

So we have to compare two machine learning algorithm to see which one is better based on our dataset.

Knn vs Logistic regression

I will inform you my answer here when i done the compare i will upload a picture of knn score and LG score !

Logistic regression score is 0.8780487804878049

knn score is 0.8926829268292683 k = 5 same as previous lab (lab 1)

thanks dr.saeed <3

```
In [1]: # Let's import the packages that we need
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
%matplotlib inline
```

```
In [2]: # we have to Load data (already we have split) so we have test and train
dtest = pd.read_csv("test_set.csv")
dtrain = pd.read_csv("Train_set.csv")
```

```
In [3]: # we sholud understanding what the data is ?
# import first 5 (haed of table ) data in table
```

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

Out[4]:

	Height	Weight	Sex
0	165.65	35.41	Female
1	148.53	74.45	Female
2	167.04	81.22	Male
3	161.54	71.47	Male
4	174.31	78.18	Male

```
In [ ]: # same as previous
```

```
In [5]: dtrain.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 2998 entries, 0 to 2997  
Data columns (total 3 columns):  
#   Column  Non-Null Count  Dtype  
---  ---  
0   Height   2998 non-null     float64  
1   Weight   2998 non-null     float64  
2   Sex       2998 non-null     object  
dtypes: float64(2), object(1)  
memory usage: 70.4+ KB
```

```
In [ ]: # to get the type
```

```
In [6]: type(dtrain)
```

```
Out[6]: pandas.core.frame.DataFrame
```

```
In [7]: dtrain.keys()
```

```
Out[7]: Index(['Height', 'Weight', 'Sex'], dtype='object')
```

```
In [ ]: # here er can also know the shape of data (row,columns)
```

```
In [8]: dtrain.shape
```

```
Out[8]: (2998, 3)
```

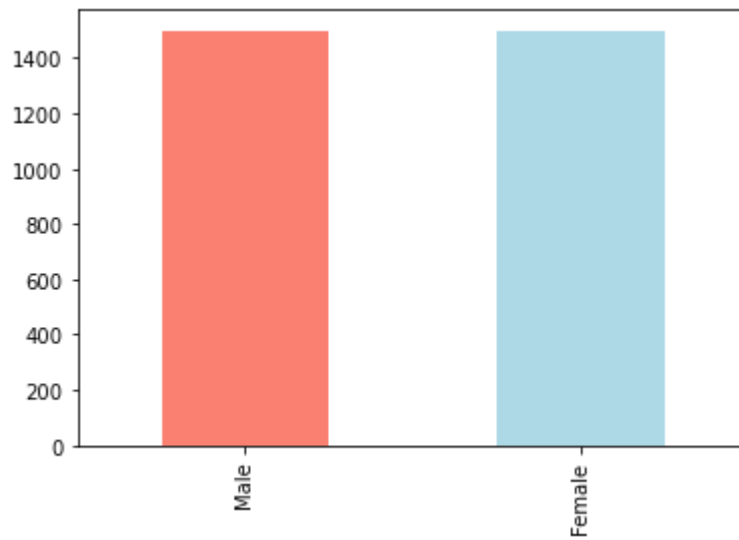
```
In [ ]: # i decided that i have to count how many female and male we have
```

```
In [9]: dtrain.Sex.value_counts()
```

```
Out[9]: Male      1500  
        Female    1498  
        Name: Sex, dtype: int64
```

```
In [ ]: # represent part
```

```
In [10]: dtrain.Sex.value_counts().plot(kind="bar", color=["salmon", "lightblue"]);
```



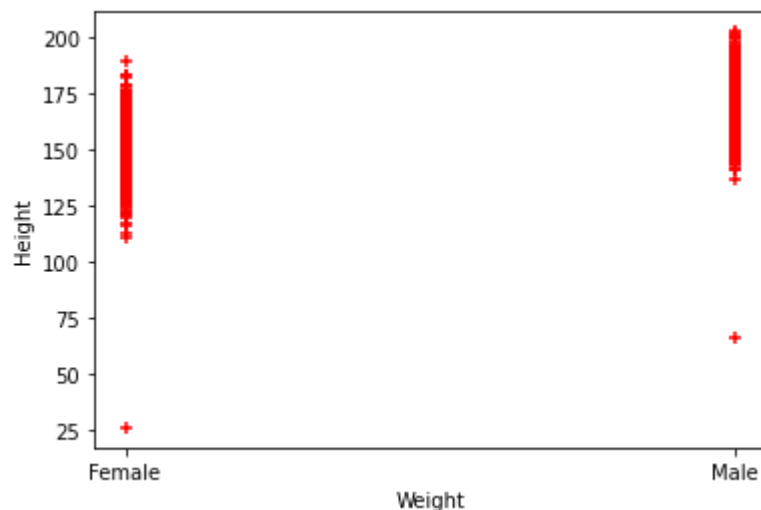
```
In [11]: plt.scatter(dtrain.Sex, dtrain.Weight, marker='+', color='blue')
```

```
Out[11]: <matplotlib.collections.PathCollection at 0x20b265cd9a0>
```



