```
# PROPERTY. SOR MANUEL MAN SORTE AND PROFET YOR MAGNET DATA SORTES.

# Input tagginhal

# Lagginhal.logit()

# PROPERTY. THE THIS CELL IN COURT TO POSET YOUR MAGNET DATA SORTES.

# PROPERTY.

# PROPER
```

Problem Description:

ABC Bank wants to sell it's term deposit product to customers and before launching the product they want to develop a model which help them in understanding whether a particular customer will buy their product or not (based on customer's past interaction with bank or other Financial Institution).

Business Understanding:

Bank wants to use ML model to shortlist customer whose chances of buying the product is more so that their marketing channel (tele marketing, SMS/email marketing etc) can focus only to those customers whose chances of buying the product is more. This will save resource and their time (which is directly involved in the cost (resource billing)). Develop model with Duration and without duration feature and report the performance of the model. Duration feature is not recommended as this will be difficult to explain the result to business and also it will be difficult for business to campaign based on duration.

Data Set Information:

The data is related with direct marketing campaigns of a Portuguese banking institution. The marketing campaigns were based on phone calls. Often, more than one contact to the same client was required, in order to access if the product (bank term deposit) would be ('yes') or not ('no') subscribed.

The classification goal is to predict if the client will subscribe (yes/no) a term deposit (variable y).

Attribute Information:

Input variables:

bank client data:

```
#1 - age (numeric
```

- #2 job : type of job (categorical: 'admin.','blue-collar','entrepreneur','housemaid','management','retired','self employed','services','student','technician','unemployed','unknown')
- #3 marital : marital status (categorical: 'divorced' married' single 'unknown': note: 'divorced' means divorced or widowed
- $\#\ 4-education\ (categorical:\ 'basic.4y',basic.6y',basic.9y',high.school',illiterate',professional.course',university.degree',unknown')$
- #5-default: has credit in default? (categorical: 'no','yes','unknown')
- # 6 housing: has housing loan? (categorical: 'no','yes','unknown
- #7 loan: has personal loan? (categorical: 'no','yes','unknown')

related with the last contact of the current campaign:

- #8 contact: contact communication type (categorical: 'cellular','telephone'
- #9 month: last contact month of year (categorical: 'jan', 'feb', 'mar', ..., 'nov', 'dec'
- # 10 day_of_week: last contact day of the week (categorical: 'mon','tue','wed','thu','fri'
- # 11 duration: last contact duration, in seconds (numeric). Important note: this attribute highly affects the output target (e.g., if duration=0 ther y='no'). Yet, the duration is not known before a call is performed. Also, after the end of the call y is obviously known. Thus, this input should only be included for benchmark purposes and should be discarded if the intention is to have a realistic predictive model.

other attributes:

- #12 campaign: number of contacts performed during this campaign and for this client (numeric, includes last contact
- #13 pdays: number of days that passed by after the client was last contacted from a previous campaign (numeric; 999 means client was not previously contacted
- #14 previous: number of contacts performed before this campaign and for this client (numeric
- ${\it \#\,15\,-}\,poutcome: outcome\ of\ the\ previous\ marketing\ campaign\ (categorical:\ 'failure', nonexistent', success')$

social and economic context attribute

- # 16 emp.var.rate: employment variation rate quarterly indicator (numeric
- #17 cons.price.idx: consumer price index monthly indicator (numeric
- # 18 cons.conf.idx: consumer confidence index monthly indicator (numeric
- #19 euribor3m: euribor 3 month rate daily indicator (numeric
- # 20 nr.employed: number of employees quarterly indicator (numeric)

Output variable (desired target)

#21 - v - has the client subscribed a term deposit? (binary: 'ves':no')

Index(['sl.no', 'age', 'job', 'marital', 'education', 'default', 'balance', 'housing', 'loan', 'contact', 'day', 'month', 'duration', 'campaign', 'pdays', 'previous', 'poutcome', 'y'], dtypem'object')

sl.	no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
0	1	58	management	married	tertiary	no	2143	yes	no	unknown	5	may	261	1	-1	0	unknown	no
1	2	44	technician	single	secondary	no	29	yes	по	unknown	5	may	151	1	-1	0	unknown	no
2	3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
3	4	47	blue-collar	married	unknown	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
4	5	33	unknown	single	unknown	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no

	sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
0	1	58	management	married	tertiary	no	2143	yes	no	unknown	5	may	261	1	-1	0	unknown	no
1	2	44	technician	single	secondary	no	29	yes	no	unknown	5	may	151	1	-1	0	unknown	no
2	3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
3	4	47	blue-collar	married	unknown	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
4	5	33	unknown	single	unknown	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no
45206	45207	51	technician	married	tertiary	no	825	no	no	cellular	17	nov	977	3	-1	0	unknown	yes
45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
45209	45210	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	no
45210	45211	37	entrepreneur	married	secondary	no	2971	no	no	cellular	17	nov	361	2	188	11	other	no
45211 rd	ows × 18 c	column	ıs														_	
4																		

```
1 Bawa_GELLinfe()

RangeIndex: 45211 entries, 0 to 45210
Data columns (total 18 columns):

# Column Non-Mull Count Otype

0 11 non 45211 non-null int64
1 age 45211 non-null int64
2 job 45211 non-null object
4 education 45211 non-null object
5 default 45211 non-null object
6 balance 45211 non-null object
7 housing 45211 non-null object
8 loan 45211 non-null object
10 day 45211 non-null object
11 month 45211 non-null object
12 duration 45211 non-null object
13 campaign 45211 non-null object
14 dpays 45211 non-null int64
15 previous 45211 non-null int64
16 poutcome 45211 non-null int64
17 y 45211 non-null object
17 y 45121 non-null object
dtypes: int64(8), object(18)
memory usage: 6.2+ MB
```

