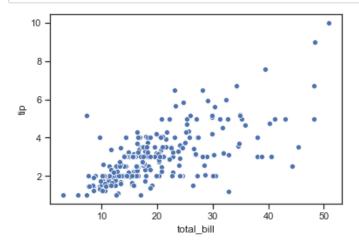
Seaborn - Visualizing statistical relationships

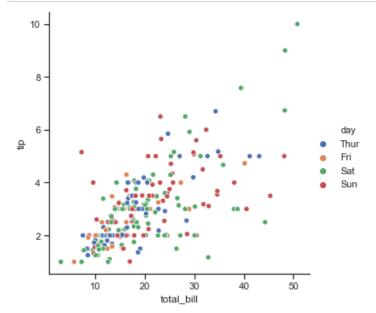
• relplot() - scatter and line plots

```
import numpy as np
In [1]:
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]: sns.set(style="ticks")
In [3]: tips = sns.load_dataset("tips")
         tips.head()
Out[3]:
            total_bill
                                             time size
                     tip
                            sex smoker day
          0
               16.99 1.01 Female
                                                     2
                                    No
                                        Sun Dinner
          1
               10.34 1.66
                                                     3
                           Male
                                    No Sun Dinner
          2
               21.01 3.50
                                    No Sun Dinner
                                                     3
                           Male
               23.68 3.31
                           Male
                                        Sun Dinner
                                                     2
                                    No
                                                     4
               24.59 3.61 Female
                                    No Sun Dinner
In [4]: # default kind="scatter"
         sns.relplot(x="total_bill", y="tip", data=tips);
            10
             8
             6
          ф
            4
             2
                                                   50
                                            40
                                    30
                                total_bill
```

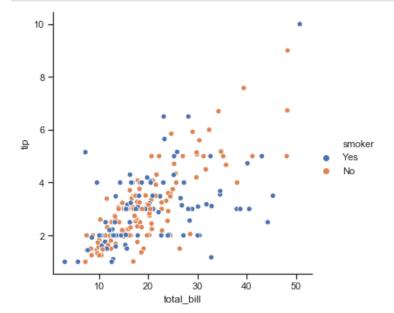
```
In [5]: sns.scatterplot(x="total_bill", y="tip", data=tips);
```

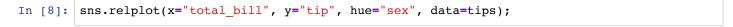


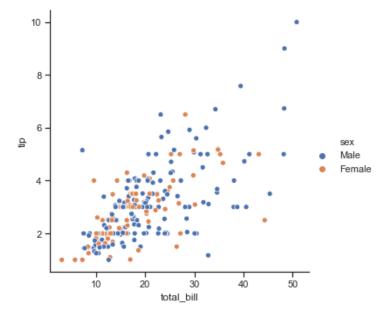
In [6]: # Add a third dimension - color by a third variable (Group by)
hue semantic - color of a point now has a meaning
sns.relplot(x="total_bill", y="tip", hue="day", data=tips);



```
In [7]: sns.relplot(x="total_bill", y="tip", hue="smoker", data=tips);
```

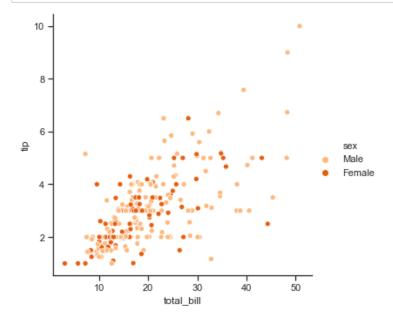


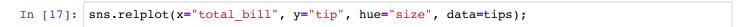


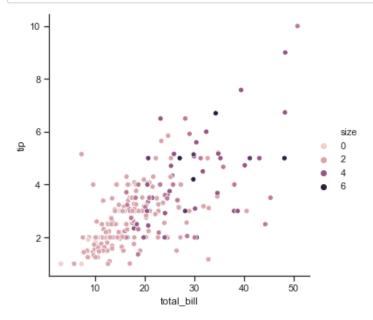


In [9]: # hue semantic categorical in above examples (qualitative palette)
If hue semantic is numeric, sequential palette is used for color

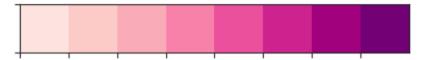
```
In [10]: current_palette = sns.color_palette()
         sns.palplot(current palette)
In [11]: sns.choose colorbrewer palette(data type="qualitative")
               name
                     Set1
                                         9
               desat
                                        1.00
Out[11]: [(0.8941176470588235, 0.10196078431372557, 0.10980392156862737),
          (0.21568627450980393, 0.4941176470588236, 0.7215686274509804),
          (0.3019607843137256, 0.6862745098039216, 0.29019607843137263),
          (0.5960784313725492, 0.3058823529411765, 0.6392156862745098),
          (1.0, 0.4980392156862745, 0.0),
          (0.99999999999998, 1.0, 0.19999999999999),
          (0.6509803921568629, 0.33725490196078434, 0.1568627450980391),
          (0.9686274509803922, 0.5058823529411766, 0.7490196078431374),
          (0.6, 0.6, 0.6)]
In [12]: colors = sns.choose_colorbrewer_palette(data_type="sequential")
               name
                     Oranges
                                         2
                                        1.00
               desat
              variant
                     regular
In [15]: colors
Out[15]: [(0.9921568627450981, 0.726797385620915, 0.49150326797385624),
          (0.9137254901960783, 0.36862745098039207, 0.0509803921568629)]
```

