

```

import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from scipy.stats import norm, chi2, f
from scipy.stats import ttest_ind, f_oneway, kruskal
from scipy.stats import chisquare, chi2_contingency
from scipy.stats import kstest
from statsmodels.distributions.empirical_distribution import ECDF

from IPython.display import display, HTML
display(HTML("<style>.container { width:100% !important; }</style>"))

<IPython.core.display.HTML object>

import matplotlib_inline
matplotlib_inline.backend_inline.set_matplotlib_formats('svg')

df_yulu =
pd.read_csv("https://d2beiqkhq929f0.cloudfront.net/public_assets/assets/000/001/428/original/bike_sharing.csv?1642089089")

df_yulu.head()

```

	datetime	season	holiday	workingday	weather	temp
atemp \						
0	2011-01-01 00:00:00	1	0	0	1	9.84
14.395						
1	2011-01-01 01:00:00	1	0	0	1	9.02
13.635						
2	2011-01-01 02:00:00	1	0	0	1	9.02
13.635						
3	2011-01-01 03:00:00	1	0	0	1	9.84
14.395						
4	2011-01-01 04:00:00	1	0	0	1	9.84
14.395						

	humidity	windspeed	casual	registered	count
0	81	0.0	3	13	16
1	80	0.0	8	32	40
2	80	0.0	5	27	32
3	75	0.0	3	10	13
4	75	0.0	0	1	1

```

df_yulu.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10886 entries, 0 to 10885
Data columns (total 12 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   datetime        10886 non-null object

```

```

1  season      10886 non-null int64
2  holiday     10886 non-null int64
3  workingday  10886 non-null int64
4  weather     10886 non-null int64
5  temp        10886 non-null float64
6  atemp       10886 non-null float64
7  humidity    10886 non-null int64
8  windspeed   10886 non-null float64
9  casual      10886 non-null int64
10 registered  10886 non-null int64
11 count       10886 non-null int64

```

dtypes: float64(3), int64(8), object(1)

memory usage: 1020.7+ KB

df\_yulu.describe()

	season	holiday	workingday	weather
temp \				
count	10886.000000	10886.000000	10886.000000	10886.000000
10886.000000				
mean	2.506614	0.028569	0.680875	1.418427
20.23086				
std	1.116174	0.166599	0.466159	0.633839
7.79159				
min	1.000000	0.000000	0.000000	1.000000
0.82000				
25%	2.000000	0.000000	0.000000	1.000000
13.94000				
50%	3.000000	0.000000	1.000000	1.000000
20.50000				
75%	4.000000	0.000000	1.000000	2.000000
26.24000				
max	4.000000	1.000000	1.000000	4.000000
41.00000				

	atemp	humidity	windspeed	casual
registered \				
count	10886.000000	10886.000000	10886.000000	10886.000000
10886.000000				
mean	23.655084	61.886460	12.799395	36.021955
155.552177				
std	8.474601	19.245033	8.164537	49.960477
151.039033				
min	0.760000	0.000000	0.000000	0.000000
0.000000				
25%	16.665000	47.000000	7.001500	4.000000
36.000000				
50%	24.240000	62.000000	12.998000	17.000000
118.000000				
75%	31.060000	77.000000	16.997900	49.000000

```
222.000000
max      45.455000    100.000000    56.996900    367.000000
886.000000
```

```
count      count
count  10886.000000
mean    191.574132
std     181.144454
min      1.000000
25%     42.000000
50%    145.000000
75%    284.000000
max     977.000000
```

```
df_yulu.nunique()
```

```
datetime    10886
season        4
holiday       2
workingday    2
weather       4
temp         49
atemp        60
humidity     89
windspeed    28
casual       309
registered   731
count        822
dtype: int64
```

```
df_yulu.columns
```

```
Index(['datetime', 'season', 'holiday', 'workingday', 'weather',
      'temp',
      'atemp', 'humidity', 'windspeed', 'casual', 'registered',
      'count'],
      dtype='object')
```

```
df_yulu.shape
```

```
(10886, 12)
```

```
df_yulu.isnull().sum()
```

```
datetime    0
season      0
holiday     0
workingday  0
weather     0
temp        0
atemp       0
humidity    0
```

```
windspeed    0
casual        0
registered    0
count         0
dtype: int64
```

```
df_yulu.isna().sum()
```

```
datetime      0
season         0
holiday        0
workingday     0
weather        0
temp           0
atemp          0
humidity       0
windspeed      0
casual         0
registered     0
count          0
dtype: int64
```

```
df_yulu.head()
```

