

```
In [76]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

import sklearn
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsRegressor
from sklearn.linear_model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score

from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
import warnings
warnings.filterwarnings('ignore')
```

```
In [10]: df=pd.read_csv('loan.csv')
df
```

```
Out[10]:
```

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
0	56	1	999	0	1	0	0
1	57	1	999	0	1	0	0
2	37	1	999	0	1	0	0
3	40	1	999	0	1	0	1
4	56	1	999	0	1	0	0
...
41183	73	1	999	0	1	1	0
41184	46	1	999	0	1	0	0
41185	56	2	999	0	1	1	0
41186	44	1	999	0	1	0	0
41187	74	3	999	1	1	1	0

41188 rows × 60 columns



```
In [6]: df.describe()
```

Out[6]:

	age	campaign	pdays	previous	no_previous_contact	no
count	41188.00000	41188.000000	41188.000000	41188.000000	41188.000000	411
mean	40.02406	2.567593	962.475454	0.172963	0.963217	
std	10.42125	2.770014	186.910907	0.494901	0.188230	
min	17.00000	1.000000	0.000000	0.000000	0.000000	
25%	32.00000	1.000000	999.000000	0.000000	1.000000	
50%	38.00000	2.000000	999.000000	0.000000	1.000000	
75%	47.00000	3.000000	999.000000	0.000000	1.000000	
max	98.00000	56.000000	999.000000	7.000000	1.000000	

8 rows × 60 columns



In [9]: `df.shape`

Out[9]: (41188, 60)

In [10]: `df.columns`

Out[10]: Index(['age', 'campaign', 'pdays', 'previous', 'no_previous_contact', 'not_working', 'job_admin.', 'job_blue-collar', 'job_entrepreneur', 'job_housemaid', 'job_management', 'job_retired', 'job_self-employed', 'job_services', 'job_student', 'job_technician', 'job_unemployed', 'job_unknown', 'marital_divorced', 'marital_married', 'marital_single', 'marital_unknown', 'education_basic.4y', 'education_basic.6y', 'education_basic.9y', 'education_high.school', 'education_illiterate', 'education_professional.course', 'education_university.degree', 'education_unknown', 'default_no', 'default_unknown', 'default_yes', 'housing_no', 'housing_unknown', 'housing_yes', 'loan_no', 'loan_unknown', 'loan_yes', 'contact_cellular', 'contact_telephone', 'month_apr', 'month_aug', 'month_dec', 'month_jul', 'month_jun', 'month_mar', 'month_may', 'month_nov', 'month_oct', 'month_sep', 'day_of_week_fri', 'day_of_week_mon', 'day_of_week_thu', 'day_of_week_tue', 'day_of_week_wed', 'poutcome_failure', 'poutcome_nonexistent', 'poutcome_success', 'Loan_Status_label'], dtype='object')

In [11]: `df.info()`

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 41188 entries, 0 to 41187

Data columns (total 60 columns):

#	Column	Non-Null Count	Dtype
0	age	41188 non-null	int64
1	campaign	41188 non-null	int64
2	pdays	41188 non-null	int64
3	previous	41188 non-null	int64
4	no_previous_contact	41188 non-null	int64
5	not_working	41188 non-null	int64
6	job_admin.	41188 non-null	int64
7	job_blue-collar	41188 non-null	int64
8	job_entrepreneur	41188 non-null	int64
9	job_housemaid	41188 non-null	int64
10	job_management	41188 non-null	int64
11	job_retired	41188 non-null	int64
12	job_self-employed	41188 non-null	int64
13	job_services	41188 non-null	int64
14	job_student	41188 non-null	int64
15	job_technician	41188 non-null	int64
16	job_unemployed	41188 non-null	int64
17	job_unknown	41188 non-null	int64
18	marital_divorced	41188 non-null	int64
19	marital_married	41188 non-null	int64
20	marital_single	41188 non-null	int64
21	marital_unknown	41188 non-null	int64
22	education_basic.4y	41188 non-null	int64
23	education_basic.6y	41188 non-null	int64
24	education_basic.9y	41188 non-null	int64
25	education_high.school	41188 non-null	int64
26	education_illiterate	41188 non-null	int64
27	education_professional.course	41188 non-null	int64
28	education_university.degree	41188 non-null	int64
29	education_unknown	41188 non-null	int64
30	default_no	41188 non-null	int64
31	default_unknown	41188 non-null	int64
32	default_yes	41188 non-null	int64
33	housing_no	41188 non-null	int64
34	housing_unknown	41188 non-null	int64
35	housing_yes	41188 non-null	int64
36	loan_no	41188 non-null	int64
37	loan_unknown	41188 non-null	int64
38	loan_yes	41188 non-null	int64
39	contact_cellular	41188 non-null	int64
40	contact_telephone	41188 non-null	int64
41	month_apr	41188 non-null	int64
42	month_aug	41188 non-null	int64
43	month_dec	41188 non-null	int64
44	month_jul	41188 non-null	int64
45	month_jun	41188 non-null	int64
46	month_mar	41188 non-null	int64
47	month_may	41188 non-null	int64
48	month_nov	41188 non-null	int64
49	month_oct	41188 non-null	int64
50	month_sep	41188 non-null	int64
51	day_of_week_fri	41188 non-null	int64
52	day_of_week_mon	41188 non-null	int64
53	day_of_week_thu	41188 non-null	int64
54	day_of_week_tue	41188 non-null	int64

```

55  day_of_week_wed          41188 non-null  int64
56  poutcome_failure        41188 non-null  int64
57  poutcome_nonexistent    41188 non-null  int64
58  poutcome_success        41188 non-null  int64
59  Loan_Status_label        41188 non-null  int64
dtypes: int64(60)
memory usage: 18.9 MB

```

```
In [12]: df.describe()
```

```
Out[12]:
```

	age	campaign	pdays	previous	no_previous_contact	no
count	41188.00000	41188.000000	41188.000000	41188.000000	41188.000000	411
mean	40.02406	2.567593	962.475454	0.172963	0.963217	
std	10.42125	2.770014	186.910907	0.494901	0.188230	
min	17.00000	1.000000	0.000000	0.000000	0.000000	
25%	32.00000	1.000000	999.000000	0.000000	1.000000	
50%	38.00000	2.000000	999.000000	0.000000	1.000000	
75%	47.00000	3.000000	999.000000	0.000000	1.000000	
max	98.00000	56.000000	999.000000	7.000000	1.000000	

