

CSE 566 Spring 2023

Whole-Genome Comparison

Instructor: Mingfu Shao

Sequence Comparison

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- Orthologous genes may have different sequences.

Seq1: A C G T A G A C C G C T A C A G

Seq2: T C C T A G A T C C G T T C G

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Seq2: T C C T A G A T C C G T T C G

- Point mutations
 - Nucleotide substitutions
 - Nucleotide indels

C T A T A G C
 ↑↓ substitution
C T A G A G C

C T A X A G C
 ↑↓ insertion/deletion
C T A - A G C

Pairwise Sequence Comparison

- **Edit distance:** the **minimum** number of substitutions and indels that can transform one sequence into the other.

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Seq1: A C G T A G A - C C G C T ~~A~~ C ~~A~~ G
Seq2: T C C T A G A ~~T~~ C C G T - C - G

The diagram shows the alignment of Seq1 and Seq2. Red circles highlight the characters involved in substitutions: A in Seq1 aligned with T in Seq2, G in Seq1 aligned with C in Seq2, and C in Seq1 aligned with T in Seq2. Red 'X' marks indicate deletions: A in Seq1 and A in Seq2. Vertical lines connect the aligned characters.

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Seq1: A C G T A G A - C C G C T ~~A~~ C ~~G~~ G

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The diagram shows a pairwise sequence alignment between Seq1 and Seq2. Seq1 is A C G T A G A - C C G C T A C G and Seq2 is T C C T A G A T C C G T T C G. The alignment is shown with vertical lines connecting matching nucleotides. Red circles highlight mismatches: A in Seq1 aligned with T in Seq2, G in Seq1 aligned with C in Seq2, C in Seq1 aligned with T in Seq2, and T in Seq1 aligned with A in Seq2. Red X marks indicate deletions in Seq1 (A and G) and insertions in Seq2 (T and -).

- Determines the **distance/similarity** and the **one-to-one correspondence** between the nucleotides.

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- Determines the **distance/similarity** and the **one-to-one correspondence** between the nucleotides.
- **Efficient algorithm:** dynamic programming

Multiple Sequence Comparison

Multiple Sequence Comparison

Seq1: A C G C A G A C C G T A C A

Seq2: A C G T A G A C C G C T A C A G

Seq3: T C C T A G A T C C G T T C G

Multiple Sequence Comparison

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- **NP-hard** for almost all formulations.

