In [1]: import pandas as pd
 import numpy as np
 import seaborn as sns
 import matplotlib.pyplot as plt

In [2]: df = pd.read_csv("Ecommerce Customers.csv")

In [3]: df.head()

Out[3]:

	Email	Address	Avatar	Avg. Session Length	Time on App	Time on Website	Length of Membership	
0	mstephenson@fernandez.com	835 Frank Tunnel\nWrightmouth, MI 82180-9605	Violet	34.497268	12.655651	39.577668	4.082621	Ę
1	hduke@hotmail.com	4547 Archer Common\nDiazchester, CA 06566-8576	DarkGreen	31.926272	11.109461	37.268959	2.664034	3
2	pallen@yahoo.com	24645 Valerie Unions Suite 582\nCobbborough, D	Bisque	33.000915	11.330278	37.110597	4.104543	۷
3	riverarebecca@gmail.com	1414 David Throughway\nPort Jason, OH 22070-1220	SaddleBrown	34.305557	13.717514	36.721283	3.120179	Ę
4	mstephens@davidson- herman.com	14023 Rodriguez Passage∖nPort Jacobville, PR 3	MediumAquaMarine	33.330673	12.795189	37.536653	4.446308	Ę
4								

In [4]: df.tail()

Out[4]:

	Email	Address	Avatar	Avg. Session Length	Time on App	Time on Website	Length of Membership	Ye Amı Sı
495	lewisjessica@craig- evans.com	4483 Jones Motorway Suite 872\nLake Jamiefurt,	Tan	33.237660	13.566160	36.417985	3.746573	573.847
496	katrina56@gmail.com	172 Owen Divide Suite 497\nWest Richard, CA 19320	PaleVioletRed	34.702529	11.695736	37.190268	3.576526	529.049
497	dale88@hotmail.com	0787 Andrews Ranch Apt. 633\nSouth Chadburgh,	Cornsilk	32.646777	11.499409	38.332576	4.958264	551.620
498	cwilson@hotmail.com	680 Jennifer Lodge Apt. 808\nBrendachester, TX	Teal	33.322501	12.391423	36.840086	2.336485	456.469
499	hannahwilson@davidson.com	49791 Rachel Heights Apt. 898\nEast Drewboroug	DarkMagenta	33.715981	12.418808	35.771016	2.735160	497.778

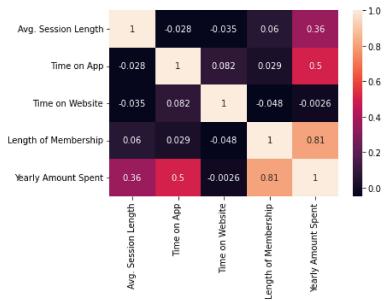
In [5]: df.shape

Out[5]: (500, 8)

```
In [6]: df.dtypes
Out[6]: Email
                                     object
                                     object
          Address
                                     object
          Avatar
          Avg. Session Length
                                    float64
          Time on App
                                    float64
          Time on Website
                                    float64
          Length of Membership
                                    float64
          Yearly Amount Spent
                                    float64
          dtype: object
 In [7]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 500 entries, 0 to 499
          Data columns (total 8 columns):
           #
               Column
                                       Non-Null Count Dtype
                                       -----
                                                        ----
           a
               Email
                                                         object
                                       500 non-null
               Address
                                       500 non-null
           1
                                                         object
           2
               Avatar
                                       500 non-null
                                                         object
                                                         float64
           3
               Avg. Session Length
                                       500 non-null
           4
               Time on App
                                       500 non-null
                                                         float64
               Time on Website
                                       500 non-null
                                                         float64
           6
               Length of Membership 500 non-null
                                                         float64
               Yearly Amount Spent
                                       500 non-null
                                                         float64
          dtypes: float64(5), object(3)
          memory usage: 31.4+ KB
 In [8]: df.describe()
Out[8]:
                 Avg. Session Length Time on App Time on Website Length of Membership Yearly Amount Spent
                         500.000000
                                     500.000000
                                                    500.000000
                                                                        500.000000
                                                                                           500.000000
           count
           mean
                          33.053194
                                      12.052488
                                                     37.060445
                                                                         3.533462
                                                                                          499.314038
                           0.992563
                                       0.994216
                                                      1.010489
                                                                         0.999278
                                                                                           79.314782
             std
                                                                         0.269901
                                                                                          256.670582
            min
                          29.532429
                                      8.508152
                                                     33.913847
            25%
                          32.341822
                                      11.388153
                                                     36.349257
                                                                         2.930450
                                                                                          445.038277
            50%
                          33.082008
                                      11.983231
                                                     37.069367
                                                                         3.533975
                                                                                          498.887875
            75%
                          33.711985
                                      12.753850
                                                     37.716432
                                                                         4.126502
                                                                                           549.313828
                          36.139662
                                      15.126994
                                                     40.005182
                                                                         6.922689
                                                                                          765.518462
            max
In [9]: df.duplicated().sum()
Out[9]: 0
In [10]: | df.isnull().sum()
Out[10]: Email
                                    0
          Address
                                    0
          Avatar
                                    0
          Avg. Session Length
                                    0
          Time on App
                                    0
          Time on Website
                                    0
          Length of Membership
                                    0
          Yearly Amount Spent
          dtype: int64
```

```
In [11]: corr = df.corr()
In [12]: sns.heatmap(corr,annot=True)
```