NALYSIS1.jpg

Dimage.png

image.png

Pimage.png

Double-click (or enter) to edit

Data Cleaning

Handle Duplicates

Remove duplicates

```
1 Bank_data = Bank_data.drop_duplicates()
2 Bank_data
                       job marital education default balance housing loan contact day month duration campaign pdays previous poutcome
                                                             29 yes no unknown 5 may
2 yes yes unknown 5 may
                                                                                                                1 -1 0 unknown no
1 -1 0 unknown no
              2 44 technician single secondary
              3 33 entrepreneur married secondary
              4 47 blue-collar married unknown
                                                                                                   198
              5 33
                                                                     no no unknown 5 may
                                                                                                                            0 unknown no
     45206 45207 51 technician married tertiary no 825 no no cellular 17 nov
                        retired divorced primary no 1729 no no cellular 17 nov
retired married secondary no 5715 no no cellular 17 nov
                                                                                                   456 2 -1 0 unknown yes
1127 5 184 3 success yes
508 4 -1 0 unknown no
     45207 45208 71
     45208 45209 72
                                                   no 668 no no telephone 17 nov
     45209 45210 57 blue-collar married secondary
                                                    no 2971 no no cellular 17 nov
     45210 45211 37 entrepreneur married secondary
                                                                                                              2 188
                                                                                                                           11 other no
```

Handle missing values

```
1 # Check for missing values in all datasets 2 print(News_data.ismall().set())

$\frac{1}{2}$ sl. no  
$\text{0}$ age  
$\text{0}$ job  

marital  
$\text{0}$ deducation  
$\text{default}$  
$\text{0}$ balance  

housing  
loan  
$\text{0}$ contact  
$\text{0}$ day  
$\text{0}$ month  
$\text{0}$ duration  
$\text{campaign}$ e 
pdays  
$\text{0}$ previous  
$\text{0}$ poutcome  
$\text{0}$ y  
$\text{0}$ dtype: int64
```

Exploratory Data Analysis (EDA)

```
1 = importing the external libraries
2 import pandas as pd
3
4 = Importing the data
5 Band_data-pd.read_cvv("/kaggle/imput/bank-full-versioni/bank-full.csv")
```

1 58 management married tertiary no 2143 yes no unknown 5 may 261 1 -1 0 unknown no no 1506 yes no unknown 5 may 4 47 blue-collar married unknown 45206 45207 51 technician married tertiary 0 unknown yes no no cellular 17 nov **45207** 45208 71 retired divorced primary no 1729 no no cellular 17 nov 456 2 -1 0 unknown yes 3 success yes retired married secondary no 5715 no no cellular 17 nov-plue-collar married secondary no 668 no no telephone 17 nov-1127 5 184 508 4 -1 **45208** 45209 72 45209 45210 57 blue-collar married secondary 0 unknown no 45210 45211 37 entrepreneur married secondary 2 188

```
1 mod_land_data_data_cort_values(by-balance', according-from)

1 mod_land_data_cort_values(by-balance', according-from)

1 mod_land_data_cort_values(by-balance', according-from)

2 mod_land_data_cort_values(by-balance', according-from)

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4 mod_land_data_cort_values(by-balance', according-from)

3 mod_land_data_cort_values(by-balance', according-from)

4 mod_land_data_cort_values(by-balance', according-from)

5 mod_land_data_cort_values(by-balance', according-from)

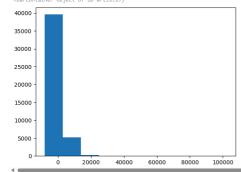
6 mod_land_
```

```
08/11/2024, 11:04
```

[40482 rows x 18 columns]

1 # Histogram
2 import matplotlib.pyplot as plt
3 plt.hist(new_bank_data['balance'])

(array([3.9587e+04, 5.2330e+03, 2.9400e+02, 5.8000e+01, 2.1000e+01, 7.0000e+00, 6.0000e+00, 1.0000e+00, 2.0000e+00]), array([-8019, 2995.6, 14010.2, 25024.8, 36039.4, 47054., 58068.6, 69083.2, 80097.8, 91112.4, 102127.]), cBarContainer object of 10 artists>)



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-		s1. ı	no a	gρ	iob	marital	education	default	halance	housing	loan	contact	dav	month	duration	campaign	ndavs	previous	noutcome	V
_				_									_							
	0		1	58	management	married	tertiary	no	2143	yes	no	unknown	5	may	261	1	-1	0	unknown	no
	1		2	44	technician	single	secondary	no	29	yes	no	unknown	5	may	151	1	-1	0	unknown	no
	2		3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
	3		4	47	blue-collar	married	unknown	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
	4		5	33	unknown	single	unknown	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no
45	5206	4520	07	51	technician	married	tertiary	no	825	no	no	cellular	17	nov	977	3	-1	0	unknown	yes
45	5207	4520	08	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
45	5208	4520	09	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
45	5209	452	10	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	no
45	5210	452	11 :	37	entrepreneur	married	secondary	no	2971	no	no	cellular	17	nov	361	2	188	11	other	no
453	211 rov	vs×1	8 colu	ımn	S														_	

