

Algorithms for computing the Family-Free Genomic Similarity under DCJ

Università di Pisa

Course: *Bioinformatics*

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Outline

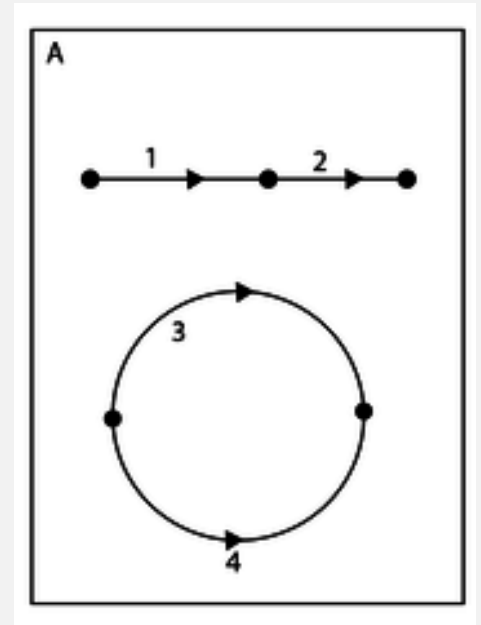
- Introduction
- ILP formulation
- Faster heuristics
- Experimental results
- Conclusions

Introduction



Introduction: DCJ model

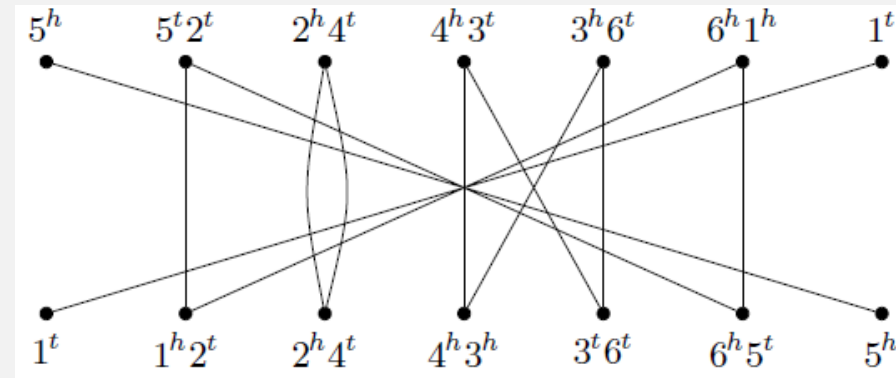
- **Genes** represented by a **number** and an **orientation**
- **Global measure** for distance: number of genome rearrangements
- **Double cut and join op.:** *cut* a genome in two distinct positions and joining the four resultant open ends
- **Similarity:** more rearrangements needed to get equal genomes, less similarity



Introduction: Genes and Adjacency Graph

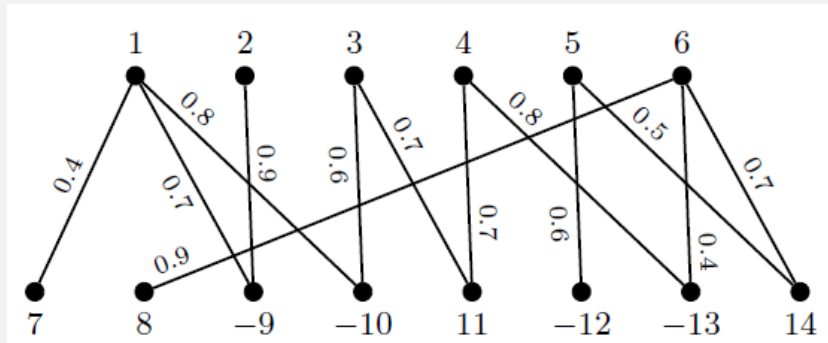
- Each **gene** g of a genome has two extremities (*tail* and *head*): g^t and g^h
- **Adjacency**: Pair of *consecutive* gene extremities or extremity of a gene adjacent to a telomere
- **Telomere**: extremity of a linear chromosome

- **Adjacency Graph**:

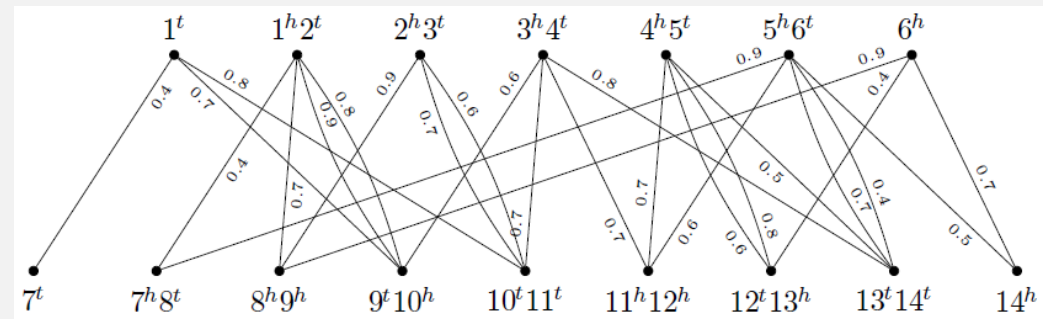


Introduction: Weighted Adjacency Graph

- **Family-Free** setting: genes represented by a unique (signed) symbol
- **No assumptions** on genes families
- **Distinct** genes: *normalized gene similarity* $[0,1]$
- With this setting vertices can have **degree greater than 2**