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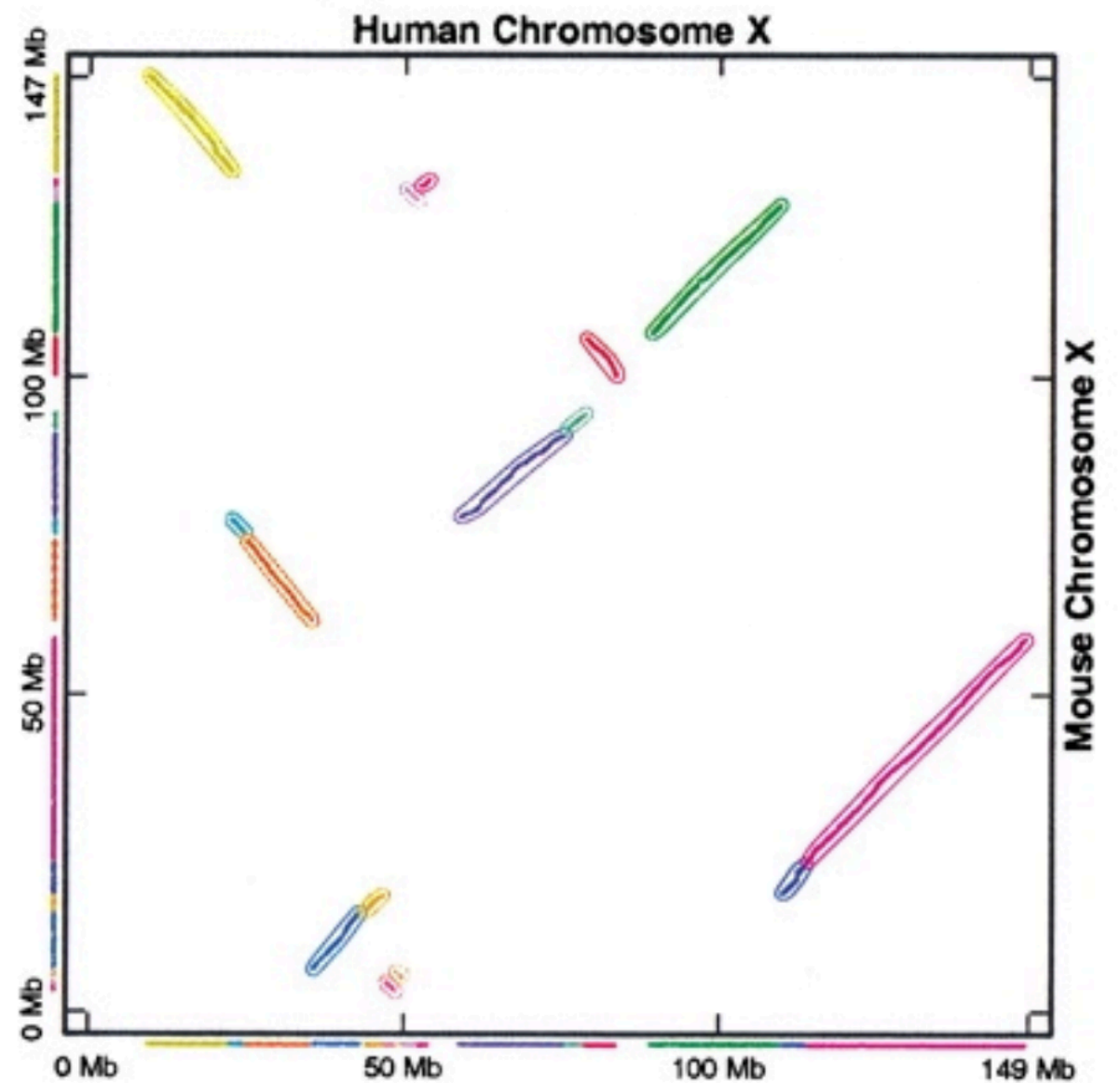
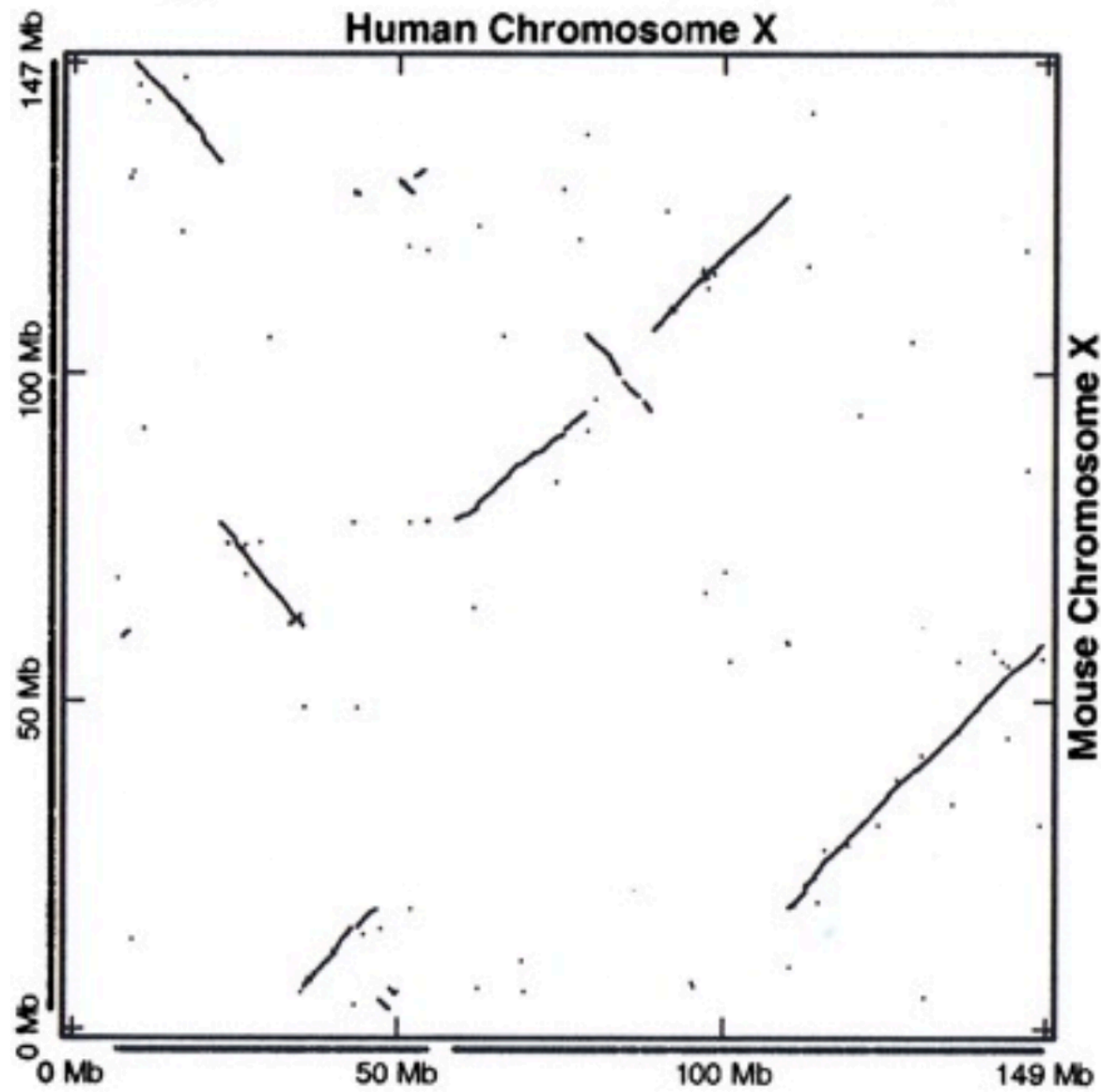
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- **NP-hard** for almost all formulations.
- **Heuristics:** progressive methods, iterative algorithms.

