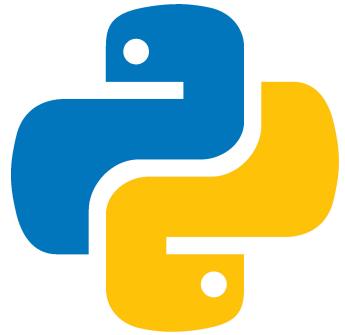


Generic data visualization tools for Python

By Vincent Roger

11/23/2023





About me

PhD in computer science.

Kiviak instrument's data scientist and machine learning guy.

LinkedIn

<https://www.linkedin.com/in/vroger11/>

Website

<https://website.vincent-roger.fr/>

Why data visualization with Python?

- Extract and prepare data is simplified with tools such as Pandas, Dask and others.
- It is a goto language for data scientists, machine learning and MLOPS people.
- So why learn a new language just for data visualization?

What types of tools

Specific (to create graphs)

- NetworkX
- Graph-tool
- PyGraphviz → Graphviz
- Pyvis

Generic

Focus of this presentation!



Plan

- Non web-oriented
- Web-oriented

Data we will use

```
import pandas as pd  
  
df_iris = pd.read_csv('https://raw.githubusercontent.com/mwaskom/seaborn-data/master/iris.csv')  
  
df_iris
```

	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
..
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

[150 rows x 5 columns]

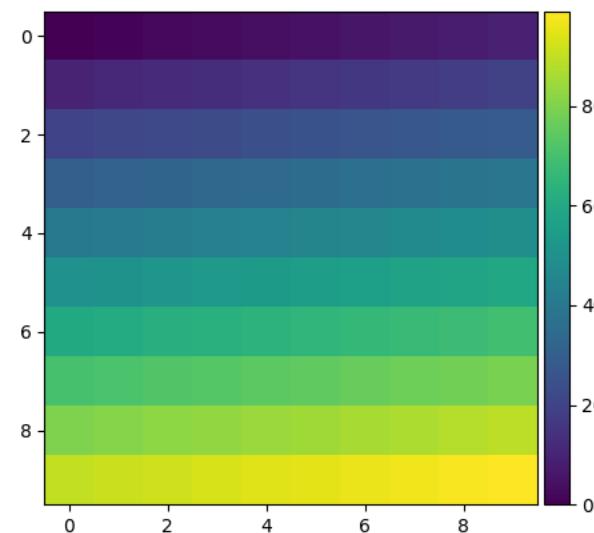
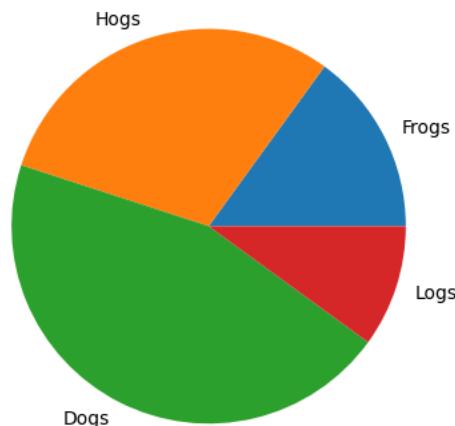
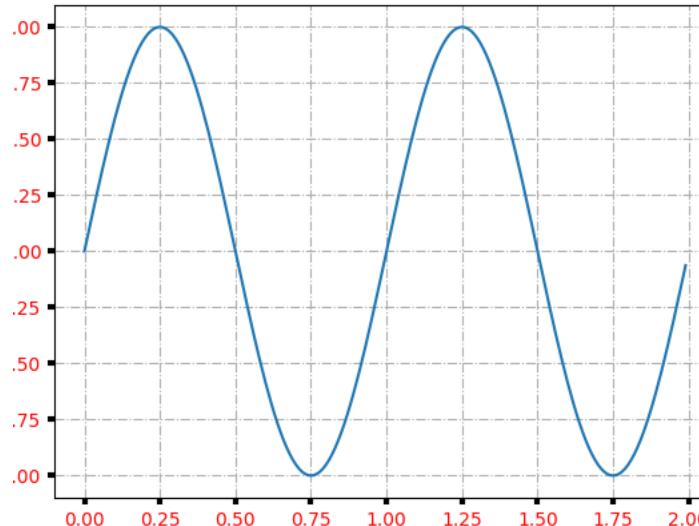


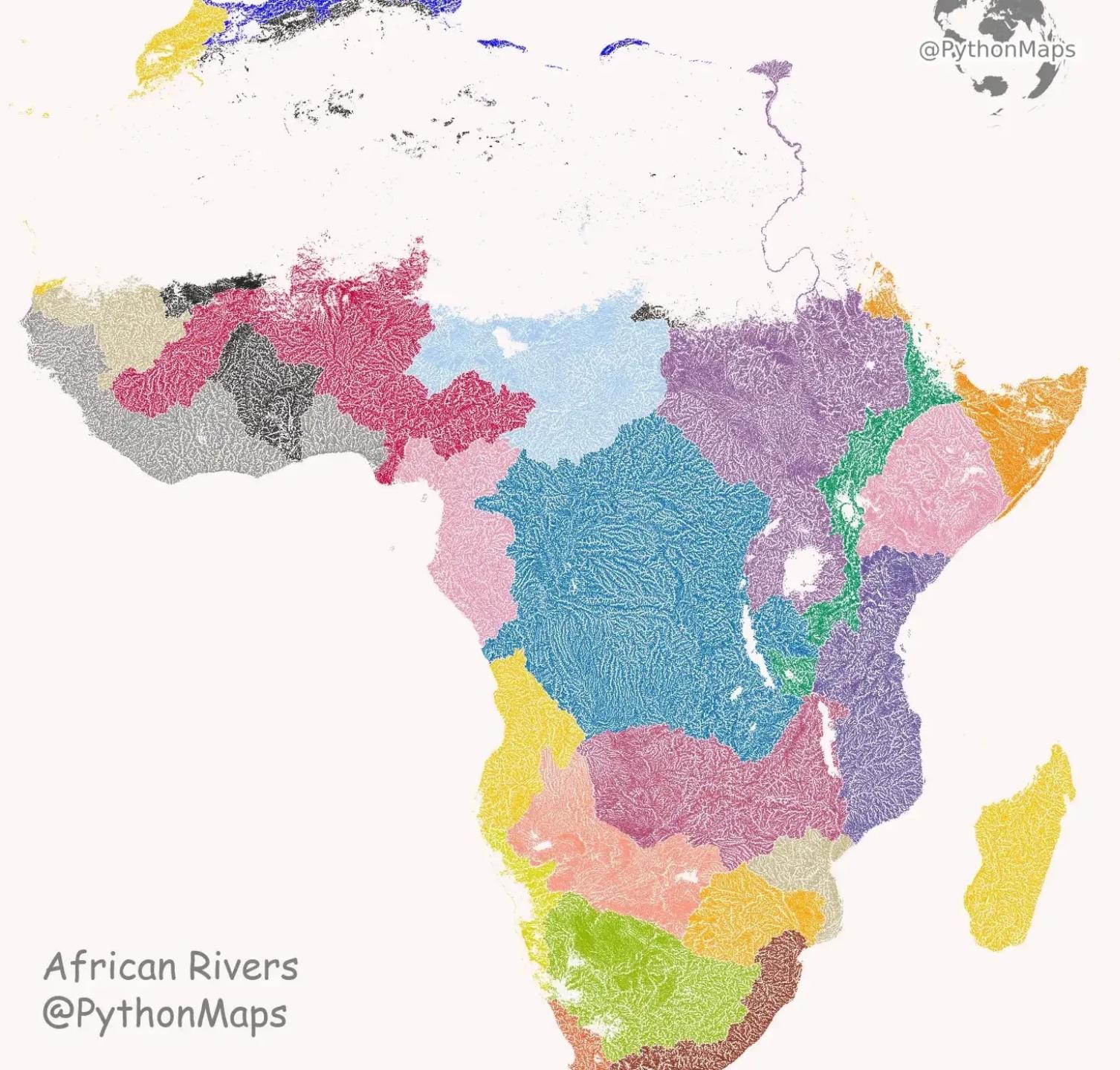


Non web-oriented

Matplotlib

- Based on Qt.
- Standard library.
- Default for multiple tools like Pandas.





African Rivers
@PythonMaps



Matplotlib

- Powerful if mastered!

