import pandas as pd
import numpy as np

import matplotlib.pyplot as plt
from sklearn.cluster import KMeans

dataset = pd.read\_csv('Daily Household Transactions.csv')

### dataset.head()

<del>_</del>		Date	Mode	Category	Subcategory	Note	Amount	Income/Expense	Currency
	0	20/09/2018 12:04:08	Cash	Transportation	Train	2 Place 5 to Place 0	30.0	Expense	INR
	1	20/09/2018 12:03:15	Cash	Food	snacks	ldli medu Vada mix 2 plates	60.0	Expense	INR
	2	19/09/2018	Saving Bank account 1	subscription	Netflix	1 month subscription	199.0	Expense	INR
	3	17/09/2018 23:41:17	Saving Bank account 1	subscription	Mobile Service Provider	Data booster pack	19.0	Expense	INR
	<b>1</b>	16/00/2019 17·15·09	Cach	Factivale	Canach Duian	Ganash idal	251 0	Evnanca	INID

#### dataset.tail()

	Date	Mode	Category	Subcategory	Note	Amount	Income/Expense	Currency
594	23/01/2018 21:28:59	Credit Card	Household	Kirana	Smart point	273.0	Expense	INR
595	22/01/2018 19:05:26	Saving Bank account 1	Food	flour mill	2 kg Bajari	14.0	Expense	INR
596	22/01/2018 12:08:08	Cash	Food	flour mill	M D sure 3kg atta	162.0	Expense	INR
597	22/01/2018 10:25:23	Saving Bank account 1	Health	Medicine	Cough-sills 4 pcs + sinarest 3 pcs	30.0	Expense	INR
£00	22/04/2018 10:25:44	Savina Rank account 1	Ecod	Milk	1 li+	60 N	Evnonco	IND

### dataset.describe()

<del></del> *		Amount
_		
	count	599.000000
	mean	2026.219983
	std	8478.642674
	min	2.000000
	25%	36.000000
	50%	77.000000
	75%	489.680000
	mav	70255 000000
	1	

## dataset.info()

<class 'pandas.core.frame.DataFrame'>
 RangeIndex: 599 entries, 0 to 598
 Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	Date	599 non-null	object
1	Mode	599 non-null	object
2	Category	599 non-null	object
3	Subcategory	489 non-null	object
4	Note	495 non-null	object
5	Amount	599 non-null	float64
6	Income/Expense	599 non-null	object
7	Currency	599 non-null	object

dtypes: float64(1), object(7)
memory usage: 37.6+ KB

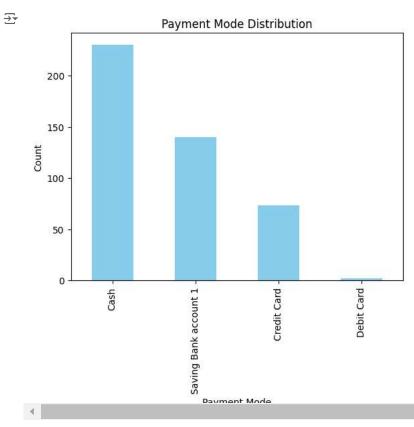
dataset.isnull().sum()

```
<del>_</del>
                          0
            Date
                          0
           Mode
                          0
          Category
                          0
        Subcategory
                        110
            Note
                        104
          Amount
                          0
      Income/Expense
                          0
          Currency
                          0
dataset.count()
\overrightarrow{\Rightarrow}
                          0
            Date
                       599
           Mode
                       599
          Category
                       599
        Subcategory
                       489
            Note
                       495
          Amount
                       599
      Income/Expense
                       599
          Currency
                       599
dataset.shape
 → (599, 8)
dataset.dropna(inplace=True)
dataset.count()
 ₹
                          0
            Date
                       445
           Mode
                       445
          Category
                       445
        Subcategory
                       445
            Note
                       445
          Amount
                       445
      Income/Expense
                       445
          Currency
                       445
dataset.isnull().sum()
```