		datetime	season	holiday	workingday	weather	temp
atemp \							
0	2011-01-01 00:00:00		1	0	0	1	9.84
14.395							
1	2011-01-01 01:00:00		1	0	0	1	9.02
13.635							
2	2011-01-01 02:00:00		1	0	0	1	9.02
13.635							
3	2011-01-01 03:00:00		1	0	0	1	9.84
14.395							
4	2011-01-01 04:00:00		1	0	0	1	9.84
14.395							

	humidity	windspeed	casual	registered	count
0	81	0.0	3	13	16
1	80	0.0	8	32	40
2	80	0.0	5	27	32
3	75	0.0	3	10	13
4	75	0.0	0	1	1

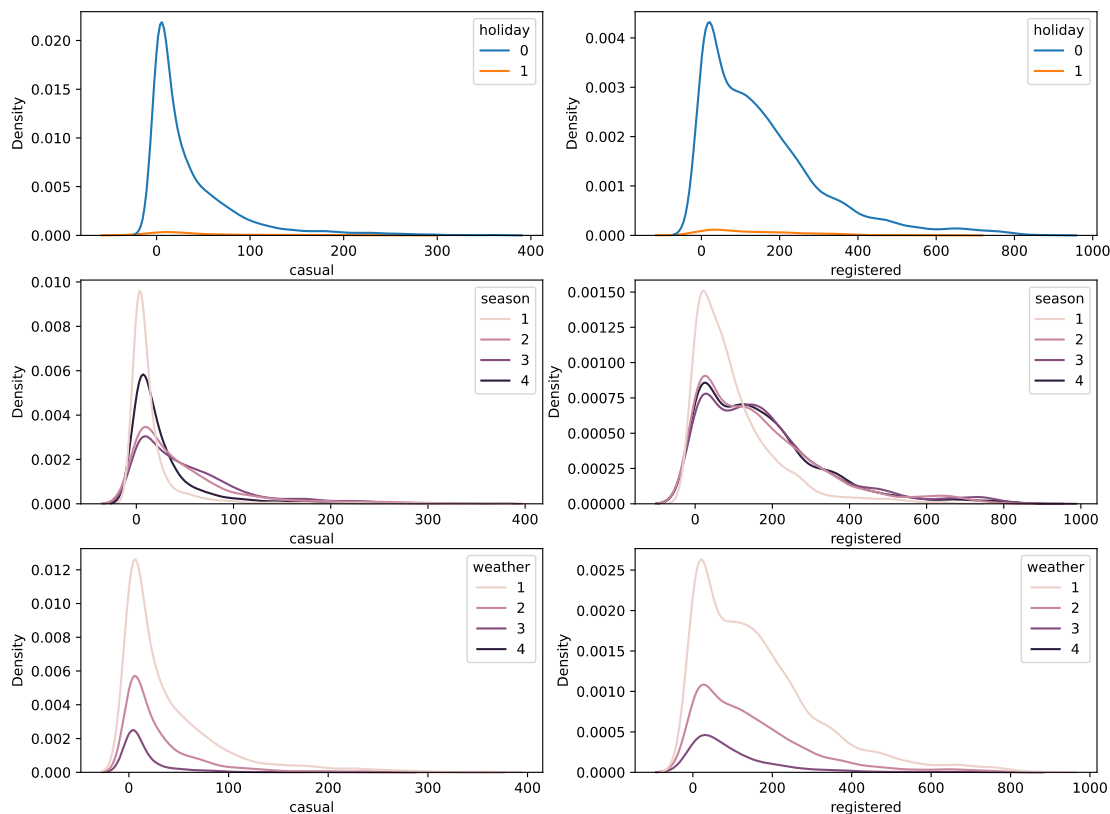
```
plt.figure(figsize=(13,10))
plt.subplot(321)
sns.kdeplot(x=df_yulu["casual"],hue=df_yulu["holiday"])
plt.subplot(322)
sns.kdeplot(x=df_yulu["registered"],hue=df_yulu["holiday"])
plt.subplot(323)
sns.kdeplot(x=df_yulu["casual"],hue=df_yulu["season"])
```

```
plt.subplot(324)
sns.kdeplot(x=df_yulu["registered"],hue=df_yulu["season"])
plt.subplot(325)
sns.kdeplot(x=df_yulu["casual"],hue=df_yulu["weather"])
plt.subplot(326)
sns.kdeplot(x=df_yulu["registered"],hue=df_yulu["weather"])
```

