

```
In [1]: print("hello world")
```

```
hello world
```

```
In [2]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [3]: df = pd.read_csv('Ecommerce Customers.csv')
```

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

```
Out[5]:
```

	Email	Address	Avatar	Avg Session Length
0	mstephenson@fernandez.com	835 Frank Tunnell\nWrightmouth, MI 82180-9605	Violet	34.497266
1	hduke@hotmail.com	4547 Archer Common\nDiazchester, CA 06566-8576	DarkGreen	31.926272
2	pallen@yahoo.com	24645 Valerie Unions Suite 582\nCobbborough, D...	Bisque	33.000915
3	riverarebecca@gmail.com	1414 David Throughway\nPort Jason, OH 22070-1220	SaddleBrown	34.305557
4	mstephens@davidson-herman.com	14023 Rodriguez Passage\nPort Jacobville, PR 3...	MediumAquaMarine	33.330673

```
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
---  -
0   Email                                500 non-null    object
1   Address                              500 non-null    object
2   Avatar                               500 non-null    object
3   Avg. Session Length                  500 non-null    float64
4   Time on App                           500 non-null    float64
5   Time on Website                       500 non-null    float64
6   Length of Membership                  500 non-null    float64
7   Yearly Amount Spent                   500 non-null    float64
dtypes: float64(5), object(3)
memory usage: 31.4+ KB
```

