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## **Scatter Plots**

Scatter plots can show how different features are related to one another, the main theme between all relational plot types is they display how features are interconnected to each other. There are many different types of plots that can be used to show this, so let's explore the scatterplot() as well as general seaborn parameters applicable to other plot types.

#### **Data**

We'll use some generated data from: <a href="http://roycekimmons.com/tools/generated\_data">http://roycekimmons.com/tools/generated\_data</a> <a href="http://roycekimmons.com/tools/generated\_data">http://roycekimmons.com/tools/generated\_data</a>)

```
In [1]: import pandas as pd
import seaborn as sns

In [3]: df = pd.read_csv("dm_office_sales.csv")
```

In [4]: df.head()

Out[4]:		division	level of education	training level	work experience	salary	sales
	0	printers	some college	2	6	91684	372302
	1	printers	associate's degree	2	10	119679	495660
	2	peripherals	high school	0	9	82045	320453
	3	office supplies	associate's degree	2	5	92949	377148
	4	office supplies	high school	1	5	71280	312802

```
Column
                       Non-Null Count Dtype
    -----
                        -----
                                       ----
0
   division
                                       object
                       1000 non-null
1
    level of education 1000 non-null
                                       object
                                       int64
2
   training level
                       1000 non-null
                       1000 non-null
                                       int64
3
   work experience
4
    salary
                       1000 non-null
                                       int64
    sales
5
                       1000 non-null
                                       int64
```

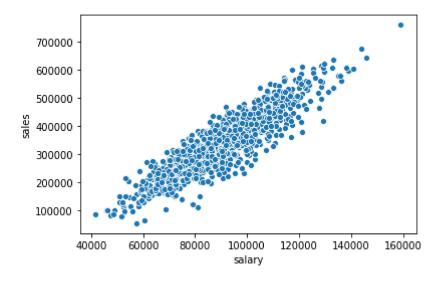
dtypes: int64(4), object(2)
memory usage: 47.0+ KB

Data columns (total 6 columns):

# **Scatterplot**

```
In [6]: sns.scatterplot(x='salary',y='sales',data=df)
```

Out[6]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089e370088>



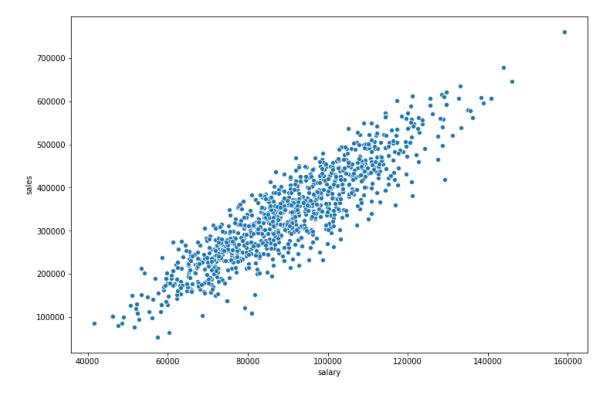
## Connecting to Figure in Matplotlib

Note how matplotlib is still connected to seaborn underneath (even without importing matplotlib.pyplot), since seaborn itself is directly making a Figure call with matplotlib. We can import matplotlib.pyplot and make calls to directly effect the seaborn figure.

```
In [7]: import matplotlib.pyplot as plt
```

```
In [8]: plt.figure(figsize=(12,8))
sns.scatterplot(x='salary',y='sales',data=df)
```

Out[8]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089fb16a08>



### **Seaborn Parameters**

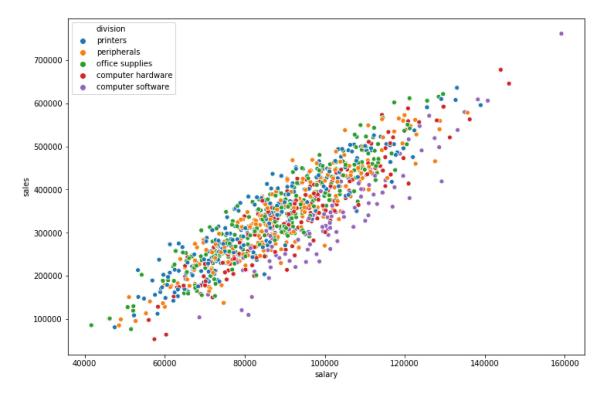
The hue and palette parameters are commonly available around many plot calls in seaborn.