Comparing Whole-Genome

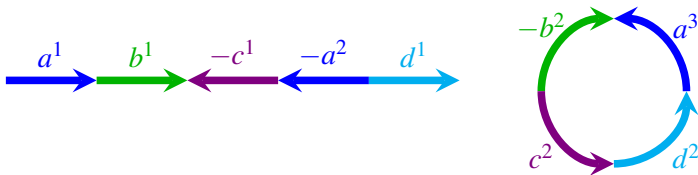


Model of Whole-genomes

- **Genome:** a set of chromosomes.
- **Chromosome:** a linear/circular list of genes (synteny blocks).

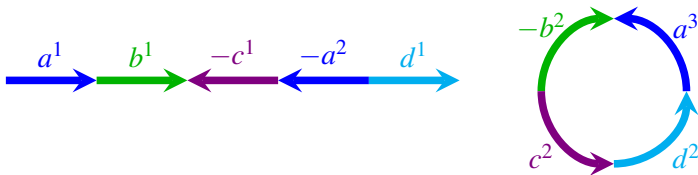
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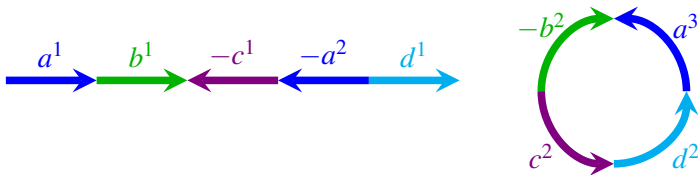
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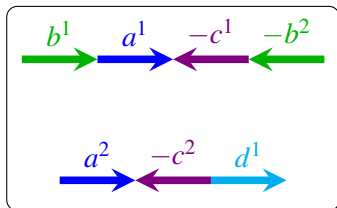
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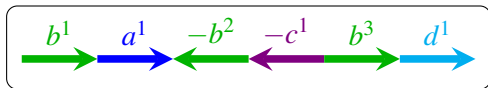
- Each gene has a **sign**, indicating its transcriptional direction.
- All genes are grouped into **gene families**.

Whole-genome Comparison

Whole-genome Comparison

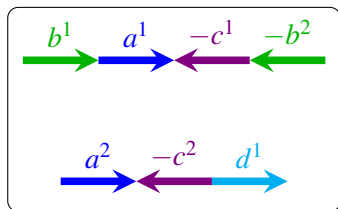


(G_1)

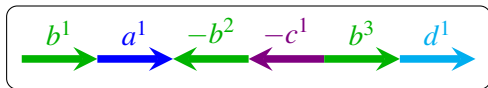


(G_2)

Whole-genome Comparison



(G_1)



(G_2)

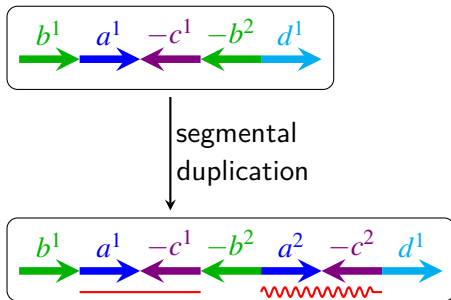
- Different copy numbers for some gene families.
- Different number of chromosomes.
- Different gene orders and orientations.

