KEY_Lesson18B_Scatterplots

August 14, 2019

1 Scatterplots

Scatterplots are used to examine the relationship between two variables.

```
[7]: # import seaborn, matplotlib
import seaborn as sns
import matplotlib.pyplot as plt

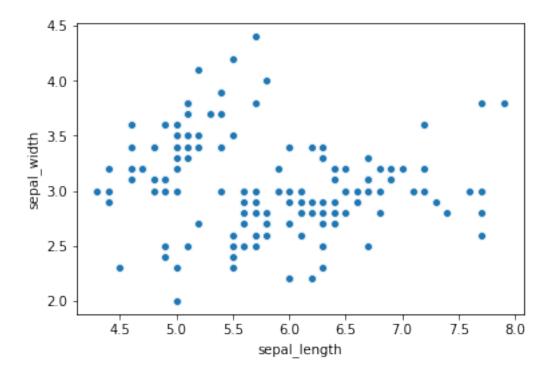
[2]: # load iris and preview the data
iris = sns.load_dataset("iris")
iris.head(10)
```

[2]:	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
5	5.4	3.9	1.7	0.4	setosa
6	4.6	3.4	1.4	0.3	setosa
7	5.0	3.4	1.5	0.2	setosa
8	4.4	2.9	1.4	0.2	setosa
9	4.9	3.1	1.5	0.1	setosa

Say we want to look at the relationship between sepal_length and sepal_width within our dataset. We'll use the sns.scatterplot function to plot this.

```
[3]: # plot sepal_length vs sepal_width sns.scatterplot('sepal_length', 'sepal_width', data=iris)
```

[3]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2137f390>



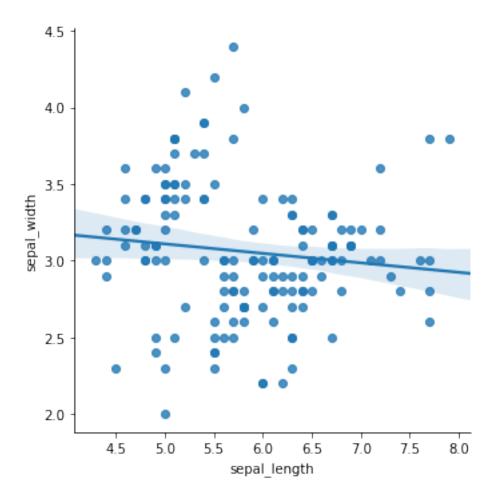
Can you remember what method in the statistics lessons we learned about that tells us about the relationship between two variables?

Correlation!

There is an easy way we can visualize the strength of the correlation on the plot using the lmplot function.

```
[4]: # plot sepal_length vs sepal_width with trendline sns.lmplot('sepal_length', 'sepal_width', data=iris)
```

[4]: <seaborn.axisgrid.FacetGrid at 0x1a216b0b70>



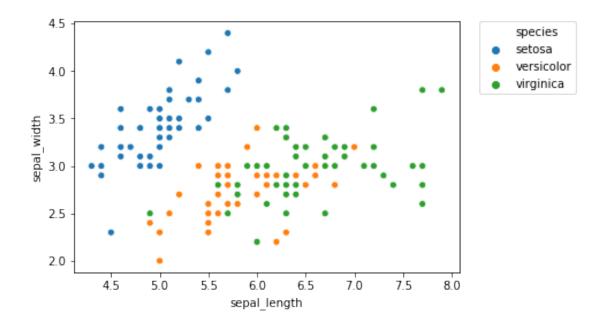
Based on this plot do you think there is a strong relationship between sepal_length and sepal_width in our data?