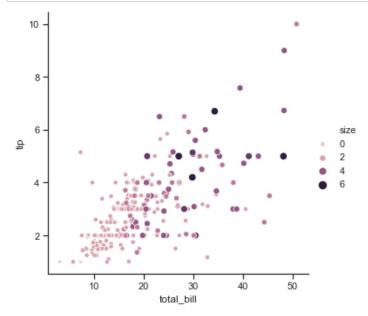


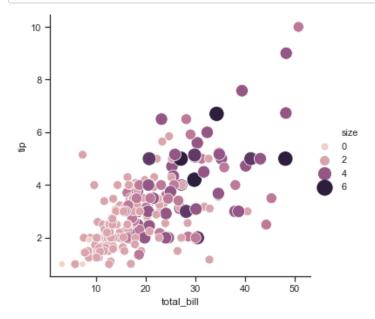


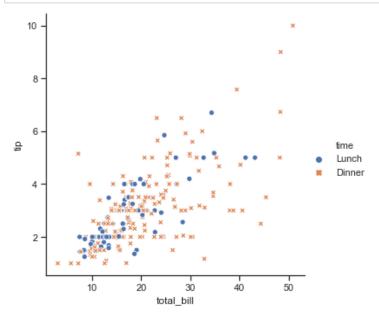


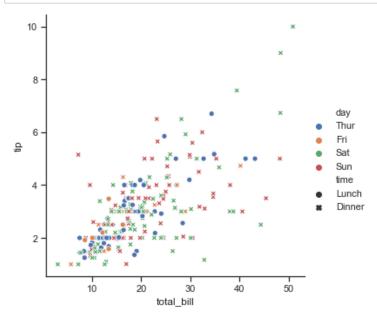
```
In [18]: sequential_colors = sns.color_palette("RdPu", 8)
sns.palplot(sequential_colors)
```

