

PREDICT MEAT PRODUCT CONSUMPTION

PROYECTO FINAL VARINI-BERRONDO



Table of Contents



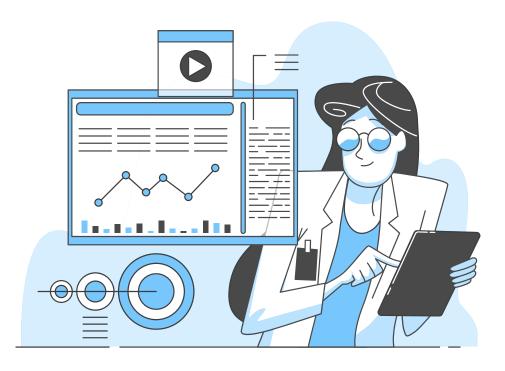


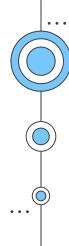






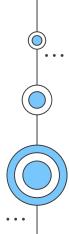


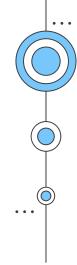




O1 SUPERSTORE

Food retail store!

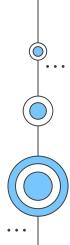


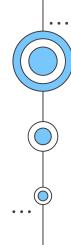


Discover the Universe of "Superstore"

For two years, we have painstakingly captured our clients' secrets and the data they revealed to us.

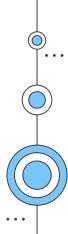
Dive into a world of demographic information where every detail helps us better understand our customers.

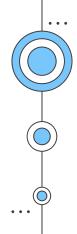




OZOBJECTIVE

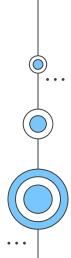
What we want to achieve?

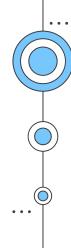




Our goal is clear: increase meat sales and decipher consumer behavior in the face of an exciting marketing campaign. How will we achieve it? Through the magic of data.



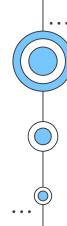




O3 CONTEXT

What we have?

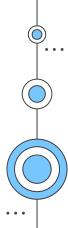


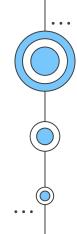


COMMERCIAL CONTEXT

In the current commercial context, increasing the sale of meat can be based on:

- 1) Diversification of the offer to satisfy customer preferences
- 2) Cater to the current consumer base
- 3) Educate about sustainable options and promote a gradual transition
- 4) Look for innovation opportunities in the meat market.

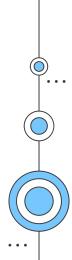


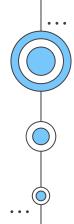


COMMERCIAL PROBLEM

- 1) Identify factors that influence purchasing decisions.
- 2) Detect patterns and groups of clients.
- 3) Data-backed personalized marketing campaigns.

Let's promote meat consumption with a strategic and data-based approach.

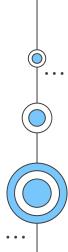




ANALYTICAL CONTEXT

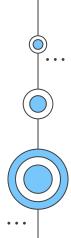
We will take an analytical approach that will examine all the factors that influence our meat eaters. We will explore educational levels, ages, genders, household types.

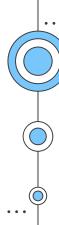
We will also closely analyze the behavior of those customers who prefer other foods. We will look for hidden patterns and trends that give us valuable insights to reposition our business strategies.





04 DATASET





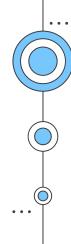
The Dataset contains customer information, including age, education, marital status, income, and purchase history, used to predict customer response to a marketing campaign.

