



# SEARCH MOVES IN THE LOCAL OPTIMA NETWORKS OF PERMUTATION SPACES

## THE QAP CASE

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# MOTIVATIONS AND GOAL

- One of the achieved objectives of Fitness Landscape Analysis (FLA) is: “estimate **how many search moves** need to be performed in order to escape an attraction basin” ...
- ... but FLA does not identify **which moves** have to be performed!
- Our goal: **identify the “escaping moves”**

# OUTLINE

- Quadratic Assignment Problem
- Local Optima Network
- Algebraic framework for Evolutionary Computation
- Qualitative analysis of the “escaping moves”
- Future lines of research

# QUADRATIC ASSIGNMENT PROBLEM (QAP)

- There are  $n$  factories and  $n$  cities.
- A distance  $a_{ij}$  is specified for each pair of cities.
- A flow  $b_{ij}$  is specified for each pair of factories.
- The problem is to assign all factories to different cities with the goal of minimizing the sum of the distances multiplied by the corresponding flows.

$$C(\pi) = \sum_{i=1}^n \sum_{j=1}^n a_{ij} b_{\pi_i \pi_j}$$

# FITNESS LANDSCAPE (FL)

A FL is a triplet  $(X, N, f)$  where:

- $X$  is the **set of solutions**  
(all the permutations of  $n$  items in QAP)
- $N$  is a **neighborhood structure** among the solutions  
(exchange neighborhood in QAP)
- $f$  is a **fitness function**  
(the QAP objective function)

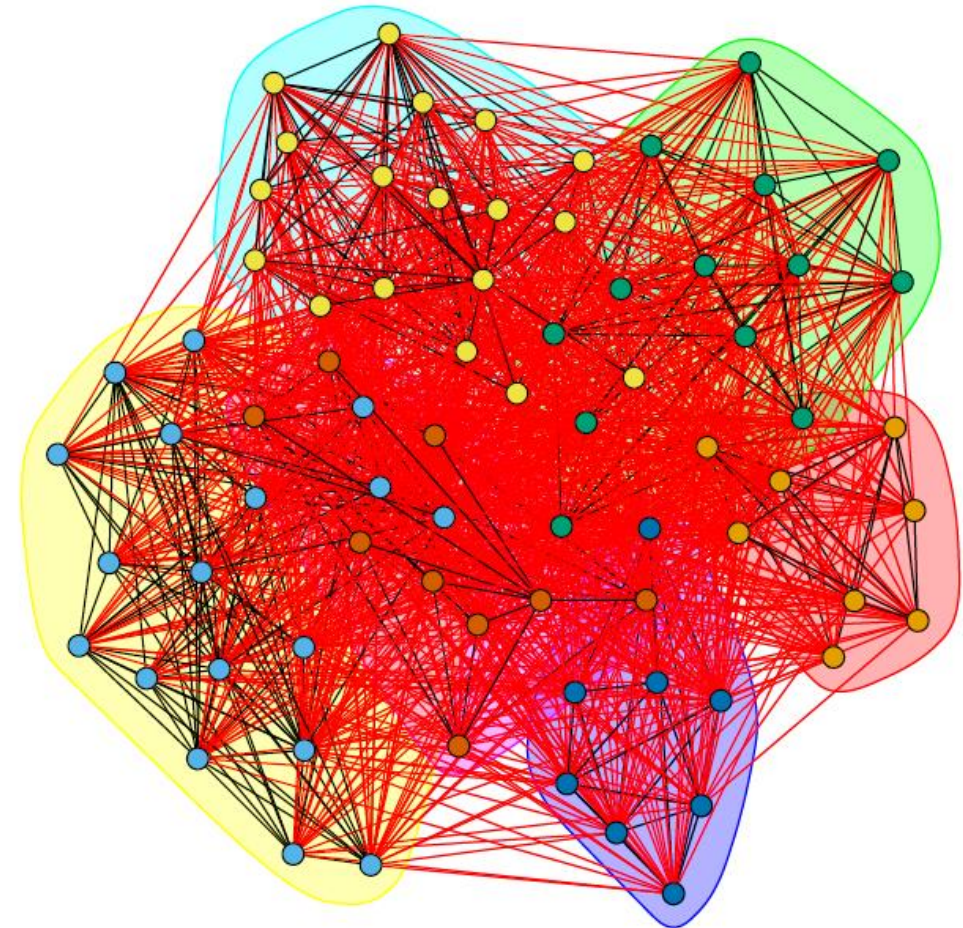
# LOCAL OPTIMA NETWORK (LON)

