# An aproximmation to the Minimum Branch Vertex Spanning Tree problem

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#### Introduction

Introduction

- Spanning trees : a way to modelize flow structure.
- Aim: to obtain enough good approximated solutions of the Minimum Branch Vertex Spanning Tree (MBVST)

#### Our project phases :

- Operational Research part :
  - Construction of the ILP
  - 2 Nodes Preprocessing
- Data Science part :
  - Estimation of the number of branch vertex and the time to solve the ILP
  - 2 Estimation of branch vertex nodes
- Final deliverable : algorithm given a spanning tree

## The ILP problem

$$\begin{aligned} & \min \sum_{v \in V} y_v \\ & \text{subject to} : \sum_{e \in E} x_e = n - 1, \\ & \sum_{(s,v) \in A^+(s)} f_{sv} - \sum_{(v,s) \in A^-(s)} f_{vs} = n - 1, \\ & \sum_{(v,u) \in A^+(v)} f_{vu} - \sum_{(u,v) \in A^-(v)} f_{uv} = -1 \ \forall v \in V \setminus \{s\}, \\ & f_{uv} \leq (n-1)x_e \ \forall e \in \{u,v\} \in E, \\ & f_{vu} \leq (n-1)x_e \ \forall e \in \{u,v\} \in E, \\ & \sum_{e \in A(v)} x_e - 2 \leq (n-1)y_v \ v \in V, \\ & x_e \in \{0,1\} \ \forall e \in E, \quad y_v \in \{0,1\} \forall v \in V, \quad f_{u,v}, f_{vu} \geq 0 \ \forall e = \{u,v\} \in E, \\ \end{aligned}$$

Following [Merabet et al., 2018], we can clasify our nodes in three types :

- $V_1$ : nodes with no more than k edges.
- $V_3$ : nodes that deleting them and its edges will result in at least k+1connected components.
- $V_2: V \setminus (V_1 \cup V_3)$ , these are our candidate nodes.

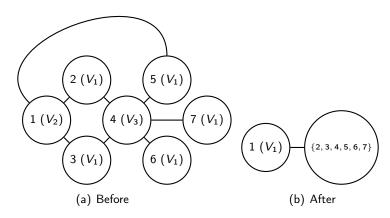


Figure – Tranformation using the method for a graph G.

# Effect of graph reduction

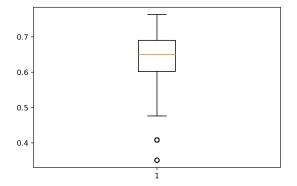


Figure – Relation between the number of nodes after the reduction of the graph and the initial number of node of the graph.

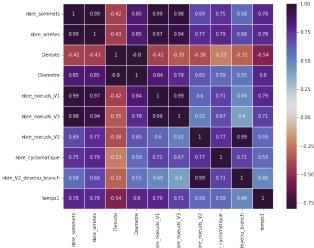
In a first step we constructed a data frame where each row/observation represents a given graph. For each graph we have retrieved several pieces of information, here they are:

- nbre\_sommets
- nbre\_arretes
- Densite
- Diametre
- nbre\_noeuds\_V1nbre\_noeuds\_V3
- nbre\_noeuds\_V2
- nbre\_cyclomatique nbre\_V2\_devenu\_branch
- temps1

This dataframe was used to build two types of Machine Learning models, which we will see later in the presentation

### The first data frame

#### Study of correlations:



### Estimation of the time to solve the ILP

Using the same dataframe, this model predicts, for a given graph, the time it will take the LP to find the solution graph of the problem.

Models

- Results
- Utility for Linear Programming

Using the previous dataframe, this model predicts, for a given graph, the number of branching nodes that the associated solution graph will have.

Models

- Results
- Quantiles Regression
- Utility for Linear Programming

A model that for a given node of a given graph, predicts whether that node will be a branching node in the associated solution graph or not. The dataframe for this kind of models:

- nbre\_voisins\_V1 nbre\_voisins\_V2
- nbre\_voisins\_V3
- nbre\_comp\_connexes nbre\_base\_cycle
- Is V2

### Estimation of the branch vertex nodes

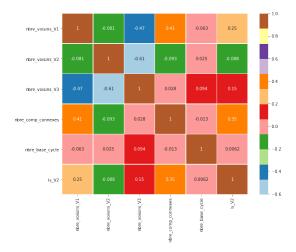
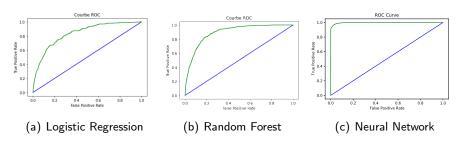


Figure – Correlations of the second dataframe

#### Models and results:

Introduction



Data Science part

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Figure – ROC Curves for the considered models.

## Final algorithm

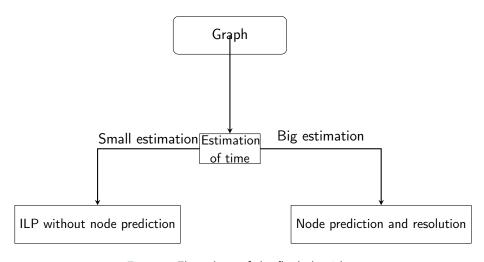


Figure – Flow chart of the final algorithm

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# Further improvements

- To improve node estimation.
- To measure the trade-off between the ILP time resolution and the probability of being considered branch or no-branch.
- To apply other problem reduction tools.

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# Références principales



Introduction

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