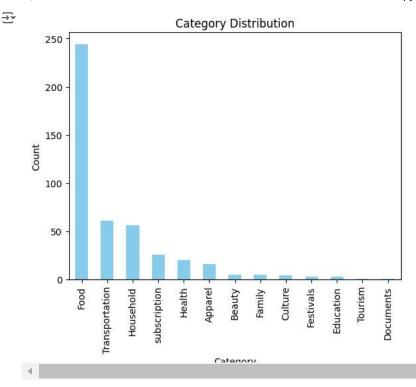
.....

```
<del>_</del>
                        0
            Date
                        0
           Mode
                        0
         Category
       Subcategory
                        0
            Note
                        0
          Amount
                        0
     Income/Expense
                       0
         Currency
                        0
```

```
payment_mode_counts = dataset['Mode'].value_counts()
payment_mode_counts.plot(kind='bar', color='skyblue')
plt.title('Payment Mode Distribution')
plt.xlabel('Payment Mode')
plt.ylabel('Count')
plt.show()
```



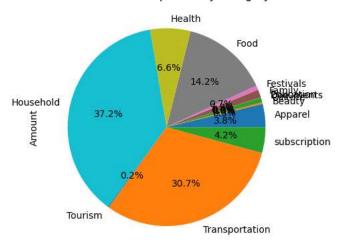
```
category_counts = dataset['Category'].value_counts()
category_counts.plot(kind='bar', color='skyblue')
plt.title('Category Distribution')
plt.xlabel('Category')
plt.ylabel('Count')
plt.show()
```



dataset.groupby('Category')['Amount'].sum().plot(kind='pie', autopct= '%1.1f%%')
plt.title('Distribution of Expenses by Category')

→ Text(0.5, 1.0, 'Distribution of Expenses by Category')

## Distribution of Expenses by Category



```
category_sums = dataset.groupby('Category')['Amount'].sum()

# Calculate the percentage of each category
category_percentages = category_sums / category_sums.sum() * 100

# Create a new category for values below 0.5%
other_categories = category_percentages[category_percentages < 2].sum()

# Filter categories above 0.5%
main_categories = category_percentages[category_percentages >= 2]

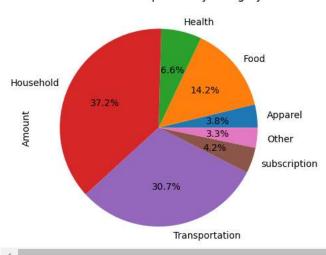
# Add the "Other" category
main_categories['Other'] = other_categories

# Plot the pie chart
main_categories.plot(kind='pie', autopct='%1.1f%%')
plt.title('Distribution of Expenses by Category')
```

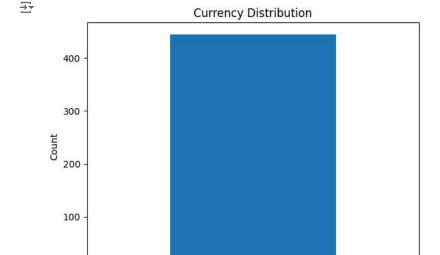
plt.show()

# $\overline{\Rightarrow}$

## Distribution of Expenses by Category



```
curr_count = dataset['Currency'].value_counts()
curr_count.plot(kind='bar')
plt.title('Currency Distribution')
plt.xlabel('Currency')
plt.ylabel('Count')
plt.show()
```



```
selected_columns = ['Amount', 'Category', 'Subcategory']
selected_data = dataset[selected_columns]

data_encoded = pd.get_dummies(selected_data, columns=['Category', 'Subcategory'])

from sklearn.preprocessing import StandardScaler
scalar = StandardScaler()
data_Scaled = scalar.fit_transform(data_encoded)

data_Scaled_df = pd.DataFrame(data_Scaled, columns=data_encoded.columns)
data_Scaled_df.head()
```