A LON is a **graph** extracted from a given FL by using a (best-improvement) hill-climber  $hc$  where:

- the **nodes are the local optima** of the given FL
- there is an **(escape) edge**  $e_{ij}$  between  $LO_i$  and  $LO_j$  if a solution  $x$  exists such that  $dist(x, LO_i) \leq D$  and  $hc(x) = LO_j$
- the **edge  $e_{ij}$  has weight**  $w_{ij} = v_{ij} / \sum_i v_{ij}$  where  $v_{ij} = \#\{ x \in X \mid dist(x, LO_i) \leq D \text{ and } hc(x) = LO_j \}$

# COMMUNITIES OF OPTIMA IN LONs

- LONs are complex networks which can be studied with methods of network science
- LONs can have a clustered structure, thus **the local optima can be divided in communities**



# ALGEBRAIC STRUCTURE OF THE PERMUTATION SPACE

□ A permutation of  $[n] = \{1, 2, \dots, n\}$  is a bijective discrete function from  $[n]$  to  $[n]$ , thus it is possible to **invert** and **compose permutations**:

$$\sigma = \pi \circ \rho \text{ iff } \sigma(i) = \pi(\rho(i)) \text{ for } 1 \leq i \leq n$$

□ Permutations of  $[n]$  form the **symmetric group**  $\mathcal{S}(n)$

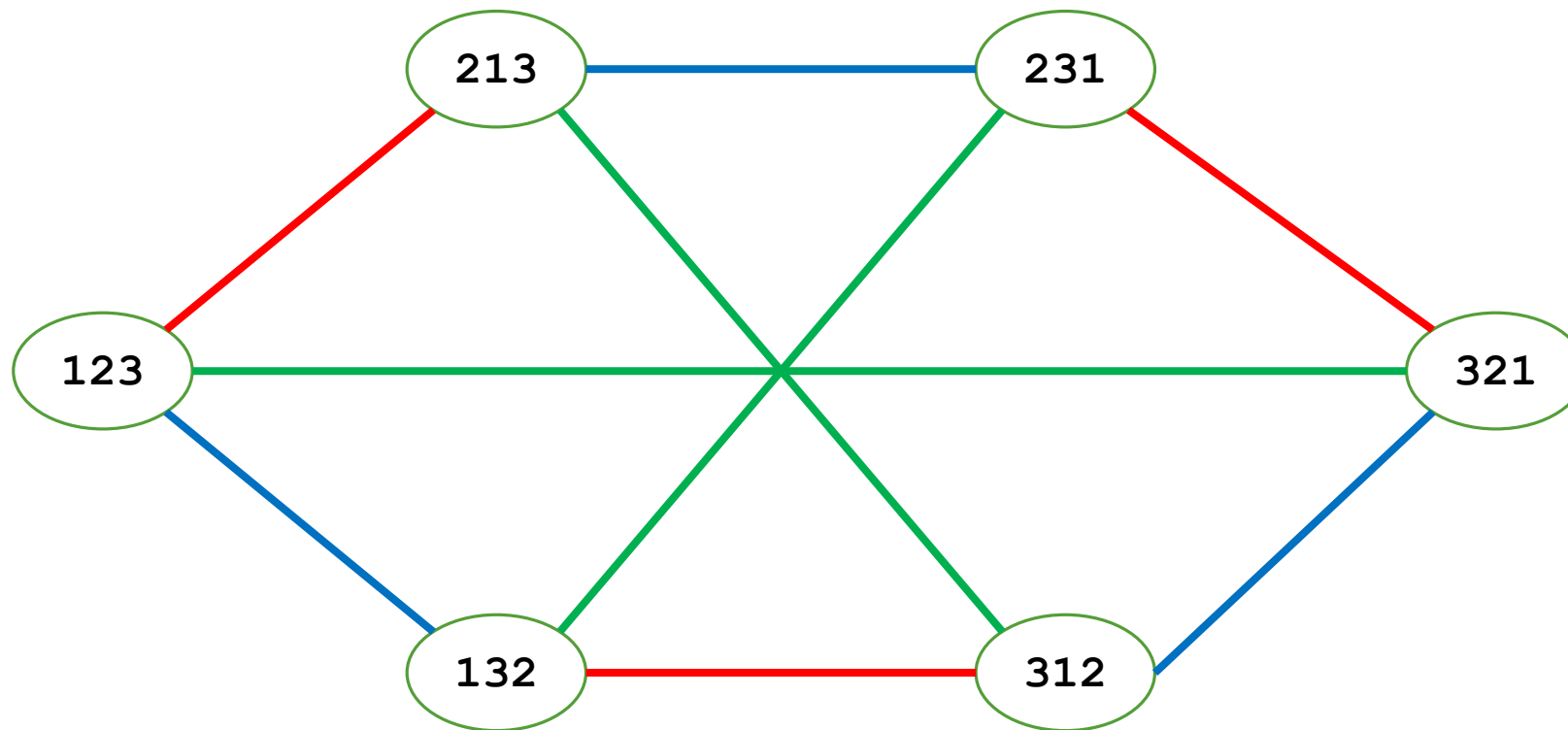
□  $\mathcal{S}(n)$  is **finitely generated by the exchange permutations**  $\epsilon_{ij}$  s.t.

