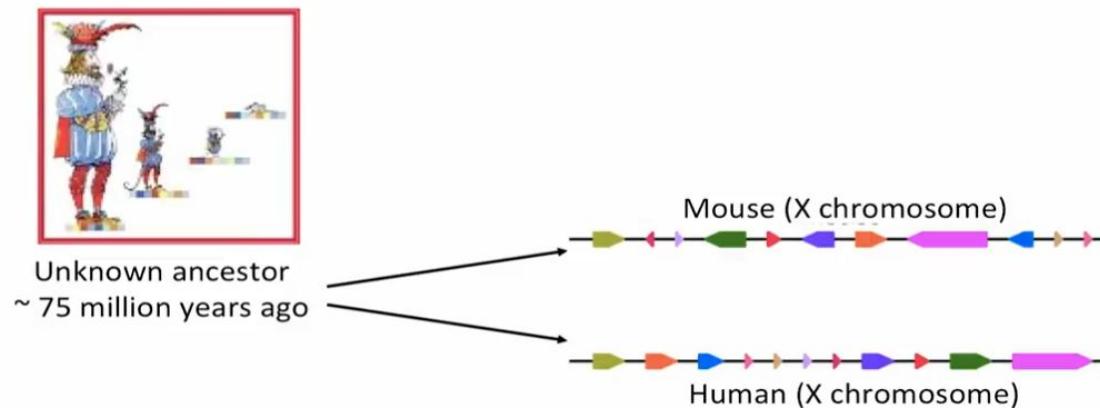
Maybe this algorithm works almost always correctly

# Outline of the lecture

We are going to see a **bioinformatic problem** that we try to solve with **different greedy algorithms** and we are going to **evaluate their goodness** in finding the **best solution**

# Let us see a problem in Biology

## Genome Rearrangements



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

[https://www.youtube.com/watch?v=lCoUp2Bq8OA&list=PLQ-85IQIPqFOcGz6A3g2ZArRL09Ffpp\\_N](https://www.youtube.com/watch?v=lCoUp2Bq8OA&list=PLQ-85IQIPqFOcGz6A3g2ZArRL09Ffpp_N) (until 8:52)

# Reversal Distance Problem

- Goal: Given two permutations, find the shortest series of reversals that transforms one into another
- Input: Permutations  $\pi$  and  $\sigma$
- Output: A series of reversals  $\rho_1, \dots, \rho_t$  transforming  $\pi$  into  $\sigma$ , such that  $t$  is minimum
- $t$  - reversal distance between  $\pi$  and  $\sigma$
- $d(\pi, \sigma)$  - smallest possible value of  $t$ , given  $\pi$  and  $\sigma$

# Sorting By Reversals Problem

- Goal: Given a permutation (i.e., a vector in a random order), find a shortest series of reversals that transforms it into the identity permutation  $(1 \ 2 \ \dots \ n)$
- Input: Permutation  $\pi$
- Output: A series of reversals  $\rho_1, \dots, \rho_t$  transforming  $\pi$  into the identity permutation such that  $t$  is minimum