Large-scale Events

- **Content-modifying** events: change copy numbers
 - Segmental duplication
 - Tandem duplication
 - Lateral gene transfer
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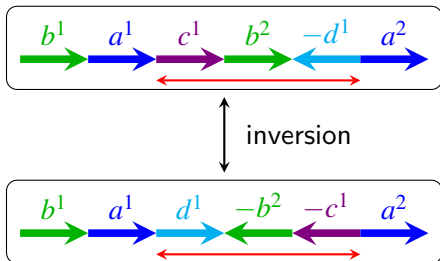


Large-scale Events

- **Rearrangements:** change number of chromosomes, gene orders and orientations.
 - Inversion
 - Translocation
 - Chromosomal fission
 - Chromosomal fusion

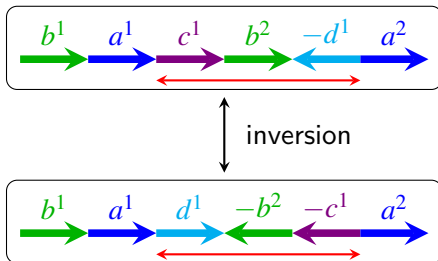
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- Almost all rearrangements can be represented by the universal **DCJ** (double-cut-and-join) operation.

DCJ (double-cut-and-join) Operation

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- Some definitions



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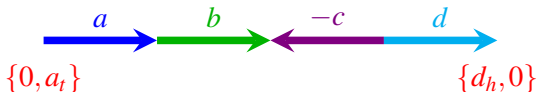
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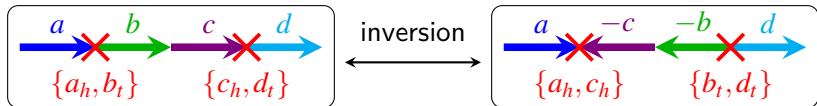
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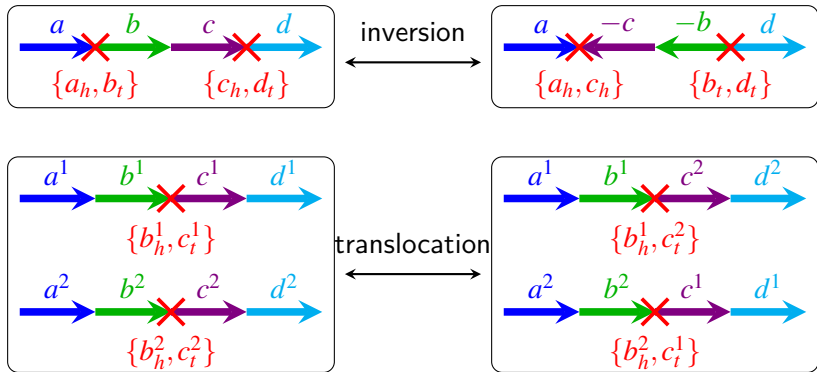
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- DCJ operation: $\{p, q\} + \{r, s\} \implies \{p, r\} + \{q, s\}$

Various Cases of DCJ Operation

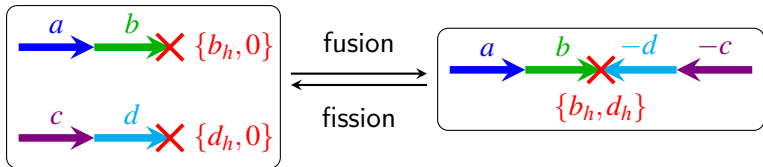
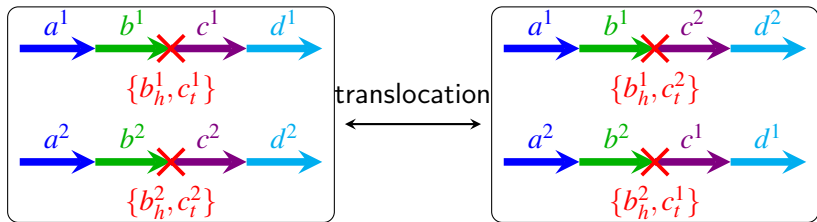
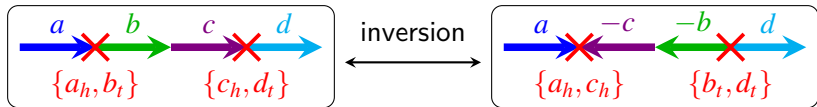
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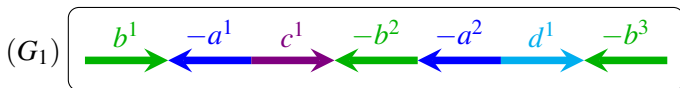
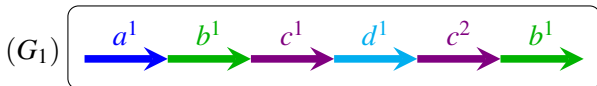


Pairwise Whole-genome Comparison

- **Edit distance:** given two genomes G_1 and G_2 , to compute the **minimum** number events (specified by an evolutionary model) that can transform G_1 into G_2 .

