

# Seaborn

March 21, 2019

## Seaborn Tutorial

```
In [6]: import pandas as pd
```

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

```
In [15]: import seaborn as sb
```

```
In [17]: sb.get_dataset_names()
```

```
C:\Users\by3001pm\AppData\Local\Continuum\anaconda3\lib\site-packages\bs4\__init__.py:181: UserWarning: Using
```

```
The code that caused this warning is on line 193 of the file C:\Users\by3001pm\AppData\Local\Continuum\anaconda3\lib\site-packages\bs4\__init__.py:181: UserWarning: Using
```

```
BeautifulSoup(YOUR_MARKUP})
```

to this:

```
BeautifulSoup(YOUR_MARKUP, "lxml")
```

```
markup_type=markup_type))
```

```
Out[17]: ['anscombe',  
          'attention',  
          'brain_networks',  
          'car_crashes',  
          'diamonds',  
          'dots',  
          'exercise',  
          'flights',  
          'fmri',  
          'gammas',  
          'iris',  
          'mpg',  
          'planets',  
          'tips',  
          'titanic']
```

```
In [11]: df = sb.load_dataset('tips')
```

```
In [12]: df.head()
```

```
Out[12]:
```

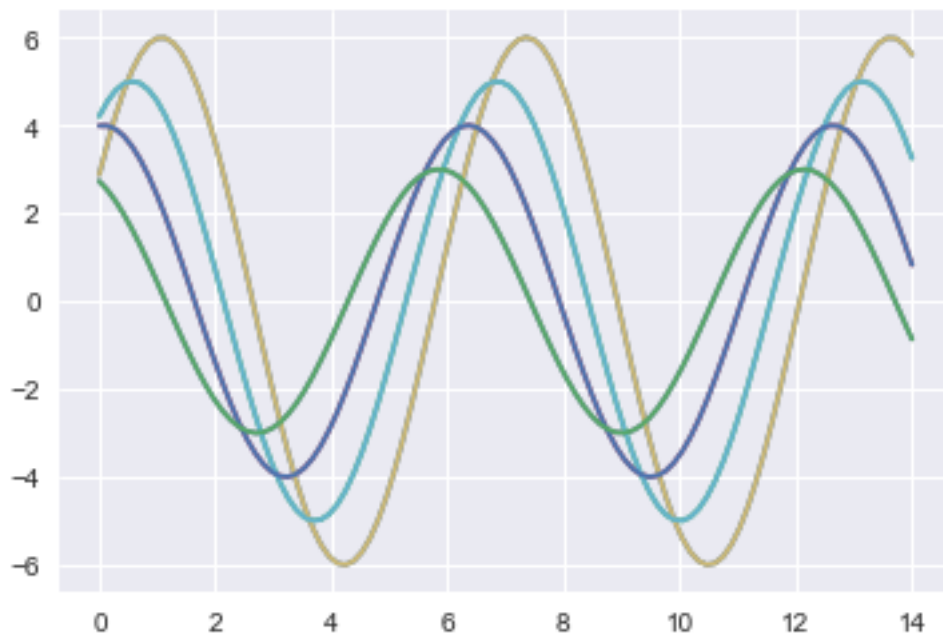
|   | total_bill | tip  | sex    | smoker | day | time   | size |
|---|------------|------|--------|--------|-----|--------|------|
| 0 | 16.99      | 1.01 | Female | No     | Sun | Dinner | 2    |
| 1 | 10.34      | 1.66 | Male   | No     | Sun | Dinner | 3    |
| 2 | 21.01      | 3.50 | Male   | No     | Sun | Dinner | 3    |
| 3 | 23.68      | 3.31 | Male   | No     | Sun | Dinner | 2    |
| 4 | 24.59      | 3.61 | Female | No     | Sun | Dinner | 4    |

```
In [13]: df.describe()
```

```
Out[13]:
```

