

Seaborn

```
import seaborn as sns
```

```
#getting seaborn available datasets
```

```
sns.get_dataset_names()
```

```
data = sns.load_dataset('dataset_name')
```

Seaborn Plots

Categorical	Distribution	Relational	Regression	Matrix
countplot	histplot	scatterplot	regplot	heatmap
barplot	kdeplot	lineplot		
boxplot	rugplot	relplot		
violinplot	ecdfplot			
stripplot	displot			
swarmplot	jointplot			
catplot	pairplot			

Note:

Hue should be only a categorical variable.

Data specifies dataframe.

x should be usually categoric variable

y should be numeric variable

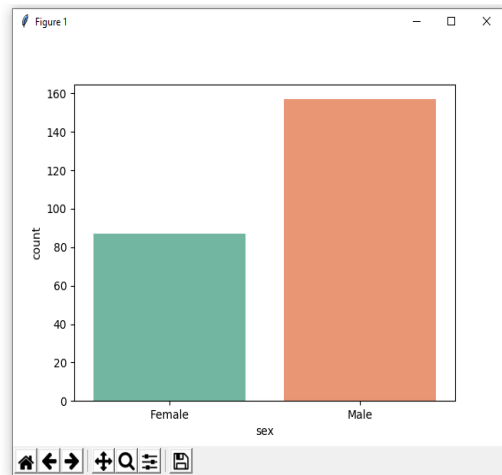
Changing the order might change the orientation. (i.e vertical to horizontal)

x, y need not be provided at the same time. They can be used alone.

Press shift + tab to get function details

Press tab for auto completion

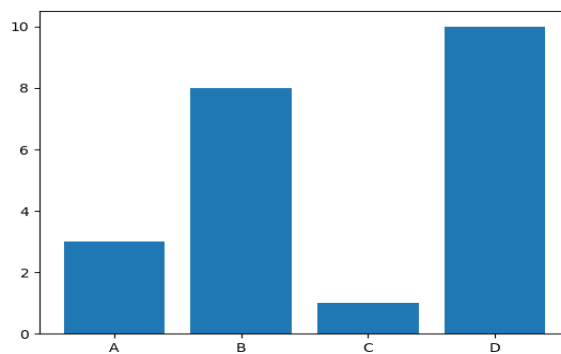
1) Count Plot



vertical plot
`sns.countplot(x, data, palette, hue)`

horizontal plot
`sns.countplot(y, data)`

2) Bar Plot



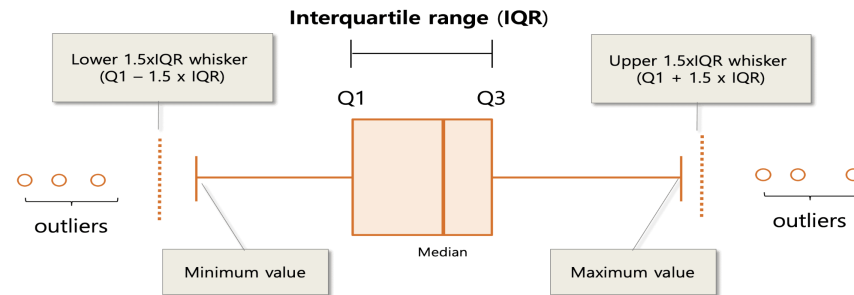
`sns.barplot(x, y, data, hue, order, hue_order, color, palette, errorbar=None, estimator=len)`

order - list of columns by which the plot should be plotted.

hue_order - list of values for hue

estimator - np.mean (default), np.median, np.sum, len (actual value)

3) Box Plot



Note:

*Box plot cannot be plotted for categorical variables alone. But it can be plotted for numerical alone