Matplotlib - scatter plot of iris sepal

```
import matplotlib.pyplot as plt

# Create a scatter plot with legends based on species
plt.figure(figsize=(8, 6))

# Iterate over each species and plot scatter points
for species in df_iris["species"].unique():
    species_data = df_iris[df_iris['species'] == species]
    plt.scatter(species_data['sepal_length'], species_data['sepal_width'], label=species)

# Add labels and title
plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
plt.title('Scatter Plot of Iris Dataset')

# Add legend
plt.legend()

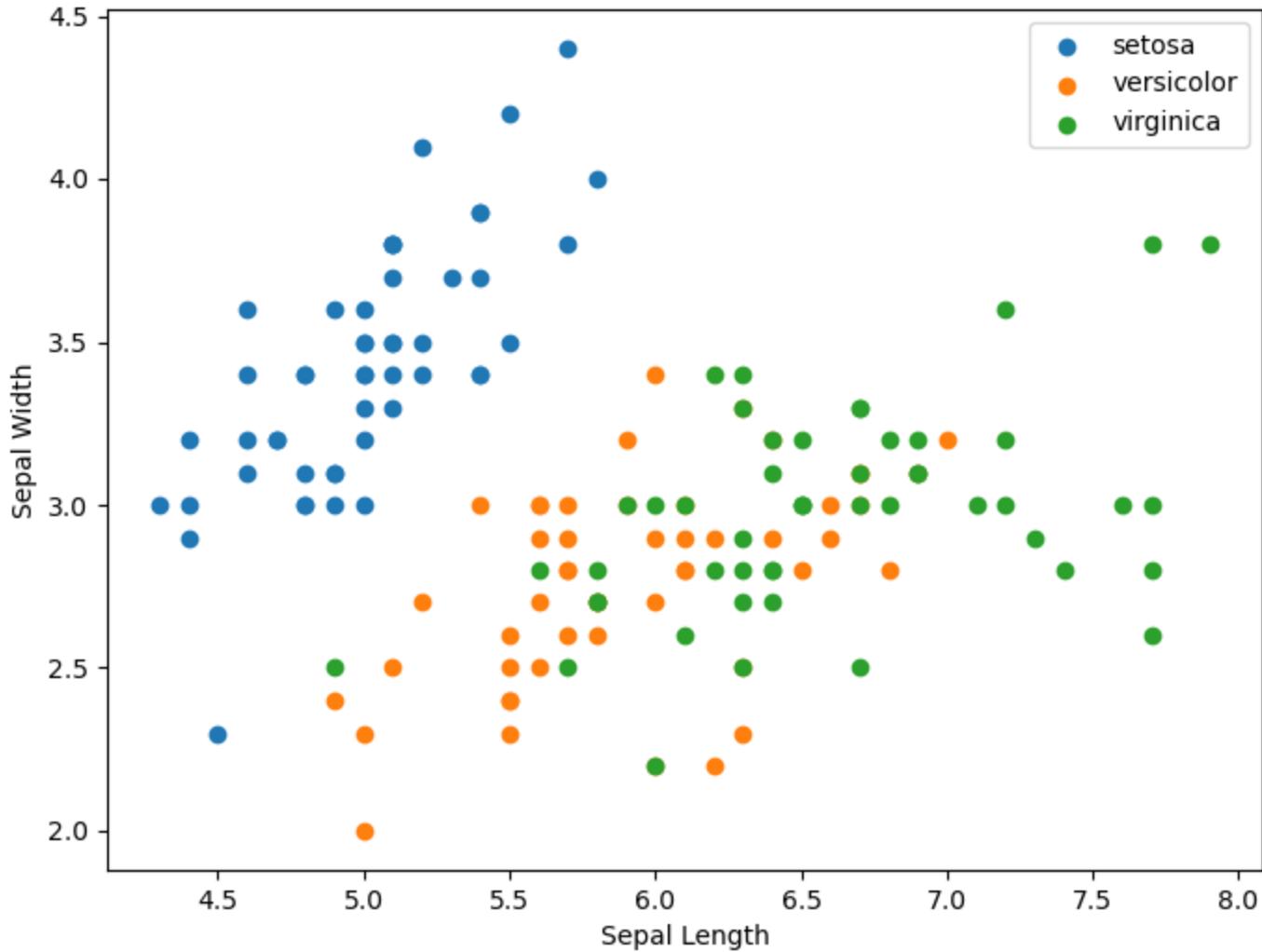
# Show the plot
plt.show()
```

x ^ v

Figure 1



Scatter Plot of Iris Dataset



Matplotlib

Advantages:

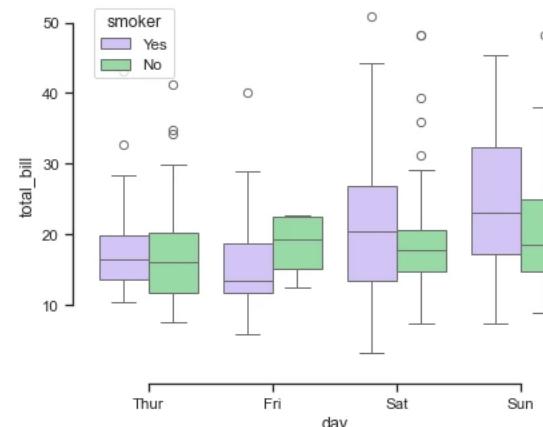
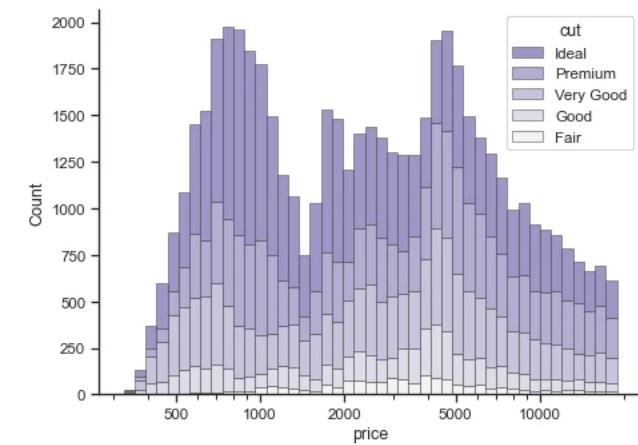
- The most commonly used.
- Work locally.
- Huge community.

Disadvantages:

- Verbose.
- Not well adapted for dynamic plots (especially with many data points).
- Can be difficult to do some specific plots.

Seaborn

- Based on Matplotlib.
- Designed for statistical plots.



Seaborn - scatter plot of iris sepal

```
import matplotlib.pyplot as plt
import seaborn as sns

# Create a scatter plot
sns.scatterplot(data=df_iris, x="sepal_length", y="sepal_width", hue="species")

# Add labels and title
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.title('Scatter Plot of Sepal')

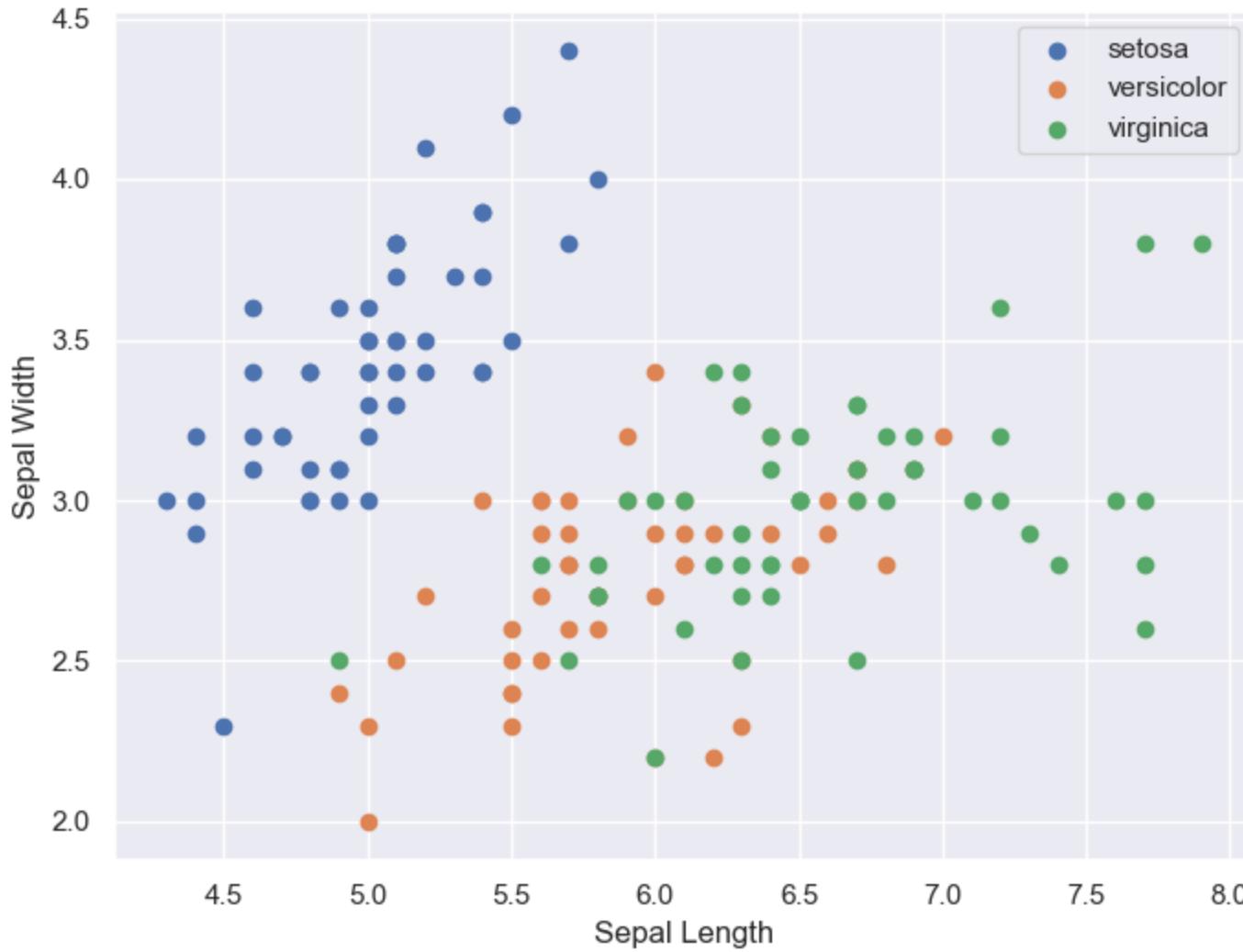
# Display the plot
plt.show()
```

