

# Business Understanding

Virtual Internship

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# Agenda

**Data Understanding** 

Clients

Last contact

Campaign

Bivarial analysis

Recommended models



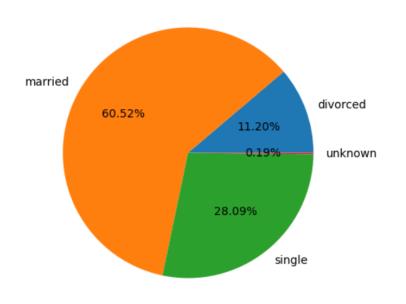
### Data Understanding

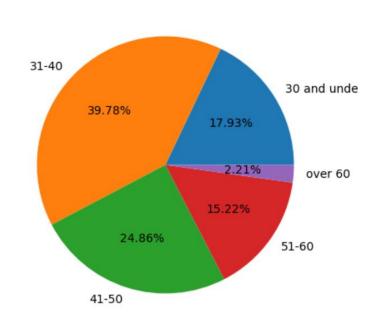
#### Features are divided into 4 groups:

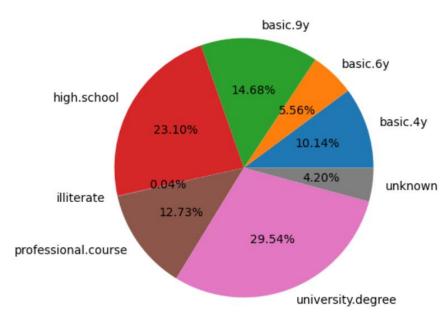
- 1. Data related to clients (age, job, marital, education, default, housing and loan).
- 2. Data related with the last contact of the current campaign (contact, month, day\_of\_week and duration).
- 3. Data related with campaigns (campaign, odays, previous and poutcome).
- 4. Data related with socio economic context attributes (emp.var.rate, cons.Price.idx, cons.conf.idx, euribor3m, nr.employed).
- 5. TARGET: "y".

```
Rows = 41188 | Columns = 21
['age', 'job', 'marital', 'education', 'default', 'housing', 'loan',
    'contact', 'month', 'day_of_week', 'duration', 'campaign', 'pdays',
    'previous', 'poutcome', 'emp.var.rate', 'cons.price.idx',
    'cons.conf.idx', 'euribor3m', 'nr.employed', 'y'],
```

#### Client







#### Customers job (%):

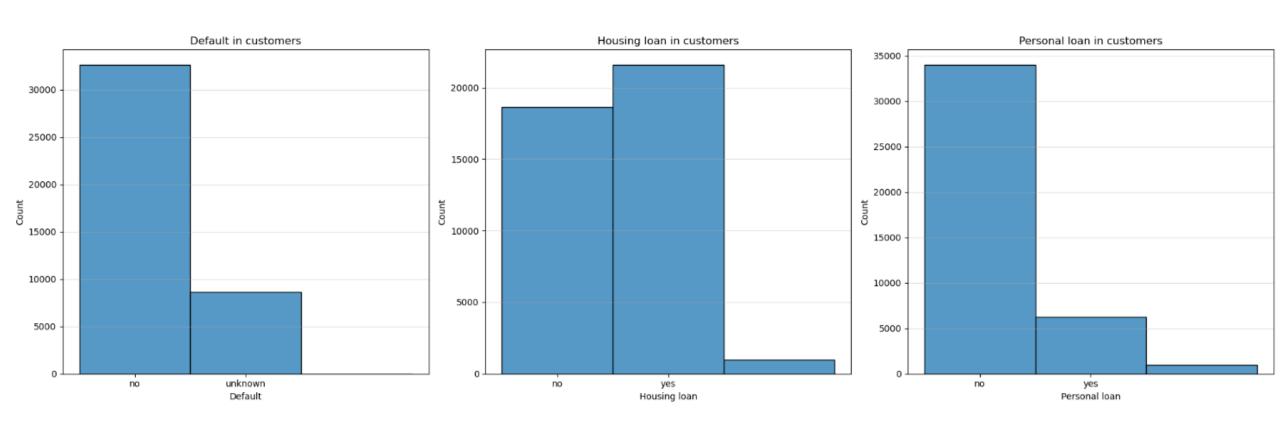
job			
admin.	25.303486	self-employed	3.450034
blue-collar	22.467709	housemaid	2.573565
technician	16.371273	unemployed	2.461882
services	9.636302	student	2.124405
management	7.099155	unknown	0.801204
retired	4.175974		
entrepreneur	3.535010		

