

```
In [1]: # Pandas for managing datasets
import pandas as pd

# Matplotlib for additional customization
from matplotlib import pyplot as plt

import matplotlib as mp

%matplotlib inline
#In Jupyter Notebook, you can also include %matplotlib inline
#to display your plots inside your notebook.

# Seaborn for plotting and styling
import seaborn as sns

import os
```

```
In [2]: # https://elitedatascience.com/wp-content/uploads/2017/04/Pokemon.csv

os.getcwd()
```

```
Out[2]: 'C:\\Users\\admin'
```

```
In [3]: os.chdir('C:\\Users\\admin\\pandas\\DataSets')
```

```
In [4]: df = pd.read_csv("Pokemon.csv")

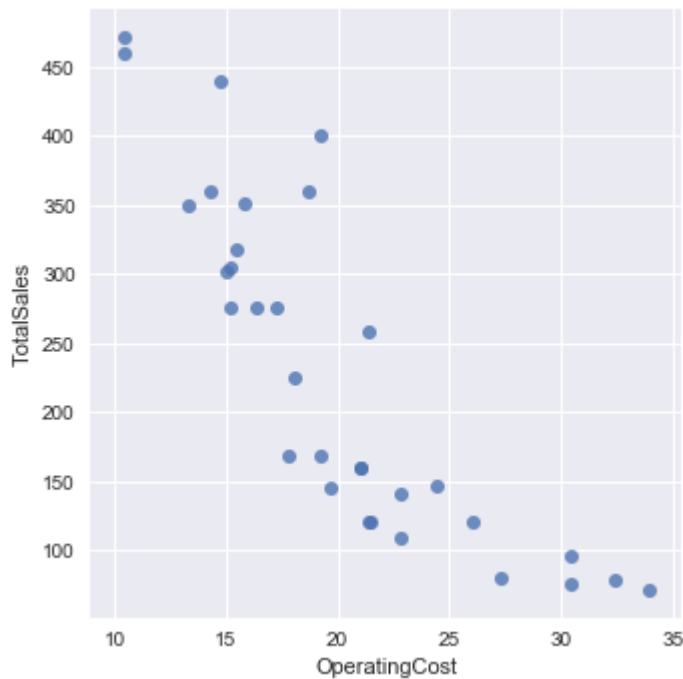
stores = pd.read_csv("stores.csv")
```

```
In [ ]: # Drawing a scatter plot
# Recommended way
sns.lmplot(x='Attack', y='Defense', data=df)

# Alternative way
# sns.lmplot(x=df.Attack, y=df.Defense)
```

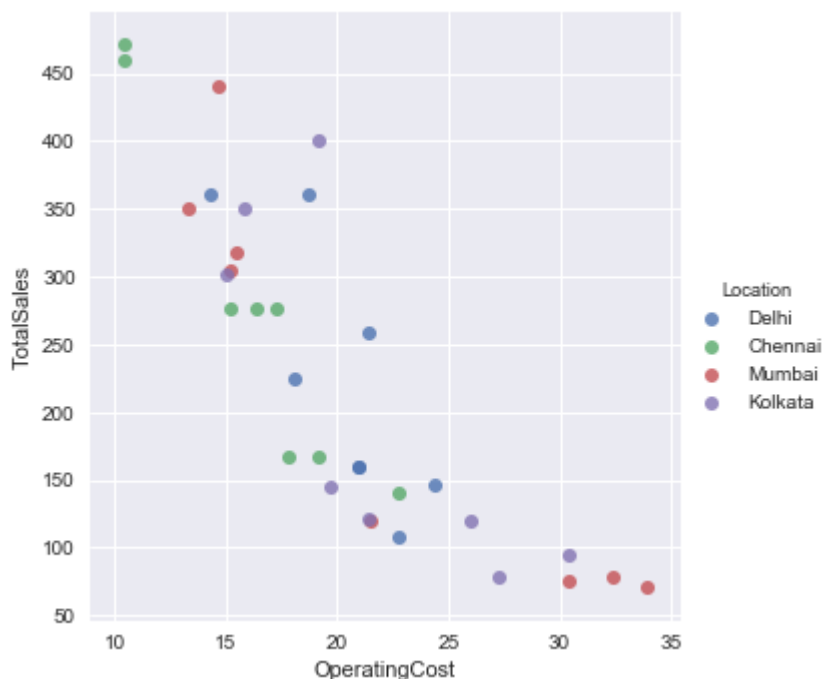
```
In [9]: # Scatter plot between TotalSales and OperatingCost
sns.lmplot(x = "OperatingCost", y = "TotalSales", data = stores, fit_reg=False)
# fit_reg = False
```