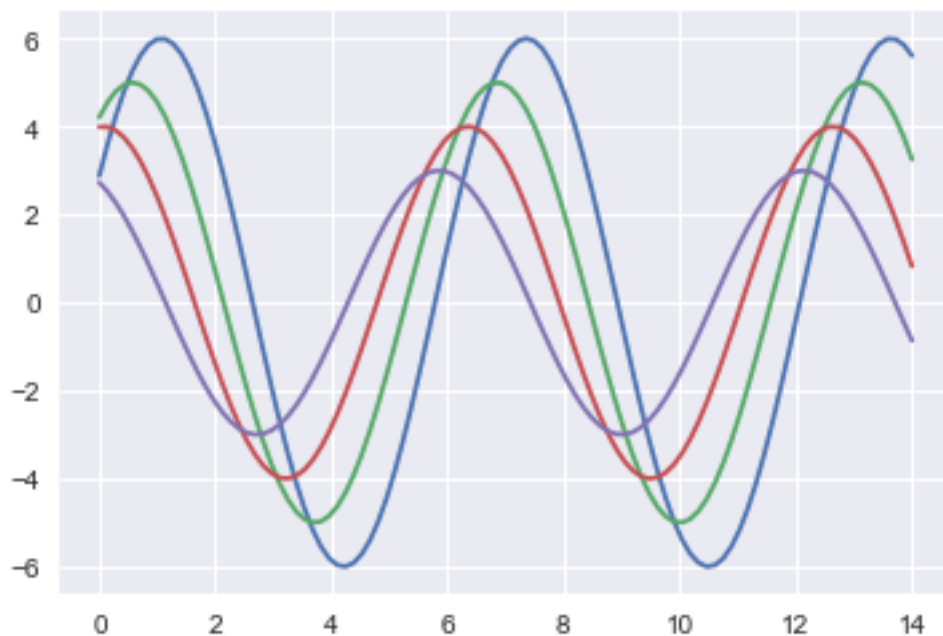
|       | total_bill | tip        | size       |
|-------|------------|------------|------------|
| count | 244.000000 | 244.000000 | 244.000000 |
| mean  | 19.785943  | 2.998279   | 2.569672   |
| std   | 8.902412   | 1.383638   | 0.951100   |
| min   | 3.070000   | 1.000000   | 1.000000   |
| 25%   | 13.347500  | 2.000000   | 2.000000   |
| 50%   | 17.795000  | 2.900000   | 2.000000   |
| 75%   | 24.127500  | 3.562500   | 3.000000   |
| max   | 50.810000  | 10.000000  | 6.000000   |

```
In [20]: import numpy as np
from matplotlib import pyplot as plt
def sinplot(flip = 1):
    x = np.linspace(0, 14, 100)
    for i in range(1, 5):
        plt.plot(x, np.sin(x + i * .5) * (7 - i) * flip)
sinplot()
plt.show()
```



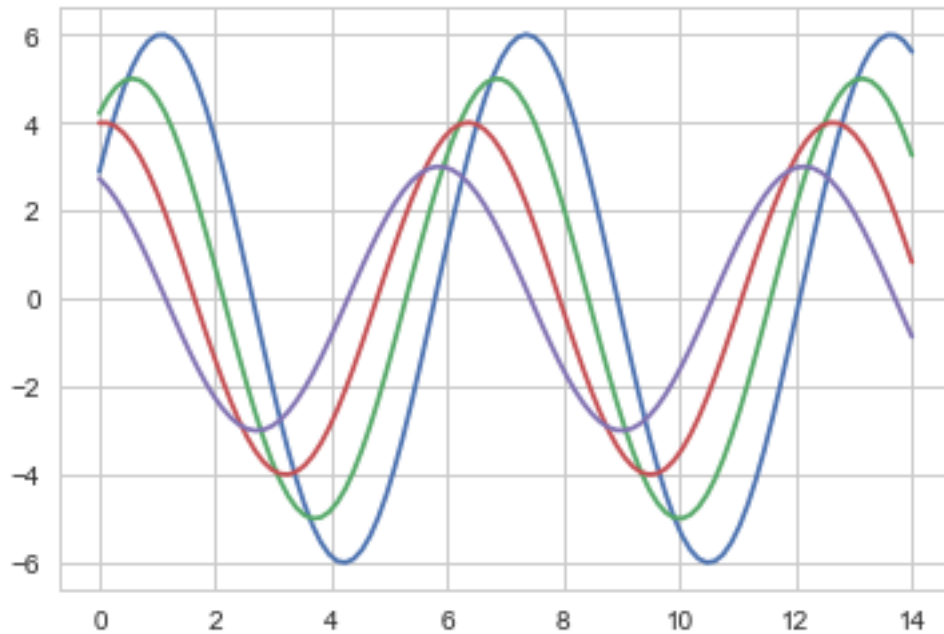
You can change the above plot to Seaborn with set() function

```
In [18]: import numpy as np
from matplotlib import pyplot as plt
def sinplot(flip = 1):
    x = np.linspace(0, 14, 100)
    for i in range(1, 5):
        plt.plot(x, np.sin(x + i * .5) * (7 - i) * flip)
import seaborn as sb
sb.set()
sinplot()
plt.show()
```



Using the set\_style() function, you can set the theme of the plot. The default theme is darkgrid. The other themes are whitegrid, dark, white, and ticks. Let's change the theme to whitegrid.

```
In [21]: import numpy as np
from matplotlib import pyplot as plt
def sinplot(flip=1):
    x = np.linspace(0, 14, 100)
    for i in range(1, 5):
        plt.plot(x, np.sin(x + i * .5) * (7 - i) * flip)
import seaborn as sb
sb.set_style("whitegrid")
sinplot()
plt.show()
```



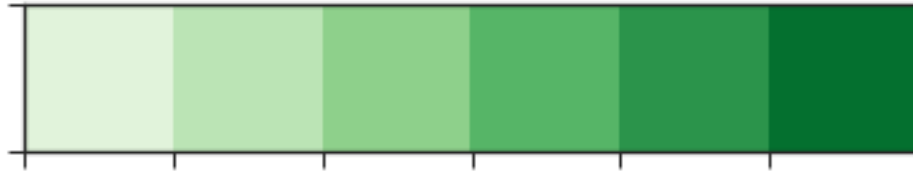
In Seaborn, you can use `color_palette()`, to give colors and aesthetics to a plot. For example:  
`seaborn.color_palette(palette = None, n_colors = None, desat = None, n_colors: Number of colors in the palette. Default is 6 colors. desat: desatuation gradient`  
`seaborn.palplot()` can also be used for color palettes.

```
In [23]: from matplotlib import pyplot as plt
import seaborn as sb
current_palette = sb.color_palette()
sb.palplot(current_palette)
plt.show()
```



