

Customer Segmentation Project Virtual Internship

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Group Information

Group Name: cust_seg

Specialization: Data Science

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Internship Batch: LISUM 12

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Problem Description

XYZ bank wants to roll out Christmas offers to their customers. But Bank does not want to roll out the same offer to all customers. Instead, they want to roll out personalized offers to a particular set of customers. If they manually start understanding the category of the customer then this will be not efficient and also they will not be able to uncover the hidden pattern in the data (pattern which group certain kind of customer in one category). Bank approached ABC analytics company to solve their problem. Bank also shared information with ABC analytics that they don't want **more than 5 groups** as this will be inefficient for their campaign.

ABC analytics assigned this talk to their analytics team and instructed their team to come up with the approach and feature which group similar behavior in one category and others in a different category.

Data Information and Goals

Total Customers: 1000000

Goal: We want to segment customers based on their behavior. The main things we are interested in for our Exploratory Data Analysis are:

- 1. What products are most used by the customers?
- 2. What channel has the most amount of customers?
- 3. What age range has the most customers? Does the age ranges have an effect on the types of products being bought?
- 4. What products are most used based on the seniority of the customer from when they started?
- 5. Does the city/province of the customer have an effect on the products that are bought or used?
- 6. What effect does gender have on the products being bought?
- 7. Income by city, age, seniority, and gender

Product Distribution

 Overall, most customers have a Current Account, but also could have a Particular Account, along with Direct Debit, Payroll, and an e-account.

