

DEPARTMENT OF INFORMATION TECHNOLOGY

Academic Year: 2020-2021

Subject: Data Mining and Business Intelligence

Evaluation of Bike Sharing in Washington D.C.

A mini-project submitted for **Business Intelligence Lab (Sem VI)**By

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Problem Statement

Bike sharing systems are a new generation of traditional bike rentals where the whole process from membership, rental and return back has become automatic. Through these systems, user is able to easily rent a bike from a particular position and return back to another position. Currently, there are about over 500 bike-sharing programs around the world which are composed of over 500 thousands bicycles. Today, there exists great interest in these systems due to their important role in traffic, environmental and health issues.

This dataset contains the hourly and daily count of rental bikes between years 2011 and 2012 in Capital bikeshare system in Washington, DC with the corresponding weather and seasonal information.

Dataset link: https://www.kaggle.com/marklvl/bike-sharing-dataset

```
In [ ]:
import numpy as np
import pandas as pd
In [ ]:
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
In [ ]:
day df=pd.read csv('day.csv')
hour df=pd.read csv('hour.csv')
In [ ]:
day df.head()
Out[]:
   instant dteday season yr mnth holiday weekday workingday weathersit
                                                                          temp
                                                                                 atemp
                                                                                            hum windspeed ca
           2011-
0
                                      0
                                               6
                                                          0
                                                                    2 0.344167 0.363625 0.805833
       1
                         0
                               1
                                                                                                   0.160446
           01-01
           2011-
                                                          0
                                                                    2 0.363478 0.353739 0.696087
                                                                                                   0.248539
1
       2
                         0
                                      0
                                               0
           01-02
           2011-
                                                                       0.196364 0.189405 0.437273
                                                                                                   0.248309
2
       3
                         0
                                      0
                                               1
           01-03
           2011-
                                      n
                                               2
                                                                                                   0.160296
3
                         0
                               1
                                                          1
                                                                    1 0.200000 0.212122 0.590435
           01-04
```

```
In [ ]:
```

2011-

01-05

0

```
hour_df.head()
```

1 0.226957 0.229270 0.436957

0.186900

3

	instant	dteday	season	yr	mnth	hr	holiday	weekday	workingday	weathersit	temp	atemp	hum	windspeed	casual
0	1	2011- 01-01	1	0	1	0	0	6	0	1	0.24	0.2879	0.81	0.0	3
1	2	2011- 01-01	1	0	1	1	0	6	0	1	0.22	0.2727	0.80	0.0	8

```
2 instant dteday
                 season yo mnth he holiday weekday workingday weathersit telmas atternio numb windspeed casual
           2011-
3
                                                    6
                                                                              0.24 0.2879 0.75
                                                                                                               3
                       1 0
                                   3
                                                                                                      0.0
           01-01
           2011-
                       1 0
                                                                          1 0.24 0.2879 0.75
                                                                                                       0.0
                                                                                                               0
           01-01
```

```
day_df.info()
<class 'pandas.core.frame.DataFrame'>
```

RangeIndex: 731 entries, 0 to 730
Data columns (total 16 columns):

#	Column	Non-Null Count	Dtype						
0	instant	731 non-null	int64						
1	dteday	731 non-null	object						
2	season	731 non-null	int64						
3	yr	731 non-null	int64						
4	mnth	731 non-null	int64						
5	holiday	731 non-null	int64						
6	weekday	731 non-null	int64						
7	workingday	731 non-null	int64						
8	weathersit	731 non-null	int64						
9	temp	731 non-null	float64						
10	atemp	731 non-null	float64						
11	hum	731 non-null	float64						
12	windspeed	731 non-null	float64						
13	casual	731 non-null	int64						
14	registered	731 non-null	int64						
15	cnt	731 non-null	int64						
dtypes: float64(4), int64(11), object(1)									
memory usage: 91.5+ KB									

In []:

hour df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 17379 entries, 0 to 17378
Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype						
0 1 2 3 4 5 6 7 8 9 10 11 12 13 14	instant dteday season yr mnth hr holiday weekday workingday weathersit temp atemp hum windspeed casual	17379 non-null	int64 object int64 int64 int64 int64 int64 int64 int64 float64 float64 float64 float64 float64						
15		17379 non-null							
15	registered	17379 non-null	int64						
16 cnt 17379 non-null int64 dtypes: float64(4), int64(12), object(1) memory usage: 2.3+ MB									

