

Line Graphs in Seaborn

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: # plt.style.use('ggplot') # Other styles to use; fivethirtyeight
sns.set_theme(style="whitegrid")
sns.set(rc={'figure.figsize':(20,10)}) # Set figure size
sns.set(font_scale = 2)
```

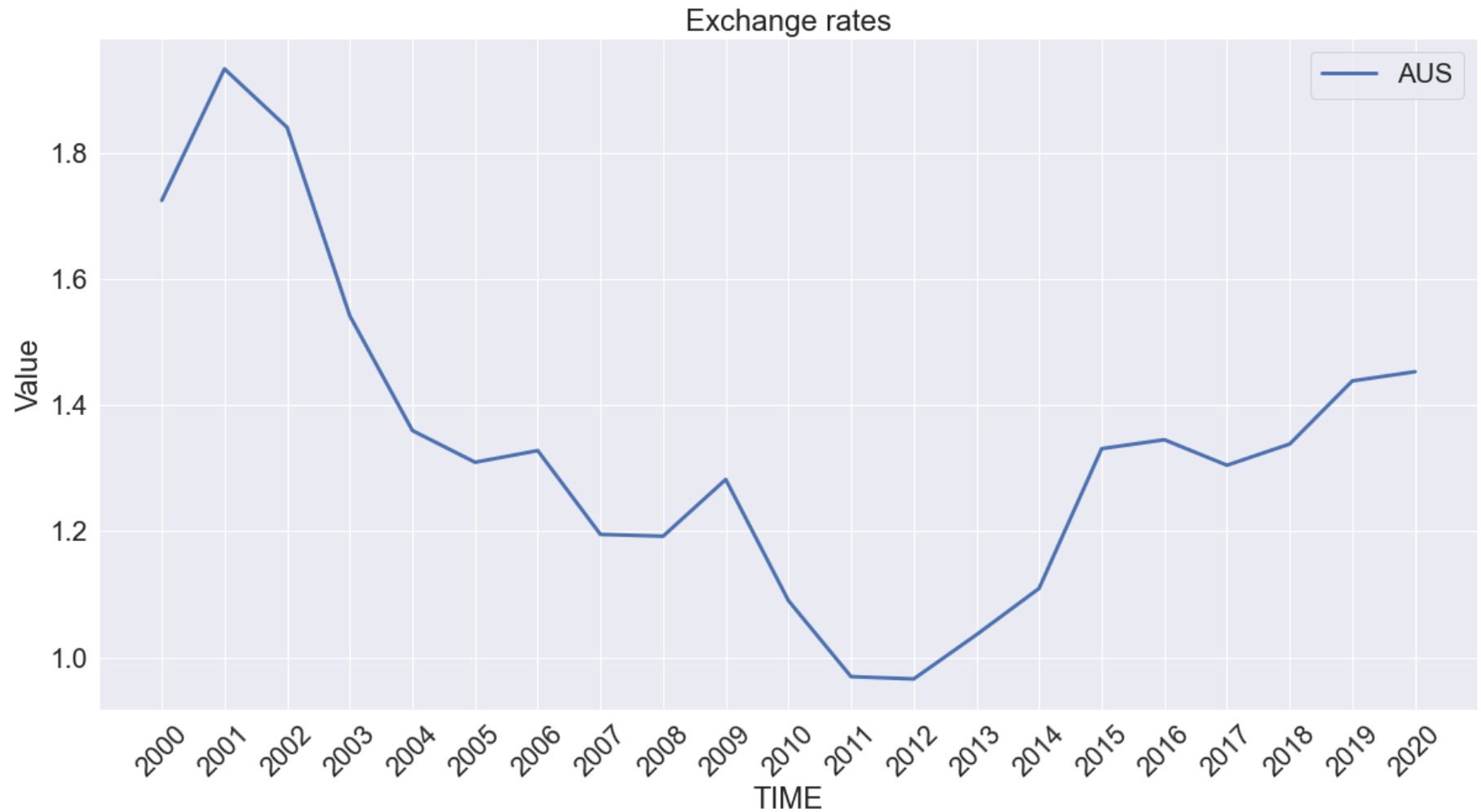
Simple line graph

```
In [3]: ex_rate_df=pd.read_csv('Exchange_Rates.csv')
ex_rate_df['TIME']=ex_rate_df['TIME'].astype(str)
simple_line_graph_ex_rate_df=ex_rate_df.query("LOCATION=='AUS'")
simple_line_graph_ex_rate_df.head()
```

Out[3]:

	LOCATION	INDICATOR	SUBJECT	MEASURE	FREQUENCY	TIME	Value	Flag Codes
0	AUS	EXCH	TOT	NATUSD	A	2000	1.724827	NaN
1	AUS	EXCH	TOT	NATUSD	A	2001	1.933443	NaN
2	AUS	EXCH	TOT	NATUSD	A	2002	1.840563	NaN
3	AUS	EXCH	TOT	NATUSD	A	2003	1.541914	NaN
4	AUS	EXCH	TOT	NATUSD	A	2004	1.359752	NaN

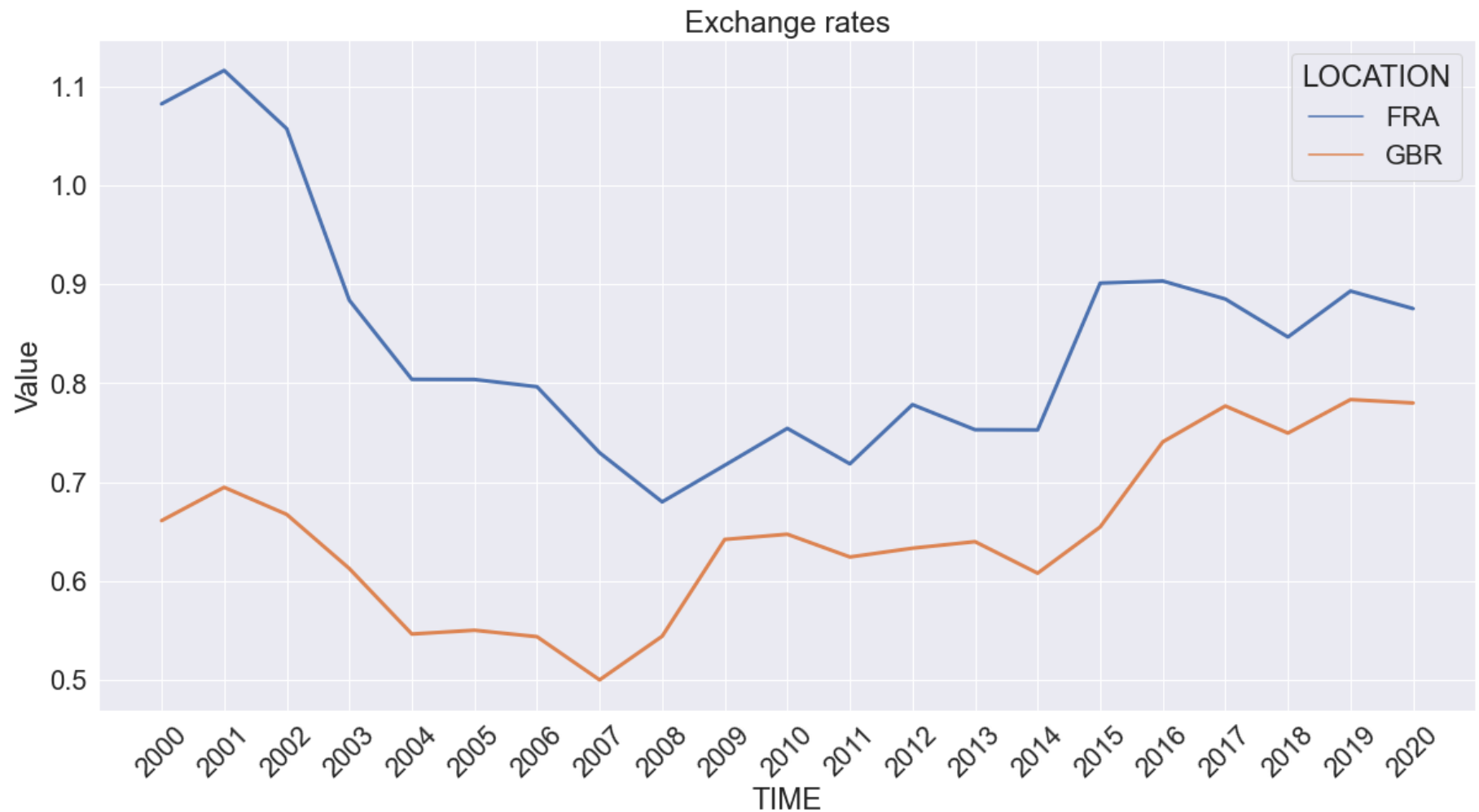
```
In [5]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.lineplot(data=simple_line_graph_ex_rate_df, x=simple_line_graph_ex_rate_df['TIME'],
              y=simple_line_graph_ex_rate_df['Value'],linewidth=3)
plt.title('Exchange rates',fontsize=24)
plt.legend(['AUS'], loc='upper right')
plt.xticks(rotation=45)
plt.show()
```



Multiple line graphs

```
In [6]: ex_rate_df=ex_rate_df[ex_rate_df['LOCATION'].isin(['FRA','GBR','USD'])]

plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.lineplot(data=ex_rate_df, x=ex_rate_df['TIME'], y=ex_rate_df['Value'],hue=ex_rate_df['LOCATION'],
              linewidth=3)
plt.title('Exchange rates',fontsize=24)
plt.xticks(rotation=45)
plt.show()
```



Simple Linear Regression Line

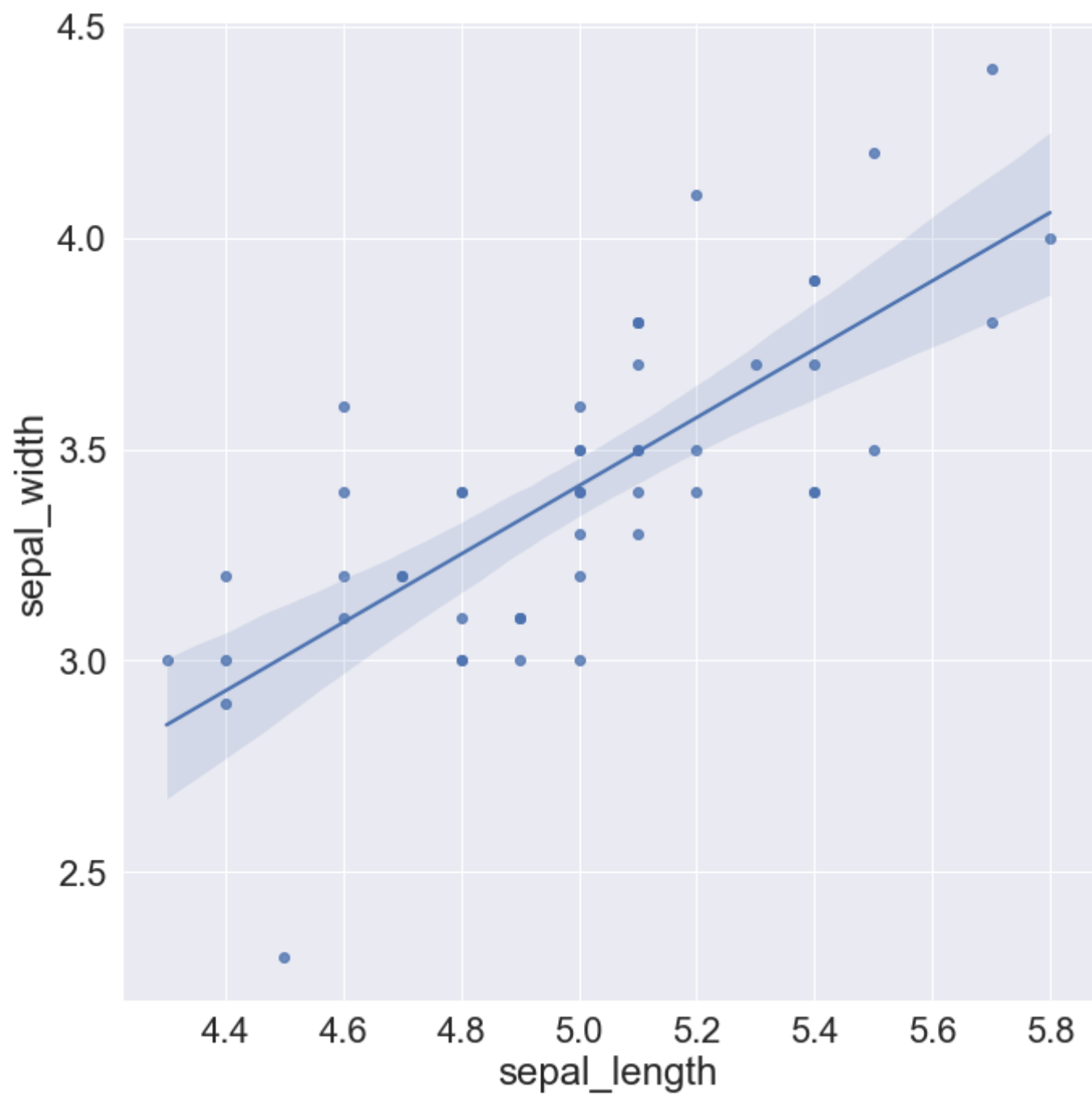
```
In [7]: iris_df=pd.read_csv('iris.csv')  
iris_df.head()
```

Out[7]:

	sepal_length	sepal_width	petal_length	petal_width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

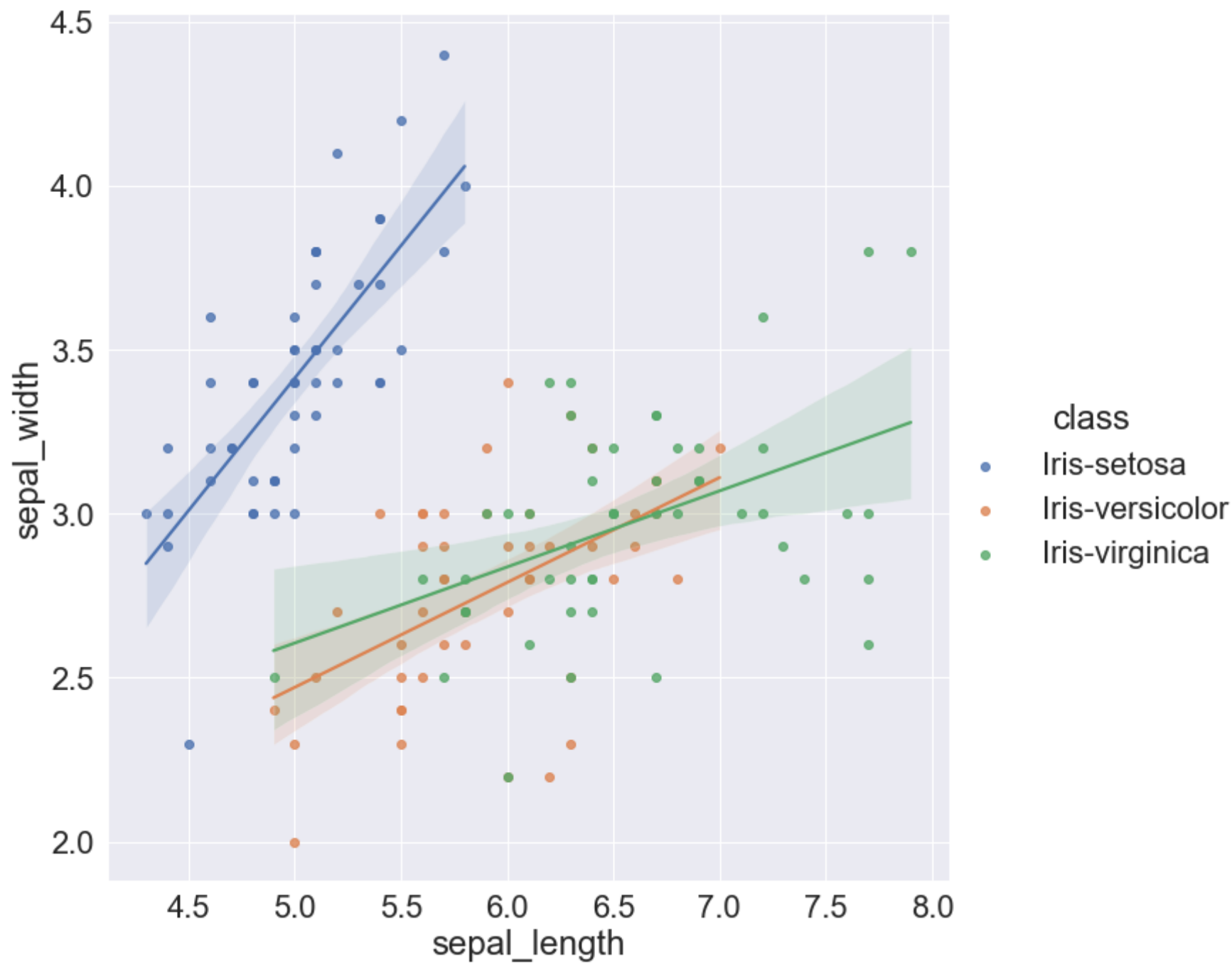
```
In [8]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)

sns.lmplot(data=iris_df[iris_df['class']=='Iris-setosa'], x="sepal_length", y="sepal_width",height=10)
plt.show()
```



Multiple Linear Regression