Out[12]: <AxesSubplot:>



```
In [13]: df.columns
Out[13]: Index(['Email', 'Address', 'Avatar', 'Avg. Session Length', 'Time on App', 'Time on Website', 'Length of Membership', 'Yearly Amount Spent'],
                  dtype='object')
In [14]: def yr_amount_spent(a):
                    if a>200 and a<=300:
                         group = "200-300"
                    elif a>300 and a<=400:
                         group = "300-400"
                    elif a>400 and a<=500:
                         group = "400-500"
                    elif a>500 and a<=600:
                         group = "500-600"
                    elif a>600 and a<=700:
                         group = "600-700"
                    elif a>700 and a<=800:
                         group = "700-800"
                    return (group)
```

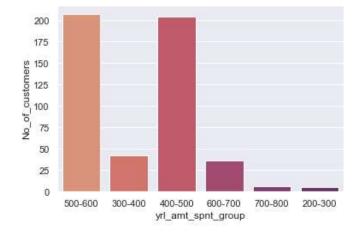
```
In [15]: df["yrl_amt_spnt_group"] = df["Yearly Amount Spent"].apply(yr_amount_spent)
```

```
In [16]: df.head()
```

Out[16]:

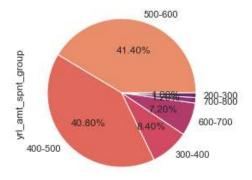
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3	riverarebecca@gmail.com	1414 David Throughway\nPort Jason, OH 22070-1220	SaddleBrown	34,305557	13.717514	36.721283	3.120179	Ę
4	mstephens@davidson- herman.com	14023 Rodriguez Passage\nPort Jacobville, PR 3	MediumAquaMarine	33.330673	12.795189	37.536653	4.446308	Ę

```
In [17]: sns.set_theme(style='darkgrid', palette='flare')
sns.countplot(data = df, x = "yrl_amt_spnt_group")
plt.ylabel("No_of_customers")
plt.show()
```



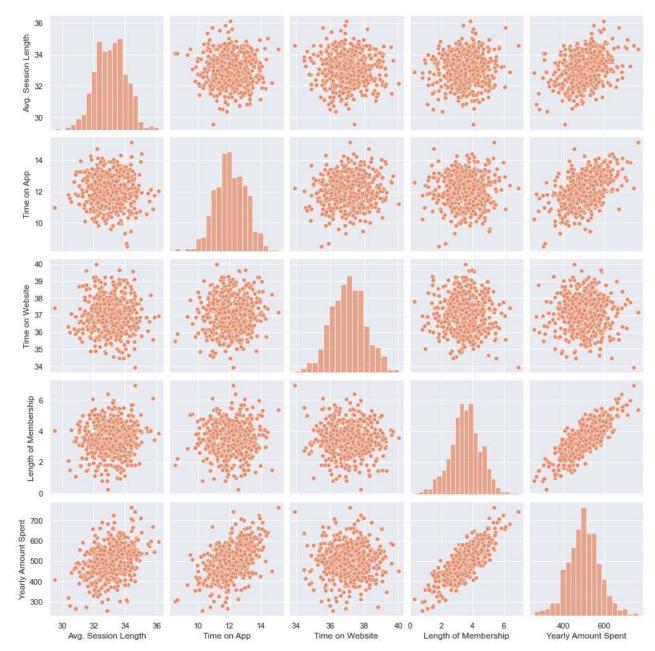
```
In [18]: sns.set_theme(style='darkgrid', palette='flare')
df["yrl_amt_spnt_group"].value_counts().plot(kind="pie",autopct = "%.2f%")
```

Out[18]: <AxesSubplot:ylabel='yrl_amt_spnt_group'>



```
In [19]: sns.pairplot(df)
```

Out[19]: <seaborn.axisgrid.PairGrid at 0x1bacb82c3a0>



In [20]: #As we can see their is some sort of Linearity between yearly amount spent and Length of membersh #Linear regression model to it

```
In [21]: x1 = df.iloc[:,3:7]
    x = df.iloc[:,3:7].values
    y = df.iloc[:,-2].values
```

In [22]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.3,random_state = (0))

In [23]: from sklearn.linear_model import LinearRegression
lm = LinearRegression()

```
In [24]: |lm.fit(x_train,y_train)
Out[24]: LinearRegression()
In [25]: pred = lm.predict(x test)
In [26]: pred[0:5]
Out[26]: array([438.05361824, 489.88569198, 370.69103491, 514.760391 ,
                496.7189217 ])
In [27]: y_test[0:5]
Out[27]: array([449.07031944, 482.60246733, 374.26967454, 513.15311185,
                502.77107457])
In [28]: import sklearn.metrics as metric
         metric.mean_absolute_error(y_test,pred)
Out[28]: 7.85137717086146
In [29]: | mse = metric.mean squared error(y test,pred)
         mse
Out[29]: 94.55779479273302
In [30]: np.sqrt(mse)
Out[30]: 9.724083236620974
In [32]: c = pd.DataFrame(lm.coef_,index=x1.columns)
         C
Out[32]:
                                    0
           Avg. Session Length 25.767530
                  Time on App 38.800394
              Time on Website -0.018041
          Length of Membership 61.852568
In [ ]: # As we can see that Time on app is more so recommended to invest more on app rather then Website
```