COLUMN

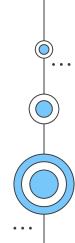
DESCRIPTION

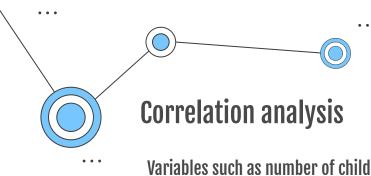
DATATVDE

COLUMN	DESCRIPTION	DATATYPE
ID	Unique ID of each customer	Integer
Year_Birth	Age of the customer	Integer
Complain	1 if the customer complained in the last 2 years	String
Dt_Customer	date of customer's enrollment with the company	DateTime
Education	customer's level of education	String
Marital	customer's marital status	String
Kidhome	number of small children in customer's household	Integer
Teenhome	number of teenagers in customer's household	Integer
Income	customer's yearly household income	Decimal
MntFishProducts	the amount spent on fish products in the last 2 years	Integer
MntMeatProducts	the amount spent on meat products in the last 2 years	Integer
MntFruits	the amount spent on fruits products in the last 2 years	Integer
MntSweetProducts	amount spent on sweet products in the last 2 years	Integer
MntWines	the amount spent on wine products in the last 2 years	Integer
MntGoldProds	the amount spent on gold products in the last 2 years	Integer
NumDealsPurchases	number of purchases made with discount	Integer
NumCatalogPurchases	number of purchases made using catalog (buying goods to be shipped through the mail)	Integer
NumStorePurchases	number of purchases made directly in stores	Integer
NumWebPurchases	number of purchases made through the company's website	Integer
NumWebVisitsMonth	number of visits to company's website in the last month	Integer
Recency	number of days since the last purchase	Integer
Response	1 if customer accepted the offer in the last campaign, 0 otherwise	Integer
Complain	1 if the customer complained in the last 2 years	Integer



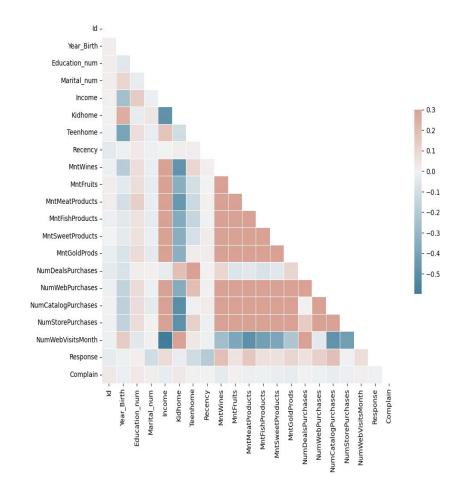
O5 EXPLORATORY DATA ANALYSIS

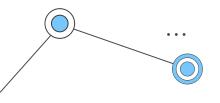




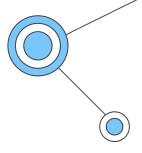
Variables such as number of children, year of birth, and adolescents in the home would not show a correlation with our impact variables (type of consumption), so they would not be relevant to explain the purchasing behavior of our customers.

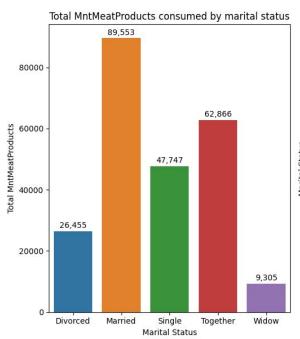
On the other hand, regarding the purchase or consultation channel, a strong correlation is observed, with a higher prevalence of certain products depending on the channel of origin.

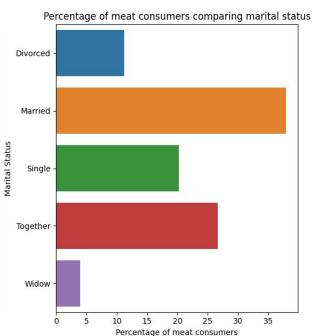


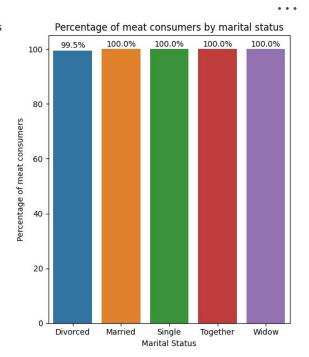


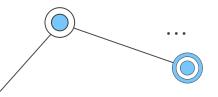
What are the marital states that consume more meat and its percentage?









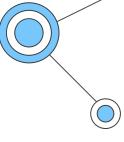


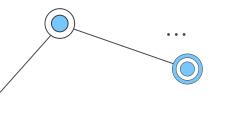
Marital Status

It indicates that the majority of individuals across all marital statuses are meat consumers, with slight variations observed among the categories.

