# **Exploratory Data Analysis**

**Customer Segmentation** 

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### **Problem**

#### **Customer Segmentation**

XYZ bank wants to roll out Christmas offers to their customers. But the bank does not want to roll out the same offer to all customers, instead they want to roll out personalized offers to particular sets of customers. If they manually start understanding the category of customer then this will not be efficient and also they will not be able to uncover the hidden pattern in the data ( pattern which groups certain kinds 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.

### What kind of data does each column hold in the raw data?

```
Unnamed: 0: [ 0 1 2 ... 999997 999998 999999]

fecha_dato: ['2015-01-28' '2015-02-28']

ncodpers: [1375586 1050611 1050612 ... 1149999 1150908 1183305]

ind_empleado: ['N' nan 'A' 'B' 'F' 'S']

pais_residencia: ['ES' nan 'CA' 'CH' 'CL' 'IE' 'AT' 'NL' 'FR' 'GB' 'DE' 'DO' 'BE' 'AR' 'VE' 'US' 'MX' 'BR' 'IT' 'EC' 'PE' 'CO' 'HN' 'FI' 'SE' 'AL' 'PT' 'MZ' 'CN'
```

## How many NA are in each column?

Unnamed: 0: 0

fecha dato: 0

ncodpers: 0

ind empleado: 10782

pais\_residencia: 10782

sexo: 10786

age: 10782

fecha alta: 10782

ind nuevo: 10782

antiguedad: 10782

indrel: 10782

ult\_fec\_cli\_1t: 998899

indrel\_1mes: 10782

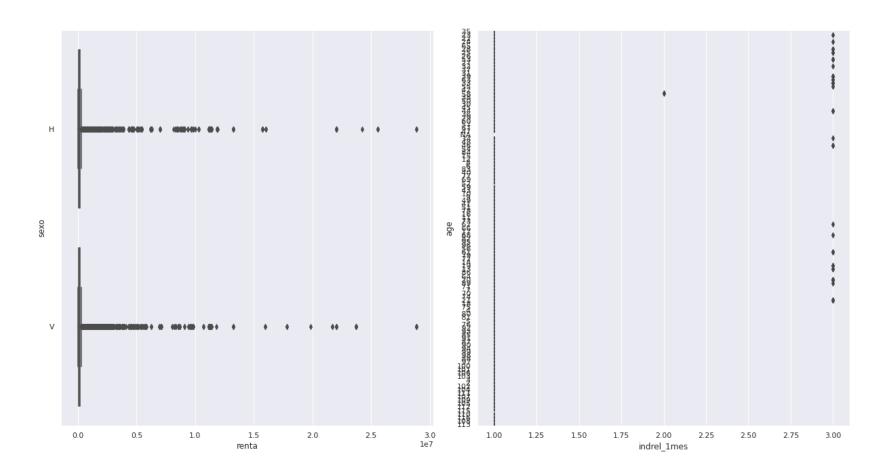
### Table 1

	Unn ame d: 0	fech a_da to	nco dper s	ind_e mplea do	pais_re sidenci a	sexo	age	fech a_alt a	ind_ nue vo	anti gue dad		ind_hip _fin_ult 1	ind_pla n_fin_ul t1	ind_pre s_fin_ul t1	ind_rec a_fin_ul t1	ind_tjcr_fin_ ult1	ind_ valo _fin _ult 1	ind_viv _fin_ult 1	ind_no mina_ul t1	ind_nom _pens_ul t1	ind_recibo_ult1
	10782 261	rows_× 2015 -01- 28	48 col 1050 741	umns NaN	NaN	NaN	NaN	NaN	NaN	NaN		0	0	0	0	0	0	0	0.0	0.0	0
1029	1029	2015 -01- 28	1051 017	NaN	NaN	NaN	NaN	NaN	NaN	NaN	:	0	0	0	0	0	0	0	0.0	0.0	0
1063	1063	2015 -01- 28	1051 064	NaN	NaN	NaN	NaN	NaN	NaN	NaN	i	0	0	0	0	0	0	0	0.0	0.0	0
1154	1154	2015 -01- 28	1051 387	NaN	NaN	NaN	NaN	NaN	NaN	NaN	:	0	0	0	0	0	0	0	0.0	0.0	0
1779	1779	2015 -01- 28	1048 660	NaN	NaN	NaN	NaN	NaN	NaN	NaN	:	0	0	0	0	0	0	0	0.0	0.0	0
	::	:		::					:												

