#### **Banking Dataset**

Term deposits serve as a significant revenue stream for banks, representing cash investments held within financial institutions. These investments involve committing funds for a predetermined period, during which they accrue interest at an agreed-upon rate. To promote term deposits, banks employ various outreach strategies including email marketing, advertisements, telephonic marketing, and digital marketing.

Despite the advent of digital channels, telephonic marketing campaigns persist as one of the most effective means of engaging customers. However, they necessitate substantial investment due to the requirement of large call centers to execute these campaigns. Therefore, it becomes essential to pre-identify potential customers likely to convert, enabling targeted outreach efforts via phone calls.

The data is related to direct marketing campaigns (phone calls) of a Portuguese banking institution. The classification goal is to predict if the client will subscribe to a term deposit (variable y).

#### Content

The data is related to the 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 by the customer or not. The data folder contains two datasets:-

Banking\_data.csv: 45,211 rows and 18 columns ordered by date (from May 2008 to November 2010)

**Detailed Column Descriptions:** 

**age**: This column represents the age of the bank client. It's a numeric variable indicating the age in years.

**job**: This column indicates the type of job the client has. It's a categorical variable with options such as "admin.", "unknown", "unemployed", "management", etc.

marital: This column represents the marital status of the client. It's a categorical variable with options such as "married", "divorced", or "single".

**education**: This column indicates the level of education of the client. It's a categorical variable with options such as "unknown", "secondary", "primary", or "tertiary".

**default**: This column indicates whether the client has credit in default. It's a binary variable with options "yes" or "no".

**balance**: This column represents the average yearly balance in euros for the client. It's a numeric variable.

**housing**: This column indicates whether the client has a housing loan. It's a binary variable with options "yes" or "no".

**loan**: This column indicates whether the client has a personal loan. It's a binary variable with options "yes" or "no".

**contact**: This column represents the type of communication used to contact the client. It's a categorical variable with options such as "unknown", "telephone", or "cellular".

day: This column represents the last contact day of the month. It's a numeric variable.

**month**: This column represents the last contact month of the year. It's a categorical variable with options such as "jan", "feb", "mar", etc.

**duration**: This column represents the duration of the last contact in seconds. It's a numeric variable.

**campaign**: This column represents the number of contacts performed during this campaign and for this client. It's a numeric variable.

**pdays**: This column represents the number of days that passed by after the client was last contacted from a previous campaign. It's a numeric variable where -1 means the client was not previously contacted.

**previous**: This column represents the number of contacts performed before this campaign and for this client. It's a numeric variable.

**poutcome**: This column represents the outcome of the previous marketing campaign. It's a categorical variable with options such as "unknown", "other", "failure", or "success".

```
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
df=pd.read_csv(r'project1.csv')
pd.set_option('display.max_rows',None)
pd.set_option("display.max_columns",None)
print(df)
```

### Q1 - What is the distribution of age among the clients?

```
age_1=df["age"].value_counts()

print(age_1)

plt.hist(age_1,color="blue",edgecolor="skyblue")

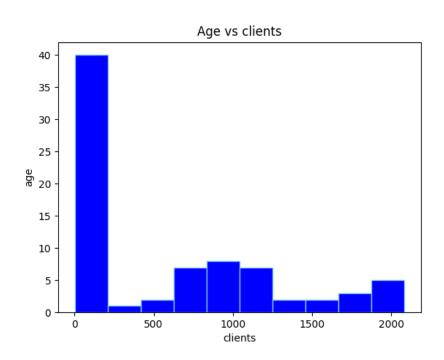
plt.xlabel("clients")

plt.ylabel("age")

plt.title("Age vs clients")

plt.savefig('graph1.png')

plt.show()
```