For Categorical - x

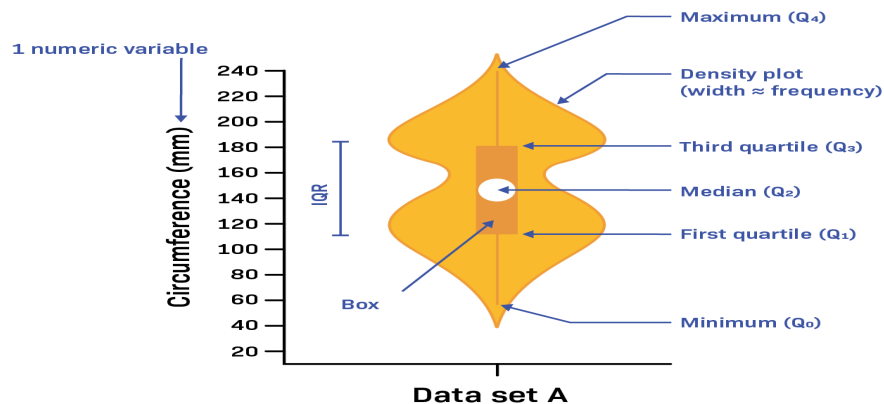
For Numerical - y

```
sns.boxplot(x, y, data, linewidth, hue, showmeans=True,
            meanprops={'marker':'o', 'markerfacecolor':'white',
                      'markeredgecolor':'black', 'markersize':5},
            palette
)
```

Box plot for all numerical variables of the dataframe

```
sns.boxplot(data)
```

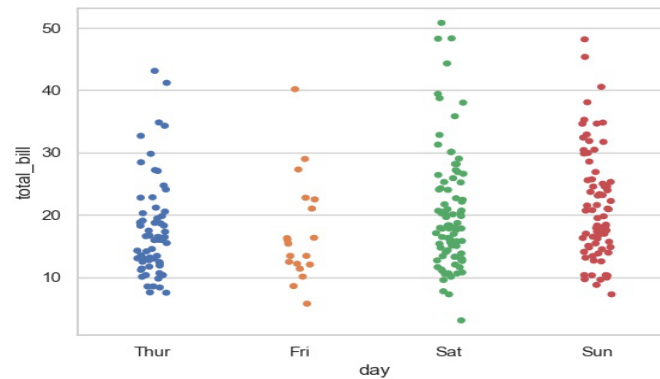
4) Violin Plot



Same note as that of boxplot.

```
sns.violinplot(x, y, data, hue, order, hue_order, color, palette, split=True, bw)
```

5) Strip Plot



Strip plot can be plotted for both numeric and categorical variables.

```
sns.stripplot(x, y, data, jitter=0.2, linewidth=0.8, hue, dodge=True, color, palette)
```

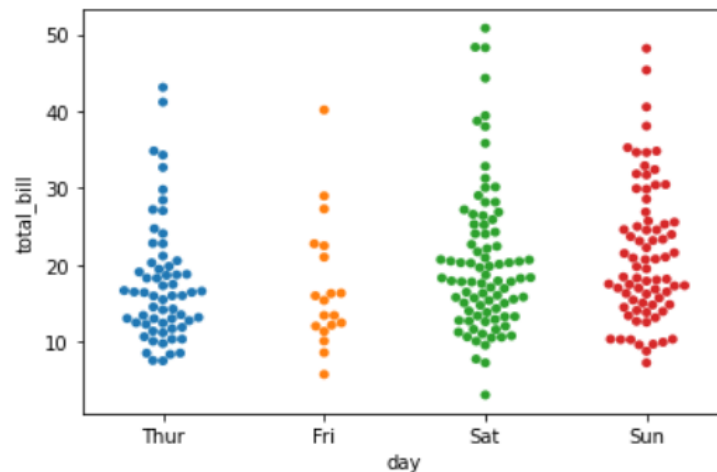
jitter - provides space, dodge=True - separates hue

Drawing strip plot on top of violin plot (also used for box plot)

```
sns.stripplot(x='Payment', y='Total', data=data, jitter=0.3, palette='Pastel1')
```

```
sns.violinplot(x='Payment', y='Total', data=data)
```