Shades of green.

```
In [24]: from matplotlib import pyplot as plt
import seaborn as sb
current_palette = sb.color_palette()
sb.palplot(sb.color_palette("Greens"))
plt.show()
```



The following is the histogram of the Iris built-in dataset using Seaborn.

```
In [44]: df = sb.load_dataset('iris')
```

```
In [46]: df.head()
```

```
Out[46]:
```

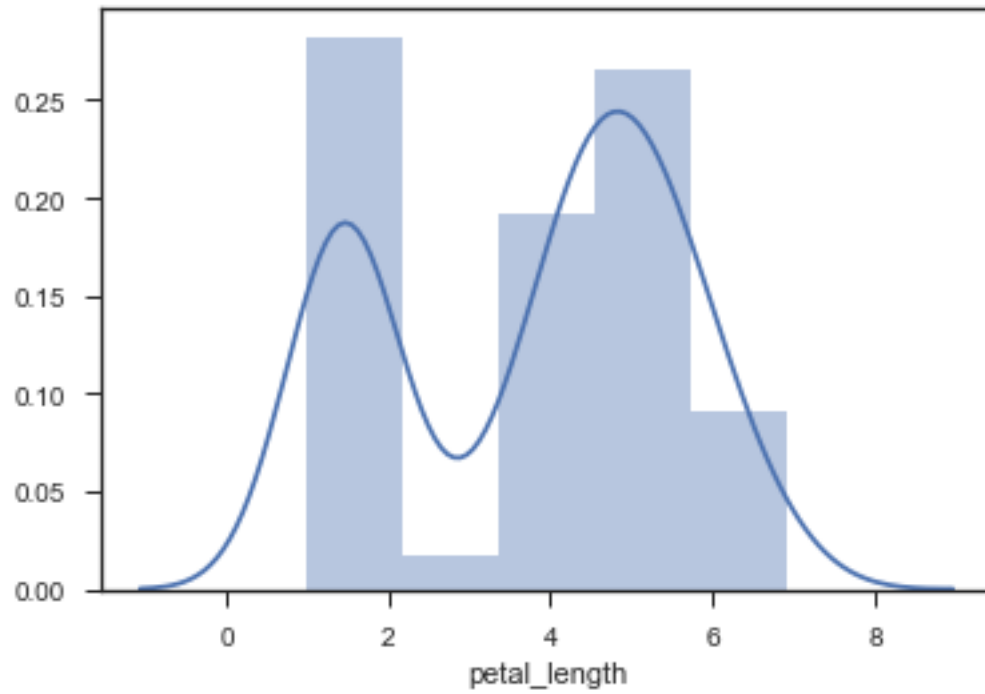
|   | sepal_length | sepal_width | petal_length | petal_width | species |
|---|--------------|-------------|--------------|-------------|---------|
| 0 | 5.1          | 3.5         | 1.4          | 0.2         | setosa  |
| 1 | 4.9          | 3.0         | 1.4          | 0.2         | setosa  |
| 2 | 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  |

```
In [45]: df.describe()
```

```
Out[45]:
```

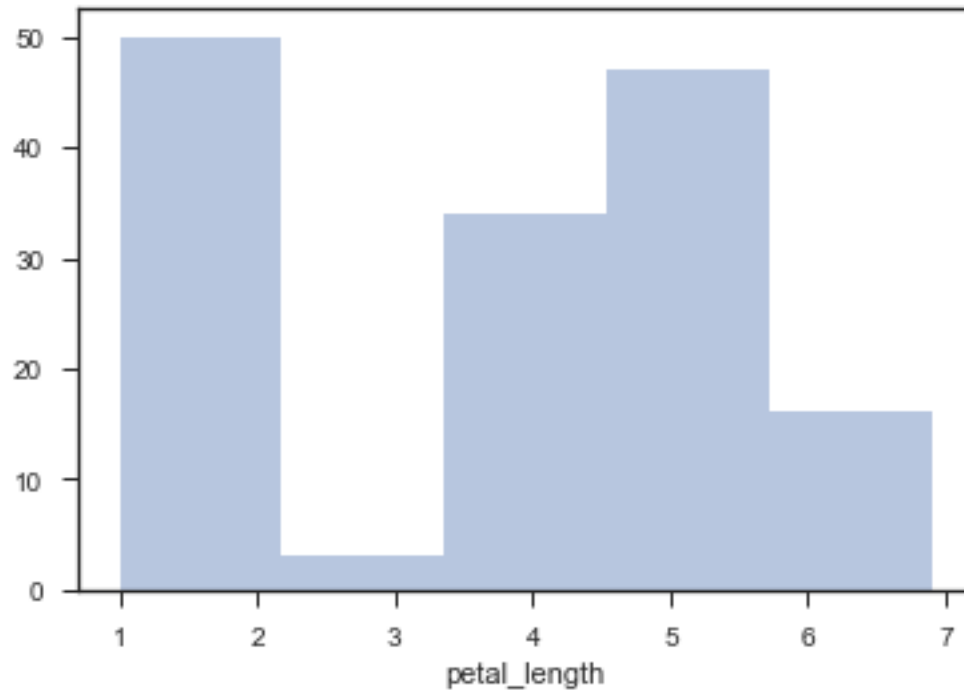
|       | sepal_length | sepal_width | petal_length | petal_width |
|-------|--------------|-------------|--------------|-------------|
| count | 150.000000   | 150.000000  | 150.000000   | 150.000000  |
| mean  | 5.843333     | 3.057333    | 3.758000     | 1.199333    |
| std   | 0.828066     | 0.435866    | 1.765298     | 0.762238    |
| min   | 4.300000     | 2.000000    | 1.000000     | 0.100000    |
| 25%   | 5.100000     | 2.800000    | 1.600000     | 0.300000    |
| 50%   | 5.800000     | 3.000000    | 4.350000     | 1.300000    |
| 75%   | 6.400000     | 3.300000    | 5.100000     | 1.800000    |
| max   | 7.900000     | 4.400000    | 6.900000     | 2.500000    |

```
In [26]: sb.distplot(df['petal_length'],kde = True)
plt.show()
```