>.▼		sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
	0	1	58	management	married	tertiary	по	2143	yes	по	unknown	5	may	261	1	-1	0	unknown	no
	1	2	44	technician	single	secondary	no	29	yes	по	unknown	5	may	151	1	-1	0	unknown	по
	2	3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
	3	4	47	blue-collar	married	unknown	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
	4	5	33	unknown	single	unknown	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no
	45206	45207	51	technician	married	tertiary	no	825	no	по	cellular	17	nov	977	3	-1	0	unknown	yes
	45207	45208	71	retired	divorced	primary	no	1729	no	по	cellular	17	nov	456	2	-1	0	unknown	yes
	45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
	45209	45210	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	no
	45210	45211	37	entrepreneur	married	secondary	no	2971	no	no	cellular	17	nov	361	2	188	11	other	no
	5211 rov	ws × 18	columi	ns															

4 . 1
...
45206 825
45207 1729
45208 5715
45209 668
45210 2971
Name: balance, Length: 45211, dtype: int64

1 ... 825 ... 1729 ... 5715 ... 668 ... 2971 balance, Length: 45211, dtype: int64

```
(array([3.9587e+04, 5.2330e+03, 2.9400e+02, 5.8000e+01, 2.1000e+01, 7.0000e+00, 6.0000e+00, 1.0000e+00, 2.0000e+00, 2.0000e+00]), array([-8019, 2995.6, 14010.2, 25024.8, 36039.4, 47054., 50666.6, 60083.2, 80097.8, 91112.4, 102127.]), cBarContainer object of 10 artists>)
          40000
          35000
          30000
          25000
          20000
          15000
          10000
            5000
                                           20000
                                                            40000
                                                                           60000
                                                                                            80000
                                                                                                           100000
sl. no age 0 job 0 marital 0 education 0 default 0 balance 0 housing 0 contact 0 day 0 day 0 day 0
         poutcome
         dtype: int64
 Index(['sl. no', 'age', 'job', 'marital', 'education', 'default', 'balance', 'housing', 'loan', 'contact', 'day', 'month', 'duration', 'campaign', 'pdays', 'previous', 'poutcome', 'y'], dtypee'object')
 array([[1, 58, 'management', ..., 0, 'unknown', 'no'],
[2, 44, 'technician', ..., 0, 'unknown', 'no'],
[3, 33, 'entrepreneur', ..., 0, 'unknown', 'no'],
                   ..., [45209, 72, 'retired', ..., 3, 'success', 'yes'], [45210, 57, 'blue-collar', ..., 0, 'unknown', 'no'], [45211, 37, 'entrepreneur', ..., 11, 'other', 'no']], dtype=object)
sl. no age job marital education default balance housing loan contact (day month duration cape.
        pdays
previous
poutcome
         dtvpe: object
         count 45211.000000 45211.000000 45211.000000 45211.000000 45211.000000 45211.000000 45211.000000

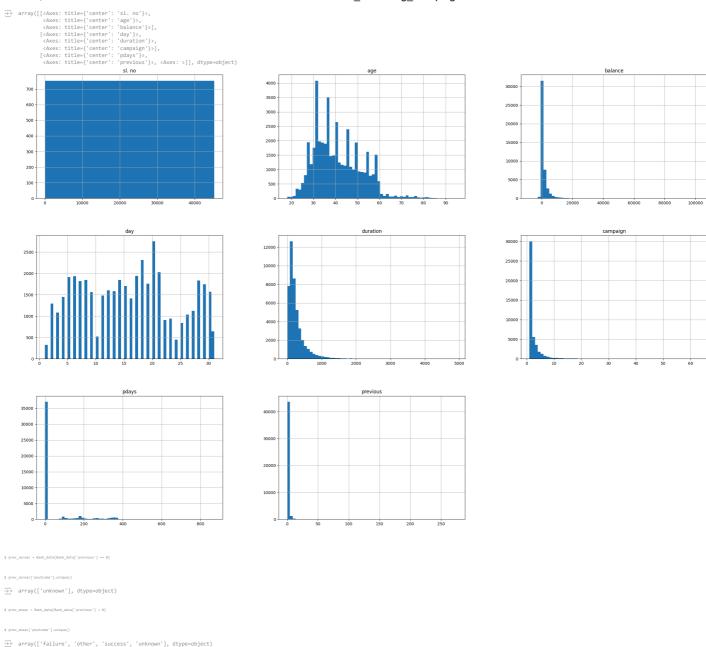
        mean
        22606.000000
        40.936210
        1362.272058
        15.806419
        258.163080
        2.763841
        40.197828
        0.580323

                                                                                     8.322476 257.527812
1.000000 0.000000
           std 13051.435847
                                           10.618762 3044.765829
                                                                                                                                3.098021 100.128746
                                                                                                                                                                          2.303441
          min 1.000000 18.000000 -8019.000000
                                                                                                                               1.000000
                                                                                                                                                    -1.000000
          25% 11303.500000 33.000000 72.000000
                                                                                     8.000000 103.000000 1.000000 -1.000000 0.000000
          50% 22606.000000
                                           39.000000 448.000000
                                                                                      16.000000 180.000000
                                                                                                                                2.000000
                                                                                                                                                     -1.000000
                                                                                                                                                                         0.000000
          75% 33908.500000
                                           48.000000 1428.000000
                                                                                      21.000000 319.000000
                                                                                                                                3.000000
                                                                                                                                                     -1.000000
                                                                                                                                                                          0.000000
```