#### Q2 -What is the marital status distribution of the clients?

```
m_status=df["marital_status"].value_counts()
print(m_status)

plt.hist(m_status,bins=10,color="blue",edgecolor="lightblue"

plt.xlabel("clients") plt.ylabel("no

of individuals") plt.title("marital

status")

plt.savefig("graph2.png")

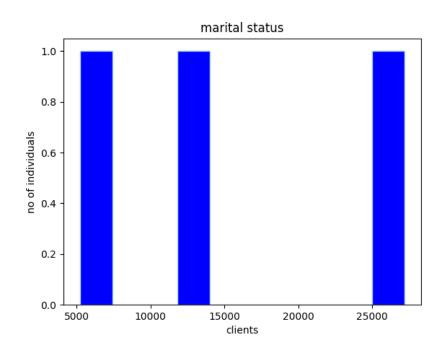
plt.show()
```

marital\_status

married 27216

single 12790

divorced 5207



### Q3- What is the level of education among the clients?

```
edu=df["education"].value_counts()
print(edu)
plt.hist(edu,bins=10,color="red",edgecolor="orange")
plt.xlabel("education")
plt.savefig("graph3.png")
plt.show()
```

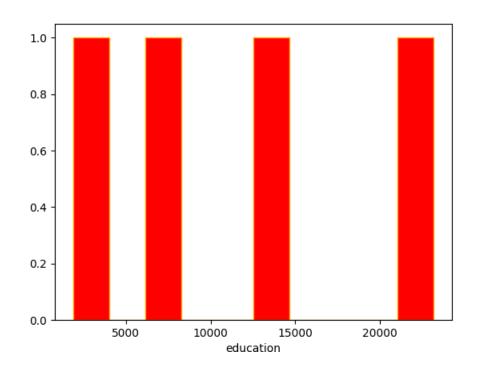
#### **education**

secondary 23204

tertiary 13301

primary 6851

unknown 1857



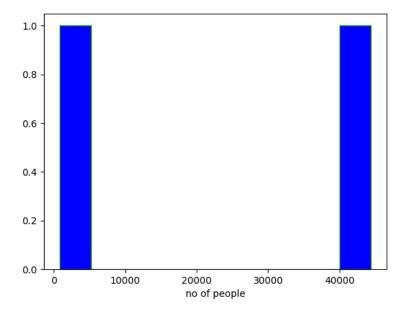
### Q4 - What proportion of clients have credit in default?

```
credit_customer=df["default"].value_counts()
print(credit_customer)
plt.hist(credit_customer,bins=10,color="blue",edgecolor="green")
plt.xlabel("no of people")
plt.savefig("graph4.png")
plt.show()
```

#### <mark>default</mark>

no 44401

yes 815



### Q5 - - How many clients have housing loans?

```
print(df.columns)

loan=df["housing"].value_counts() print(loan)

plt.hist(loan,bins=10,color="blue",edgecolor="green")

plt.xlabel("clients")

plt.ylabel("housing loans")

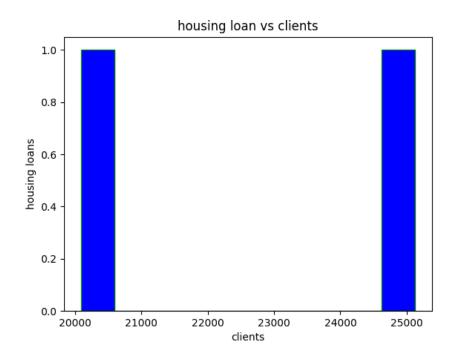
plt.title("housing loan vs clients")

plt.savefig("graph5.png")

plt.show()
```

yes 25130

no 20086



#### Q6 How many clients have personal loans?

```
loan_1=df["loan"].value_counts()
print(loan_1)

plt.hist(loan_1,bins=10,color="green",edgecolor="yellow")

plt.xlabel("no of clients")

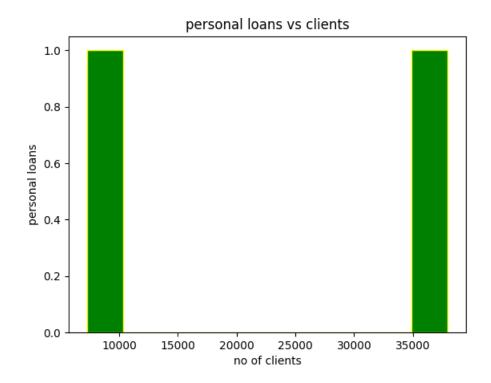
plt.ylabel("personal loans")

plt.title("personal loans vs clients")

plt.savefig("graph6.png")

plt.show()
```