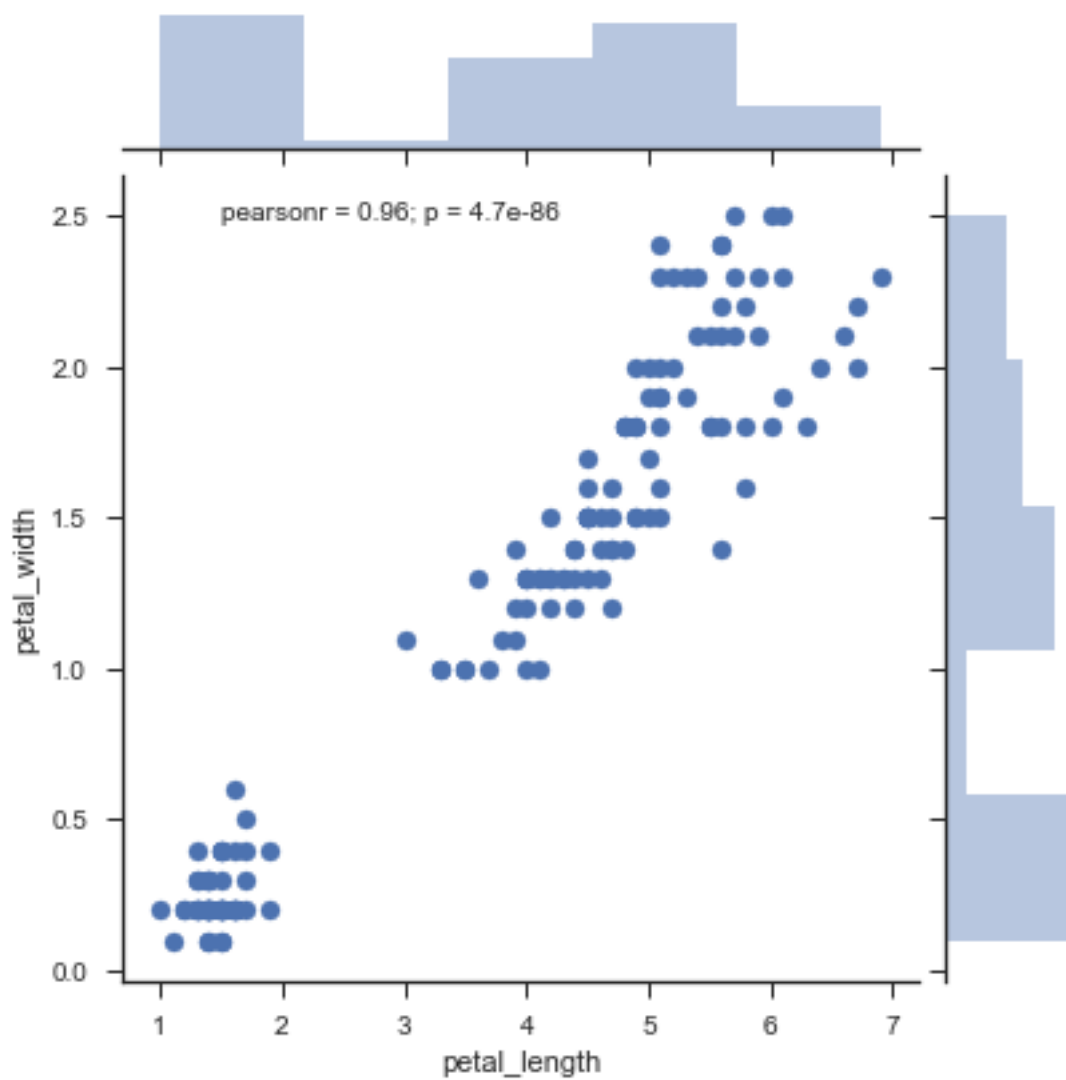
You generally don't need to set `kde` to `true`. It will give you a gaussian kernel density estimate (used in signal processing.)

```
In [25]: df = sb.load_dataset('iris')
         sb.distplot(df['petal_length'], kde = False)
         plt.show()
```



Scatterplot in Seaborn using the Iris dataset. There are mutiple ways to do this. The folowing is a jointplot in Seaborn that plots two variables and shows their relationship. It also gives you Pearson r and marginal histogram.