8 rows × 60 columns



```
In [15]: df['Loan_Status_label'].value_counts()
```

```
Out[15]: Loan_Status_label
0      36548
1       4640
Name: count, dtype: int64
```

```
In [20]: print(round(len(df[df['Loan_Status_label']==0])/len(df),2))
print(round(len(df[df['Loan_Status_label']==1])/len(df),2))
```

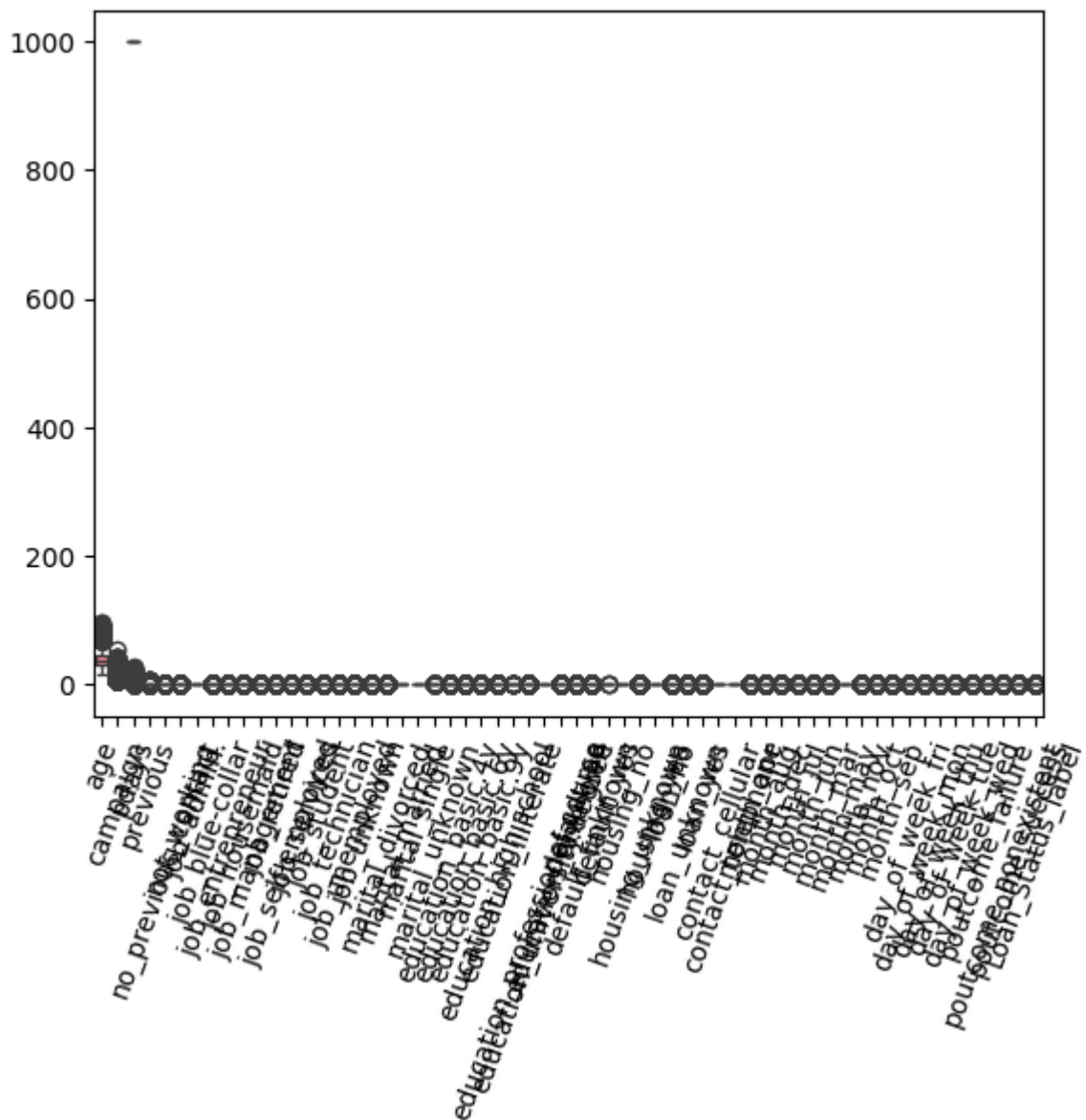
```
0.89
0.11
```

```
In [21]: df.isnull().sum()
```

```
Out[21]: age 0
campaign 0
pdays 0
previous 0
no_previous_contact 0
not_working 0
job_admin. 0
job_blue-collar 0
job_entrepreneur 0
job_housemaid 0
job_management 0
job_retired 0
job_self-employed 0
job_services 0
job_student 0
job_technician 0
job_unemployed 0
job_unknown 0
marital_divorced 0
marital_married 0
marital_single 0
marital_unknown 0
education_basic.4y 0
education_basic.6y 0
education_basic.9y 0
education_high.school 0
education_illiterate 0
education_professional.course 0
education_university.degree 0
education_unknown 0
default_no 0
default_unknown 0
default_yes 0
housing_no 0
housing_unknown 0
housing_yes 0
loan_no 0
loan_unknown 0
loan_yes 0
contact_cellular 0
contact_telephone 0
month_apr 0
month_aug 0
month_dec 0
month_jul 0
month_jun 0
month_mar 0
month_may 0
month_nov 0
month_oct 0
month_sep 0
day_of_week_fri 0
day_of_week_mon 0
day_of_week_thu 0
day_of_week_tue 0
day_of_week_wed 0
poutcome_failure 0
poutcome_nonexistent 0
poutcome_success 0
```

```
Loan_Status_label      0
dtype: int64
```

```
In [34]: sns.boxplot(df)
plt.xticks(rotation =70)
plt.show()
```



```
In [38]: Q1 = df.quantile(0.25)
Q3 = df.quantile(0.75)
IQR = Q3 - Q1
print(f'IQR : \n\n{IQR}')
```

