

Hands-On-Activity 9.2: Customized Visualizations using Seaborn

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Section: CPE22S3

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Instructions:

- Create a Python notebook to answer all shown procedures, exercises and analysis in this section.

Resources:

- Download the following datasets: [fb_stock_prices_2018.csv](#) [earthquakes-1.csv](#)

Procedures:

- [9.4 Introduction to Seaborn](#)
- [9.5 Formatting Plots](#)
- [9.6 Customizing Visualizations](#)

Data Analysis:

- Provide comments on output from the procedures.

Supplementary Activity:

Using the CSV files provided and what we have learned so far in this module complete the following exercises:

1. Using seaborn, create a heatmap to visualize the correlation coefficients between earthquake magnitude and whether there was a tsunami with the magType of mb.
2. Create a box plot of Facebook volume traded and closing prices, and draw reference lines for the bounds of a Tukey fence with a multiplier of 1.5. The bounds will be at $Q1 - 1.5 * IQR$ and $Q3 + 1.5 * IQR$. Be sure to use the quantile() method on the data to make this easier. (Pick whichever orientation you prefer for the plot, but make sure to use subplots.)
3. Fill in the area between the bounds in the plot from exercise #2.
4. Use axes[0] to shade a rectangle from '2018-07-25' to '2018-07-31' which marks the large decline in Facebook price on a line plot of the closing price.
5. Using the Facebook stock price data, annotate the following three events on a line plot of the closing price:
 - Disappointing user growth announced after close on July 25, 2018
 - Cambridge Analytica story breaks on March 19, 2018 (when it affected the market)
 - FTC launches investigation on March 20, 2018
6. Modify the reg_resid_plot() function to use a matplotlib colormap instead of cycling between two colors. Remember, for this use case, we should pick a qualitative colormap or make our own.

Summary/Conclusion:

- Provide a summary of your learnings and the conclusion for this activity.

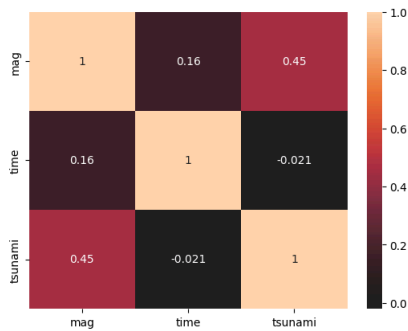
```
%matplotlib inline
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
fb = pd.read_csv('/content/fb_stock_prices_2018.csv' , index_col='date', parse_dates=True)
quakes = pd.read_csv('/content/earthquakes-1.csv')
```

```
# setup data visualization
```

```
quakesMb = quakes.query('magType == "mb"')
numeric_columns = quakesMb.select_dtypes(include=['number'])
sns.heatmap(numeric_columns.corr(), annot=True, center=0)
# now we create a heatmap to visualize the data with the correlation of the magnitude and tsunami, magType as mb
```

```
<Axes: >
```



```
# box plot of Facebook's closing and volume traded
```

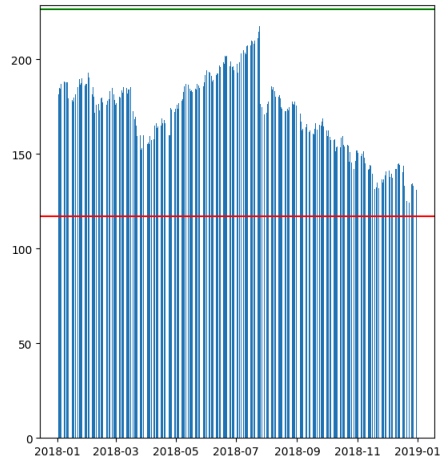
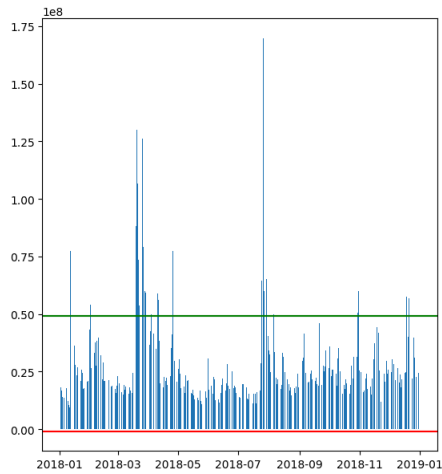
```
iqrVolume = fb['volume'].quantile(0.75) - fb['volume'].quantile(0.25)
iqrClose = fb['close'].quantile(0.75) - fb['close'].quantile(0.25)
```

```
tukeyLowerVolume = fb['volume'].quantile(0.25) - 1.5 * iqrVolume
tukeyUpperVolume = fb['volume'].quantile(0.75) + 1.5 * iqrVolume
tukeyLowerClose = fb['close'].quantile(0.25) - 1.5 * iqrClose
tukeyUpperClose = fb['close'].quantile(0.75) + 1.5 * iqrClose
```

```
fig, ax = plt.subplots(1, 2, figsize=(14, 7))
ax[0].bar(fb.index, fb.volume)
ax[0].axhline(tukeyLowerVolume, color='r')
ax[0].axhline(tukeyUpperVolume, color='g')
```

```
ax[1].bar(fb.index, fb.close)
ax[1].axhline(tukeyLowerClose, color='r')
ax[1].axhline(tukeyUpperClose, color='g')
```