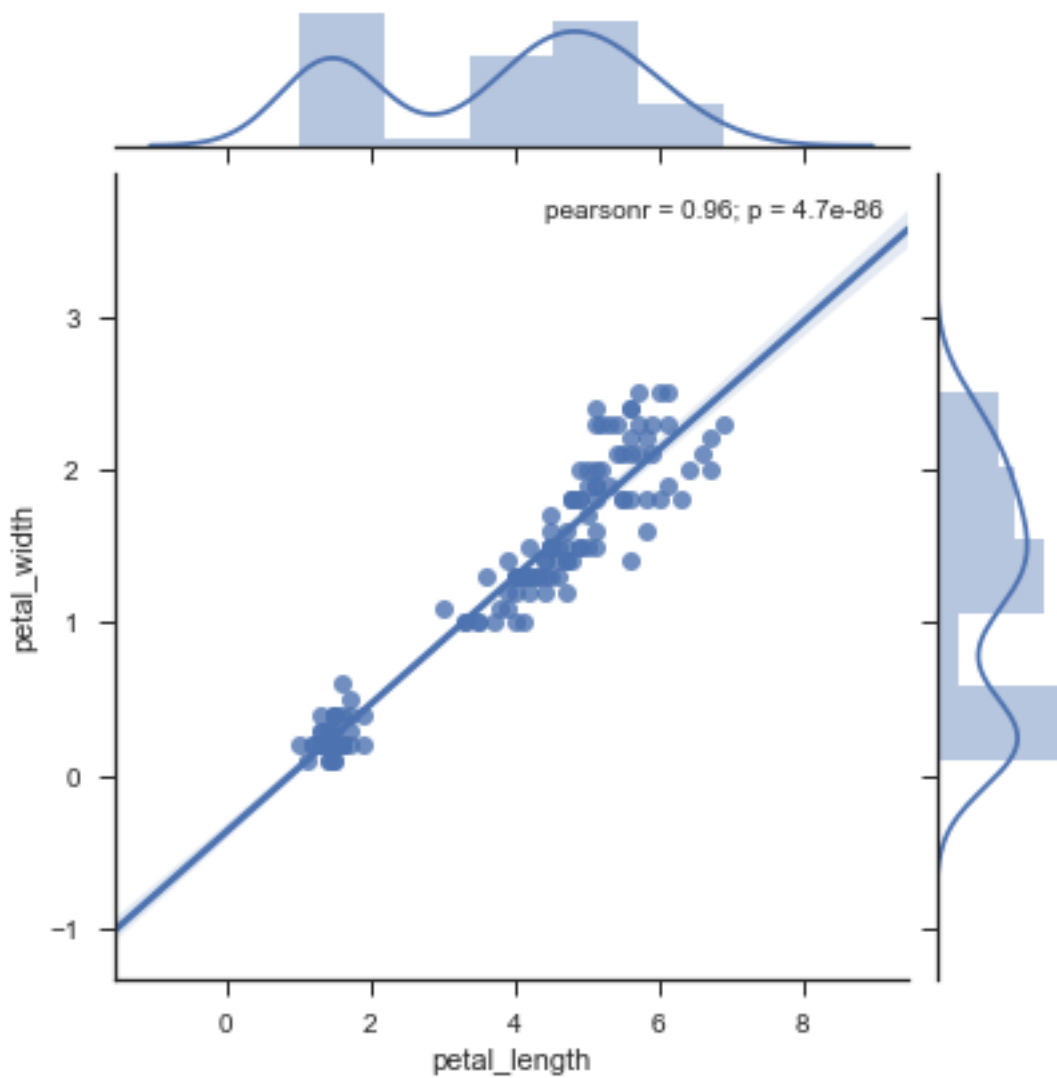
```
In [27]: sb.jointplot(x = 'petal_length',y = 'petal_width',data = df)
plt.show()
```



You can add regression and kernel density fit.