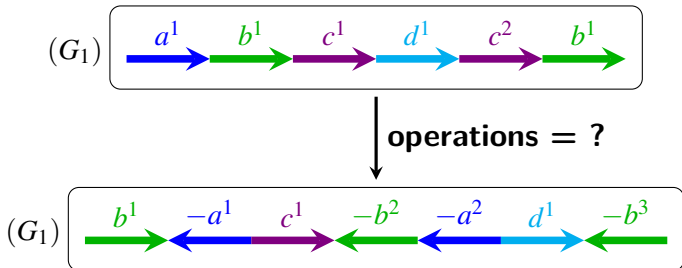
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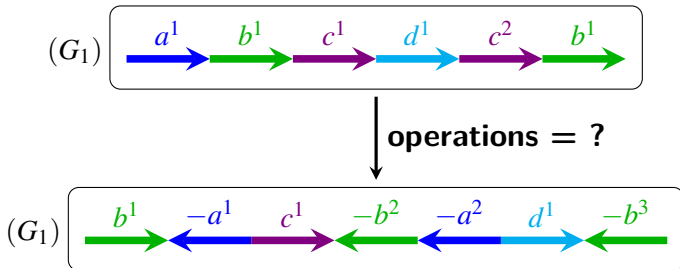
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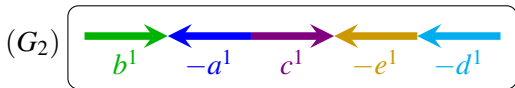
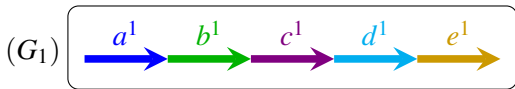
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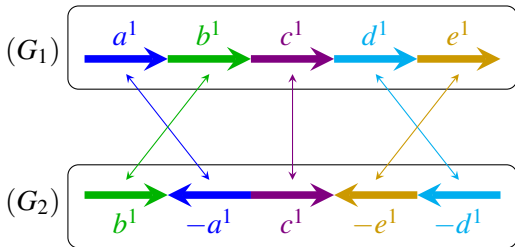


- Diversity of the edit distance problems
 - Assumptions on the given genomes.
 - Which types of events are in the evolutionary model.

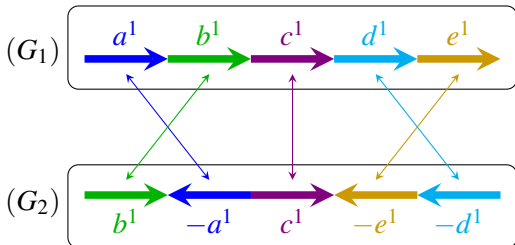
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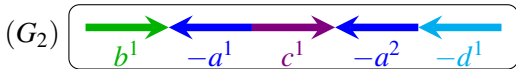
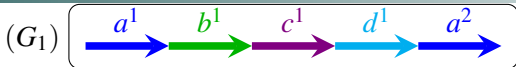


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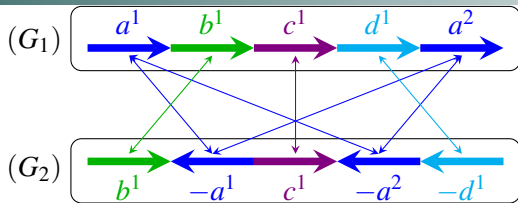


- Edit distances that can be computed in **linear time**:
 - **Inversion distance** (*Hannenhalli and Pevzner, 1995; Bader et al., 2001*)
 - **DCJ distance** (*Bergeron et al., 2006*)
 - **DCJ + Insertion + Deletion** (*Braga et al., 2010; Compeau et al., 2013*)

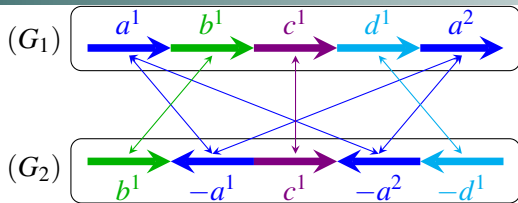
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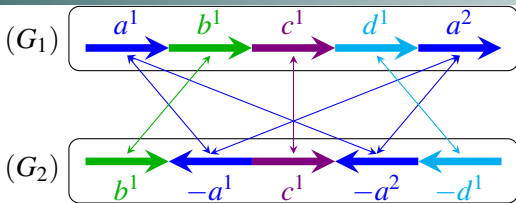


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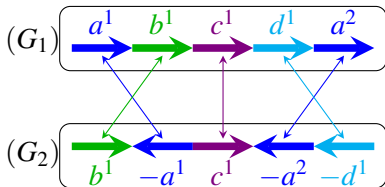


- **Matching:** To build a matching between homologous genes, such that a certain distance is minimized.

