

PREDICTING ACADEMIC SUCCESS IN HIGHER EDUCATION



Team Presentation



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<https://github.com/Mariana13-31/ST0245-001/tree/master/proyecto>



Algorithm Design

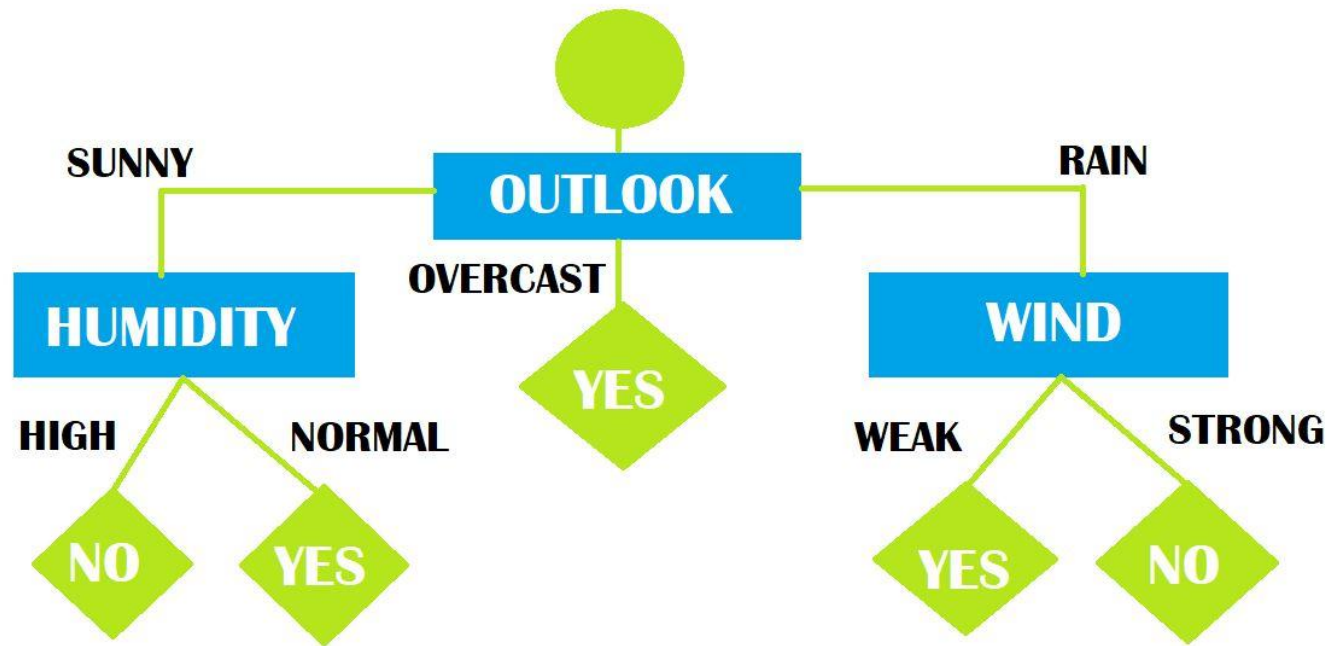
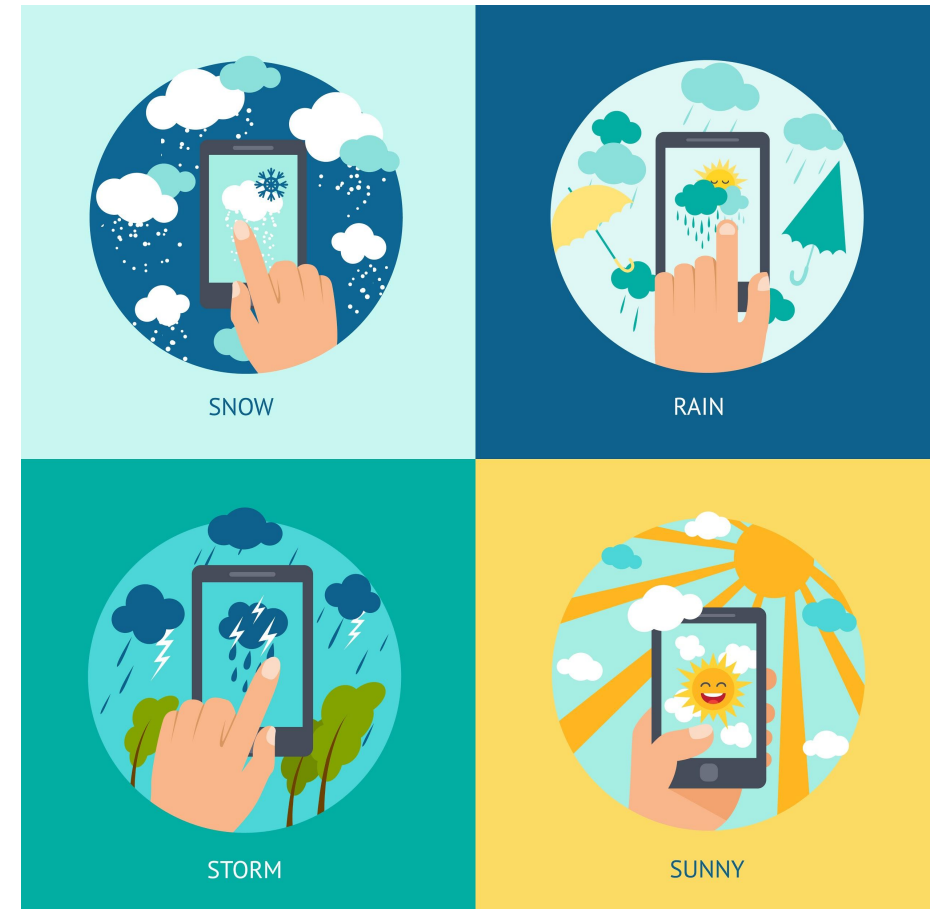