1 Bank_data.nunique

```
sl. no 45211
age 77
job 12
marital 3
education 4
default 2
balance 7168
housing 2
loan 2
contact 3
day 31
month 12
duration 1573
campaign 48
pdays 559
previous 41
poutcome 4
y 2
dtype: int64
```

1 Bank_data.hist(bins=60, figsize=(30,20))



ərray(['failure', 'other', 'success', 'unknown'], dtype=object)

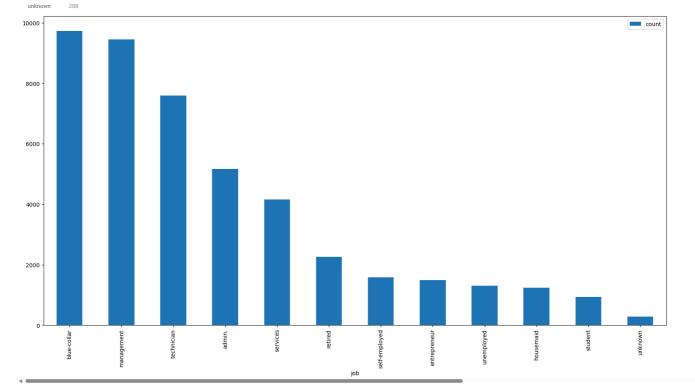
$\overline{\Rightarrow}$		sl. no	200	iob	marital	education	default	halance	housing	loan	contact	day	month	duration	camnaign	ndavs	nrevious	noutcome	V
		521 110	uge	300	mor zeuz	Cuucucion	ucruuze	Duzunec	110032116	20011	concacc	uuy	morren	001002011	сиприды	padys	previous	pouccome	
	12660	12661	31	technician	single	secondary	no	19	no	no	telephone	4	jul	65	2	-1	0	unknown	no
	12665	12666	25	blue-collar	single	secondary	no	192	no	no	telephone	4	jul	112	2	-1	0	unknown	no
	12674	12675	36	unemployed	divorced	primary	no	0	no	no	telephone	4	jul	304	3	-1	0	unknown	no
	12716	12717	49	technician	married	primary	no	486	no	no	telephone	7	jul	229	5	-1	0	unknown	no
	12750	12751	46	unknown	married	unknown	no	2911	no	no	telephone	7	jul	41	9	-1	0	unknown	no
	45141	45142	77	unknown	married	unknown	no	397	no	no	telephone	8	nov	207	1	185	3	success	no
	45154	45155	63	retired	married	primary	no	3738	no	no	telephone	9	nov	301	1	456	4	failure	no
	45166	45167	41	unemployed	single	tertiary	no	79	no	no	telephone	9	nov	394	1	390	2	success	yes
	45170	45171	19	student	single	primary	no	245	no	no	telephone	10	nov	98	2	110	2	other	no
	45209	45210	57	blue-collar	married	secondary	no	668	no	по	telephone	17	nov	508	4	-1	0	unknown	по
	1509 rov	vs × 18 cc	lumns																

 \rightarrow array(['no', 'yes'], dtype=object)

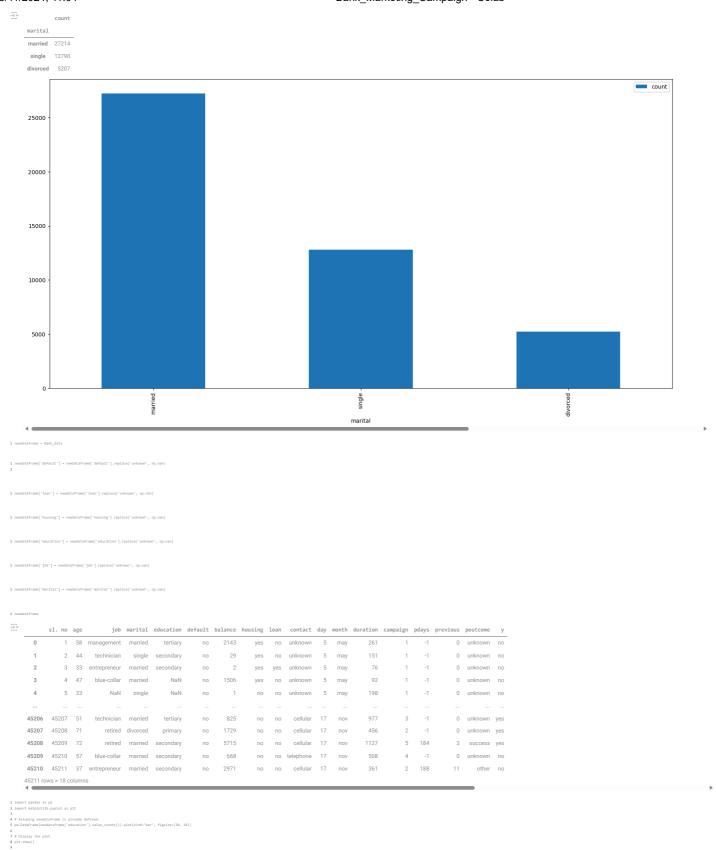
	sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome)
0	1	58	management	married	tertiary	no	2143	yes	по	unknown	5	may	261	1	-1	0	unknown	n
1	2	44	technician	single	secondary	no	29	yes	no	unknown	5	may	151	1	-1	0	unknown	n
2	3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	n
3	4	47	blue-collar	married	unknown	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	n
4	5	33	unknown	single	unknown	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	n
45206	45207	51	technician	married	tertiary	no	825	no	no	cellular	17	nov	977	3	-1	0	unknown	ye
45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	ye
45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	ye
45209	45210	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	n
45210	45211	37	entrepreneur	married	secondary	no	2971	no	no	cellular	17	nov	361	2	188	11	other	п
45211 rows × 18 columns																		