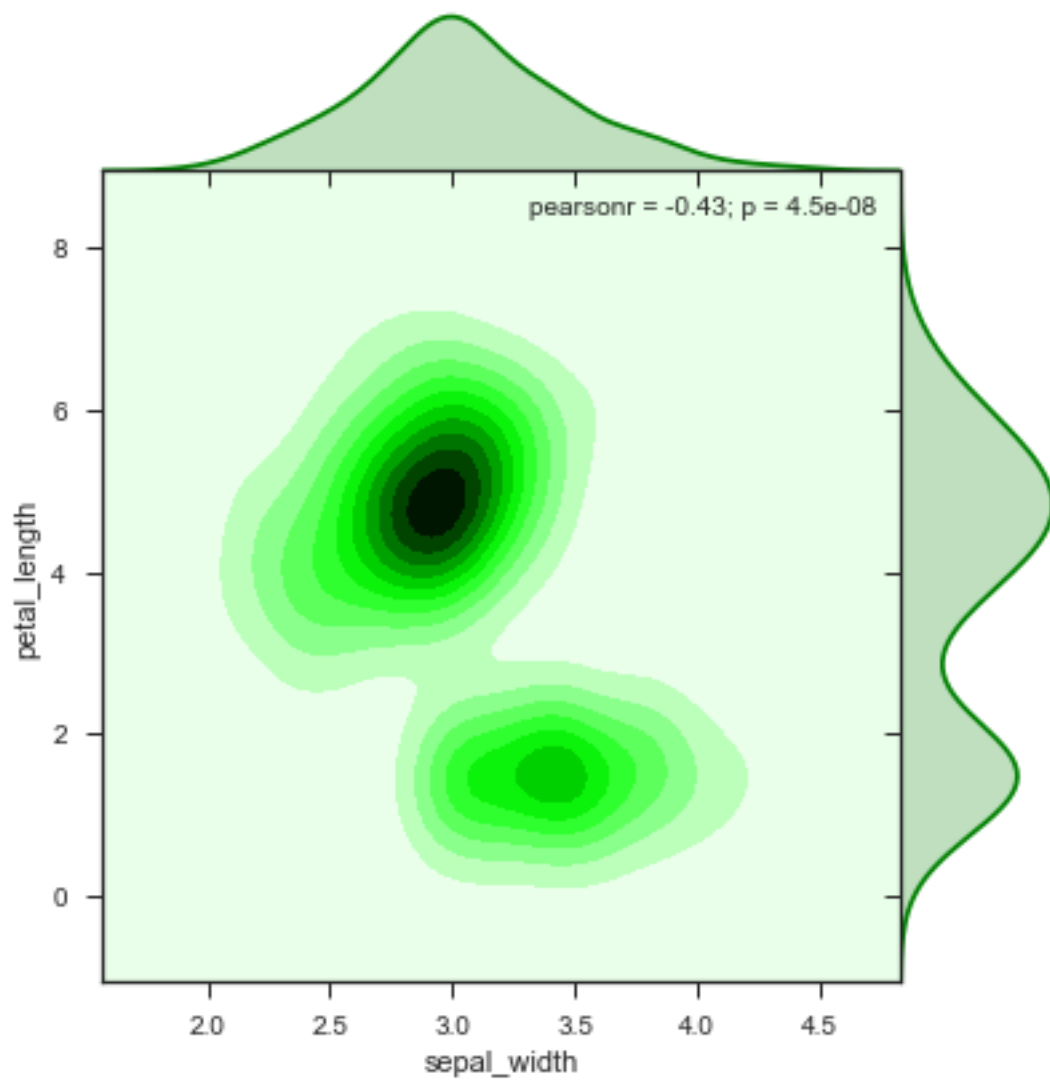
```
In [33]: sb.jointplot("petal_length", "petal_width", data=df, kind="reg")  
plt.show()
```





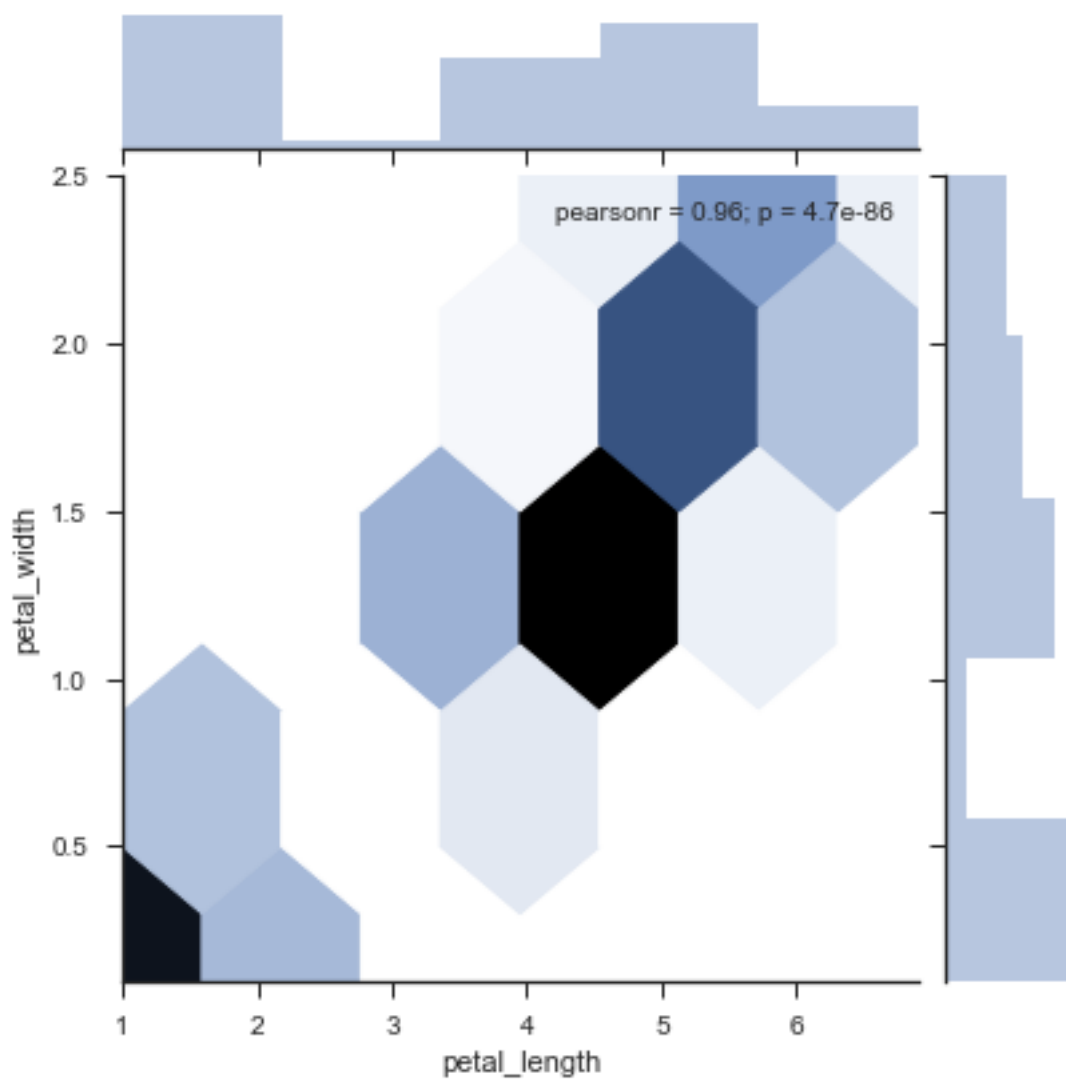
There appears to be a strong linear relationship between petal\_width and petal\_length.  
Density estimates.

```
In [37]: sb.jointplot("sepal_width", "petal_length", data=df, kind="kde", space=0, color="g")
plt.show()
```