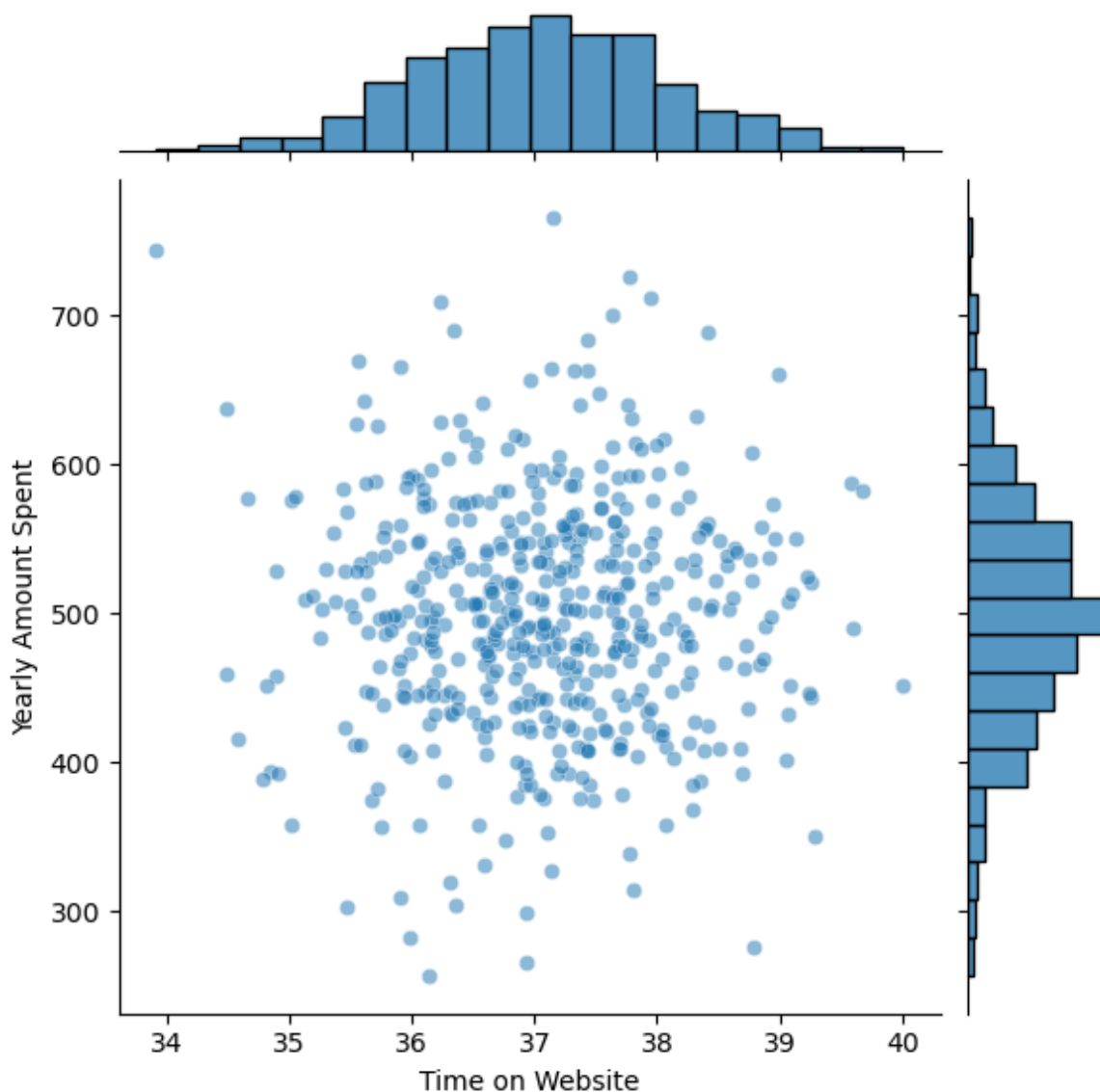
```
In [8]: df.describe()
```

Out [8]:

	Avg. Session Length	Time on App	Time on Website	Length of Membership	Yearly Amount Spent
count	500.000000	500.000000	500.000000	500.000000	500.000000
mean	33.053194	12.052488	37.060445	3.533462	499.314038
std	0.992563	0.994216	1.010489	0.999278	79.314782
min	29.532429	8.508152	33.913847	0.269901	256.670582
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
max	36.139662	15.126994	40.005182	6.922689	765.518462

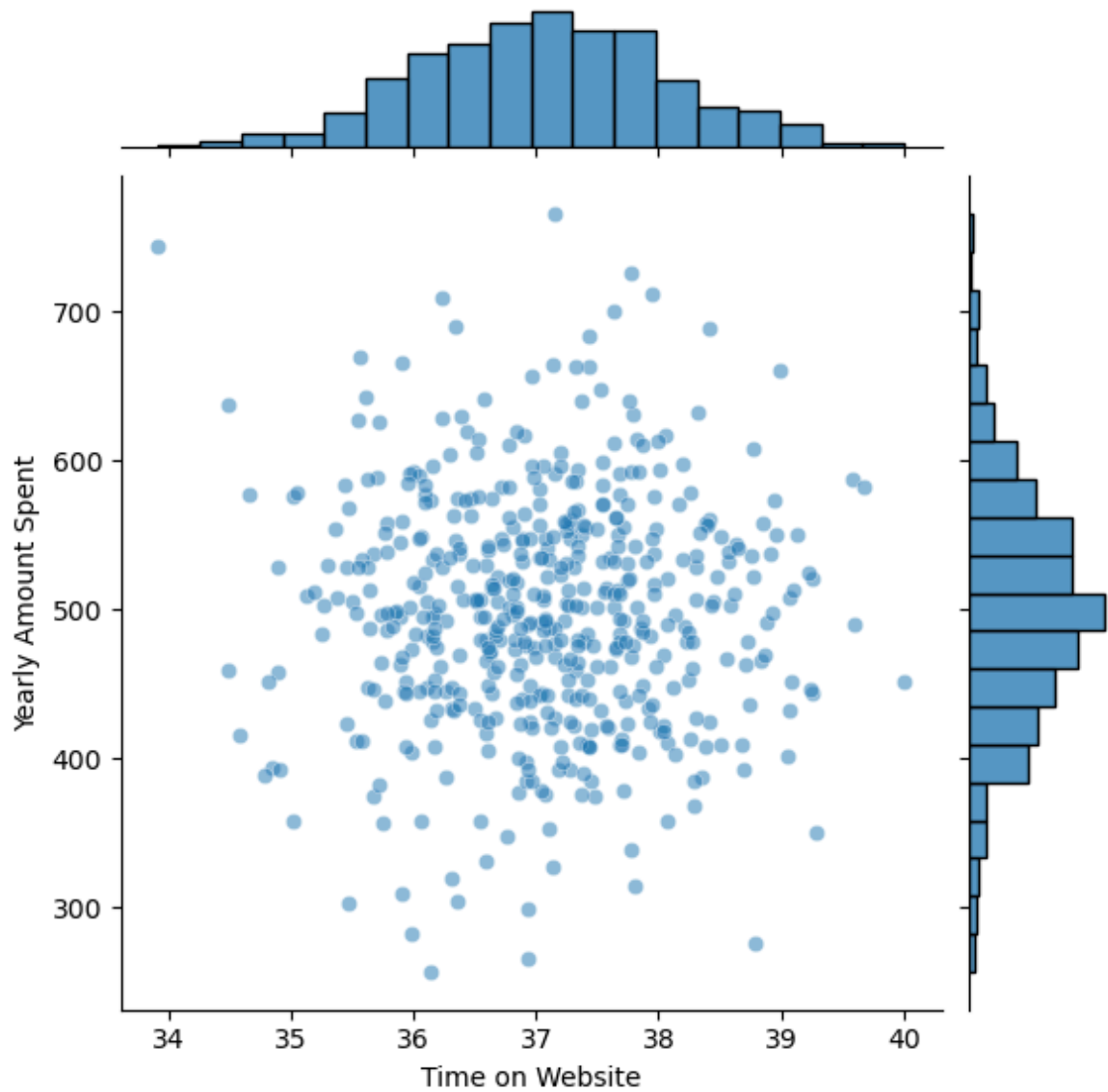
```
In [10]: sns.jointplot(x = "Time on Website", y = "Yearly Amount Spent", data = df)
```

```
Out[10]: <seaborn.axisgrid.JointGrid at 0x15cf77230>
```



