

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

In [3]: df=pd.read_csv('Task 3 Mall_Customers.csv')
```

```
In [4]: df.head()
```

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

```
In [5]: df.tail()
```

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
195	196	Female	35	120	79
196	197	Female	45	126	28
197	198	Male	32	126	74
198	199	Male	32	137	18
199	200	Male	30	137	83

```
In [6]: df.rename(columns={'Genre': 'Gender'}, inplace=True)
```

```
In [7]: df.isna().sum()
```

```
Out[7]: CustomerID      0
Gender      0
Age      0
Annual Income (k$)  0
Spending Score (1-100)  0
dtype: int64
```

```
In [8]: df.columns
```

```
Out[8]: Index(['CustomerID', 'Gender', 'Age', 'Annual Income (k$)',
'Spending Score (1-100)'],
dtype='object')
```

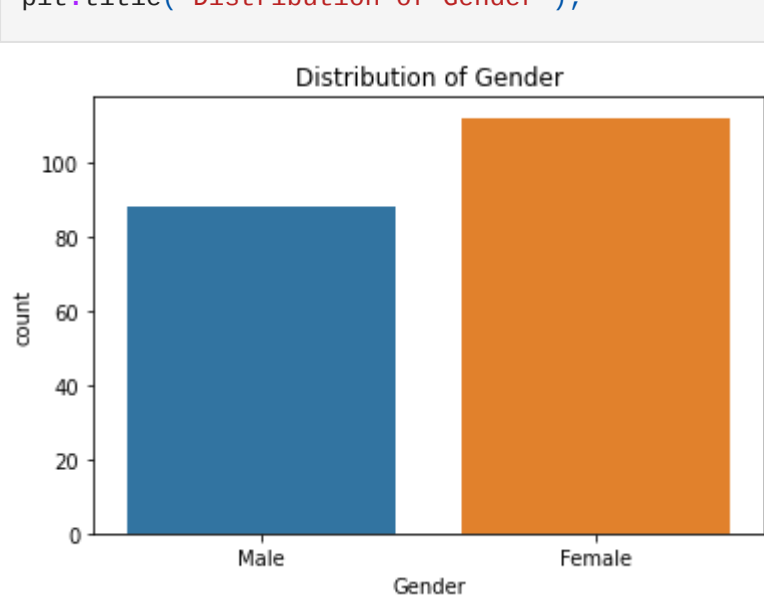
```
In [9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column              Non-Null Count  Dtype
---  --
0   CustomerID          200 non-null   int64
