

# Data Science Project

## Week 9 Deliverable

**Group name:** Data Explorers

**Specialization:** Data Science

**Batch:** LISUM12

### Team members Details:

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**Project Title:** Bank Customer Segmentation

**Problem Description:** A bank wants to create customer segments/categories to send personalized Christmas offers to its customers. We need to identify feature(s) to group customers into different categories.

Customer segmentation is an approach to creating smaller customer groups relevant to the bank's product marketing and services. Based on customer accounts and usage details, sending different offers to customers increases the chances of increasing the business for a bank and reaching maximum customers as possible.

We need to create at most five customer segments. We plan to employ Data Science techniques to understand, pre-process and create customer segments.

### Data Source:

<https://drive.google.com/drive/folders/1bfCpJIKmp6IHxiLPWvOS2nU1dc24pViB>

**GitHub Repo link:** [https://github.com/jalpa015/DataScience\\_Project/tree/master/week9](https://github.com/jalpa015/DataScience_Project/tree/master/week9)

## **Data Understanding:**

- The dataset consists of 47 columns.
- The dataset consists of bank customer information. It consists of different account details, customer information (age, address, gender), and information on banking services (pensions, loans, credit cards, etc.)
- The data contains information about 1,000,000 customers of XYZ bank.
- The employee information (Present, ex-employee, active, passive) is present for each customer.
- Customer Details available are –
  - Age
  - Gender
  - date of joining the bank
  - customer type (owner, co-owner)
  - customer relation (active, inactive, potential, former)
  - Residence
  - Foreigner (different birth country)
  - Deceased
  - Address
  - Gross income
- Customer account and services information present in the dataset is as follows –
  - Saving account
  - Guarantees
  - Current Account
  - Derived Account
  - Payroll account
  - Junior account
  - Particular account, particular plus account
  - Short, medium, long-term deposits
  - E-account
  - Funds
  - Mortgage
  - Pensions
  - Loans
  - Taxes
  - Credit card
  - Securities
  - Home account
  - Direct debit
- As the account types and services are present in different columns, there is a possibility that not all customers would opt for all types of accounts and services.

- The data information column wise is as follows –

Data columns (total 48 columns):

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	Unnamed: 0	1000000 non-null	int64
1	fecha_dato	1000000 non-null	object
2	ncodpers	1000000 non-null	int64
3	ind_empleado	989218 non-null	object
4	pais_residencia	989218 non-null	object
5	sexo	989214 non-null	object
6	age	1000000 non-null	object
7	fecha_alta	989218 non-null	object
8	ind_nuevo	989218 non-null	float64
9	antiguedad	1000000 non-null	object
10	indrel	989218 non-null	float64
11	ult_fec_cli_1t	1101 non-null	object
12	indrel_1mes	989218 non-null	float64
13	tiprel_1mes	989218 non-null	object
14	indresi	989218 non-null	object
15	indext	989218 non-null	object
16	conyuemp	178 non-null	object
17	canal_entrada	989139 non-null	object
18	indfall	989218 non-null	object
19	tipodom	989218 non-null	float64
20	cod_prov	982266 non-null	float64
21	nomprov	982266 non-null	object
22	ind_actividad_cliente	989218 non-null	float64
23	renta	824817 non-null	float64
24	ind_ahor_fin_ult1	1000000 non-null	int64
25	ind_aval_fin_ult1	1000000 non-null	int64
26	ind_cco_fin_ult1	1000000 non-null	int64
27	ind_cder_fin_ult1	1000000 non-null	int64

28	ind_cno_fin_ult1	1000000	non-null	int64
29	ind_ctju_fin_ult1	1000000	non-null	int64
30	ind_ctma_fin_ult1	1000000	non-null	int64
31	ind_ctop_fin_ult1	1000000	non-null	int64
32	ind_ctpp_fin_ult1	1000000	non-null	int64
33	ind_deco_fin_ult1	1000000	non-null	int64
34	ind_deme_fin_ult1	1000000	non-null	int64
35	ind_dela_fin_ult1	1000000	non-null	int64
36	ind_ecue_fin_ult1	1000000	non-null	int64
37	ind_fond_fin_ult1	1000000	non-null	int64
38	ind_hip_fin_ult1	1000000	non-null	int64
39	ind_plan_fin_ult1	1000000	non-null	int64
40	ind_pres_fin_ult1	1000000	non-null	int64
41	ind_reca_fin_ult1	1000000	non-null	int64
42	ind_tjcr_fin_ult1	1000000	non-null	int64
43	ind_valo_fin_ult1	1000000	non-null	int64
44	ind_viv_fin_ult1	1000000	non-null	int64
45	ind_nomina_ult1	994598	non-null	float64
46	ind_nom_pens_ult1	994598	non-null	float64
47	ind_recibo_ult1	1000000	non-null	int64

dtypes: float64(9), int64(24), object(15)