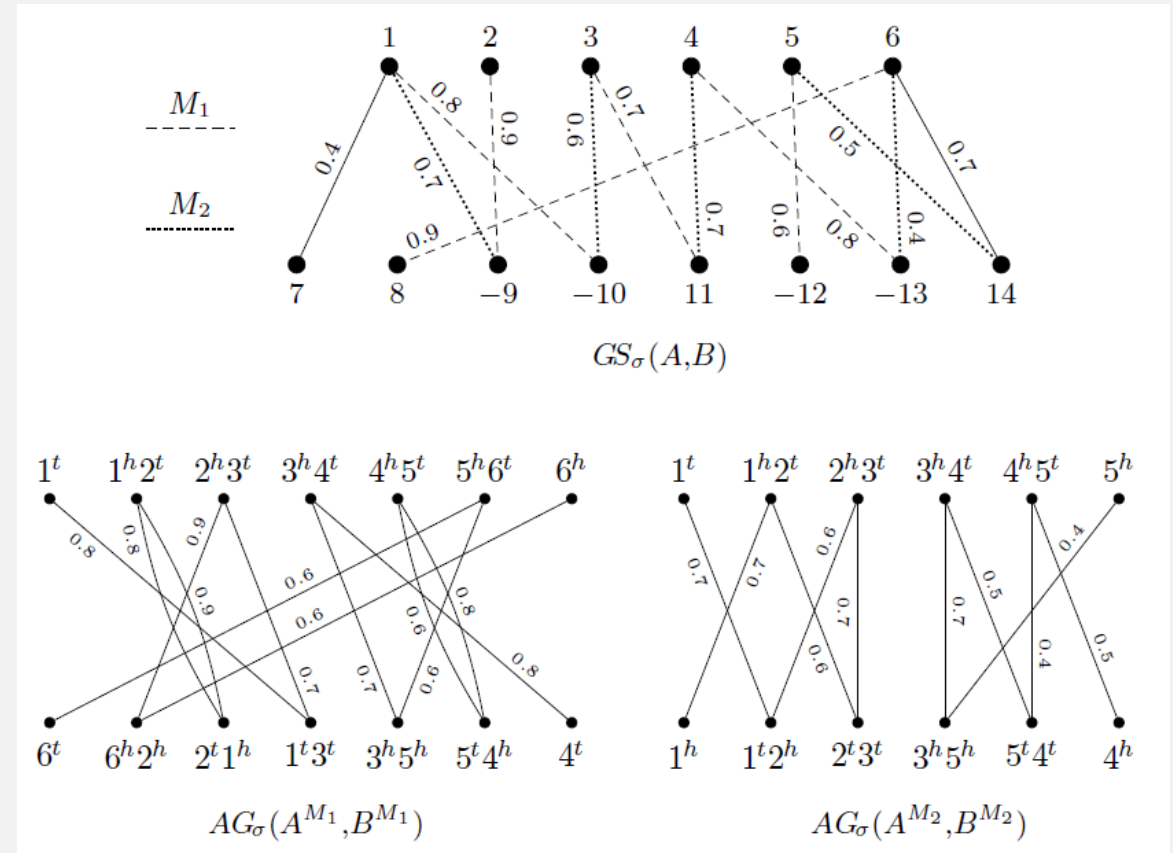
Gene similarity graph



Weighted adjacency graph

Introduction: Matchings

- **Matching:** set of edges without common vertices
- **Maximal matching** (A^M, B^M) : if any edge not in M is added to M , it is no longer a matching
- **Decomposition** in components is simplified
- For each gene what is the **best connection** with the other genome?



Introduction: FFDCJ Similarity

- **FFDCJ-SIM(A, B)**: given genomes A and B and their gene sim. σ , calculate their family-free DCJ similarity:

$$s_{FFDCJ}(A, B) = \max_{M \in M} \{s_{\sigma}(A^M, B^M)\}$$

where $M = \text{set of all maximal matchings in } GS_{\sigma}(A, B)$

- $s_{\sigma}(A^M, B^M) = \sum_{C \in P} \left(\frac{w(C)}{|C|+2}\right) + \sum_{C \in I} \left(\frac{w(C)}{|C|+1}\right) + \sum_{C \in C} \left(\frac{w(C)}{|C|}\right)$

where P, I, C are *even paths, odd paths* and *cycles*

- **Cycles** are more important than **paths**

ILP formulation for an
exact solution



ILP formulation: constraints

- **Integer Linear Programming (ILP):** mathematical optimization in which there are constraints and some or all variables are *integers*
- $x_{ab} = 1$ iff **edge** between a and b is included in the **solution**
- $y_{ri} = 1$ iff **vertex** v_r belongs to the cycle i in the **solution**
- E_G = set of edges in **Gene Adj. Graph**
- Core constraints (on binary variables):
 - **Valid edges:** $x_{a^h b^h} = x_{a^t b^t}, \forall ab \in E_G$
 - **Maximal matching:** $x_{a^h a^t} + x_{b^h b^t} \leq 1, \forall ab \in E_G$ (self-edge \rightarrow gene removed from solution)
 - Each vertex can only belong to one cycle: $\sum_{i=1}^r y_{ri} \leq 1, \forall v_r \in V_H$

ILP formulation: objective function

- We have to:

$$\text{maximize } \sum_{i=1}^k \sum_{l \in L} \sum_{e \in E_m} \frac{w_e x_{ei}^l}{l}$$

- E_m = matching edges set
- k = number of vertices in Adjacency Graph
- $l \in L$ = every possible length for the cycles
- $x_{ei}^l = 1$ if edge $e \in$ cycle i with length l

ILP formulation: complexity

- We have to:

$$\text{maximize } \sum_{i=1}^k \sum_{l \in L} \sum_{e \in E_m} \frac{w_e x_{ei}^l}{l}$$