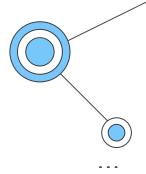
Also, It can be observed that Married individuals tend to consume the highest amount of meat products compared to other marital statuses, followed by Together and Single

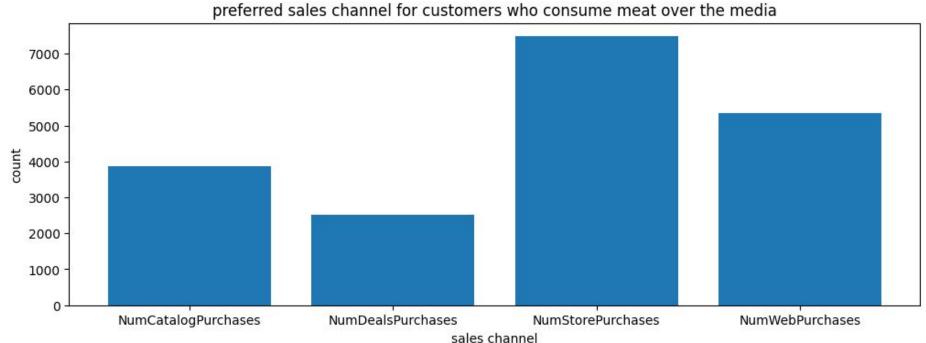
This suggests that marital status plays a role in determining meat consumption patterns.

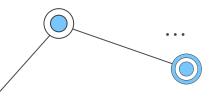




What is the preferred sales channel for customers who consume meat and are over the mean value?



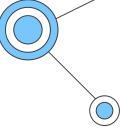


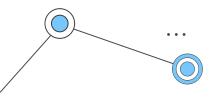


Sales channel

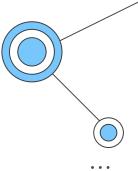
Among the identified sales channels, customers who consume meat above the average tend to make a relatively higher count of purchases through the NumStorePurchases and NumWebPurchases channels.

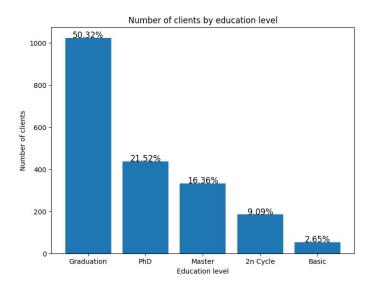
This implies that these channels are influential in driving meat product sales and should be targeted in marketing strategies.

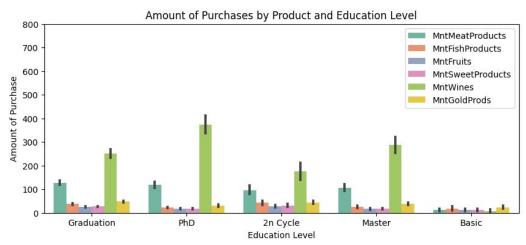


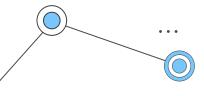


What kind of education do customers who consume meat have?





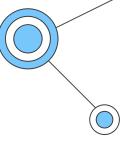


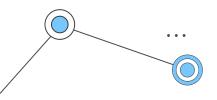


Education

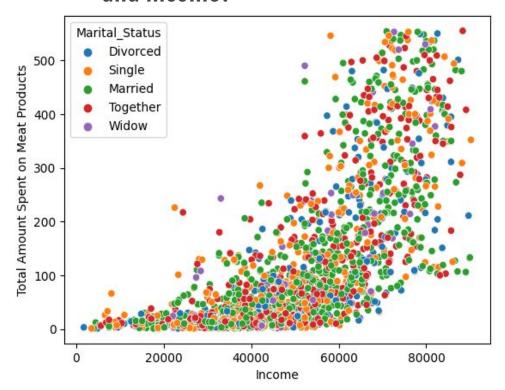
There is a clear difference between the different types of education, being Graduation the one with the highest number of client followed by PhD and Master, but we can see that expecting Basic all the other education types consume same percentage of meat.