# Reversals

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

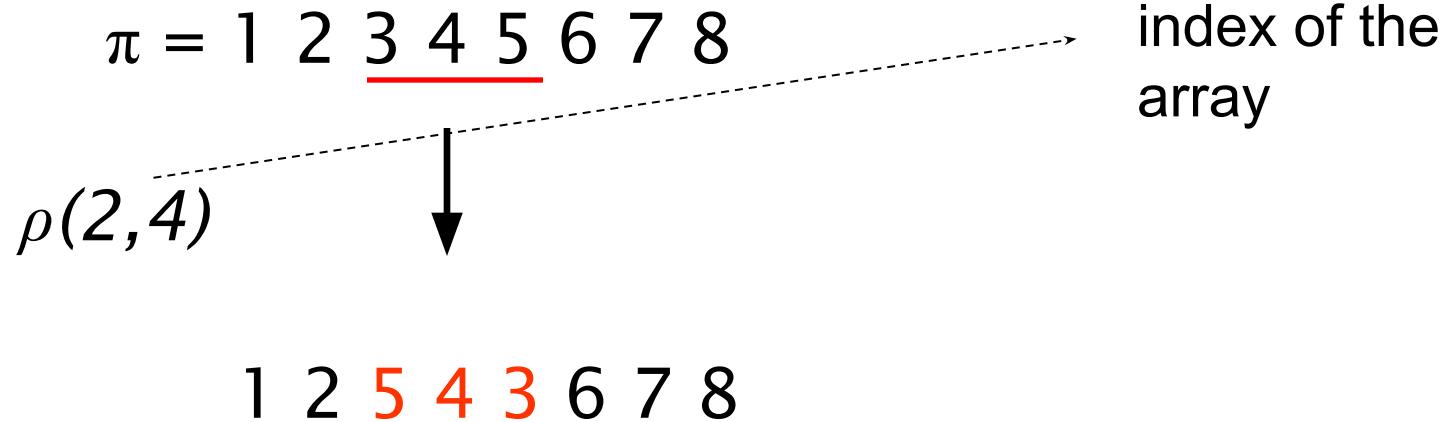
$$\pi = \pi_1 \dots \pi_{i-1} \pi_i \pi_{i+1} \dots \pi_{j-1} \pi_j \pi_{j+1} \dots \pi_n$$