1   Gender              200 non-null   object
2   Age                 200 non-null   int64
3   Annual Income (k$)  200 non-null   int64
4   Spending Score (1-100) 200 non-null   int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB
```

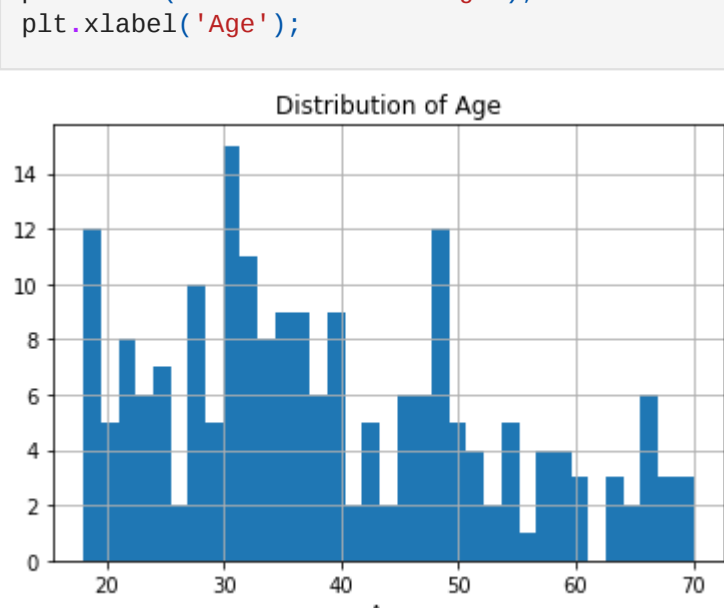
```
In [10]: df.describe()
```

	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

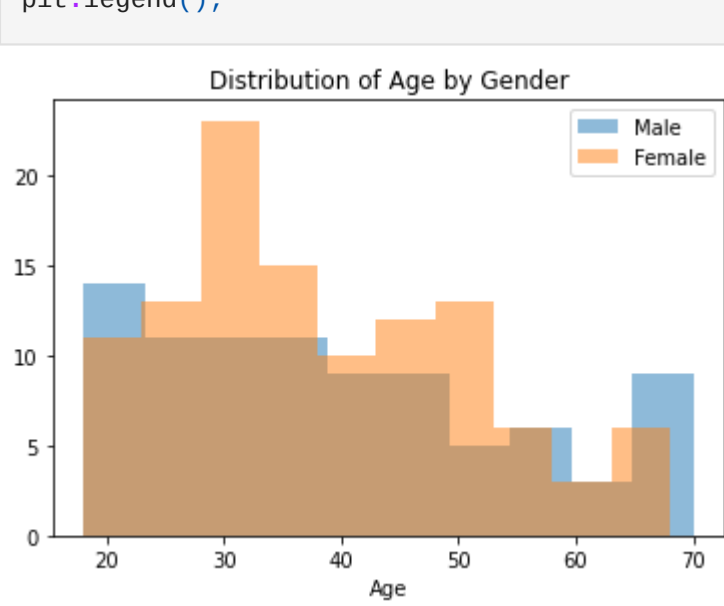
```
In [11]: sns.countplot(x='Gender', data=df);
plt.title('Distribution of Gender');
```



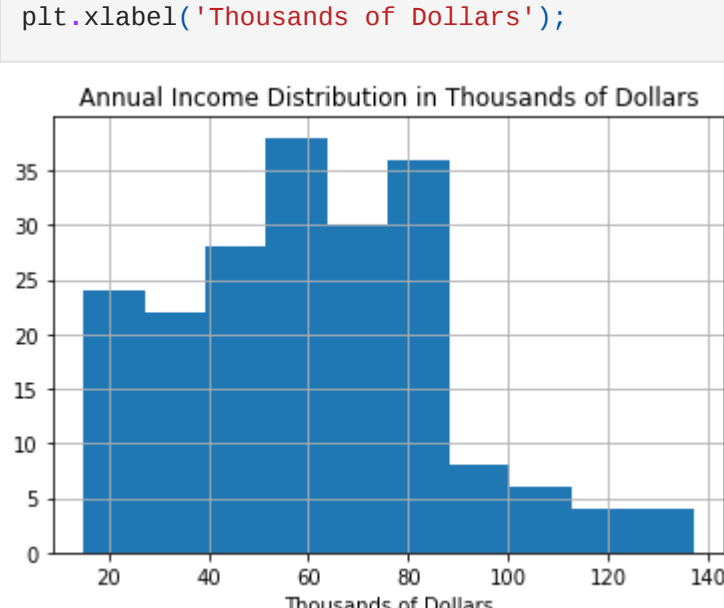
```
In [12]: df.hist('Age', bins=35);
plt.title('Distribution of Age');
plt.xlabel('Age');
```



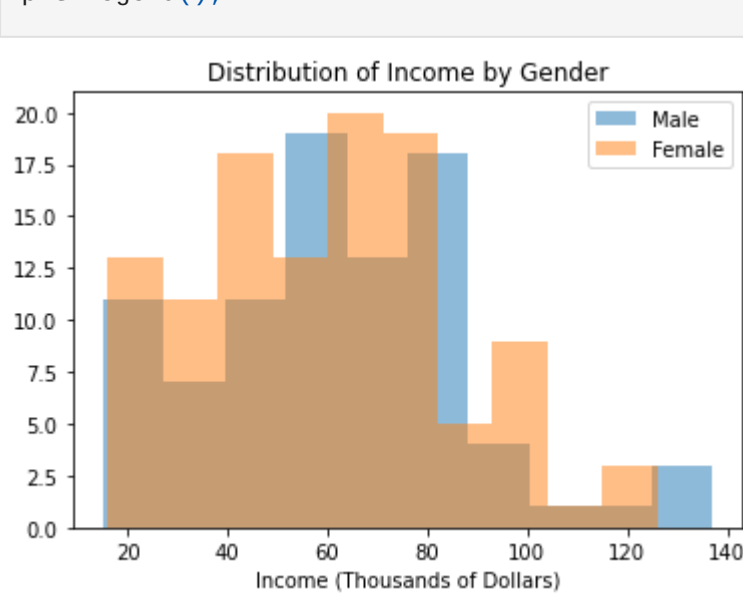
```
In [13]: plt.hist('Age', data=df[df['Gender'] == 'Male'], alpha=0.5, label='Male');
plt.hist('Age', data=df[df['Gender'] == 'Female'], alpha=0.5, label='Female');
plt.title('Distribution of Age by Gender');
plt.xlabel('Age');
plt.legend();
```