/Users/mrunmay/opt/anaconda3/lib/python3.9/site-packages/seaborn/distributions.py:316: UserWarning: Dataset has 0 variance; skipping density estimate. Pass `warn\_singular=False` to disable this warning.  
warnings.warn(msg, UserWarning)

/Users/mrunmay/opt/anaconda3/lib/python3.9/site-packages/seaborn/distributions.py:316: UserWarning: Dataset has 0 variance; skipping density estimate. Pass `warn\_singular=False` to disable this warning.  
warnings.warn(msg, UserWarning)

<AxesSubplot:xlabel='registered', ylabel='Density'>



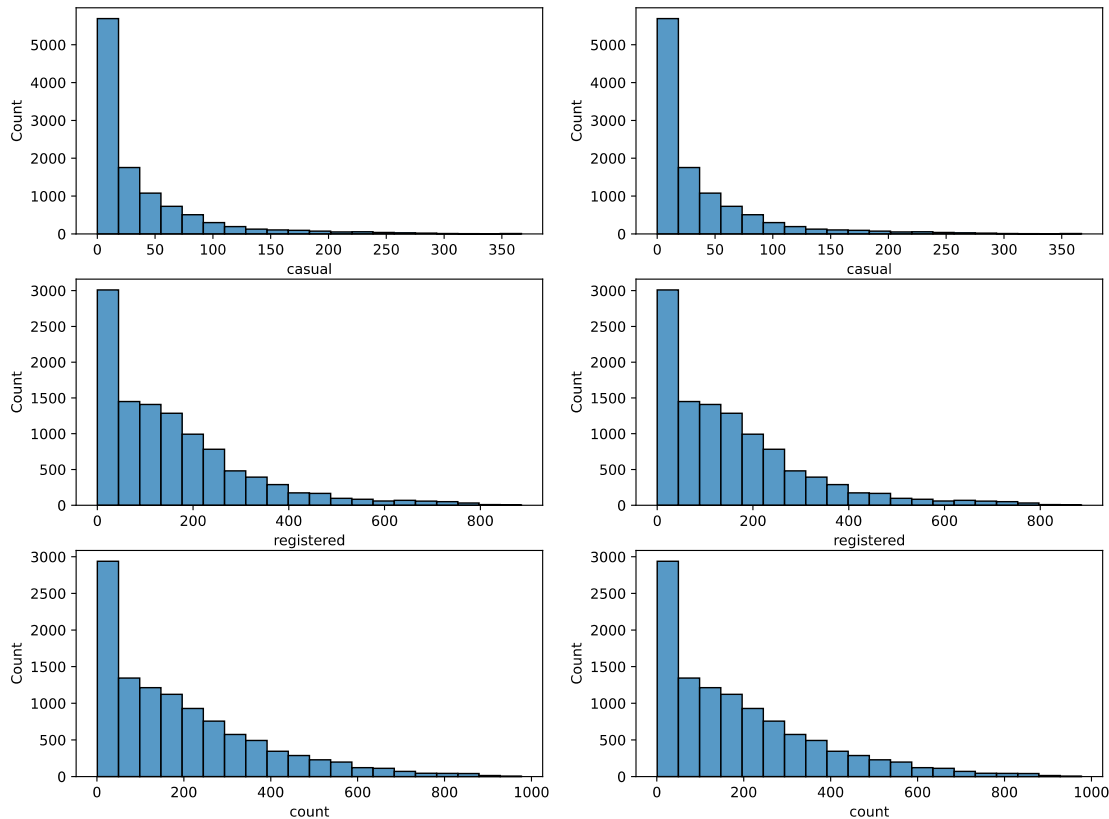
```
plt.figure(figsize=(13,10))
plt.subplot(321)
sns.histplot(df_yulu["casual"],bins=20)
plt.subplot(322)
sns.histplot(df_yulu["casual"],bins=20)
plt.subplot(323)
sns.histplot(df_yulu["registered"],bins=20)
plt.subplot(324)
```

```

sns.histplot(df_yulu["registered"],bins=20)
plt.subplot(325)
sns.histplot(df_yulu["count"],bins=20)
plt.subplot(326)
sns.histplot(df_yulu["count"],bins=20)