$\rho(i, j)$

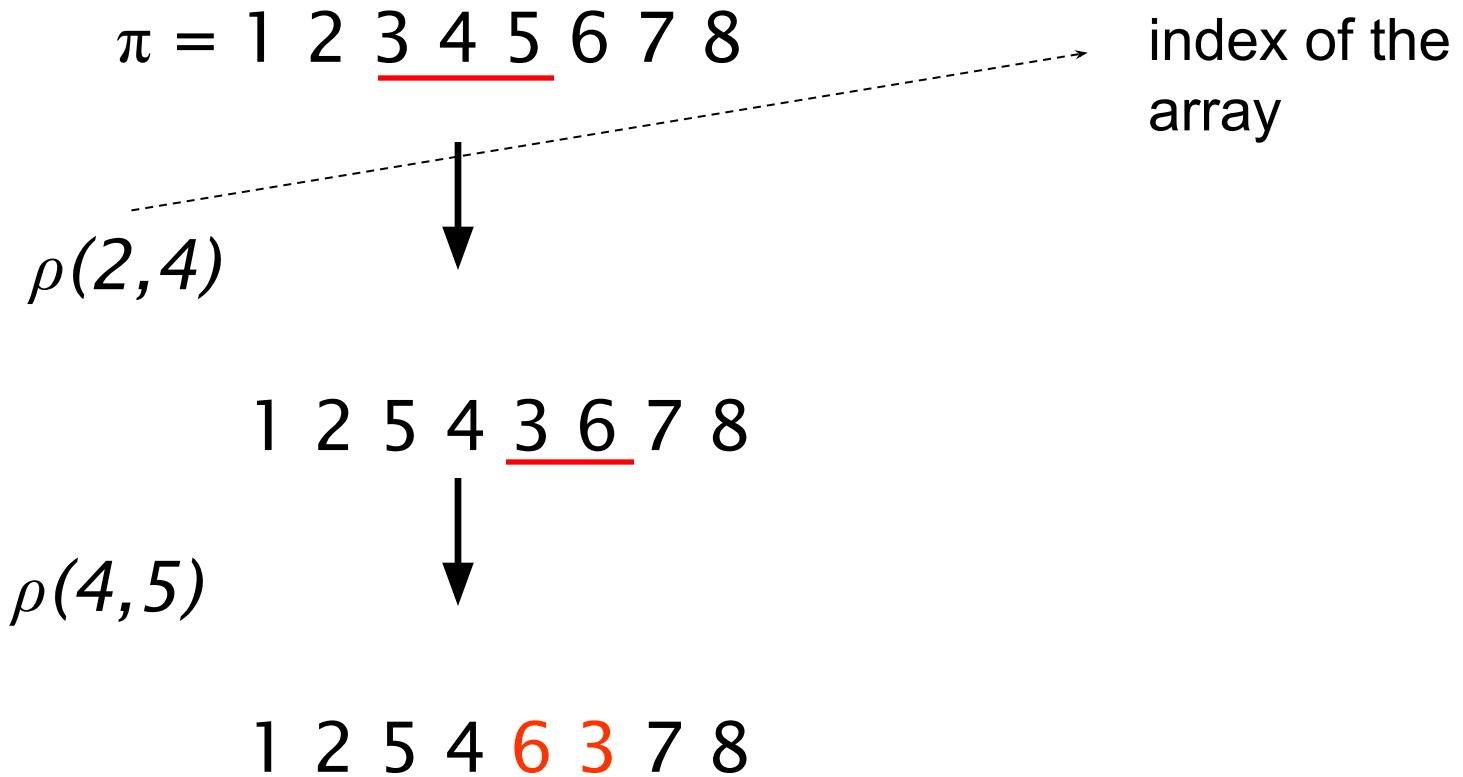


$$\pi_1 \dots \pi_{i-1} \pi_j \pi_{j-1} \dots \pi_{i+1} \pi_i \pi_{j+1} \dots \pi_n$$

# Reversals: Example



# Reversals: Example



# Sorting By Reversals: Example

- $t = d(\pi)$  - reversal distance of  $\pi$
- Example :

$$\begin{array}{rcl} \pi & = & \underline{3 \ 4} \ 2 \ 1 \ 5 \ 6 \ 7 \ 10 \ 9 \ 8 \\ & & 4 \ 3 \ 2 \ 1 \ 5 \ 6 \ 7 \ \underline{10 \ 9 \ 8} \\ & & \underline{4 \ 3 \ 2 \ 1} \ 5 \ 6 \ 7 \ 8 \ 9 \ 10 \\ & & 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7 \ 8 \ 9 \ 10 \end{array}$$

So  $d(\pi) = 3$

# Sorting by reversals: 5 steps

**Step 0:**  $\pi$  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:** 2 3 4 5 6 7 8 -1

**Step 4:** -8 -7 -6 -5 -4 -3 -2 -1

**Step 5:**  $\gamma$  1 2 3 4 5 6 7 8

# Sorting by reversals: 4 steps

**Step 0:**  $\pi$  2 -4 -3 5 -8 -7 -6 1

**Step 1:** 2 3 4 5 -8 -7 -6 1

**Step 2:** -5 -4 -3 -2 -8 -7 -6 1

**Step 3:** -5 -4 -3 -2 -1 6 7 8

**Step 4:**  $\gamma$  1 2 3 4 5 6 7 8

# Sorting by reversals: 4 steps

**Step 0:**  $\pi$  2 -4 -3 5 -8 -7 -6 1

**Step 1:** 2 3 4 5 -8 -7 -6 1

**Step 2:** -5 -4 -3 -2 -8 -7 -6 1

**Step 3:** -5 -4 -3 -2 -1 6 7 8

**Step 4:**  $\gamma$  1 2 3 4 5 6 7 8

What is the reversal distance for this permutation? Can it be sorted in 3 steps?