X ^ v

Figure 1



Scatter Plot of Iris Dataset

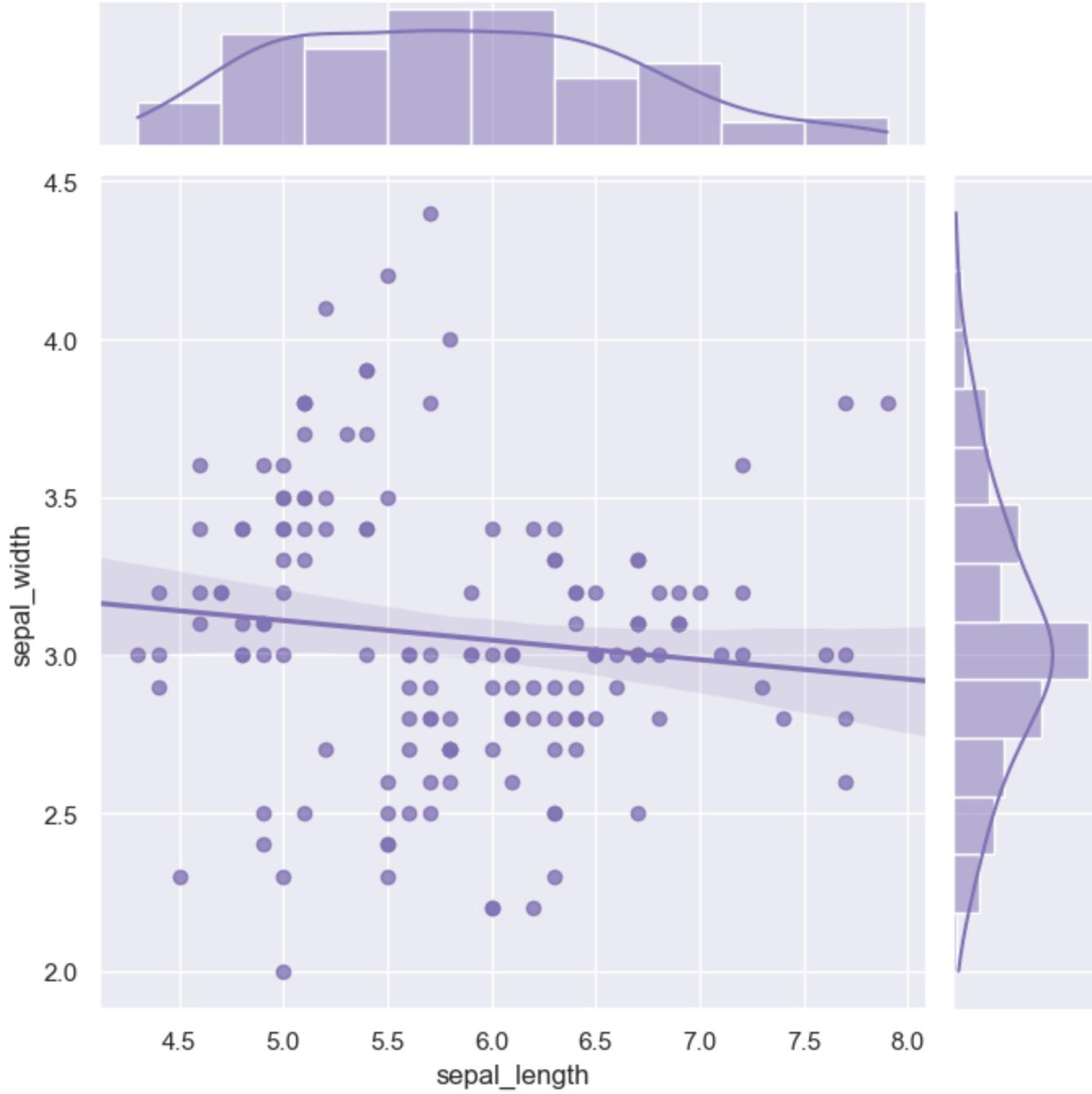


Seaborn - quickly do stats

```
import matplotlib.pyplot as plt
import seaborn as sns

# Compute jointplot
sns.set_theme(style="darkgrid")
g = sns.jointplot(data=df_iris, x="sepal_length", y="sepal_width",
                   kind="reg", truncate=False,
                   color="m", height=7)

# Display the plot
plt.show()
```



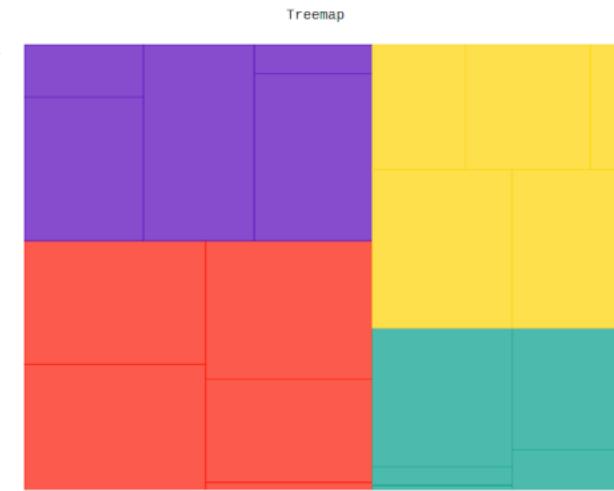
Seaborn

Advantages:

- Provides themes.
- Provide high-level functions for statistical analysis to ease some plots.

Disadvantages:

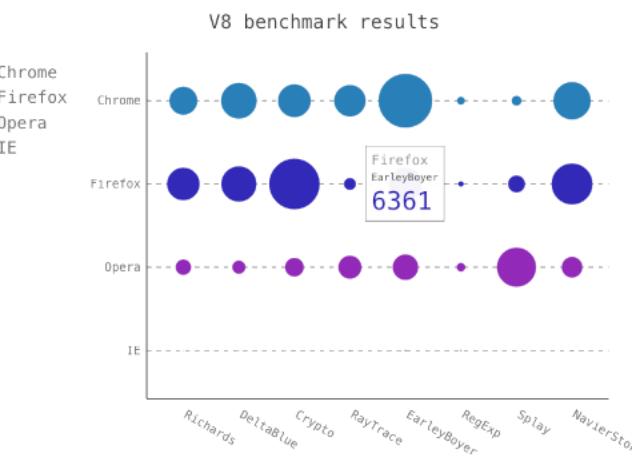
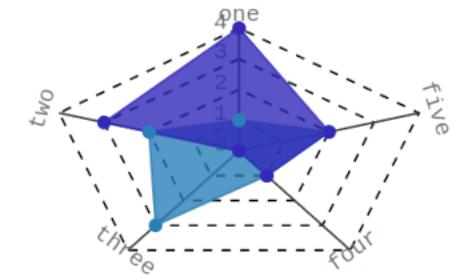
- Still a little bit verbose when modifying some plots.
- Not designed for dynamic plots.
- Not able to generate all kinds of plots.



Pygal

- Produce dynamic SVG.
- Written in Python.

Radar Chart with Fill

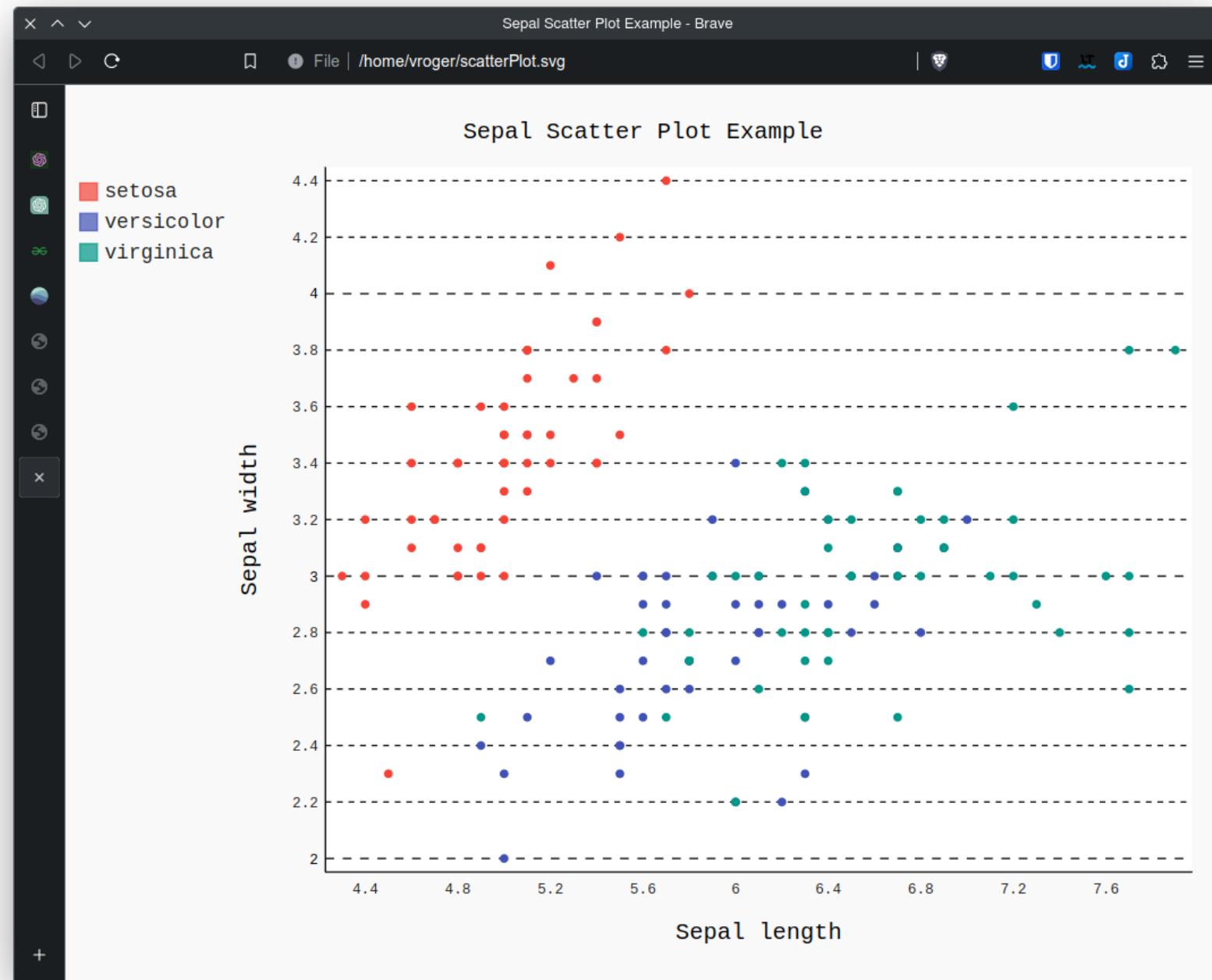


Pygal - scatter plot of iris sepal

```
import pygal

scatter_chart = pygal.XY(stroke=False)
for specie in df_iris["species"].unique():
    df_s = df_iris[df_iris["species"] == specie]
    scatter_chart.add(f'{specie}', list(zip(df_s["sepal_length"], df_s["sepal_width"])))

scatter_chart.title = 'Sepal Scatter Plot Example'
scatter_chart.x_title = 'Sepal length'
scatter_chart.y_title = 'Sepal width'
scatter_chart.render_to_file('scatterPlot.svg')
```