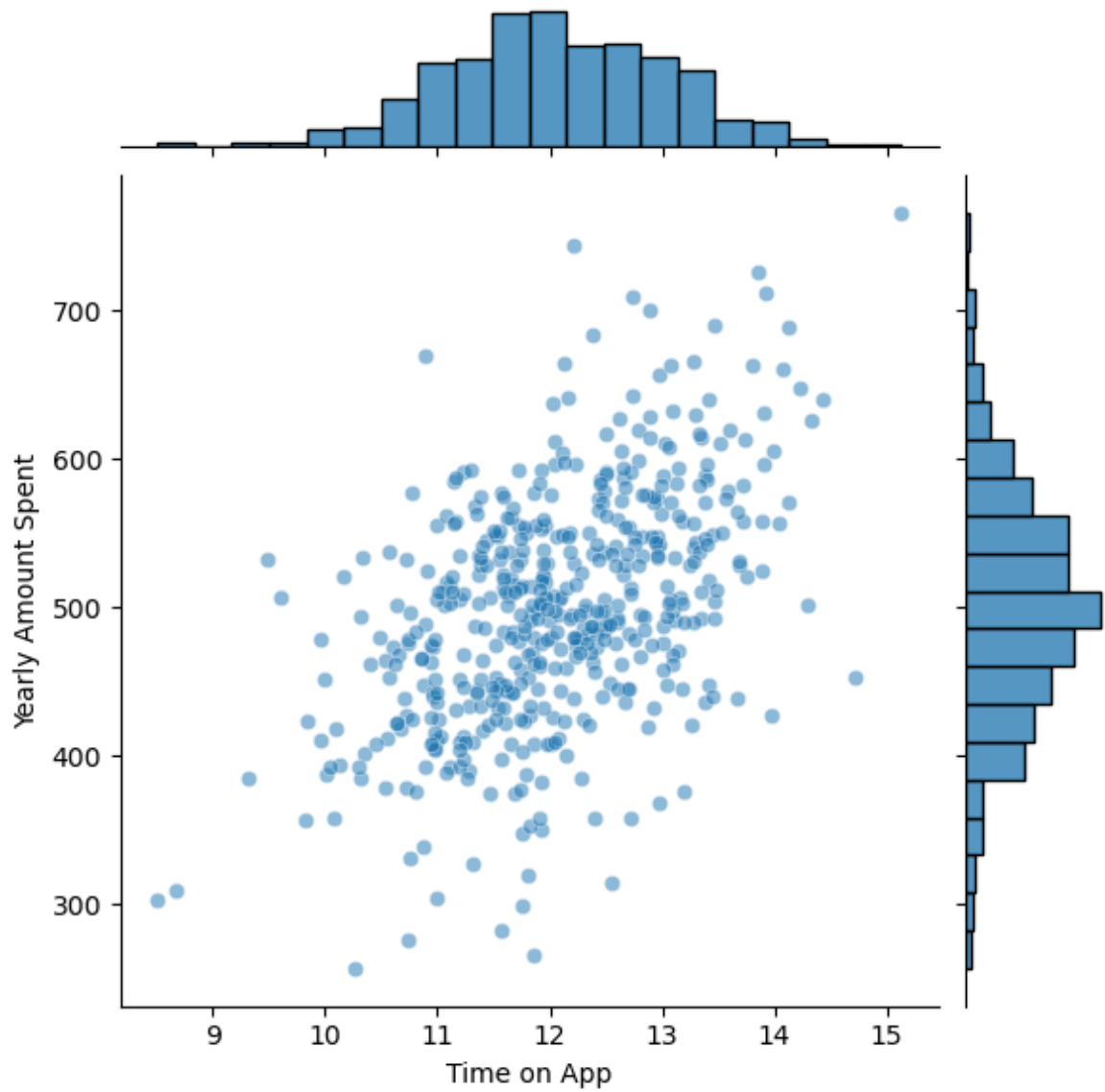
```
In [10]: sns.jointplot(x = "Time on Website", y = "Yearly Amount Spent", data = df)
```

```
Out[10]: <seaborn.axisgrid.JointGrid at 0x15cf77230>
```



