Logistic Regression

April 24, 2019

1 Logistic Regression Project

In this project I worked with a fake advertising data set, indicating whether or not a particular internet user clicked on an Advertisement. I have tried to create a logistic regression model that would predict whether or not they will click on an ad based off the features of that user.

The data set had the following features:

- 'Daily Time Spent on Site': consumer time on site in minutes
- 'Age': cutomer age in years
- 'Area Income': Avg. Income of geographical area of consumer
- 'Daily Internet Usage': Avg. minutes a day consumer is on the internet
- 'Ad Topic Line': Headline of the advertisement
- 'City': City of consumer
- 'Male': Whether or not consumer was male
- 'Country': Country of consumer
- 'Timestamp': Time at which consumer clicked on Ad or closed window
- 'Clicked on Ad': 0 or 1 indicated clicking on Ad

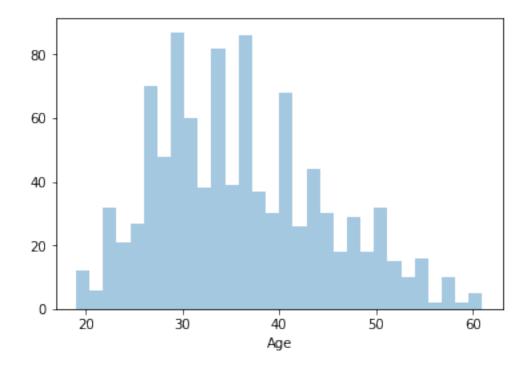
```
In [7]: import sklearn
       import numpy as np
       import seaborn as sns
In [2]: from sklearn.linear_model import LogisticRegression
In [3]: import pandas as pd
In [5]: data = pd.read_csv('advertising.csv')
In [6]: data.head()
Out [6]:
          Daily Time Spent on Site Age Area Income Daily Internet Usage
       0
                             68.95
                                     35
                                            61833.90
                                                                    256.09
       1
                             80.23
                                     31
                                            68441.85
                                                                    193.77
                                     26
                             69.47
                                            59785.94
                                                                    236.50
        3
                             74.15
                                     29
                                            54806.18
                                                                    245.89
                             68.37
                                     35 73889.99
                                                                    225.58
                                  Ad Topic Line
                                                           City Male
                                                                          Country \
```

0	Cloned 5thgeneration orchestration	Wrightburgh	0	Tunisia
1	Monitored national standardization	West Jodi	1	Nauru
2	Organic bottom-line service-desk	Davidton	0	San Marino
3	Triple-buffered reciprocal time-frame	West Terrifurt	1	Italy
4	Robust logistical utilization	South Manuel	0	Iceland

	T	imestamp	Clicked	on Ad
0	2016-03-27 (00:53:11		0
1	2016-04-04 (01:39:02		0
2	2016-03-13	20:35:42		0
3	2016-01-10	02:31:19		0
4	2016-06-03 (03:36:18		0

In [10]: sns.distplot(data['Age'],kde=False,bins=30)

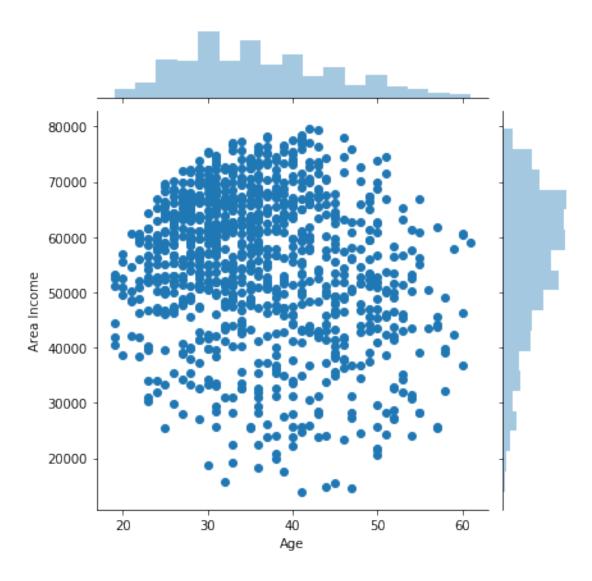
Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x23356bd16a0>