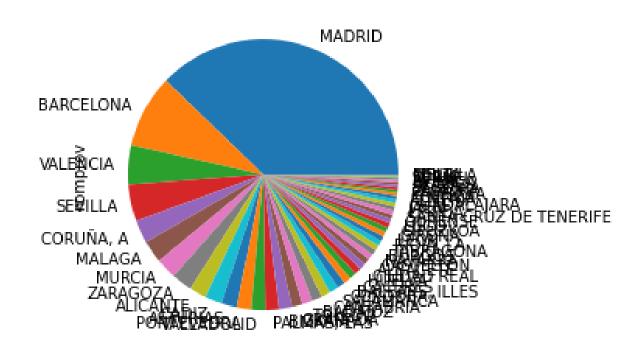
# **Correlation of variables**

	na me d: 0	nc od per s	ind _nu evo	ind rel	indr el_1 mes	tip od om	co d_ pro v	ind_acti vidad_cl iente	ren ta	ind_ah or_fin_ ult1	 ind_hi p_fin_ ult1	ind_pl an_fin _ult1	ind_pr es_fin _ult1	ind_re ca_fin _ult1	ind_tj cr_fin _ult1	ind_va lo_fin_ ult1	ind_vi v_fin_ ult1	ind_n omina _ult1	ind_no m_pens _ult1
Unname d: 0	1.0 000 00	- 0.4 471 19	0.0 083 12	- 0.0 073 74	- 0.00 0538	Na N	0.0 356 14	0.116265	0.0 445 28	0.0058 93	 0.0439 27	0.0486 69	0.0290 35	0.0926 28	0.0828 70	0.0776 29	0.0351 14	0.0692 04	0.07419
ncodper s		1.0 000 00	0.0 028 98	0.0 115 54	0.00 1253	Na N	- 0.0 407 61	- 0.187022	- 0.0 884 17	- 0.0134 69	 - 0.1000 45	- 0.1258 74	- 0.0270 66	- 0.1891 29	- 0.1902 08	- 0.1777 82	- 0.0735 52	- 0.1382 54	- 0.15262 2
ind_nue vo	0.0 083 12	0.0 028 98	1.0 000 00	0.0 266 81	0.26 8051	Na N	- 0.0 002 79	0.008165	- 0.0 009 86	- 0.0002 96	 - 0.0022 24	- 0.0027 01	- 0.0014 32	- 0.0030 34	- 0.0049 98	- 0.0042 65	- 0.0017 89	- 0.0040 81	- 0.00449 7
indrel	- 0.0 073 74	0.0 115 54	0.0 266 81	1.0 000 00	0.00 4462	Na N	0.0 017 28	- 0.030518	- 0.0 007 45	- 0.0004 47	 - 0.0027 45	- 0.0030 67	- 0.0012 19	- 0.0050 14	- 0.0056 41	- 0.0030 62	- 0.0026 99	- 0.0057 71	- 0.00591 9
indrel_1 mes		0.0 012 53	0.2 680	0.0 044 62	1.00 0000	Na N	- 0.0 007	- 0.000859	0.0 006	0.0000	 - 0.0006	0.0008	- 0.0004	- 0.0018	- 0.0011	- 0.0013	0.0005	0.0001	0.00031

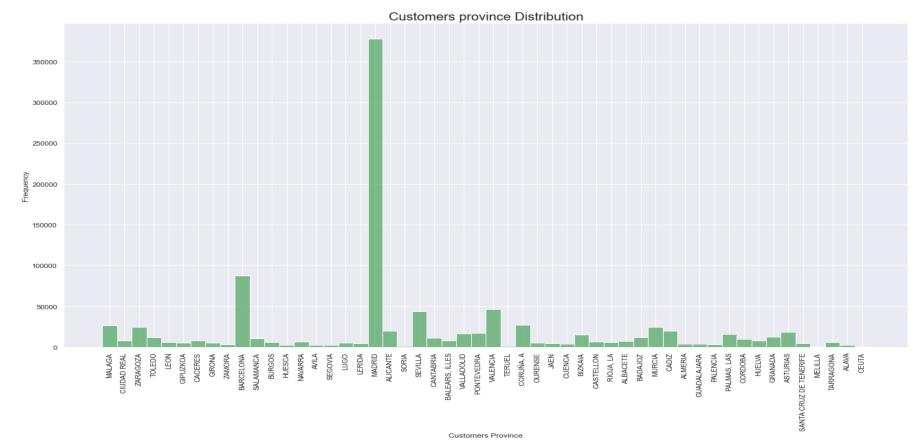
#### CHECKING FOR OUTLIERS IN (2)NUMERICAL DATA