```
In [9]: sns.lmplot(data=iris_df,x="sepal_length", y="sepal_width", hue="class",height=10)  
plt.show()
```



Bar Plot

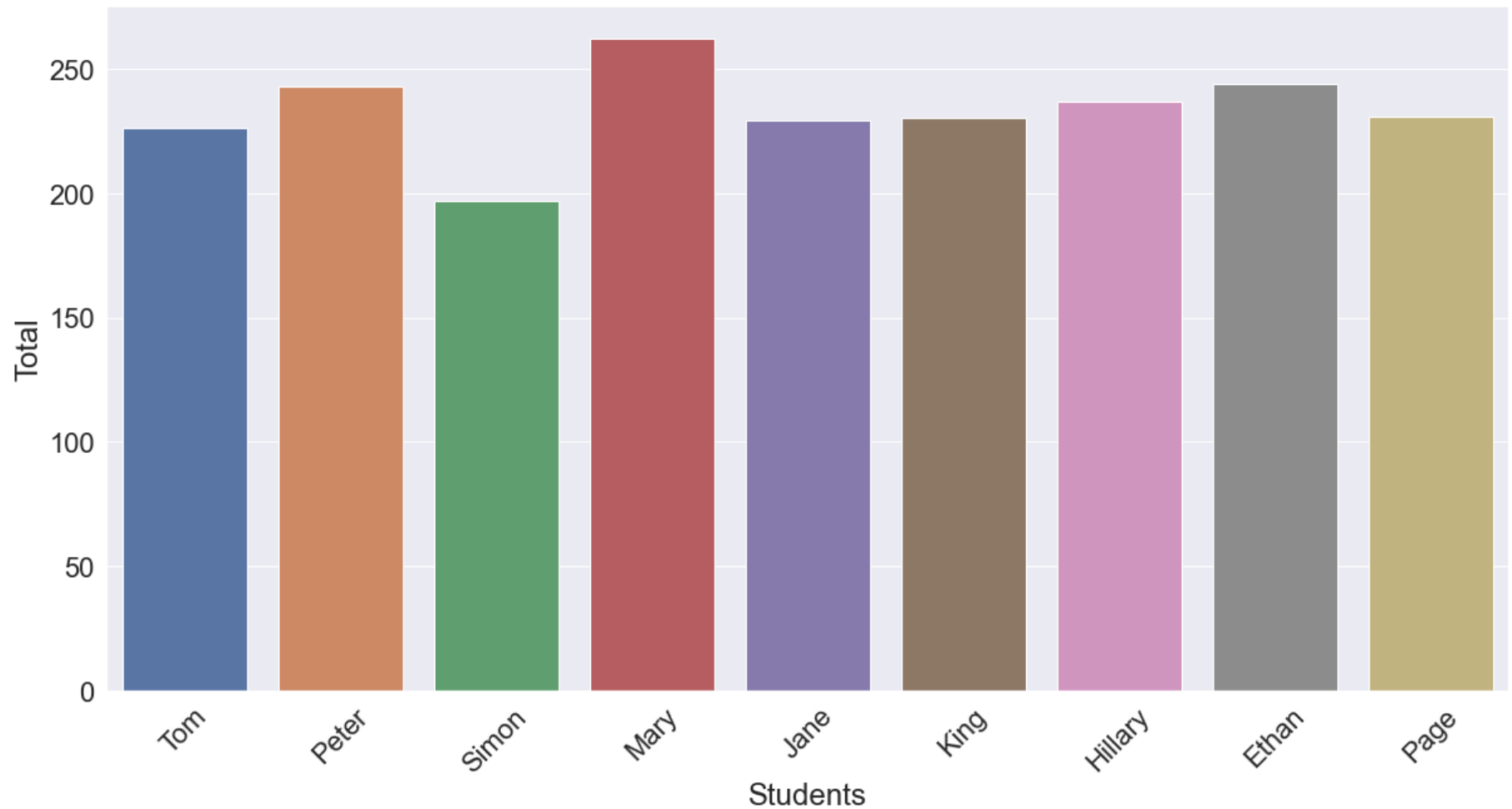
```
In [10]: score_df = pd.DataFrame(  
    {  
        "Students": ["Tom", "Peter","Simon", "Mary", "Jane","King","Hillary","Ethan","Page"],  
        "Math": [79.00, 67.00,80.00, 84.00, 70.00,60.00,90.00,76.00,75],  
        "Physics":[63.00, 98, 60.00, 90,84.00, 77.00,55.00,70,66.00],  
        "Computer":[84.00,78.00, 57.00, 88.00, 75.00,93.00,92.00,98.00,90.00],  
    },  
    index=["Tom", "Peter","Simon", "Mary", "Jane","King","Hillary","Ethan","Page"]  
)  
  
score_df['Total']=score_df[['Math','Physics','Computer']].apply(np.sum,axis=1)  
score_df
```

Out[10]:

	Students	Math	Physics	Computer	Total
Tom	Tom	79.0	63.0	84.0	226.0
Peter	Peter	67.0	98.0	78.0	243.0
Simon	Simon	80.0	60.0	57.0	197.0
Mary	Mary	84.0	90.0	88.0	262.0
Jane	Jane	70.0	84.0	75.0	229.0
King	King	60.0	77.0	93.0	230.0
Hillary	Hillary	90.0	55.0	92.0	237.0
Ethan	Ethan	76.0	70.0	98.0	244.0
Page	Page	75.0	66.0	90.0	231.0

Simple bar chart

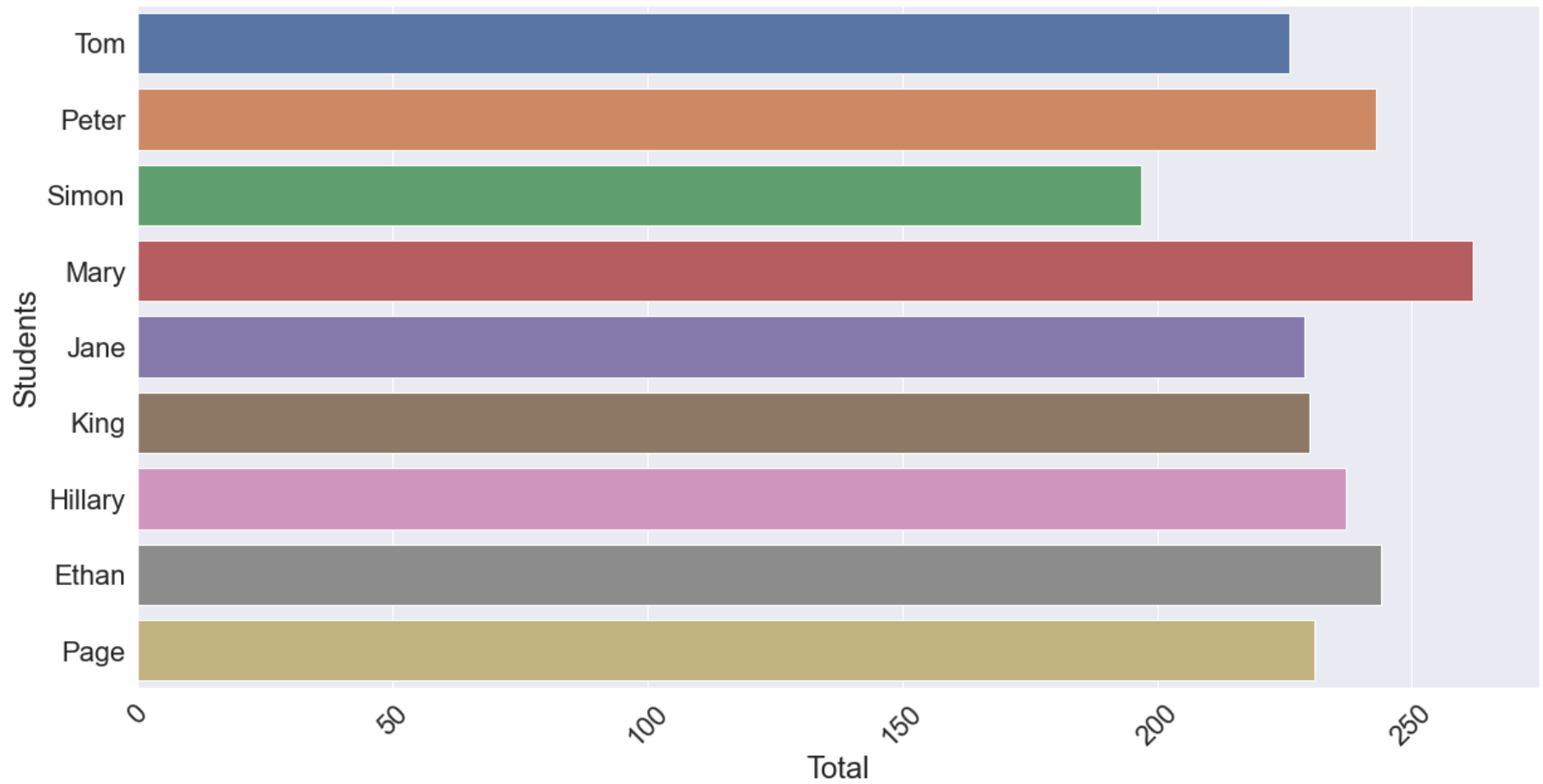
```
In [11]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.barplot(x=score_df['Students'], y=score_df['Total'])
plt.xticks(rotation=45)
plt.show()
```



Horizontal Bar chart

Simply interchange x and y axes

```
In [12]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.barplot(y=score_df['Students'], x=score_df['Total'])
plt.xticks(rotation=45)
plt.show()
```



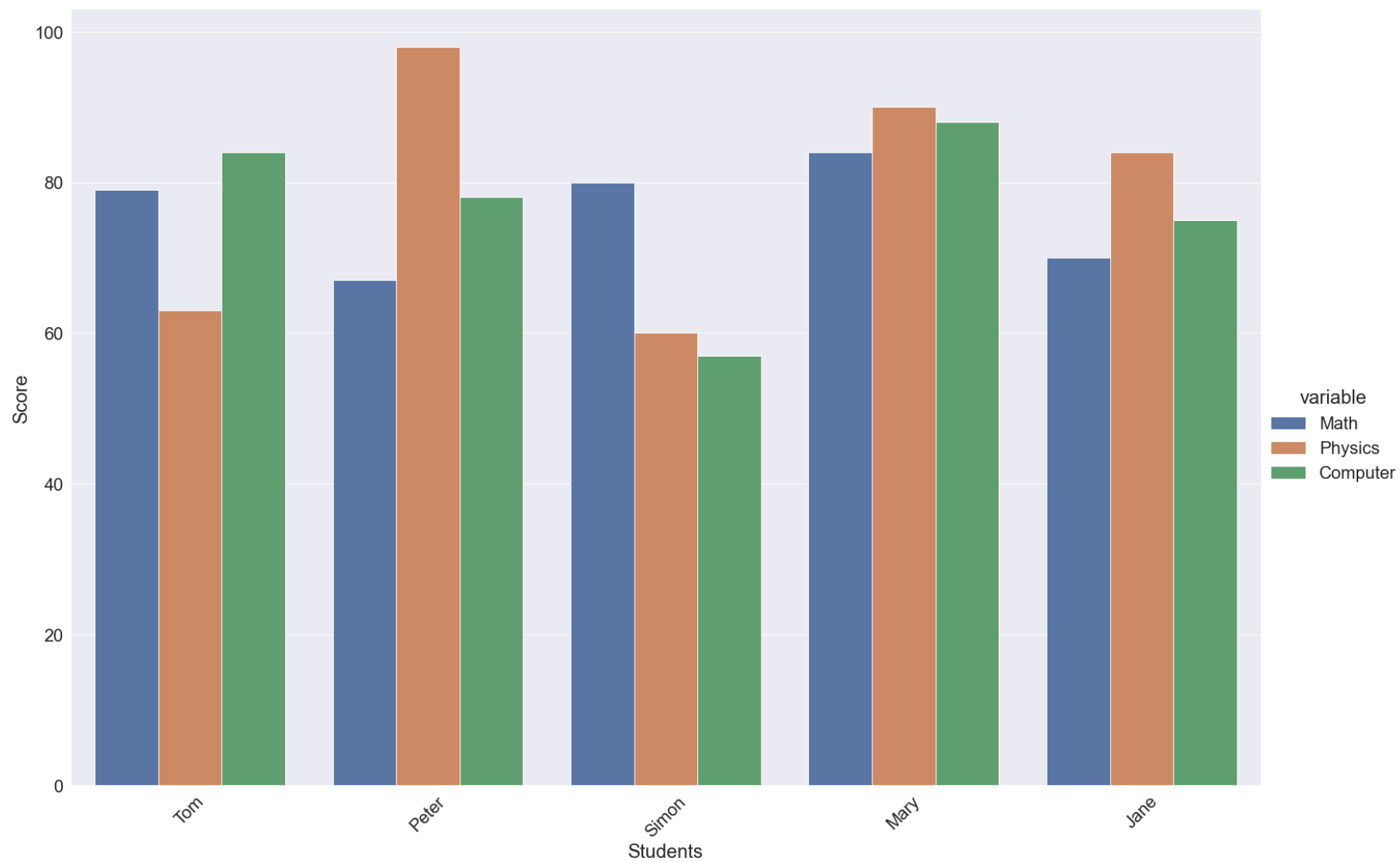
Grouped Bar chart