#### hue

Color points based off a categorical feature in the DataFrame

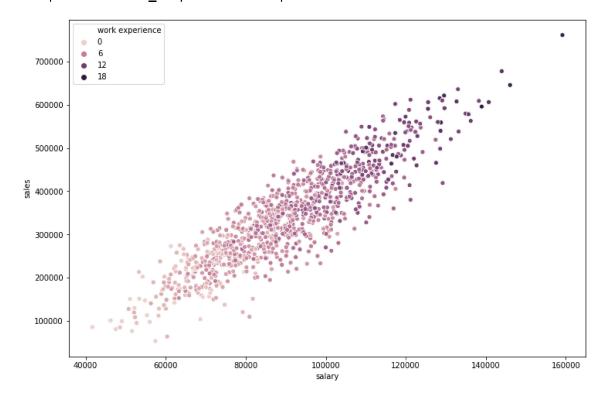
```
In [9]: plt.figure(figsize=(12,8))
sns.scatterplot(x='salary',y='sales',data=df,hue='division')
```

Out[9]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089fb0fe88>



In [10]: plt.figure(figsize=(12,8))
sns.scatterplot(x='salary',y='sales',data=df,hue='work experience')

Out[10]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089fc3d848>

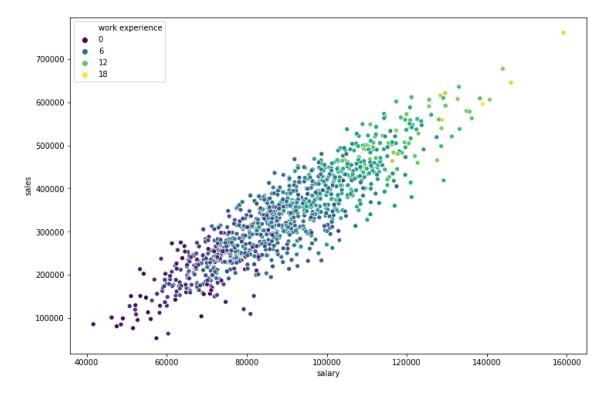


#### Choosing a palette from Matplotlib's cmap:

https://matplotlib.org/tutorials/colors/colormaps.html (https://matplotlib.org/tutorials/colors/colormaps.html)

```
In [11]: plt.figure(figsize=(12,8))
    sns.scatterplot(x='salary',y='sales',data=df,hue='work experience',palette=
```

Out[11]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089fcbbdc8>



## **Scatterplot Parameters**

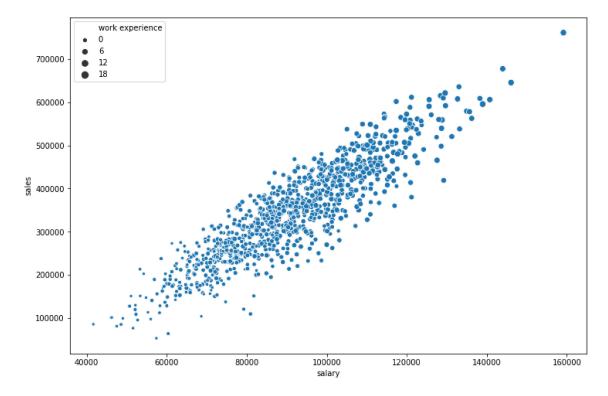
These parameters are more specific to the scatterplot() call

#### size

Allows you to size based on another column

```
In [12]: plt.figure(figsize=(12,8))
sns.scatterplot(x='salary',y='sales',data=df,size='work experience')
```

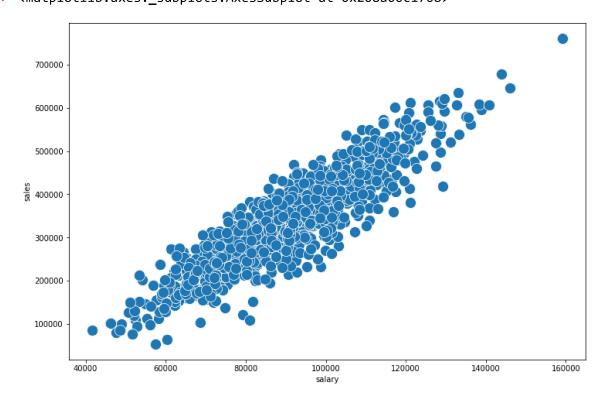
Out[12]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2089fcb7188>