- $O(N^4)$ variables
- $N = |A| + |B|$
- Solving an ILP problem is **NP-hard**

Faster heuristic algorithms

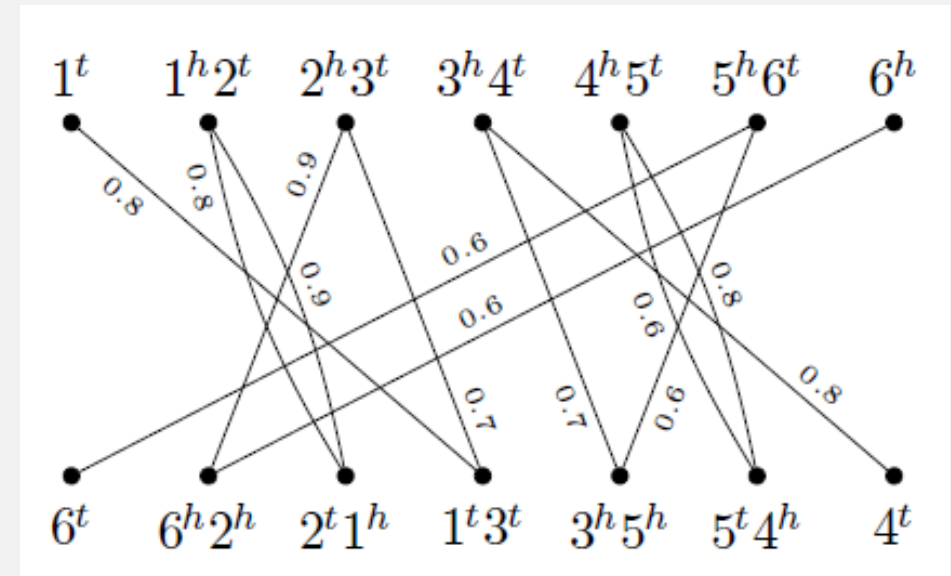


Heuristic algorithms: Best Density

- **First step:** generate all cycles in weighted adjacency graph
- **Second step:** Store all cycles C of $AG_{\sigma}(A, B)$ in decreasing order of their *density* $\frac{w(C)}{|C|^2}$
- **Third step:** *while* it is possible select and add to M the best density *consistent* cycle, remove it from AG , *loop*
- **Final step:** return $s_{\sigma}(A, B) = \sum_{C \in M} (\frac{w(C)}{|C|})$

Heuristic algorithms: Best Density example

- **First step:** two cycles of length 2
- **Second step:** $\langle c_{12}, 0.425 \rangle, \langle c_{45}, 0.350 \rangle$
- **Final result:** $s_{\sigma}(A, B) = \sum_{C \in M} \left(\frac{w(C)}{|C|} \right) = \left(\frac{0.9+0.8}{2} \right) + \left(\frac{0.8+0.6}{2} \right) = 1.55$
- *Max value for similarity is the number of components in the matching*



Heuristic algorithms: Best Length

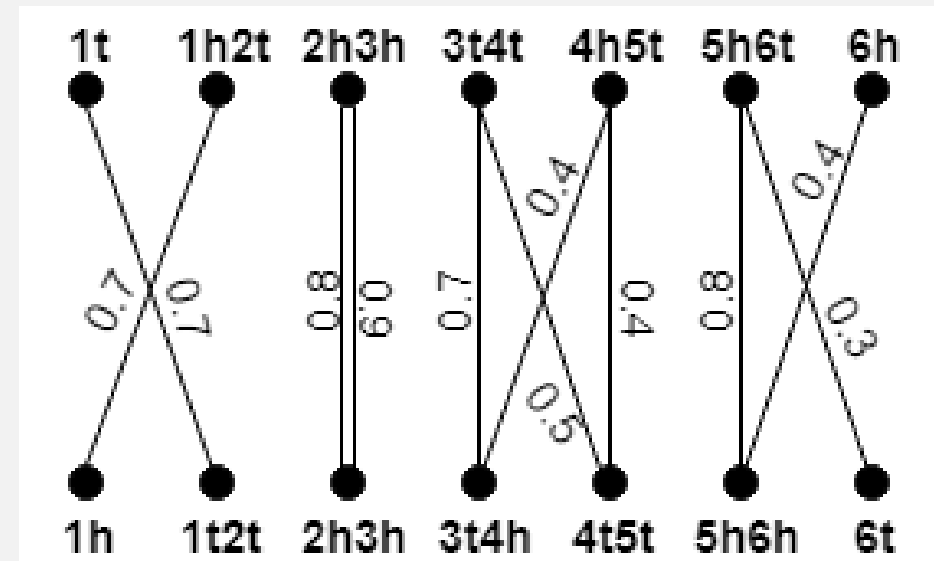
- **First step:** generate all cycles in weighted adjacency graph
- **Second step:** Store all cycles C of $AG_\sigma(A, B)$ in increasing order of their *lengths* $|C|$
- **Third step:** *while* it is possible select and add to M *consistent* cycle of length 2, then of length 4 and so on, *until* there are no more *consistent* cycles.
- **Final step:** return $s_\sigma(A, B) = \sum_{C \in M} \left(\frac{w(C)}{|C|} \right)$