```

<AxesSubplot:xlabel='count', ylabel='Count'>

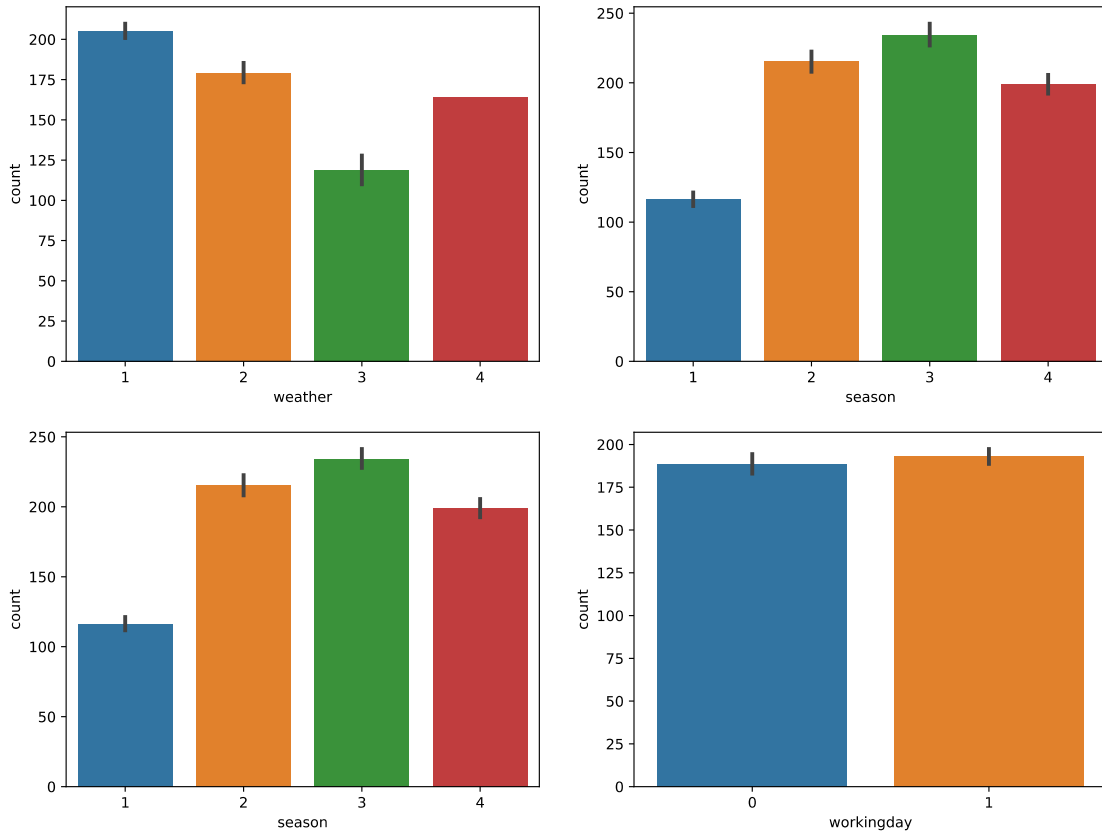


```

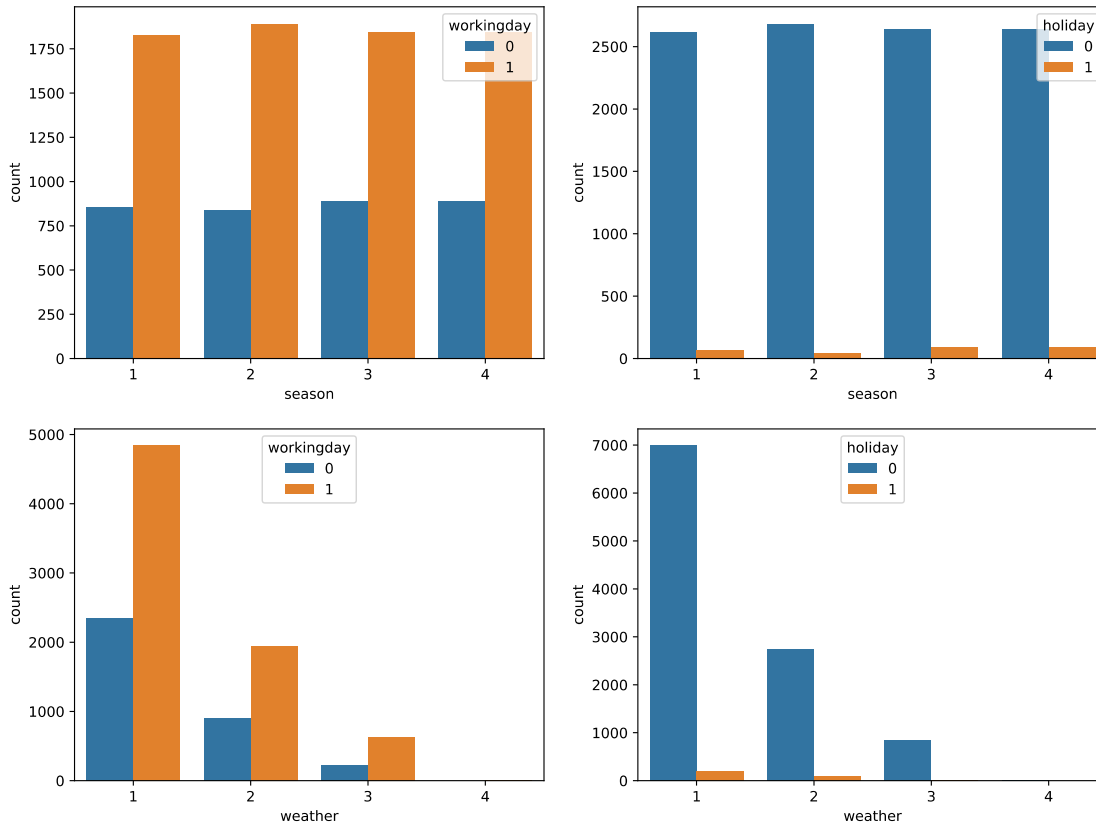
plt.figure(figsize=(13,10))
plt.subplot(221)
sns.barplot(x="weather", y="count", data=df_yulu, estimator=np.mean)
plt.subplot(222)
sns.barplot(x="season", y="count", data=df_yulu, estimator=np.mean)
plt.subplot(223)
sns.barplot(x="season", y="count", data=df_yulu, estimator=np.mean)
plt.subplot(224)
sns.barplot(x="workingday", y="count", data=df_yulu,
estimator=np.mean)

```

<AxesSubplot:xlabel='workingday', ylabel='count'>



```
plt.figure(figsize=(13,10))
plt.subplot(221)
sns.countplot(data=df_yulu, x="season", hue="workingday", dodge=True)
plt.subplot(222)
sns.countplot(data=df_yulu, x="season", hue="holiday", dodge=True)
plt.subplot(223)
sns.countplot(data=df_yulu, x="weather", hue="workingday", dodge=True)
plt.subplot(224)
sns.countplot(data=df_yulu, x="weather", hue="holiday", dodge=True)
<AxesSubplot:xlabel='weather', ylabel='count'>
```