# Sorting By Reversals: A Greedy Algorithm

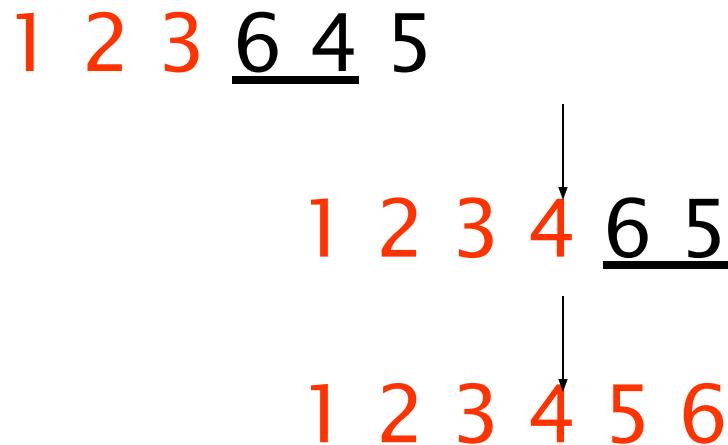
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- If sorting permutation  $\pi = 1 \ 2 \ 3 \ 6 \ 4 \ 5$ , the first three elements are already in order so it does not make any sense to break them.
- The length of the already sorted prefix of  $\pi$  is denoted  $\text{prefix}(\pi)$ 
  - $\text{prefix}(\pi)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

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# Greedy Algorithm: An Example

- Doing so,  $\pi$  can be sorted



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

# Greedy Algorithm: Pseudocode

SimpleReversalSort( $\pi$ )

- 1 **for**  $i \square 1$  to  $n - 1$  *# assuming that elements are from 1 to n*
- 2    $j \square$  position of element  $i$  in  $\pi$  (i.e.,  $\pi_j = i$ )
- 3   **if**  $j \neq i$
- 4      $\pi \square$  we apply  $\rho(i, j)$  on  $\pi$
- 5   **if**  $\pi$  is the identity permutation
- 6   **return**  $\pi$

# Analyzing SimpleReversalSort

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

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

## Analyzing SimpleReversalSort (cont'd)

- But it can be sorted in two steps:

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

- Step 1: 5 4 3 2 1 6
- Step 2: 1 2 3 4 5 6
- So, SimpleReversalSort( $\pi$ ) is not optimal
- But how good is it?

# Analyzing SimpleReversalSort (cont'd)

- But it can be sorted in two steps:

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

- Step 1: 5 4 3 2 1 6
- Step 2: 1 2 3 4 5 6
- So, SimpleReversalSort( $\pi$ ) is not optimal
- But how good is it?
- Optimal algorithms are **unknown** for many problems; approximation algorithms are used

# Approximation Algorithms

- These algorithms find approximate solutions rather than optimal solutions
- The approximation ratio of an algorithm A on the problem with input  $\pi$  is:

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

where

$A(\pi)$  - solution produced by algorithm A

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

- + (in our case,  $\pi$  is an instance of the reversal sorting problem)

# Approximation Algorithms

- If an algorithm has an approximation ratio = 1.5, it means that the solution it finds is never more than 150% of the optimal one.
  - For example, if the minimum sorting requires 10 reversals, the approx algorithm will use at most 15.

# Approximation Ratio

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- For algorithm A that minimizes objective function (minimization algorithm):
  - $\max_{|\pi| = n} A(\pi) / \text{OPT}(\pi)$
- For maximization algorithm:
  - $\min_{|\pi| = n} A(\pi) / \text{OPT}(\pi)$

Can we do better than  
SimpleReversalSort( $\pi$ )?

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Yes

Sometimes we need to characterize  
better the problem to make more  
sophisticated techniques

# DISCLAIMER: DON'T BE AFRAID

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You are requested to understand the general idea of greedy not how to elaborate this more sophisticated techniques (right now..)