$$\epsilon_{ij}(k) = \begin{cases} k & \text{if } k \neq i \text{ and } k \neq j \\ j & \text{if } k = i \\ i & \text{if } k = j \end{cases}$$

□ Given any  $\pi \in \mathcal{S}(n)$ ,  $\pi \circ \epsilon_{ij}$  corresponds to **exchanging the items at positions  $i$  and  $j$  in the permutation  $\pi$**



# THE CAYLEY GRAPH



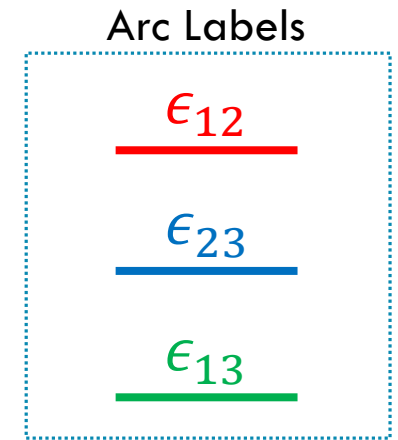
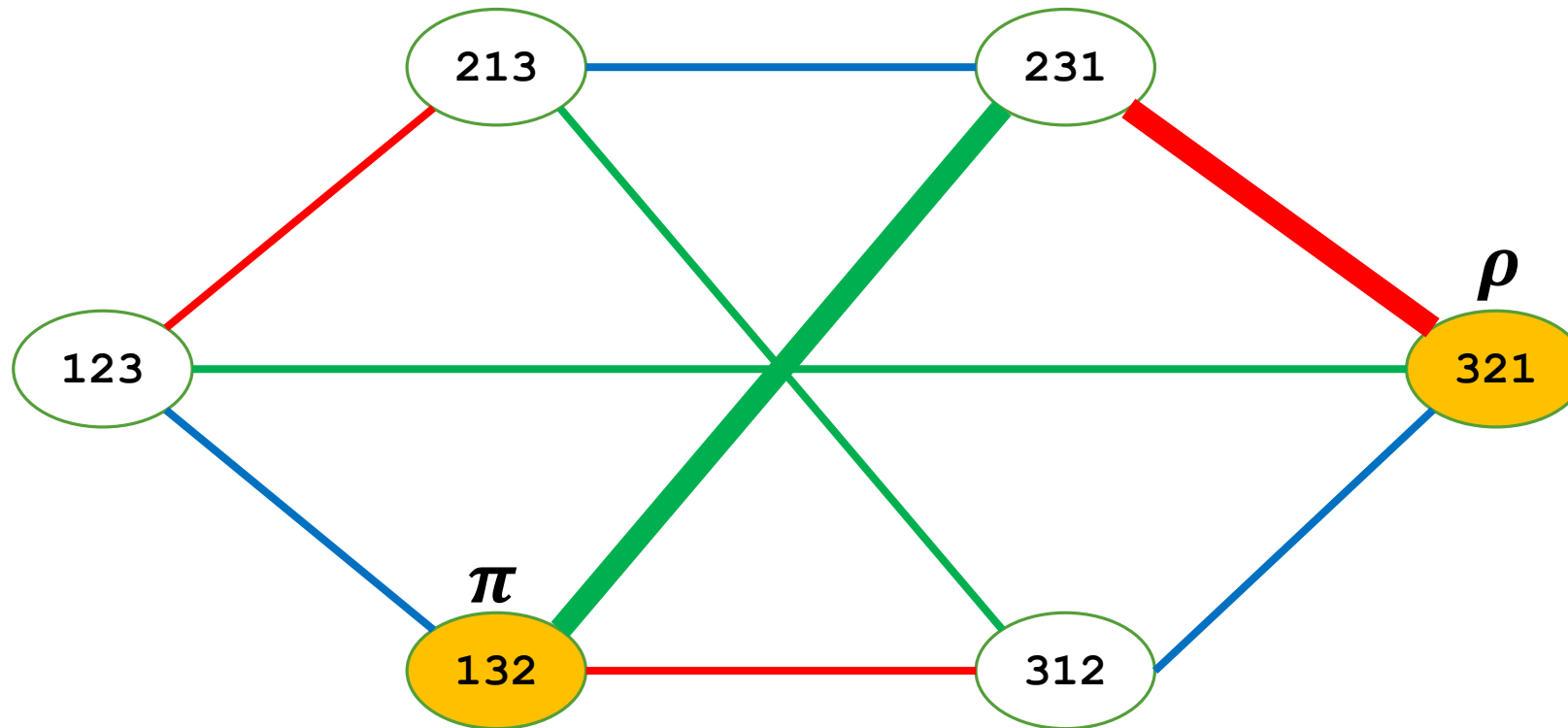
Arc Labels