Figure 1. The algorithm used to build the binary decision tree was CART. In this figure we can see an example that processes different conditions in order to predict the weather using as root node the variable outlook, which has a lower impurity Gini so it is the best to divide the data that make up this example.



Node Splitting

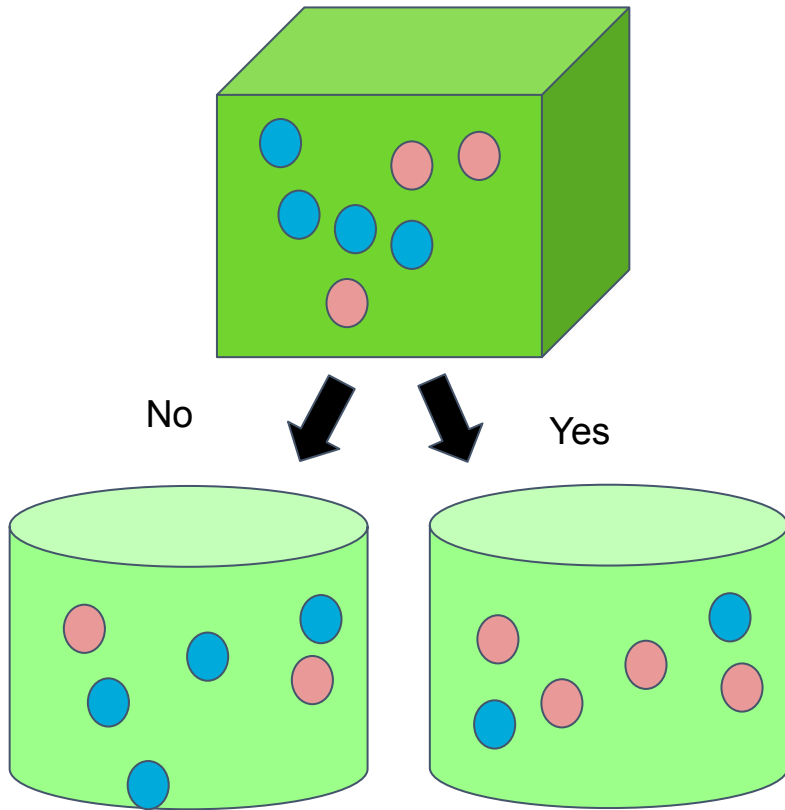


Figure 2. As an example, this split is based on the condition "English ≤ 82 ." For this case, left Gini impurity is 0.42, right Gini impurity is 0.41, and weighed Gini impurity is 0.41.