Pygal

Advantages

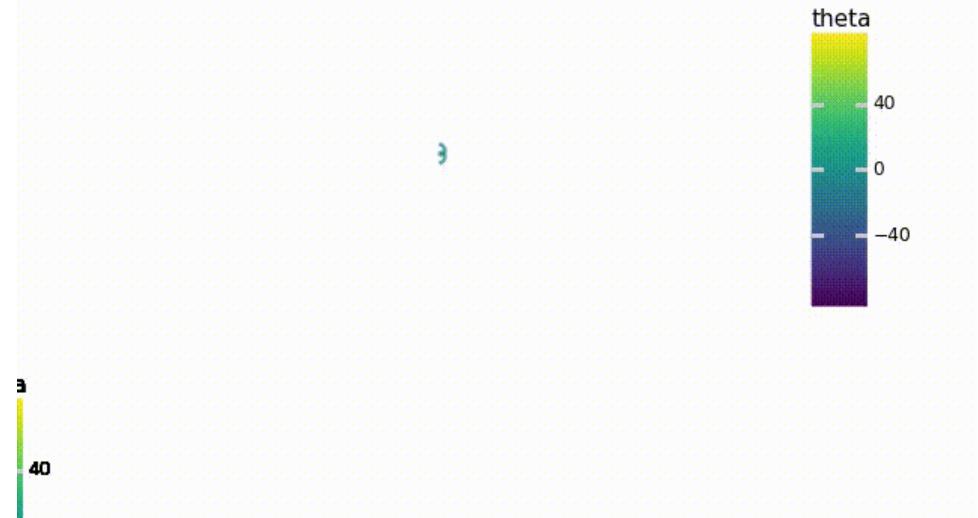
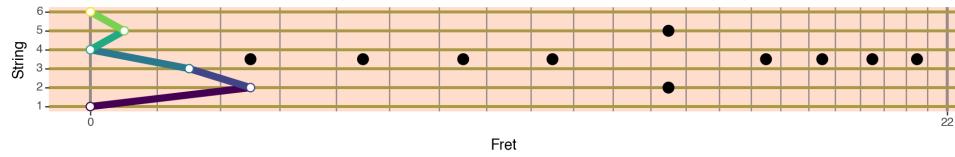
- Even if the output is SVG files, it is interactive by default!
- Great variety of charts.

Disadvantages

- SVG compatible with web browsers, but not with software such as Inkscape.
- Chart is not clear with the defaults (especially the grid).

Plotnine

- Implementation of ggplot2's grammar in Python.



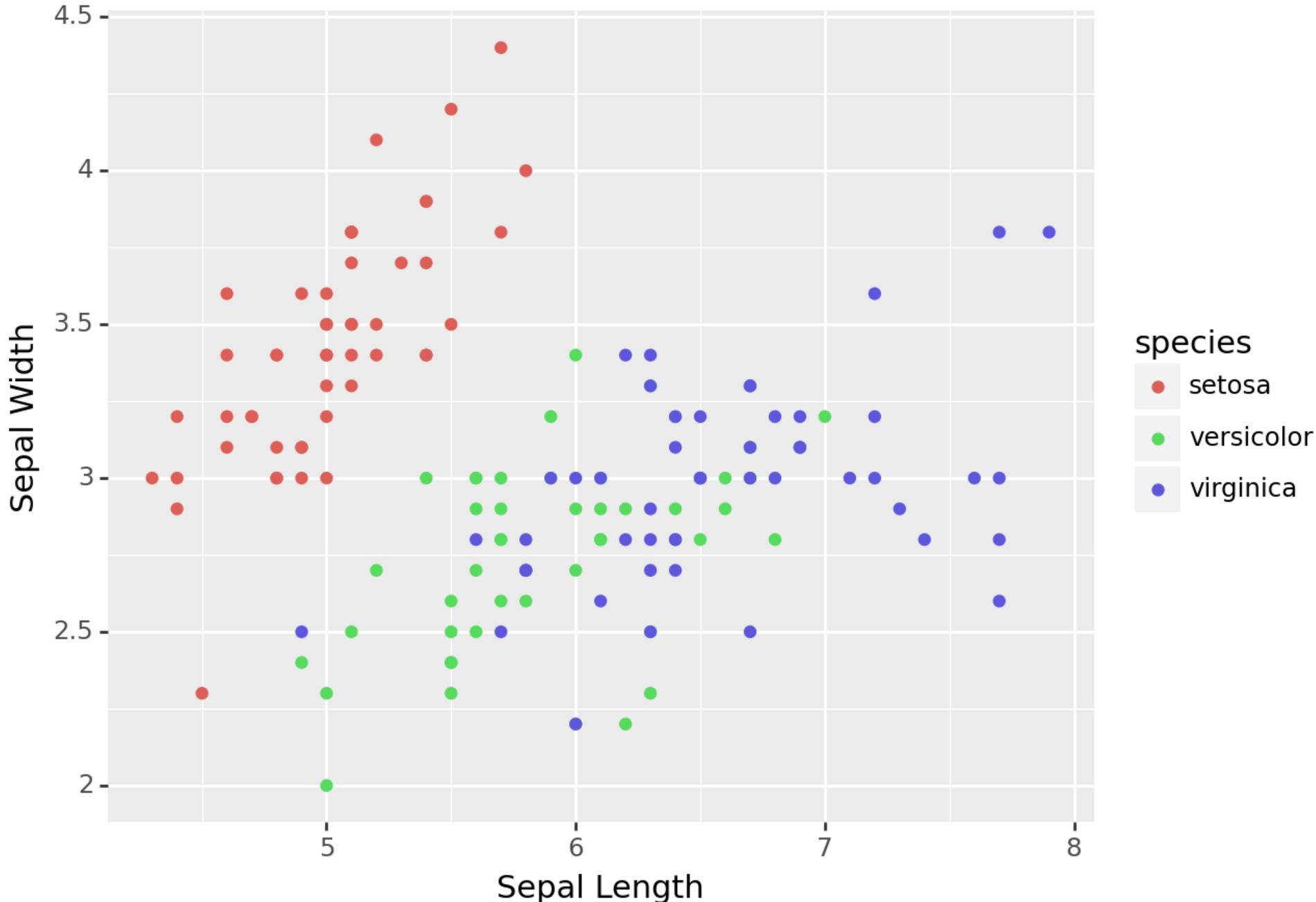
Plotnine - scatter plot of iris sepal

```
from plotnine import ggplot, aes, geom_point, labs

# Create a scatter plot of sepal size
scatter_plot = (
    ggplot(df_iris, aes(x='sepal_length', y='sepal_width', color='species')) +
    geom_point() +
    labs(title='Iris Dataset Scatter Plot', x='Sepal Length', y='Sepal Width')
)

# Display the plot
print(scatter_plot)
```

Scatter Plot of Sepal Size



Plotnine

Avantages

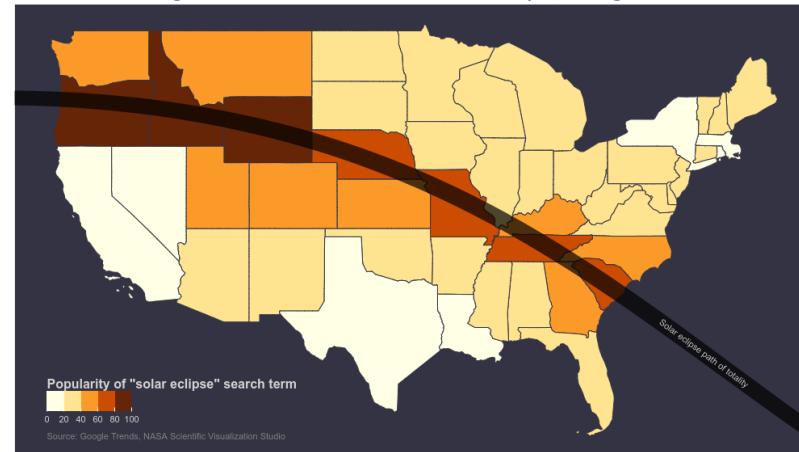
- As powerful as ggplot2.
- Tool adapted for R users using Python.
- Huge variety of plots possible.

Disadvantages

- Syntax is not Pythonic.
- Not for plotting large datasets.
- Plots limited in interactivity.

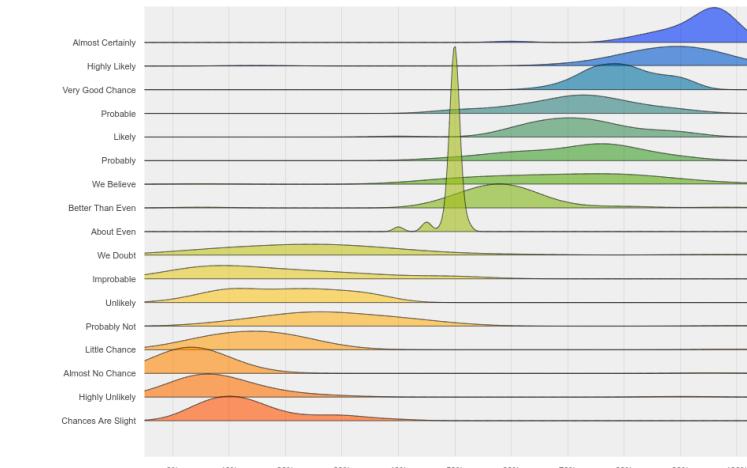
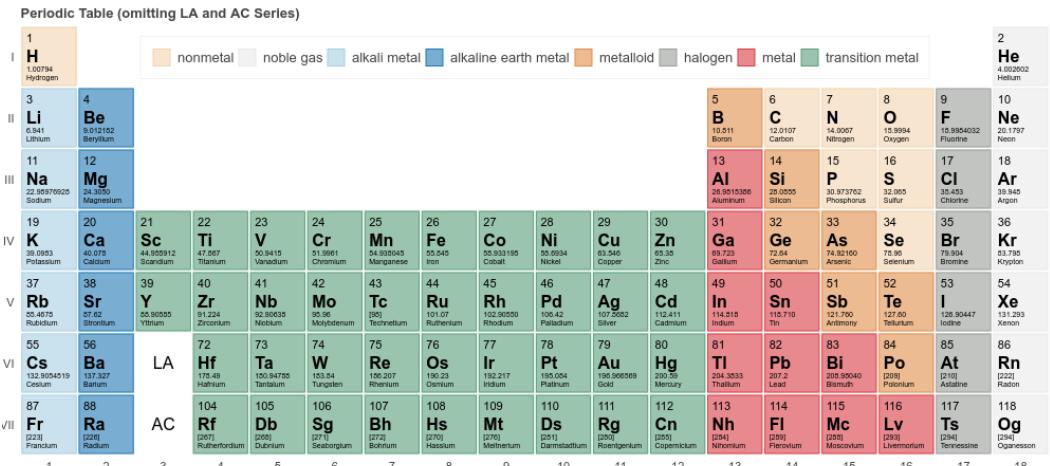


Web oriented



Bokeh

- Usable in Jupyter notebooks and IPython (output HTML by default).
- Capacity to create dashboards.
- Usage of WebGL is possible.