#### Marital state of customers (%):

marital	
divorced	11.197436
married	60.522482
single	28.085850
unknown	0.194231

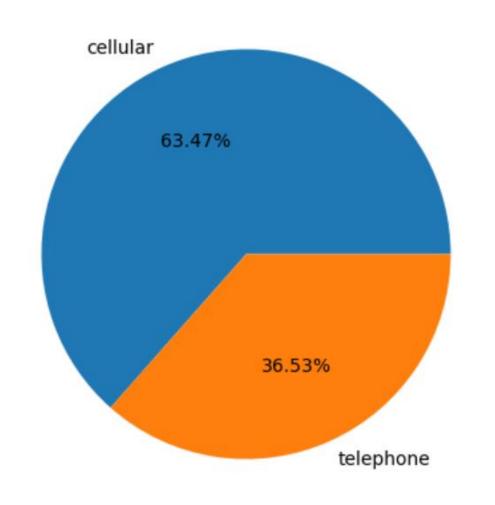
#### Client

#### Customers and Default, Housing Loan and Personal Loan:



#### **Last Contact**

#### **Contact to Costumers**



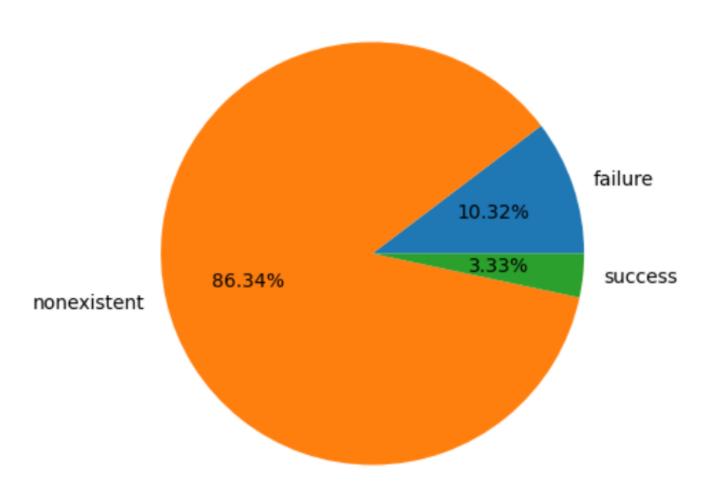
	contact				
month					
may	13769				
jul	7174				
aug	6178				
jun	5318				
nov	4101				

duration_ra	rige
0-200	22760
201-400	11264
401-600	3700
601-800	1707
801-1000	801
over 1000	956
contact	
cellular	63.474798
telephone	36.525202