 <matplotlib.lines.Line2D at 0x7a4587d94070>



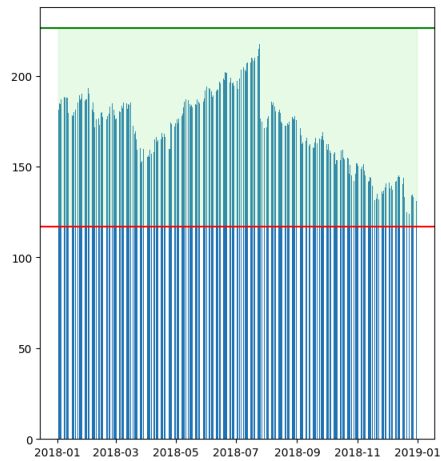
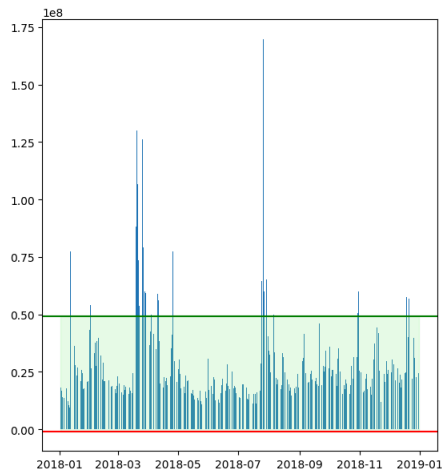
```
# Filling the area of the bounds of the above graph.
# Creating a box plot of Facebook volume traded and closing prices.
iqrVolume = fb['volume'].quantile(0.75) - fb['volume'].quantile(0.25)
iqrClose = fb['close'].quantile(0.75) - fb['close'].quantile(0.25)

tukeyLowerVolume = fb['volume'].quantile(0.25) - 1.5 * iqrVolume
tukeyUpperVolume = fb['volume'].quantile(0.75) + 1.5 * iqrVolume
tukeyLowerClose = fb['close'].quantile(0.25) - 1.5 * iqrClose
tukeyUpperClose = fb['close'].quantile(0.75) + 1.5 * iqrClose

fig, ax = plt.subplots(1, 2, figsize=(14, 7))
ax[0].bar(fb.index, fb.volume)
ax[0].axhline(tukeyLowerVolume, color='r')
ax[0].axhline(tukeyUpperVolume, color='g')
ax[0].fill_between(fb.index, tukeyLowerVolume, tukeyUpperVolume, color='lightgreen', alpha=0.2)

ax[1].bar(fb.index, fb.close)
ax[1].axhline(tukeyLowerClose, color='r')
ax[1].axhline(tukeyUpperClose, color='g')
ax[1].fill_between(fb.index, tukeyLowerClose, tukeyUpperClose, color='lightgreen', alpha=0.2)
```

 <matplotlib.collections.PolyCollection at 0x7a458730f760>



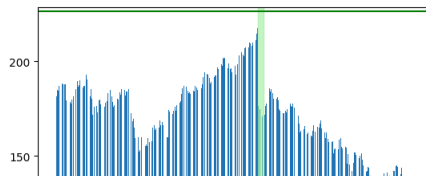
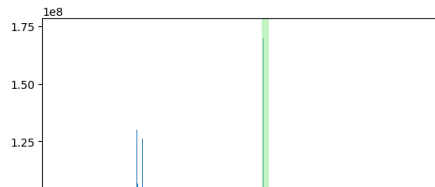
```
# Using axvspan() to shade a rectangle which marks the decline of Facebook price.
iqrVolume = fb['volume'].quantile(0.75) - fb['volume'].quantile(0.25)
iqrClose = fb['close'].quantile(0.75) - fb['close'].quantile(0.25)

tukeyLowerVolume = fb['volume'].quantile(0.25) - 1.5 * iqrVolume
tukeyUpperVolume = fb['volume'].quantile(0.75) + 1.5 * iqrVolume
tukeyLowerClose = fb['close'].quantile(0.25) - 1.5 * iqrClose
tukeyUpperClose = fb['close'].quantile(0.75) + 1.5 * iqrClose

fig, ax = plt.subplots(1, 2, figsize=(14, 7))
ax[0].bar(fb.index, fb.volume)
ax[0].axhline(tukeyLowerVolume, color='r')
ax[0].axhline(tukeyUpperVolume, color='g')
ax[0].axvspan('2018-07-25', '2018-07-31', color='lightgreen', alpha=0.5)

ax[1].bar(fb.index, fb.close)
ax[1].axhline(tukeyLowerClose, color='r')
ax[1].axhline(tukeyUpperClose, color='g')
ax[1].axvspan('2018-07-25', '2018-07-31', color='lightgreen', alpha=0.5)
```