Bokeh - scatter plot of iris sepal

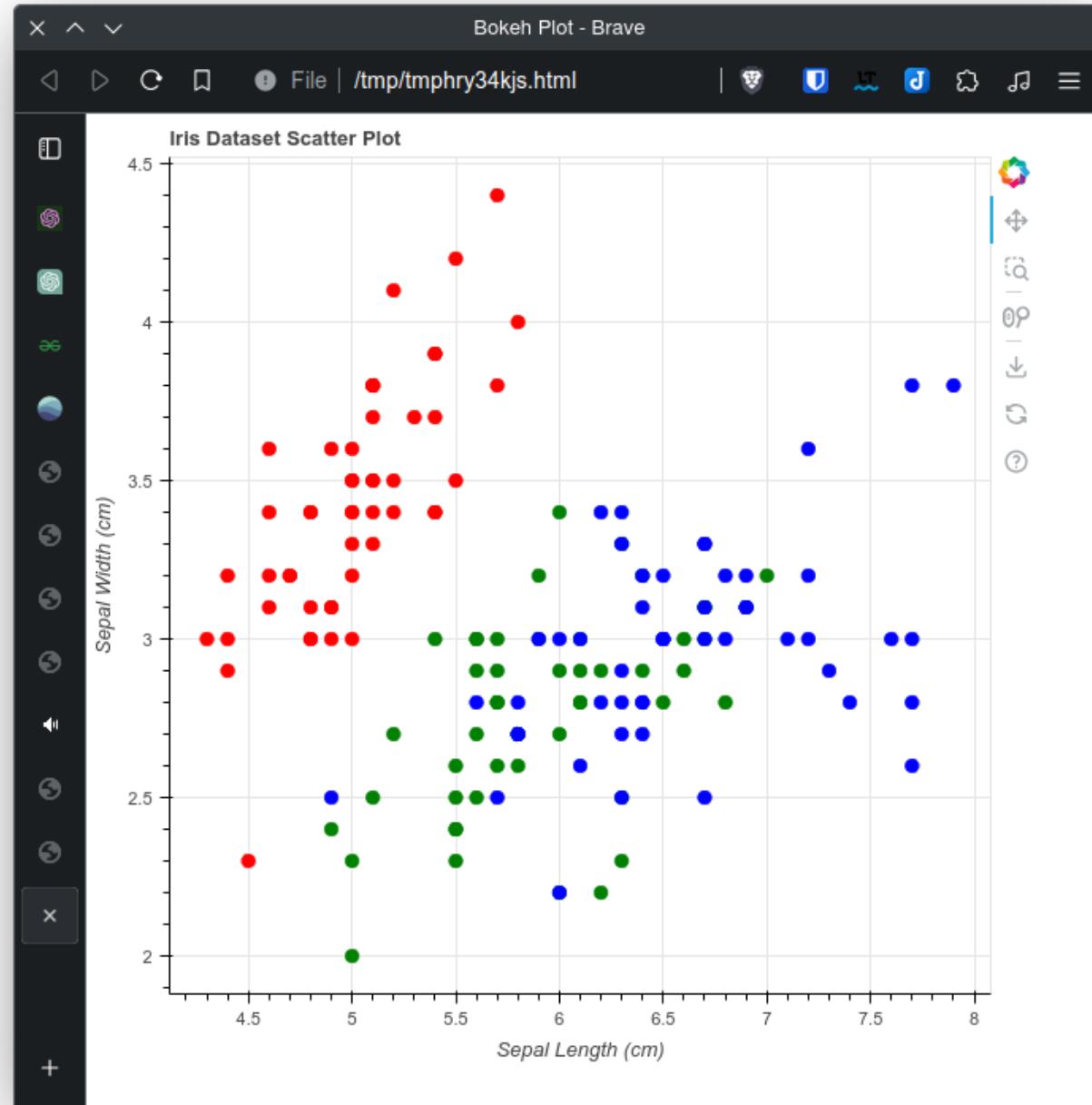
```
from bokeh.plotting import figure, show
from bokeh.transform import factor_cmap
from bokeh.models import ColumnDataSource

# Create a ColumnDataSource
source = ColumnDataSource(df_iris)

# Create a scatter plot with titles
p = figure(title='Iris Dataset Scatter Plot', x_axis_label='Sepal Length (cm)', y_axis_label='Sepal Width (cm)')

p.circle(x='sepal_length', y='sepal_width', source=source,
          size=8,
          color=factor_cmap(
              'species',
              palette=['red', 'green', 'blue'],
              factors=df_iris["species"].unique()
          )
)

# Show the plot
show(p)
```



Bokeh

Advantages

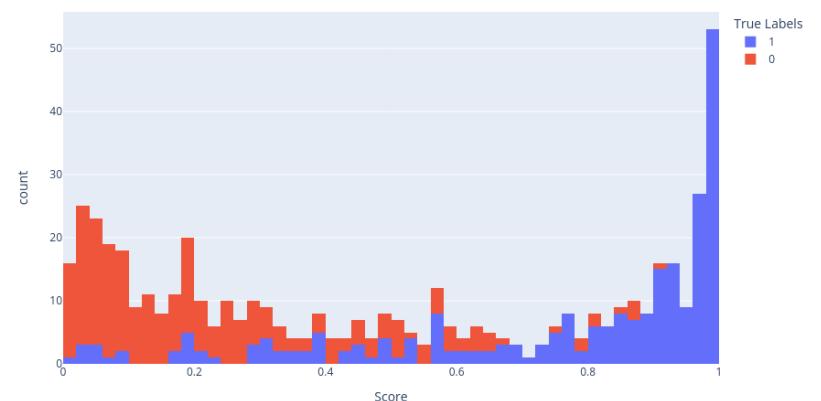
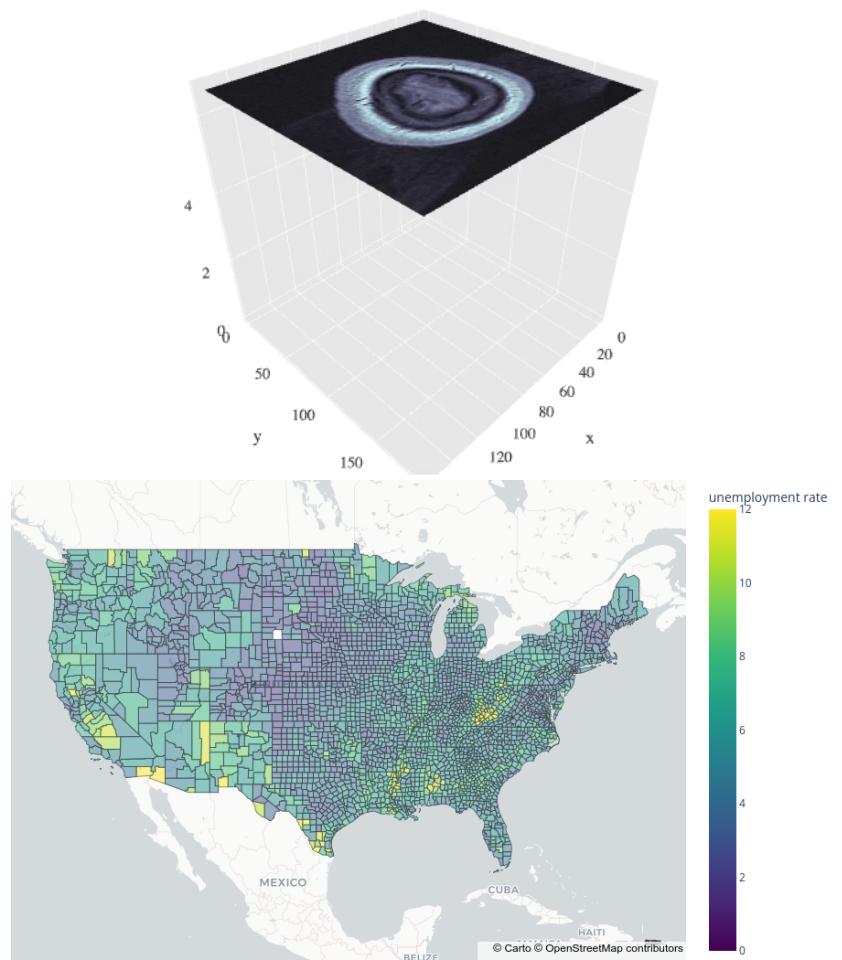
- Low level and "high" level API.
- Can handle a million points easily.
- Interactivity (widgets and dashboards).
- Capacity to do many different plots.

Disadvantages

- Is harder to learn compared to other libraries.
- Less complete for 3D plots.
- If bokeh servers are down, your plots may not show properly (or at all).

Plotly

- Designed for interactive visualizations.
- Usage of WebGL is possible.
- Usable with Python, R and JavaScript.