duration range

### Campaing

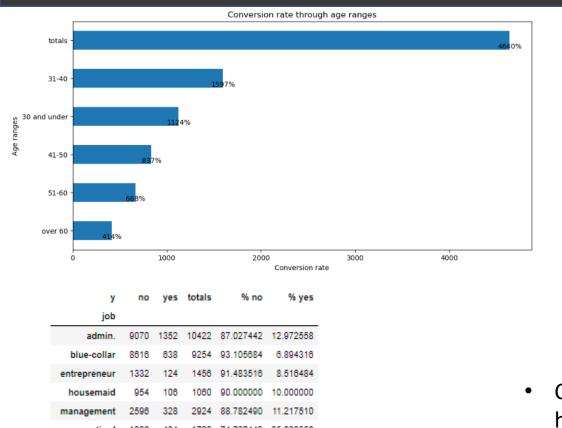




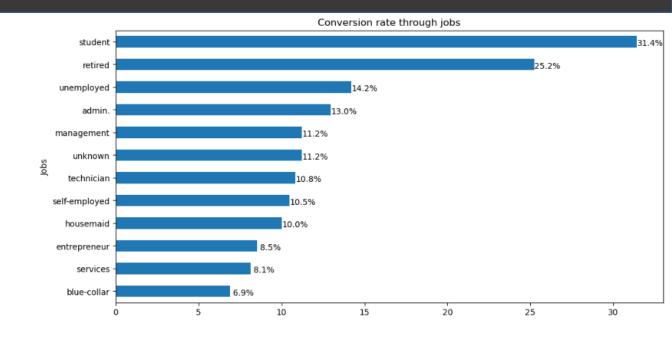
#### campaign\_range

0-3	81.463047
4-6	12.695445
7-9	3.185394
10-12	1.279499
More than 12	1.37661

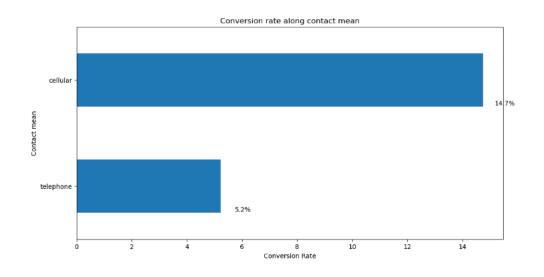
- The majority of customers (86%) have no previous relation with the bank
- Only 14% of customers have had previous contacts with the bank
- Among those who had previous contacts with the bank, the success rate is 3.3%.



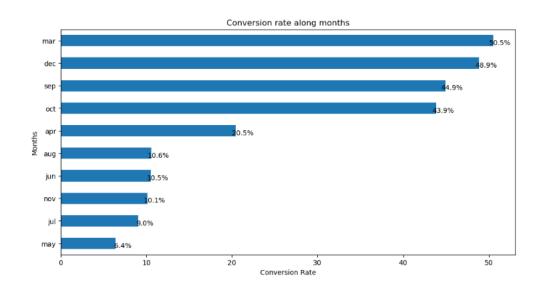
totals 36548 4640 41188 88.734583 11.265417



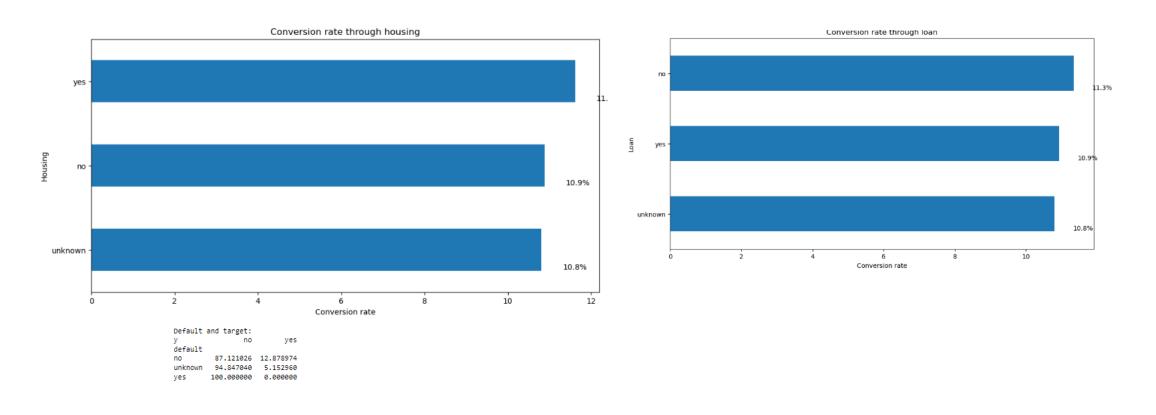
- Customers under 30 and over 60 years old have higher term deposit acceptance rates (15% and 45%, respectively)
- Student and retired are the top jobs with the highest acceptance rates.