### **NULL Values –**

- The null values columns are Employee Index, Customer country of residence, gender, date of customer joining, new customer index, Primary customer, late date for primary customer, customer type, address details (province, type), gross income, nominee payroll and pensions. The details are as follows –

ind_empleado	10782
pais_residencia	10782
sexo	10786
fecha_alta	10782
ind_nuevo	10782
indrel	10782
ult_fec_cli_1t	998899
indrel_lmes	10782
tiprel_lmes	10782
indresi	10782
indext	10782
conyuemp	999822

canal_entrada	10861
indfall	10782
tipodom	10782
cod_prov	17734
nomprov	17734
ind_actividad_cliente	10782
renta	175183
ind_nomina_ult1	5402
ind_nom_pens_ult1	5402

### **Data Distribution** –

Data distribution helps us determine if any outliers are present in the data and how data is arranged.

The skewness of data helps understand the probability distribution of the data. It can be performed on financial data. The distribution for the gross income of customers is not a bell shaped distribution. There are many customers whose income lies between \$1202 – \$300,000. The maximum income recorded is 28894395.

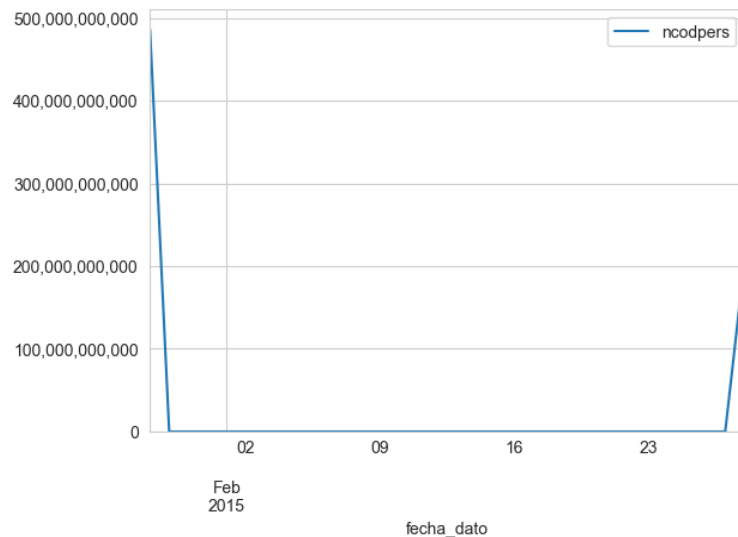
The distribution of numeric data is identified as follows –

#### **Distribution of numeric data**

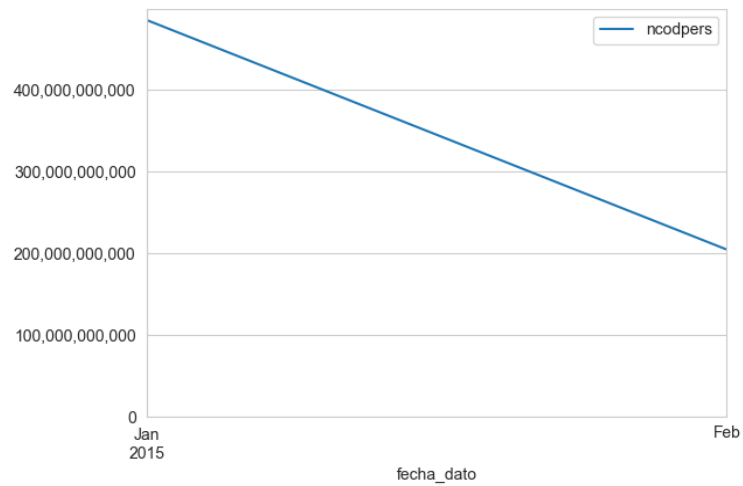
The time series classification of data is computed for the number of customers for the date when customer joined bank and the date the records were fetched. The results as follows –