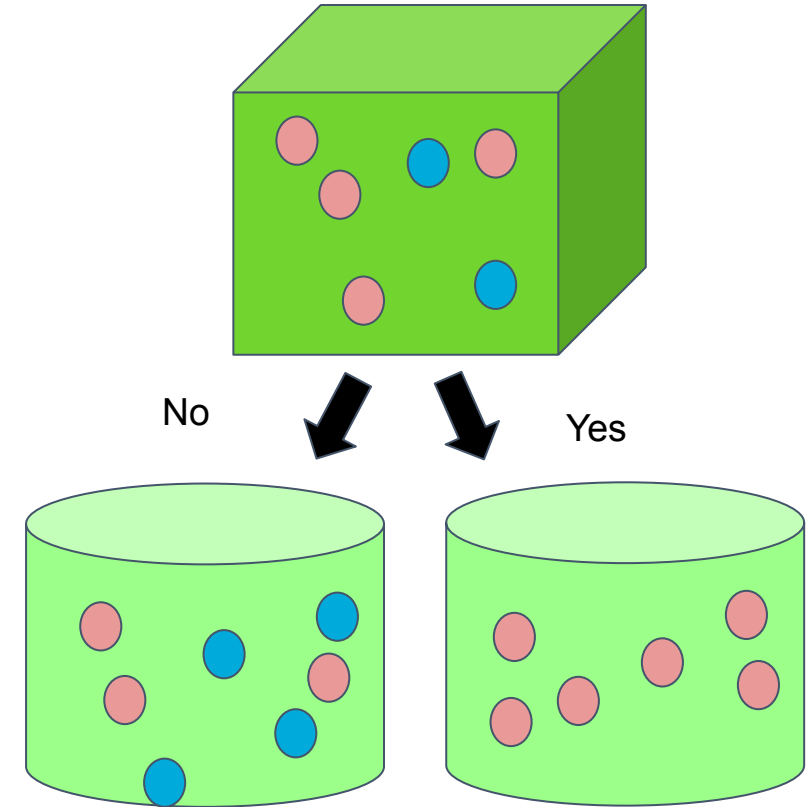


Figure 3. As an example, this split is based on the condition "Math ≤ 68 ." For this case, left Gini impurity is 0.48, right Gini impurity is 0, and weighed Gini impurity is 0.26.



Random Forest. To put it simply: the Random Forest creates multiple decision trees and combines them for a more accurate and stable prediction.