$\epsilon_{12}$

$\epsilon_{23}$

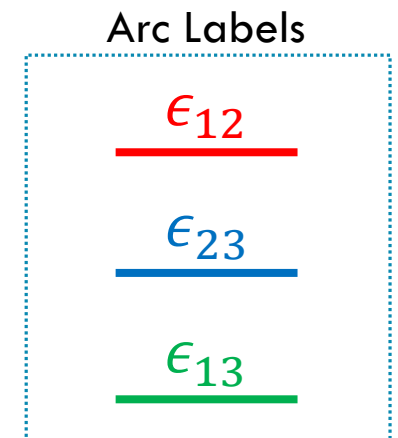
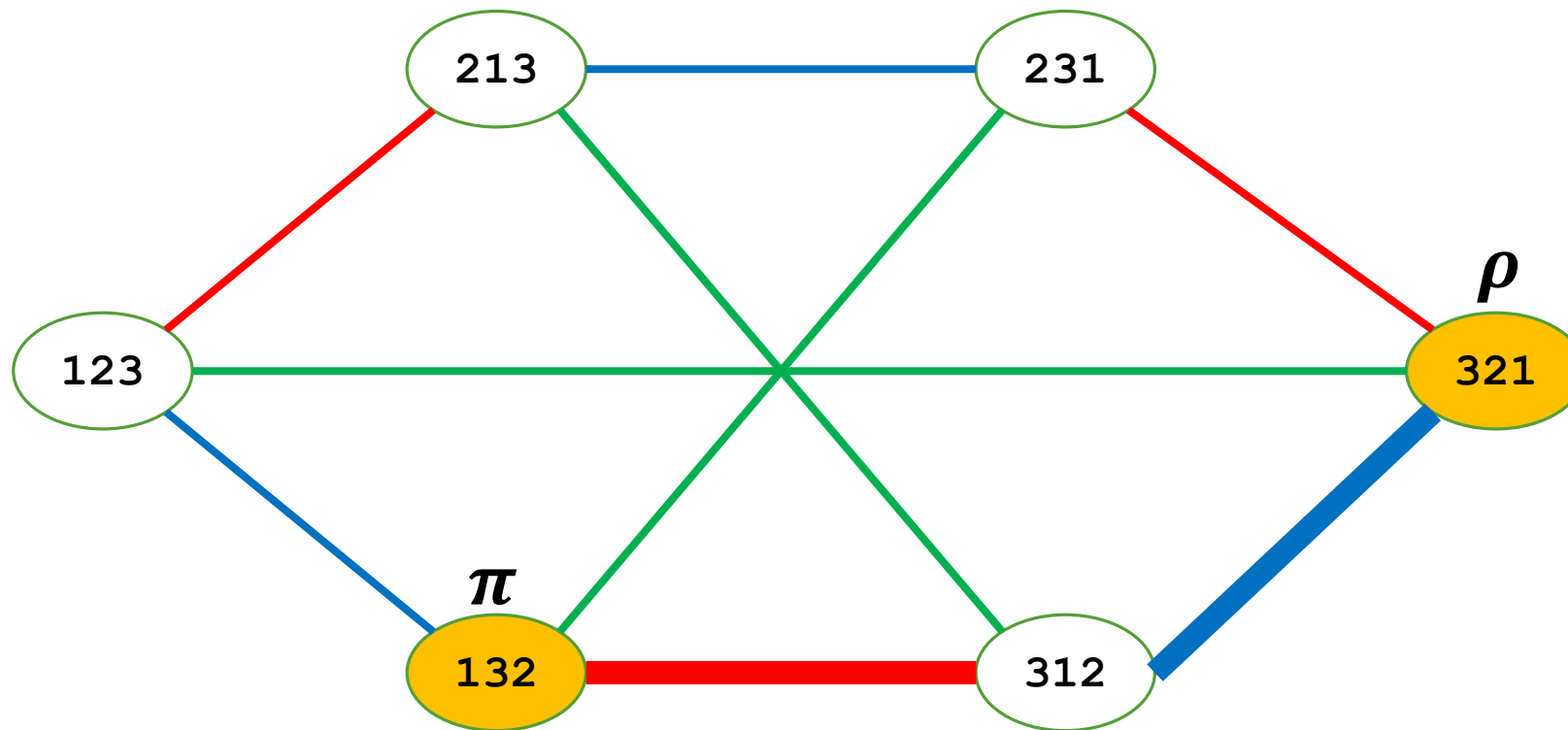
$\epsilon_{13}$