L pd.DataFrame(Bank_data['job'].value_counts()).plot(kind='bar', figsize=(20,10)

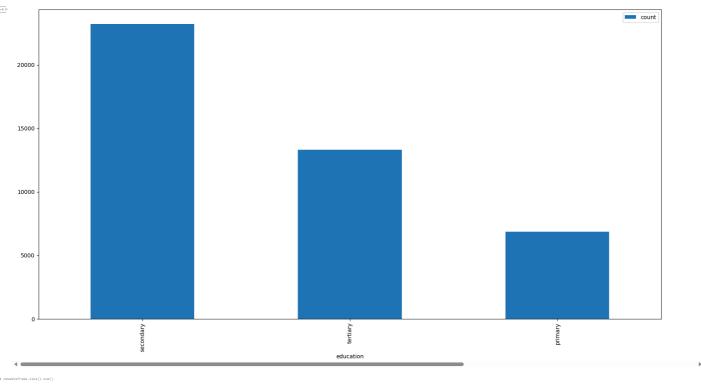
₹		count
	job	
	blue-collar	9732
	management	9458
	technician	7597
	admin.	5171
	services	4154
	retired	2264
	self-employed	1579
	entrepreneur	1487
	unemployed	1303
	housemaid	1240
	student	938



¹ pd.DataFrame(Bank_data['marital'].value_counts()).plot(kind='bar', figsize=(20,10))
2 pd.DataFrame(Bank_data['marital'].value_counts())



https://colab.research.google.com/#fileId=https%3A//storage.googleap is.com/kaggle-colab-exported-notebooks/bank-marketing-campaign-d5c2...



1 meditarrame.isma().sum()

3 age 0
job 288
marital 0
education 1857
default 0
balance 0
housing 0
contact 0
day 0
month 0
duration 0
campaign 0
pdays 0
previous 0
poutcome 0
y these inted y dtype: int64

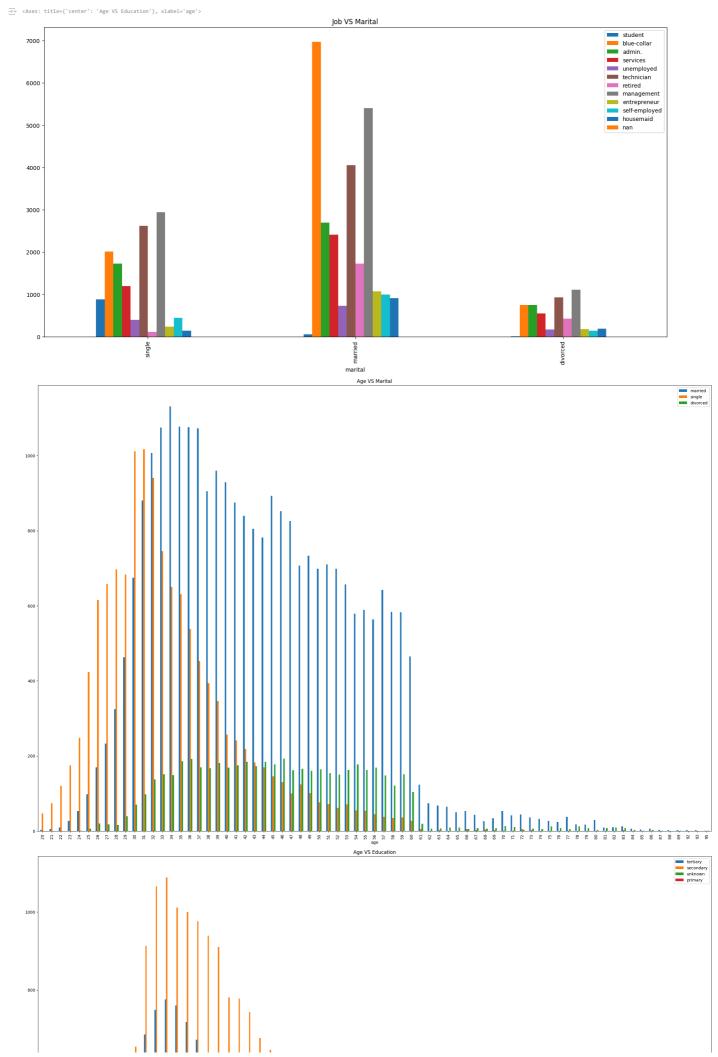
		sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
-	42954	42955	18	student	single	NaN	no	108	no	no	cellular	9	feb	92	1	183	1	success	yes
	41273	41274	18	student	single	NaN	no	3	no	no	cellular	25	aug	130	2	-1	0	unknown	yes
	42146	42147	18	student	single	secondary	no	156	no	no	cellular	4	nov	298	2	82	4	other	no
	43637	43638	18	student	single	NaN	no	348	no	no	cellular	5	may	443	4	-1	0	unknown	yes
	40736	40737	18	student	single	primary	no	1944	no	no	telephone	10	aug	122	3	-1	0	unknown	no
	44420	44421	93	retired	married	NaN	no	775	no	no	cellular	4	aug	476	2	13	9	success	yes
	44262	44263	93	retired	married	NaN	no	775	no	no	cellular	22	jul	860	2	177	7	success	yes
	31233	31234	94	retired	divorced	secondary	no	1234	no	no	cellular	3	mar	212	1	-1	0	unknown	по
	41663	41664	95	retired	married	secondary	no	0	no	no	telephone	1	oct	215	1	-1	0	unknown	no
	33699	33700	95	retired	divorced	primary	no	2282	no	no	telephone	21	apr	207	17	-1	0	unknown	yes
4	5211 ro	ws × 18 c	olumr	IS															