```
In [13]: score_melt_df=pd.melt(score_df.reset_index().head(),id_vars=['Students'],value_vars=['Math','Physics','Computer'],
                                value_name='Score')
score_melt_df.head()
```

Out[13]:

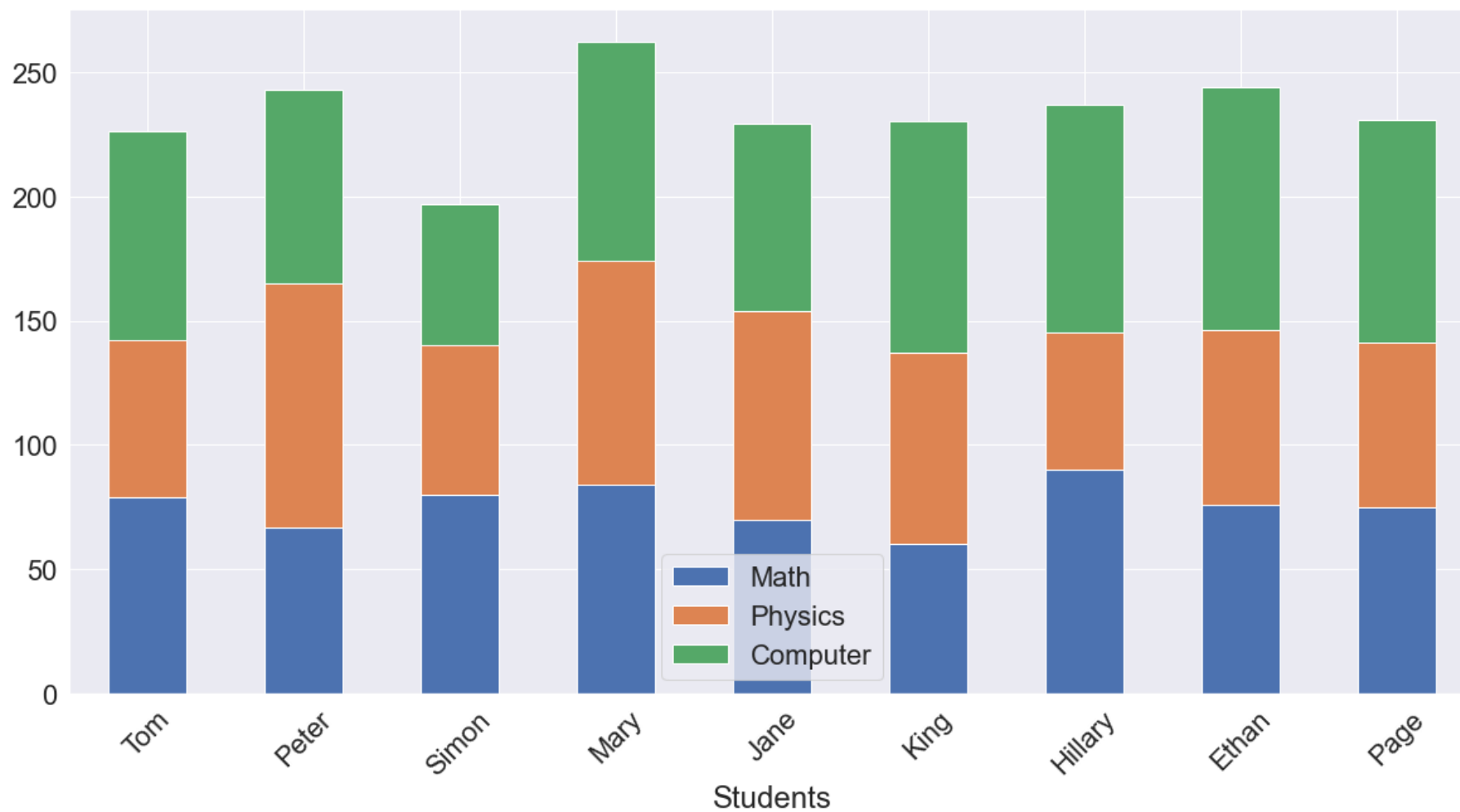
	Students	variable	Score
0	Tom	Math	79.0
1	Peter	Math	67.0
2	Simon	Math	80.0
3	Mary	Math	84.0
4	Jane	Math	70.0

```
In [14]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.catplot(data=score_melt_df, kind="bar", x="Students", y="Score", hue="variable",height=15, aspect=1.5)
plt.xticks(rotation=45)
plt.show()
```



Stacked Bar Chart

```
In [15]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
score_df[['Students', 'Math', 'Physics', 'Computer']].set_index('Students').plot(kind='bar', stacked=True)
plt.xticks(rotation=45)
plt.show()
```



Scatter Plot

```
In [16]: iris_df=pd.read_csv('iris.csv')
iris_df.head()
```

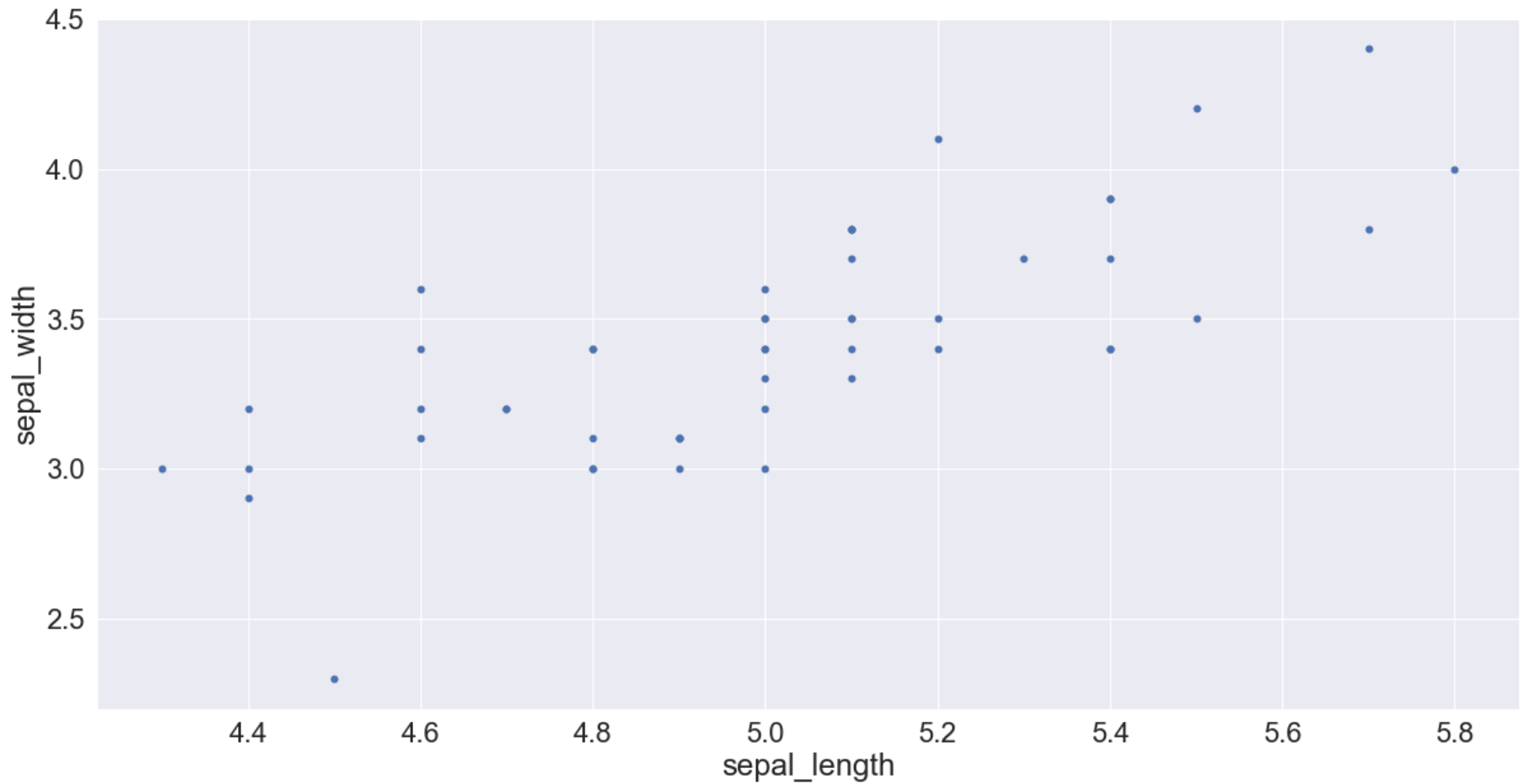
Out[16]:

	sepal_length	sepal_width	petal_length	petal_width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

Simple Scatter plot

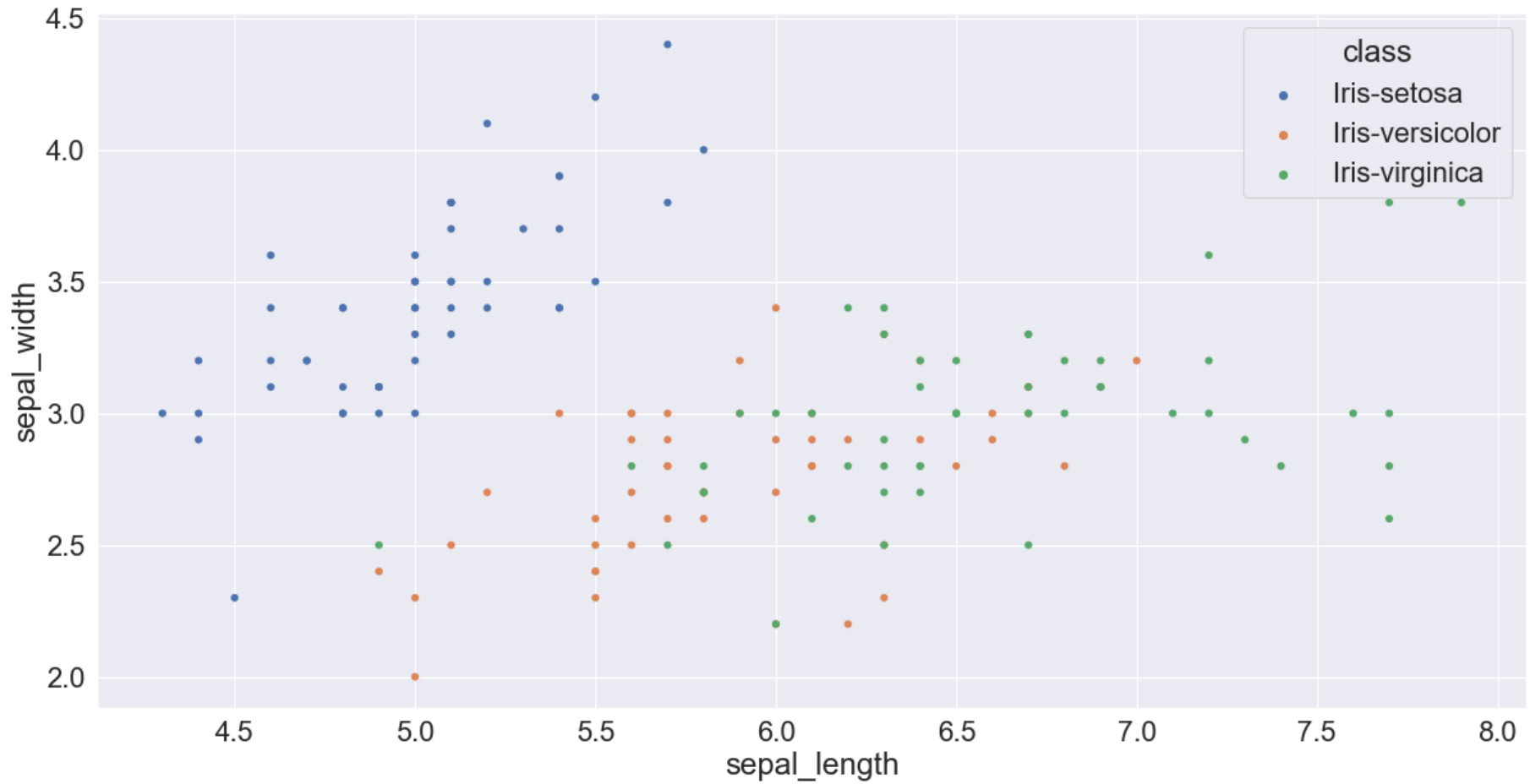
```
In [17]: setosa_df=iris_df[iris_df['class']=='Iris-setosa']

plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.scatterplot(x="sepal_length", y="sepal_width", sizes=(1, 8), linewidth=0,data=setosa_df)
plt.show()
```



Multi-class Scatter plot

```
In [18]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.scatterplot(x="sepal_length", y="sepal_width", sizes=(1, 8), linewidth=0,data=iris_df,hue='class')
plt.show()
```



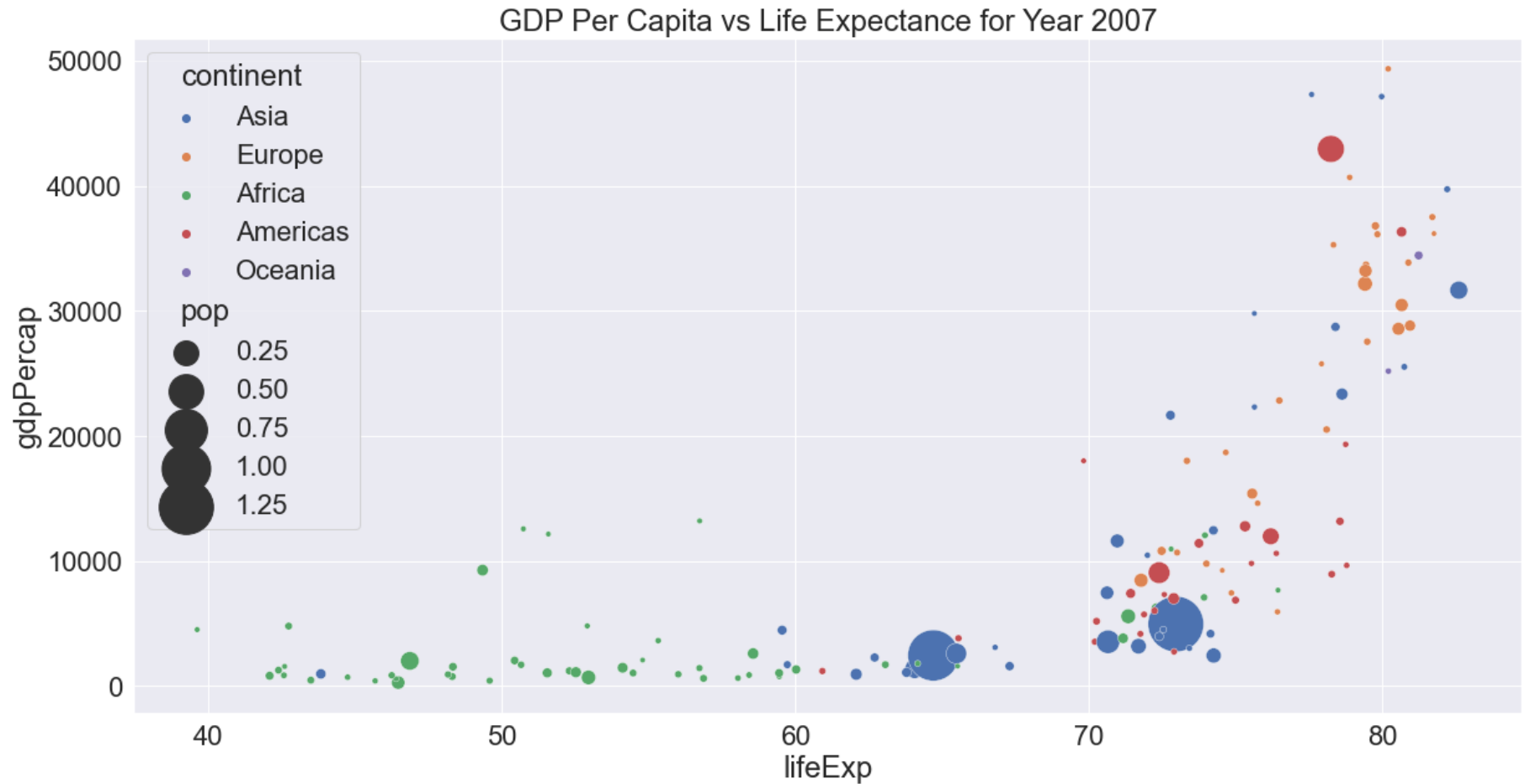
Buble chart