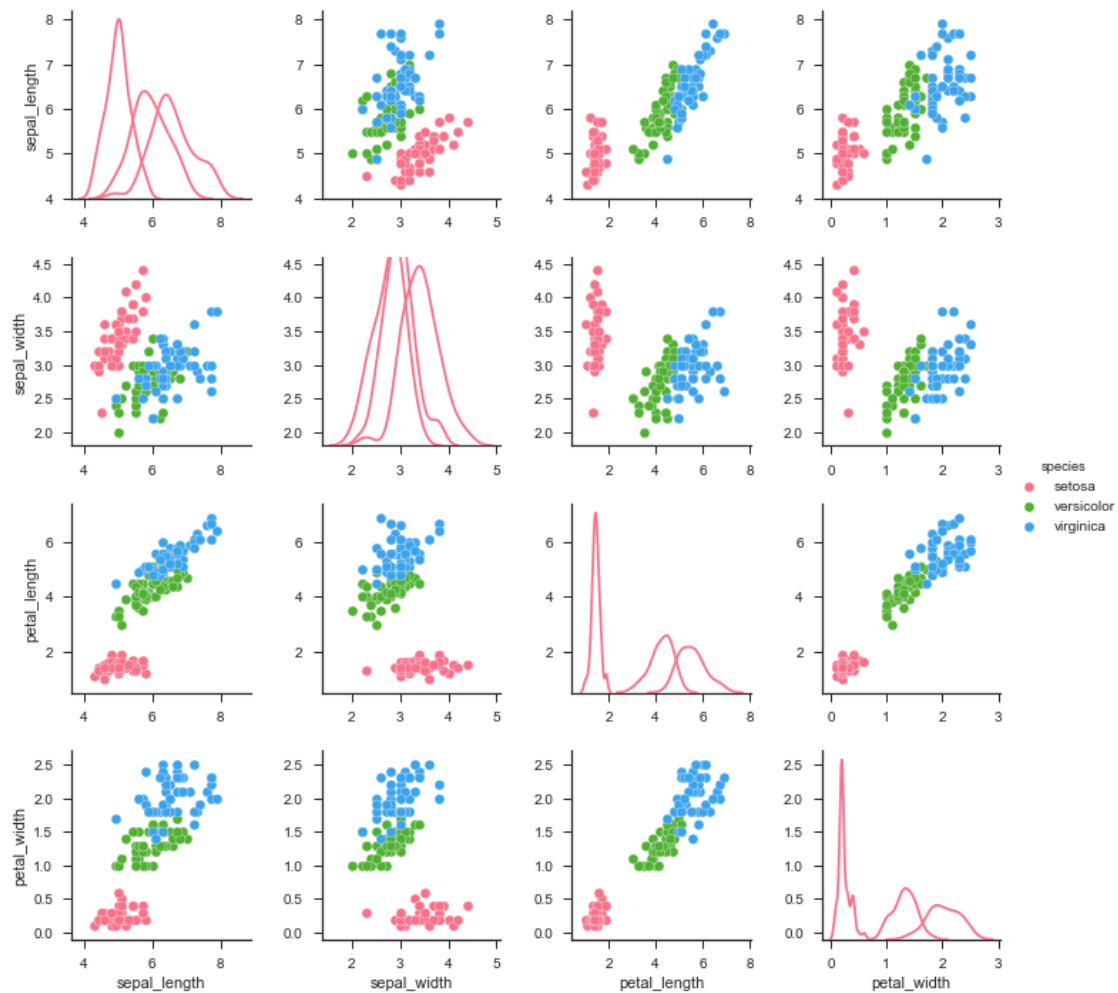
If your data is very scattered, you may use hexbin plot instead of scatterplot.

```
In [38]: sb.jointplot(x = 'petal_length',y = 'petal_width',data = df,kind = 'hex')  
plt.show()
```