Currency



```
Amount Category_Apparel Category_Beauty Category_Culture Category_Documents Category_Education Category_Family Category_Fest
 0 -0.154891
                       -0.193122
                                           -0.1066
                                                            -0.095238
                                                                                 -0.047458
                                                                                                      -0.082385
                                                                                                                          -0.1066
                                                                                                                                             -0.0
 1 -0.142333
                       -0.193122
                                                            -0.095238
                                                                                 -0.047458
                                                                                                      -0.082385
                                           -0.1066
                                                                                                                          -0.1066
                                                                                                                                             -0.0
 2 -0.084149
                       -0.193122
                                           -0.1066
                                                            -0.095238
                                                                                 -0.047458
                                                                                                      -0.082385
                                                                                                                          -0.1066
                                                                                                                                             -0.0
 3 -0.159495
                       -0.193122
                                                            -0.095238
                                                                                 -0.047458
                                                                                                      -0.082385
                                                                                                                                              -0.0
                                           -0.1066
                                                                                                                          -0.1066
 4 -0.062383
                       -0.193122
                                           -0.1066
                                                            -0.095238
                                                                                 -0.047458
                                                                                                      -0.082385
                                                                                                                          -0.1066
                                                                                                                                             12.1
5 rows × 75 columns
```

```
inertia = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i,random_state= 0)
    kmeans.fit(data_Scaled_df)
    inertia.append(kmeans.inertia_)

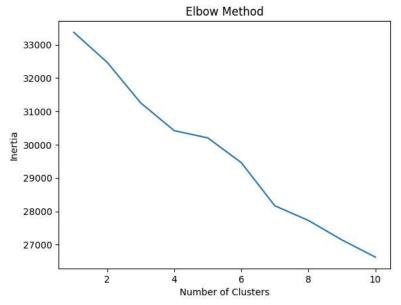
plt.plot(range(1, 11), inertia)
plt.title('Elbow Method')
plt.xlabel('Number of Clusters')
plt.ylabel('Inertia')
plt.show
```

```
🚁 /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
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    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
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    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
      super()._check_params_vs_input(X, default_n_init=10)
     matplotlib.pyplot.show
```

```
matplotlib.pyplot.show
def show(*args, **kwargs)

Display all open figures.

Parameters
------
block : bool, optional
Whether to wait for all figures to be closed before returning.
```



```
optimal_clusters = 8
kmeans = KMeans(n_clusters=optimal_clusters, random_state=0)
kmeans.fit(data_Scaled_df)
dataset['Cluster'] = kmeans.labels
print(dataset.head())
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change from 1
       super()._check_params_vs_input(X, default_n_init=10)
                       Date
                                              Mode
                                                          Category \
      20/09/2018 12:04:08
                                              Cash
                                                    Transportation
       20/09/2018 12:03:15
                                              Cash
    1
                                                              Food
                19/09/2018
                             Saving Bank account 1
                                                      subscription
       17/09/2018 23:41:17
                                                      subscription
    3
                             Saving Bank account 1
       16/09/2018 17:15:08
                                                         Festivals
                                              Cash
                    Subcategory
                                                        Note Amount
    0
                          Train
                                        2 Place 5 to Place 0
                                                                30.0
                         snacks
                                 Idli medu Vada mix 2 plates
                                                                60.0
    2
                        Netflix
                                        1 month subscription
                                                               199.0
    3
       Mobile Service Provider
                                           Data booster pack
                                                                19 A
                   Ganesh Pujan
                                                 Ganesh idol
                                                               251.0
```

```
Income/Expense Currency Cluster
Expense INR 1
                 INR
0
1
        Expense
                   INR
2
        Expense
                 INR
                   INR
                              1
3
        Expense
                              6
                 INR
        Expense
4
                              1
```

```
for cluster in range(optimal_clusters):
    print(f"Cluster {cluster}")
    respect (dataset[cluster]] == cluster[ head())
```