In [13]: sns.jointplot(data['Age'],data['Area Income'])

C:\ProgramData\Anaconda3\lib\site-packages\scipy\stats\stats.py:1713: FutureWarning: Using a new return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval

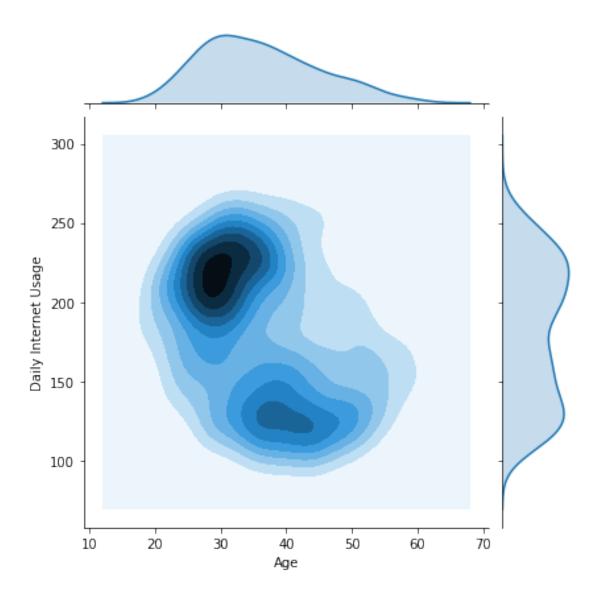
Out[13]: <seaborn.axisgrid.JointGrid at 0x233572e3cc0>



In [14]: sns.jointplot(data['Age'],data['Daily Internet Usage'],kind='kde')

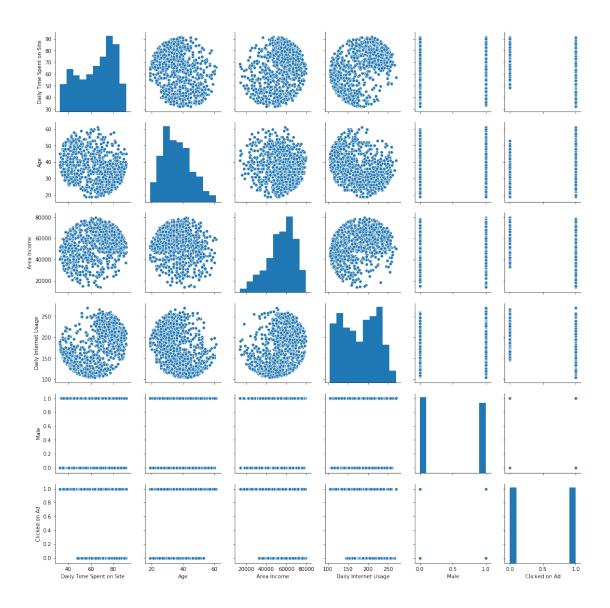
C:\ProgramData\Anaconda3\lib\site-packages\scipy\stats\py:1713: FutureWarning: Using a new return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval

Out[14]: <seaborn.axisgrid.JointGrid at 0x233573ca7f0>



In [15]: sns.pairplot(data)

Out[15]: <seaborn.axisgrid.PairGrid at 0x23357363748>



In [16]: data.corr()

Out[16]:	Daily Time Spent on Site Age Area Income \
Daily Time Spent on Site	1.000000 -0.331513 0.310954
Age	-0.331513 1.000000 -0.182605
Area Income	0.310954 -0.182605 1.000000
Daily Internet Usage	0.518658 -0.367209 0.337496
Male	-0.018951 -0.021044 0.001322
Clicked on Ad	-0.748117 0.492531 -0.476255
	Daily Internet Usage Male Clicked on Ad
Daily Time Spent on Site	0.518658 -0.018951 -0.748117
Age	-0.367209 -0.021044 0.492531

```
Area Income 0.337496 0.001322 -0.476255
Daily Internet Usage 1.000000 0.028012 -0.786539
Male 0.028012 1.000000 -0.038027
Clicked on Ad -0.786539 -0.038027 1.000000
```