Decision-Tree Model

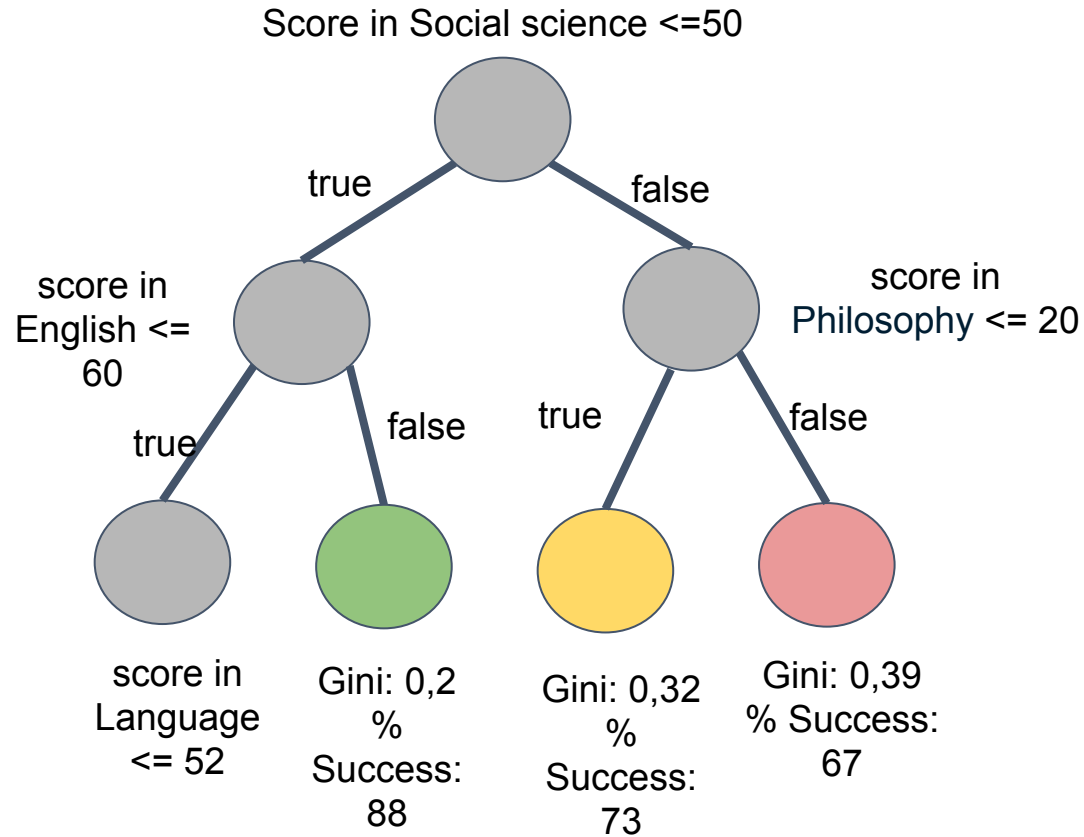


Figure 4. A binary decision tree to predict Saber Pro scores based on the results of Saber 11. Green nodes represent those with a high probability of success, yellow medium probability and red a low probability of success.

Most Relevant Features:



Social Sciences



English



Philosophy

Algorithm Complexity



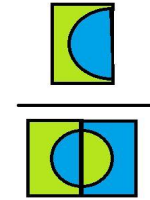
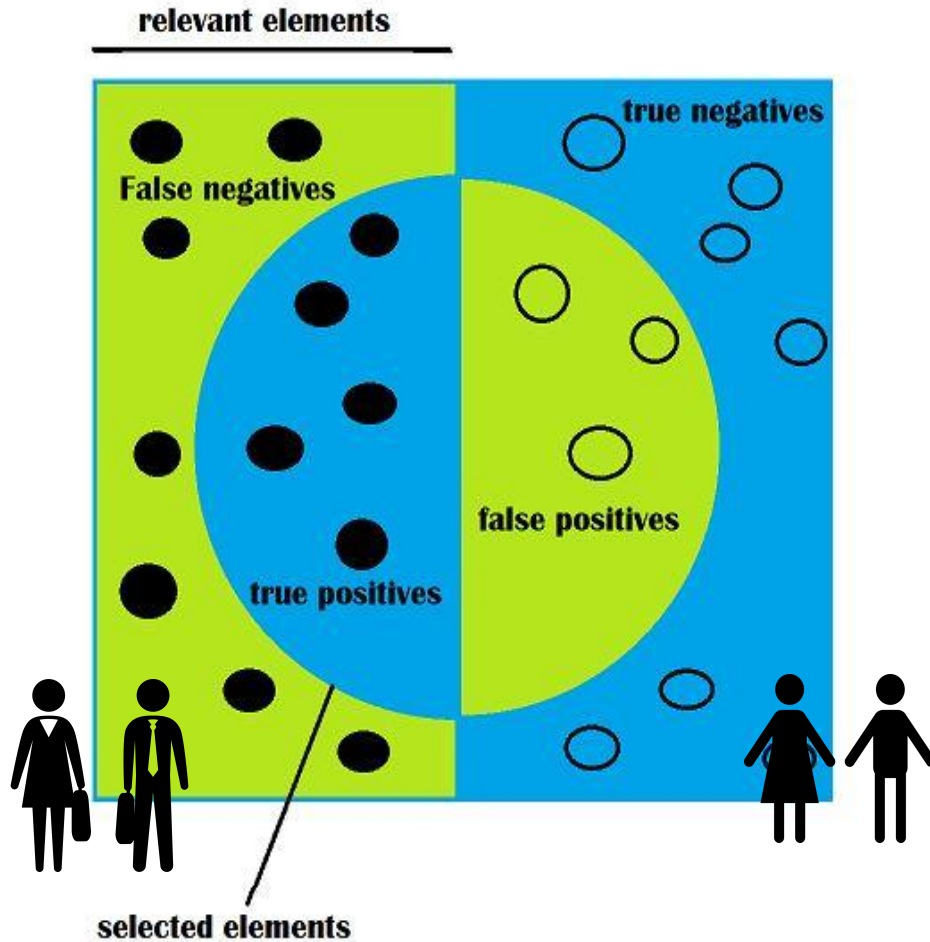
	Time Complexity	Memory Complexity
Training the model	$O(m * t * n \log n)$	$O(m * n)$
Testing the Model	$O(n * t)$	$O(m * n)$

Complexity in time and memory of the algorithm used that implements random forests based on the CART algorithm. Here we can interpret the variable **n** as the number of students, **m** the number of variables to be taken into account as relevant and **t** as the number of trees that make up the forest