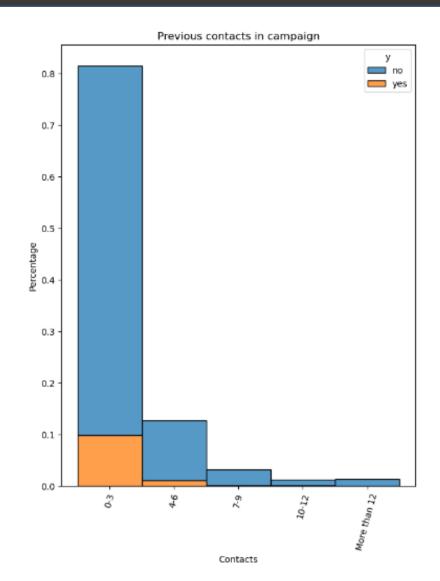
у	no	yes	totals	% no	% yes		
duration_range							
0-300	28466	1518	29984	94.937300	5.062700		
301-600			7740	81.421189	18.578811		
601-900			2183	57.672927	42.327073		
901-1200	329	439	768	42.838542	57.161458		
1201-1500	201-1500 115		307	37.459283	62.540717		
1501-1800	36	69	105	34.285714	65.714286		
1801-2100	2101-2400 6 8		56	39.285714	60.714286		
2101-2400			<b>2101-2400</b> 6		14	42.857143	57.142857
More than 2400			31	41.935484	58.064516		
totals	36548	4640	41188	88.734583	11.265417		

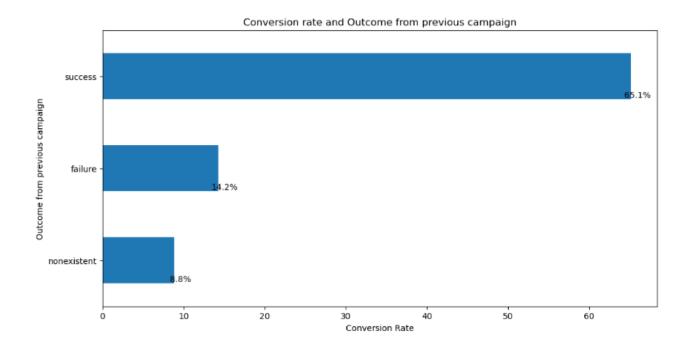


- Contacting customers by cell phone resulted in a higher term deposit acceptance rate (15%) compared to telephone contact (5%)
- Low customer numbers in December, March,
   October, and September had a high conversion rate (over 40%)
- Longer call duration was associated with higher conversion rates, but only a small proportion of