IQR :

age	15.0
campaign	2.0
pdays	0.0
previous	0.0
no_previous_contact	0.0
not_working	0.0
job_admin.	1.0
job_blue-collar	0.0
job_entrepreneur	0.0
job_housemaid	0.0
job_management	0.0
job_retired	0.0
job_self-employed	0.0
job_services	0.0
job_student	0.0
job_technician	0.0
job_unemployed	0.0
job_unknown	0.0
marital_divorced	0.0
marital_married	1.0
marital_single	1.0
marital_unknown	0.0
education_basic.4y	0.0
education_basic.6y	0.0
education_basic.9y	0.0
education_high.school	0.0
education_illiterate	0.0
education_professional.course	0.0
education_university.degree	1.0
education_unknown	0.0
default_no	0.0
default_unknown	0.0
default_yes	0.0
housing_no	1.0
housing_unknown	0.0
housing_yes	1.0
loan_no	0.0
loan_unknown	0.0
loan_yes	0.0
contact_cellular	1.0
contact_telephone	1.0
month_apr	0.0
month_aug	0.0
month_dec	0.0
month_jul	0.0
month_jun	0.0
month_mar	0.0
month_may	1.0
month_nov	0.0
month_oct	0.0
month_sep	0.0
day_of_week_fri	0.0
day_of_week_mon	0.0
day_of_week_thu	0.0
day_of_week_tue	0.0
day_of_week_wed	0.0
poutcome_failure	0.0
poutcome_nonexistent	0.0

```
poutcome_success          0.0  
Loan_Status_label          0.0  
dtype: float64
```

```
In [41]: lower_bound= Q1 - 1.5*IQR  
         upper_bound= Q3 +1.5*IQR  
         print(f'Lower bound for outliers: \n\n{lower_bound}\n\n')  
         print(f'Upper bound for outliers : \n\n{upper_bound}\n\n')
```


Lower bound for outliers:

age	9.5
campaign	-2.0
pdays	999.0
previous	0.0
no_previous_contact	1.0
not_working	0.0
job_admin.	-1.5
job_blue-collar	0.0
job_entrepreneur	0.0
job_housemaid	0.0
job_management	0.0
job_retired	0.0
job_self-employed	0.0
job_services	0.0
job_student	0.0
job_technician	0.0
job_unemployed	0.0
job_unknown	0.0
marital_divorced	0.0
marital_married	-1.5
marital_single	-1.5
marital_unknown	0.0
education_basic.4y	0.0
education_basic.6y	0.0
education_basic.9y	0.0
education_high.school	0.0
education_illiterate	0.0
education_professional.course	0.0
education_university.degree	-1.5
education_unknown	0.0
default_no	1.0
default_unknown	0.0
default_yes	0.0
housing_no	-1.5
housing_unknown	0.0
housing_yes	-1.5
loan_no	1.0
loan_unknown	0.0
loan_yes	0.0
contact_cellular	-1.5
contact_telephone	-1.5
month_apr	0.0
month_aug	0.0
month_dec	0.0
month_jul	0.0
month_jun	0.0
month_mar	0.0
month_may	-1.5
month_nov	0.0
month_oct	0.0
month_sep	0.0
day_of_week_fri	0.0
day_of_week_mon	0.0
day_of_week_thu	0.0
day_of_week_tue	0.0
day_of_week_wed	0.0
poutcome_failure	0.0
poutcome_nonexistent	1.0

poutcome_success	0.0
Loan_Status_label	0.0
dtype: float64	

Upper bound for outliers :

age	69.5
campaign	6.0
pdays	999.0
previous	0.0
no_previous_contact	1.0
not_working	0.0
job_admin.	2.5
job_blue-collar	0.0
job_entrepreneur	0.0
job_housemaid	0.0
job_management	0.0
job_retired	0.0
job_self-employed	0.0
job_services	0.0
job_student	0.0
job_technician	0.0
job_unemployed	0.0
job_unknown	0.0
marital_divorced	0.0
marital_married	2.5
marital_single	2.5
marital_unknown	0.0
education_basic.4y	0.0
education_basic.6y	0.0
education_basic.9y	0.0
education_high.school	0.0
education_illiterate	0.0
education_professional.course	0.0
education_university.degree	2.5
education_unknown	0.0
default_no	1.0
default_unknown	0.0
default_yes	0.0
housing_no	2.5
housing_unknown	0.0
housing_yes	2.5
loan_no	1.0
loan_unknown	0.0
loan_yes	0.0
contact_cellular	2.5
contact_telephone	2.5
month_apr	0.0
month_aug	0.0
month_dec	0.0
month_jul	0.0
month_jun	0.0
month_mar	0.0
month_may	2.5
month_nov	0.0
month_oct	0.0
month_sep	0.0
day_of_week_fri	0.0
day_of_week_mon	0.0

```

day_of_week_thu          0.0
day_of_week_tue          0.0
day_of_week_wed          0.0
poutcome_failure         0.0
poutcome_nonexistent     1.0
poutcome_success         0.0
Loan_Status_label        0.0
dtype: float64n

```

In [42]: `df.shape`

Out[42]: (41188, 60)

In [43]: `df[((df<lower_bound) | (df>upper_bound)).any(axis=1)]`

Out[43]:

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
0	56	1	999	0	1	0	0
1	57	1	999	0	1	0	0
2	37	1	999	0	1	0	0
3	40	1	999	0	1	0	1
4	56	1	999	0	1	0	0
...
41183	73	1	999	0	1	1	0
41184	46	1	999	0	1	0	0
41185	56	2	999	0	1	1	0
41186	44	1	999	0	1	0	0
41187	74	3	999	1	1	1	0

41188 rows × 60 columns



In [44]: `print(lower_bound)`
`print(upper_bound)`

age	9.5
campaign	-2.0
pdays	999.0
previous	0.0
no_previous_contact	1.0
not_working	0.0
job_admin.	-1.5
job_blue-collar	0.0
job_entrepreneur	0.0
job_housemaid	0.0
job_management	0.0
job_retired	0.0
job_self-employed	0.0
job_services	0.0
job_student	0.0
job_technician	0.0
job_unemployed	0.0
job_unknown	0.0
marital_divorced	0.0
marital_married	-1.5
marital_single	-1.5
marital_unknown	0.0
education_basic.4y	0.0
education_basic.6y	0.0
education_basic.9y	0.0
education_high.school	0.0
education_illiterate	0.0
education_professional.course	0.0
education_university.degree	-1.5
education_unknown	0.0
default_no	1.0
default_unknown	0.0
default_yes	0.0
housing_no	-1.5
housing_unknown	0.0
housing_yes	-1.5
loan_no	1.0
loan_unknown	0.0
loan_yes	0.0
contact_cellular	-1.5
contact_telephone	-1.5
month_apr	0.0
month_aug	0.0
month_dec	0.0
month_jul	0.0
month_jun	0.0
month_mar	0.0
month_may	-1.5
month_nov	0.0
month_oct	0.0
month_sep	0.0
day_of_week_fri	0.0
day_of_week_mon	0.0
day_of_week_thu	0.0
day_of_week_tue	0.0
day_of_week_wed	0.0
poutcome_failure	0.0
poutcome_nonexistent	1.0
poutcome_success	0.0
Loan_Status_label	0.0

