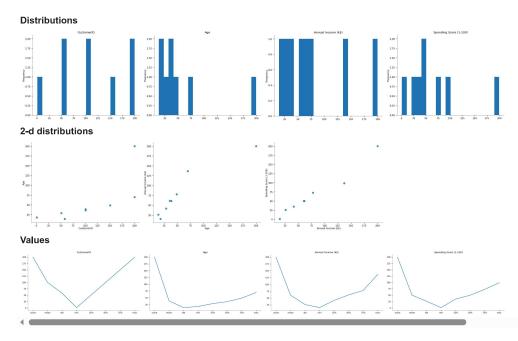
```
import pandas as pd
import seaborn as sns
dt = pd.read_csv('/content/Mall_Customers.csv')
dt.info()
<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 200 entries, 0 to 199
     Data columns (total 5 columns):
                                 Non-Null Count Dtype
     # Column
    ---
     0
         CustomerID
                                 200 non-null
                                 200 non-null
         Gender
                                                 object
     1
     2
         Age
                                 200 non-null
                                                 int64
         Annual Income (k$)
                                 200 non-null
                                                 int64
        Spending Score (1-100) 200 non-null
                                                 int64
     dtypes: int64(4), object(1)
     memory usage: 7.9+ KB
x = dt["Annual Income (k$)"]
y = dt["Spending Score (1-100)"]
```

dt.describe()

→

| | CustomerID | Age | Annual Income (k\$) | Spending Score (1-100) |
|-------|------------|------------|---------------------|------------------------|
| count | 200.000000 | 200.000000 | 200.000000 | 200.000000 |
| mean | 100.500000 | 38.850000 | 60.560000 | 50.200000 |
| std | 57.879185 | 13.969007 | 26.264721 | 25.823522 |
| min | 1.000000 | 18.000000 | 15.000000 | 1.000000 |
| 25% | 50.750000 | 28.750000 | 41.500000 | 34.750000 |
| 50% | 100.500000 | 36.000000 | 61.500000 | 50.000000 |
| 75% | 150.250000 | 49.000000 | 78.000000 | 73.000000 |
| max | 200.000000 | 70.000000 | 137.000000 | 99.000000 |



dt.dtypes



Annual Income (k\$)

Age int64

int64

Spending Score (1-100) int64

dtvne: object

dt['Gender'].value_counts()



count

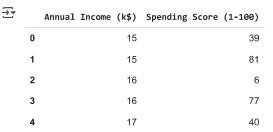
Gender

Female 112

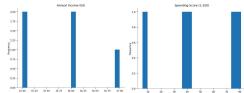
Male 88

dtvne: int64

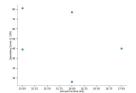
x = dt.iloc[:,3:]
x.head()



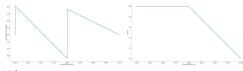
Distributions



2-d distributions



Time series



Values

```
from sklearn.cluster import KMeans
wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state= 42)
    kmeans.fit(x)
```

```
wcss.append(kmeans.inertia_)
```

```
import matplotlib.pyplot as plt
plt.plot(range(1, 11), wcss)
plt.title('Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```




```
kmeans = kmeans = KMeans(n_clusters=5, init='k-means++', random_state = 42)
y_kmeans = kmeans.fit(x)

label = kmeans.labels_
dt['cluster'] = label

plt.scatter(dt['Annual Income (k$)'], dt['Spending Score (1-100)'], c=dt['cluster'])
plt.title('mall customer segmentation')
plt.xlabel('age')
plt.ylabel('spending score')
plt.show
```

```
matplotlib.pyplot.show
def show(*args, **kwargs) -> None

Display all open figures.

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