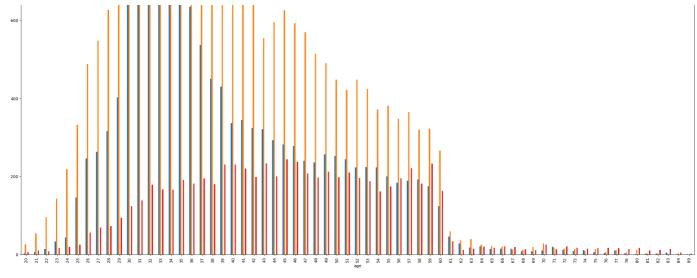
```
array(['management', 'technician', 'entrepreneur', 'blue-collar', nan, 'retired', 'admin.', 'services', 'self-employed', 'unemployed', 'housemaid', 'student'], dtype=object)
```

$\overline{\geq}$		student	blue-collar	admin.	services	unemployed	technician	retired	management	entrepreneur	self-employed	housemaid	nan
	marital												
	single	878	2014	1728	1198	401	2620	108	2947	238	446	144	NaN
	married	54	6968	2693	2407	731	4052	1731	5400	1070	993	912	NaN
	divorced	6	750	750	549	171	925	425	1111	179	140	184	NaN

⇒ array(['married', 'single', 'divorced'], dtype=object)

```
1131 650.0
1077 631.0
1076 538.0
1075 746.0
1073 453.0
   [73 rows x 3 columns]
       married single divorced
    20
    22
          9 120.0
                      NaN
    23
          27 175.0
         53 248.0
          2 NaN
                      1.0
           2 NaN
                      NaN
                       1.0
= array(['tertiary', 'secondary', nan, 'primary'], dtype=object)
                                179
167
139
166
124
...
12
10
17
2
   [65 rows x 4 columns]
       tertiary secondary unknown primary
                  26.0
    20
    21
                  54.0
    22
                  96.0
         33 143.0 NaN
    23
    24
                  219.0
                                 19
    81
                  3.0 NaN 10
                   3.0
                                 12
    83
                 2.0 NaN
                                14
                  3.0 NaN
```