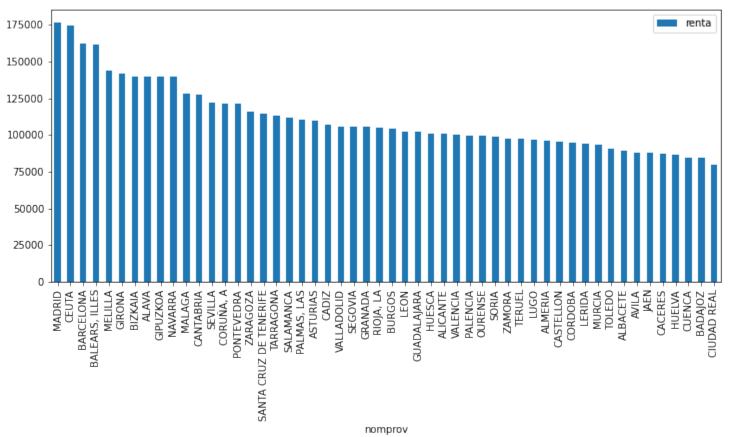
### Share of customers by province



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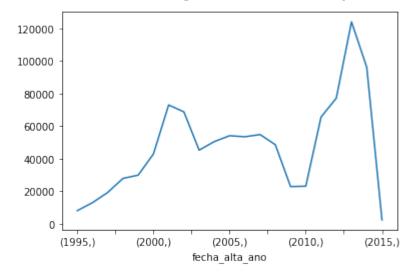


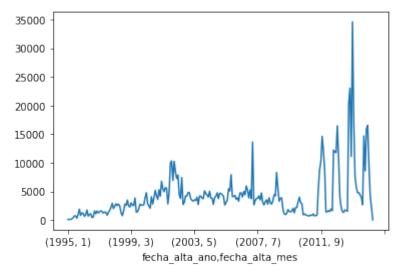
### Average household income by province



### Monthly and Annual Trends in Customer Registration

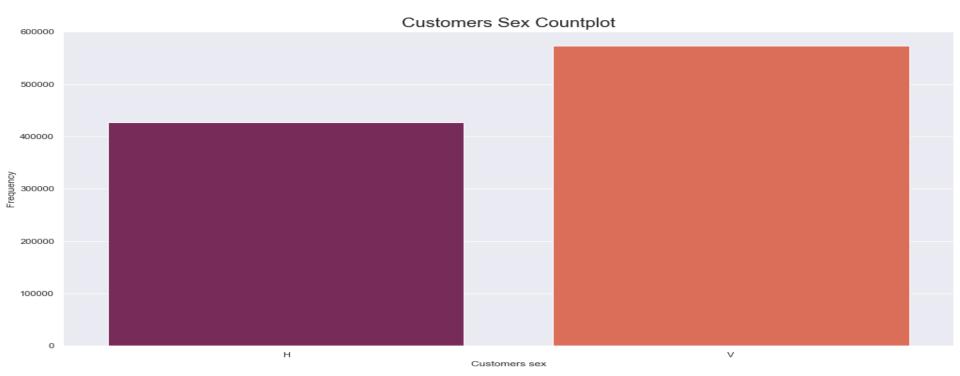
Number of customer registrations by year and by month. Note that the period after 2011 has more registrations compared to before.



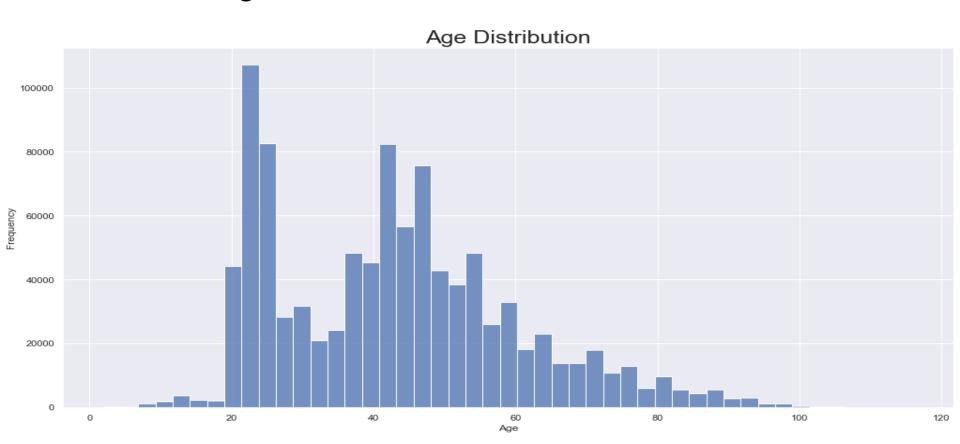


### **Customers Sex Category Distribution**

Most of the customers belong to the sex category "V".



# **Customers Age Distribution**



#### Recommendation

- We recommend to create a customer segment specially for the Madrid province residents and separately for residents outside Madrid, considering that they make up the largest group of customers and have a higher household income than other provinces.
- Another segment would be for customers who have registered from 2011, as the customers from those years represent a relatively large portion of customers compared to those from previous years.
- A sex category segment for customers will be appropriate as well. (Not sure if this is very appropriate - is this discrimination? I dont think its discrimination because we have a separate column that specifies the sex of all the customers, this is necessary because one wouldnt want christmas package designed for sex "v" to be going to Sex "h" remember it is a personalized christmas package, I think the company should have alot of varieties for future purposes as well.)
- Customer Demographic Segment: a segment for customers and their respective province.