Week4

September 14, 2020

1 Pandas Visualization

```
In [ ]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        %matplotlib notebook
In []: # see the pre-defined styles provided.
        plt.style.available
In [ ]: # use the 'seaborn-colorblind' style
        plt.style.use('seaborn-colorblind')
1.0.1 DataFrame.plot
In [ ]: np.random.seed(123)
        df = pd.DataFrame({'A': np.random.randn(365).cumsum(0),
                            'B': np.random.randn(365).cumsum(0) + 20,
                            'C': np.random.randn(365).cumsum(0) - 20},
                           index=pd.date_range('1/1/2017', periods=365))
        df.head()
In []: df.plot(); # add a semi-colon to the end of the plotting call to suppress a
  We can select which plot we want to use by passing it into the 'kind' parameter.
In [ ]: df.plot('A', 'B', kind = 'scatter');
```

You can also choose the plot kind by using the DataFrame.plot.kind methods instead of

providing the kind keyword argument.

```
kind:-'line': line plot (default) - 'bar': vertical bar plot - 'barh': horizontal bar plot
- 'hist': histogram - 'box': boxplot - 'kde': Kernel Density Estimation plot - 'density'
: same as 'kde' - 'area': area plot - 'pie': pie plot - 'scatter': scatter plot - 'hexbin':
hexbin plot
```

```
In []: ax = df.plot.scatter('A', 'C', c='B', s=df['B'], colormap='viridis')
          ax.set_aspect('equal')
In []: df.plot.box();
In []: df.plot.hist(alpha=0.7);
```

Kernel density estimation plots are useful for deriving a smooth continuous function from a given sample.

```
In [ ]: df.plot.kde();
```

1.0.2 pandas.tools.plotting

Iris flower data set

2 Seaborn

```
In [ ]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        %matplotlib notebook
In [ ]: np.random.seed(1234)
        v1 = pd.Series(np.random.normal(0,10,1000), name='v1')
        v2 = pd.Series(2*v1 + np.random.normal(60, 15, 1000), name='v2')
In [ ]: plt.figure()
        plt.hist(v1, alpha=0.7, bins=np.arange(-50,150,5), label='v1');
        plt.hist(v2, alpha=0.7, bins=np.arange(-50, 150, 5), label='v2');
        plt.legend();
In []: # plot a kernel density estimation over a stacked barchart
        plt.figure()
        plt.hist([v1, v2], histtype='barstacked', normed=True);
        v3 = np.concatenate((v1, v2))
        sns.kdeplot(v3);
```

```
In [ ]: plt.figure()
        # we can pass keyword arguments for each individual component of the plot
        sns.distplot(v3, hist_kws={'color': 'Teal'}, kde_kws={'color': 'Navy'});
In [ ]: sns.jointplot(v1, v2, alpha=0.4);
In [ ]: grid = sns.jointplot(v1, v2, alpha=0.4);
        grid.ax_joint.set_aspect('equal')
In [ ]: sns.jointplot(v1, v2, kind='hex');
In [ ]: # set the seaborn style for all the following plots
        sns.set_style('white')
        sns.jointplot(v1, v2, kind='kde', space=0);
In [ ]: iris = pd.read_csv('iris.csv')
        iris.head()
In []: sns.pairplot(iris, hue='Name', diag_kind='kde', size=2);
In [ ]: plt.figure(figsize=(8,6))
       plt.subplot(121)
        sns.swarmplot('Name', 'PetalLength', data=iris);
       plt.subplot(122)
        sns.violinplot('Name', 'PetalLength', data=iris);
```