#### yes 7244



# Q7 What are the communication types used for contacting clients during the campaign?

```
type=df["contact"].value_counts()
print(type)

plt.hist(type,bins=10,color="blue",edgecolor="green")

plt.xlabel("type")

plt.ylabel("clients")

plt.title("type vs clients")

plt.savefig("graph7.png")

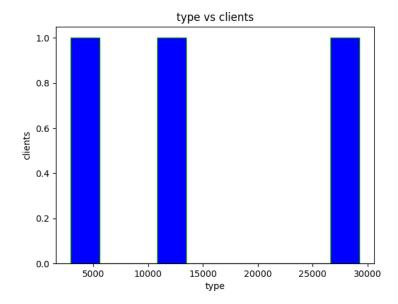
plt.show()
```

### <mark>contact</mark>

cellular 29290

unknown 13020

telephone 2906



#### Q8 What is the distribution of the last contact day of the month?

```
day_1=df["day"].value_counts()
print(day_1)
plt.hist(day_1,bins=10,color="blue",edgecolor="green")
plt.xlabel("clients")
plt.ylabel("day")
plt.title("day vs clients")
plt.savefig("graph8.png")
plt.show()
```

day

20 2752

**18 2308** 

**21 2026** 

**17 1942** 

<mark>6 1932</mark>

**5** 1910

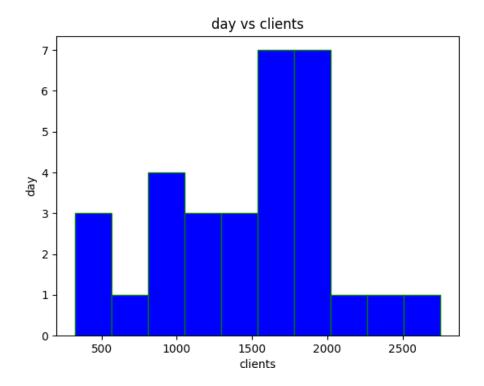
14 1848

**8** 1842

**28 1830** 

<mark>7 1817</mark>

- <mark>19 1757</mark>
- <mark>29 1745</mark>
- <mark>15 1703</mark>
- 12 1603
- 13 1585
- <mark>30 1566</mark>
- 9 1561
- <mark>11 1479</mark>
- 4 1445
- <mark>16 1417</mark>
- 2 1293
- <mark>27 1121</mark>
- 3 1079
- <mark>26 1035</mark>
- <mark>23 939</mark>
- <mark>22 905</mark>
- <mark>25 840</mark>
- **31** 643
- 10 524
- <mark>24 447</mark>
- 1 322



### Q9 How does the last contact month vary among the clients?

```
month1=df["month"].value_counts()
print(month1)

plt.hist(month1,bins=10,color="orange",edgecolor="green")
plt.xlabel("clients")
plt.ylabel("month")
plt.title("month vs clients")
plt.savefig("graph9.png")
plt.show()
```

