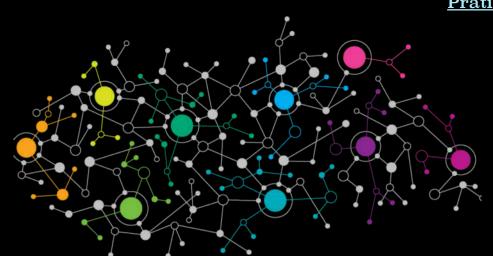




# Supervised Learning: Predicting video games user review scores

Pratical assignment developed for EIADC Class





#### Machine Learn Problem

o In this practical work (<u>assignment nº 2</u>) we will analyze a dataset that contains various atributes like year of release, genre, number of follows, and the target atribute of prediction: rating of 6,000 games.

• The goal of this assignment is to predict if the average score users gave the video game a bad, mediocre, good, or great

score.

# · Tools and Algorithms used

- <u>Tools</u>:
- Jupyter Notebook: integrated development environment (IDE);

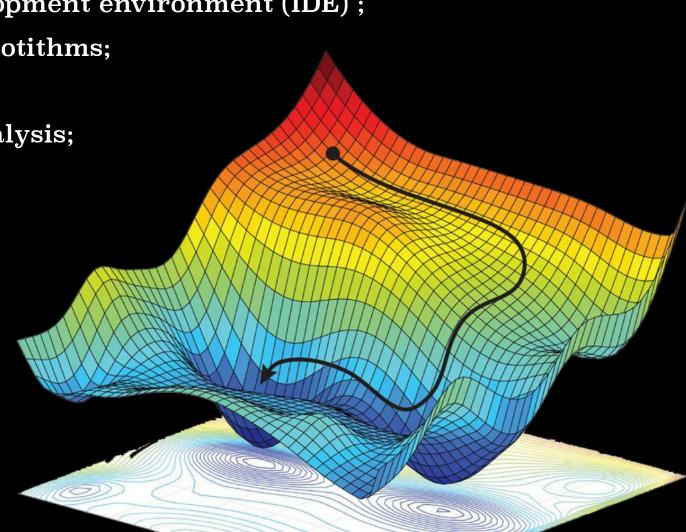
Scikitlearn: Machine Learning Algrotithms;

Seaborn: Plotting;

Pandas: Manipulation and Data Analysis;

o Numpy: Mathematical Functions.

- Algorithms:
- Dummy classifier;
- Tree decision classifier;
- K-NN.



### · Data preprocessing:

Data Normalization;

Removal of unwanted data with missing values;

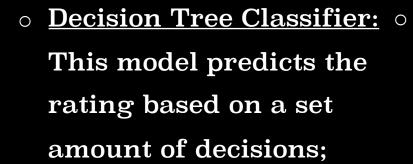
Column	Non-Null Count	Dtype
id	5753 non-null	int64
name	5753 non-null	object
category	5753 non-null	object
dlcs	5753 non-null	int64
expansions	5753 non-null	int64
year	5753 non-null	int64
follows	5753 non-null	int64
in_franchise	5753 non-null	bool

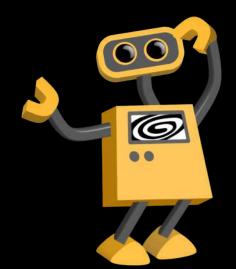
Visualization of data correlations.

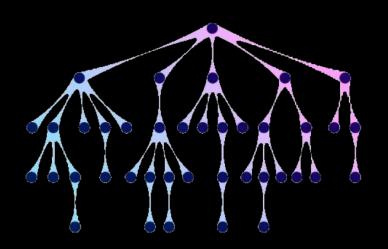


## Used Predicting Models:

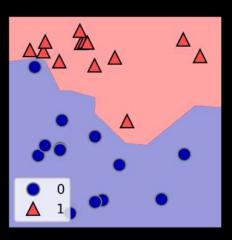
 Dummy Classifier: Used as a baseline to compare with the other models;