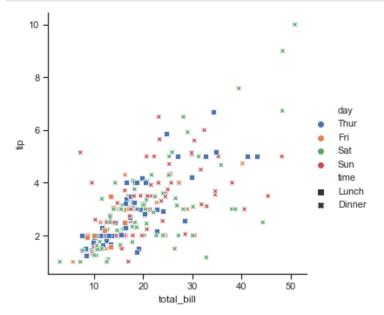






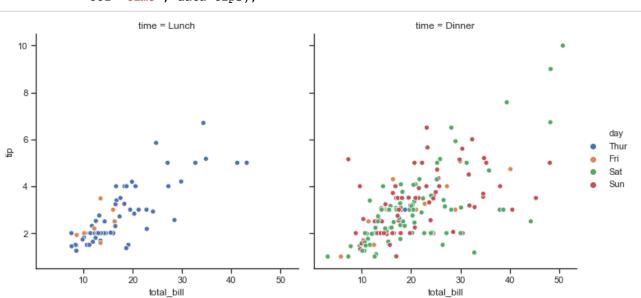


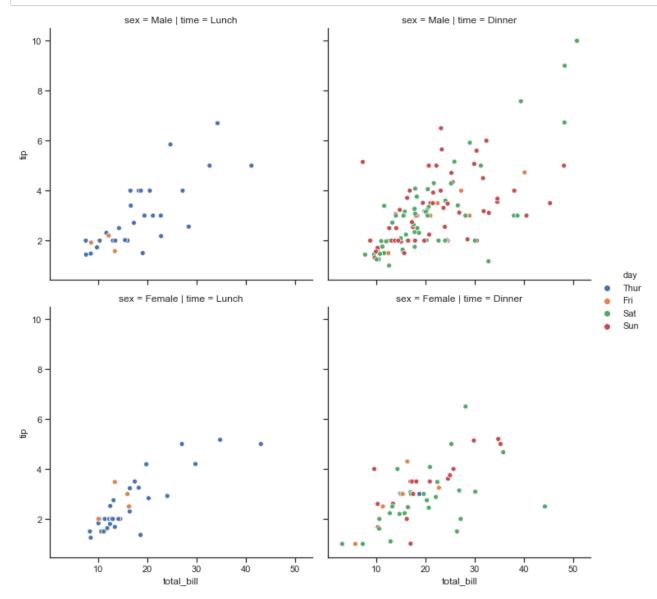


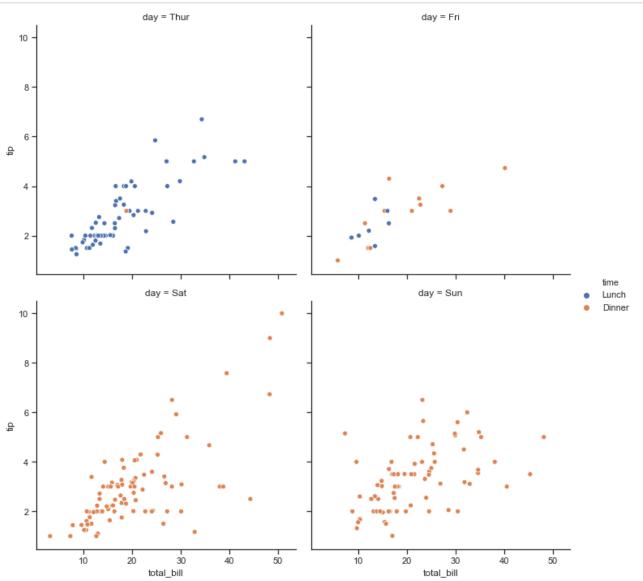


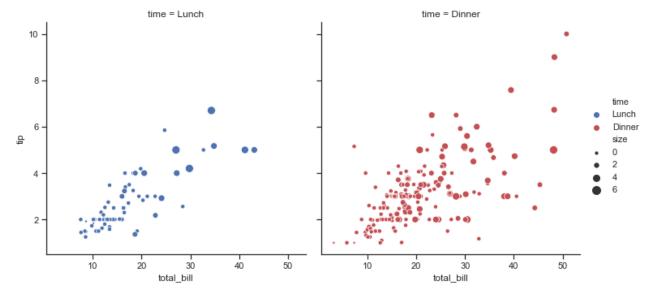
Showing multiple relationships

• when a relationship between two variables depends on more than one variable?









Line Plots

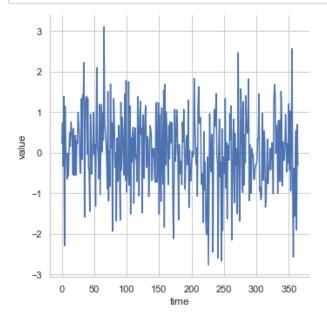
```
In [28]: sns.set(style="whitegrid")
    np.random.seed(54321)

In [29]: df = pd.DataFrame(dict(time=np.arange(365),
```

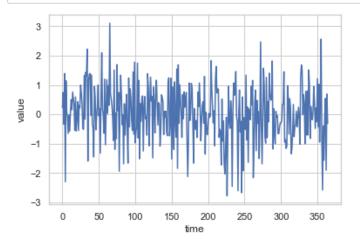
Out[29]:

value	time	
0.223979	0	0
0.744591	1	1
-0.334269	2	2
1.389172	3	3
-2.296095	4	4

```
In [30]: sns.relplot(x="time", y="value", kind="line", data=df);
```



In [31]: sns.lineplot(x="time", y="value", data=df);

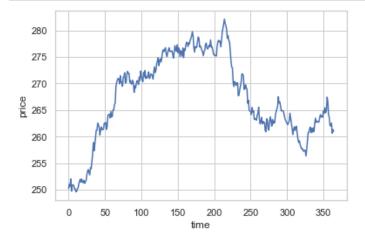


```
In [32]: df["price"] = 250 + df["value"].cumsum()
    df.head()
```

Out[32]:

	time	value	price
0	0	0.223979	250.223979
1	1	0.744591	250.968570
2	2	-0.334269	250.634301
3	3	1.389172	252.023472
4	4	-2.296095	249.727378

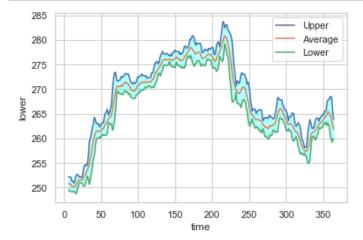
```
In [33]: sns.lineplot(x="time", y="price", data=df);
```



```
In [34]: df["mean"] = df['price'].rolling(7).mean()
    df["sd"] = df["price"].rolling(7).std()
    df["upper"] = df["mean"] + (2 * df["sd"])
    df["lower"] = df["mean"] - (2 * df["sd"])
```

Out[34]:

lower	upper	sd	mean	price	value	time	
NaN	NaN	NaN	NaN	250.223979	0.223979	0	0
NaN	NaN	NaN	NaN	250.968570	0.744591	1	1
NaN	NaN	NaN	NaN	250.634301	-0.334269	2	2
NaN	NaN	NaN	NaN	252.023472	1.389172	3	3
NaN	NaN	NaN	NaN	249.727378	-2.296095	4	4
NaN	NaN	NaN	NaN	250.868583	1.141205	5	5
249.341793	252.192582	0.712697	250.767188	250.924030	0.055448	6	6
249.512836	252.197908	0.671268	250.855372	250.841270	-0.082760	7	7
249.318752	252.169425	0.712668	250.744088	250.189583	-0.651688	8	8
249.188190	252.168346	0.745039	250.678268	250.173561	-0.016022	9	9
249.222611	251.441246	0.554659	250.331929	249.599096	-0.574465	10	10
249.180808	251.460153	0.569836	250.320480	249.647239	0.048143	11	11



In []:	
In []:	