# DIFFERENCES BETWEEN PERMUTATIONS



$$\rho \ominus \pi = \epsilon_{13} \circ \epsilon_{12}$$

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$$\rho \ominus \pi = \epsilon_{13} \circ \epsilon_{12} = \epsilon_{12} \circ \epsilon_{23}$$

# DIFFERENCES BETWEEN PERMUTATIONS

□ All the paths connecting  $\pi$  to  $\rho$  in the Cayley graph are all the possible factorizations of  $\rho \ominus \pi$

□ Given any pair of permutations  $\pi, \rho$  their difference is

$$\rho \ominus \pi = \pi^{-1} \circ \rho$$

□ ... but the factorizations of  $\rho \ominus \pi$  indicate the sequences of pairs of positions to exchange ...

□ ... while we want the sequences of pairs of items to exchange!!!

# ITEMS TO EXCHANGE (AND NOT POSITIONS!!!)

□ A permutation is a **bijection from positions to items**

□ We can **exchange two generic items  $i$  and  $j$  from  $\pi$**  as follows

$$\left(\pi^{-1} \circ \epsilon_{ij}\right)^{-1} = \epsilon_{ij} \circ \pi$$

□ The **sequences of pairs of items to be exchanged for moving from  $\pi$  to  $\rho$**  correspond to all the possible factorizations of

$$\pi^{-1} \ominus \rho^{-1} = \rho \circ \pi^{-1}$$

# COMPACT REPRESENTATION OF THE EXCHANGES

- Given two permutations, the **number of alternative (shortest) paths connecting them is exponential in their distance**
- We want to identify the pairs of items to exchange **independently of where they appear in the factorizations**
- We use the **cycle decomposition of a permutation**

$$\begin{array}{c} 12345678 \\ \langle 26745831 \rangle \end{array} \longrightarrow (1268)(37)(4)(5)$$

$(1, 2), (1, 6), (1, 8), (2, 6), (2, 8), (6, 8), (3, 7)$

# WEIGH EXCHANGES BY IMPORTANCE (1/2)

- Given two permutations, the exchanges appearing in multiple (shortest) paths between them are more important
- If the two permutations are local optima, an exchange appearing in a large number of (shortest) paths connecting them is more useful for escaping a basin of attraction
- Let consider that a factorization in terms of exchanges can be obtained by iteratively exchanging two items belonging to the same cycle:
  - The cycle breaks into two (smaller) cycles
  - The identity permutation is the only one with  $n$  cycles (of length 1)