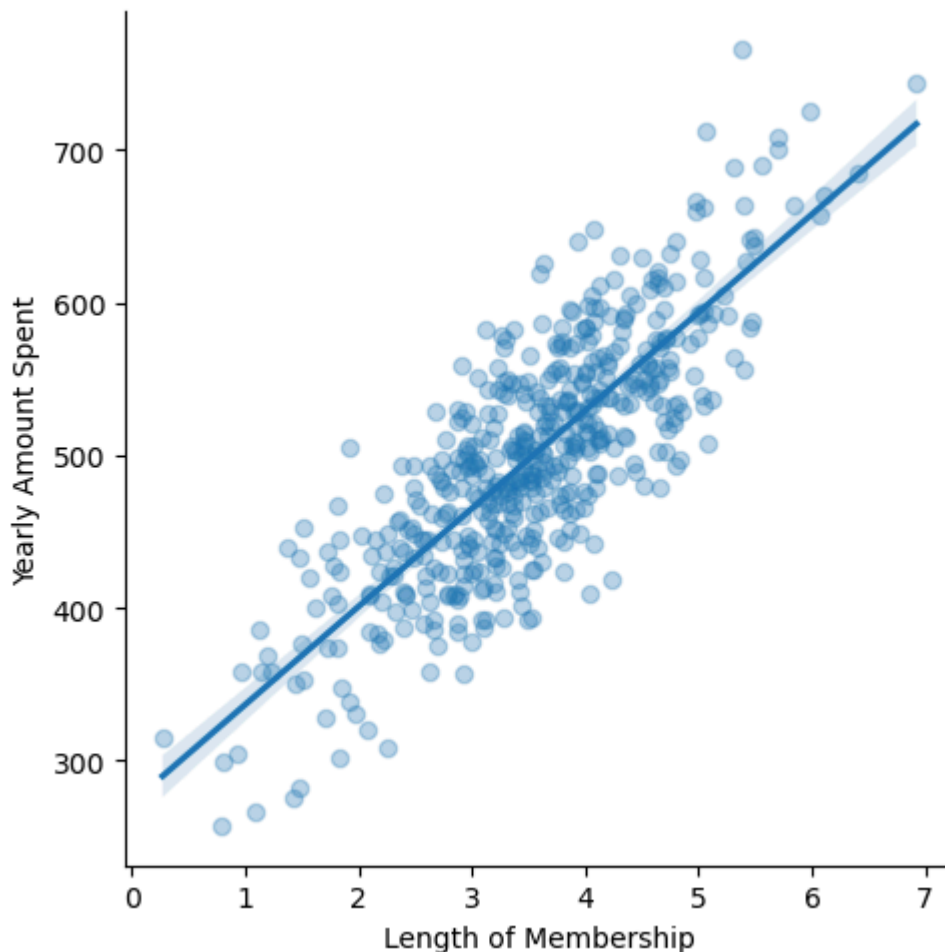
```
In [11]: sns.jointplot(x = "Time on App", y = "Yearly Amount Spent", data = df, al
```

```
Out[11]: <seaborn.axisgrid.JointGrid at 0x160ac7110>
```



```
In [15]: sns.lmplot(x = 'Length of Membership',  
                  y = 'Yearly Amount Spent',  
                  data = df,  
                  scatter_kws = {'alpha':0.3})
```

```
Out[15]: <seaborn.axisgrid.FacetGrid at 0x160c706b0>
```



