

# COMPUTING SIMILARITY BETWEEN PHARMACOPHORES USING GRAPH EDIT DISTANCES

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May 3, 2019

AGAC

*Analyse de Graphes Appliquée à la Chémoinformatique*

Financial support from the Normandy region



- Context
- Objective and Principle
- Implementation
  - Preliminary implementation
  - Deviation
- Conclusion and Perspective.

## Terminologies:

- **Ligand:** an organic molecule that binds with a biomolecule
- **Target:** a part of the organism to which a ligand binds, producing a change of behaviour. Proteins is a common class of targets.
- **Pharmacophore:** a part of a molecular structure that is responsible for a particular biological or pharmacological interaction that it undergoes.

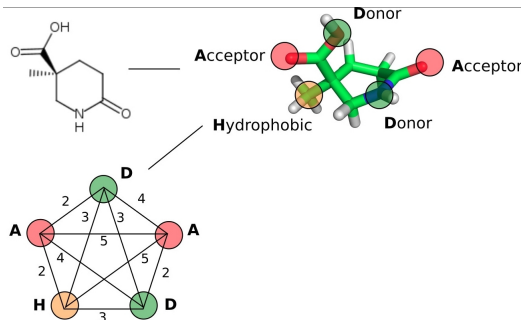


Figure : A ligand-pharmacophore model [Kutlushina et al., 2018]

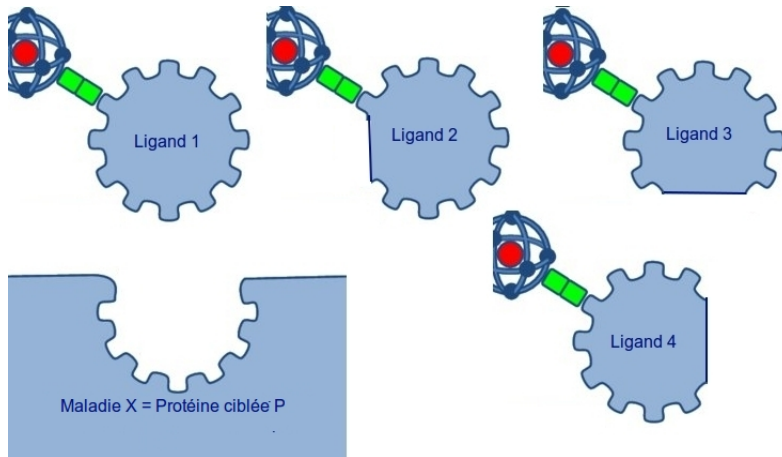


Figure : Ligand-Target Model.

Given a set of studied ligands and their biological profiles/bioactivity (to a target)  
→ Predict the potential bioactivity of ligand of which the biological profile is currently unknown.

- Compute similarity between studied ligands
- Elaborate a mechanism for a supervised classification
- Apply the supervised classification to the candidate ligand of which the bioactivity has not been known yet.

- Preliminary implementation
- Deviation.

- Compute similarity between ligands using Graph Edit Distance (GED)
- Compute Graph Edit Distance using a Branch-and-Bound procedure  
*[Abu-Aisheh et al., 2017]*
- Apply GED to 1492 ligands (from 8 to 13 vertex) from a ligand set for ABL-target
  - It exists instances without any optimality within 10 computation minutes
  - Size of graph should be reduced by using sub-parts of ligands which are responsible for pharmacological interactions.



Example: H-H-H-4-3-3 vs. A-H-H-H-4-3-3-4-7-8

→ Assumption: good cost configuration

→ 3 Node substitutions and 1 Node insertion:  $3! \times 4 = 24$  possibilities

- 1 Node substitution: H → A
- 1 Node insertion: H
- 3 Edge insertions: A-H (3); H-H (7); H-H (8)
- 3 Edge substitutions: H-H (4) → A-H (4); H-H (3) → H-H (4); H-H (3) → A-H (3).

- Apply **Norns** (by projet [Minomics](#)) to the dataset to extract a space of pharmacophores which satisfy the following constraints [*Métivier et al., 2018*]:
  - Size of graph: from 1 to 6 vertex
  - Support/Frequency: at least 10
  - Growth Rate: used to classify pharmacophores
  - With and without MMRFS algorithm
- Without MMRFS ( $S_1$ ): 87175 pharmacophores
- With MMRFS ( $S_2$ ): 137 pharmacophores
- Edit Cost Configuration:
  - Node Deletion/Insertion: 6
  - Node Substitution: 10
  - Edge Deletion/Insertion: 0
  - Edge Substitution:  $difference(edge1, edge2)$
- Computation time for GED: 0.5 second

- From viewpoint of pharmacophores in  $S_1$ :
  - Compute GED between 2 pharmacophores in  $S_2$  and all pharmacophores in  $S_1$
  - Represent the nearest pharmacophores in  $S_1$  to the 2 pharmacophores in  $S_2$

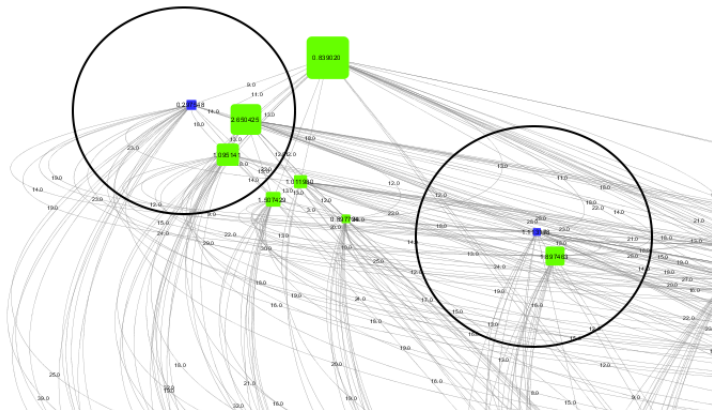


Figure : An example of representation processing used GED.

- From viewpoint of each candidate ligand with currently unknown biological profile:
  - Compute GED between all pharmacophores in  $S_2$  belonging to the ligand and all pharmacophores in  $S_1$
  - Predict the bioactivity of the candidate ligand using profiles of the nearest pharmacophores.

- Use GED as similarity between ligands and between pharmacophores
- Apply a Branch-and-Bound procedure to GED computation
  - Is there any better way for an exact solution?
- Use the nearest neighbours to represent pharmacophores and to predict the bioactivity of the candidate ligand
  - Is there any better way for a prediction with relational values?