Out[9]: <seaborn.axisgrid.FacetGrid at 0x12781668>



```
In [10]: # Adding color to scatterplot
sns.lmplot(x = "OperatingCost", y = "TotalSales", data = stores, fit_reg=False, hue=
```

Out[10]: <seaborn.axisgrid.FacetGrid at 0x123abbe0>



```
In [ ]: sns.lmplot(x='Attack', y='Defense', data=df,
                  fit_reg=False, # No regression line
                  hue='Stage') # Color by evolution stage
```

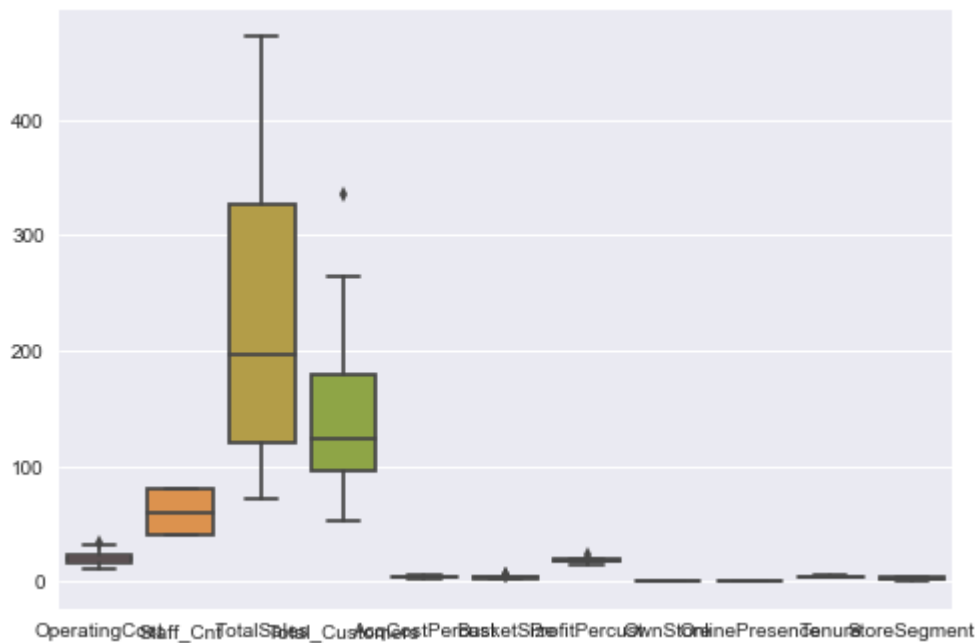
```
In [ ]: # Plot using Seaborn
sns.lmplot(x='Attack', y='Defense', data=df,
          fit_reg=False,
          hue='Stage')

# Tweak using Matplotlib
plt.ylim(0, 500)
plt.xlim(0, 300)
```

```
In [ ]: # Whiskers in Seaborn
```

```
In [12]: sns.boxplot(data=stores)
```

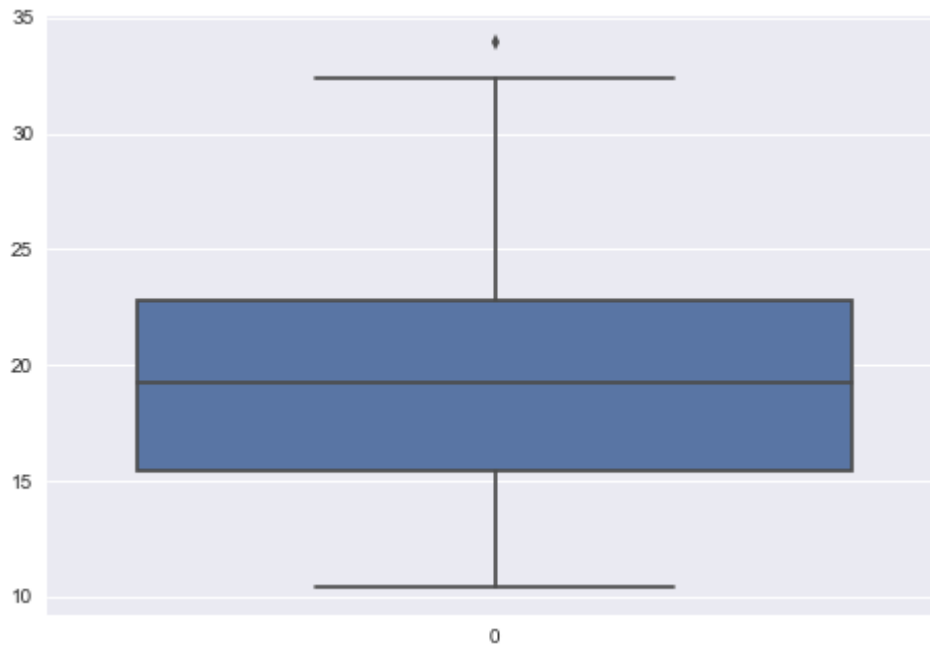
```
Out[12]: <matplotlib.axes._subplots.AxesSubplot at 0x13608dd8>
```



```
In [11]: sns.boxplot(data=stores.OperatingCost)
```