*#Alpha is 0.05*

```
df_yulu_working = df_yulu[df_yulu["workingday"]==1]
```

```
df_yulu_Not_working = df_yulu[df_yulu["workingday"]==0]
```

```
ttest_ind(df_yulu_working["count"],df_yulu_Not_working["count"])
```

```
Ttest_indResult(statistic=1.2096277376026694,  
pvalue=0.22644804226361348)
```

```
df_yulu_season_a = df_yulu[df_yulu["season"] == 1]["count"]
```

```
df_yulu_season_b = df_yulu[df_yulu["season"] == 2]["count"]
```

```
df_yulu_season_c = df_yulu[df_yulu["season"] == 3]["count"]
```

```
df_yulu_season_d = df_yulu[df_yulu["season"] == 4]["count"]
```

*#H0 = season has no effect on the no of cycles rented*

```
f_oneway(df_yulu_season_a,df_yulu_season_b,df_yulu_season_c,df_yulu_season_d)
```

```
F_onewayResult(statistic=236.94671081032106,  
pvalue=6.164843386499654e-149)
```

```
e1 = ECDF(df_yulu_season_a)
```

```
e2 = ECDF(df_yulu_season_b)
```

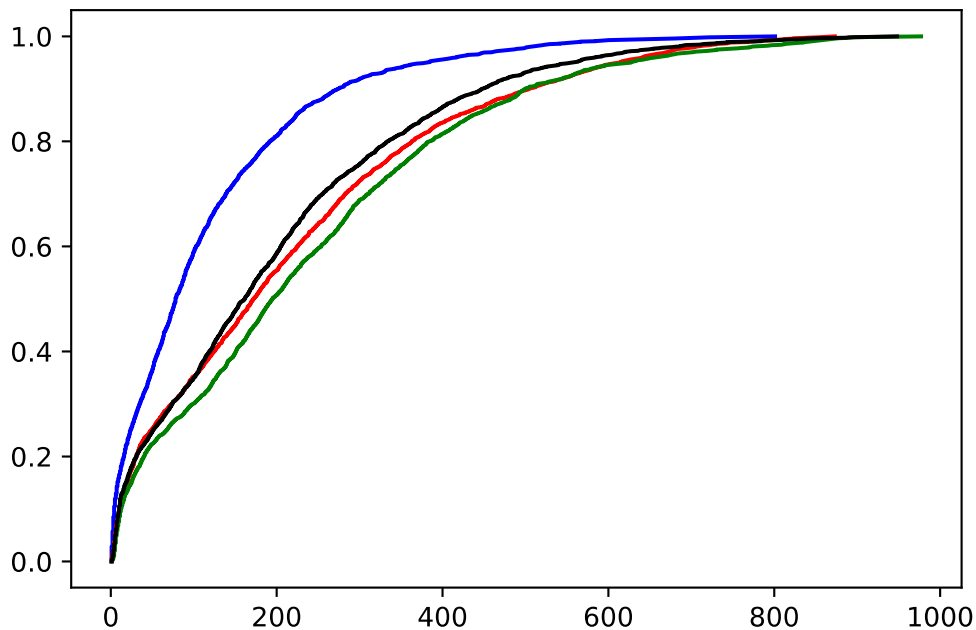
```
e3 = ECDF(df_yulu_season_c)
```

```
e4 = ECDF(df_yulu_season_d)
```



```
plt.plot(e1.x, e1.y,c='b')
plt.plot(e2.x, e2.y,c='r')
plt.plot(e3.x, e3.y,c='g')
plt.plot(e4.x, e4.y,c='k')

[<matplotlib.lines.Line2D at 0x7fafb233ed00>]
```



*#P-value is very low means we can reject null hypothesis and consider that seasons has an effect on the no of cycles rented*

```
df_yulu_weather_a = df_yulu[df_yulu["weather"] == 1]["count"]
df_yulu_weather_b = df_yulu[df_yulu["weather"] == 2]["count"]
df_yulu_weather_c = df_yulu[df_yulu["weather"] == 3]["count"]
df_yulu_weather_d = df_yulu[df_yulu["weather"] == 4]["count"]
```