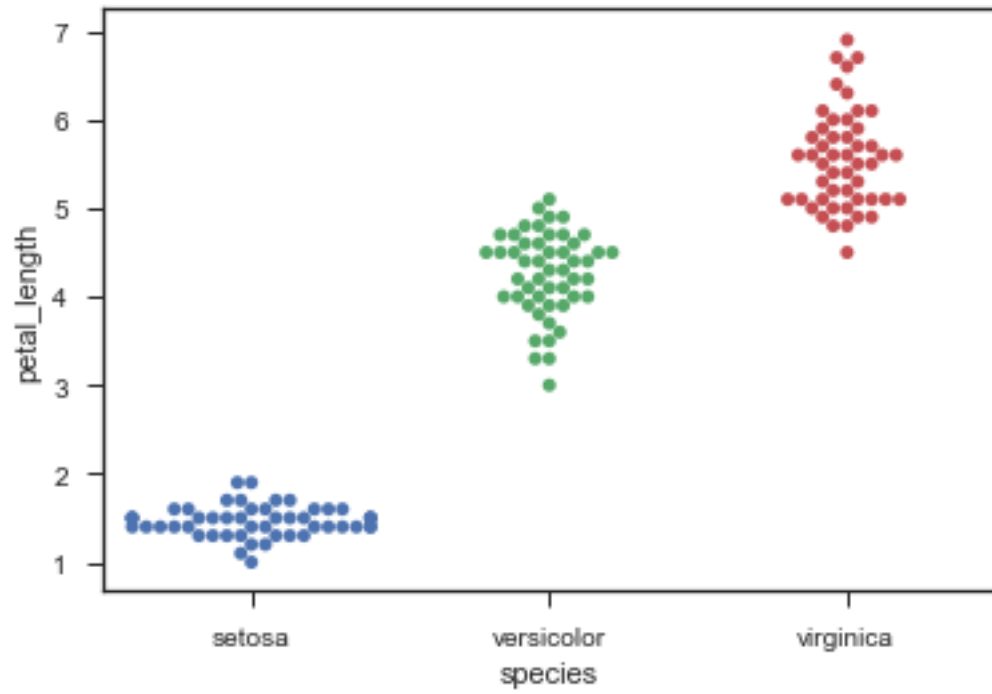
Use `pairplot()` to plot multiple pairwise bivariate distributions in a dataset.

```
In [39]: sb.set_style("ticks")
         sb.pairplot(df, hue = 'species', diag_kind = "kde", kind = "scatter", palette = "husl")
         plt.show()
```



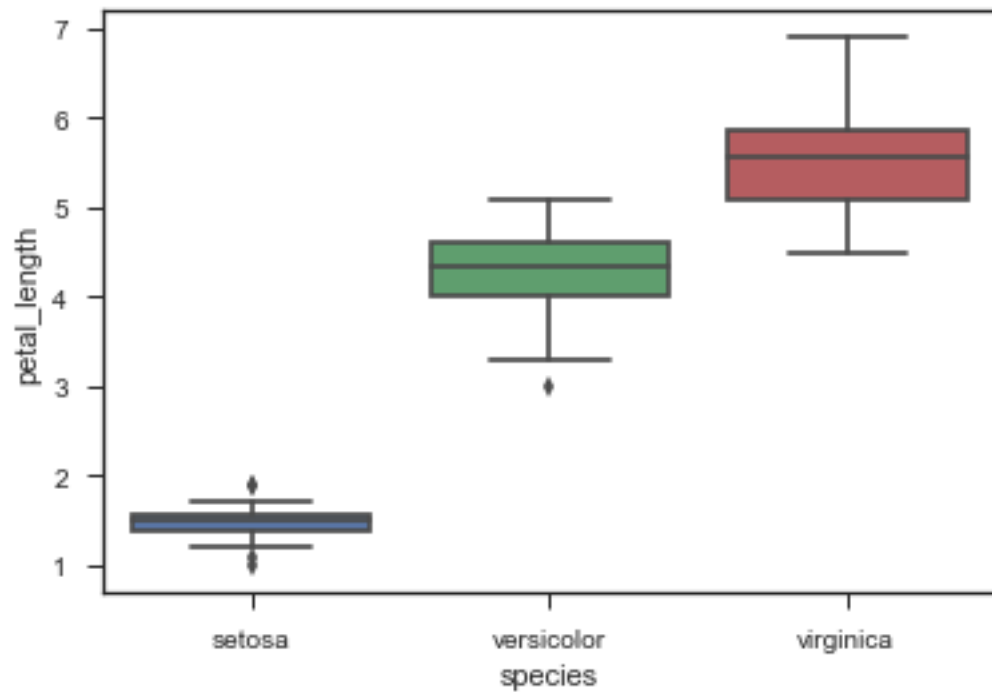
## Swarmplot

```
In [40]: sb.swarmplot(x = "species", y = "petal_length", data = df)
plt.show()
```



Boxplot

```
In [42]: sb.boxplot(x = "species", y = "petal_length", data = df)  
plt.show()
```



Violin plot

```
In [43]: sb.violinplot(x = "species", y = "petal_length", data = df)  
plt.show()
```