# Adjacencies and Breakpoints

$$\pi = \pi_1 \pi_2 \pi_3 \dots \pi_{n-1} \pi_n$$

- A pair of elements  $\pi_i$  and  $\pi_{i+1}$  are **adjacent** if

$$\pi_{i+1} = \pi_i \pm 1$$

- For example:

$$\pi = 1 \ 9 \ 3 \ 4 \ \underline{7 \ 8} \ 2 \ \underline{6 \ 5}$$

- (3, 4) or (7, 8) and (6,5) are adjacent pairs

# Breakpoints: An Example

There is a **breakpoint** between any adjacent element that are non-consecutive:

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

- Pairs  $(1,9)$ ,  $(9,3)$ ,  $(4,7)$ ,  $(8,2)$  and  $(2,5)$  form breakpoints of permutation  $\pi$

# Breakpoints: An Example

There is a **breakpoint** between any adjacent element that are non-consecutive:

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

- Pairs  $(1,9)$ ,  $(9,3)$ ,  $(4,7)$ ,  $(8,2)$  and  $(2,5)$  form breakpoints of permutation  $\pi$
- $b(\pi)$  - # breakpoints in permutation  $\pi$

# Extending Permutations

- We want to ensure that also the first and the last element are in the **right positions**. To do that...
- We put two elements  $\pi_0 = 0$  and  $\pi_{n+1} = n+1$  at the ends of  $\pi$

Example:

$$\begin{array}{cccccccccc} \pi = 1 & | 9 & | 3 & 4 & | 7 & 8 & 2 & | 6 & 5 \\ & \downarrow & & & & & & & \\ & & & & & & & & \\ \pi = 0 & 1 & | 9 & | 3 & 4 & | 7 & 8 & | 2 & 6 & 5 & | 10 \end{array}$$

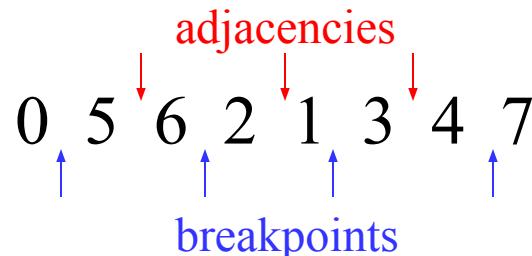
Extending with 0 and 10

Note: A new breakpoint was created after extending

# Sum up: Adjacency & Breakpoints

- An **adjacency** - a pair of adjacent elements that are **consecutive**
- A **breakpoint** - a pair of adjacent elements that are **not consecutive**

$\pi = 5 \ 6 \ 2 \ 1 \ 3 \ 4$   $\longrightarrow$  Extend  $\pi$  with  $\pi_0 = 0$  and  $\pi_7 = 7$



# Reversal Distance and Breakpoints

- Observation: each reversal eliminates at most 2 breakpoints.

$$\pi = |2 \ 3 | 1 \ 4 \ 6 \ 5 |$$

$$0 \ \underline{2} | 3 \ 1 | 4 | 6 \ 5 | 7$$

$$b(\pi) = 5$$

$$0 \ 1 \ \underline{3 \ 2} \ 4 | 6 \ 5 | 7$$

$$b(\pi) = 4$$

$$0 \ 1 \ 2 \ 3 \ 4 \ \underline{6 \ 5} \ 7$$

$$b(\pi) = 2$$

$$0 \ 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7$$

$$b(\pi) = 0$$

# Reversal Distance and Breakpoints

- Observation: each reversal eliminates at most 2 breakpoints.
- This implies:

$$\text{reversal distance} \geq \#\text{breakpoints} / 2$$

$\pi =  2 \ 3   1 \ 4 \   6 \ 5  $	
$0 \ 2   3 \ 1   4   6 \ 5   7$	$b(\pi) = 5$
$0 \ 1 \ 3 \ 2 \ 4   6 \ 5   7$	$b(\pi) = 4$
$0 \ 1 \ 2 \ 3 \ 4 \ 6 \ 5 \ 7$	$b(\pi) = 2$
$0 \ 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7$	$b(\pi) = 0$

# Sorting By Reversals: A Better Greedy Algorithm

---

## BreakPointReversalSort( $\pi$ )

1 **while**  $b(\pi) > 0$

2     Among all possible reversals,  
choose reversal  $\rho$  minimizing  $b(\pi)$  after  
its application

3      $\pi \square$  apply  $\rho(i, j)$  on  $\pi$

4 **return**  $\pi$

---

# Sorting By Reversals: A Better Greedy Algorithm

---

## BreakPointReversalSort( $\pi$ )

1 **while**  $b(\pi) > 0$

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3      $\pi \square$  apply  $\rho(i, j)$  on  $\pi$

4 **return**  $\pi$

Problem: this algorithm may work forever (we cannot reduce the number of breakpoints anymore)

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Can we do better than  
BreakPointReversalSort( $\pi$ )?