```
In [19]: gdpPercap_df=pd.read_csv('gdpPercap.csv')
gdpPercap_2007_df=gdpPercap_df.query("year=='2007'")
gdpPercap_2007_df.head()
```

Out[19]:

	country	continent	year	lifeExp	pop	gdpPercap
11	Afghanistan	Asia	2007	43.828	31889923	974.580338
23	Albania	Europe	2007	76.423	3600523	5937.029526
35	Algeria	Africa	2007	72.301	33333216	6223.367465
47	Angola	Africa	2007	42.731	12420476	4797.231267
59	Argentina	Americas	2007	75.320	40301927	12779.379640

```
In [20]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
sns.scatterplot(data=gdpPercap_2007_df, x="lifeExp", y="gdpPercap", size='pop', hue='continent',
               legend=True, sizes=(20, 2000))
plt.title('GDP Per Capita vs Life Expectance for Year 2007')
plt.show()
```



Pie Chart

```
In [21]: gdpPercap_df=pd.read_csv('gdpPercap.csv')
gdpPercap_2007_df=gdpPercap_df.query("year=='2007'")
gdpPercap_2007_df.head()
```

Out[21]:

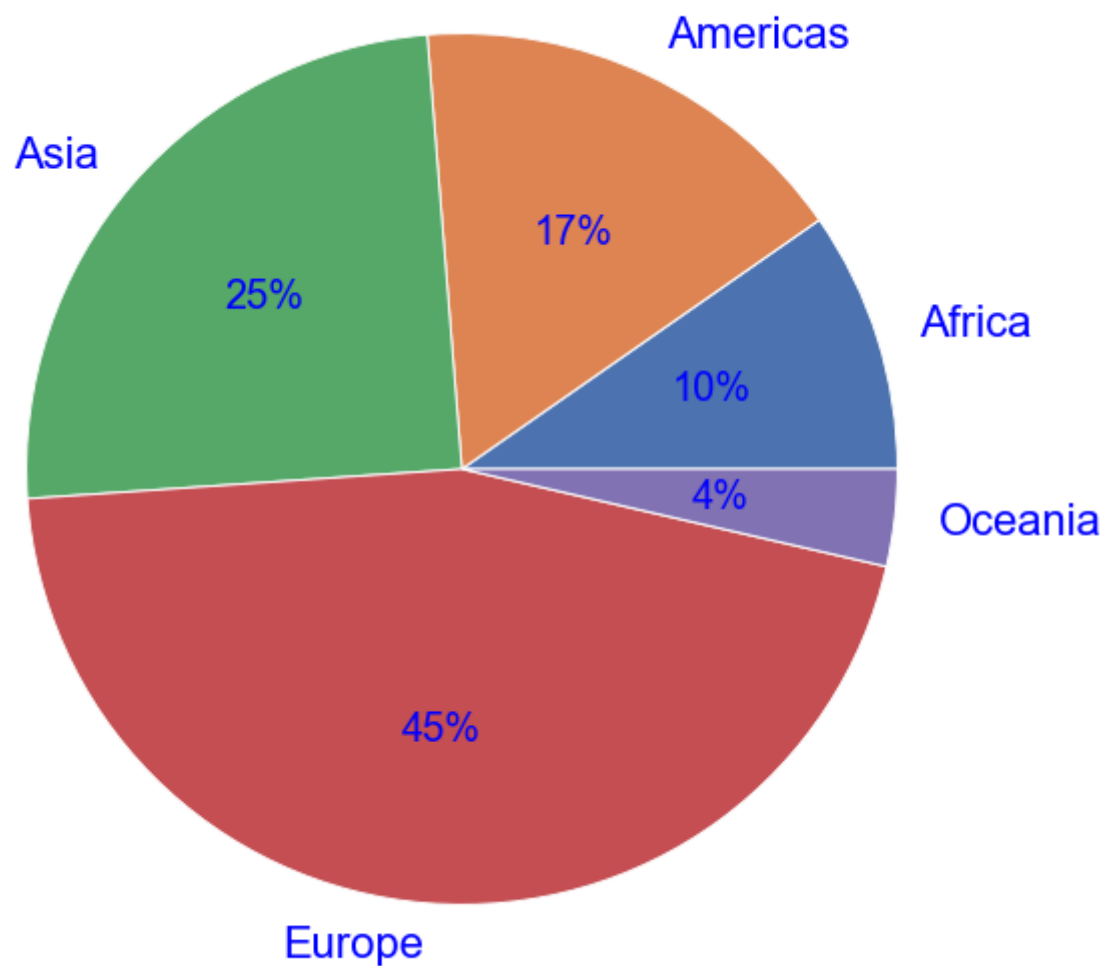
	country	continent	year	lifeExp	pop	gdpPercap
11	Afghanistan	Asia	2007	43.828	31889923	974.580338
23	Albania	Europe	2007	76.423	3600523	5937.029526
35	Algeria	Africa	2007	72.301	33333216	6223.367465
47	Angola	Africa	2007	42.731	12420476	4797.231267
59	Argentina	Americas	2007	75.320	40301927	12779.379640

```
In [22]: gdpPercap_2007_df=pd.DataFrame(gdpPercap_2007_df.groupby('continent')['gdpPercap'].sum())
gdpPercap_2007_df
```

Out[22]:

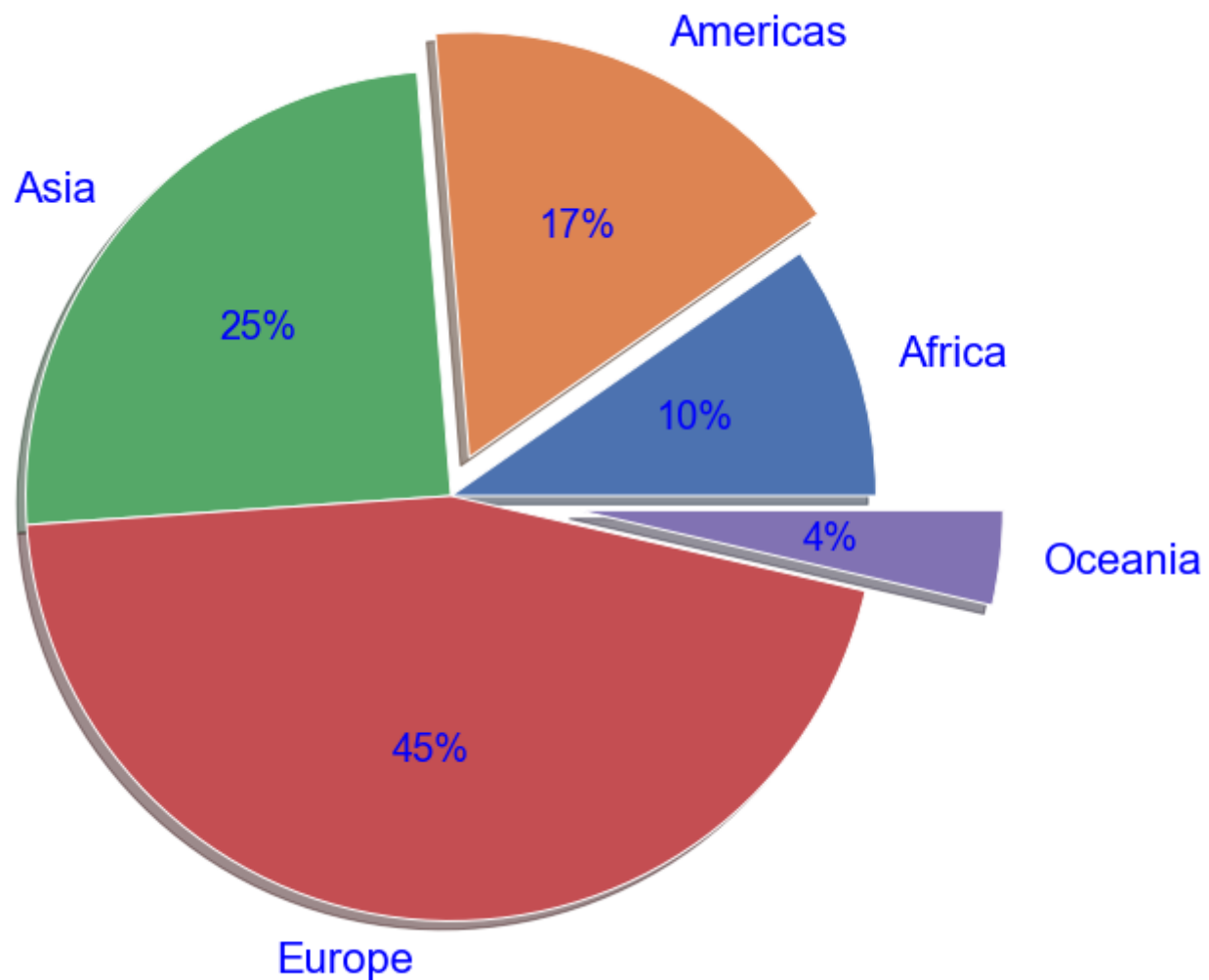
	gdpPercap
continent	
Africa	160629.695446
Americas	275075.790634
Asia	411609.886714
Europe	751634.449078
Oceania	59620.376550

```
In [23]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20
plt.pie(gdpPercap_2007_df['gdpPercap'], labels =gdpPercap_2007_df.index, autopct='%.0f%%')
plt.show()
```



Explode Pie

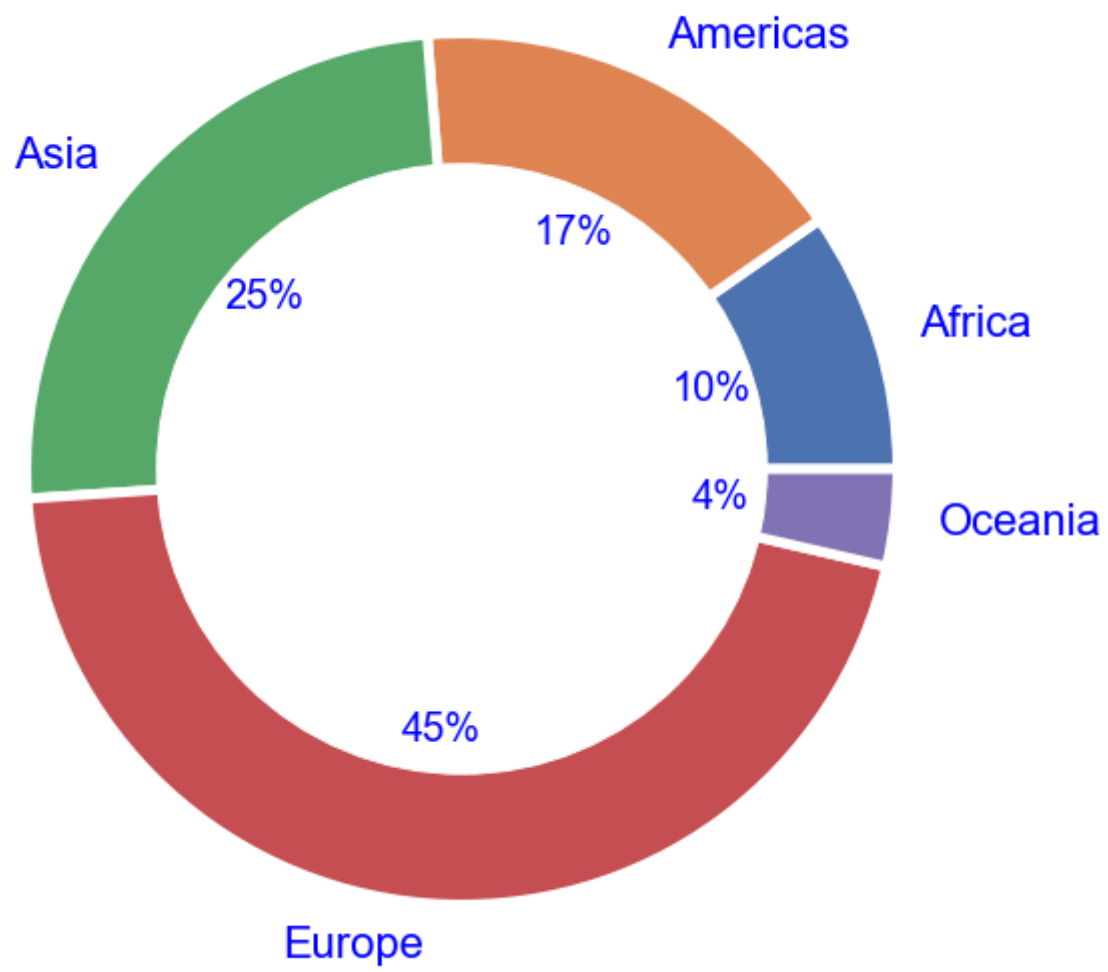

```
In [24]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20
plt.pie(gdpPercap_2007_df['gdpPercap'], labels =gdpPercap_2007_df.index, autopct='%.0f%%',
        explode=(0,0.1,0,0,0.3),shadow=True)
plt.show()
```



Donought

```
In [25]: inner_circle = plt.Circle( (0,0), 0.7, color='white')

plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20
plt.pie(gdpPercap_2007_df['gdpPercap'], labels =gdpPercap_2007_df.index, autopct='%.0f%%',
        wedgeprops = { 'linewidth' : 5, 'edgecolor' : 'white' })
p = plt.gcf()
p.gca().add_artist(inner_circle)
plt.show()
```



Box Plot