dtype: float64	
age	69.5
campaign	6.0
pdays	999.0
previous	0.0
no_previous_contact	1.0
not_working	0.0
job_admin.	2.5
job_blue-collar	0.0
job_entrepreneur	0.0
job_housemaid	0.0
job_management	0.0
job_retired	0.0
job_self-employed	0.0
job_services	0.0
job_student	0.0
job_technician	0.0
job_unemployed	0.0
job_unknown	0.0
marital_divorced	0.0
marital_married	2.5
marital_single	2.5
marital_unknown	0.0
education_basic.4y	0.0
education_basic.6y	0.0
education_basic.9y	0.0
education_high.school	0.0
education_illiterate	0.0
education_professional.course	0.0
education_university.degree	2.5
education_unknown	0.0
default_no	1.0
default_unknown	0.0
default_yes	0.0
housing_no	2.5
housing_unknown	0.0
housing_yes	2.5
loan_no	1.0
loan_unknown	0.0
loan_yes	0.0
contact_cellular	2.5
contact_telephone	2.5
month_apr	0.0
month_aug	0.0
month_dec	0.0
month_jul	0.0
month_jun	0.0
month_mar	0.0
month_may	2.5
month_nov	0.0
month_oct	0.0
month_sep	0.0
day_of_week_fri	0.0
day_of_week_mon	0.0
day_of_week_thu	0.0
day_of_week_tue	0.0
day_of_week_wed	0.0
poutcome_failure	0.0
poutcome_nonexistent	1.0
poutcome_success	0.0

```
Loan_Status_label          0.0  
dtype: float64
```

```
In [45]: ((df<lower_bound) | (df>upper_bound)).sum()
```

```

Out[45]: age                                469
         campaign                            2406
         pdays                             1515
         previous                           5625
         no_previous_contact                1515
         not_working                       3609
         job_admin.                         0
         job_blue-collar                   9254
         job_entrepreneur                   1456
         job_housemaid                     1060
         job_management                     2924
         job_retired                       1720
         job_self-employed                 1421
         job_services                       3969
         job_student                       875
         job_technician                    6743
         job_unemployed                    1014
         job_unknown                       330
         marital_divorced                  4612
         marital_married                   0
         marital_single                     0
         marital_unknown                   80
         education_basic.4y                4176
         education_basic.6y                2292
         education_basic.9y                6045
         education_high.school             9515
         education_illiterate              18
         education_professional.course     5243
         education_university.degree       0
         education_unknown                 1731
         default_no                        8600
         default_unknown                   8597
         default_yes                       3
         housing_no                        0
         housing_unknown                   990
         housing_yes                       0
         loan_no                           7238
         loan_unknown                      990
         loan_yes                           6248
         contact_cellular                  0
         contact_telephone                 0
         month_apr                         2632
         month_aug                         6178
         month_dec                         182
         month_jul                         7174
         month_jun                         5318
         month_mar                         546
         month_may                         0
         month_nov                        4101
         month_oct                         718
         month_sep                         570
         day_of_week_fri                   7827
         day_of_week_mon                   8514
         day_of_week_thu                   8623
         day_of_week_tue                   8090
         day_of_week_wed                   8134
         poutcome_failure                  4252
         poutcome_nonexistent              5625
         poutcome_success                  1373

```

```
Loan_Status_label          4640  
dtype: int64
```

```
In [46]: outliers = df[((df<lower_bound)|(df>upper_bound)).any(axis =1)]  
print(outliers.shape)
```

```
(41188, 60)
```

```
In [47]: df_no_outliers=df[~((df<lower_bound)|(df>upper_bound)).any(axis=1)]  
df_no_outliers.shape
```

```
Out[47]: (0, 60)
```

```
In [48]: df.corr()
```


Out[48]:

	age	campaign	pdays	previous	no_previous_cc
age	1.000000	0.004594	-0.034369	0.024365	-0.0
campaign	0.004594	1.000000	0.052584	-0.079141	0.0
pdays	-0.034369	0.052584	1.000000	-0.587514	0.9
previous	0.024365	-0.079141	-0.587514	1.000000	-0.5
no_previous_contact	-0.034292	0.052569	0.999992	-0.587462	1.0
not_working	0.207547	-0.017529	-0.112797	0.104066	-0.1
job_admin.	-0.102584	0.011745	-0.025398	0.018888	-0.0
job_blue-collar	-0.024191	-0.001775	0.065335	-0.054845	0.0
job_entrepreneur	0.031213	-0.002203	0.019246	-0.013239	0.0
job_housemaid	0.085403	0.004226	-0.001649	-0.011569	-0.0
job_management	0.062040	-0.009135	0.000254	0.006735	0.0
job_retired	0.440772	-0.006847	-0.072084	0.065113	-0.0
job_self-employed	-0.001356	0.006361	0.014307	-0.011231	0.0
job_services	-0.065731	0.002383	0.030573	-0.011885	0.0
job_student	-0.199749	-0.024657	-0.096367	0.104671	-0.0
job_technician	-0.064383	0.001557	0.004575	-0.017142	0.0
job_unemployed	-0.004426	-0.000200	-0.023085	0.008426	-0.0
job_unknown	0.047773	0.002625	-0.011419	-0.003345	-0.0
marital_divorced	0.166125	0.005872	0.011711	-0.003066	0.0
marital_married	0.271266	0.002332	0.031614	-0.043420	0.0
marital_single	-0.411703	-0.007624	-0.042015	0.048485	-0.0
marital_unknown	0.001062	0.009873	-0.005989	0.009095	-0.0
education_basic.4y	0.244079	0.004000	0.005804	-0.017601	0.0
education_basic.6y	0.009897	-0.000991	0.023238	-0.021493	0.0
education_basic.9y	-0.038320	-0.005278	0.036963	-0.026418	0.0
education_high.school	-0.106550	0.000195	0.005522	0.014347	0.0
education_illiterate	0.017007	-0.002188	-0.002086	-0.002613	-0.0
education_professional.course	0.002054	0.002554	-0.003493	-0.007631	-0.0
education_university.degree	-0.071138	-0.000950	-0.036958	0.025419	-0.0
education_unknown	0.069485	0.002162	-0.022014	0.022641	-0.0
default_no	-0.165019	-0.032923	-0.080072	0.102533	-0.0
default_unknown	0.165001	0.033007	0.080047	-0.102604	0.0

