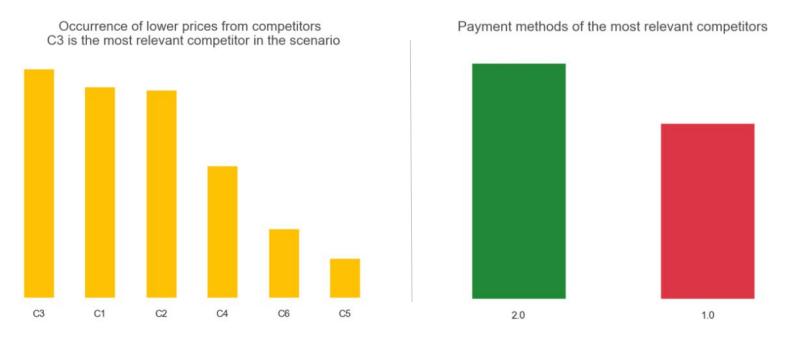
B2W Challenge

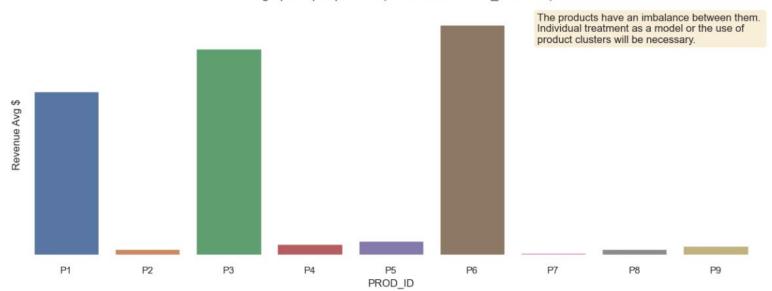
Model forecasting

Relating competitors



Understanding the difference between products

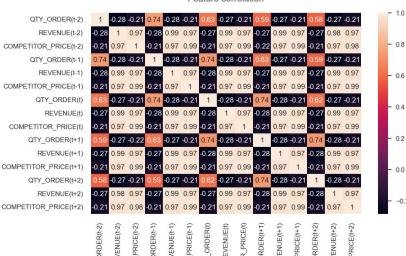
Average price per product (REVENUE / QTY_ORDER)

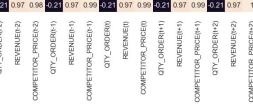


Transformation to supervised model

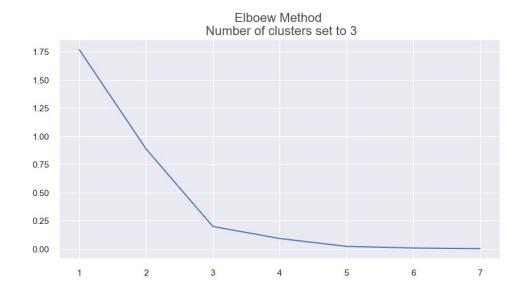
- Various possibilities for parameterization and diversity of models
- Used shift to shift the data, thus creating temporal features (t-2, t-1, t+1, t+2)

Feature correlation



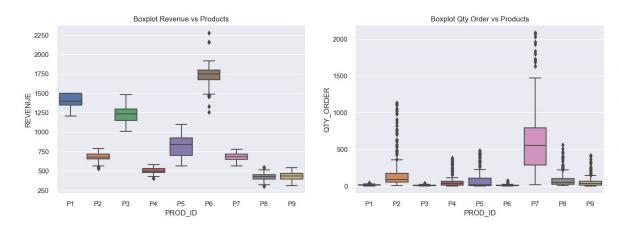


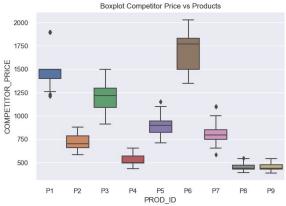
K-means for product clustering



	PROD_ID	QTY_ORDER	REVENUE	LABEL
0	P1	3597.143168	351287.831255	0
1	P2	49132.265996	194956.364284	1
2	P3	2470.633106	308111.805211	0
3	P4	12431.904156	99455.123366	1
4	P5	14940.066835	149408.145137	1
5	P6	2723.437701	491197.768680	0
6	P7	177069.942954	194498.385509	2
7	P8	18049.193527	88309.741122	1
8	P9	11923.209047	91042.659545	1