6) Swarm Plot



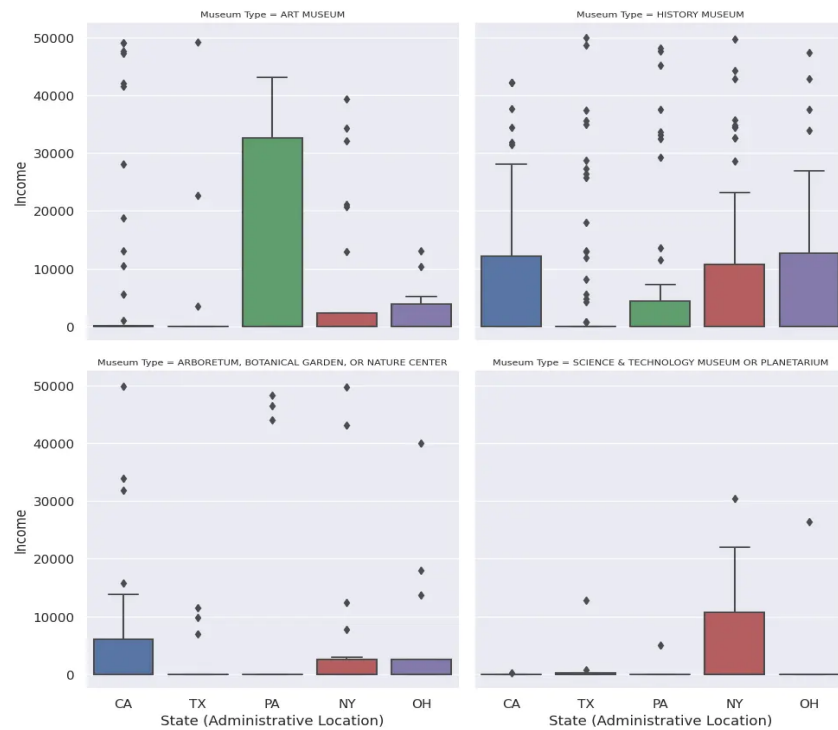
```
sns.swarmplot(x, y, data, hue, dodge=True, color, palette, marker, size)
```

Drawing swarm plot on top of violin plot (also used for box plot)

```
sns.swarmplot(x='Payment', y='Total', data=data, palette='Pastel1')
```

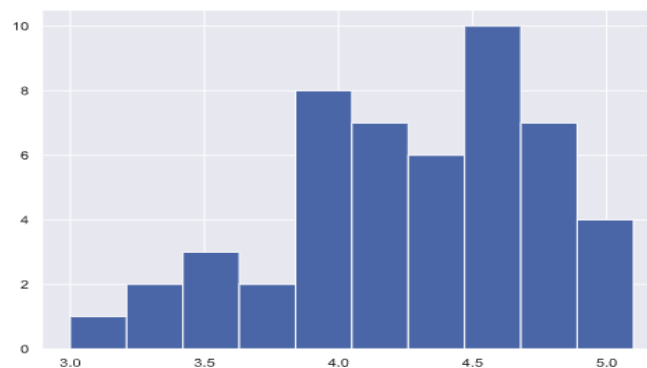
```
sns.violinplot(x='Payment', y='Total', data=data)
```

7) Cat Plot



`sns.catplot(x, y, data, kind='bar/violin/box/strip/swarm', hue, row, col, palette)`

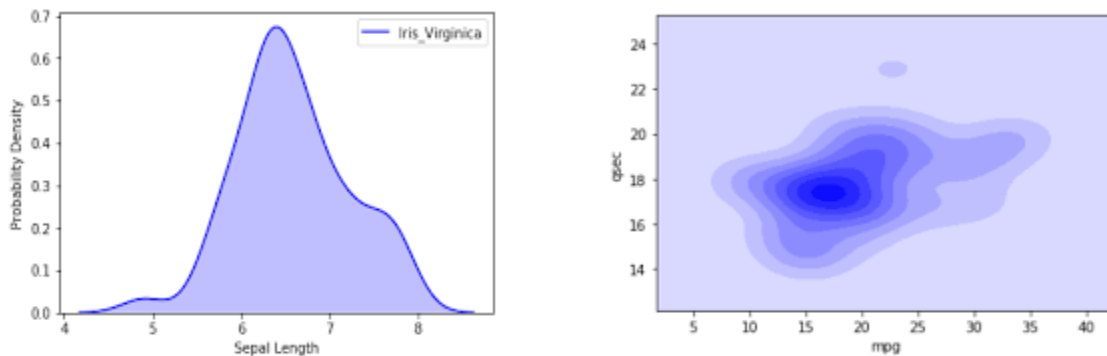
8) Histogram



Can be uni as well as bivariate.