Heuristic algorithms: Best Length example

- **First step:** one 2-cycle and one 4-cycle
- **Second step:** $\langle c_{23}, 2, 0.425 \rangle, \langle c_{45}, 4, 0.125 \rangle$
- **Final result:** $s_{\sigma}(A, B) =$

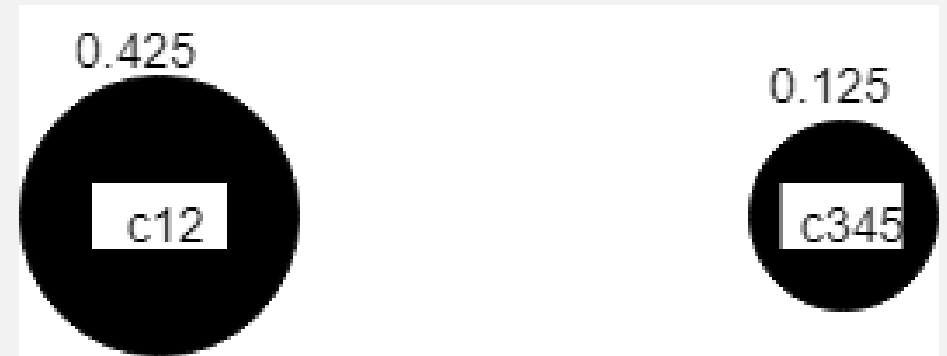
$$\sum_{C \in M} \left(\frac{w(C)}{|C|} \right) = \left(\frac{0.9 + 0.8}{2} \right)$$

$$+ \left(\frac{0.7 + 0.4 + 0.4 + 0.5}{4} \right) = 1.35$$
- *Lower similarity than before: 4-cycles are less important*



Heuristic algorithms: Best Length with Max Independent Set

- **Cycle Graph:** vertex is a cycle, edge if cycles are *inconsistent*
- **Weight of vertex:** *density* of cycle
- **Algorithm:** find an *independent set* with the greatest weight in the cycle graph



Experimental results



Experimental results: tests characteristics

- **Single machine:** 16GB memory, *gurobi* library for ILP computation, 1800sec time limit
- Each dataset with **10 simulated genome samples:** each genome of sizes around 25, 50 or 1000 genes (only used for heuristics)
- **45 comparisons** of pairs per dataset
- Tests on different proportions of **genome level events: r-fold** with rates 1, 2 and 5
- $Gap = \left(\frac{upper\ bound}{best\ solution} - 1 \right) * 100$