# WEIGH EXCHANGES BY IMPORTANCE (2/2)

- ❑ Pair of items in shorter cycles (w.r.t. all the other cycles) appear in a large number of factorizations
- ❑ Closer are two items in a cycle, more are the factorizations where they appear
- ❑ Some approximated formulae and tabulations in the paper (obtained by considering a recursive variant of our factorization algorithm)



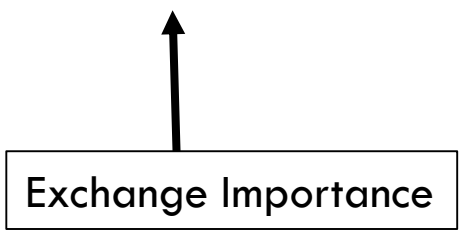
# THE EXPERIMENTAL ANALYSIS

- ❑ **LONs** of few QAP real-like instances (**thanks to Sebastien Verel**)
- ❑ **Clustered** by means of two community finding algorithms (R package *igraph*)
- ❑ **Intra-Community Analysis**: are there more relevant exchanges for moving between local optima of the **same community**?
- ❑ **Inter-Community Analysis**: are there more relevant exchanges for moving between local optima of **different communities**?

# THE ANALYZEEXCHANGES ALGORITHM

**Algorithm 2** Computation of the importance of the exchanges in order to move among a set of permutations

```
1: function ANALYZEEXCHANGES( $S \subset \mathcal{S}_n \times \mathcal{S}_n$ )
2:    $Z_{ij} \leftarrow 0$  for all  $1 < i < j \leq n$ 
3:   for all pairs of permutations  $\pi, \rho \in S$  do
4:     Compute the cycles decomposition of  $\pi^{-1} \ominus \rho^{-1}$ 
5:     for all cycles  $c$  of  $\pi^{-1} \ominus \rho^{-1}$  do
6:       for all pairs of items  $i, j \in c$  do
7:          $Z_{ij} \leftarrow Z_{ij} + Q_k(\epsilon_{ij})$ 
8:       end for
9:     end for
10:  end for
11:  return  $Z$ 
12: end function
```



## □ INPUT:

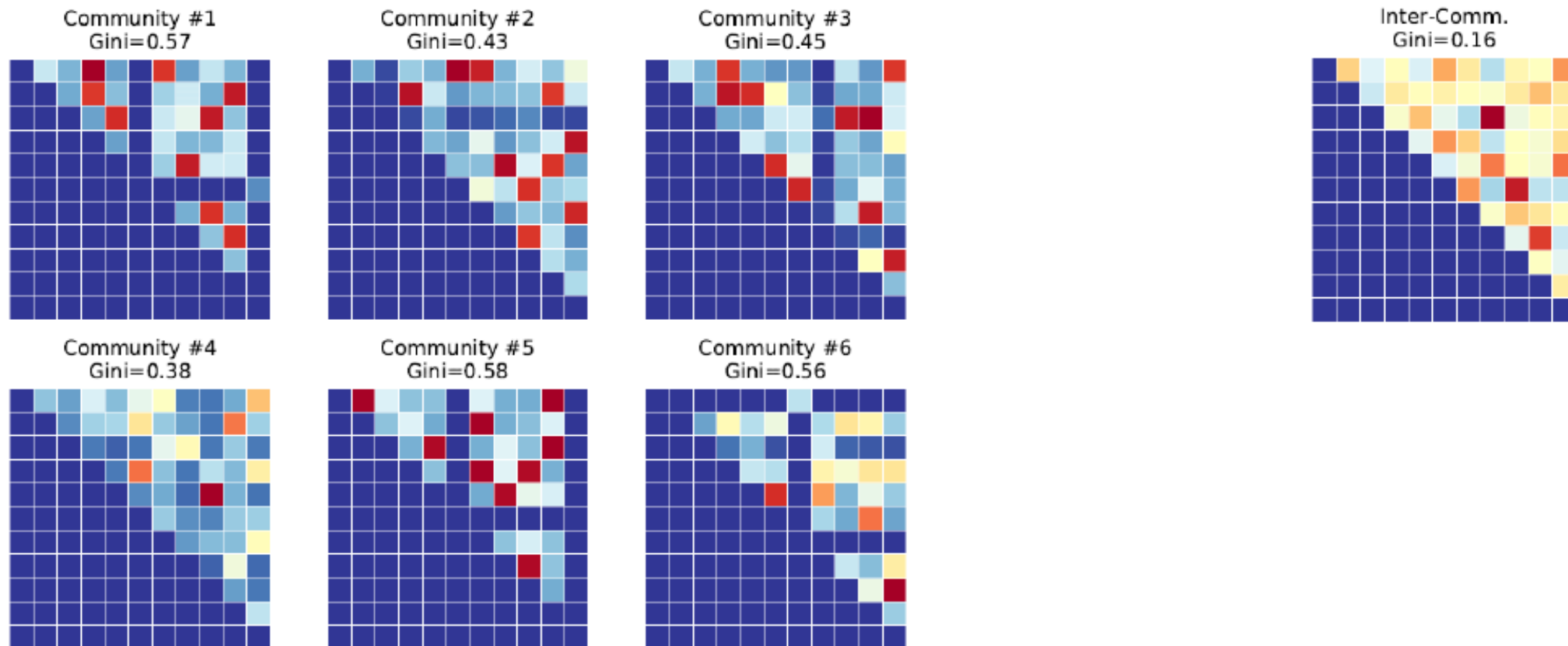
- (Intra-Comm. An.) set of local optima in a same community
- (Inter-Comm. An.) set of local optima in different communities