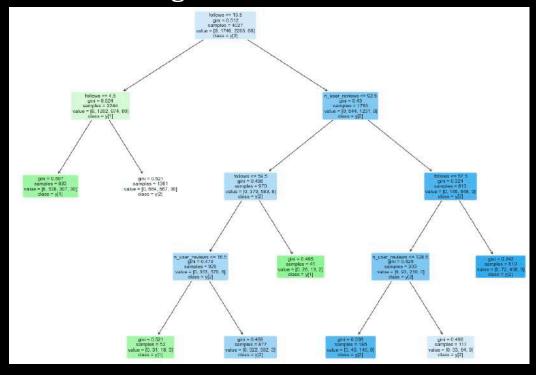


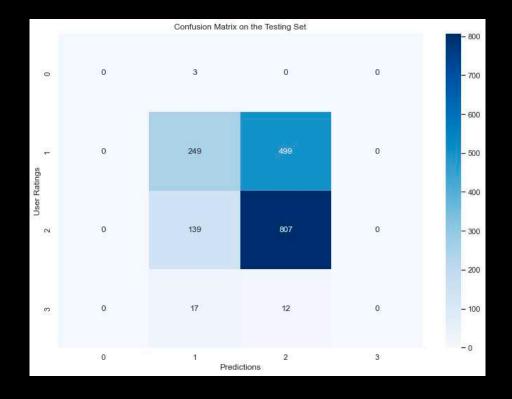
K- Nearest Neighbors
algorithm: The game is
classified by a plurality vote of
its neighbors, with the game's
rating being assigned to the
rating most common among
its k nearest neighbors.



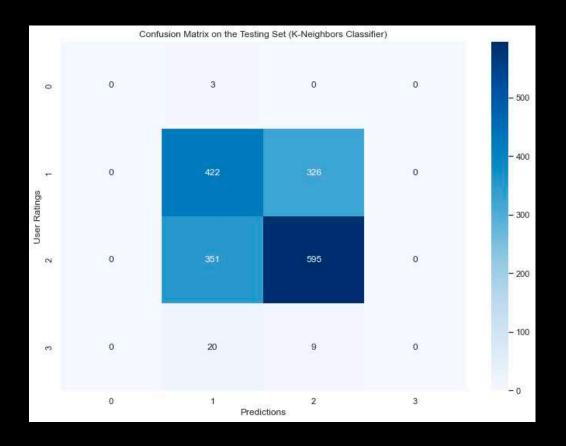
#### Model vizualization and evaluation

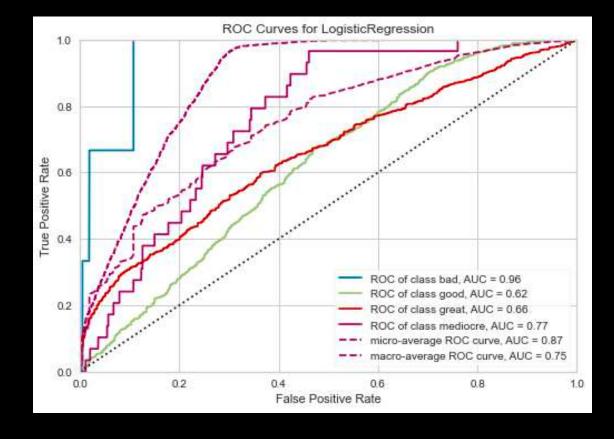
- Dummy Classifier: The baseline accuracy was 55%.
- Tree Decision Classifier:
- ✓ The average accuracy was 61%;
- ✓ It failed the most predicting the worst games (bad and mediocre);
- ✓ The average amount of leaves was 6.





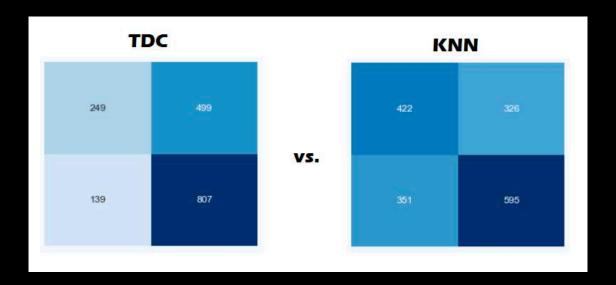
- <u>K-NN</u>:
- ✓ The average accuracy was 62%;
- ✓ It would cluster more accurately between the good and great games;
- ✓ The average amount of neighbors was 85.





## Model Comparison

- o Both Models compared with the dummy classifier did better with around 6-7% higher accuracy;
- The Models had difficulties predicting the rating of the low-end games, but generally predicted better the higher-end games, with K-NN being the best;



- o In terms of code execution time, K-NN took far more time (35.2 s) than the Tree Decision Classifier (0.3 s) because the former has higher expensive real time execution;
- o To conclude we can say that the K-NN had a better perfomance than the Tree Decision Classifier.

#### • References

- Seaborn Documents: <a href="https://seaborn.pydata.org/#">https://seaborn.pydata.org/#</a>;
- ScikitLearn Documents: <a href="https://scikit-learn.org/stable/modules/classes.html">https://scikit-learn.org/stable/modules/classes.html</a>;
- Decision Tree Classifier Tutorial:

https://www.kaggle.com/code/prashant111/decision-tree-classifier-tutorial;

Files available on classroom Moodle.