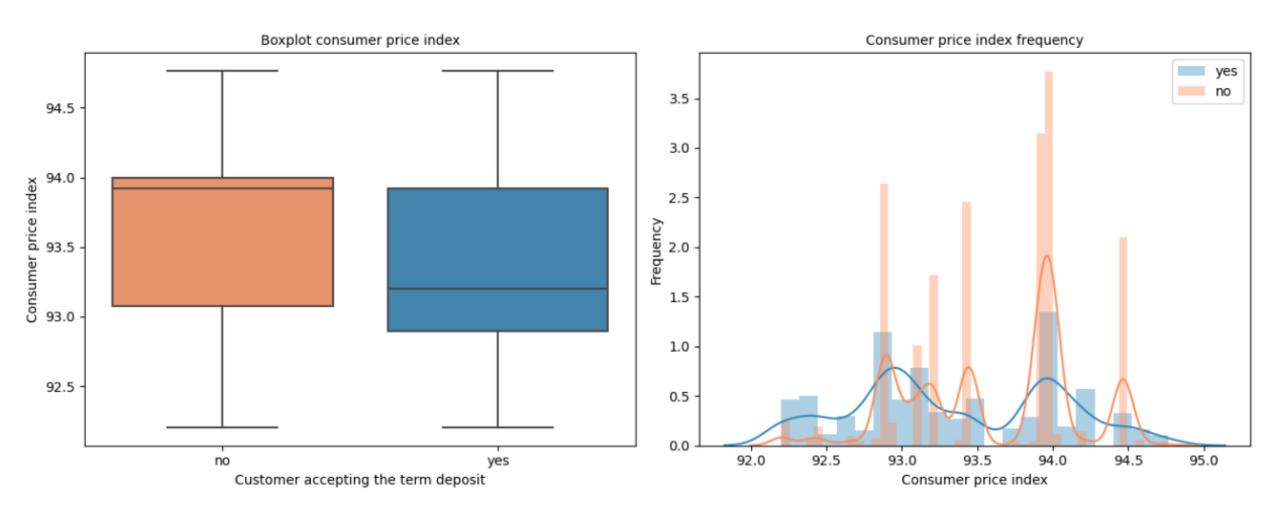


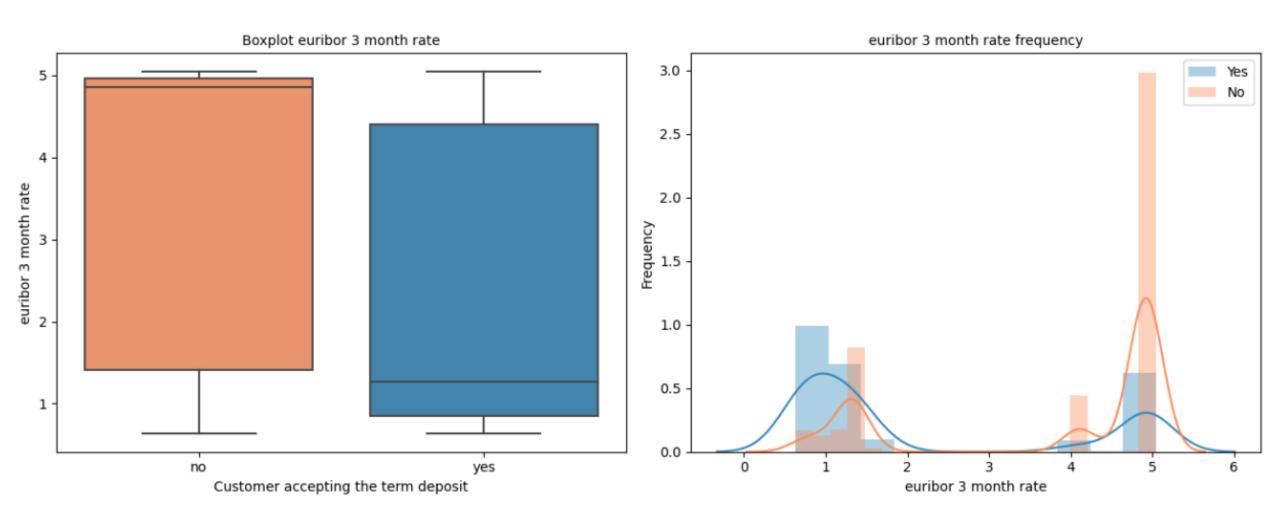
- Conversion rates for Loan and Housing products are similar across all categories of people (known, unknown, and without relevant information)
- Default product conversion rate varies based on default





- A lower number of contacts during the campaign is associated with higher conversion rates: 12% for 0-3 contacts and 9% for 4-6 contacts
- Previous campaign success is a strong predictor of term deposit acceptance, with 65% of customers who had previously accepted now also accepting it.





#### Recommended Model

Used Pycaret library to select best model for business problem based on AUC Preprocessed three datasets using different methods and found treating "Unknown" as separate category produced best results

	Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
lightgbm	Light Gradient Boosting Machine	0.8970	0.8007	0.3267	0.5767	0.4167	0.3649	0.3829	0.2900
catboost	CatBoost Classifier	0.8973	0.7962	0.3045	0.5855	0.4001	0.3501	0.3728	7.1710
lr	Logistic Regression	0.8231	0.7903	0.6352	0.3458	0.4476	0.3532	0.3764	0.3960
lda	Linear Discriminant Analysis	0.8210	0.7895	0.6352	0.3421	0.4446	0.3492	0.3730	0.1390

Selected LightGBM and CatBoost as top-performing models for hyperparameter tuning Plan to implement Grid Search and optimize hyperparameters next week to further

Plan to implement Grid Search and optimize hyperparameters next week to further improve model performance.

# Thank You