*#H0 = weather has no effect on the no of cycles rented*

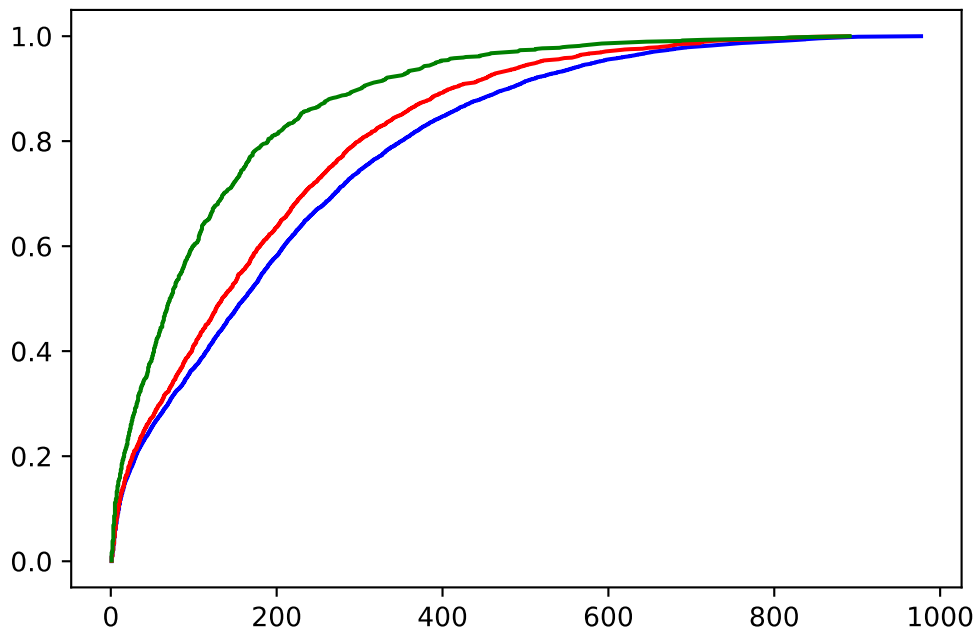
```
f_oneway(df_yulu_weather_a,df_yulu_weather_b,df_yulu_weather_c,df_yulu_weather_d)
```

```
F_onewayResult(statistic=65.53024112793271, pvalue=5.482069475935669e-42)
```

```
e1 = ECDF(df_yulu_weather_a)
e2 = ECDF(df_yulu_weather_b)
e3 = ECDF(df_yulu_weather_c)
e4 = ECDF(df_yulu_weather_d)
```

```
plt.plot(e1.x, e1.y,c='b')
plt.plot(e2.x, e2.y,c='r')
plt.plot(e3.x, e3.y,c='g')
plt.plot(e4.x, e4.y,c='k')
```

```
[<matplotlib.lines.Line2D at 0x7fafb1669be0>]
```



*#P-value is very low means we can reject null hypothesis and consider that weather has an effect on the no of cycles rented*

```
WS = pd.crosstab(index = df_yulu["weather"],columns =  
df_yulu["season"])
```

WS

season	1	2	3	4
weather				
1	1759	1801	1930	1702
2	715	708	604	807
3	211	224	199	225
4	1	0	0	0

*#H0 = weather and season are independent*

```
chi2_contingency(WS)
```

```
(49.15865559689363,  
1.5499250736864862e-07,  
9,  
array([[1.77454639e+03, 1.80559765e+03, 1.80559765e+03,  
1.80625831e+03],  
[6.99258130e+02, 7.11493845e+02, 7.11493845e+02,  
7.11754180e+02],  
[2.11948742e+02, 2.15657450e+02, 2.15657450e+02,  
2.15736359e+02],  
[2.46738931e-01, 2.51056403e-01, 2.51056403e-01, 2.51148264e-  
01]]))
```

*#P-value is very less so we can reject null hypothesis means weather and season are dependant*

```
plt.figure(figsize=(13,10))  
sns.heatmap(data=df_yulu.corr(),cmap="Blues",annot=True)
```

<AxesSubplot:>