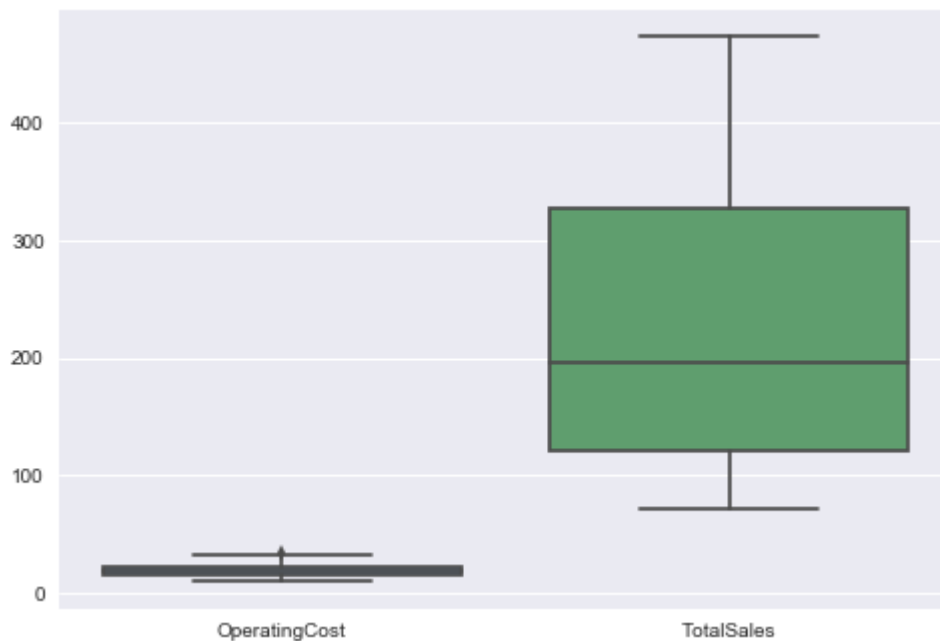
```
C:\Users\admin\Anaconda2\lib\site-packages\seaborn\categorical.py:454: FutureWarning: remove_na is deprecated and is a private function. Do not use.  
    box_data = remove_na(group_data)
```

```
Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x133777b8>
```



```
In [13]: sns.boxplot(data = stores.loc[:,["OperatingCost","TotalSales"]])
```

```
Out[13]: <matplotlib.axes._subplots.AxesSubplot at 0x13c4e898>
```



```
In [ ]: # Violin plots are useful alternatives to box plots.

# They show the distribution (through the thickness of the violin) instead of
# only the summary statistics.

# Set theme
sns.set_style('whitegrid')

# Violin plot
sns.violinplot(x='Location', y='TotalSales', data=stores)
```

```
In [ ]: col_Pallete = ["red","blue","green","yellow"]
sns.violinplot(x='Location', y='TotalSales', data=stores,palette=col_Pallete)
```

```
In [ ]: # color reference can be found at
# http://www.stat.columbia.edu/~tzheng/files/Rcolor.pdf

# Generating colors automatically

# Getting all color names so as to generate random colors for the graphs
colorDict = mp.colors.cnames
# from matplotlib package, get the colors as mentioned in the above pdf

colors = []
for i in colorDict:
    colors.append(i)

# colors variable is a list data structure of all available colors

# to get n random values from colors list
import random
c = random.sample(colors,4)
print c

sns.violinplot(x='Location', y='TotalSales', data=stores,palette=c)
```

```
In [ ]: # Violin plots are great for visualizing distributions.

# However, for a dataset, we may want to simply display each point.

# That's where the swarm plot comes in.

# This visualization will show each point, while "stacking" those with similar va

# Swarm plot with Pokemon color palette

sns.swarmplot(x='Location', y='TotalSales', data=stores,palette=c)
```

```
In [ ]: # Overlaying plots.

# It's pretty straightforward to overlay plots using Seaborn,
# and it works the same way as with Matplotlib.

# First, we'll make our figure larger using Matplotlib.
# Then, we'll plot the violin plot. However, we'll set inner=None to remove the box plot.
# Next, we'll plot the swarm plot. This time, we'll make the points black so they are visible.
# Finally, we'll set a title using Matplotlib.
```

```
In [ ]: # Set figure size with matplotlib
plt.figure(figsize=(10,6))
sns.violinplot(x='Location', y='TotalSales', data=stores,palette=c,alpha=0.1)
sns.swarmplot(x='Location', y='TotalSales', data=stores,color='k')

# Set title with matplotlib
plt.title('TotalSales by Location')
```