Plotly - scatter plot of iris sepal

```
import plotly.express as px

fig = px.scatter(df_iris, x='sepal_length', y='sepal_width', color='species',
                  title='Iris Dataset Scatter Plot', labels={'sepal_length': 'Sepal Length', 'sepal_width': 'Sepal Width'})

# Show the plot
fig.show()
```



Plotly

Advantages

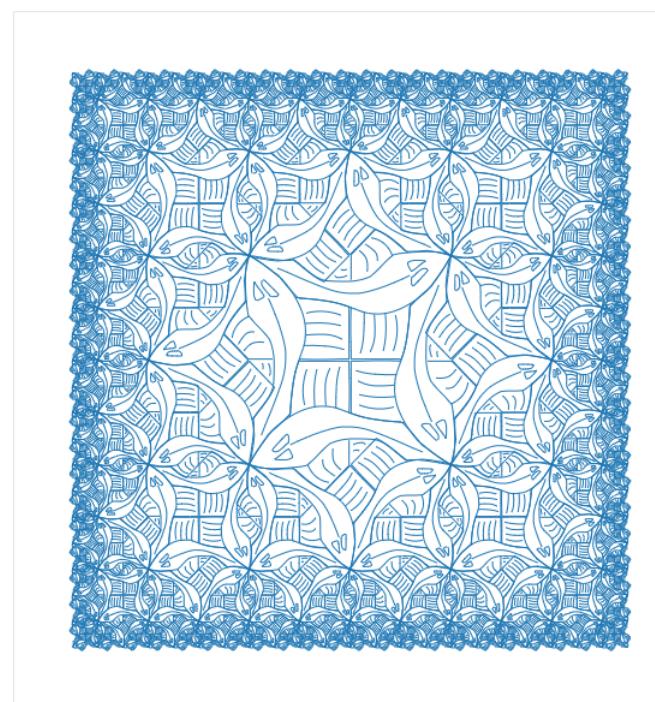
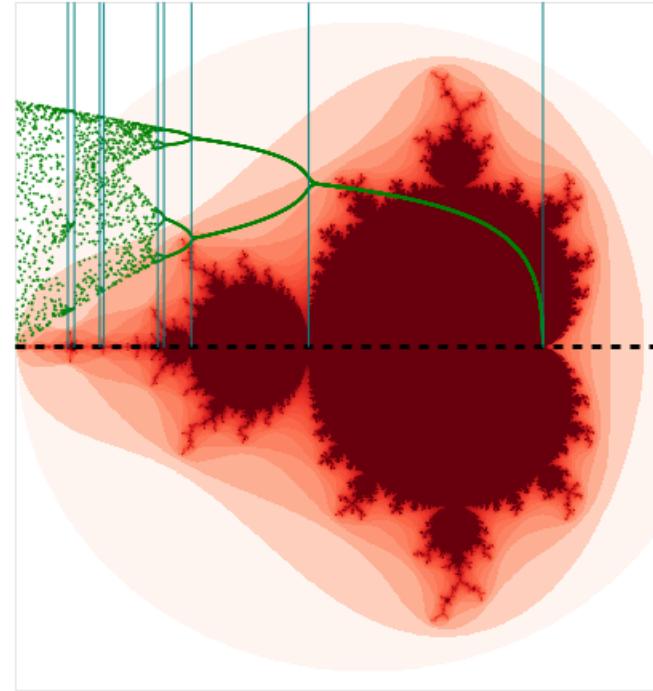
- The documentation is crystal clear for simple plots.
- Dashboard possible with Dash (well integrated).
- Auto resize plot by default.

Disadvantages

- If Plotly servers are down, your plots may not show properly (or at all).
- Hard to use images as ticks.

Holoviews

A high-level library for Bokeh,
Matplotlib and Plotly.



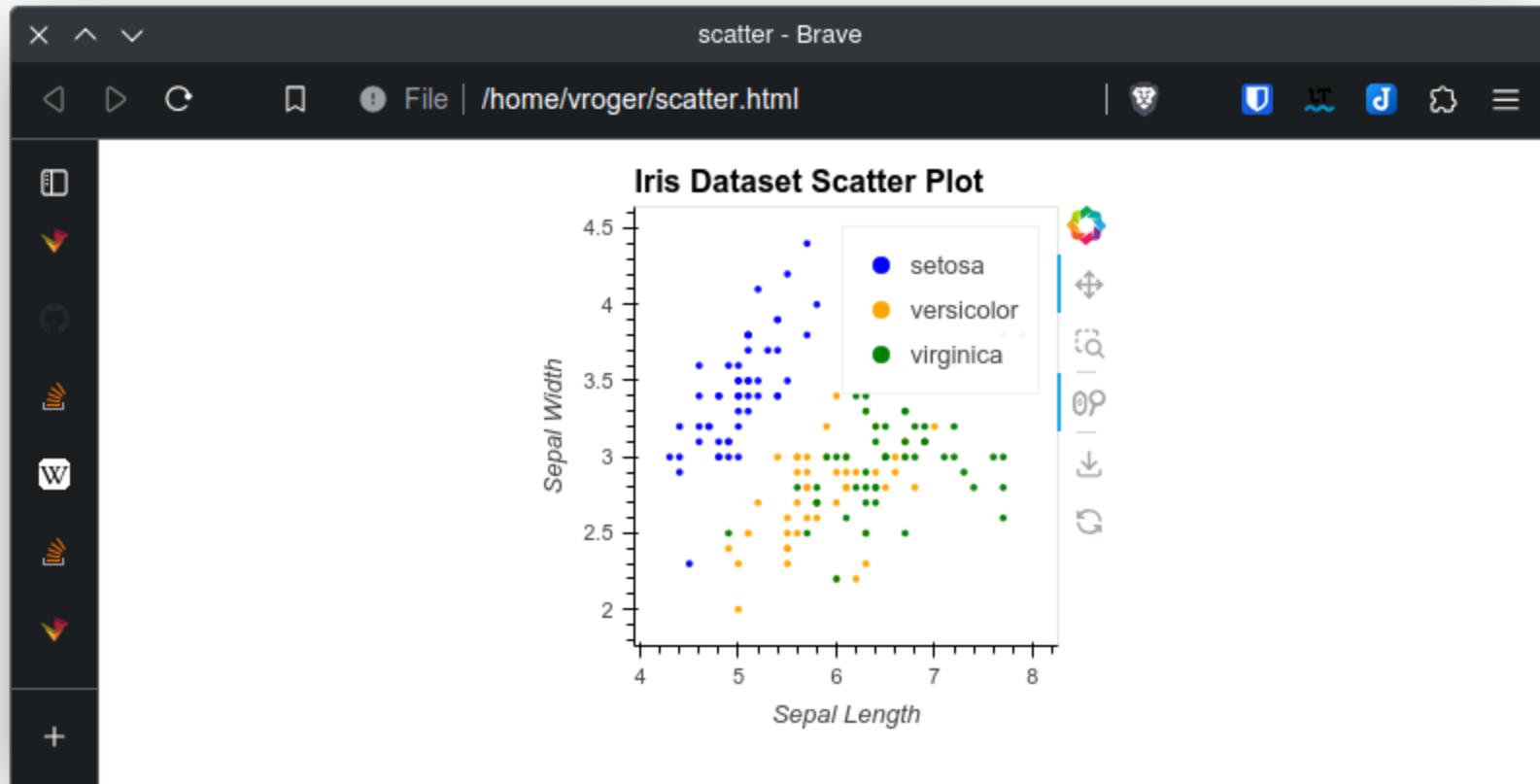
Holoviews - scatter plot of iris sepal

```
import holoviews as hv

scatter = hv.Scatter(
    data=df_iris, kdims=['sepal_length'], vdims=['sepal_width', 'species'],
).opts(color='species', cmap=['blue', 'orange', 'green'])

scatter = scatter.opts(
    title="Iris Dataset Scatter Plot",
    xlabel='Sepal Length',
    ylabel='Sepal Width',
)

hv.save(scatter, 'scatter.html', backend='bokeh')
```