In []:

```
day_df.drop('instant',axis=1,inplace=True)
```

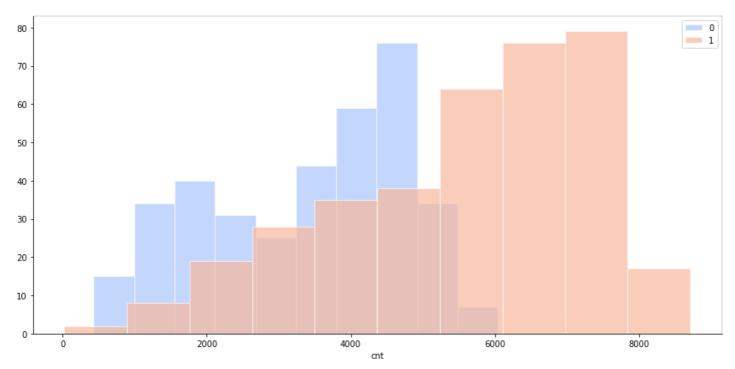
In []:

```
g=sns.FacetGrid(day_df, hue='yr', palette='coolwarm', size=6, aspect=2)
g=g.map(plt.hist,'cnt',alpha=0.7, edgecolor='w')
plt.legend()
```

/usr/local/lib/python3.7/dist-packages/seaborn/axisgrid.py:316: UserWarning: The `size` p arameter has been renamed to `height`; please update your code.
warnings.warn(msg, UserWarning)

Out[]:

<matplotlib.legend.Legend at 0x7f1334f90fd0>



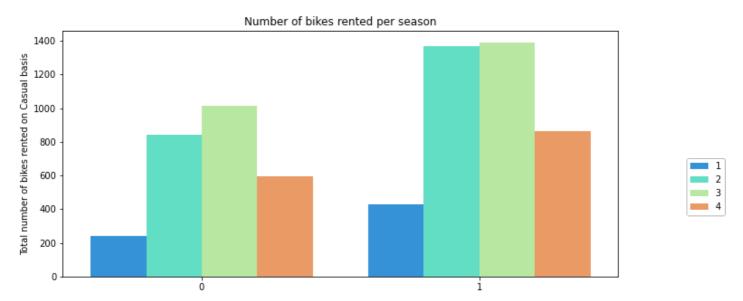
In []:

```
plt.figure(figsize=(11,5))
sns.barplot('yr','casual',hue='season', data=day_df,palette='rainbow', ci=None)
plt.legend(loc='upper right',bbox_to_anchor=(1.2,0.5))
plt.xlabel('Year')
plt.ylabel('Total number of bikes rented on Casual basis')
plt.title('Number of bikes rented per season')
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be 'data', and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Text(0.5, 1.0, 'Number of bikes rented per season')



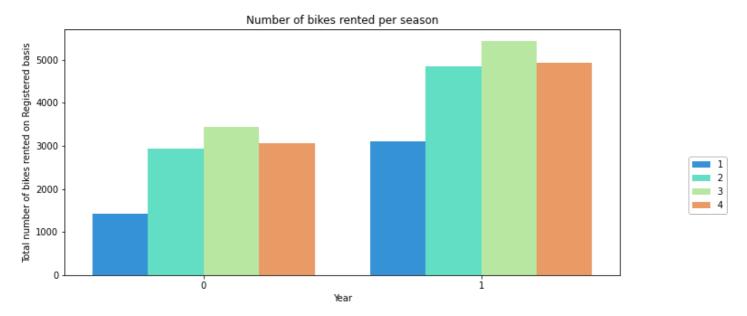
```
plt.figure(figsize=(11,5))
sns.barplot('yr','registered',hue='season', data=day_df,palette='rainbow', ci=None)
plt.legend(loc='upper right',bbox_to_anchor=(1.2,0.5))
plt.xlabel('Year')
plt.ylabel('Total number of bikes rented on Registered basis')
plt.title('Number of bikes rented per season')
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Out[]:

Text(0.5, 1.0, 'Number of bikes rented per season')



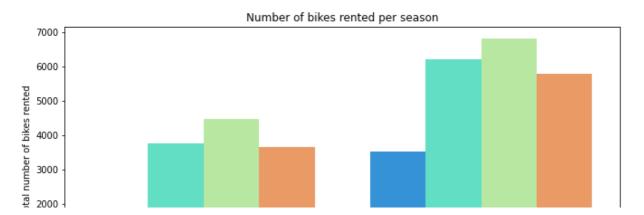
In []:

```
plt.figure(figsize=(11,5))
sns.barplot('yr','cnt',hue='season', data=day_df,palette='rainbow', ci=None)
plt.legend(loc='upper right',bbox_to_anchor=(1.2,0.5))
plt.xlabel('Year')
plt.ylabel('Total number of bikes rented')
plt.title('Number of bikes rented per season')
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Text(0.5, 1.0, 'Number of bikes rented per season')