Performed various visualizations to find if there is any multi collinearity present in the independent variables

```
In [25]: data x = data
In [27]: data_y = data_x.pop('Clicked on Ad')
   Splitting the dataset into training and testing data
In [28]: from sklearn.model_selection import train_test_split
In [108]: train x, test x, train y, test y = train test split(data x,
                                                                 data_y, test_size=0.30,
                                                                 random_state=101)
   Using Logistic Regression from Scikit Learn and fitting it on training data and then using it to
predict targets on test data
In [30]: from sklearn.linear_model import LogisticRegression
In [99]: logmodel = LogisticRegression()
         logmodel.fit(train_x,train_y)
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:433: FutureWarning
  FutureWarning)
Out[99]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=True,
                    intercept_scaling=1, max_iter=100, multi_class='warn',
                   n_jobs=None, penalty='12', random_state=None, solver='warn',
                    tol=0.0001, verbose=0, warm_start=False)
In [37]: predict_y = logmodel.predict(test_x)
In [102]: logmodel.coef_
Out[102]: array([[-4.45816498e-02, 2.18379839e-01, -7.63621393e-06,
                   -2.45264007e-02, 1.13334440e-03]])
   In the following lines, I have created a confusion matrix to check the classification rate
In [38]: from sklearn.metrics import confusion_matrix
```

In [39]: cm =confusion_matrix(test_y,predict_y)

cm

In the following lines, I have played with pandas to add predicted target and actual target into dataframe and created a new variable which reports whether classification was correct or not.

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm

```
C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm This is separate from the ipykernel package so we can avoid doing imports until

Out[40]:	Daily Time Spent	on Site	Age	Area Income	Daily Internet Usage	Male	\
545		42.60	55	55121.65	168.29	0	
298		80.39	31	66269.49	214.74	0	
109		74.02	32	72272.90	210.54	0	
837		55.77	49	55942.04	117.33	1	
194		77.20	33	49325.48	254.05	1	
605		64.67	51	24316.61	138.35	1	
246		39.96	45	59610.81	146.13	1	
693		65.15	29	41335.84	117.30	0	
802		50.63	50	25767.16	142.23	0	
406		56.34	50	68713.70	139.02	1	
824		68.47	28	67033.34	226.64	0	
977		54.97	31	51900.03	116.38	1	
742		71.90	29	72203.96	193.29	1	
341		72.23	48	52736.33	115.35	0	
601		52.13	50	40926.93	118.27	1	
319		62.06	44	44174.25	105.00	0	
121		49.58	26	56791.75	231.94	0	
299		65.80	25	60843.32	231.49	1	
32		59.05	57	25583.29	169.23	1	
269		79.15	26	62312.23	203.23	0	
887		52.17	44	57594.70	115.37	1	
668		52.56	31	33147.19	250.36	1	
820		57.51	38	47682.28	105.71	0	

754		76.87	36	72948.76	212.59	0
731		61.87	35	66629.61	250.20	1
225		78.70	30	53441.69	133.99	0
569		71.28	37	67307.43	246.72	1
56		65.19	36	75254.88	150.61	0
408		51.68	49	51067.54	258.62	0
734		56.57	26	56435.60	131.98	0
532		79.53		46132.18	244.91	0
866		86.58	32	60151.77	195.93	1
555		72.53	37	73474.82	223.93	0
716		39.96	35	53898.89	138.52	1
81		73.46	28	65653.47	222.75	1
756		56.93	37	57887.64	111.80	0
796		79.22	27	63429.18	198.79	1
612		75.19	31	33502.57	245.76	1
699		73.88	29	63109.74	233.61	0
915		34.04	34	40182.84	174.88	1
801		53.33	34	44275.13	111.63	1
729		79.60	28	56570.06	227.37	1
577		81.21	36	63394.41	233.04	0
260		74.65	28	67669.06	212.56	0
938		32.60	38	40159.20	190.05	0
389		63.88	38	19991.72	136.85	0
476		78.76	24	46422.76	219.98	1
202		35.65	40	31265.75	172.58	1
442		36.44	39	52400.88	147.64	1
113		46.13	46	37838.72	123.64	1
620		81.75	24	52656.13	190.08	1
550		75.80	36	71222.40	224.90	0
27		51.95	52	58295.82	129.23	0
720		75.71	34	62109.80	246.06	0
945		89.80	36	57330.43	198.24	0
847		88.82	36	58638.75	169.10	0
682		43.57	36	50971.73	125.20	1
884		70.92	39	66522.79	249.81	1
325		36.87	36	29398.61	195.91	0
482		69.42	25	65791.17	213.38	0
	Predicted Target	Actual				
545	1		1			
298	0		0			
109	0		0			
837	1		1			
194	0		0			
605	1		1			
246	1		1			
693	1		1			

000	4	4
802	1	1
406	1	1
824	0	0
977	1	1
742	0	0
341	1	1
601	1	1
319	1	1
121	0	0
299	0	0
32	1	1
269	0	0
887	1	1
668	0	1
820	1	1
754	0	0
731	0	0
225	0	1
569	0	0
56	1	1
408	1	1
734	0	1
• •	• • •	
532	1	0
866	0	0
555	0	0
716	1	1
81	0	0
756	1	1
796	0	0
612	0	0
699	0	0
915	1	1
801	1	1
729	0	0
577	0	0
260	0	0
938	1	1
389	1	1
476	0	0
202	1	1
442	1	1
113	1	1
620	0	0
550	0	0
27	1	1
720	0	0
945	0	0
- ==	· ·	J