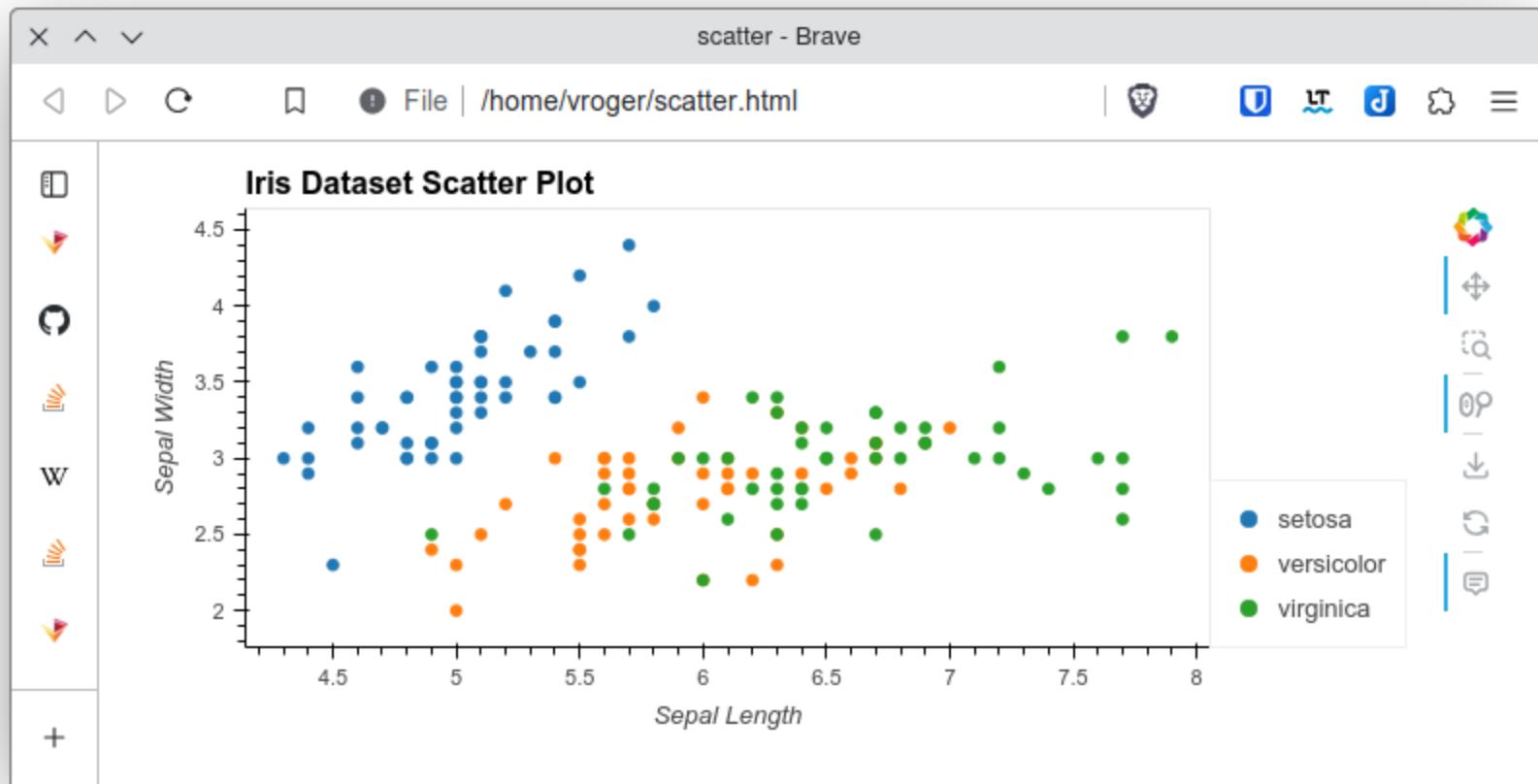


Holoviews - scatter plot of iris sepal using pandas API

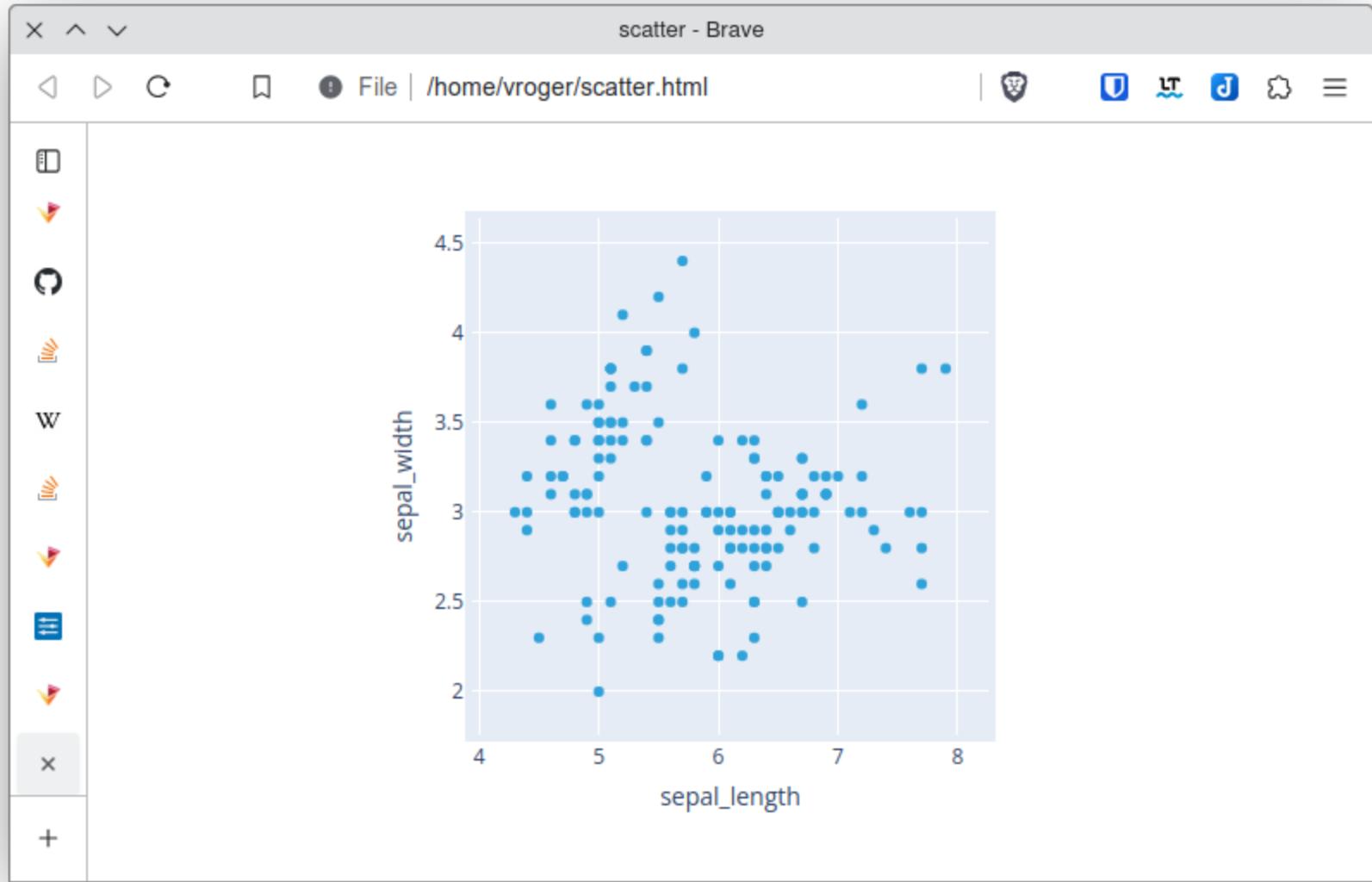
```
import hvplot
import hvplot.pandas
import holoviews as hv

scatter = df_iris.hvplot(kind='scatter', x='sepal_length', y='sepal_width', color='species')

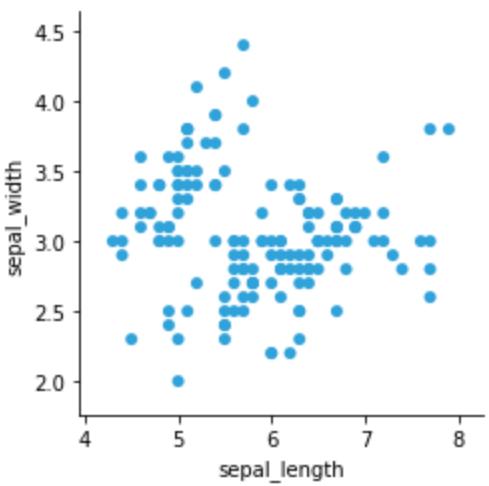
scatter = scatter.opts(
    title="Iris Dataset Scatter Plot",
    xlabel='Sepal Length',
    ylabel='Sepal Width',
)
hv.save(scatter, 'scatter.html', backend='bokeh')
```

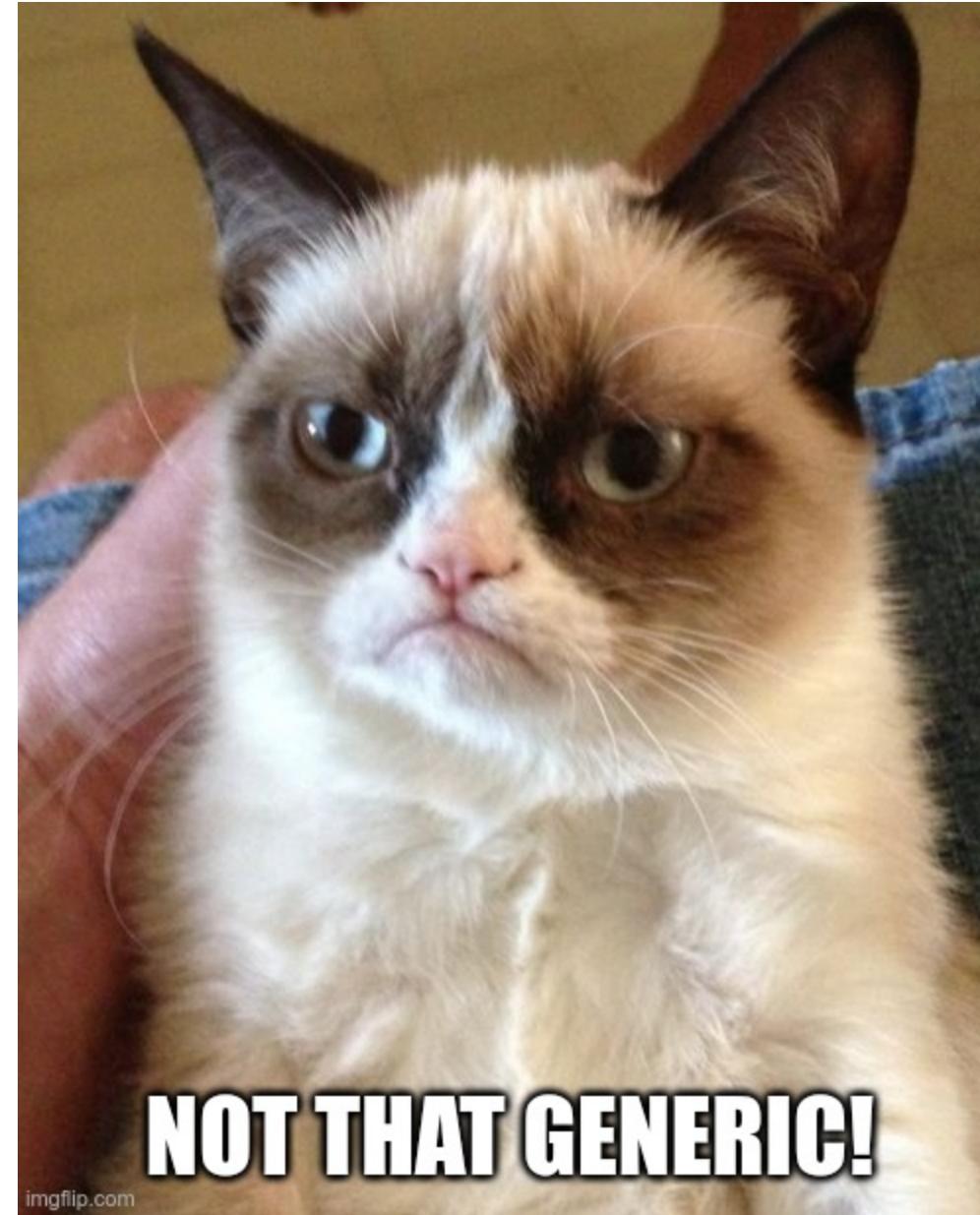


```
hv.save(scatter, 'scatter.html', backend='plotly')
```



```
hv.save(scatter, 'scatter.png', backend='matplotlib')
```





NOT THAT GENERIC!

imgflip.com

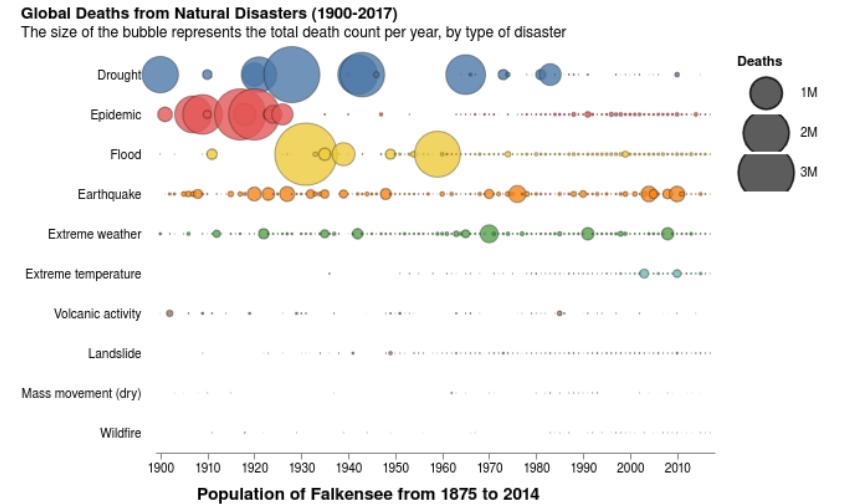
Holoviews

Advantages

- Works well with Bokeh.
- High level API.