`sns.histplot(x, y, data, binwidth=10, bins=value/list, kde=True, hue, color, palette,
multiple='stack',
element='step'/ 'poly',
fill=False,
stat= 'count' (default) /density/ probability/ frequency/ percent)`

9) KDE Plot (Kernel Density Estimation)



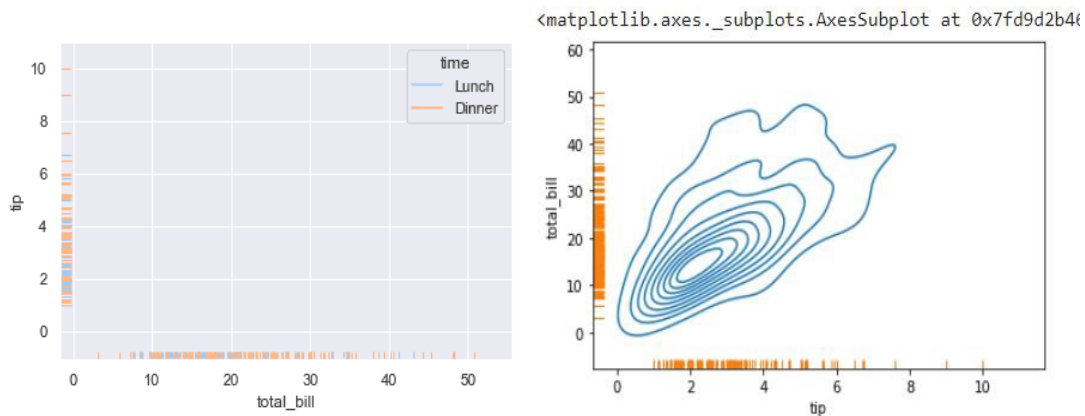
Can be uni as well as bivariate.

```
sns.kdeplot(x, y, data, fill=True, bw_adjust=0.2, hue, multiple='stack', color, palette, alpha, levels)
```

levels - specify the number of contours.

```
# kde plot for all numerical variables of dataframe
sns.kdeplot(data)
```

10) Rug Plot



Can be uni/bi variate. Can be used for both categorical and numeric.

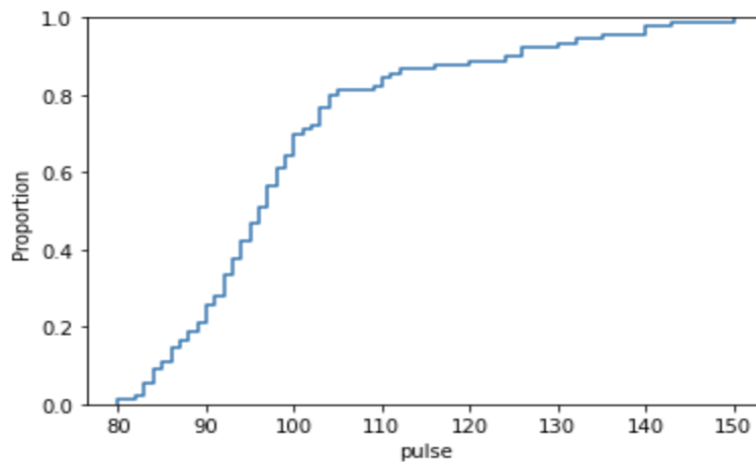
```
sns.rugplot(x, y, data, hue, height=0.1, color)
```

```
#Combining with kde plot
```

```
sns.rugplot(x='gross income', y='Quantity', data=data, height=0.05)
```

```
sns.kdeplot(x='gross income', y='Quantity', data=data, fill=True, color='purple')
```

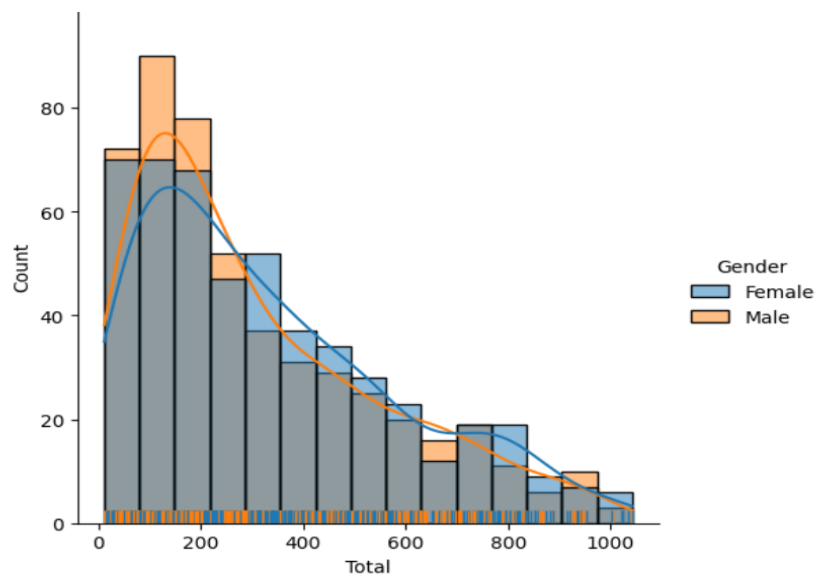
11) ECDF Plot (Empirical Cumulative Distribution Function)



It is a univariate plot. Can be used for both categoric and numeric.