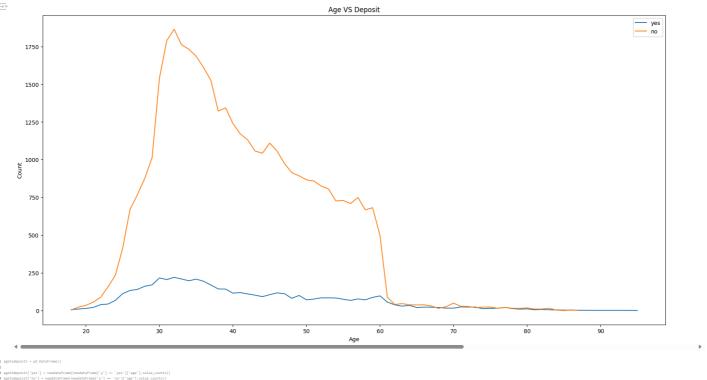
1 newdat	neudata Franc																			
\equiv		sl.	no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
	0		1	58	management	married	tertiary	no	2143	yes	no	unknown	5	may	261	1	-1	0	unknown	no
	1		2	44	technician	single	secondary	no	29	yes	no	unknown	5	may	151	1	-1	0	unknown	no
	2		3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
	3		4	47	blue-collar	married	NaN	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
	4		5	33	NaN	single	NaN	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no
	45206	45	207	51	technician	married	tertiary	no	825	no	no	cellular	17	nov	977	3	-1	0	unknown	yes
	45207	45	208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
	45208	45	209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
	45209	45	210	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	no
	45210	45	211	37	entrepreneur	married	secondary	no	2971	no	no	cellular	17	nov	361	2	188	11	other	no
4	15211 rd)WS ×	18 c	olumn	IS														_	

```
1 import pendas as pd

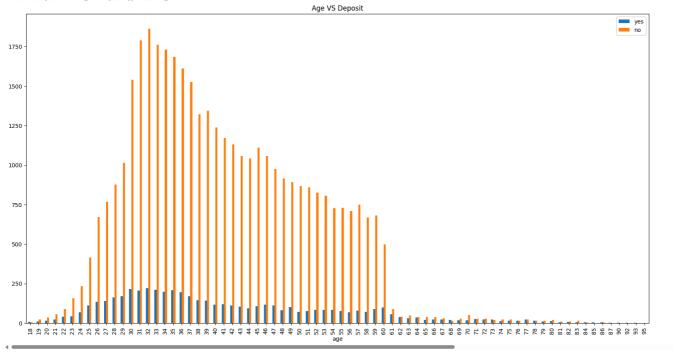
2
1
2
2
3
2
3
3
4 pathologist! - pd.fast/rame()
5 agethologist!("m") - modataframe(posudsaframe("y") -- "yes")["age"].value_counts()
7
8 priot(agethologist!("m") - modataframe(modataframe("y") -- "no")["age"].value_counts()
7
8 priot(agethologist!("m") - modataframe(modataframe("y") -- "no")["age"].value_counts()
7
8 priot(agethologist)
30 2221 1864.0
30 227 1540.0
33 220 1655.0
31 206 1790.0
.......
87 3 1.0
90 2 NaN
92 2 NaN
95 1 1.0

[74 rOWS X 2 columns]

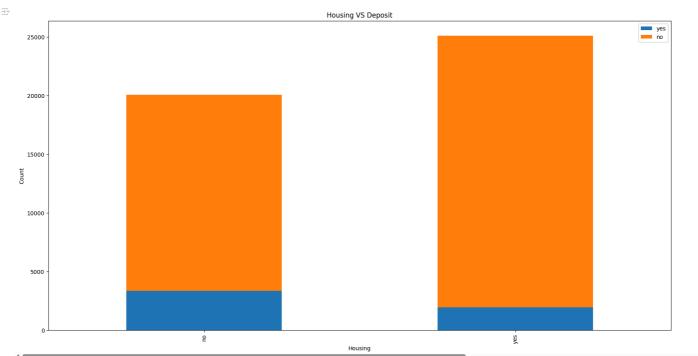
1 import pendas as pd
2 import emploitlib.typylat as plt
4
4
4
4
4
4
4
4
4
4
4
4
4
4
4
5 agethologist!("m") - modataframe()
6 agethologist!("m") - modataframe("y") -- "yes")["age"].value_counts()
7 agethologist!("m") - modataframe(modataframe("y") -- "no")["age"].value_counts()
9 af Platting the line plat
1 agethologist!("m") - modataframe(modataframe("y") -- "no")["age"].value_counts()
9 af Platting the line plat
1 agethologist!.com("interplate") - modataframe("y") -- "no")["age"].value_counts()
9 af Platting the line plat
1 agethologist!.com("interplate") -- modataframe("y") -- "no")["age"].value_counts()
9 af Platting the line plat
1 agethologist!.com("interplate") -- modataframe("y") -- "no")["age"].value_counts()
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1 agethologist!.com("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("interplate("in
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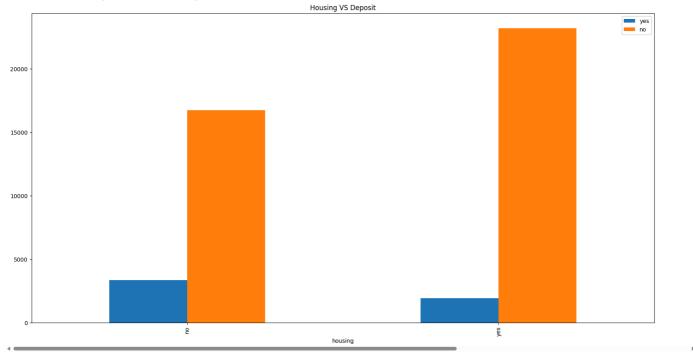




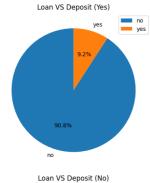


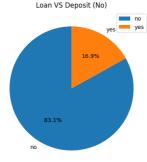
- 1 housingtodeposit = pd.DataFrame(
- 3 housingtodeposit['yes'] = newdataframe[newdataframe['y'] == 'yes']['housing'].value_counts(4 housingtodeposit['no'] = newdataframe[newdataframe['y'] == 'no']['housing'].value_counts()
- 6 housingtodeposit.sort_index().plot.bar(title = "Housing VS Deposit", figsize=(20,10))

 $\stackrel{\textstyle \longleftarrow}{\longrightarrow}$ <Axes: title={'center': 'Housing VS Deposit'}, xlabel='housing'>







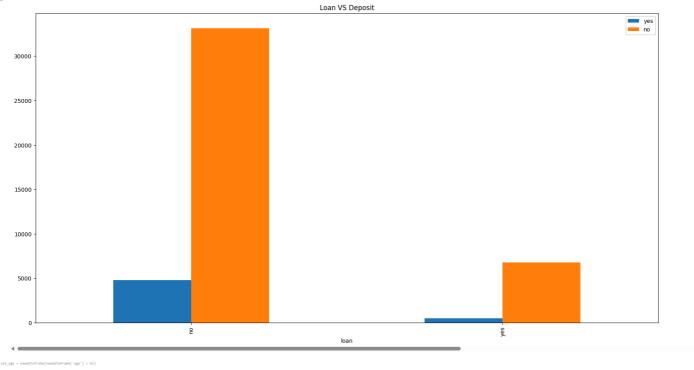


1 loantodeposit = pd.DataFrame

 \Rightarrow

- 3 loantodeposit['yes'] = newdataframe[newdataframe['y'] == 'yes']['loan'].value_counts(
 4 loantodeposit['no'] = newdataframe[newdataframe['y'] == 'no']['loan'].value_counts()
- 6 loamtodeposit.sort_index().plot.bar(title = "Loan VS Deposit", figsize=(20,10))

 $\overrightarrow{\exists_{v}}$ <Axes: title={'center': 'Loan VS Deposit'}, xlabel='loan'>



\exists		sl.	no	age	jol
28	906	28	907	66	housemai
28	924	28	925	62	technicia

₹		sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
	28906	28907	66	housemaid	married	secondary	no	1929	no	no	cellular	2	feb	169	1	-1	0	unknown	no
	28924	28925	62	technician	married	secondary	no	973	no	no	cellular	2	feb	326	1	-1	0	unknown	yes
	28965	28966	62	retired	married	secondary	no	1495	no	no	cellular	2	feb	265	1	-1	0	unknown	yes
	29158	29159	83	retired	married	primary	no	425	no	no	telephone	2	feb	912	1	-1	0	unknown	no
	29261	29262	75	retired	divorced	primary	no	46	no	no	cellular	2	feb	294	1	-1	0	unknown	no