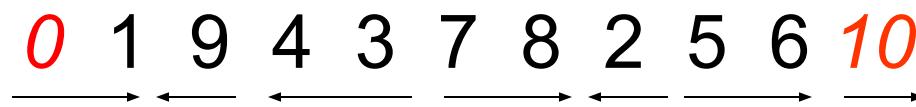
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Yes

We need to characterize even better  
the problem to exploit further aspects

# Strips

- Strip: an interval between two consecutive breakpoints in a permutation
  - Decreasing strip: *strip* of elements in decreasing order (e.g. 6 5 and 3 2 ).
  - Increasing strip: *strip* of elements in increasing order (e.g. 7 8)



- A single-element strip can be declared either increasing or decreasing. We will choose to declare them as decreasing with exception of the strips with 0 and  $n+1$

# Reducing the Number of Breakpoints

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Observation 1:

If permutation  $\pi$  contains at least one decreasing strip, then there exists a reversal  $\rho$  which decreases the number of breakpoints (i.e.  $b(\pi)$  after  $\rho$   $<$   $b(\pi)$  )

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# Things To Consider

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

$$0 \ 1 \ 4 \ 6 \ 5 \ 7 \ 8 \ 3 \ 2 \ 9 \quad b(\pi) = 5$$

- Choose decreasing strip with the smallest element  $k$  in  $\pi$  (  $k = 2$  in this case)

## Things To Consider (cont'd)

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

$$0 \ 1 \ 4 \ 6 \ 5 \ 7 \ 8 \ 3 \ 2 \ 9 \quad b(\pi) = 5$$

- Choose decreasing strip with the smallest element  $k$  in  $\pi$  (  $k = 2$  in this case)

## Things To Consider (cont'd)

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

$$0 \ 1 \mid 4 \mid 6 \ 5 \mid 7 \ 8 \mid 3 \ 2 \mid 9 \quad b(\pi) = 5$$

- Choose decreasing strip with the smallest element  $k$  in  $\pi$  ( $k = 2$  in this case)
- Find  $k - 1$  in the permutation

# Things To Consider (cont'd)

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

$$0 \ 1 \ 4 \ 6 \ 5 \ 7 \ 8 \ 3 \ 2 \ 9 \quad b(\pi) = 5$$

- Choose decreasing strip with the smallest element  $k$  in  $\pi$  ( $k = 2$  in this case)
- Find  $k - 1$  in the permutation
- Reverse the segment between  $k$  and  $k-1$ :

- $0 \ 1 \ 4 \ 6 \ 5 \ 7 \ 8 \ 3 \ 2 \ 9 \quad b(\pi) = 5$



- $0 \ 1 \ 2 \ 3 \ 8 \ 7 \ 5 \ 6 \ 4 \ 9 \quad b(\pi) = 4$

# Reducing the Number of Breakpoints Again

- If there is no decreasing strip, there may be no reversal  $\rho$  that reduces the number of breakpoints (i.e.  $b(\pi)$  after  $\rho \geq b(\pi)$  for any reversal  $\rho$ ).
- By reversing an increasing strip ( # of breakpoints stay unchanged ), we will create a decreasing strip at the next step. Then the number of breakpoints will be reduced in the next step (observation 1).