	age	campaign	pdays	previous	no_previous_cc
default_yes	0.001891	-0.003803	0.001668	0.002766	0.0
housing_no	0.001755	0.010959	0.010131	-0.021677	0.0
housing_unknown	-0.001092	-0.000396	0.002033	0.003769	0.0
housing_yes	-0.001414	-0.010800	-0.010719	0.020446	-0.0
loan_no	0.007224	-0.004831	-0.000737	0.000504	-0.0
loan_unknown	-0.001092	-0.000396	0.002033	0.003769	0.0
loan_yes	-0.007198	0.005294	-0.000086	-0.002143	-0.0
contact_cellular	-0.007021	-0.077368	-0.117970	0.212848	-0.1
contact_telephone	0.007021	0.077368	0.117970	-0.212848	0.1
month_apr	0.014285	-0.058519	-0.009629	0.079384	-0.0
month_aug	0.065309	0.015009	-0.003176	-0.048852	-0.0
month_dec	0.046563	-0.009819	-0.076425	0.059557	-0.0
month_jul	-0.043563	0.098385	0.049695	-0.116785	0.0
month_jun	-0.009532	0.070144	0.015957	-0.069905	0.0
month_mar	0.008390	-0.015244	-0.078769	0.072733	-0.0
month_may	-0.067523	-0.032704	0.070690	-0.012534	0.0
month_nov	0.030045	-0.079028	-0.016946	0.081050	-0.0
month_oct	0.053008	-0.049126	-0.128677	0.127754	-0.1
month_sep	0.040889	-0.033953	-0.154629	0.157651	-0.1
day_of_week_fri	0.007148	0.026641	0.013782	0.004404	0.0
day_of_week_mon	0.019027	0.014405	0.001031	-0.002012	0.0
day_of_week_thu	-0.021356	0.005141	-0.008186	0.001512	-0.0
day_of_week_tue	0.019137	-0.025125	-0.007300	0.000090	-0.0
day_of_week_wed	-0.023671	-0.021088	0.001021	-0.003929	0.0
poutcome_failure	-0.002537	-0.068890	0.006354	0.682608	0.0
poutcome_nonexistent	-0.016376	0.087645	0.491353	-0.878776	0.4
poutcome_success	0.035626	-0.050893	-0.950700	0.524045	-0.9
Loan_Status_label	0.030399	-0.066357	-0.324914	0.230181	-0.3

60 rows × 60 columns

```
In [53]: df.corr()['Loan_Status_label']
```

```

Out[53]: age                                0.030399
         campaign                           -0.066357
         pdays                             -0.324914
         previous                           0.230181
         no_previous_contact                -0.324877
         not_working                        0.121246
         job_admin.                         0.031426
         job_blue-collar                   -0.074423
         job_entrepreneur                   -0.016644
         job_housemaid                     -0.006505
         job_management                     -0.000419
         job_retired                       0.092221
         job_self-employed                  -0.004663
         job_services                       -0.032301
         job_student                        0.093955
         job_technician                     -0.006149
         job_unemployed                     0.014752
         job_unknown                       -0.000151
         marital_divorced                   -0.010608
         marital_married                    -0.043398
         marital_single                     0.054133
         marital_unknown                    0.005211
         education_basic.4y                 -0.010798
         education_basic.6y                 -0.023517
         education_basic.9y                 -0.045135
         education_high.school              -0.007452
         education_illiterate               0.007246
         education_professional.course      0.001003
         education_university.degree        0.050364
         education_unknown                  0.021430
         default_no                         0.099344
         default_unknown                    -0.099293
         default_yes                        -0.003041
         housing_no                         -0.011085
         housing_unknown                    -0.002270
         housing_yes                        0.011743
         loan_no                            0.005123
         loan_unknown                       -0.002270
         loan_yes                           -0.004466
         contact_cellular                   0.144773
         contact_telephone                  -0.144773
         month_apr                          0.076136
         month_aug                           -0.008813
         month_dec                           0.079303
         month_jul                           -0.032230
         month_jun                           -0.009182
         month_mar                           0.144014
         month_may                           -0.108271
         month_nov                           -0.011796
         month_oct                           0.137366
         month_sep                           0.126067
         day_of_week_fri                     -0.006996
         day_of_week_mon                     -0.021265
         day_of_week_thu                     0.013888
         day_of_week_tue                     0.008046
         day_of_week_wed                     0.006302
         poutcome_failure                    0.031799
         poutcome_nonexistent                -0.193507
         poutcome_success                    0.316269

```

```
Loan_Status_label          1.000000
Name: Loan_Status_label, dtype: float64
```

```
In [11]: x=df.drop(columns=['Loan_Status_label'],axis=1)
y=df['Loan_Status_label']
```

```
In [56]: x
```

```
Out[56]:
```

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
0	56	1	999	0	1	0	0
1	57	1	999	0	1	0	0
2	37	1	999	0	1	0	0
3	40	1	999	0	1	0	1
4	56	1	999	0	1	0	0
...
41183	73	1	999	0	1	1	0
41184	46	1	999	0	1	0	0
41185	56	2	999	0	1	1	0
41186	44	1	999	0	1	0	0
41187	74	3	999	1	1	1	0

41188 rows × 59 columns



```
In [57]: y
```

```
Out[57]: 0      0
1      0
2      0
3      0
4      0
..
41183   1
41184   0
41185   0
41186   1
41187   0
Name: Loan_Status_label, Length: 41188, dtype: int64
```

```
In [17]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
```

```
In [18]: x_train
```

Out[18]:

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
12556	40	2	999	0	1	0	0
35451	31	4	999	0	1	0	1
30592	59	6	999	1	1	1	0
17914	43	5	999	0	1	0	0
3315	39	2	999	0	1	0	1
...
6265	58	2	999	0	1	1	0
11284	37	1	999	0	1	0	0
38158	35	1	4	1	0	0	1
860	40	2	999	0	1	0	0
15795	29	2	999	0	1	0	1

32950 rows × 59 columns

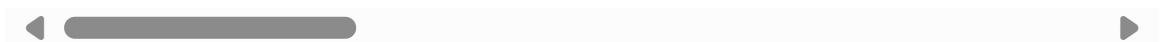


In [62]: x_test

Out[62]:

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
32884	57	1	999	1	1	0	0
3169	55	2	999	0	1	0	0
32206	33	1	999	1	1	0	0
9403	36	4	999	0	1	0	1
14020	27	2	999	0	1	0	0
...
12322	27	1	999	0	1	0	0
23440	41	7	999	0	1	0	0
29431	46	12	999	1	1	0	0
16627	31	2	999	0	1	0	1
1871	59	2	999	0	1	0	1

8238 rows × 59 columns



In []:

In [20]: `from sklearn.preprocessing import StandardScaler`

```
In [24]: sc = StandardScaler()

x_train_sc= sc.fit_transform(x_train)
x_test_sc=sc.fit_transform(x_test)
```

```
In [22]: x_train
```

```
Out[22]:
```

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
12556	40	2	999	0	1	0	0
35451	31	4	999	0	1	0	1
30592	59	6	999	1	1	1	0
17914	43	5	999	0	1	0	0
3315	39	2	999	0	1	0	1
...
6265	58	2	999	0	1	1	0
11284	37	1	999	0	1	0	0
38158	35	1	4	1	0	0	1
860	40	2	999	0	1	0	0
15795	29	2	999	0	1	0	1

32950 rows × 59 columns



```
In [23]: x_train_sc
```

```
Out[23]: array([[ -1.66930454e-03, -2.06241614e-01,  1.94660673e-01, ...,
        -3.40481710e-01,  3.98373809e-01, -1.84965343e-01],
       [-8.64094846e-01,  5.13675879e-01,  1.94660673e-01, ...,
        -3.40481710e-01,  3.98373809e-01, -1.84965343e-01],
       [ 1.81900684e+00,  1.23359337e+00,  1.94660673e-01, ...,
        2.93701532e+00, -2.51020518e+00, -1.84965343e-01],
       ...,
       [-4.80794606e-01, -5.66200360e-01, -5.14775262e+00, ...,
        -3.40481710e-01, -2.51020518e+00,  5.40641820e+00],
       [-1.66930454e-03, -2.06241614e-01,  1.94660673e-01, ...,
        -3.40481710e-01,  3.98373809e-01, -1.84965343e-01],
       [-1.05574497e+00, -2.06241614e-01,  1.94660673e-01, ...,
        -3.40481710e-01,  3.98373809e-01, -1.84965343e-01]],
      shape=(32950, 59))
```

```
In [26]: x_test_sc
```

```
Out[26]: array([[ 1.63562386, -0.56486335,  0.19840223, ...,  2.98952736,
                 -2.53145677, -0.18861365],
                [ 1.44262285, -0.19952267,  0.19840223, ..., -0.33450104,
                 0.39502946, -0.18861365],
                [-0.68038832, -0.56486335,  0.19840223, ...,  2.98952736,
                 -2.53145677, -0.18861365],
                ...,
                [ 0.57411828,  3.45388416,  0.19840223, ...,  2.98952736,
                 -2.53145677, -0.18861365],
                [-0.87338934, -0.19952267,  0.19840223, ..., -0.33450104,
                 0.39502946, -0.18861365],
                [ 1.82862488, -0.19952267,  0.19840223, ..., -0.33450104,
                 0.39502946, -0.18861365]], shape=(8238, 59))
```

```
In [27]: lr = LogisticRegression()
lr.fit(x_train_sc,y_train)
```

```
Out[27]: ▼ LogisticRegression ⓘ ?
          ► Parameters
```

```
In [28]: lr.score(x_train,y_train)
```

```
Out[28]: 0.8911380880121396
```

```
In [29]: lr.score(x_test,y_test)
```

```
Out[29]: 0.8880796309783928
```

```
In [30]: print(f'Training Accuracy :{round(lr.score(x_train,y_train),2)*100}%')
print(f'Test Accuracy      :{round(lr.score(x_test,y_test),2)*100}%')
```

```
Training Accuracy :89.0%
Test Accuracy      :89.0%
```

```
In [ ]:
```

```
In [37]: knn = KNeighborsRegressor(n_neighbors=7)
knn.fit(x_train,y_train)
```

```
Out[37]: ▼ KNeighborsRegressor ⓘ ?
          ► Parameters
```

```
In [45]: print(f'Training Accuracy :{round(knn.score(x_train,y_train),2)*100}%')
print(f'Test Accuracy      :{round(knn.score(x_test,y_test),2)*100}%')
```

```
Training Accuracy :31.0%
Test Accuracy      :5.0%
```

```
In [48]: dtree = DecisionTreeRegressor(max_depth=9)
dtree.fit(x_train,y_train)
```

Out[48]:

▼ DecisionTreeRegressor ⓘ ?
► Parameters

```
In [49]: print(f'Training Accuracy:{round(dtree.score(x_train,y_train),2)*100}%')  
print(f'Test Accuracy:{round(dtree.score(x_test,y_test),2)*100}%')
```

Training Accuracy:25.0%

Test Accuracy:11.0%

```
In [62]: lr = LinearRegression()  
lr.fit(x_train,y_train)
```

Out[62]:

▼ LinearRegression ⓘ ?
► Parameters

```
In [54]: lr.coef_
```

```
Out[54]: array([[ 0.00201149, -0.08745845, -0.00212815,  0.16239052,  0.00085786,  
                  0.47454794,  0.05473844, -0.15383082, -0.07670865, -0.0420984 ,  
                 -0.06475622,  0.26319342, -0.04324801, -0.13699808,  0.24708782,  
                 -0.05459497, -0.0357333 , -0.01857775, -0.15697032, -0.00646726,  
                  0.09798183,  0.00392923, -0.00727861, -0.00354071, -0.11126238,  
                 -0.05912369,  0.00692359, -0.041064 ,  0.07190036,  0.08191892,  
                  0.31694894, -0.37661627, -0.00185919, -0.05130146,  0.01889007,  
                 -0.02911512, -0.0091884 ,  0.01889007, -0.07122818,  0.39499941,  
                 -0.45652593,  0.32585489, -0.66403745,  0.15324375, -0.49246554,  
                  0.30350426,  0.66817654, -0.60034406, -0.65798655,  0.53600784,  
                  0.36651981, -0.02030538, -0.20352329,  0.03823005,  0.04925475,  
                  0.07481735, -0.0494228 , -0.03076258,  0.01865887]])
```

```
In [55]: x.columns
```

```
Out[55]: Index(['age', 'campaign', 'pdays', 'previous', 'no_previous_contact',  
                'not_working', 'job_admin.', 'job_blue-collar', 'job_entrepreneur',  
                'job_housemaid', 'job_management', 'job_retired', 'job_self-employed',  
                'job_services', 'job_student', 'job_technician', 'job_unemployed',  
                'job_unknown', 'marital_divorced', 'marital_married', 'marital_single',  
                'marital_unknown', 'education_basic.4y', 'education_basic.6y',  
                'education_basic.9y', 'education_high.school', 'education_illiterate',  
                'education_professional.course', 'education_university.degree',  
                'education_unknown', 'default_no', 'default_unknown', 'default_yes',  
                'housing_no', 'housing_unknown', 'housing_yes', 'loan_no',  
                'loan_unknown', 'loan_yes', 'contact_cellular', 'contact_telephone',  
                'month_apr', 'month_aug', 'month_dec', 'month_jul', 'month_jun',  
                'month_mar', 'month_may', 'month_nov', 'month_oct', 'month_sep',  
                'day_of_week_fri', 'day_of_week_mon', 'day_of_week_thu',  
                'day_of_week_tue', 'day_of_week_wed', 'poutcome_failure',  
                'poutcome_nonexistent', 'poutcome_success'],  
               dtype='object')
```

```
In [63]: pd.DataFrame(lr.coef_, index=x.columns, columns=['Coefficients/weights'])
```


Out[63]:

	Coefficients/weights
age	0.000468
campaign	-0.004046
pdays	-0.002870
previous	0.038687
no_previous_contact	2.543796
not_working	0.042511
job_admin.	0.006092
job_blue-collar	-0.008288
job_entrepreneur	-0.007919
job_housemaid	-0.003928
job_management	-0.004909
job_retired	0.022905
job_self-employed	-0.007474
job_services	-0.007328
job_student	0.044514
job_technician	0.001932
job_unemployed	-0.024909
job_unknown	-0.010688
marital_divorced	-0.010373
marital_married	-0.003805
marital_single	0.006759
marital_unknown	0.007420
education_basic.4y	-0.014540
education_basic.6y	-0.012719
education_basic.9y	-0.020613
education_high.school	-0.016809
education_illiterate	0.089458
education_professional.course	-0.016766
education_university.degree	-0.002353
education_unknown	-0.005658
default_no	0.042344
default_unknown	0.010154
default_yes	-0.052498

Coefficients/weights	
housing_no	-0.000328
housing_unknown	0.002211
housing_yes	-0.001883
loan_no	-0.000511
loan_unknown	0.002211
loan_yes	-0.001700
contact_cellular	0.036052
contact_telephone	-0.036052
month_apr	-0.026038
month_aug	-0.127250
month_dec	0.121278
month_jul	-0.112827
month_jun	-0.052237
month_mar	0.215443
month_may	-0.109137
month_nov	-0.129622
month_oct	0.127199
month_sep	0.093191
day_of_week_fri	-0.002314
day_of_week_mon	-0.015840
day_of_week_thu	0.005647
day_of_week_tue	0.004898
day_of_week_wed	0.007608
poutcome_failure	-0.065673
poutcome_nonexistent	-0.010402
poutcome_success	0.076076

In [64]: lr

Out[64]:

LinearRegression

Parameters

In [68]: y_pred_train=lr.predict(x_train)
y_pred_test=lr.predict(x_test)

```
In [65]: x_train[:3]
```

```
Out[65]:
```

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
12556	40	2	999	0	1	0	0
35451	31	4	999	0	1	0	1
30592	59	6	999	1	1	1	0

3 rows × 59 columns

```
In [66]: y_train[:3]
```

```
Out[66]: 12556    0
          35451    0
          30592    0
          Name: Loan_Status_label, dtype: int64
```

```
In [69]: y_pred_train[:3]
```

```
Out[69]: array([-0.03354994,  0.09632921,  0.13188504])
```

```
In [70]: x_test[:3]
```

```
Out[70]:
```

	age	campaign	pdays	previous	no_previous_contact	not_working	job_admin.
32884	57	1	999	1	1	0	0
3169	55	2	999	0	1	0	0
32206	33	1	999	1	1	0	0

3 rows × 59 columns

```
In [71]: y_test[:3]
```

```
Out[71]: 32884    0
          3169    0
          32206    0
          Name: Loan_Status_label, dtype: int64
```

```
In [72]: y_pred_test
```

```
Out[72]: array([0.08423694, 0.01119656, 0.07370406, ..., 0.13130264, 0.13318661,
                0.05738234], shape=(8238,))
```

```
In [78]: def training_evaluation(actual,predicted):
          mse =mean_squared_error(actual,predicted)
          rmse=np.sqrt(mean_squared_error(actual,predicted))
          mae=mean_absolute_error(actual,predicted)
          r2=r2_score(actual,predicted)

          print(f'MSE      :{mse}')
```

```
print(f'RMSE    :{rmse}')
print(f'MAE     :{mae}')
print(f'R2_score :{r2}')
print()

plt.scatter(actual,predicted,color='r',label='Actual price')
plt.plot([actual.min(),actual.max()], [predicted.min(),predicted.max()],color='b')
plt.title('Best fit line - Training data')
plt.xlabel('Actual price')
plt.ylabel('Predicted price')
plt.legend()
plt.show()
```

In [79]: training_evaluation(y_train,y_pred_train)

```
MSE      :0.08261930553215971
RMSE     :0.28743574157045904
MAE      :0.1667332223085798
R2_score :0.17214830947233095
```

AttributeError Traceback (most recent call last)

