```
In [1]: #!pip install plotly
```

In [2]: #import libraries
 import pandas as pd
 import plotly.express as px
 import matplotlib.pyplot as plt

### In [3]: #Load dataset

### Out[4]:

	transaction_id	transaction_amount	location	merchant	age	gender	fraud_label
0	1	1000.0	New York	ABC Corp	35	М	0
1	2	500.0	Chicago	XYZ Inc	45	F	0
2	3	2000.0	Los Angeles	ABC Corp	28	М	1
3	4	1500.0	San Francisco	XYZ Inc	30	F	0
4	5	800.0	Chicago	ABC Corp	50	F	0
81	82	1500.0	Los Angeles	XYZ Inc	31	М	0
82	83	2800.0	San Francisco	ABC Corp	50	F	1
83	84	1350.0	Chicago	XYZ Inc	28	М	0
84	85	920.0	New York	ABC Corp	47	F	0
85	86	2000.0	Los Angeles	XYZ Inc	36	М	0

86 rows × 7 columns

```
In [5]: #First we check null values
         dataset.isnull().sum()
Out[5]: transaction_id
         transaction_amount
                               0
         location
         merchant
                               0
         age
         gender
         fraud_label
         dtype: int64
In [6]: dataset.shape
Out[6]: (86, 7)
In [7]: dataset['gender'].value_counts()['M']
Out[7]: 43
In [8]: dataset['gender'].value_counts()['F']
Out[8]: 43
In [9]: dataset['location'].value_counts()
Out[9]: Chicago
                          22
         New York
                          22
         San Francisco
                          21
         Los Angeles
                          21
         Name: location, dtype: int64
In [10]: dataset['merchant'].value_counts()
Out[10]: XYZ Inc
                     43
         ABC Corp
                     43
         Name: merchant, dtype: int64
```

```
In [11]: df= dataset['gender'] == 'M']
df
```

### Out[11]:

	transaction_id	transaction_amount	location	merchant	age	gender	fraud_label
0	1	1000.0	New York	ABC Corp	35	М	0
2	3	2000.0	Los Angeles	ABC Corp	28	М	1
5	6	3000.0	New York	XYZ Inc	42	М	1
7	8	900.0	Los Angeles	XYZ Inc	37	М	0
9	10	1800.0	New York	XYZ Inc	48	М	0
11	12	2200.0	Chicago	XYZ Inc	51	М	0
13	14	1600.0	Los Angeles	XYZ Inc	26	М	0
15	16	1200.0	Chicago	XYZ Inc	34	М	0
17	18	1900.0	Los Angeles	XYZ Inc	32	М	0
19	20	4000.0	Chicago	XYZ Inc	38	М	1
21	22	1700.0	Los Angeles	XYZ Inc	49	М	0
23	24	2300.0	Chicago	XYZ Inc	27	М	1
25	26	1400.0	Los Angeles	XYZ Inc	54	М	0
27	28	1100.0	Chicago	XYZ Inc	44	М	0
29	30	2000.0	Los Angeles	XYZ Inc	46	М	0
31	32	2100.0	Chicago	XYZ Inc	43	М	0
33	34	1800.0	Los Angeles	XYZ Inc	29	М	0
35	36	1300.0	Chicago	XYZ Inc	37	М	0
37	38	2000.0	Los Angeles	XYZ Inc	33	М	0
39	40	2400.0	Chicago	XYZ Inc	26	М	0
41	42	1500.0	Los Angeles	XYZ Inc	31	М	0
43	44	1350.0	Chicago	XYZ Inc	28	М	0
45	46	2000.0	Los Angeles	XYZ Inc	36	М	0
47	48	1900.0	Chicago	XYZ Inc	38	М	1
49	50	1750.0	Los Angeles	XYZ Inc	49	М	0
51	52	2300.0	Chicago	XYZ Inc	41	М	0

	transaction_id	transaction_amount	location	merchant	age	gender	fraud_label
53	54	1600.0	Los Angeles	XYZ Inc	39	М	0
55	56	1250.0	Chicago	XYZ Inc	35	М	0
57	58	2200.0	Los Angeles	XYZ Inc	29	М	0
59	60	4000.0	Chicago	XYZ Inc	37	М	1
61	62	1700.0	Los Angeles	XYZ Inc	49	М	0
63	64	2800.0	Chicago	XYZ Inc	27	М	1
65	66	1400.0	Los Angeles	XYZ Inc	54	М	0
67	68	1100.0	Chicago	XYZ Inc	44	М	0
69	70	2000.0	Los Angeles	XYZ Inc	46	М	0
71	72	2100.0	Chicago	XYZ Inc	43	М	0
73	74	1800.0	Los Angeles	XYZ Inc	29	М	0
75	76	1300.0	Chicago	XYZ Inc	37	М	0
77	78	2000.0	Los Angeles	XYZ Inc	33	М	0
79	80	2400.0	Chicago	XYZ Inc	26	М	0
81	82	1500.0	Los Angeles	XYZ Inc	31	М	0
83	84	1350.0	Chicago	XYZ Inc	28	М	0
85	86	2000.0	Los Angeles	XYZ Inc	36	М	0