# Things To Consider (cont'd)

- There are no decreasing strips in  $\pi$ , for:

$$\begin{array}{rcl} \pi & = & 0 \ 1 \ 2 \ | \ 5 \ 6 \ 7 \ | \ 3 \ 4 \ | \ 8 \quad b(\pi) = 3 \\ \rho(6,7) \text{ on } \pi & = & \overbrace{0 \ 1 \ 2 \ | \ 5 \ 6 \ 7} \quad | \ 4 \ 3 \ | \ 8 \quad b(\pi) = 3 \end{array}$$

- ✓  $\rho(6,7)$  does not change the # of breakpoints
- ✓  $\rho(6,7)$  creates a decreasing strip thus guaranteeing that the next step will decrease the # of breakpoints by doing  $\rho(6,2)$ .

# ImprovedBreakpointReversalSort

ImprovedBreakpointReversalSort( $\pi$ )

- 1 **while**  $b(\pi) > 0$
- 2   **if**  $\pi$  has a decreasing strip
- 3       Among all possible reversals, choose reversal  $\rho$   
          that minimizes  $b(\pi)$  *after*  $\rho$
- 4   **else**
- 5       Choose a reversal  $\rho$  that flips an increasing strip in  $\pi$
- 6        $\pi \square$  *apply*  $\rho$  *on*  $\pi$
- 7 **return**  $\pi$

# ImprovedBreakpointReversalSort: performance?

- *ImprovedBreakPointReversalSort is the optimal solution?*

# ImprovedBreakpointReversalSort: performance?

- *ImprovedBreakPointReversalSort is the optimal solution? Unfortunately, no..*
- *ImprovedBreakPointReversalSort* is an approximation algorithm
  - Optimal algorithm eliminates at most 2 breakpoints in every step:  $d(\pi) \geq b(\pi) / 2$
  - It eliminates at least one breakpoint in every two steps; at most  $2b(\pi)$  steps
  - Approximation ratio:  $2b(\pi) / d(\pi)$

# ImprovedBreakpointReversalSort: performance?

- *ImprovedBreakPointReversalSort is the optimal solution? Unfortunately, no..*
- *ImprovedBreakPointReversalSort* is an approximation algorithm
  - Approximation ratio:  $2b(\pi) / d(\pi)$
- NOTE: we can compute  $d(\pi)$  for a specific instance of  $\pi$  but not for all the instance of sorting by reversal problem, in this way we can compute the approximation ratio

# Take home messages

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When should we use Greedy Algorithms?

# When should we use Greedy Algorithms?

- **Simple and easy to understand**

we follow a simple idea: for every subproblem, a greedy algorithm tries to find the best optimal solution (e.g., in ImprovedBreakpointReversalSort, choose reversal  $\rho$  that minimizes  $b(\pi)$  after  $\rho$ )

# When should we use Greedy Algorithms?

- **Simple and easy to understand**

we follow a simple idea: for every subproblem, a greedy algorithm tries to find the best optimal solution (e.g., in ImprovedBreakpointReversalSort, choose reversal  $\rho$  that minimizes  $b(\pi)$  after  $\rho$ )

**This is also the limit:** following a simple rule (or rules) for **all** the subproblems might lead to **naive solutions** (i.e., solutions that consider the characteristics of every subproblem missing details)

# When should we use Greedy Algorithms?

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- **Can be used as a building block for other algorithms:** it can be used as a starting point for developing more complex algorithms.

We started with SimpleReversalSort

We improve it with BreakPointReversalSort

We improve it with ImprovedBreakpointReversalSort

# When should we use Greedy Algorithms?

- **Fast and efficient** (compared to other techniques)

example: SimpleReversalSort runs in  $O(n)$  ( $n$  size of the vector  $\pi$ )

# When should we use Greedy Algorithms?

- **Provides a good enough solution** (we have seen how good today for one problem)

ImprovedBreakPointReversalSort still finding acceptable **approximate** solutions (not the best ones)

We still don't know an **efficient algorithm** computing the **optimal solution** for sorting by reversal