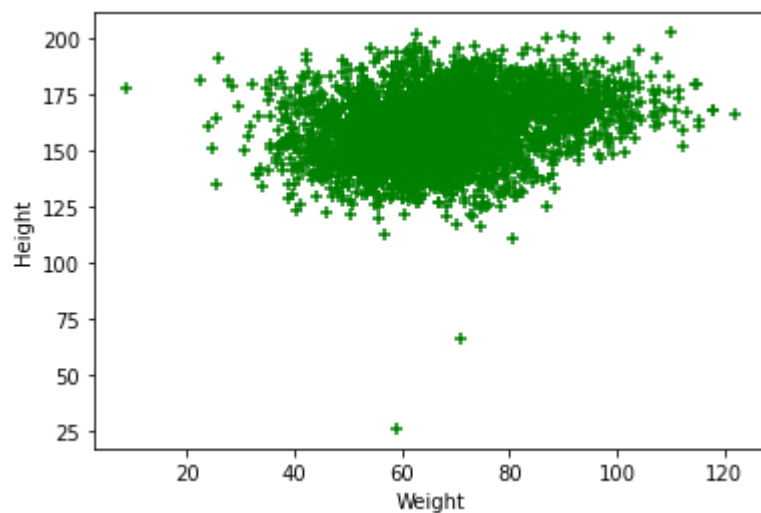
```
In [12]: plt.scatter(dtrain.Sex,dtrain.Height,marker='+',color='red')  
plt.ylabel("Height")  
plt.xlabel("Weight")
```

Out[12]: Text(0.5, 0, 'Weight')



```
In [13]: plt.scatter(dtrain.Weight,dtrain.Height,marker='+',color='green')  
plt.ylabel("Height")  
plt.xlabel("Weight")
```

Out[13]: Text(0.5, 0, 'Weight')



```
In [14]: # and we have to drop the target from table ,
#Split the dataset into features and target
X_train=dtrain.drop('Sex',axis='columns')
X_train
```

Out[14]:

	Height	Weight
0	165.65	35.41
1	148.53	74.45
2	167.04	81.22
3	161.54	71.47
4	174.31	78.18
...
2993	150.83	49.66
2994	157.09	64.34
2995	162.99	45.58
2996	154.76	48.92
2997	185.08	82.74

2998 rows × 2 columns

```
In [15]: y_train=dtrain['Sex']
y_train
```

Out[15]:

0	Female
1	Female
2	Male
3	Male
4	Male
...	
2993	Female
2994	Female
2995	Female
2996	Female
2997	Male

Name: Sex, Length: 2998, dtype: object

```
In [16]: # and we have to drop the target from table ,
#Split the dataset into features and target

X_test=dtest.drop('Sex',axis='columns')
X_test
```

Out[16]:

	Height	Weight
0	146.323241	59.861065
1	175.695412	77.863687
2	183.216164	72.131992
3	184.245269	77.546000
4	132.302261	55.188496
...
200	155.090314	77.248911
201	149.175907	93.231692
202	168.030874	63.640623
203	172.608090	55.189983
204	145.082128	45.583285

205 rows × 2 columns

```
In [17]: y_test=dtest['Sex']
y_test
```

Out[17]:

0	Female
1	Male
2	Male
3	Male
4	Female
...	...
200	Female
201	Male
202	Female
203	Male
204	Female

Name: Sex, Length: 205, dtype: object

```
In [18]: #Import package LogisticRegression

from sklearn.linear_model import LogisticRegression
lgrgmodel=LogisticRegression()
```

```
In [ ]: # fti the model !
```

```
In [19]: lgrgmodel.fit(X_train,y_train)
```

```
Out[19]: LogisticRegression()
```

```
In [20]: #optimization
lgrg_pred=lgrgmodel.predict(X_test)
lgrg_pred
```

```
Out[20]: array(['Female', 'Male', 'Male', 'Male', 'Female', 'Female', 'Male',
                'Female', 'Male', 'Male', 'Male', 'Male', 'Female', 'Female',
                'Female', 'Female', 'Male', 'Male', 'Male', 'Male', 'Female',
                'Male', 'Male', 'Female', 'Female', 'Female', 'Female', 'Male',
                'Male', 'Male', 'Female', 'Female', 'Male', 'Female', 'Female',
                'Male', 'Female', 'Male', 'Female', 'Male', 'Female', 'Female',
                'Male', 'Female', 'Male', 'Female', 'Male', 'Female', 'Male',
                'Female', 'Male', 'Male', 'Female', 'Male', 'Male', 'Male', 'Male',
                'Female', 'Female', 'Male', 'Female', 'Female', 'Male', 'Male',
                'Female', 'Male', 'Female', 'Female', 'Female', 'Female', 'Female',
                'Male', 'Male', 'Male', 'Female', 'Female', 'Female', 'Male',
                'Female', 'Male', 'Male', 'Male', 'Female', 'Female', 'Male',
                'Male', 'Female', 'Female', 'Male', 'Male', 'Female', 'Female',
                'Male', 'Female', 'Female', 'Male', 'Male', 'Female', 'Female',
                'Male', 'Female', 'Female', 'Male', 'Female', 'Male', 'Male',
                'Female', 'Male', 'Female', 'Male', 'Male', 'Female', 'Female',
                'Female', 'Male', 'Female', 'Male', 'Female', 'Male', 'Male',
                'Female', 'Male', 'Female', 'Male', 'Female', 'Male', 'Male',
                'Female', 'Male', 'Female', 'Male', 'Male', 'Male', 'Female',
                'Male', 'Female', 'Female', 'Male', 'Female', 'Male', 'Female',
                'Female', 'Female', 'Female', 'Female', 'Male', 'Male', 'Female'],
              dtype=object)
```

```
In [21]: # get the accuracy and compare it with Knn
# our accuracy is 87
lgrgmodel.score(X_test,y_test)
```

```
Out[21]: 0.8780487804878049
```

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