	45191	45192	75	retired	divorced	tertiary	no	3810	yes	no	cellular	16	nov	262	1	183	1	failure	yes
	45195	45196	68	retired	married	secondary	no	1146	no	no	cellular	16	nov	212	1	187	6	success	yes
	45204	45205	73	retired	married	secondary	no	2850	no	no	cellular	17	nov	300	1	40	8	failure	yes
	45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
	45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
	1041 row	vs × 18 co	lumns																

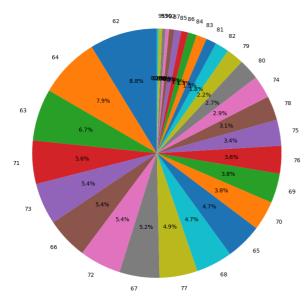
1 010_0	iReford" aRel	y j yes	J																
₹		sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	у
	28924	28925	62	technician	married	secondary	no	973	no	no	cellular	2	feb	326	1	-1	0	unknown	yes
	28965	28966	62	retired	married	secondary	no	1495	no	no	cellular	2	feb	265	1	-1	0	unknown	yes
	29464	29465	68	retired	divorced	primary	no	250	no	no	cellular	3	feb	360	1	-1	0	unknown	yes
	29865	29866	75	retired	divorced	primary	no	3881	yes	no	cellular	4	feb	136	3	-1	0	unknown	yes
	29961	29962	69	retired	married	primary	no	324	no	no	cellular	4	feb	136	4	-1	0	unknown	yes
	45191	45192	75	retired	divorced	tertiary	no	3810	yes	no	cellular	16	nov	262	1	183	1	failure	yes
	45195	45196	68	retired	married	secondary	no	1146	no	no	cellular	16	nov	212	1	187	6	success	yes
	45204	45205	73	retired	married	secondary	no	2850	no	no	cellular	17	nov	300	1	40	8	failure	yes
	45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
	45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
	445 rows	s × 18 colu	ımns																

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₹		sl. no	age	job	marital	education	default	balance	housing	loan	contact	day	month	duration	campaign	pdays	previous	poutcome	У
	28906	28907	66	housemaid	married	secondary	no	1929	no	no	cellular	2	feb	169	1	-1	0	unknown	no
	28924	28925	62	technician	married	secondary	no	973	no	no	cellular	2	feb	326	1	-1	0	unknown	yes
	28965	28966	62	retired	married	secondary	no	1495	no	no	cellular	2	feb	265	1	-1	0	unknown	yes
	29158	29159	83	retired	married	primary	no	425	no	no	telephone	2	feb	912	1	-1	0	unknown	no
	29261	29262	75	retired	divorced	primary	no	46	no	no	cellular	2	feb	294	1	-1	0	unknown	no
	45191	45192	75	retired	divorced	tertiary	no	3810	yes	no	cellular	16	nov	262	1	183	1	failure	yes
	45195	45196	68	retired	married	secondary	no	1146	no	no	cellular	16	nov	212	1	187	6	success	yes
	45204	45205	73	retired	married	secondary	no	2850	no	no	cellular	17	nov	300	1	40	8	failure	yes
	45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
	45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
	1041 row	/s × 18 co	lumns																

1 import matplotlib.pyplot as plt

→

Age Distribution for Yes Deposits



1 newdataframe

	sl. no	age	300	maritai	education	аетаитт	ратапсе	nousing	Toan	contact	aay	montn	duration	campaign	paays	previous	pourcome	У
0	1	58	management	married	tertiary	no	2143	yes	no	unknown	5	may	261	1	-1	0	unknown	no
1	2	44	technician	single	secondary	no	29	yes	no	unknown	5	may	151	1	-1	0	unknown	no
2	3	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	5	may	76	1	-1	0	unknown	no
3	4	47	blue-collar	married	NaN	no	1506	yes	no	unknown	5	may	92	1	-1	0	unknown	no
4	5	33	NaN	single	NaN	no	1	no	no	unknown	5	may	198	1	-1	0	unknown	no
45206	45207	51	technician	married	tertiary	no	825	no	no	cellular	17	nov	977	3	-1	0	unknown	yes
45207	45208	71	retired	divorced	primary	no	1729	no	no	cellular	17	nov	456	2	-1	0	unknown	yes
45208	45209	72	retired	married	secondary	no	5715	no	no	cellular	17	nov	1127	5	184	3	success	yes
45209	45210	57	blue-collar	married	secondary	no	668	no	no	telephone	17	nov	508	4	-1	0	unknown	no
45210	45211	37	entrepreneur	married	secondary	no	2971	no	по	cellular	17	nov	361	2	188	11	other	no
5211 ro	ws × 18 c	olumr	IS															

1 newdataframe.rename(columns = {'y':'deposited?'), inplace = True)

1 newdataframe['default'] = newdataframe['default'].replace({'yes': 1, 'no': θ})

1 newdataframe['deposited?'] = newdataframe['deposited?'].replace(('yes': 1, 'no': 0))

1 newdataframe['housing'] = newdataframe['housing'].replace(('yes': 1, 'no': 0)

1 newdataframe('loan') = newdataframe('loan').reolace(('ves': 1. 'no': 0))