Table of overall product distribution

```
Saving account = 0.01508%
Guarantees = 0.00372%
Current account = 78.16387%
Derivada account = 0.05072%
Payroll account = 10.17122%
Junior account = 1.30998%
Most particular account = 1.36756%
Particular account = 17.50098%
Particular plus account = 6.04837%
Short-term deposits = 0.23912%
Medium-term deposits = 0.27183%
Long-term deposits = 6.16176%
e-account = 10.23683%
Funds = 2.35362\%
Mortage = 0.85348%
Pensions_1 = 1.26572\%
Loans = 0.33371\%
Taxes = 6.27887\%
Credit card = 6.499%
Securities = 3.34419%
Home account = 0.55247%
Payroll = 7.32761\%
Pensions_2 = 8.06338%
Direct debit = 16.89642%
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Channel Distribution

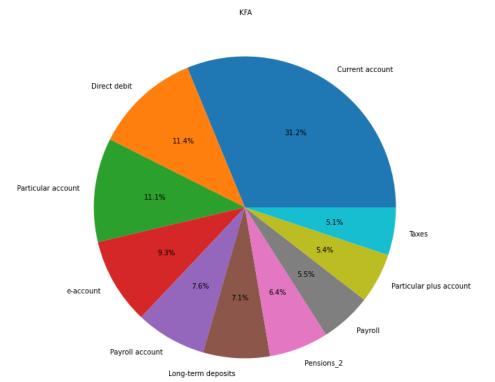
 From the table, we see that channel "KAT" is the most popular channel being used by the customers.

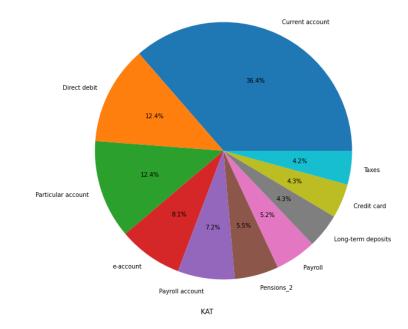
Table of the 10 most popular channels being used

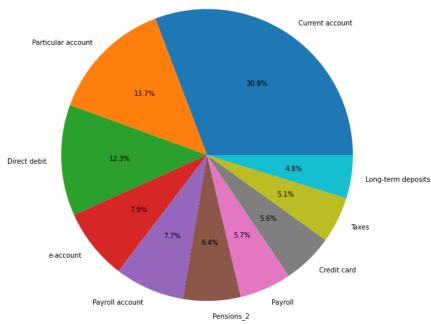
KAT	262959
KFC	215771
KHE	201024
KFA	32524
KAS	7087
KAG	6847
KAA	5342
KAB	5245
KAY	4992
KHK	4823

Channel Distribution

 For the distribution of products based on channel, we can see that for the top 3 channels that take up the most amount of data, there is no significant difference of products between customers.

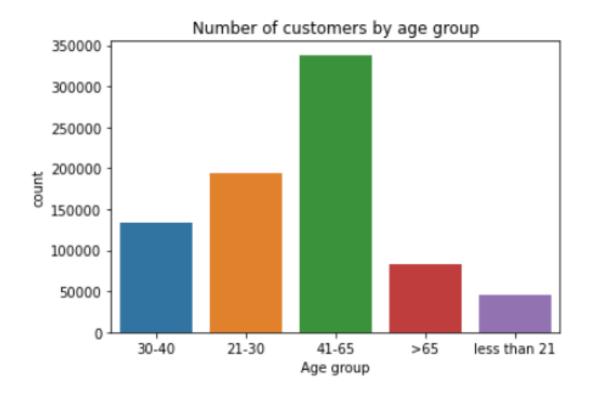






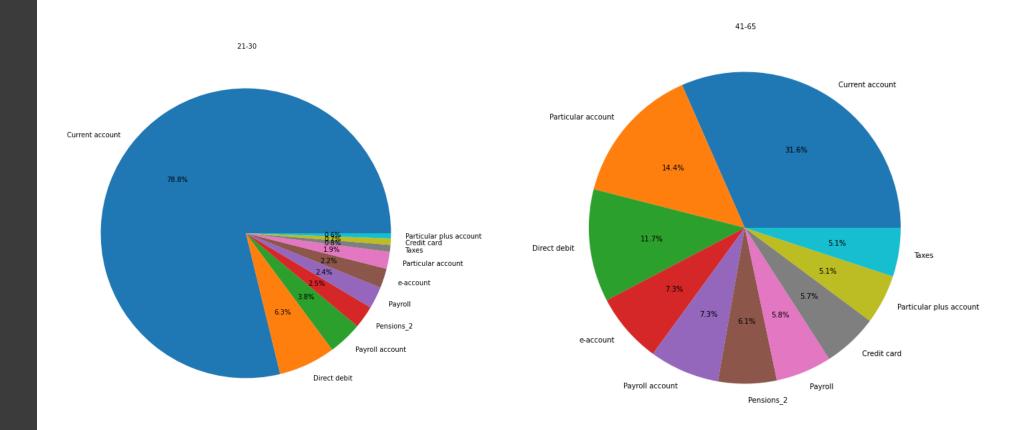
Age Distribution

 For the general customer distribution, people aged 41-65 make up the biggest portion of total customers.



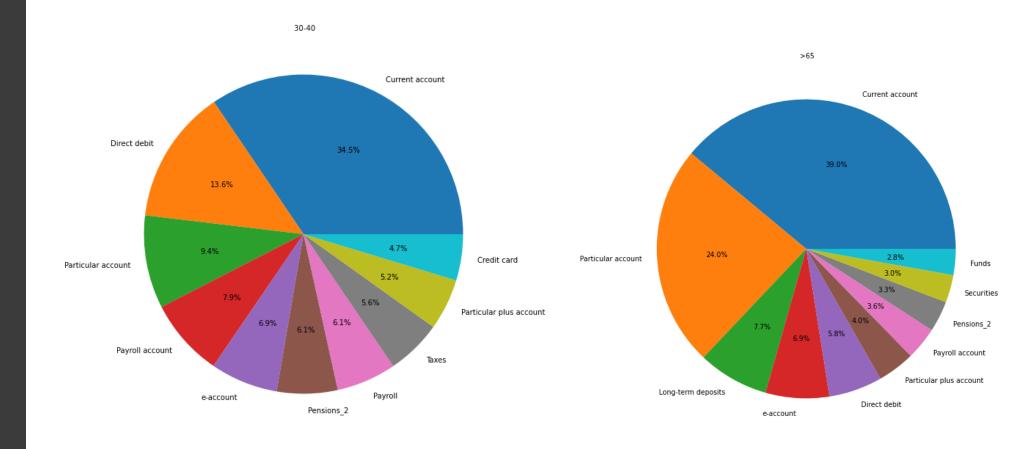
Age: Product Distribution

- However, customers who are young adults (under 21 and 21-30)