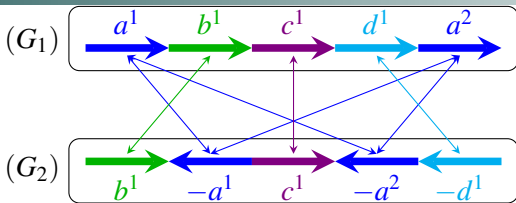
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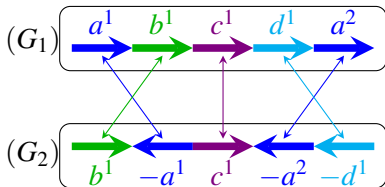
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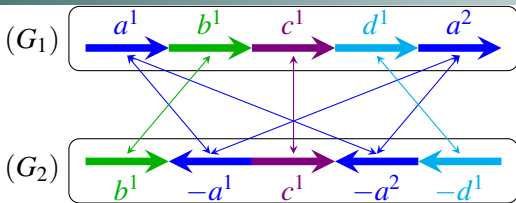


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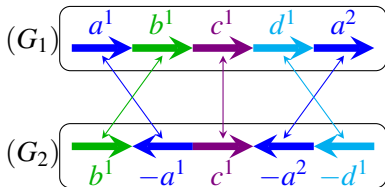


- **Exemplar (Sankoff, 1999):**
To select one gene in each family, such that a certain distance is minimized.

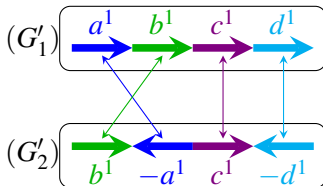
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- Under the **exemplar** strategy:
 - **Inversion distance:** Branch-and-Bound (*Sankoff et al., 1999*)
 - **Breakpoint distance:** Divide-and-Conquer (*Nguyen et al., 2005*)

Multiple Whole-genome Comparison

- **Median distance problem:** Given k genomes G_1, G_2, \dots, G_k , to compute a *median* genome G_0 such that $\sum_{i=1}^k d(G_0, G_i)$ is minimized.

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- Most of the formulations are **NP**-hard.
- **Inversion median distance:** Branch-and-Bound (*Siepel and Moret, 2001*), Heuristics (*Rajan et al., 2010*)
- **DCJ median distance:** Branch-and-Bound (*Zhang et al., 2009*), Decomposition scheme (*Xu and Sankoff, 2008*)

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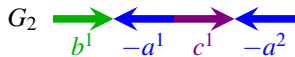
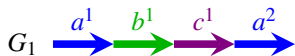
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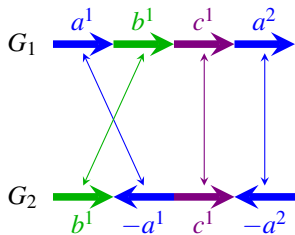
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- 2 After clarifying the evolutionary events in the history, orthologs and paralogues can be naturally inferred.
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- 4 These problems are interesting for computer scientists, since they usually have very special structures.

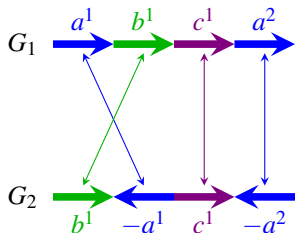
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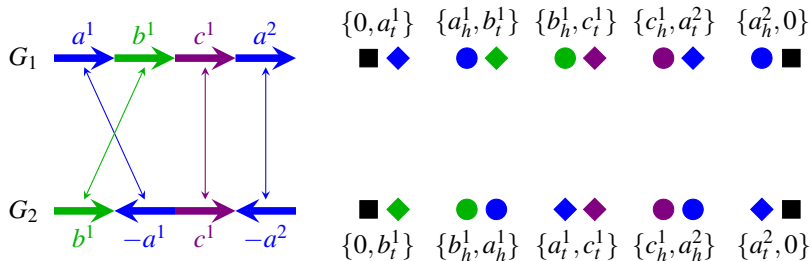


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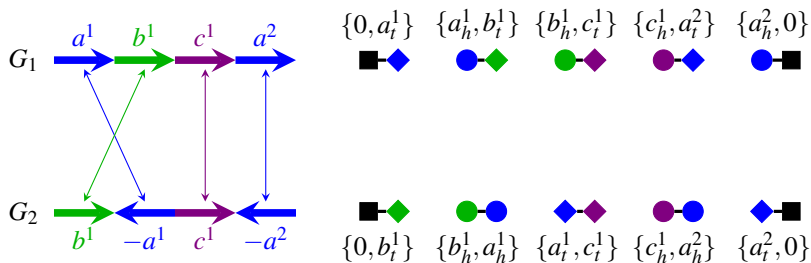
■ Adjacency Graph