```
In [26]: gdpPercap_df=pd.read_csv('gdpPercap.csv')
gdpPercap_2007_df=gdpPercap_df.query("year=='2007'")
gdpPercap_2007_df.head()
```

Out[26]:

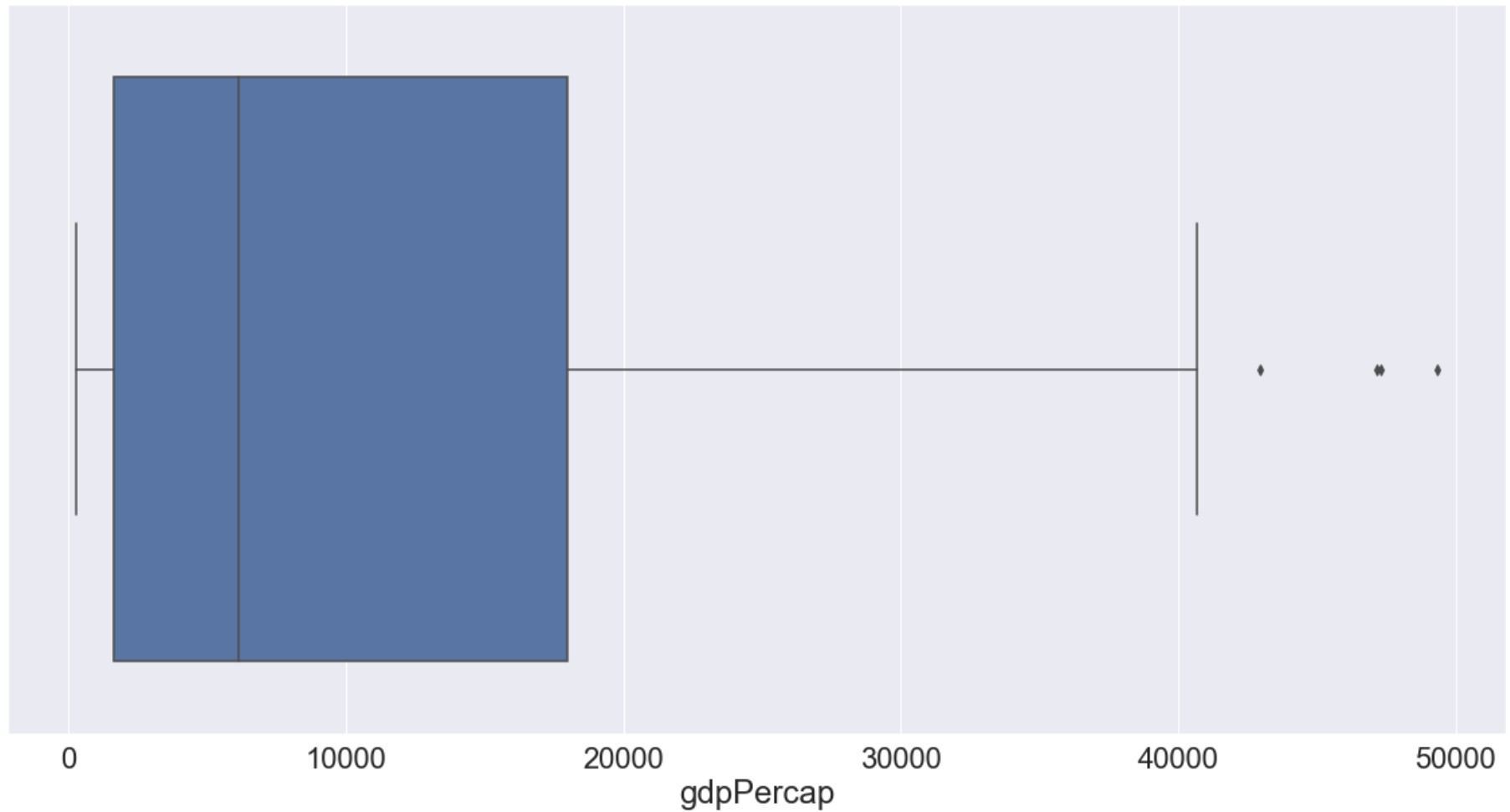
	country	continent	year	lifeExp	pop	gdpPercap
11	Afghanistan	Asia	2007	43.828	31889923	974.580338
23	Albania	Europe	2007	76.423	3600523	5937.029526
35	Algeria	Africa	2007	72.301	33333216	6223.367465
47	Angola	Africa	2007	42.731	12420476	4797.231267
59	Argentina	Americas	2007	75.320	40301927	12779.379640

Simple Boxplot

```
In [27]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

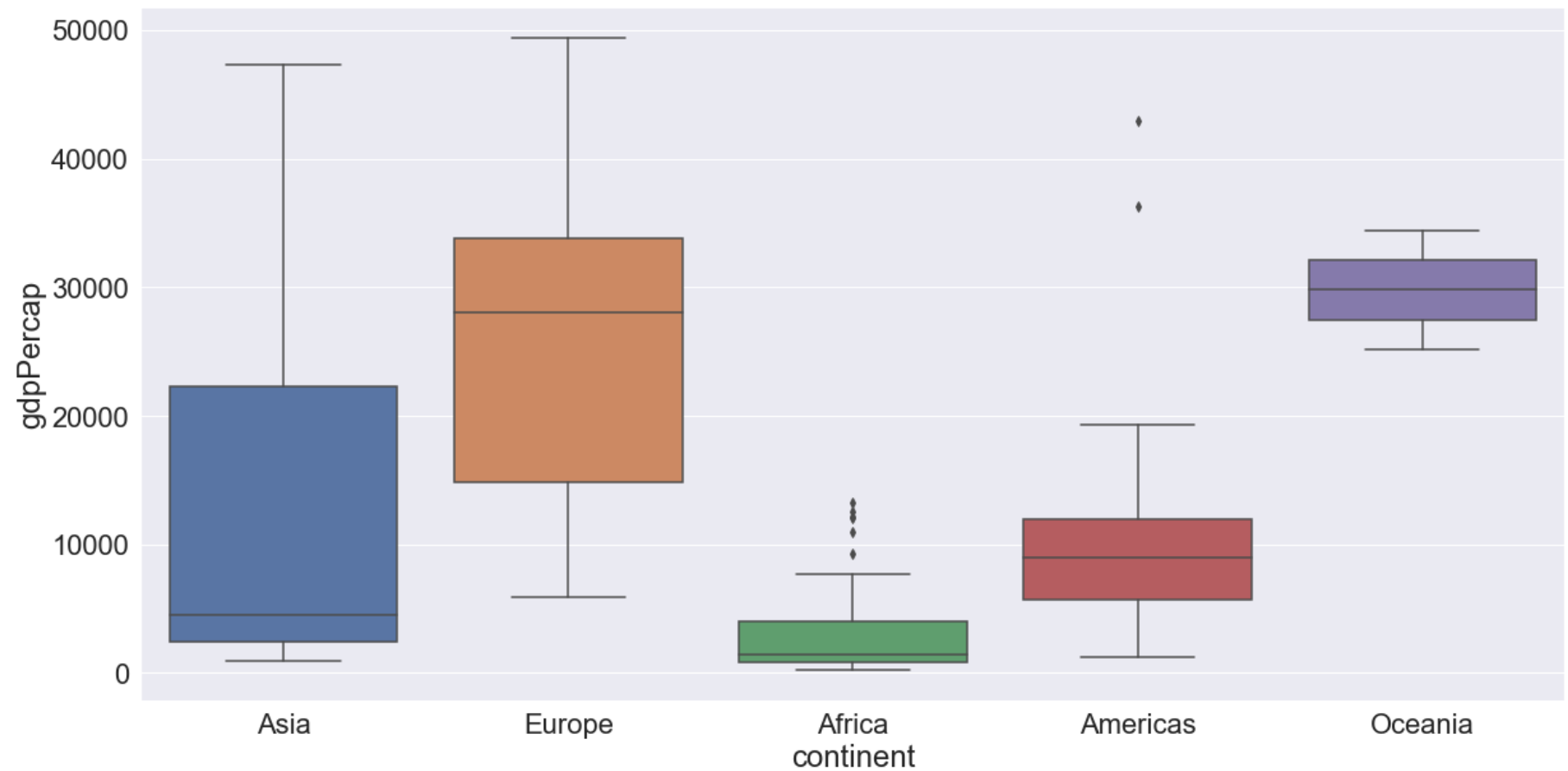
sns.boxplot(gdpPercap_2007_df['gdpPercap'])
plt.show()
```

C:\Users\soongaya\Anaconda3\lib\site-packages\seaborn_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
FutureWarning



Multiple Boxplots

```
In [28]: sns.boxplot(x='continent', y='gdpPercap', data=gdpPercap_2007_df)
plt.show()
```



Heatmap


```
In [29]: gdpPercap_df=pd.read_csv('gdpPercap.csv')
gdpPercap_heatmap_df=gdpPercap_df.pivot('year','country','gdpPercap')
gdpPercap_heatmap_df.iloc[0:,0:12]
```

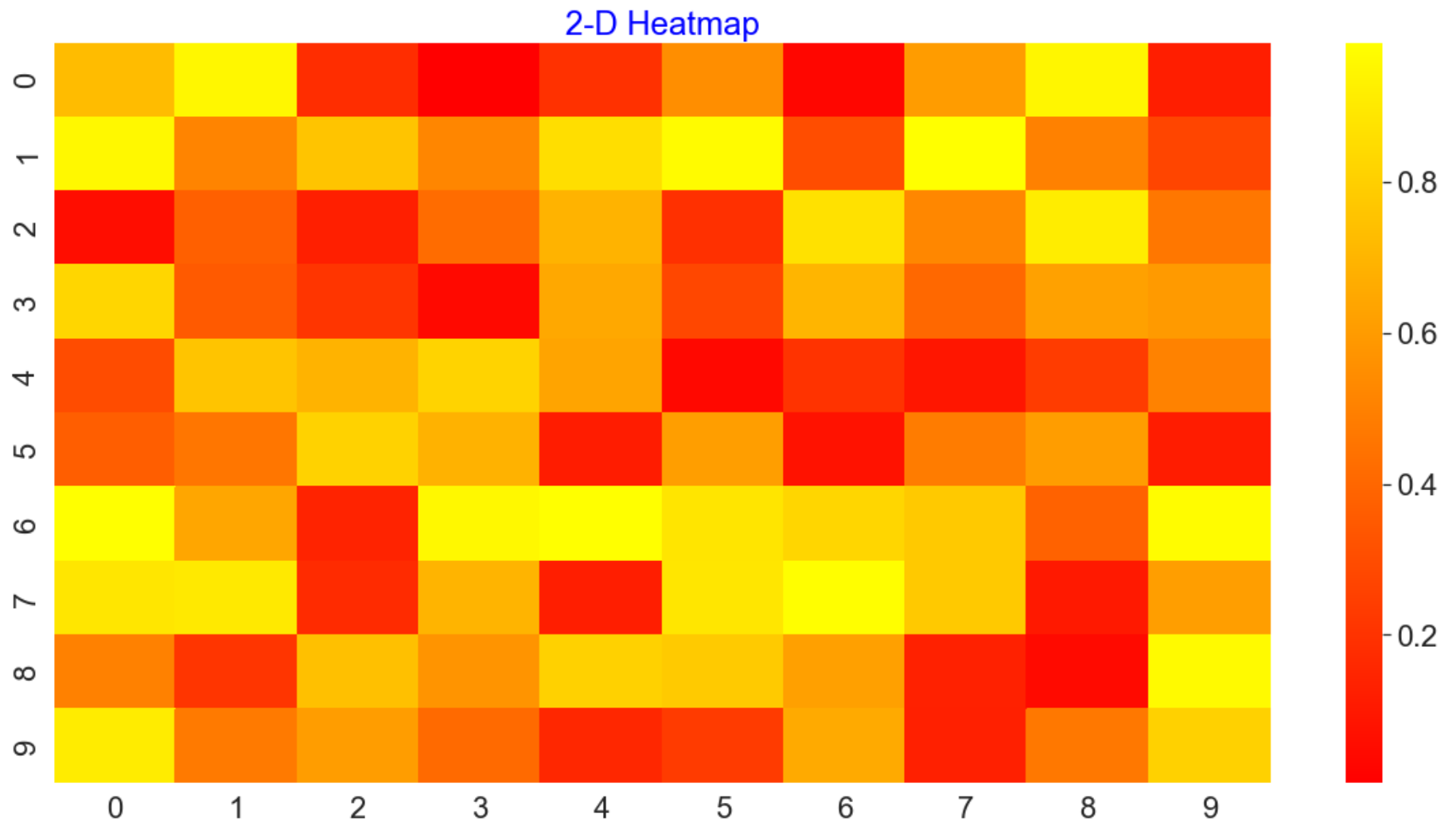
Out[29]:

country	Afghanistan	Albania	Algeria	Angola	Argentina	Australia	Austria	Bahrain	Bangladesh
year									
1952	779.445314	1601.056136	2449.008185	3520.610273	5911.315053	10039.59564	6137.076492	9867.084765	684.244172
1957	820.853030	1942.284244	3013.976023	3827.940465	6856.856212	10949.64959	8842.598030	11635.799450	661.637458
1962	853.100710	2312.888958	2550.816880	4269.276742	7133.166023	12217.22686	10750.721110	12753.275140	686.341554
1967	836.197138	2760.196931	3246.991771	5522.776375	8052.953021	14526.12465	12834.602400	14804.672700	721.186086
1972	739.981106	3313.422188	4182.663766	5473.288005	9443.038526	16788.62948	16661.625600	18268.658390	630.233627
1977	786.113360	3533.003910	4910.416756	3008.647355	10079.026740	18334.19751	19749.422300	19340.101960	659.877232
1982	978.011439	3630.880722	5745.160213	2756.953672	8997.897412	19477.00928	21597.083620	19211.147310	676.981866
1987	852.395945	3738.932735	5681.358539	2430.208311	9139.671389	21888.88903	23687.826070	18524.024060	751.979403
1992	649.341395	2497.437901	5023.216647	2627.845685	9308.418710	23424.76683	27042.018680	19035.579170	837.810164
1997	635.341351	3193.054604	4797.295051	2277.140884	10967.281950	26997.93657	29095.920660	20292.016790	972.770035
2002	726.734055	4604.211737	5288.040382	2773.287312	8797.640716	30687.75473	32417.607690	23403.559270	1136.390430
2007	974.580338	5937.029526	6223.367465	4797.231267	12779.379640	34435.36744	36126.492700	29796.048340	1391.253792



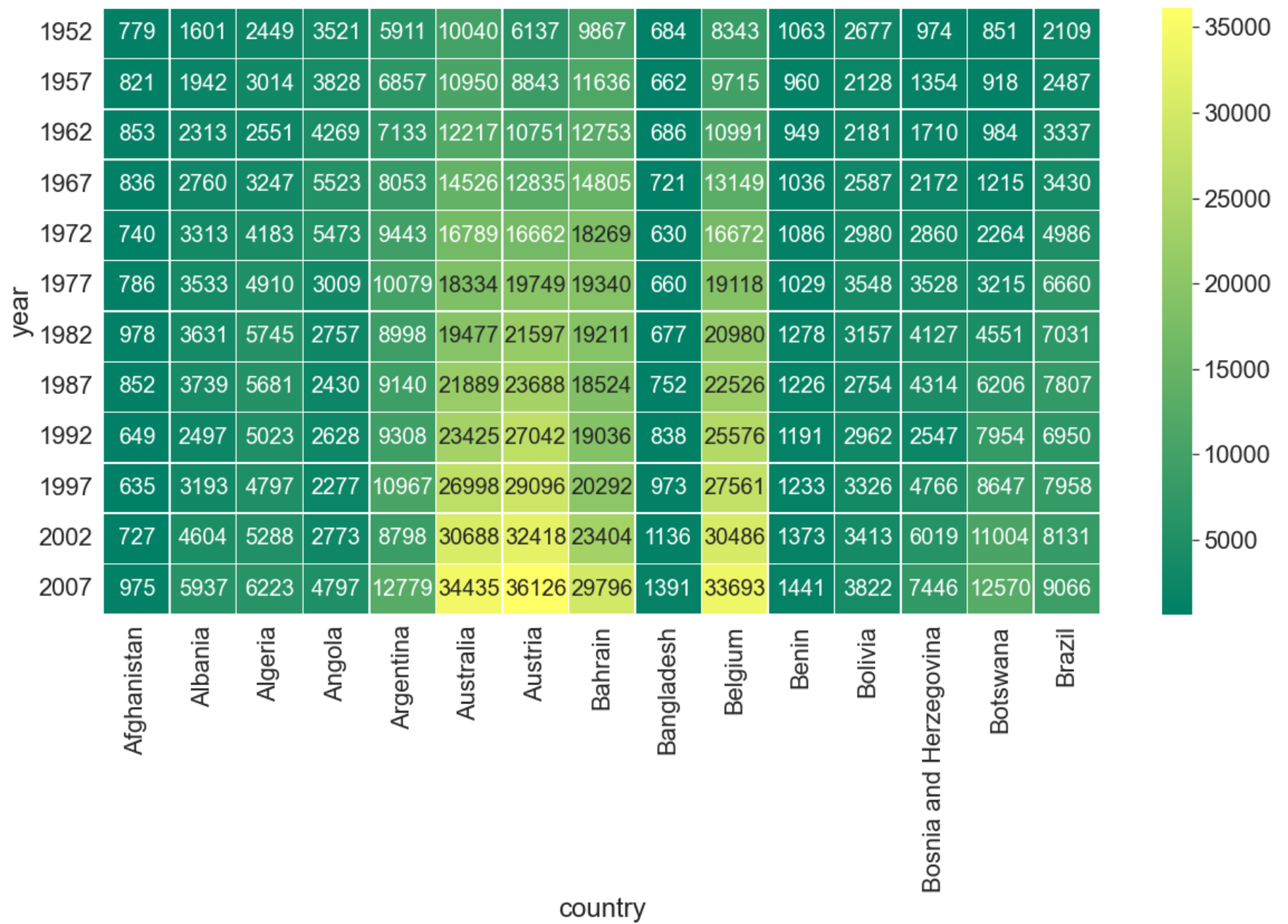
```
In [30]: heatmap_data = np.random.random(( 10 , 10 ))
```