 make up the biggest proportion of purchases with a Current
 account compared to those aged 41-65.
- Also, not shown here, but a Junior account is a popular product used by those who are under 21.



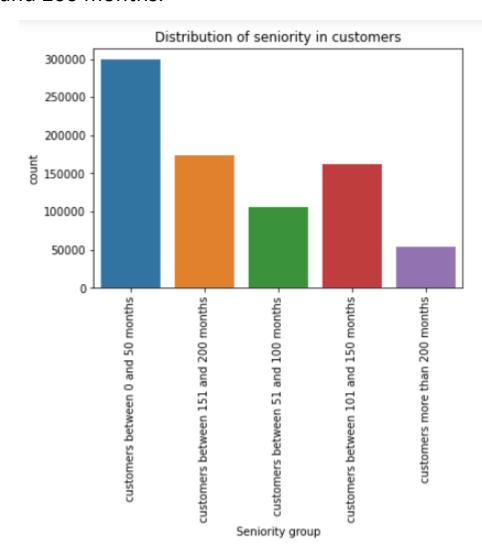
Age: Product Distribution

- For customers who are over 65, the other most popular account for them is Particular account.
- Direct debit is popular among individuals aged 30-40.



Seniority of customers

- Most customers are brand new, meaning they have been there less than 50 months (or 4 years).
- The next popular groups are those who have been at company between 100 and 200 months.

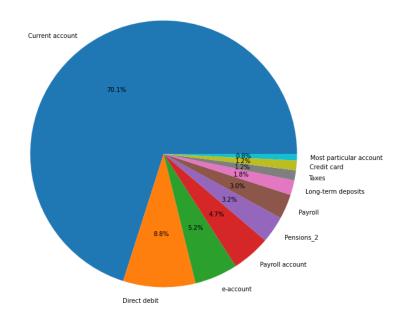


Seniority of customers: Product Distributio n

- The amount of customers who have a current account decreases as the time they have been a member increases.
- Customers who have been there 4
 years or less are most likely to have
 a Current account.

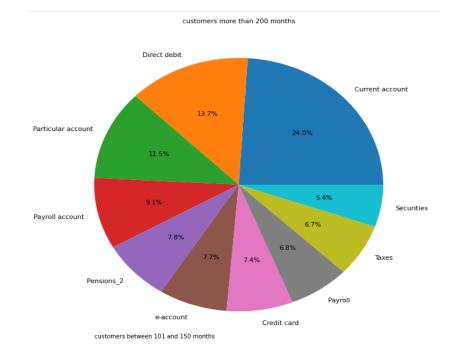
0 and 50 months (~4 years)

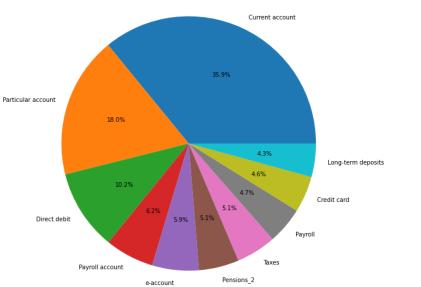
customers between 0 and 50 months



101 and 150 months

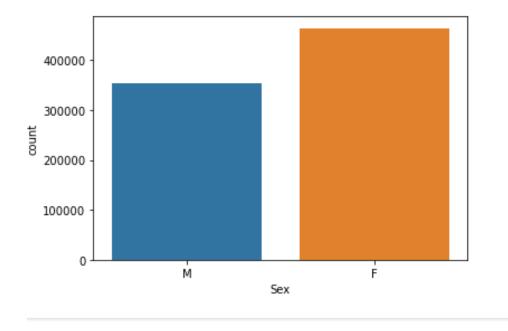
> 200 months





Customers by gender

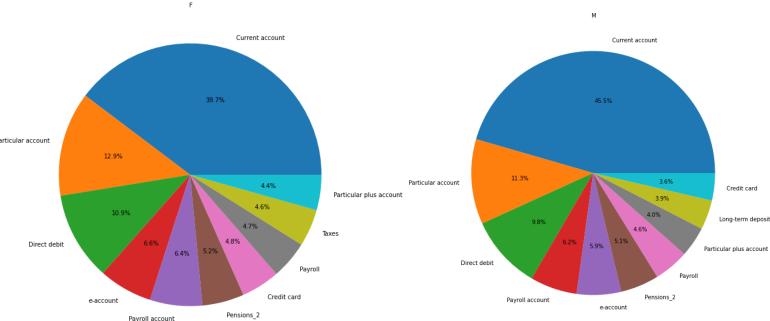
• Overall, females are the most popular customer gender.



Customers by gender: Product Distribution

Sex	F	M	
Saving account	113.0	40.0	
Guarantees	26.0	11.0	
Current account	342323.0	270689.0	
Derivada account	416.0	71.0	
Payroll account	54990.0	36653.0	
Junior account	6085.0	5663.0	
Most particular account	4830.0	3309.0	
Particular account	110931.0	67061.0	
Particular plus account	37421.0	23802.0	
Short-term deposits	974.0	680.0	