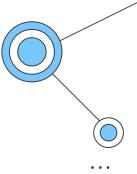
This indicates that education level may not have a significant impact on meat consumption preferences. We can also observe and separate the largest number of users who consume in the store for future analysis.

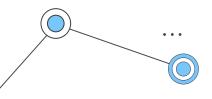




What relation is between meat consumption and income?

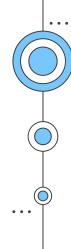






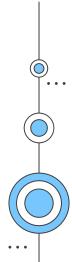
Income

First of all, income has a significant influence on the total amount spent on meat products. People with higher incomes tend to allocate a greater part of their resources to meat consumption. However, consumption is also observed in those with low incomes. Therefore. This indicates that meat consumption is not limited to a specific income group and that there are opportunities to attract customers of different economic levels.



O6 MODELING

Trying to predict



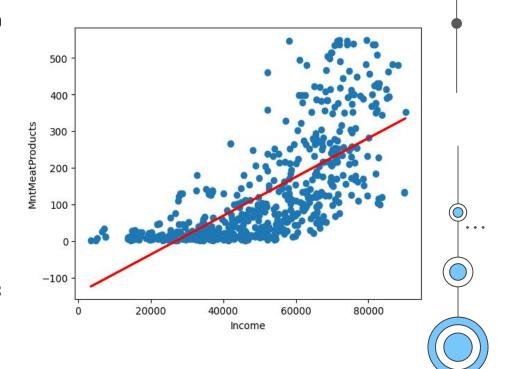


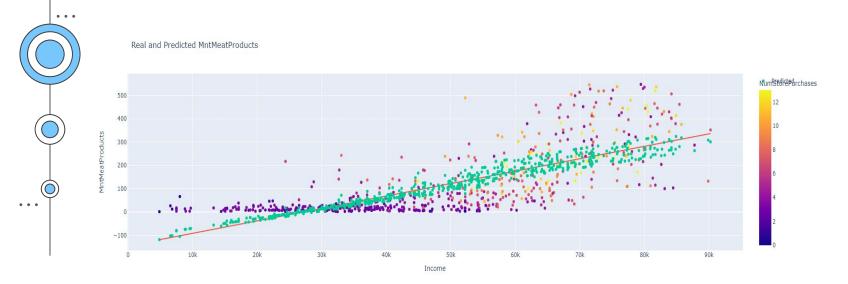
LINEAL REGRESSION

We are going to work with a linear regression model because it establishes a linear relationship between the dependent variable (meat consumption) and the independent variables (characteristics or predictor variables). If meat consumption is expected to increase or decrease proportionally to certain predictor variables, the linear regression model is appropriate.

The R-squared value is 0.50, which means that approximately 50% of the variance in the amount spent on meat products can be explained by the income variable.

However, it's important to note that the model's performance can be further improved, as there is still a significant amount of unexplained variance in the data (as indicated by the MSE and R-squared values).





By adding the variable 'NumStorePurchases', we achieved a higher R-squared value of 0.5424, indicating that it explains a larger proportion of the variance in meat product consumption compared to the first model's R-squared value of 0.50. Additionally, has a lower mean squared error (MSE) of 9165.63, indicating smaller prediction errors compared to the first model's MSE of 9992.53.

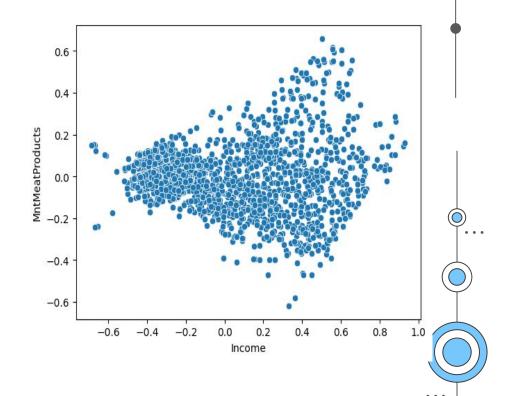
In conclusion, the second model, which incorporates both income and the number of store purchases as predictors, demonstrates superior performance in predicting meat product consumption compared to the first model that solely relied on income. The second model exhibits a higher R-squared value, lower MSE, and lower MAPE, indicating a better fit to the data and more accurate predictions. The inclusion of relevant additional variables in the modeling process can significantly improve the predictive power and accuracy of the model.