Cell In[79], line 1

```
----> 1 training_evaluation(y_train,y_pred_train)
```

Cell In[78], line 14, in training_evaluation(actual, predicted)

```
11 print()
13 plt.scatter(actual,predicted,color='r',label='Actual price')
--> 14 plt.plot([actual.min(),actual.max()], [predicted.min(),predicted.max()],co
lors=,lw=3,label=)
15 plt.title('Best fit line - Training data')
16 plt.xlabel('Actual price')
```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\pyplot.py:3838, in plot(scalex, scaley, data, *args, **kwargs)

```
3830 @_copy_docstring_and_deprecators(Axes.plot)
3831 def plot(
3832     *args: float | ArrayLike | str,
3833     (...) 3836     **kwargs,
3837 ) -> list[Line2D]:
-> 3838     return gca().plot(
3839         *args,
3840         scalex=scalex,
3841         scaley=scaley,
3842         **({ : data} if data is not None else {}),
3843         **kwargs,
3844     )
```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\axes_axes.py:1777, in Axes.plot(self, scalex, scaley, data, *args, **kwargs)

```
1534 """
1535 Plot y versus x as lines and/or markers.
1536
1537 (...) 1774 (``'green'``) or hex strings (``'#008000'``).
1775 """
1776 kwargs = cbook.normalize_kwargs(kwargs, mlines.Line2D)
-> 1777 lines = [*self._get_lines(self, *args, data=data, **kwargs)]
1778 for line in lines:
1779     self.add_line(line)
```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\axes_base.py:297, in _process_plot_var_args.__call__(self, axes, data, return_kwards, *args, **kwargs)

```
295     this += args[0],
296     args = args[1:]
--> 297 yield from self._plot_args(
298     axes, this, kwargs, ambiguous_fmt_datakey=ambiguous_fmt_datakey,
299     return_kwards=return_kwards
300 )
```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\axes_base.py:546, in _process_plot_var_args._plot_args(self, axes, tup, kwargs, return_kwards, ambiguous_fmt_datakey)

```
544     return list(result)
545 else:
--> 546     return [l[0] for l in result]
```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\axes_base.py:539, in <genexpr>(.0)

```
534 else:
```

```

535     raise ValueError(
536         f"label must be scalar or have the same length as the input "
537         f"data, but found {len(label)} for {n_datasets} datasets.")
--> 539 result = (make_artist(axes, x[:, j % ncx], y[:, j % ncy], kw,
540                        {**kwargs, : label}))
541         for j, label in enumerate(labels))
543 if return_kwargs:
544     return list(result)

```

```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\axes
\_base.py:338, in _process_plot_var_args._make_line(self, axes, x, y, kw, kwargs)
336 kw = {**kw, **kwargs} # Don't modify the original kw.
337 self._setdefaults(self._getdefaults(kw), kw)
--> 338 seg = mlines.Line2D(x, y, **kw)
339 return seg, kw

```

```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\line
s.py:421, in Line2D.__init__(self, xdata, ydata, linewidth, linestyle, color, gap
color, marker, markersize, markeredgewidth, markeredgewidth, markeredgewidth, mar
kerfacecoloralt, fillstyle, antialiased, dash_capstyle, solid_capstyle, dash_join
style, solid_joinstyle, pickradius, drawstyle, markevery, **kwargs)
417 self.set_mkeredgewidth(markedgewidth)
419 # update kwargs before updating data to give the caller a
420 # chance to init axes (and hence unit support)
--> 421 self._internal_update(kwargs)
422 self.pickradius = pickradius
423 self.ind_offset = 0

```

```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\artis
t.py:1233, in Artist._internal_update(self, kwargs)
1226 def _internal_update(self, kwargs):
1227     """
1228     Update artist properties without prenormalizing them, but generating
1229     errors as if calling `set`.
1230
1231     The lack of prenormalization is to maintain backcompatibility.
1232     """
-> 1233     return self._update_props(
1234         kwargs, {cls.__name__}
1235         {prop_name!r} )

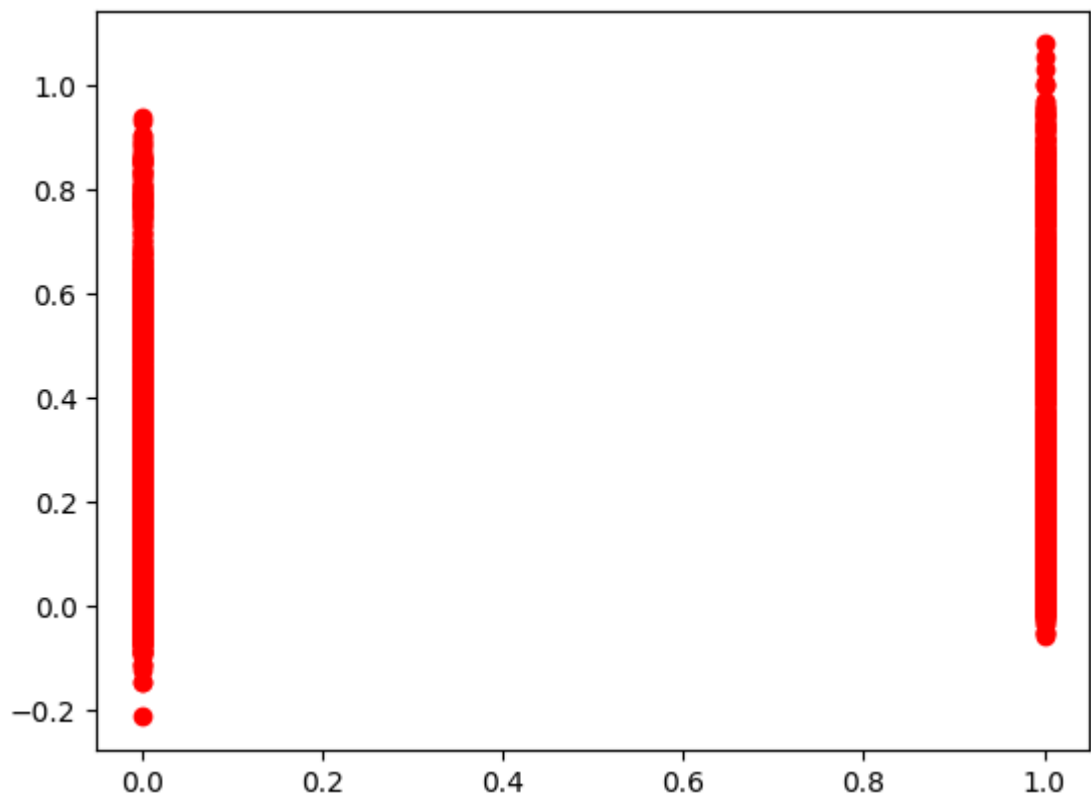
```

```

File ~\AppData\Local\Programs\Python\Python313\Lib\site-packages\matplotlib\artis
t.py:1206, in Artist._update_props(self, props, errfmt)
1204         func = getattr(self, f"set_{k}", None)
1205         if not callable(func):
-> 1206             raise AttributeError(
1207                 errfmt.format(cls=type(self), prop_name=k),
1208                 name=k)
1209         ret.append(func(v))
1210 if ret:

```

AttributeError: Line2D.set() got an unexpected keyword argument 'colors'



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