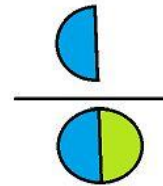


designed by freepik

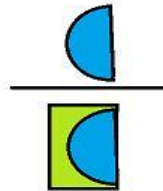
Evaluation Metrics



Accuracy: Total successful predictions over total data



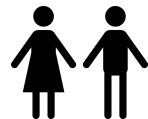
Precision: Successful prediction of success over total success



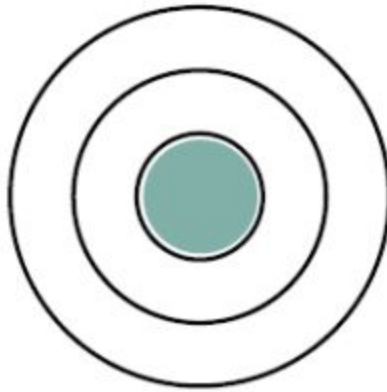
Recall: Successful prediction of success over total successful predictions



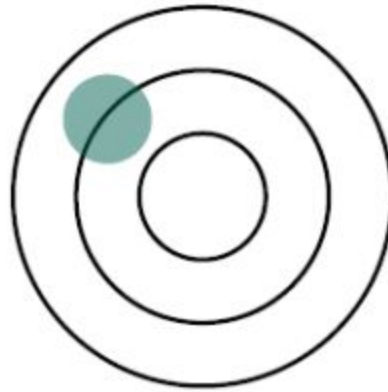
Success predictions



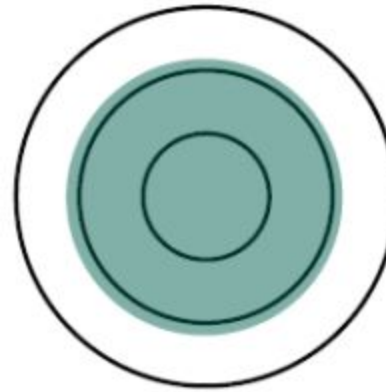
Not Success predictions