Medium-term deposits	1543.0	1107.0	
Long-term deposits	34742.0	22836.0	
e-account	56494.0	34764.0	
Funds	15834.0	7125.0	
Mortage	6188.0	2494.0	
Pensions_1	7886.0	4740.0	Partio
Loans	2295.0	1065.0	
Taxes	39577.0	21086.0	
Credit card	37272.0	19537.0	
Securities	23401.0	9493.0	
Home account	3498.0	2114.0	
Payroll	37079.0	25392.0	

 Overall, females have more products than males, but although the difference isn't significant, there are a little bit more males that have a Current account than females.



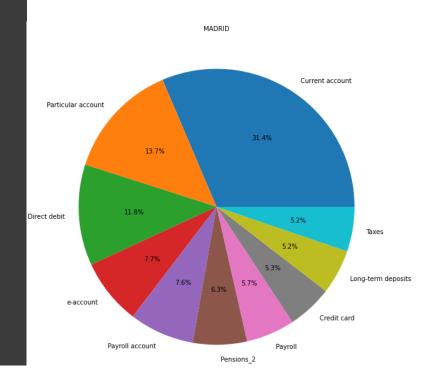
Customers by country and city

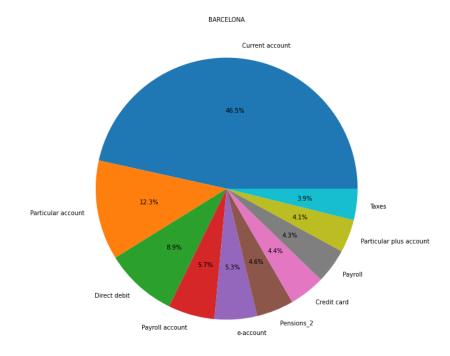
 Based on the tables below, we can see that almost all customers are from Spain, along with Madrid being the most popular city/province.

Top countri	es of customers	Top 10 cities of cus	stomers
ES MX PA BE IT DE BO PY	818243 4 2 2 2 2 2 2	MADRID BARCELONA VALENCIA SEVILLA ZARAGOZA MALAGA CORUÑA, A MURCIA CADIZ ALICANTE	329081 74838 39516 38786 21882 21501 20801 19378 16692 16198

Customers by city: Product Distribution

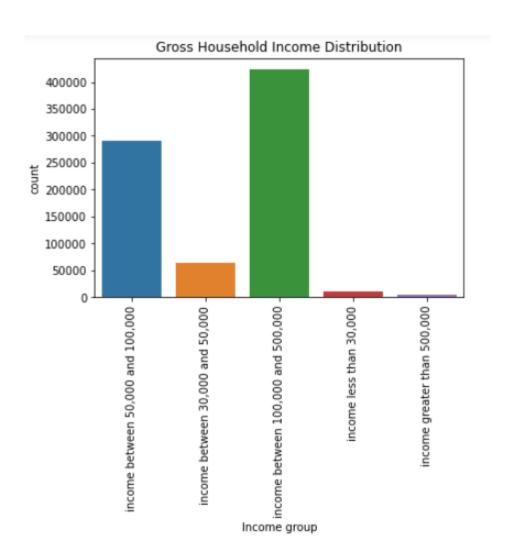
 Based on the pie charts, we can see that the city/province can have a significant effect on the product distributions from customers.





Customers by income

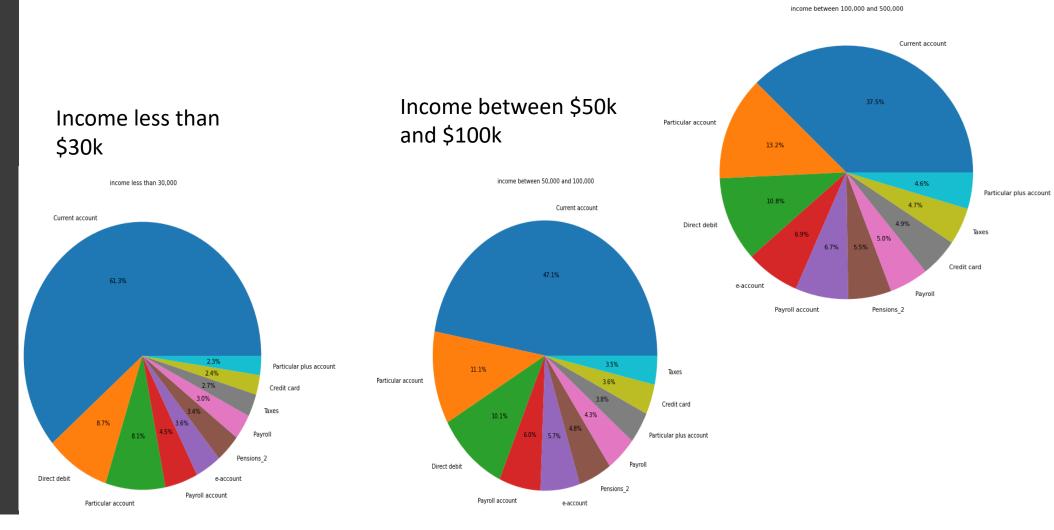
• From the Bar graph, we can see that customers usually make between \$100,000 and \$500,000.



Customers by income: Product Distribution

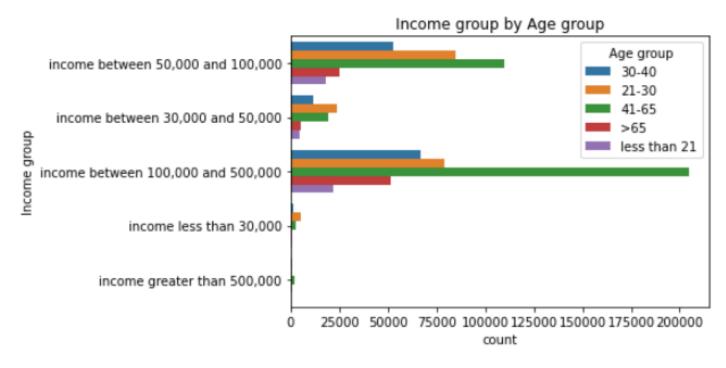
• Based on the pie charts, we can conclude that there is a significant difference between the product distributions based on the gross income of customers.

Income between \$100k and \$500k