```
In [14]: df.hist('Annual Income (k$)');
plt.title('Annual Income Distribution in Thousands of Dollars');
plt.xlabel('Thousands of Dollars');
```



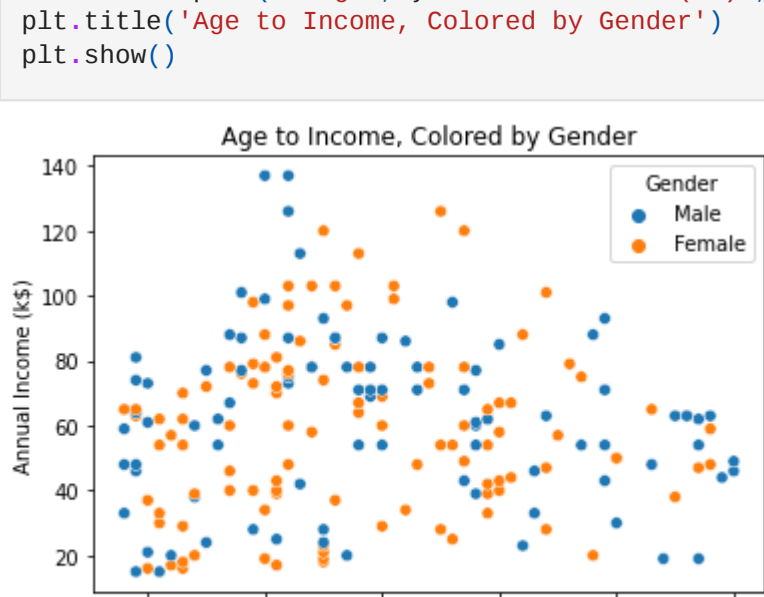
```
In [15]: plt.hist('Annual Income (k$)', data=df[df['Gender'] == 'Male'], alpha=0.5, label='Male');
plt.hist('Annual Income (k$)', data=df[df['Gender'] == 'Female'], alpha=0.5, label='Female');
plt.title('Distribution of Income by Gender');
plt.xlabel('Income (Thousands of Dollars)');
plt.legend();
```



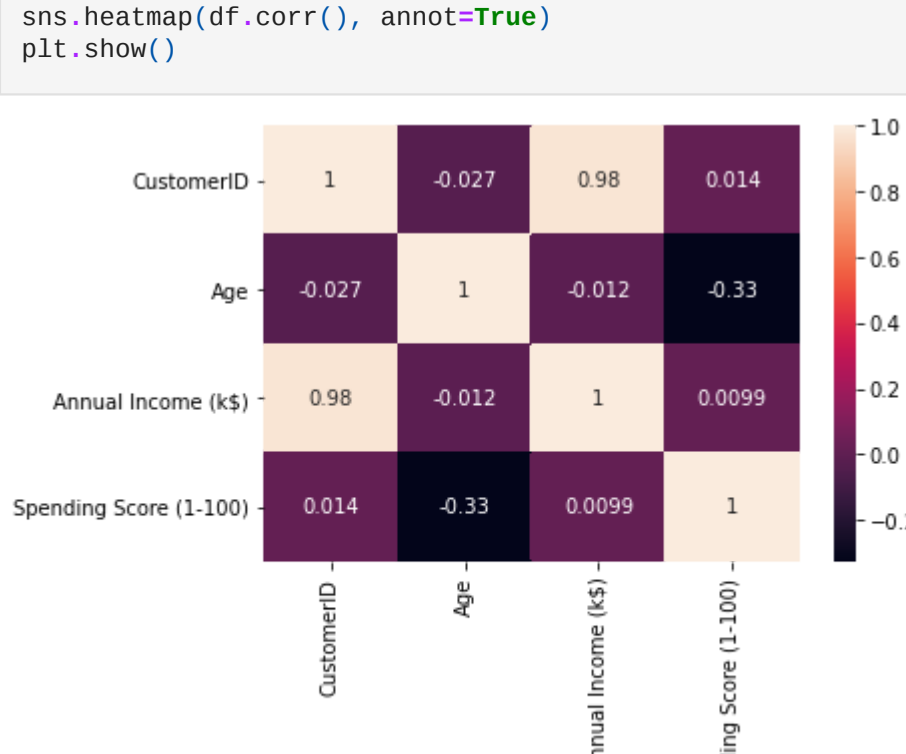
```
In [16]: male_customers = df[df['Gender'] == 'Male']
female_customers = df[df['Gender'] == 'Female']
print(male_customers['Spending Score (1-100)'].mean())
print(female_customers['Spending Score (1-100)'].mean())
```