This gives us a general idea of the trend between sepal_length and sepal_width, but what if we wanted to explore the relationship between these variables on a more granular level? For example - if we wanted to see how this relationship might differ between the different species within our dataset? We can separate our plot similar to the way we did in the line graph using the hue parameter.

```
[9]: # plot sepal_length vs sepal_width colored by species
sns.scatterplot('sepal_length', 'sepal_width', data=iris, hue = 'species')

# the line below moves the legend outside of the plot borders
# dont worry about understanding this line of code
plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```

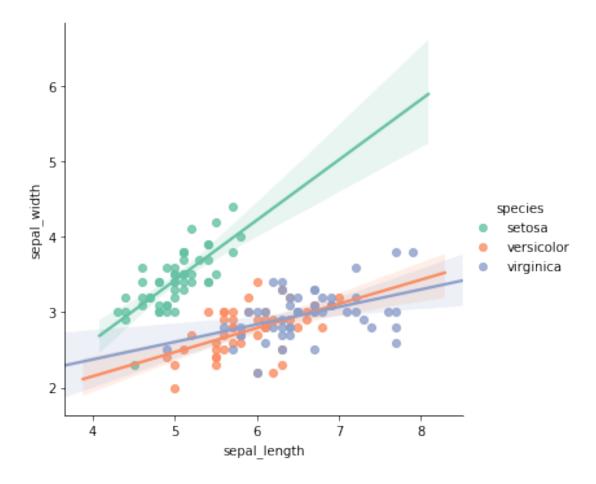
[9]: <matplotlib.legend.Legend at 0x1a21df9978>



Similarly, we can use the sns.lmplot function to add a linear trendline for each species separately. We can also change the color palette using the palette parameter

```
[11]: # plot sepal_length vs sepal_width colored by species
sns.lmplot('sepal_length', 'sepal_width', data=iris, hue = 'species',
→palette="Set2")
```

[11]: <seaborn.axisgrid.FacetGrid at 0x1a2207c278>



What do you notice about the relationship between our two variables when we separate (i.e. *stratify*) by species?

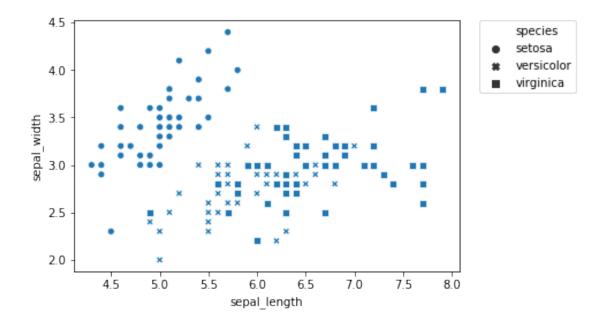
Instead of stratifying by species using color, we can do so using the marker shape with the style parameter.

```
[13]: # plot sepal_length vs sepal_width colored by species
sns.scatterplot('sepal_length', 'sepal_width', data=iris, style='species',

→palette = 'Set2')

# the line below moves the legend outside of the plot borders
plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```

[13]: <matplotlib.legend.Legend at 0x1a22417550>



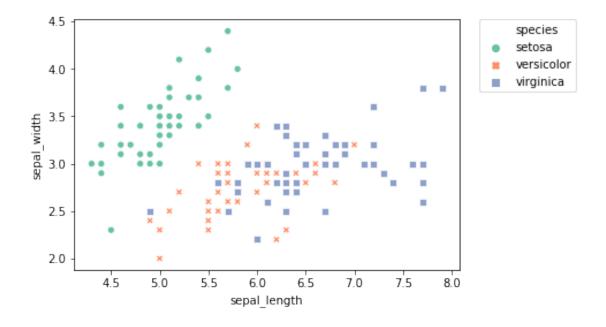
Lastly, we can combine hue, style and palette all together:

```
[17]: # plot sepal_length vs sepal_width colored by species
sns.scatterplot('sepal_length', 'sepal_width', data=iris, hue = 'species',

→style='species', palette = 'Set2')

# the line below moves the legend outside of the plot borders
plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```

[17]: <matplotlib.legend.Legend at 0x1a21926b00>



In this lesson we learned: * how to create a scatterplot in seaborn * stratifying a scatterplot by another variable using color (hue) * stratifying a scatterplot by another variable using marker shape (style) * changing the color palette of a stratified plot (palette)