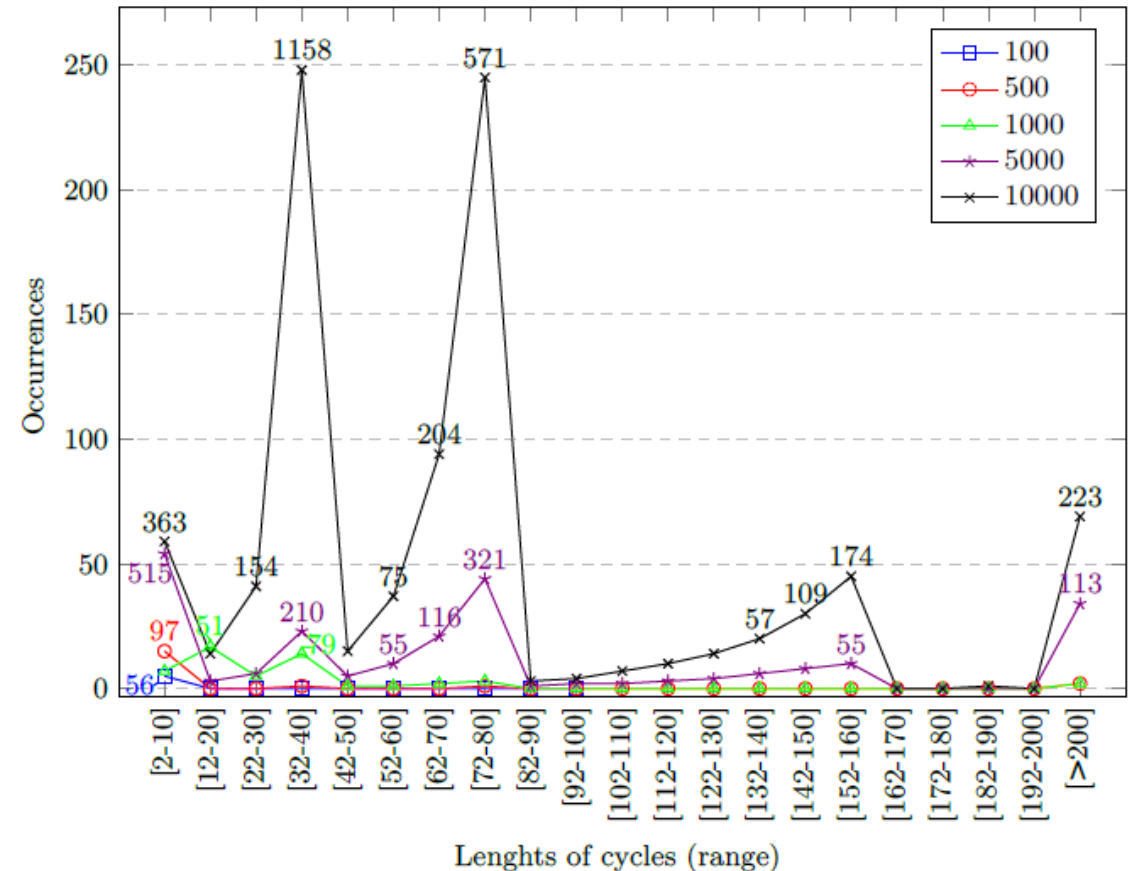
Experimental results: time and gap comparison

	ILP			GREEDY-DENSITY	GREEDY-LENGTH	GREEDY-WMIS
	Time (s)	Not finished	Gap (%)	Gap (%)	Gap (%)	Gap (%)
25 genes, $r = 1$	19.50	0	–	5.03	5.84	5.97
25 genes, $r = 2$	84.60	2	69.21	30.77	43.57	43.00
25 genes, $r = 5$	49.72	0	–	43.83	55.38	55.38
50 genes, $r = 1$	445.91	7	19.56	18.74	19.36	18.90
50 genes, $r = 2$	463.50	29	38.12	65.41	66.52	64.78
50 genes, $r = 5$	330.88	29	259.72	177.58	206.60	206.31

- **Gap** increases proportionally to the rate of reversals and translocations (association of genes is subject to this bias)
- **Greedy-density** provides almost every time the best solutions among the heuristics (balance between normalized weight and cycles length)

Experimental results: cycles distribution

- **5-fold** instances running the **Greedy-Density** algorithm on genomes with different size
- **Most of the cycles** found are **short** compared to the genome sizes
- That's why **heuristics** are faster (2896 sec on 10000 genes)
- Number of cycles could be **exponential** in theory → enumerate is bad!



Conclusion



Conclusion: ILP vs Heuristics

- ILP program is fast and accurate for smaller instances
- **Greedy-density** heuristic is probably the best choice for general use on larger instances
- **Drawback:** FF DCJ similarity values cannot be compared easily: value is between 0 and $|M|$
- **Best normalization:** divide sim. value by $|M|$



Thank you for your attention!

Reference: Diego P. Rubert, Gabriel L. Medeiros, Edna A. Hoshino, Marilia D.V. Braga, Jens Stoye, and Fabio V. Martinez

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