```
plt.rcParams['axes.labelsize'] = 20  
sns.set(font_scale = 2)  
plt.rcParams['text.color'] = 'blue'  
plt.rcParams['font.size'] = 20  
sns.heatmap(heatmap_data, cmap='autumn')  
plt.title('2-D Heatmap')  
plt.show()
```



Heatmap with DataFrame

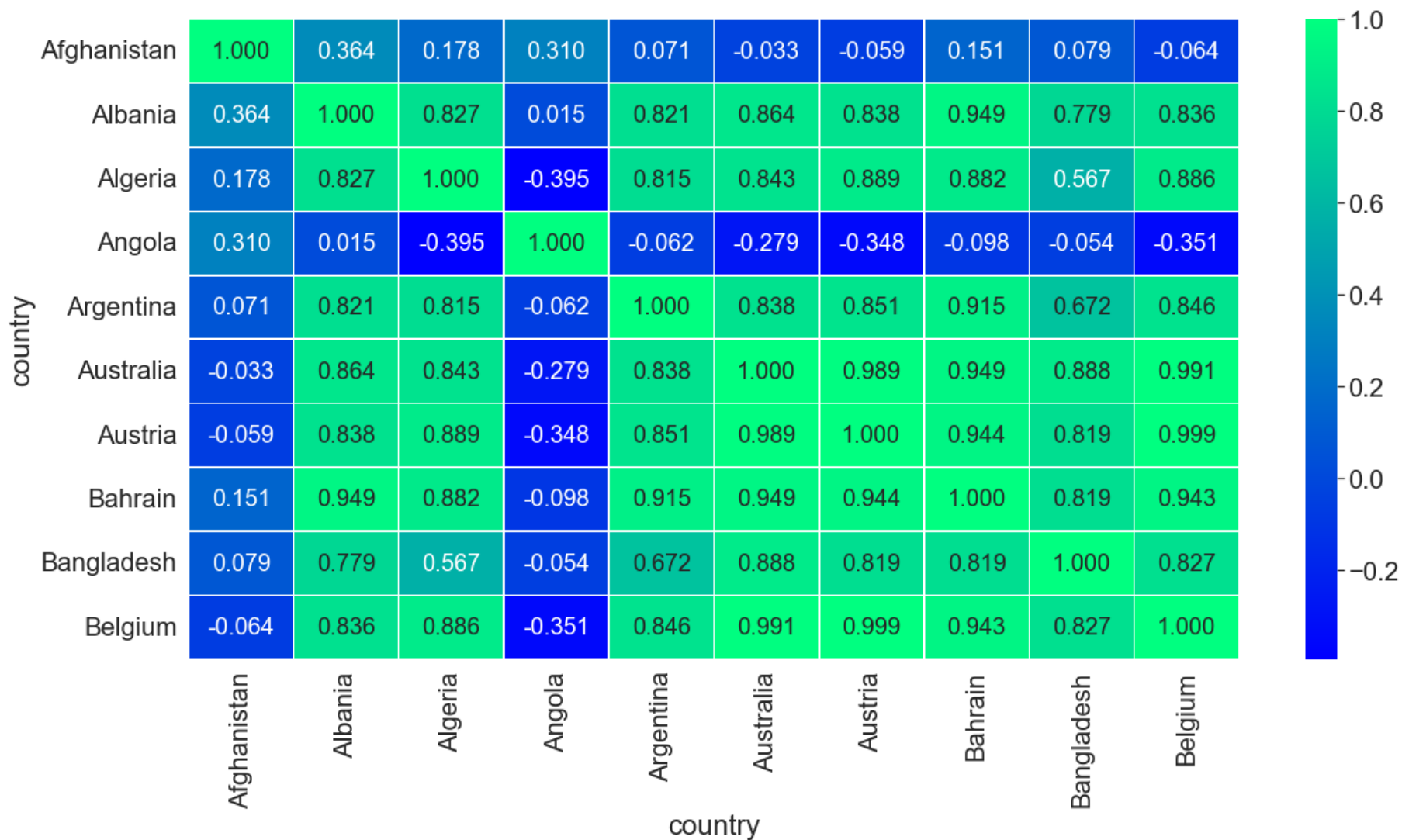
```
In [31]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20
sns.heatmap(gdpPercap_heatmap_df.iloc[0:,0:15], annot=True, fmt=".0f", linewidths=.5,cmap='summer')
plt.show()
```



Correlation Heatmap

```
In [32]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

cmap = sns.diverging_palette(230, 20, as_cmap=True)
sns.heatmap(gdpPercap_heatmap_df.corr().iloc[0:10,0:10], annot=True, fmt=".3f", linewidths=.5,cmap='winter')
plt.show()
```



Grouped Visualizations

Pairplot

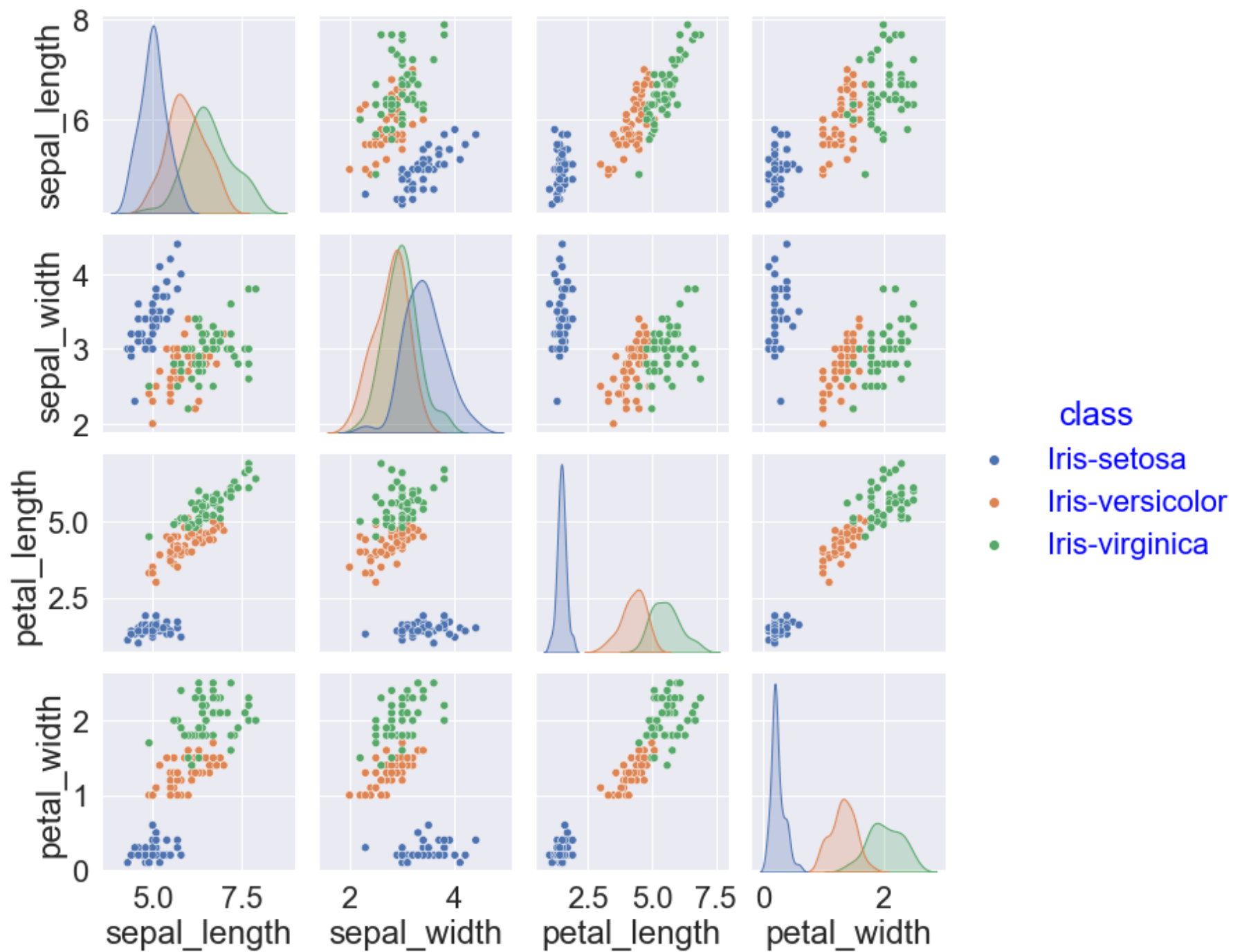
```
In [33]: iris_df=pd.read_csv('iris.csv')
iris_df.head()
```

Out[33]:

	sepal_length	sepal_width	petal_length	petal_width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa


```
In [34]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

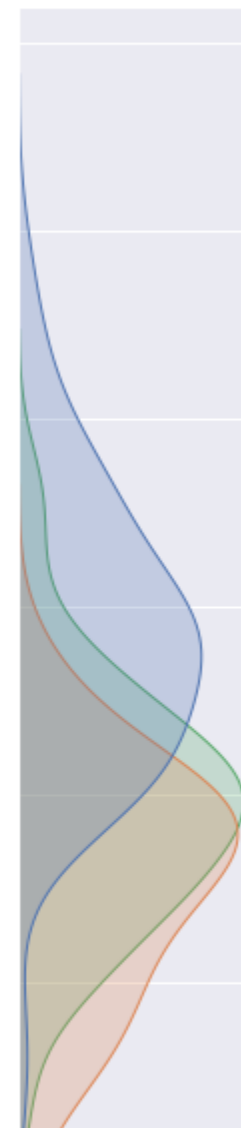
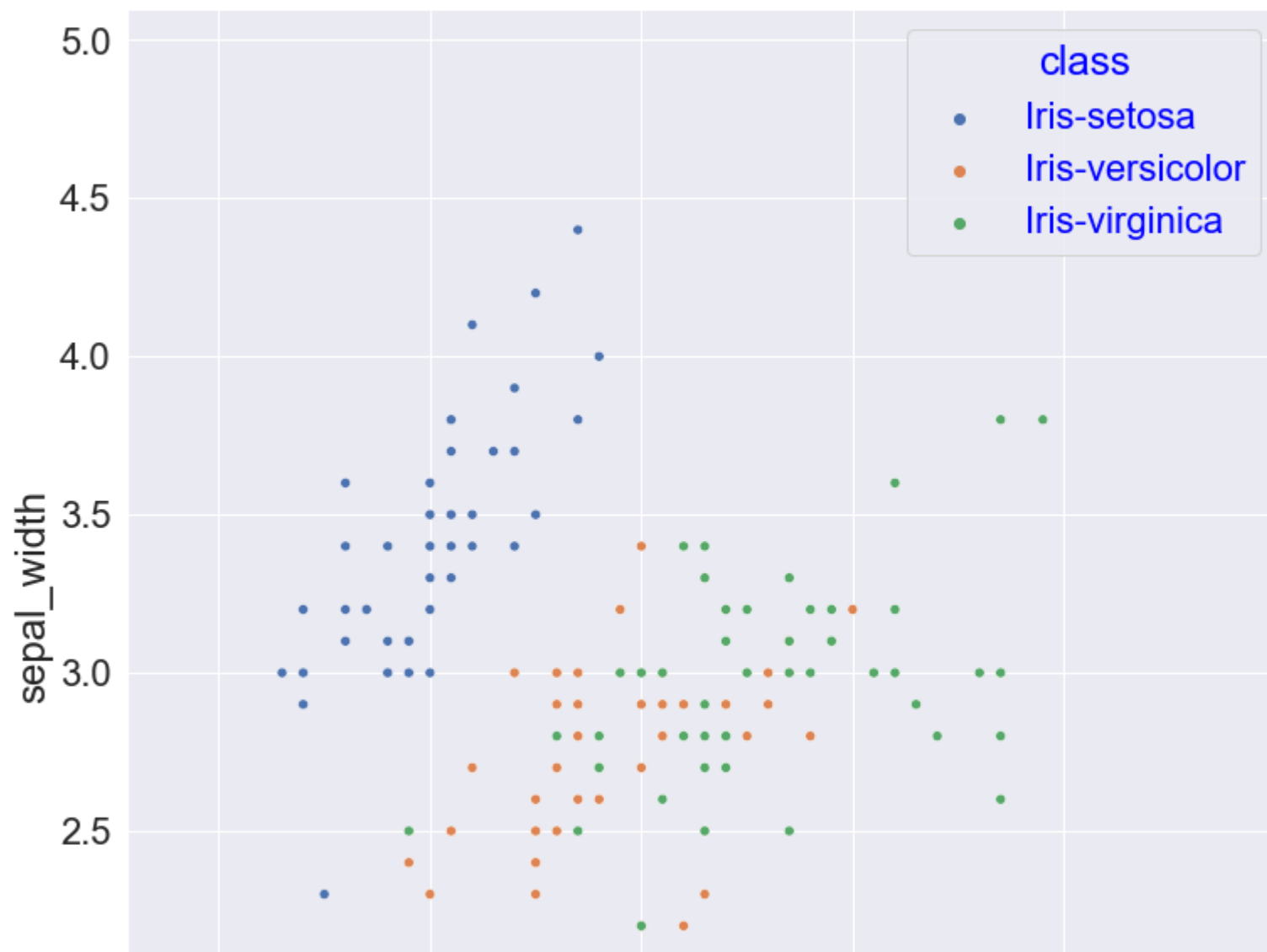
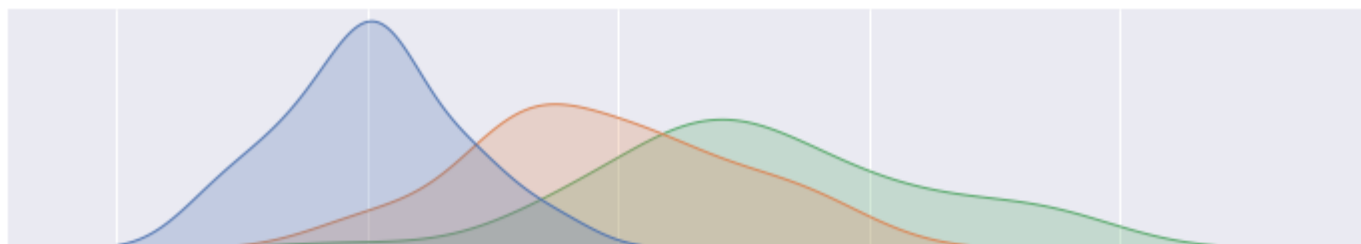
sns.pairplot(iris_df, hue='class')
plt.show()
```

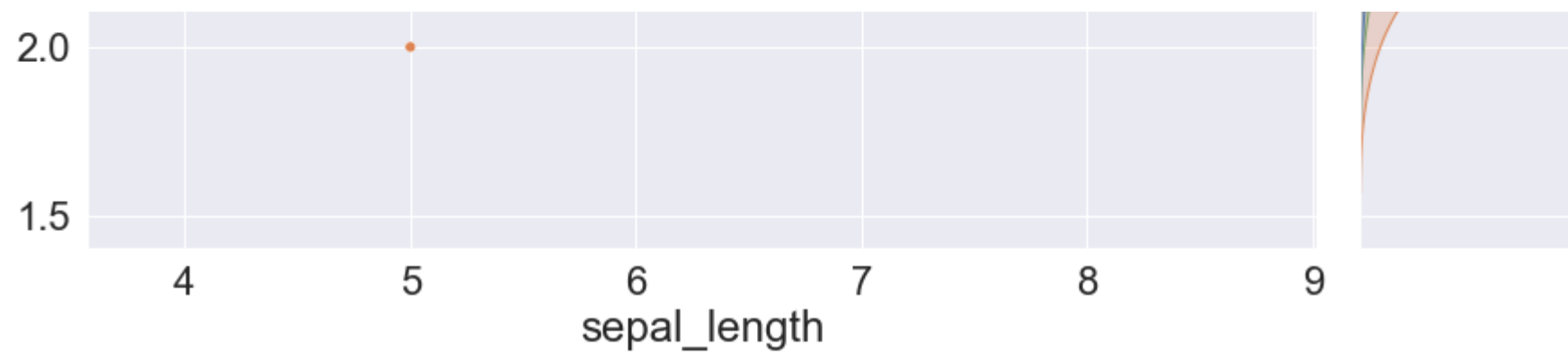


Joint plot