```
1000 - 0 Year
```

```
df_season_fall=day_df[day_df['season']==3]
```

In []:

```
df_season_fall.mnth.nunique()
```

Out[]:

Л

In []:

sns.factorplot('mnth','cnt',hue='workingday',data=df_season_fall, ci=None, palette='Set2')

/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:3714: UserWarning: The `fac torplot` function has been renamed to `catplot`. The original name will be removed in a f uture release. Please update your code. Note that the default `kind` in `factorplot` (`'p oint'`) has changed `'strip'` in `catplot`.

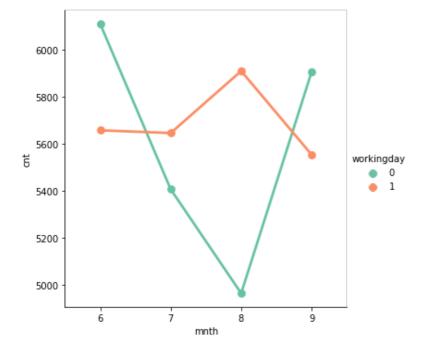
warnings.warn(msg)

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Out[]:

<seaborn.axisgrid.FacetGrid at 0x7f1333697190>



In []:

```
sns.factorplot('mnth','cnt',hue='weathersit',data=df_season_fall, ci=None, palette='Set2')
```

/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:3714: UserWarning: The `fac torplot` function has been renamed to `catplot`. The original name will be removed in a f uture release. Please update your code. Note that the default `kind` in `factorplot` (`'p oint'`) has changed `'strip'` in `catplot`.

warnings.warn(msg)

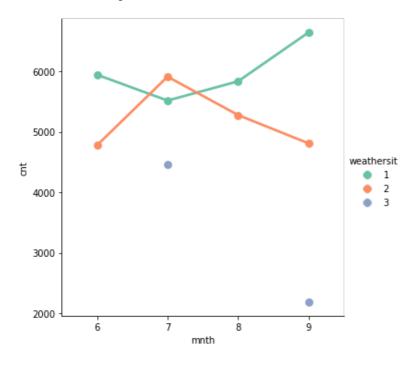
/usr/local/lib/python3.7/dist-packages/seaborn/ decorators.py:43: FutureWarning: Pass the

following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Out[]:

<seaborn.axisgrid.FacetGrid at 0x7f1333cb91d0>



In []:

sns.jointplot('temp','cnt',data=day_df,size=7)

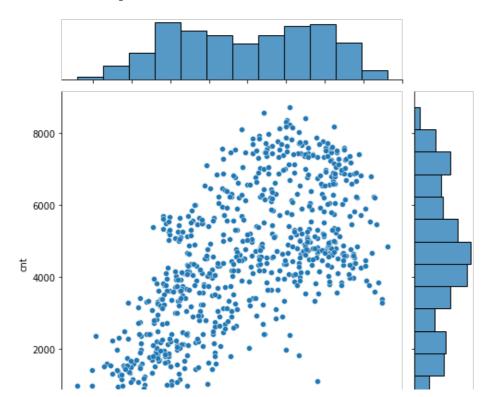
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

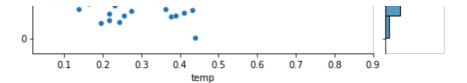
FutureWarning

/usr/local/lib/python3.7/dist-packages/seaborn/axisgrid.py:2073: UserWarning: The `size` parameter has been renamed to `height`; please update your code.
warnings.warn(msg, UserWarning)

Out[]:

<seaborn.axisgrid.JointGrid at 0x7f1333557410>





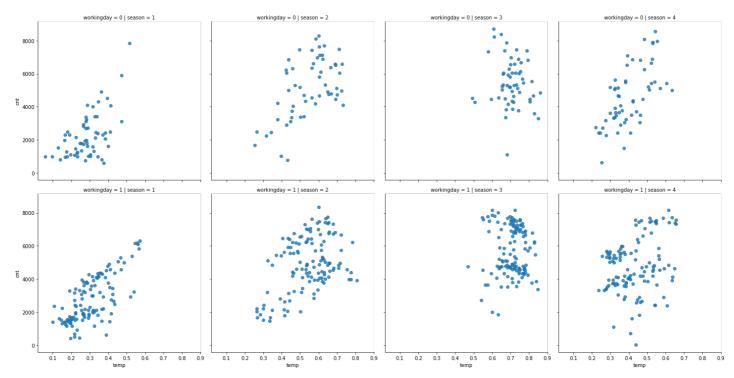
sns.lmplot('temp','cnt',row='workingday',col='season',data=day_df,palette='RdBu_r',fit_re
g=False)

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be 'data', and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Out[]:

<seaborn.axisgrid.FacetGrid at 0x7f13333a2f10>



In []:

hour_df.drop('instant',axis=1,inplace=True)

In []:

```
plt.figure(figsize=(20,5))
mask = np.zeros_like(hour_df.corr(), dtype=np.bool)
mask[np.triu_indices_from(mask)] = True
sns.heatmap(hour_df.corr(),cmap='RdBu_r',mask=mask, annot=True)
```

0.8

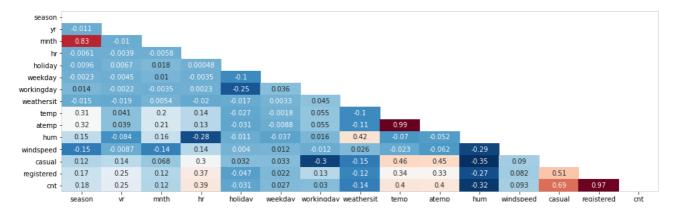
- 0.4

0.2

0.0

Out[]:

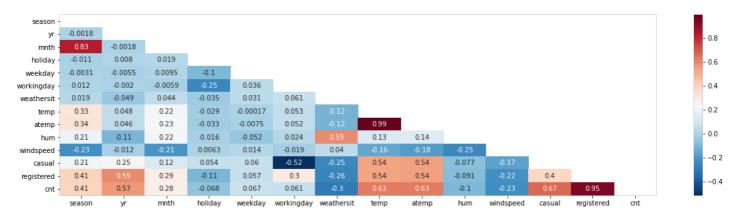
 ${\tt <matplotlib.axes._subplots.AxesSubplot}$ at ${\tt 0x7f1333599650>}$



```
plt.figure(figsize=(20,5))
mask = np.zeros_like(day_df.corr(), dtype=np.bool)
mask[np.triu_indices_from(mask)] = True
sns.heatmap(day_df.corr(),cmap='RdBu_r',mask=mask, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f1331695fd0>



In []:

```
df=pd.merge(day_df,hour_df,how='left',left_on='dteday',right_on='dteday')
```

In []:

df.head()

Out[]:

	dteday	season_x	yr_x	mnth_x	holiday_x	weekday_x	workingday_x	weathersit_x	temp_x	atemp_x	hum_x	windspe
0	2011- 01-01	1	0	1	0	6	0	2	0.344167	0.363625	0.805833	0.16
1	2011- 01-01	1	0	1	0	6	0	2	0.344167	0.363625	0.805833	0.16
2	2011- 01-01	1	0	1	0	6	0	2	0.344167	0.363625	0.805833	0.16
3	2011- 01-01	1	0	1	0	6	0	2	0.344167	0.363625	0.805833	0.16
4	2011- 01-01	1	0	1	0	6	0	2	0.344167	0.363625	0.805833	0.16
4						1						Þ

In []:

```
sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')
```

Out[]:

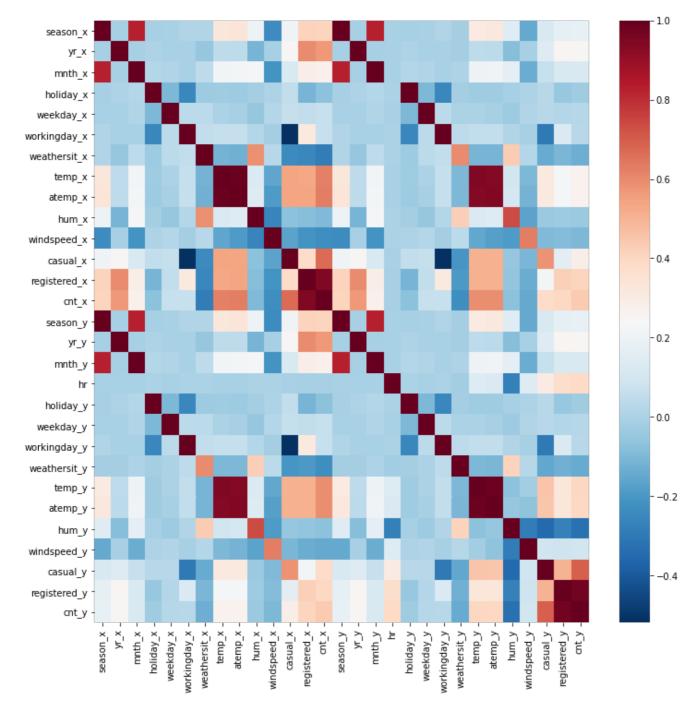
<matplotlib.axes._subplots.AxesSubplot at 0x7f13314c7610>



```
plt.figure(figsize=(12,12))
sns.heatmap(df.corr(),cmap='RdBu_r')
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f13314c2350>



In []:

```
X=df.drop(['dteday','cnt_y'],axis=1)
y=df['cnt_y']
```

In []:

```
df.columns
```

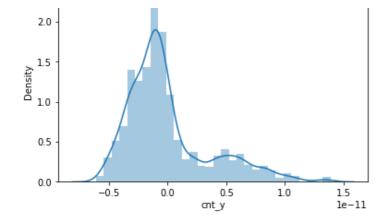
```
Index(['dteday', 'season_x', 'yr_x', 'mnth_x', 'holiday_x', 'weekday_x',
       'workingday_x', 'weathersit_x', 'temp_x', 'atemp_x', 'hum_x',
       'windspeed_x', 'casual_x', 'registered_x', 'cnt_x', 'season_y', 'yr_y',
       'mnth_y', 'hr', 'holiday_y', 'weekday_y', 'workingday_y',
       'weathersit y', 'temp y', 'atemp y', 'hum y', 'windspeed y', 'casual y',
       'registered_y', 'cnt_y'],
      dtype='object')
In [ ]:
from sklearn.model selection import train test split
In [ ]:
X train, X test, y train, y test= train test split(X,y)
In [ ]:
from sklearn.linear model import LinearRegression
In [ ]:
lm=LinearRegression()
In [ ]:
lm.fit(X train, y train)
Out[]:
LinearRegression(copy X=True, fit intercept=True, n jobs=None, normalize=False)
In [ ]:
predictions=lm.predict(X test)
In [ ]:
plt.scatter(y test, predictions)
Out[]:
<matplotlib.collections.PathCollection at 0x7f133137b050>
 1000
 800
 600
 400
 200
            200
                    400
                            600
                                   800
                                          1000
In [ ]:
from sklearn import metrics
In [ ]:
print('MAE= ', metrics.mean_absolute_error(y_test,predictions))
print('MSE= ', metrics.mean_squared_error(y_test,predictions))
print('RMS= ', np.sqrt(metrics.mean_squared_error(y_test,predictions)))
MAE= 2.6609581179598595e-12
MSE= 1.2570070510448574e-23
```

```
Coefficient
    season_x -4.060769e-15
        yr_x -7.403397e-07
     mnth_x -1.035950e-06
    holiday_x -3.548994e-07
  weekday_x 6.743899e-07
workingday_x -2.895134e-08
 weathersit_x 9.344258e-15
     temp_x -6.270922e-13
    atemp_x 1.789259e-13
      hum_x 1.123360e-14
 windspeed_x 2.788286e-14
    casual_x 7.311627e-03
 registered_x 7.311627e-03
       cnt_x -7.311627e-03
    season_y -1.963652e-15
        yr_y 7.403396e-07
     mnth_y 1.035950e-06
         hr -2.473795e-15
    holiday_y 3.548992e-07
  weekday_y -6.743899e-07
workingday_y 2.895137e-08
 weathersit_y -1.063208e-15
     temp_y 5.115475e-13
    atemp_y -1.709650e-13
      hum_y 5.480059e-14
 windspeed_y -1.628038e-15
    casual_y 1.000000e+00
 registered_y 1.000000e+00
In [ ]:
sns.distplot(y_test-predictions,bins=30)
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2557: FutureWarning: `dis
tplot` is a deprecated function and will be removed in a future version. Please adapt you
r code to use either `displot` (a figure-level function with similar flexibility) or `his
tplot` (an axes-level function for histograms).
  warnings.warn(msg, FutureWarning)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f132c1a2090>
```

RMS= 3.545429524112498e-12

pd.DataFrame(data=lm.coef , index=X.columns, columns=['Coefficient'])

In []:



In []:

#End of the project

Conclusion

The analysis gives us a detailed visualisation of the Median Bike Sharing in Washington D.C. We noticed that the major renting have a huge difference in median bike sharing with respect to the climate. Most of the sales are achieved when the weather fits perfectly. The count of the bikes are also striking, which the analysis concludes that the median count hovers around heavy density.

From these analysis, marketing team can focus on targeted marketing easily