```
847
                             0
         682
                             1
         884
                             0
         325
                              1
         482
                             0
         [300 rows x 7 columns]
In [79]: test_y = test_y.tolist()
In [81]: correct =[]
         for i in range(0, len(test_y)):
             Yo = test_y[i];Ys= predict_y[i]
             if(Yo==Ys):
                 correct.append('Yes')
             else:
                 correct.append('No')
         b['Correct Prediction?'] = correct
```

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:8: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

0

1

0

1

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm

```
In [113]: b.head()
```

Out[113]:	Daily Time	Spent on Site	Age	Area Income	Daily Internet Usage	Male	\
54	5	42.60	55	55121.65	168.29	0	
29	8	80.39	31	66269.49	214.74	0	
10	9	74.02	32	72272.90	210.54	0	
83	7	55.77	49	55942.04	117.33	1	
19	4	77.20	33	49325.48	254.05	1	

	Predicted Target	Actual Target	Correct Prediction?
545	1	1	Yes
298	0	0	Yes
109	0	0	Yes
837	1	1	Yes
194	0	0	Yes

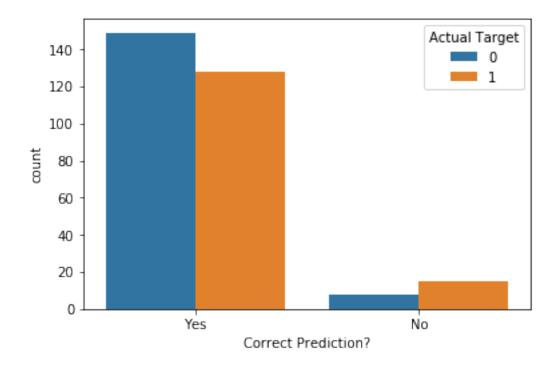
```
In [55]: from sklearn.metrics import accuracy_score
```

In [94]: accuracy = accuracy_score(test_y,predict_y)

In [95]: accuracy

Out [95]: 0.9233333333333333

```
In [115]: sns.countplot(b['Correct Prediction?'],hue= b['Actual Target'])
Out[115]: <matplotlib.axes._subplots.AxesSubplot at 0x2335e3a4cc0>
```



We have got an accuracy of 0.92 from the model created and the above visulaization shows how many correct predictions were made values 0 and 1 individually.

0.52333333333333333

I have further compared this with MLP classifier model from neural networks which produced very less accuracy.