Outlier handling

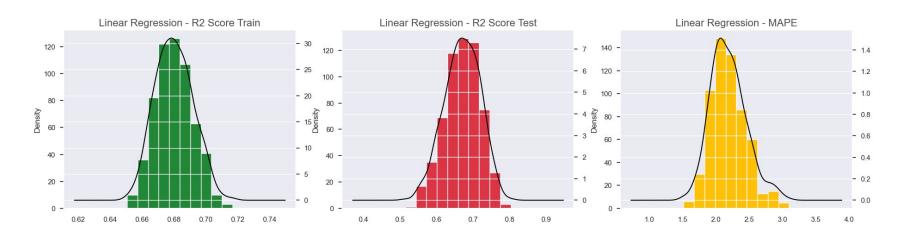




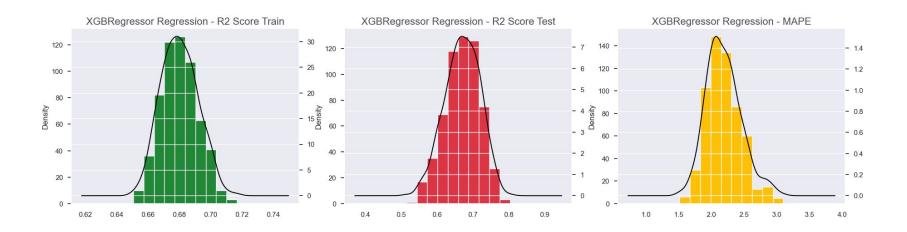
Outliers were filtered based on a standard deviation above 3 in their product dataset. These values (REVENUE) have been replaced by your average.

For COMPETITOR_PRICE the outliers were replaced by the same corresponding REVENUE value.

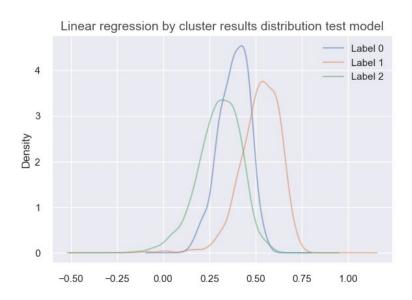
Linear Regression

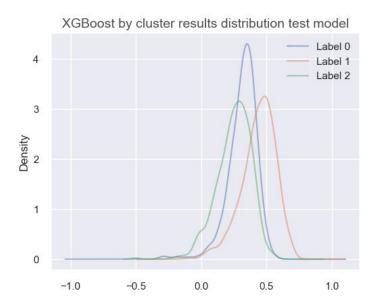


XGBoost

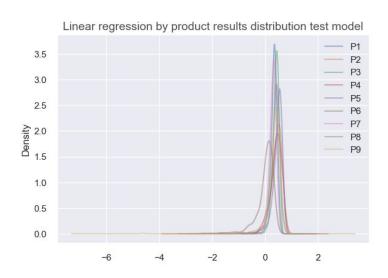


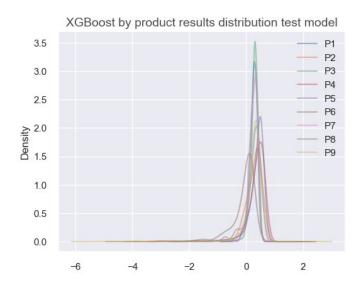
Clustering Regression Model





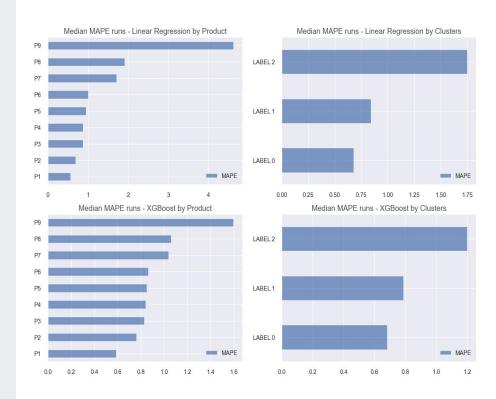
Product Regression Model





Results and Next Steps

- Despite the significant relevance of the results in the execution of the models with all data available in R2 score, there is a problem in the prediction by products analyzed in the MAPE method.
- Both algorithms presented greater prediction difficulties for the products P1, P2 and P3, even resulting in a negative R2 for P1 (worse than a horizontal line).
- Clustering in turn alleviated discrepancies in results when compared to one-to-one products.



Thank You!

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