Heatmap

Heatmaps help you visualize matrix-like data.

```
In [ ]: stats_df = df.drop(['Total', 'Stage', 'Legendary'], axis=1)

# Calculate correlations
#corr = stats_df.corr()

# Heatmap
#sns.heatmap(corr)

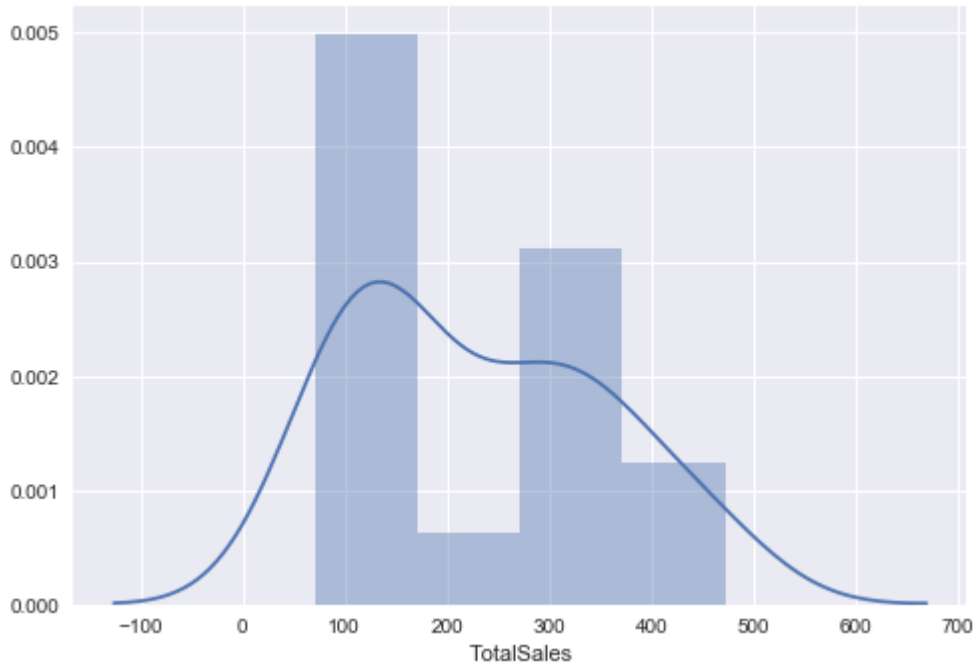
stores1 = stores.iloc[:,4:]
corrStores = stores1.corr()
sns.heatmap(corrStores)
```

Histogram

Histograms allow you to plot the distributions of numeric variables.

```
In [15]: # Distribution Plot (a.k.a. Histogram)
sns.distplot(stores.TotalSales)
```

```
Out[15]: <matplotlib.axes._subplots.AxesSubplot at 0x13e1a390>
```



Bar Plot

Bar plots help you visualize the distributions of categorical variables.

```
In [ ]: # Count Plot (a.k.a. Bar Plot)
sns.countplot(x='Type 1', data=df, palette=pkmn_type_colors)

# Rotate x-labels
plt.xticks(rotation=-45)
```

```
In [ ]: sns.countplot(x='StoreType', data=stores, palette=col_Pallete)

# Rotate x-labels
plt.xticks(rotation=-45)
```

Factor Plot

Factor plots make it easy to separate plots by categorical classes.

```
In [ ]: # Factor Plot
g = sns.factorplot(x='Type 1',
                  y='Attack',
                  data=df,
                  hue='Stage', # Color by stage
                  col='Stage', # Separate by stage
                  kind='swarm') # Swarmplot

# Rotate x-axis labels
g.set_xticklabels(rotation=-45)
```

```
In [ ]: g = sns.factorplot(x='Location',
                          y='TotalSales',
                          data=stores,
                          hue='Location', # Color by stage
                          col='StoreType', # Separate by stage
                          kind='bar') # barplot

# Rotate x-axis labels
g.set_xticklabels(rotation=-45)
```

Density Plot

Density plots display the distribution between two variables.

```
In [ ]: # Density Plot
sns.kdeplot(df.Attack, df.Defense)
```

```
In [ ]: sns.kdeplot(stores.TotalSales, stores.OperatingCost)
```

```
In [ ]:
```