```
In [49]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

sns.jointplot(data=iris_df,x='sepal_length',y='sepal_width',hue='class',height=13)
plt.show()
```



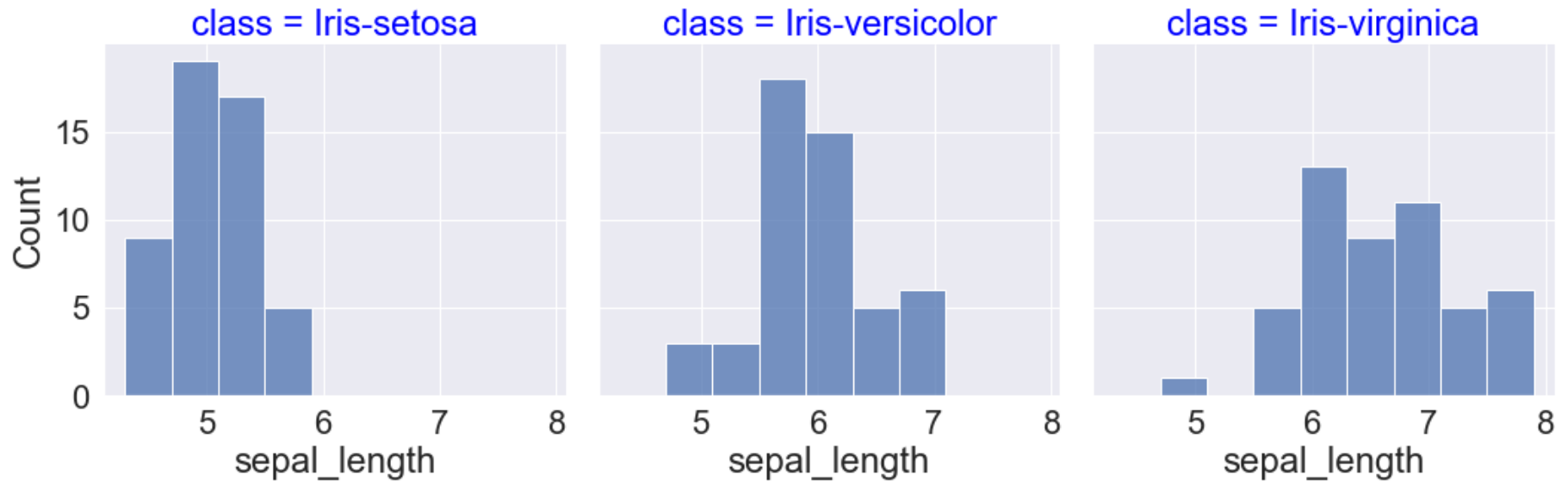


Distplot

```
In [60]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

sns.displot(iris_df, x="sepal_length", col="class", height=5, facet_kws=dict(margin_titles=True))
plt.plot()
```

Out[60]: []



Distribtuion Plot

```
In [35]: iris_df=pd.read_csv('iris.csv')
iris_df.head()
```

Out[35]:

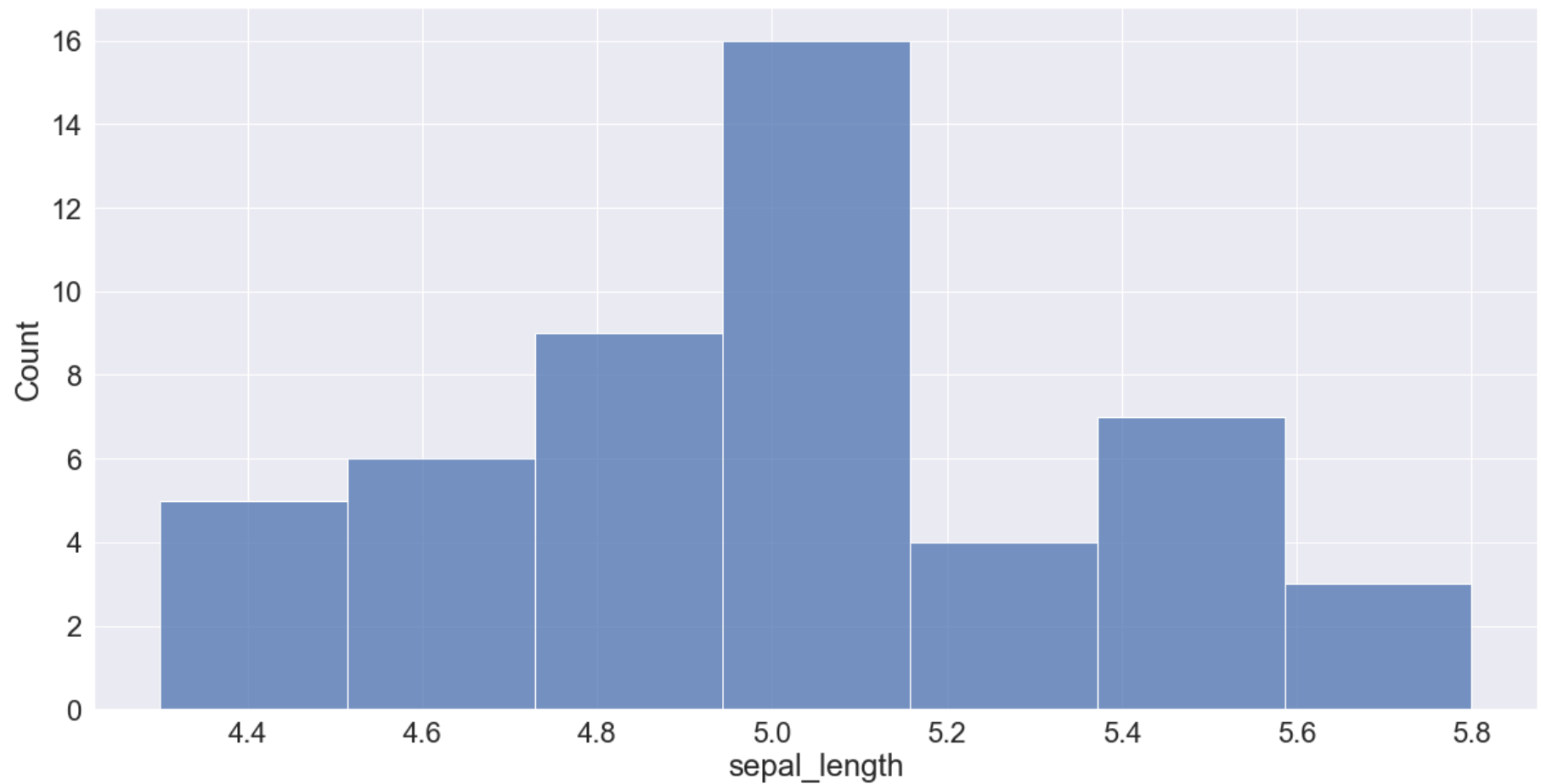
	sepal_length	sepal_width	petal_length	petal_width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

Histogram

Plotting univariate histogram


```
In [36]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

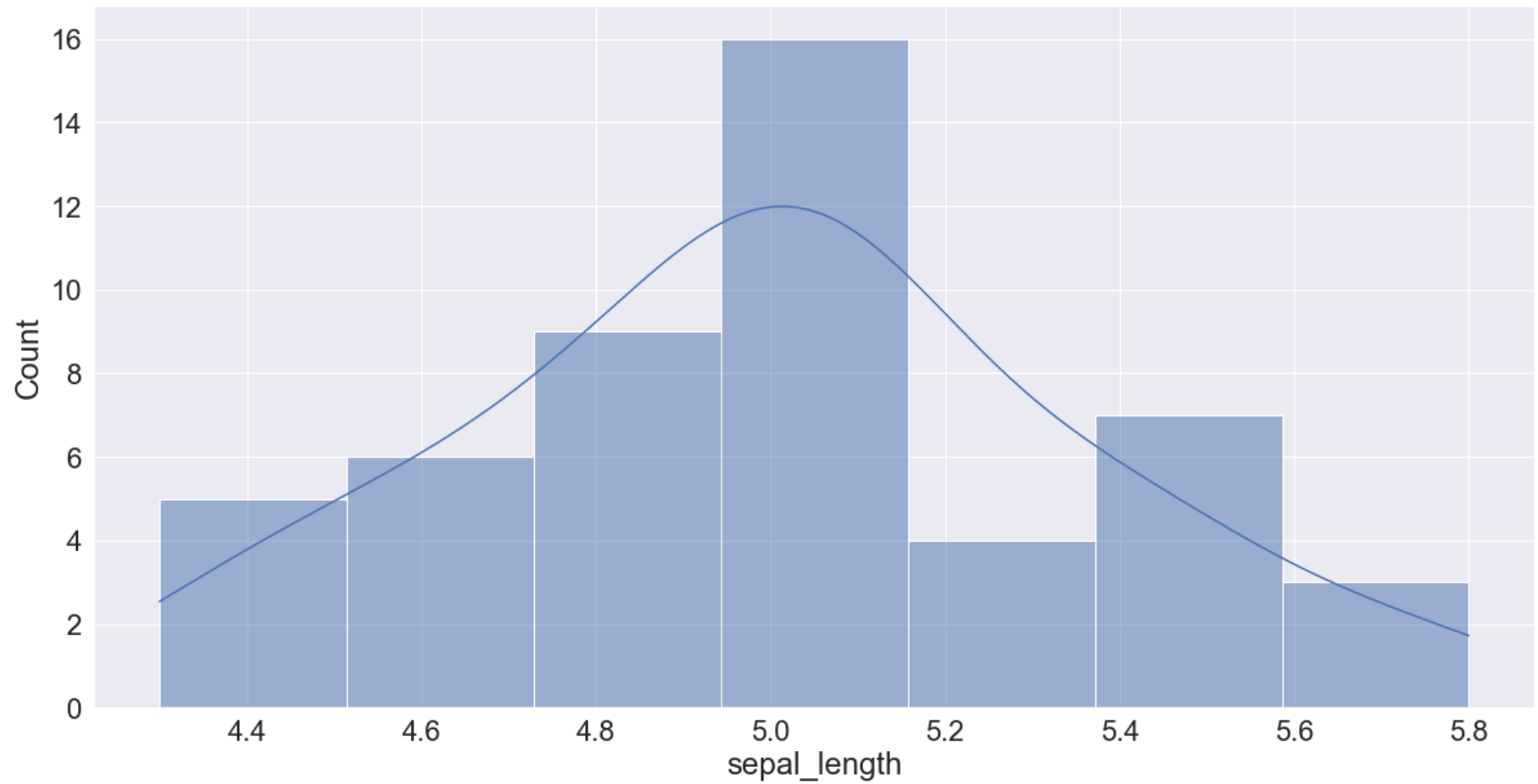
sns.histplot(iris_df[iris_df['class']=='Iris-setosa'], x='sepal_length')
plt.show()
```



Adding kernel density estimate (kde) to histogram