### <mark>month</mark>

may 13766

<mark>jul 6895</mark>

aug 6247

jun 5341

<mark>nov 3975</mark>

<mark>apr 2932</mark>

feb 2649

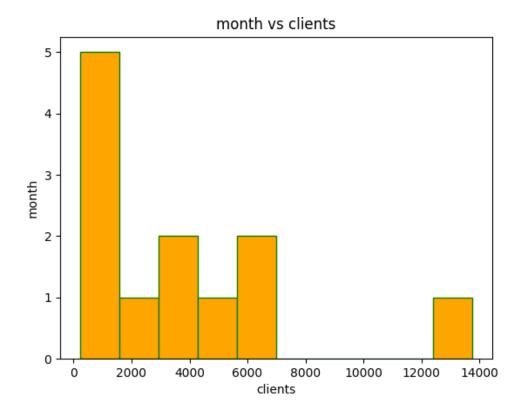
jan 1403

oct 738

<mark>sep 579</mark>

<mark>mar 477</mark>

dec 214



#### Q10 What is the distribution of the duration of the last contact?

```
# dur=df["duration"].value_counts()

# print(dur)

# plt.hist(dur,bins=10,color="violet",edgecolor="green")

# plt.xlabel("clients")

# plt.ylabel("duration")

# plt.title("clients vs duration ")

# plt.savefig("graph10.png")

# plt.show()
```

### <mark>duration</mark>

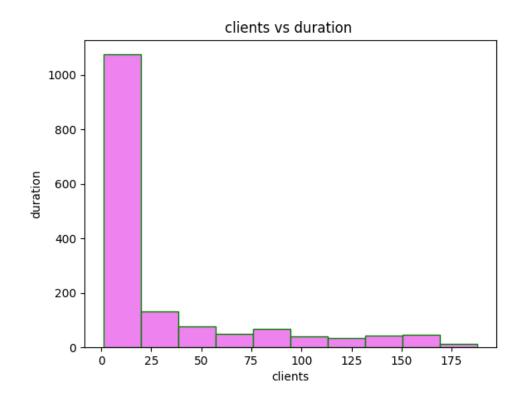
**124 188** 

90 184

1556 **1** 

1342

Name: count, Length: 1573, dtype: int64



## Q11 How many contacts were performed during the campaign for each client?

```
camp=df["campaign"].value_counts()
print(camp)

plt.hist(camp,bins=10,color="orange",edgecolor="green")

plt.xlabel("clients")

plt.ylabel("campaign")

plt.title("campaign vs clients")

plt.savefig("graph11.png")

plt.show()
```

#### <mark>campaign</mark>

- 1 17548
- <mark>2 12506</mark>
- <del>3 5521</del>
- 4 3522
- <mark>5 1764</mark>
- <mark>6 1291</mark>
- 7 735