Use s= if you want to change the marker size to be some uniform integer value

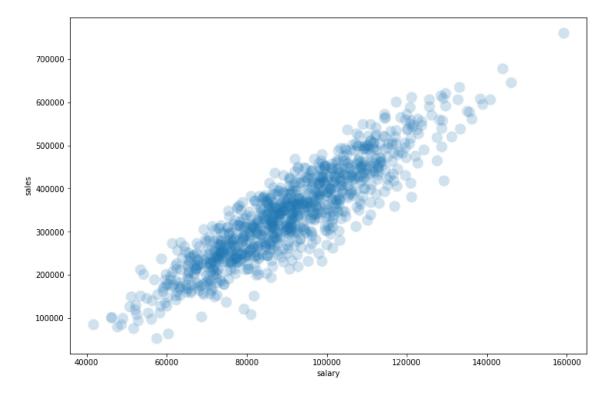
```
In [13]: plt.figure(figsize=(12,8))
sns.scatterplot(x='salary',y='sales',data=df,s=200)
```

Out[13]: <matplotlib.axes.\_subplots.AxesSubplot at 0x208a00c1708>



```
In [17]: plt.figure(figsize=(12,8))
    sns.scatterplot(x='salary',y='sales',data=df,s=200,linewidth=0,alpha=0.2)
```

Out[17]: <matplotlib.axes.\_subplots.AxesSubplot at 0x208a077b908>

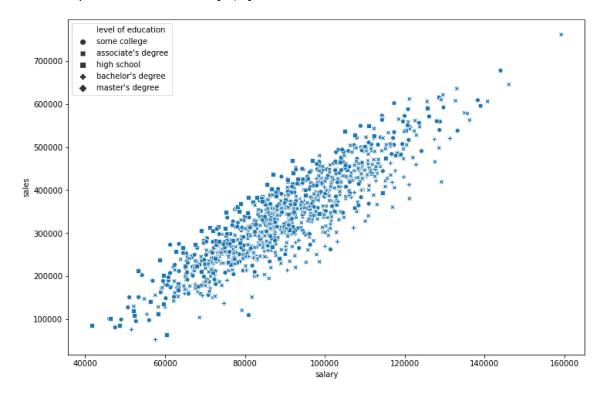


### style

Automatically choose styles based on another categorical feature in the dataset. Optionally use the **markers=** parameter to pass a list of marker choices based off matplotlib, for example: ['\*','+','o']

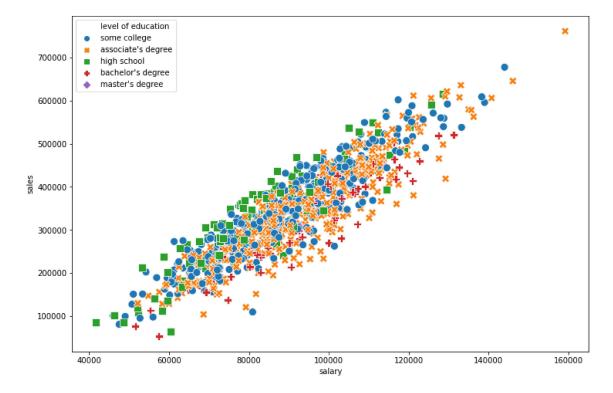
```
In [13]: plt.figure(figsize=(12,8))
    sns.scatterplot(x='salary',y='sales',data=df,style='level of education')
```

Out[13]: <AxesSubplot:xlabel='salary', ylabel='sales'>



In [14]: plt.figure(figsize=(12,8))
# Sometimes its nice to do BOTH hue and style off the same column
sns.scatterplot(x='salary',y='sales',data=df,style='level of education',hue

Out[14]: <AxesSubplot:xlabel='salary', ylabel='sales'>



# **Exporting a Seaborn Figure**

```
In [16]: plt.figure(figsize=(12,8))
    sns.scatterplot(x='salary',y='sales',data=df,style='level of education',hue

# Call savefig in the same cell
    plt.savefig('example_scatter.jpg')
```