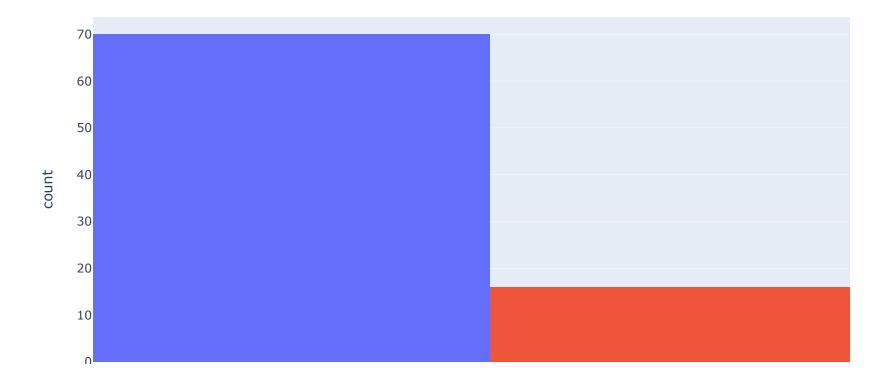
```
In [12]: dataset.columns
Out[12]: Index(['transaction_id', 'transaction_amount', 'location', 'merchant', 'age',
```

# Fraud and Non Fraud Customer details

'gender', 'fraud\_label'],

dtype='object')

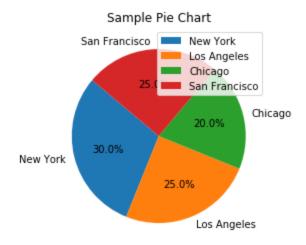
### Fraud and Non Fraud Customer



# Percentage of Customer loacationWise

```
In [14]: labels = ['New York', 'Los Angeles', 'Chicago', 'San Francisco']
    sizes = [30, 25, 20, 25]

plt.pie(sizes, labels=labels, autopct='%1.1f%%', startangle=140)
    plt.title('Sample Pie Chart')
    plt.legend(labels, loc='upper right')
    plt.show()
```



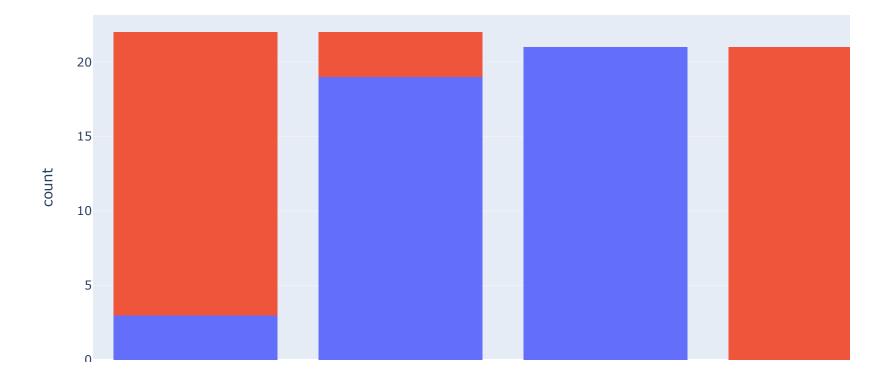
# What kind of relation between Transacation amount and Age

In [15]: dataset.corr()
Out[15]:

	transaction_id	transaction_amount	age	fraud_label
transaction_id	1.000000	0.048031	0.009045	-0.074628
transaction_amount	0.048031	1.000000	-0.096590	0.771904
age	0.009045	-0.096590	1.000000	-0.021454
fraud_label	-0.074628	0.771904	-0.021454	1.000000

#Relationship between transaction amount and age is Negative correlation but its a poor Neagtive value -0.096590

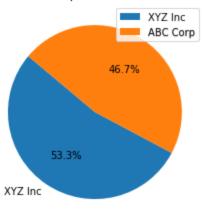
### No of customer in location wise



For Los Angeles location we only have a male candidate(21) & san fransisco location we only have a female candidates (21)

## **Bank Details**

### Sample Pie Chart



## locationwise transaction\_amount