```
48.51136363636363
51.526785714285715
```

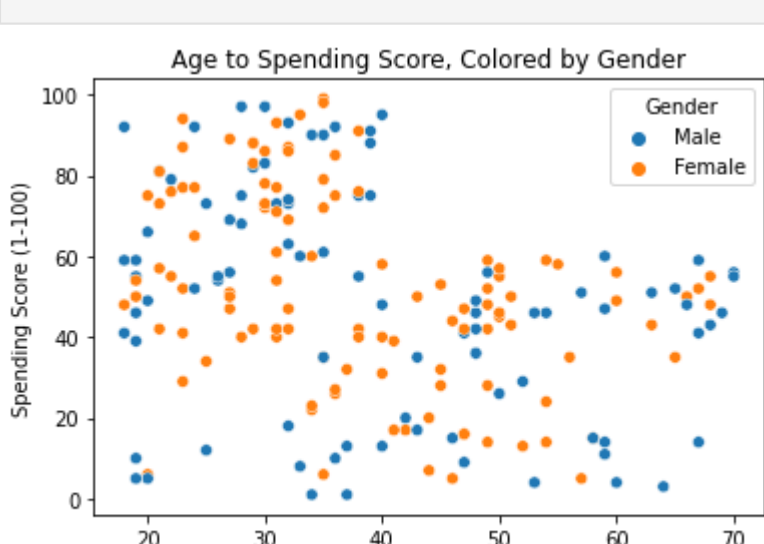
```
In [17]: sns.scatterplot(x='Age', y='Annual Income (k$)', hue='Gender', data=df)
plt.title('Age to Income, Colored by Gender')
plt.show()
```



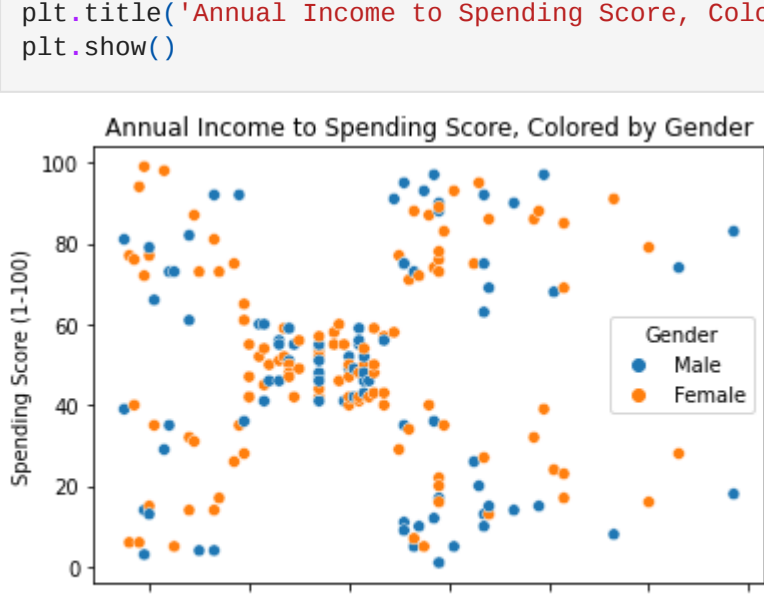
```
In [18]: sns.heatmap(df.corr(), annot=True)
plt.show()
```



```
In [19]: sns.scatterplot(x='Age', y='Spending Score (1-100)', hue='Gender', data=df)
plt.title('Age to Spending Score, Colored by Gender')
plt.show()
```



```
In [20]: sns.scatterplot(x='Annual Income (k$)', y='Spending Score (1-100)', hue='Gender', data=df)
plt.title('Annual Income to Spending Score, Colored by Gender')
plt.show()
```



```
In [ ]:
```