```
In [37]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

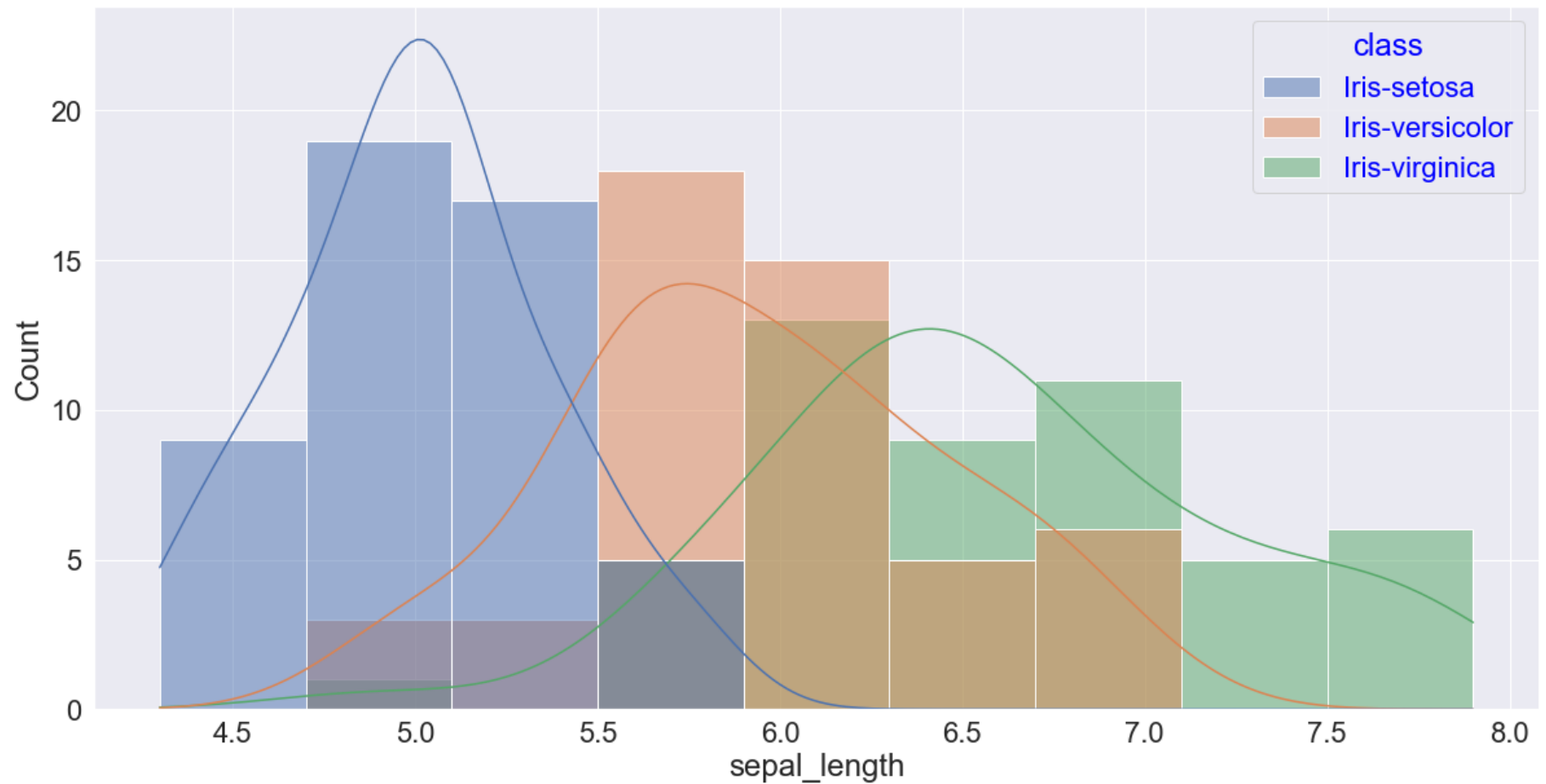
sns.histplot(iris_df[iris_df['class']=='Iris-setosa'], x='sepal_length',kde=True)
plt.show()
```



Plotting multivariate histogram

```
In [38]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

sns.histplot(iris_df, x='sepal_length', kde=True, hue='class')
plt.show()
```



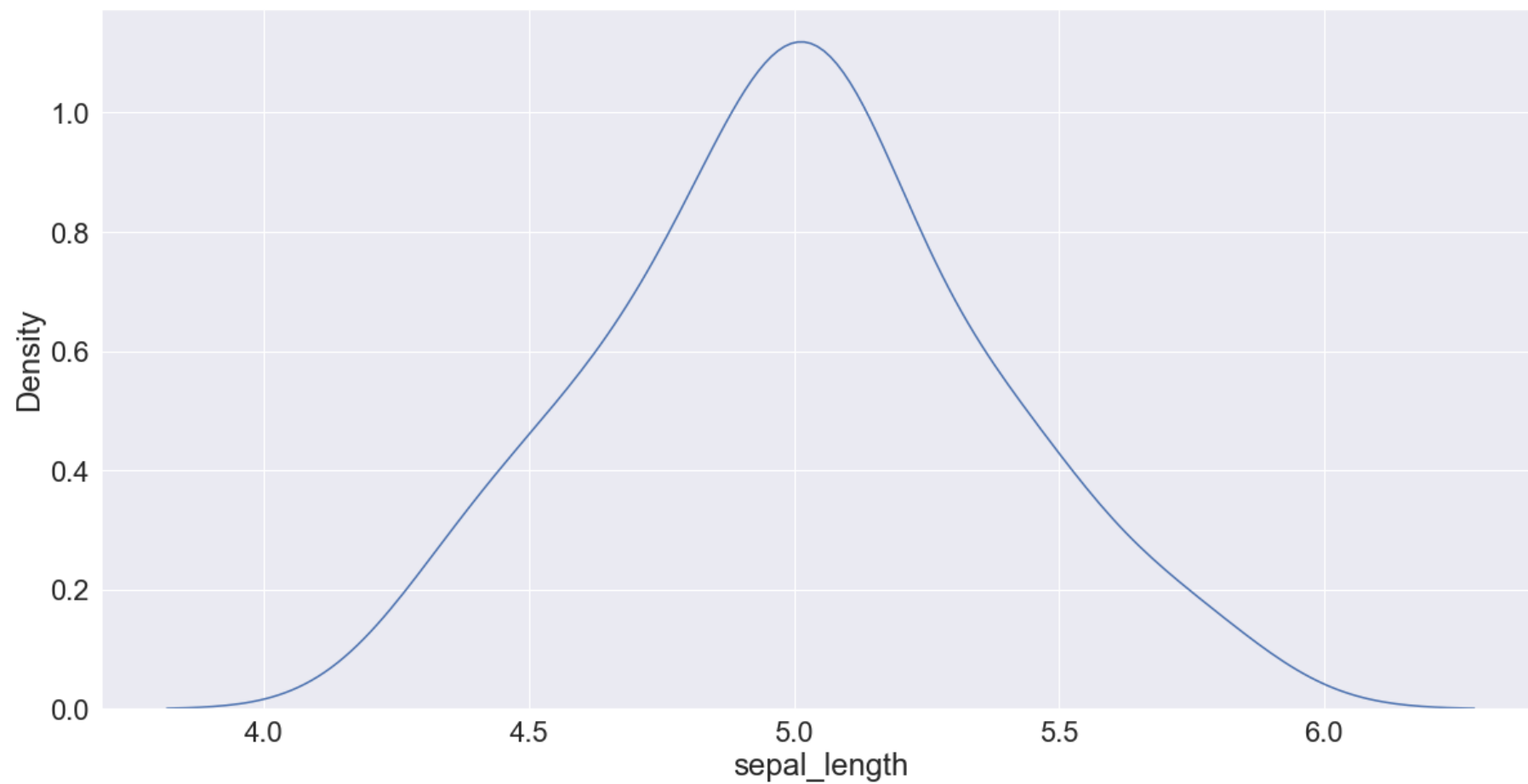
kernel density estimate (KDE) plot

A kernel density estimate (KDE) plot is a method for visualizing the distribution of observations in a dataset, analagous to a histogram. KDE represents the data using a continuous probability density curve in one or more dimensions.

Simple univariate kde distribution

```
In [39]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

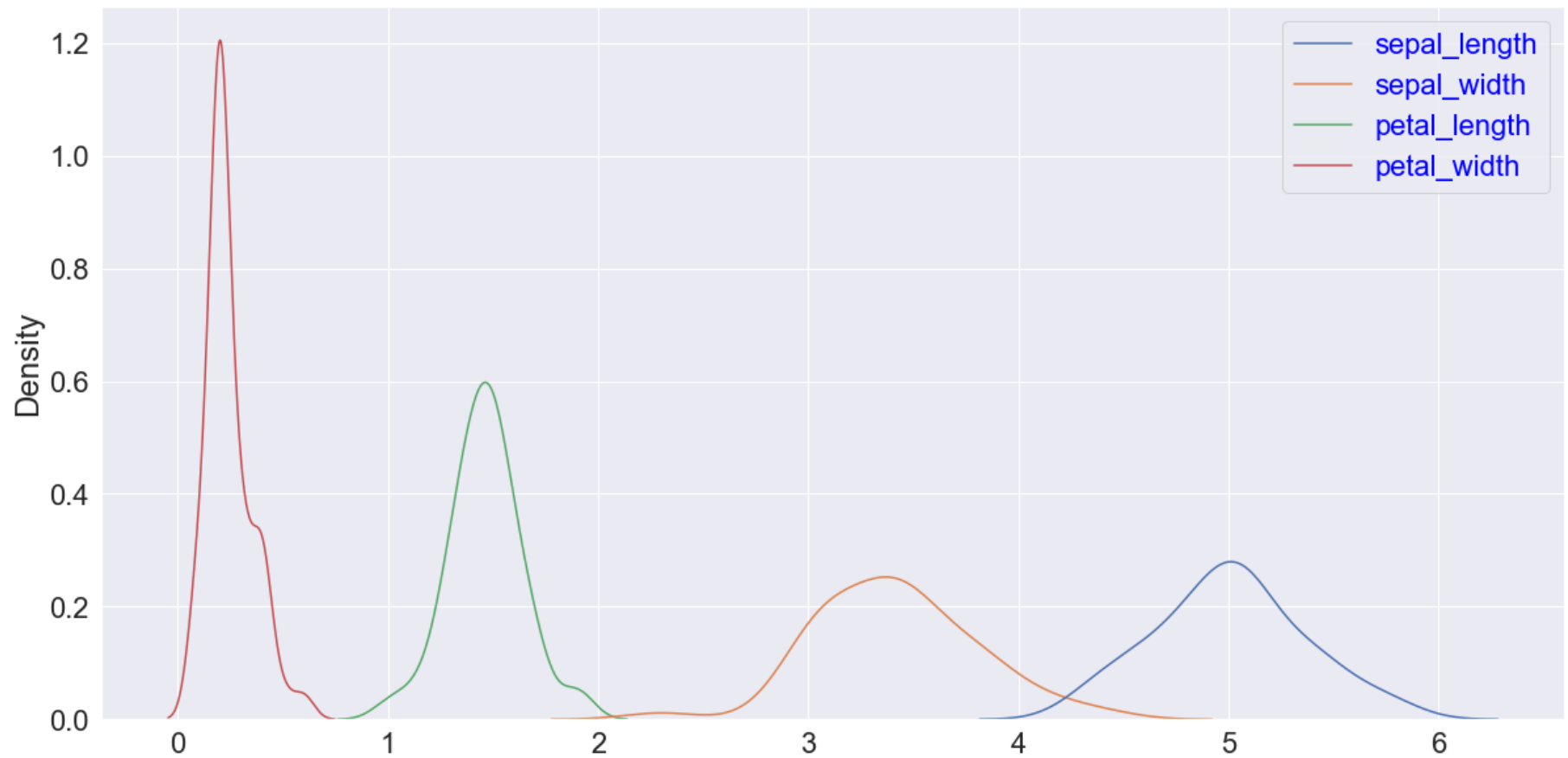
sns.kdeplot(data=iris_df[iris_df['class']=='Iris-setosa'], x='sepal_length')
plt.show()
```



kde plot for all variables

```
In [40]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

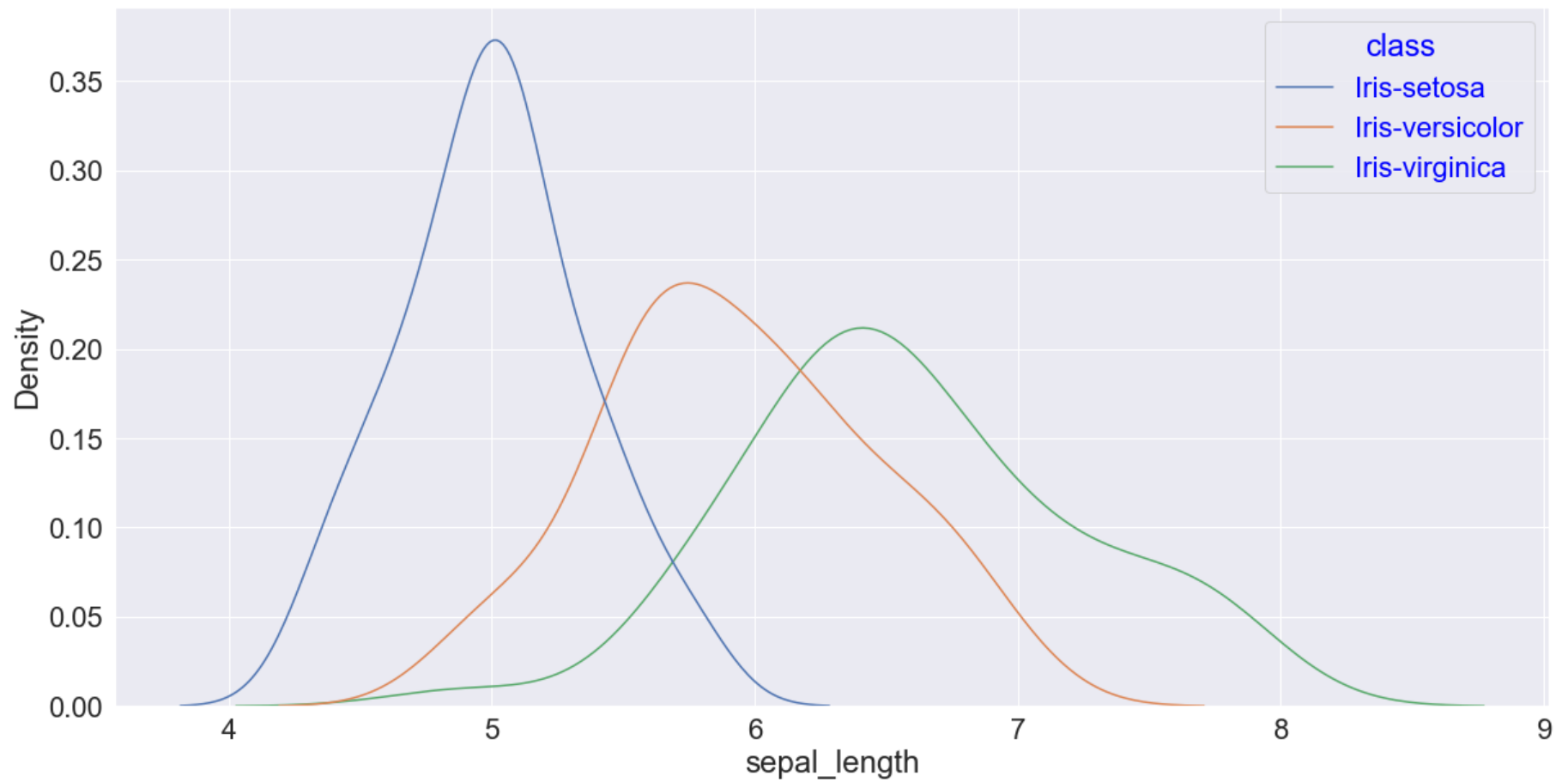
sns.kdeplot(data=iris_df[iris_df['class']=='Iris-setosa'])
plt.show()
```



kde plot for different species

```
In [41]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

sns.kdeplot(data=iris_df, x='sepal_length', hue='class')
plt.show()
```



kde plot with color fill

```
In [42]: plt.rcParams['axes.labelsize'] = 20
sns.set(font_scale = 2)
plt.rcParams['text.color'] = 'blue'
plt.rcParams['font.size'] = 20

sns.kdeplot(data=iris_df, x="sepal_length", hue="class",
            fill=True, common_norm=False, palette="crest",
            alpha=.5, linewidth=0,)
plt.show()
```