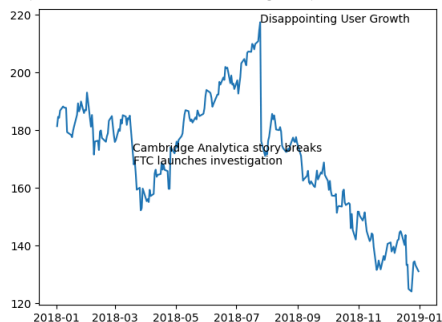
```
<matplotlib.patches.Polygon at 0x7a458720e500>
```



```
# Using the Facebook Stock Price dataset for annotating significant events.
import matplotlib.dates as plt_date
```

```
plt.plot(fb.close)
plt.annotate('Disappointing User Growth', xy=(plt_date.date2num(pd.to_datetime('2018-07-25')), fb.loc['2018-07-25', 'close']))
plt.annotate('Cambridge Analytica story breaks', xy=(plt_date.date2num(pd.to_datetime('2018-03-19')), fb.loc['2018-03-19', 'close']))
plt.annotate('FTC launches investigation', xy=(plt_date.date2num(pd.to_datetime('2018-03-20')), fb.loc['2018-03-20', 'close']))
```

```
<matplotlib.text.Text at 0x7a458720e500>
```



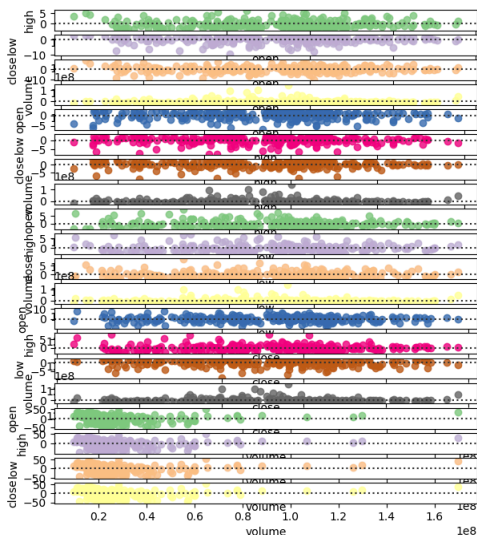
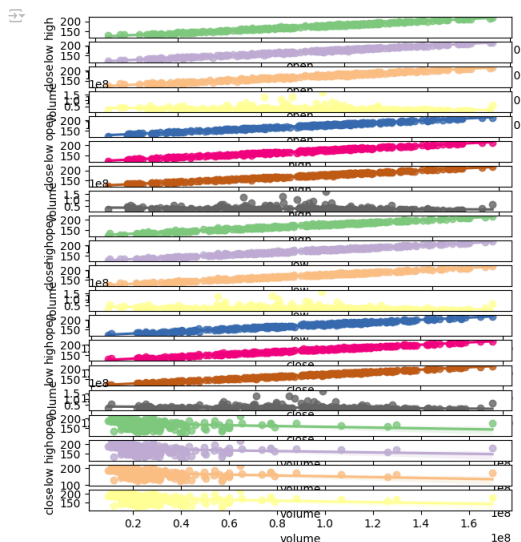
```
# Modifying the reg_red_plots() function instead of cycling between two colors.
import itertools
import matplotlib.pyplot as plt
import seaborn as sns
```

```
def reg_resid_plots(data):
    num_cols = data.shape[1]
    permutation_count = num_cols*(num_cols - 1)
    fig, ax = plt.subplots(permutation_count, 2, figsize=(15, 8))
    colormap = plt.colormaps['Accent']

    for i, (x,y), axes in zip(range(permutation_count), itertools.permutations(data.columns, 2), ax):
        color = colormap(i%colormap.N)
        for subplot, func in zip(axes, (sns.regplot, sns.residplot)):
            func(data=data, x=x, y=y, color=color, ax=subplot)
```

```
plt.close()
return fig
```

```
reg_resid_plots(fb)
```



Summary/Conclusion:

It is a no-brainer to visualize data when it comes to data science, since this increases the efficiency of gathering and annotating/interpreting data. We already have pandas, numpy, and matplotlib for data gathering, wrangling, as well as visualization, Seaborn is another option for us to visualize data in a more interactive and 'fun' way with the heatmaps as well as the reg plots.