**8** 540

9 327

**10 266** 

<u>11 201</u>

12 155

**13 133** 

<del>14</del> 93

<mark>15 84</mark>

<del>16 79</del>

**17** 69

18 <u>51</u>

<mark>19 44</mark>

20 43

21 35

22 23

**25 22** 

23 22

24 20

<mark>29 16</mark>

<mark>28 16</mark>

<mark>26 13</mark>

31 12

27 10

<mark>32 9</mark>

<mark>30 8</mark>

<mark>33 6</mark>

<mark>34 5</mark>

<mark>36 4</mark>

<mark>35 4</mark>

<mark>43 3</mark>

<mark>38 3</mark>

<mark>37 2</mark>

<mark>50 2</mark>

<mark>41 2</mark>

<mark>46 1</mark>

<mark>58 1</mark>

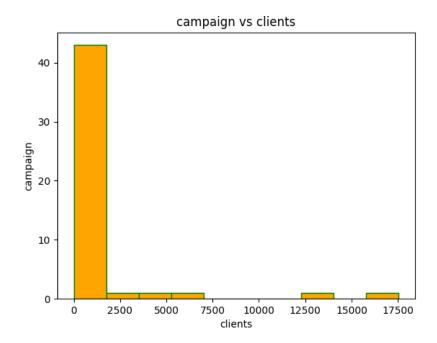
<mark>55 1</mark>

<mark>63 1</mark>

<mark>51 1</mark>

<mark>39 1</mark>

<mark>44 1</mark>

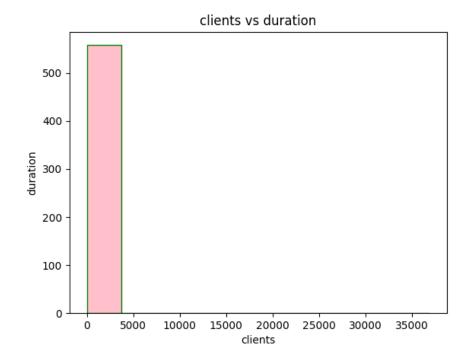


## Q12 What is the distribution of the number of days passed since the client was last contacted from a previous campaign?

```
days=df["pdays"].value_counts()
print(days)
plt.hist(days,bins=10,color="pink",edgecolor="green")
plt.xlabel("clients")
plt.ylabel("duration")
plt.title("clients vs duration")
plt.savefig("graph12.png")
plt.show()
```

pdays

Name: count, Length: 559, dtype: int64



## Q13 How many contacts were performed before the current campaign for each client?

```
contact=df["previous"].value_counts()
print(contact)
plt.hist(contact,bins=10,color="darkgreen",edgecolor="green")
plt.xlabel("clients")
plt.ylabel("previous")
plt.title("previous vs clients")
plt.savefig("graph13.png")
plt.show()
previous
0 36956
1 2772
    2106
    1142
4 715
5 459
     278
7 205
     130
     92
10 67
```

**11** 65

<mark>12 44</mark>

**13** 38

**15 20** 

<mark>14 19</mark>

**17 15** 

<mark>16 13</mark>

<mark>19 11</mark>

20 8

23 8

<del>18</del> 6

<mark>22 6</mark>

<mark>24 5</mark>

**27 5** 

21 4

<mark>29 4</mark>

<mark>25 4</mark>

30 3

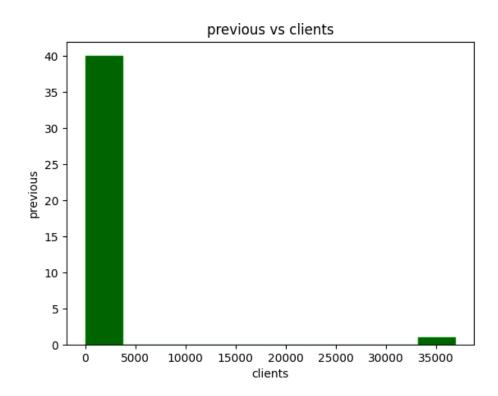
38 2

37 2

<mark>26 2</mark>

28 2

51 1



#### Q14 What were the outcomes of the previous marketing campaigns?

```
outcome=df["poutcome"].value_counts()
print(outcome)
plt.hist(outcome,bins=10,color="yellow",edgecolor="green")
plt.xlabel("clients")
plt.ylabel("outcome")
plt.title("outcome vs clients")
plt.savefig("graph14.png")
plt.show()
```

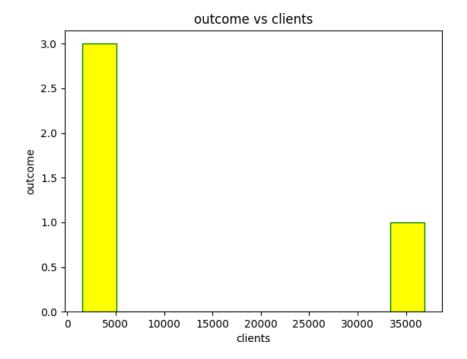
#### **poutcome**

unknown 36961

failure 4902

other 1840

success 1513



## Q15 What is the distribution of clients who subscribed to a term deposit vs. those who did not?

```
term_deposit=df["y"].value_counts()

print(term_deposit)

plt.hist(term_deposit,bins=10,color="blue",edgecolor="green")

plt.xlabel("clients")

plt.ylabel("term deposit")

plt.title("term deposit vs clients")

plt.savefig("graph15.png")

plt.show()
```

no 39922

yes 5294