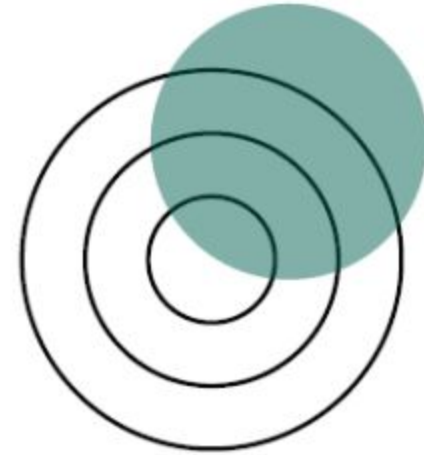
High precision
Good trueness



High precision
Bad trueness



Low precision
Good trueness



Low precision
Bad trueness

Evaluation Metrics

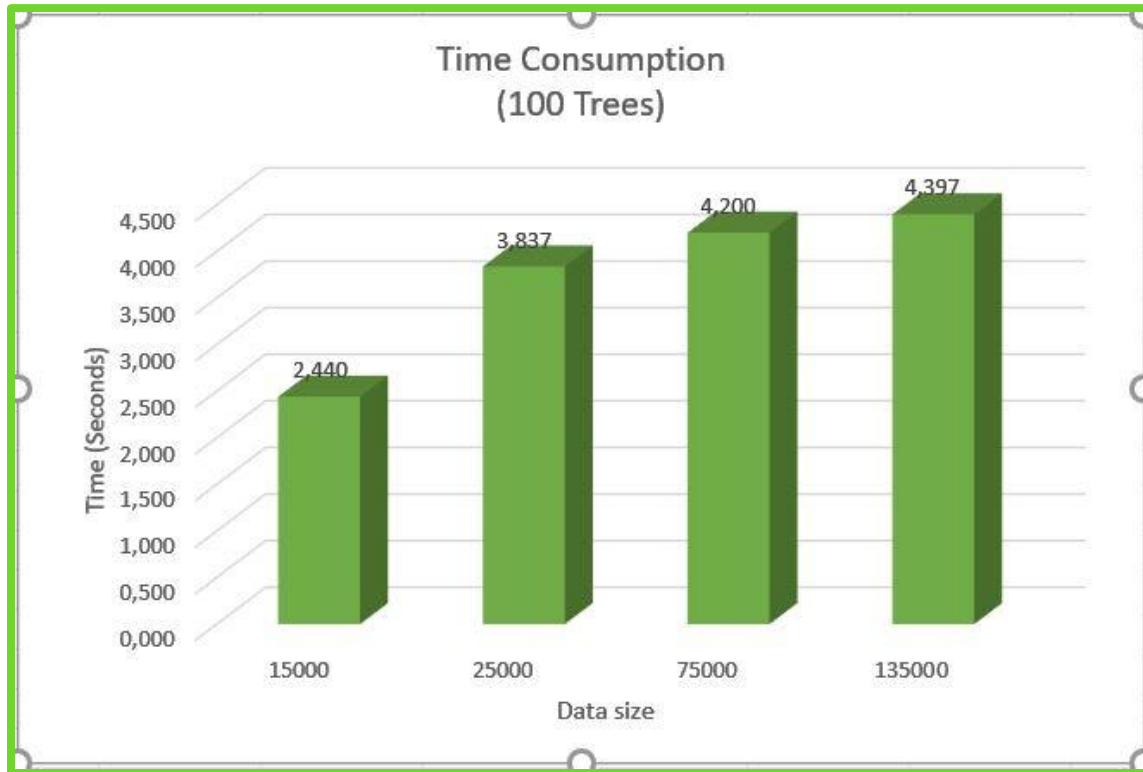


	Training data set	Testing data set
Accuracy	0.78	0.78
Precision	0.77	0.75
Recall	0.61	0.58

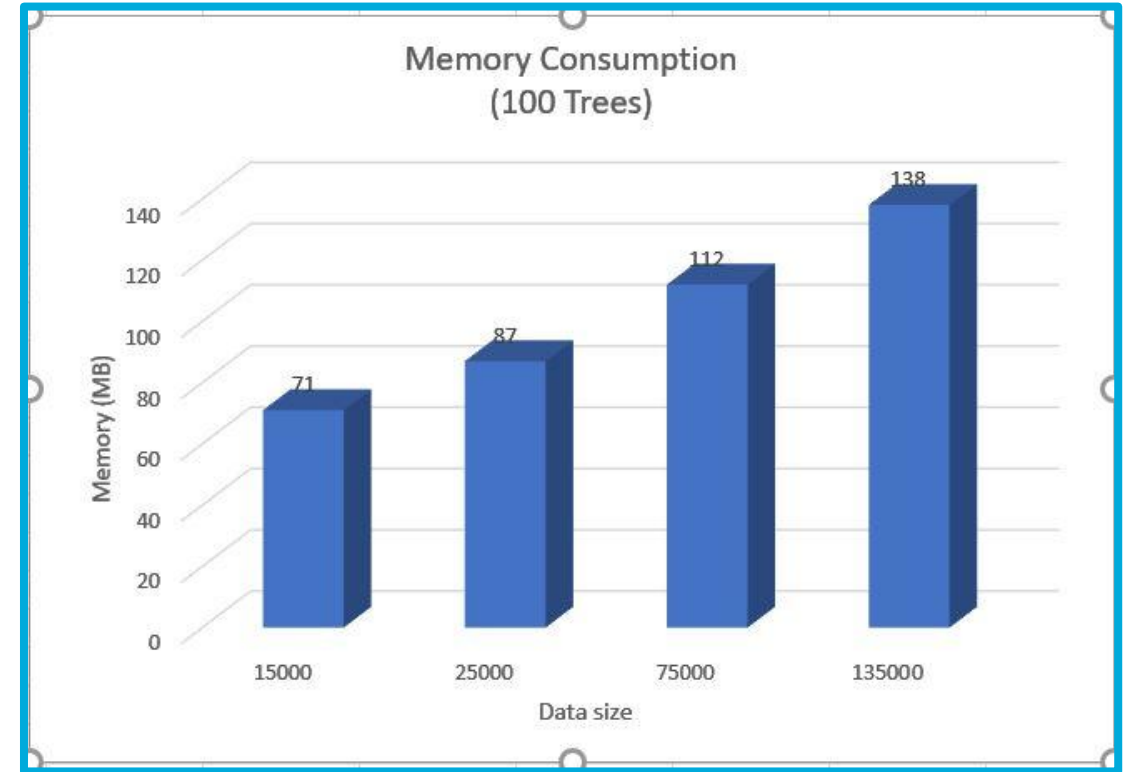
Evaluation metrics obtained with the training data set of 135,000 students and the testing data set of 25,000 students



Time and Memory Consumption



Time Consumption



Memory Consumption

THANK YOU!