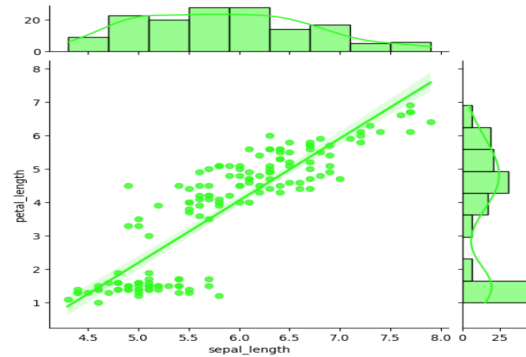
```
sns.ecdfplot(x, data, hue, color, palette, stat='proportion/count/percent/density')
```

12) Displot



```
sns.displot(x, data, kde=True, rug=True, hue, multiple='stack', element='poly', row, col,  
            color, palette  
            rug_kws=dict(height=0.1),  
            kde_kws=dict(bw_adjust=0.12))
```

13) Joint Plot



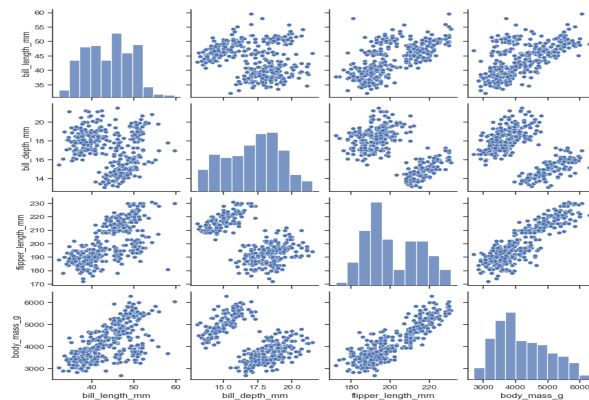
Bivariate plot

```
sns.jointplot(x, y, data, kind='scatter/hex/kde/resid/hist', hue, marginal_ticks=True, height=3,
              joint_kws=dict(marker='*', color='red'),
              marginal_kws=dict(color='pink', element='poly'))
```

drawing kde and rugplot on top of joint plot

```
pl = sns.jointplot(x='petal_length', y='petal_width', data=data, color='#BC2E12', height=4)
pl.plot_joint(sns.kdeplot, color='pink')
pl.plot_joint(sns.rugplot, height=0.1, color='green')
```

14) Pair Plot

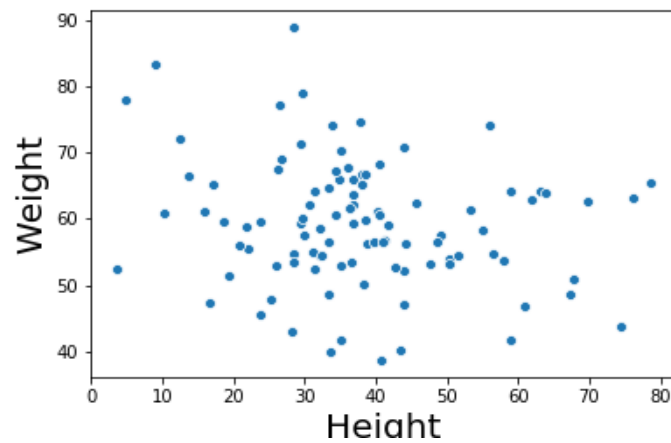


```
sns.pairplot(data, diag_kind='kde/hist/None', kind='scatter/reg/kde/hist', hue, color, palette,
             x_vars=value/list, y_vars=value/list, corner=True,
             diag_kws=dict(kde=True, color='#16FF00'),
             plot_kws=dict(color='#060047', marker='D', s=5))
```

Creating kde plot on top of pairplot

```
pl = sns.pairplot(data=data, plot_kws=dict(color='red'), diag_kws=dict(element='poly',
color='pink'))
pl.map_upper(sns.kdeplot) # For only upper
pl.map_lower(sns.kdeplot, fill=True) # For only lower
```