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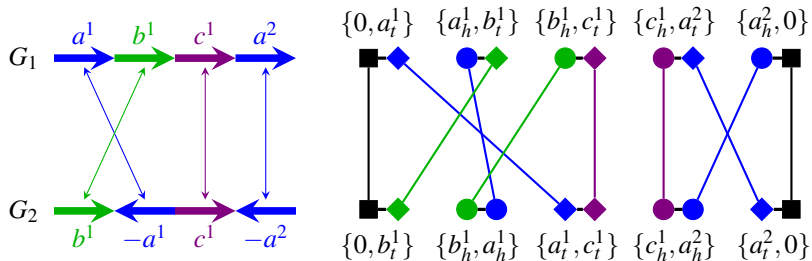
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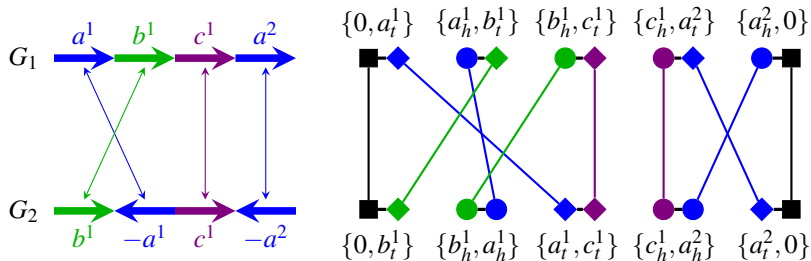
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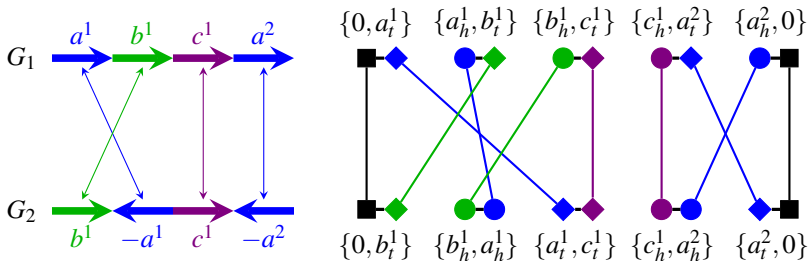
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- DCJ distance: $d(B) = \#(\text{adjacencies}) - \#(\text{cycles})$.

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- A fast and exact algorithm.
 - An ILP formulation that gives the optimal solution.
 - Efficient Algorithms to identify optimal substructures.

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■ Mathematical Formulation

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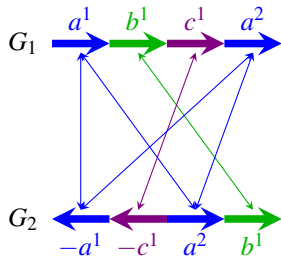
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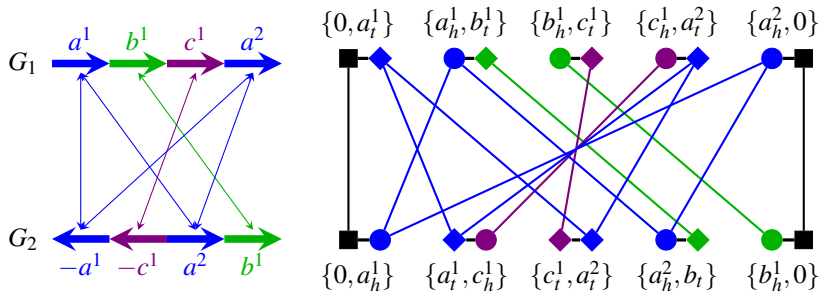
- CPLEX