□ OUTPUT: a (triangular) matrix such that  $Z_{ij}$  measures the relevance of  $\epsilon_{ij}$  as escaping move

# HEATMAPS FOR KCSo11RL-1 (WALKTRAP)

## Intra-Community Analysis

## Inter-Community Analysis



# GINI INDEXES ON THE Z-VALUES

- The Gini Index is a measure of statistical dispersion (popular in economy)
  - 0 on uniform distributions, 1 on degenerate distributions
  - in our scenario, 1 is impossible (due to the constraints among permutation items)
  - 0.5 has been empirically observed to produce a «concentrated» distribution

Table 2: Gini Indexes for every clustered LON

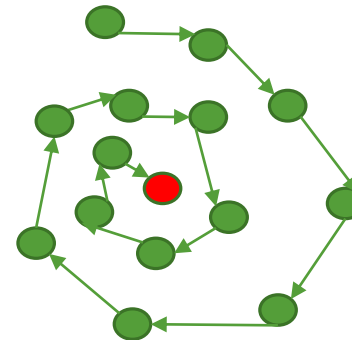
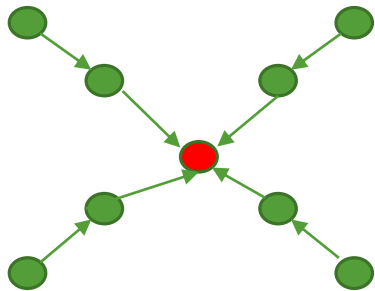
Instance	Community Finding Algorithm	#Communities	Intra-Community Gini Index	Inter-Community Gini Index
KCso11rl-1	FastGreedy	6	$0.56 \pm 0.14$	0.16
KCso11rl-1	WalkTrap	6	$0.50 \pm 0.09$	0.16
KCso11rl-2	FastGreedy	6	$0.46 \pm 0.11$	0.12
KCso11rl-2	WalkTrap	2	$0.26 \pm 0.15$	0.14
KCso11rl-3	FastGreedy	8	$0.44 \pm 0.10$	0.18
KCso11rl-3	WalkTrap	6	$0.41 \pm 0.13$	0.17
KCso11rl-4	FastGreedy	5	$0.57 \pm 0.08$	0.16
KCso11rl-5	WalkTrap	5	$0.58 \pm 0.08$	0.16

# CONCLUSIONS

- ❑ Real-like QAP instances look to have «preferred» search moves that allow to move across basins of attraction belonging to the same community
- ❑ The same does not look to be true for basins of attraction belonging to different communities
- ❑ This analysis shows that the **Algebraic Framework for EC can be useful for fitness landscape analyses**

# FUTURE WORKS

- ❑ Experiment with larger QAP instances and sampled LONs
- ❑ Consider other permutation problems
- ❑ Other applications of the Algebraic Framework to FLA:  
«Vorticity» index to discern the following type of basin of attractions



**THANKS FOR YOUR ATTENTION!**