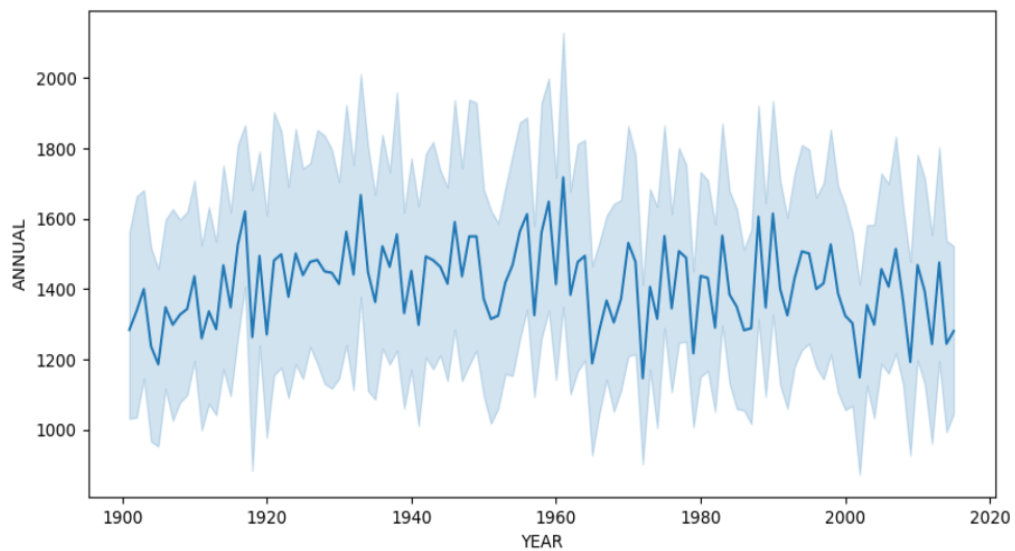

15) Scatter Plot



Bivariate plot

```
sns.scatterplot(data=data, x, y, hue, palette, s=200, edgecolor='black')
```

16) Line Plot

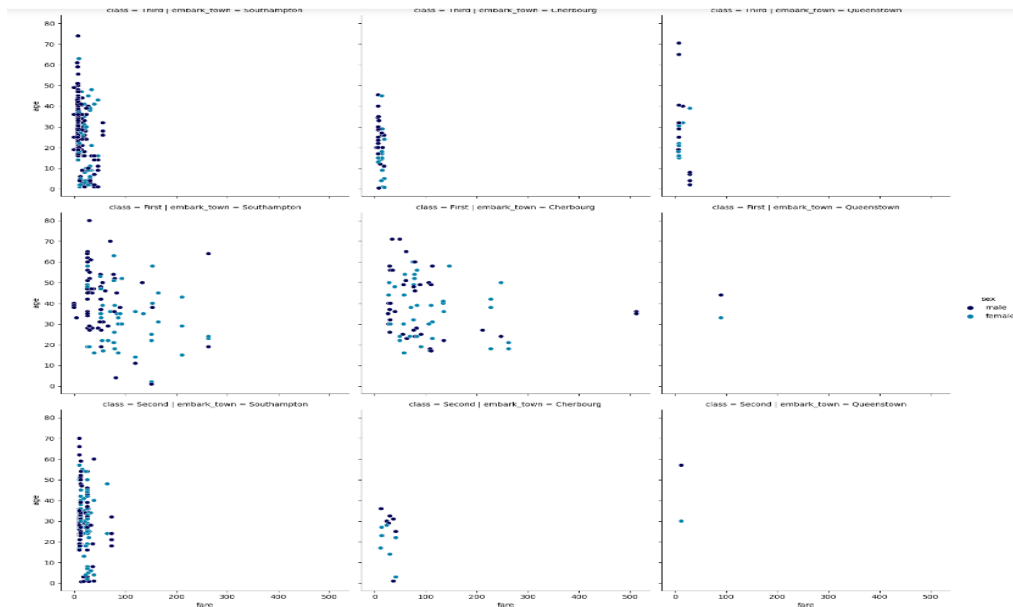


Bivariate Plot

```
sns.lineplot(data, x, y, errorbar=None, hue, color, palette, estimator='mean/sum/None')
```