- GUROBI

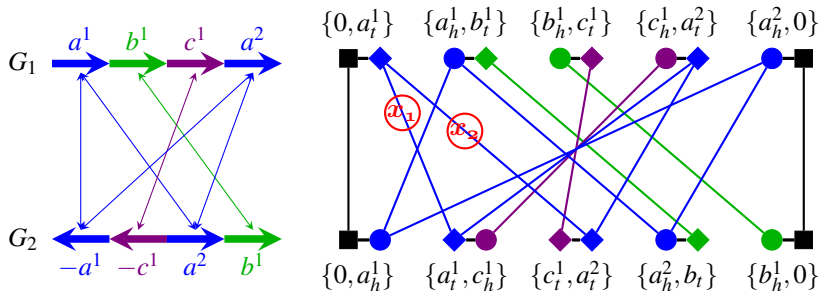
ILP Formulation



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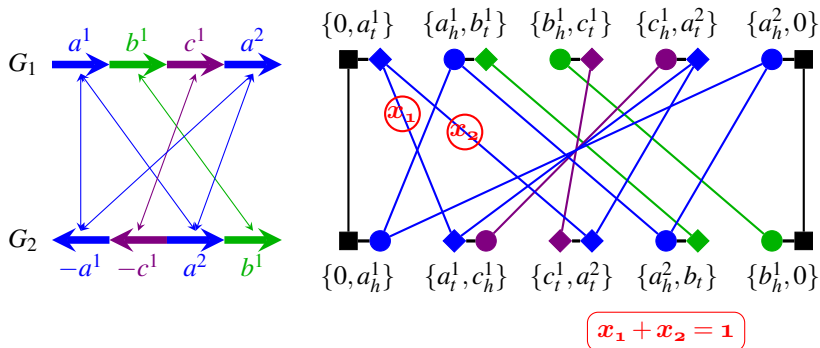


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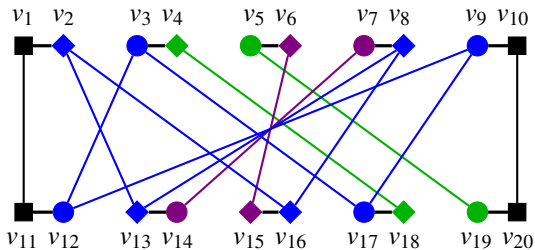
- Variables: $x_e \in \{0, 1\}$, indicating choosing e or not.

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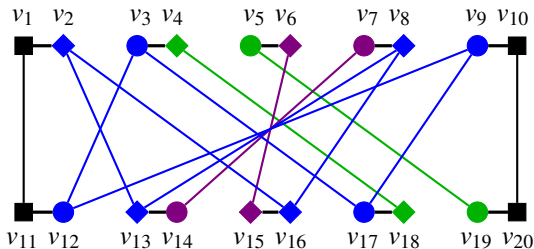


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- Constraints: Ensure a one-to-one correspondence.

ILP Formulation



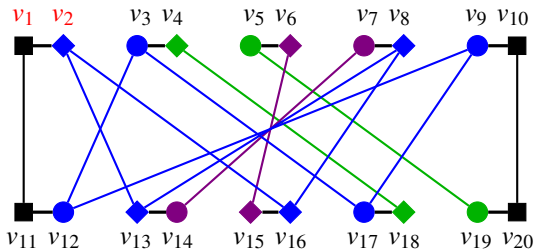
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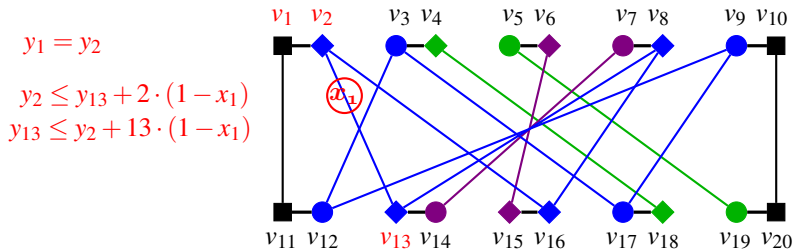
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$$y_1 = y_2$$



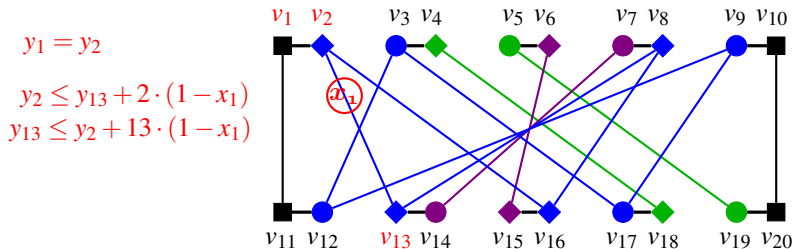
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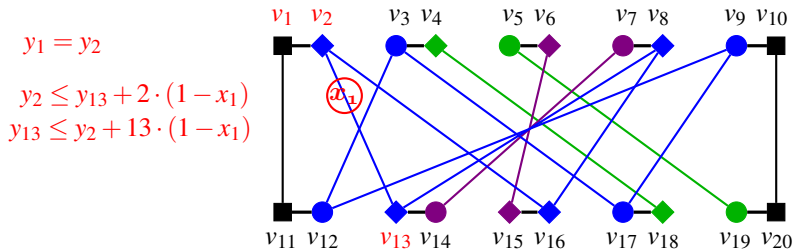
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⇒ At most one vertex in each cycle can reach the upper bound.

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- $O(|E|)$ variables and $O(|E|)$ constraints