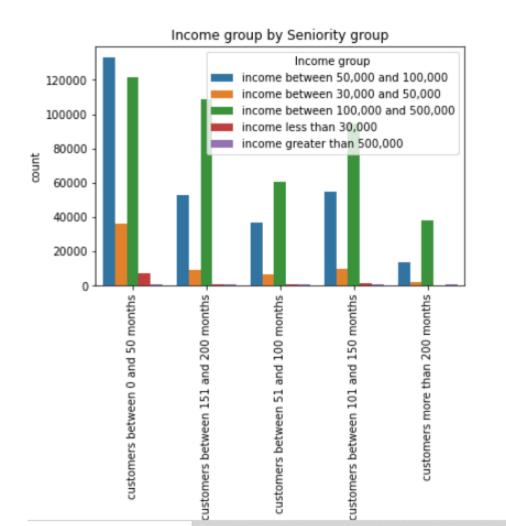
Income by age group

- Customers who are young adults (21-30) are most likely to have a gross household income of between \$50,000 and \$100,000.
- Customers who are middle aged (41-65) are more likely to have a gross household income greater than \$100,000 but less than \$500,000. This would most likely make sense as this age group has been working for a while.



Income by seniority

• Customers who have gross household income between \$50,000 and \$100,000 most likely have been a customer for less than 4 years, along with those who have income between \$100,000 and \$500,000.



Income by city

 Overall, Madrid was the most popular city by customers for most income groups. However, it was shown that for customers whose gross household income was less than \$30,000, other cities had a higher number of customers.

Province name ALBACETE	Income group	income greater	than 500,000	income less th	an 30,000
ALICANTE ALMERIA ASTURIAS ASTURIAS AVILA AVILA BADAJOZ BALEARS, ILLES BADAGOS BARCELONA BIZKAIA BIRGOS BARCELORA BURGOS BARCELOR CACERES BRODE CADIZ CANTABRIA CASTELLON BIZKAIA BOBADA CASTELLON BIRGOS BRODE CADIZ BRODE CADIZ BRODE CADIZ BRODE CASTELLON BRODE CORUÑA, A B					
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LUGO 0 97 MADRID 3092 382 MALAGA 71 206 MELILLA 4 0 MURCIA 2 562	LEON		4		137
MADRID 3092 382 MALAGA 71 206 MELILLA 4 0 MURCIA 2 562	LERIDA		3		133
MALAGA 71 206 MELILLA 4 0 MURCIA 2 562	LUG0		0		97
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MURCIA 2 562	MALAGA		71		206
	MELILLA		4		0
NAVARRA Ø Ø	MURCIA		2		562
-	NAVARRA		0		0

Final Recommendations

In order to segment our customers into 5 groups for our modeling which is the next step we can use the following recommendations to help.

- We can focus on ages that are older than 18-30 as that is where most of the customers are.
- Males have lower product usage than females, so giving away special offers to males could be a strategy.
- Since Spain has the most amount of customers overall (mainly just Spain for the data we have) along with Madrid being the most popular city, the bank can focus on other cities/areas in Spain.
- The bank can focus on individuals separately with regards to their seniority with certain offers given to people with middle and higher seniority and then to people with lower seniority.

Modeling Choices

- Since we will be working with no dependent variables, we will use unsupervised learning methods for our modeling.
- Since we are wanting to divide customers into different groups that share similar product and other demographic behavior in order to break up the groups for special Christmas offers, we will most likely be using clustering, which is when we divide data points up into a certain number of groups so that the data points in same group are similar to the other data points in same group than other groups.

K-Means Clustering: For K-Means Clustering, the goal of this method is to partition n observations into k clusters in which every observation belongs to the cluster with the nearest mean.

Hierarchical Clustering: For Hierarchical Clustering, the goal of this method is build a hierarchy of clusters.

- We will use both methods and compare for validation to finalize our analysis.

Thank You