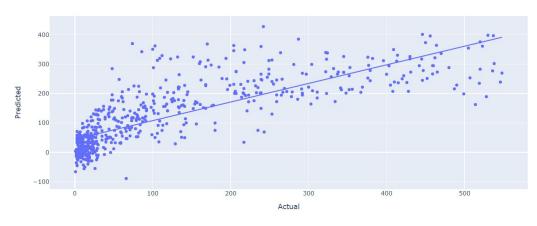
PCA (Principal Component Analysis)

The PCA results provide insights into the variance explained by each component and the cumulative variance explained by the selected number of components. This analysis aids in dimensionality reduction and understanding the relative importance of the components in capturing the variability within the data.

The results show that applying PCA with 2 components explains approximately 81.42% of the total variance of the data. The first principal component explains 67.9% of the total variance and the second principal component explains 13.51%.



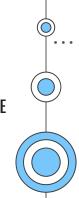


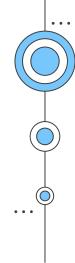


The R-squared value is 0.63. This indicates that the model captures about 63% of the variability in the data, suggesting a moderate level of predictive power. The mean squared error (MSE) is 9165.63 The mean absolute percentage error (MAPE) is 2.71.

Based on these results, we can conclude that the linear regression model, trained on the principal components obtained from PCA, demonstrates a reasonable level of predictive performance. The R-squared value indicates that the model captures a significant portion of the target variable's variability.

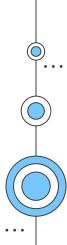
However, it is important to note that there is still room for improvement, as indicated by the moderate MSE and MAPE values. Further refinement of the model or exploration of alternative algorithms may lead to better predictive accuracy. Additionally, it is worth considering the interpretability of the transformed features derived from PCA and their impact on the model's predictions.





O7 CONCLUSIONS

Our insights



After conducting a thorough analysis to increase meat sales by predicting meat product consumption in response to a new marketing campaign, we have gained valuable insights

01

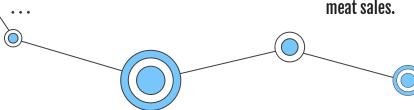
We observed that customers with higher incomes and who made more purchases through various channels, such as web and catalog, tend to consume more meat products.

02

Moreover, customers with higher education levels showed a greater preference for meat products.

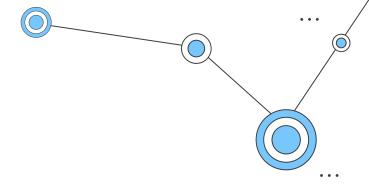
03

We found positive correlations between meat product consumption and variables such as income, number of purchases, and educational level. These factors can significantly influence meat sales.



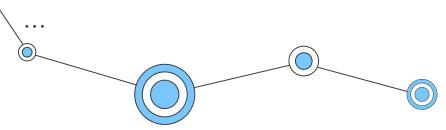
04

Using linear regression models, we predicted meat product consumption based on various customer attributes. We evaluated the models using metrics such as R-squared value, mean squared error (MSE), and mean absolute percentage error (MAPE). The results indicated that the models had moderate predictive power, capturing a significant portion of the variability in meat consumption.



05

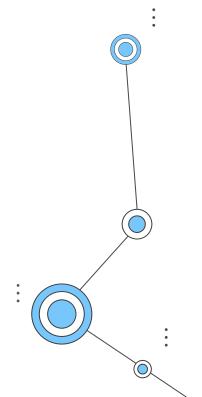
After applying the Principal Component Analysis (PCA) we reduce the dimensionality of the data and identify key components explaining the variance in meat consumption. The two principal components obtained explained around 81.43% of the variance in the data.





In conclusion, understanding the relationships between marital status, sales channels, education, and income provides valuable insights for marketing strategies aimed at increasing meat sales.

Targeting married individuals, leveraging effective sales channels, and considering income levels can help optimize marketing campaigns and product offerings to cater to the preferences and behaviors of the target audience.



Thanks!

CREDITS: This presentation template was created by Slidesgo, including icons by Flaticon, infographics & images by Freepik and illustrations by Stories