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In [16]: from sklearn.model_selection import train_test_split
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```
In [17]: x = df[['Time on Website', 'Time on App', 'Length of Membership', 'Avg. Se  
y = df['Yearly Amount Spent']
```

```
In [18]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, r
```

```
In [22]: from sklearn.linear_model import LinearRegression
```

```
In [23]: lm = LinearRegression()
```

```
In [24]: lm.fit(x_train, y_train)
```

```
Out[24]: ▼ LinearRegression ⓘ ?
```

```
LinearRegression()
```

```
In [25]: lm.coef_
```

```
Out[25]: array([ 0.45914788, 38.59713548, 61.67473243, 25.72425621])
```

```
In [29]: cdf = pd.DataFrame(lm.coef_, x.columns, columns=['Coef'])  
print(cdf)
```

	Coef
Time on Website	0.459148
Time on App	38.597135
Length of Membership	61.674732
Avg. Session Length	25.724256

```
In [30]: predictions = lm.predict(x_test)
```

```
In [31]: predictions
```

```
Out[31]: array([403.66993069, 542.57756289, 427.06591658, 502.02460425,
 410.12143559, 569.93442508, 531.93431341, 506.29650969,
 408.71870658, 473.97737105, 441.46912726, 425.33703059,
 425.1297229 , 527.61676714, 431.45684016, 424.0769184 ,
 575.76543296, 484.89856554, 458.35936863, 481.96502182,
 502.32441491, 513.63783554, 507.58877002, 646.57464283,
 450.24372141, 496.27043415, 556.40457807, 554.95630839,
 399.64237199, 325.84623136, 532.89783259, 478.12238702,
 501.05701845, 305.97335848, 505.77244448, 483.79591969,
 518.8331528 , 438.18241857, 456.71094234, 471.04609461,
 494.44008972, 445.31155755, 508.78802753, 501.04594193,
 488.83499673, 535.38079541, 595.20129802, 514.04714872,
 280.76758312, 433.10112367, 421.70823427, 481.23640152,
 584.71372272, 608.7748096 , 563.98513427, 494.72804869,
 394.52133407, 456.4197529 , 573.08767515, 499.6984241 ,
 512.83277025, 392.12434043, 480.05057697, 481.54520299,
 475.1117359 , 546.2717533 , 430.85039085, 602.16082001,
 422.3695128 , 493.57280186, 528.74970313, 581.49002635,
 620.19139276, 512.56880298, 411.76623862, 498.47637494,
 461.51337557, 446.41371051, 448.07229961, 535.44710412,
 599.45225302, 619.33717662, 494.15919062, 671.99976398,
 532.46469814, 438.90606319, 515.04975242, 546.7821954 ,
 331.94282076, 510.51987447, 536.57891032, 500.19533618,
 376.92345776, 573.73961388, 479.68031607, 588.61435483,
 485.69922203, 456.40200844, 399.25197845, 451.5098931 ,
 519.40693826, 434.71194217, 596.13049586, 487.91791966,
 407.46691799, 524.16812757, 504.12982787, 452.11540623,
 524.21791295, 457.59311643, 444.19371592, 457.80432916,
 448.76590761, 438.31789012, 677.04967982, 566.09639245,
 651.93616661, 381.08127926, 577.5577254 , 578.35797052,
 518.61431291, 538.94532336, 377.4301223 , 663.30814872,
 523.83158824, 456.86065622, 446.07594402, 388.55038282,
 521.03242183, 431.94999241, 460.08016327, 426.31959507,
 433.30417088, 634.89577554, 462.41086078, 460.71673829,
 512.49535288, 703.83033889, 411.84238624, 551.54681408,
 553.33669558, 409.68202123, 423.34491341, 509.66438623,
 509.88865178, 543.67591782, 504.31300469, 519.18802223,
 520.03155195, 535.13855037])
```

```
In [37]: sns.scatterplot(x = predictions,y = y_test)
plt.xlabel('predictions from out model')
plt.title('evaluation of our model')
```

```
Out[37]: Text(0.5, 1.0, 'evaluation of our model')
```