None = actual value

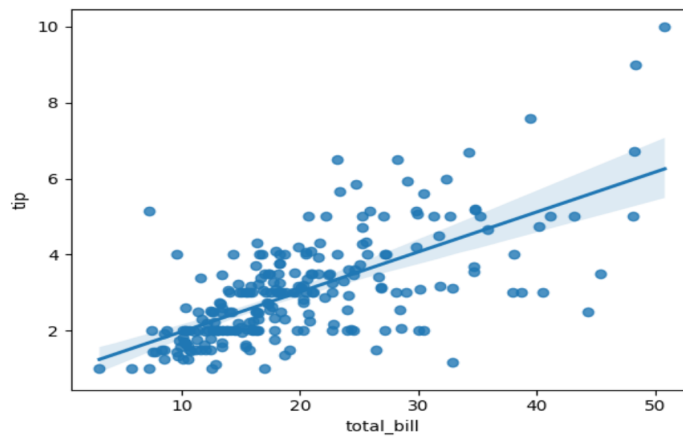
17) Relational Plot



Bivariate, both must be numerical only

`sns.relplot(data, x, y, kind='line/scatter', errorbar=None, hue, row, col, col_wrap)`

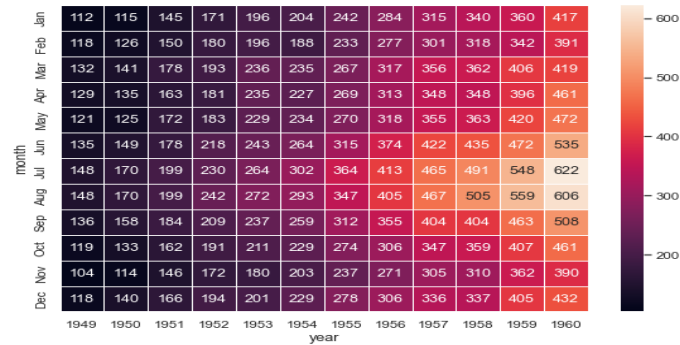
18) Regression Plot



Bivariate Plot.

```
sns.regplot(data, x, y, color, marker, ci=None,
            scatter_kws=dict(color='#5800FF', s=100, alpha=0.5),
            line_kws=dict(color='#FF0060', linestyle='--'))
```

19) HeatMap



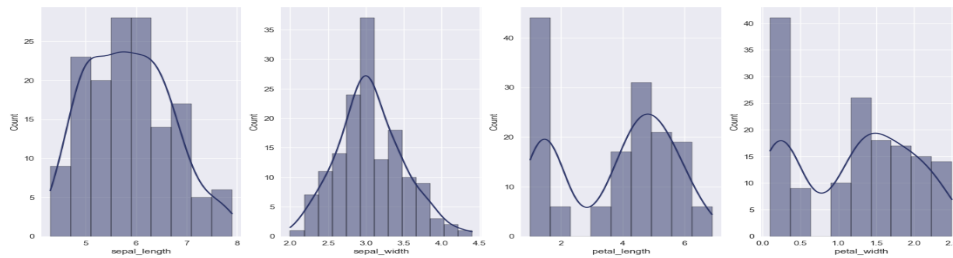
Note:

For heatmap, all columns in the dataframe should be numeric.

```
sns.heatmap(data, annot=True, fmt='.0f', linewidth=0.5, linecolor='white', cmap,
            annot_kws=dict(size=15, weight='bold'))
```

```
sns.heatmap(data.corr(), vmin=-1, vmax=1, center=0, cmap)
```

20) Subplots



Basic Syntax

```
fig, ax = plt.subplots(n_rows, n_cols, figsize=(x, y))
```

1D

```
fig, ax = plt.subplots(1, 2, figsize=(x, y))
sns.barplot(ax=ax[0], x, y, data)
sns.histplot(ax=ax[1], x, y, data)
```

1D

```
fig, ax = plt.subplots(2, 1, figsize=(x, y))
sns.barplot(ax=ax[0], x, y, data)
sns.histplot(ax=ax[1], x, y, data)
```

2D

```
fig, ax = plt.subplots(2, 2, figsize=(x, y))
sns.barplot(ax=ax[0, 0], x, y, data)
sns.histplot(ax=ax[0, 1], x, y, data)
sns.barplot(ax=ax[1, 0], x, y, data)
sns.histplot(ax=ax[1, 1], x, y, data)
```