To check the time series of numeric data by daily, monthly and yearly frequency

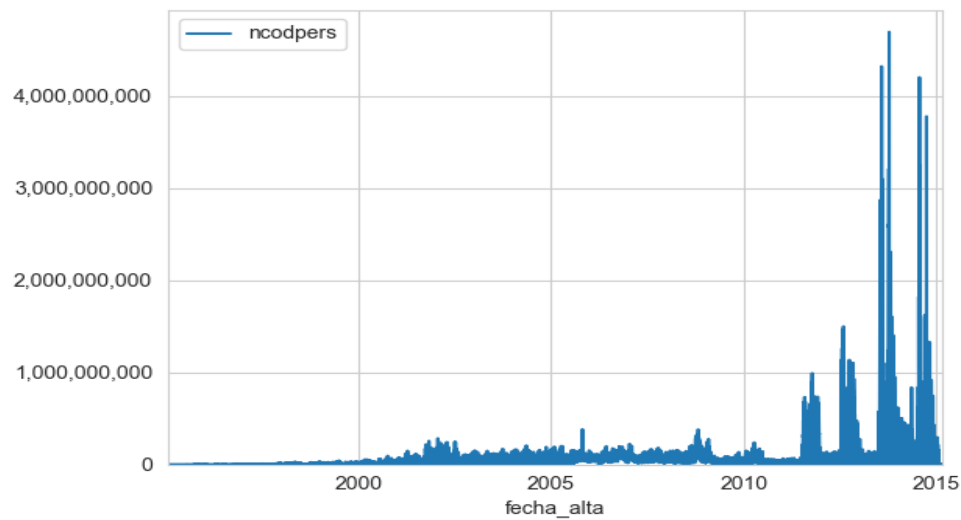
Plotting daily data



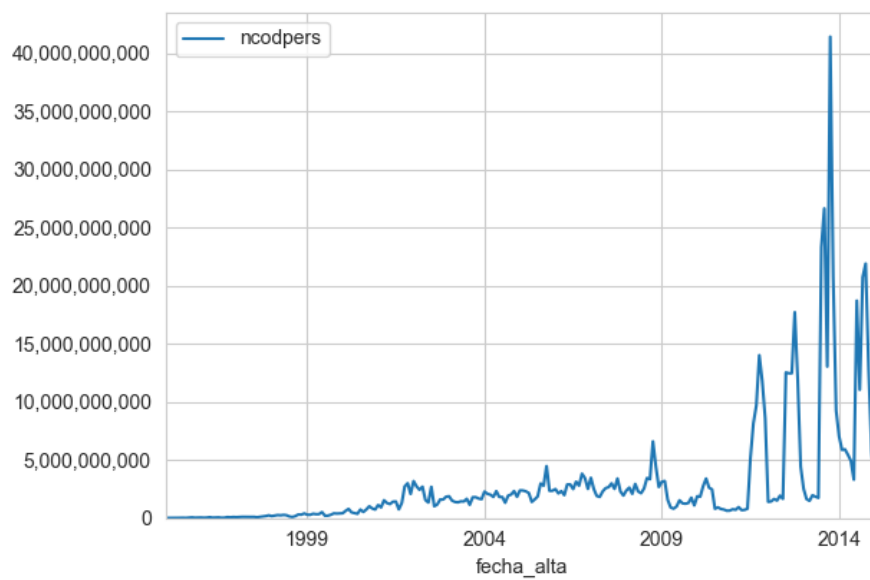
Plotting monthly data



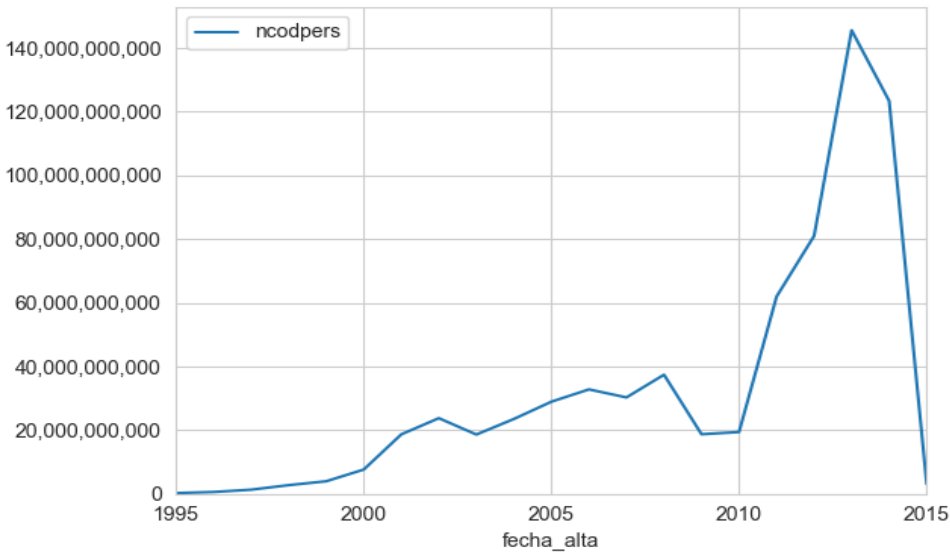
Plotting daily data



Plotting monthly data



Plotting yearly data



As observed in the graphs, the account creation of customers is not equally distributed. There was a significant increase in customers between 2010 and 2015.

Approach for Data cleaning for the upcoming week –

1. Columns with account details have Null values or 0; however, not all customers have all types of accounts and services. Hence, we can still consider information from those columns.
2. In columns such as account creation date and income, the null values can maybe be imputed by considering the mean values. We will conduct experiments to see what options fit the best.
3. We also intend to conduct experiments on treating outliers. In the initial study of data, outliers have been observed. However, by conducting a study about the bank sector and analyzing the outlier data, we decide how to treat the outlier (trimming or imputation with values).

The code for the task can be found at -

[https://github.com/jalpa015/DataScience\\_Project/blob/master/EDA.ipynb](https://github.com/jalpa015/DataScience_Project/blob/master/EDA.ipynb)