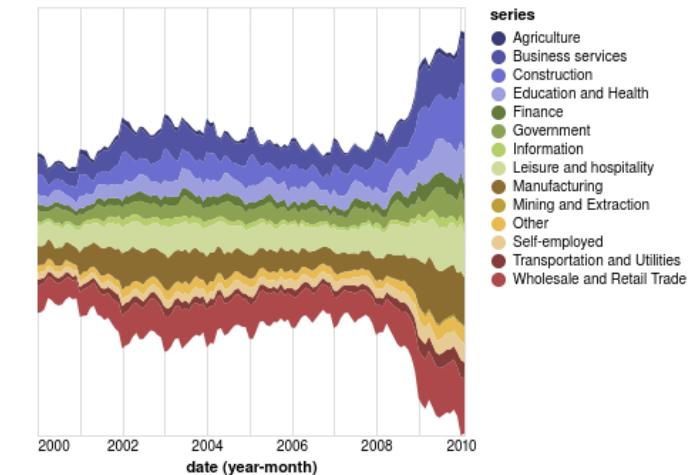
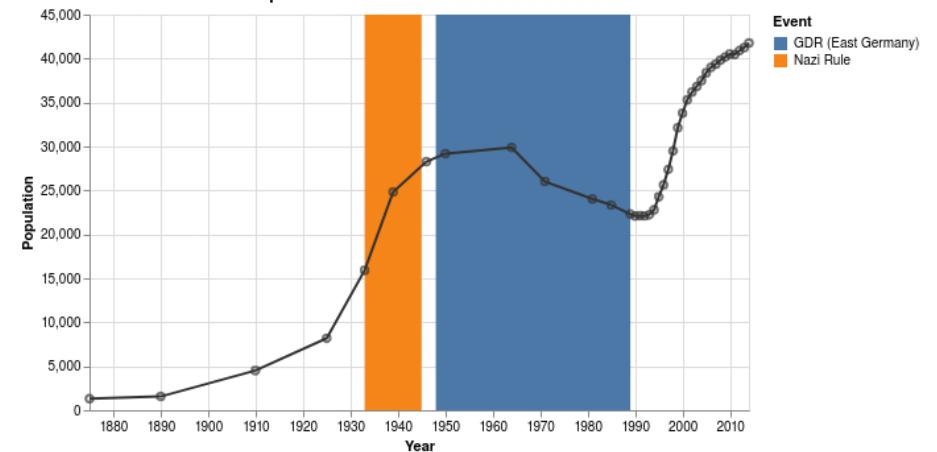
Disadvantages

- Not generic (at least for now).



Altair

- Designed for notebooks.
- Use Vega engine (which produces JavaScript code).

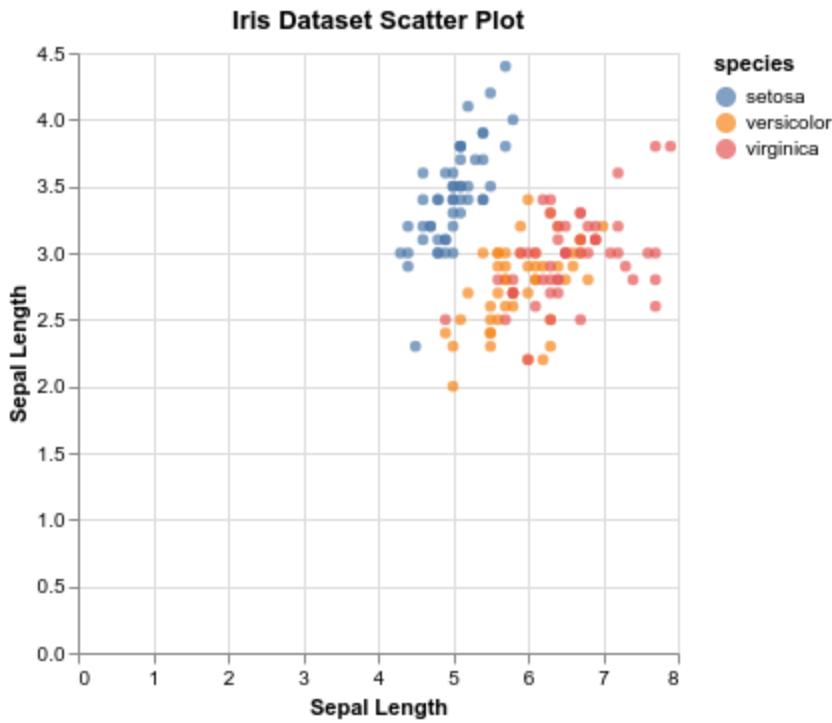


Altair - scatter plot of iris sepal

```
import altair as alt

# Create the scatter plot
scatter_plot = alt.Chart(df_iris).mark_circle().encode(
    x=alt.X('sepal_length', title='Sepal Length'),
    y=alt.Y('sepal_width', title='Sepal Width'),
    color='species'
).properties(
    title='Iris Dataset Scatter Plot'
)

# Show the plot
scatter_plot
```



Altair

Advantages

- Simple use of dataframes.
- Better performance than Plotly in a VS Code notebook.

Disadvantages

- Defaults are not that great for statistical plots.
- No viewer for the last versions (5.x).
- Limited chart customization.

Conclusion



My choice

Plotly

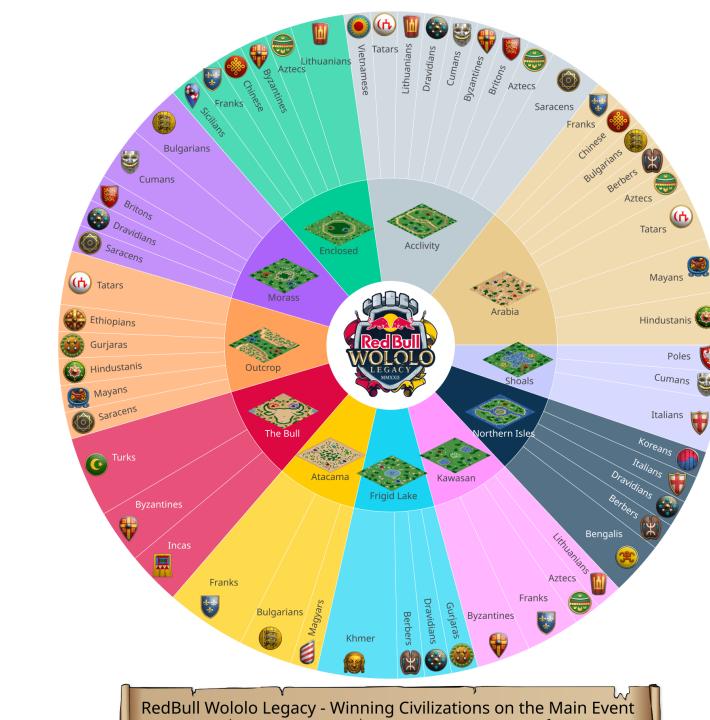
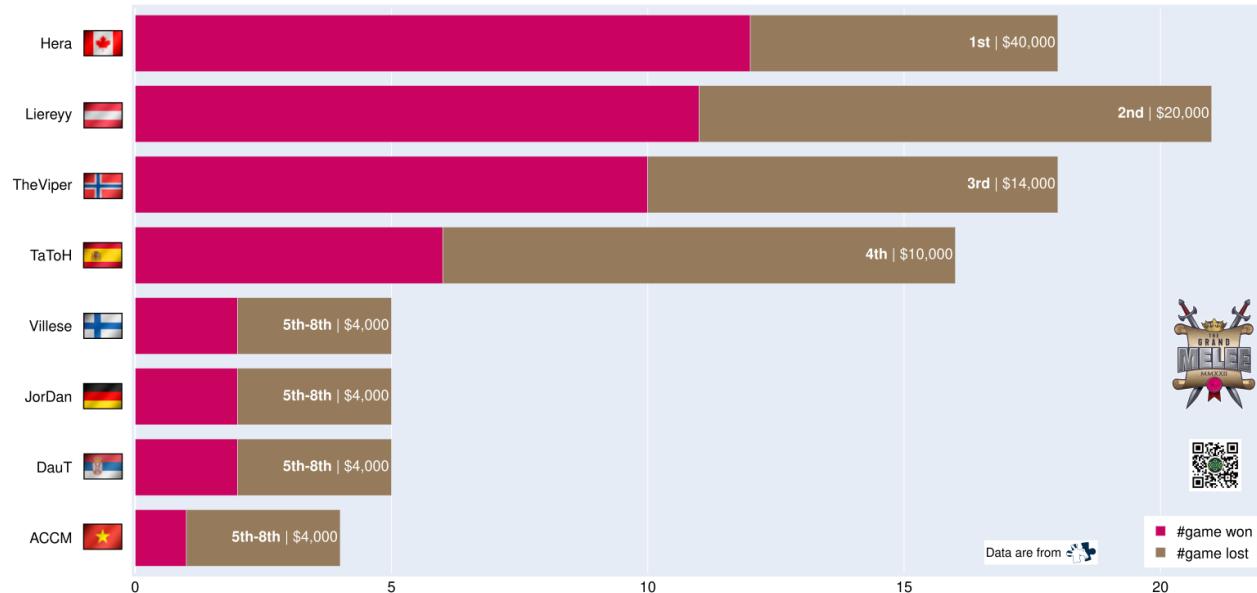
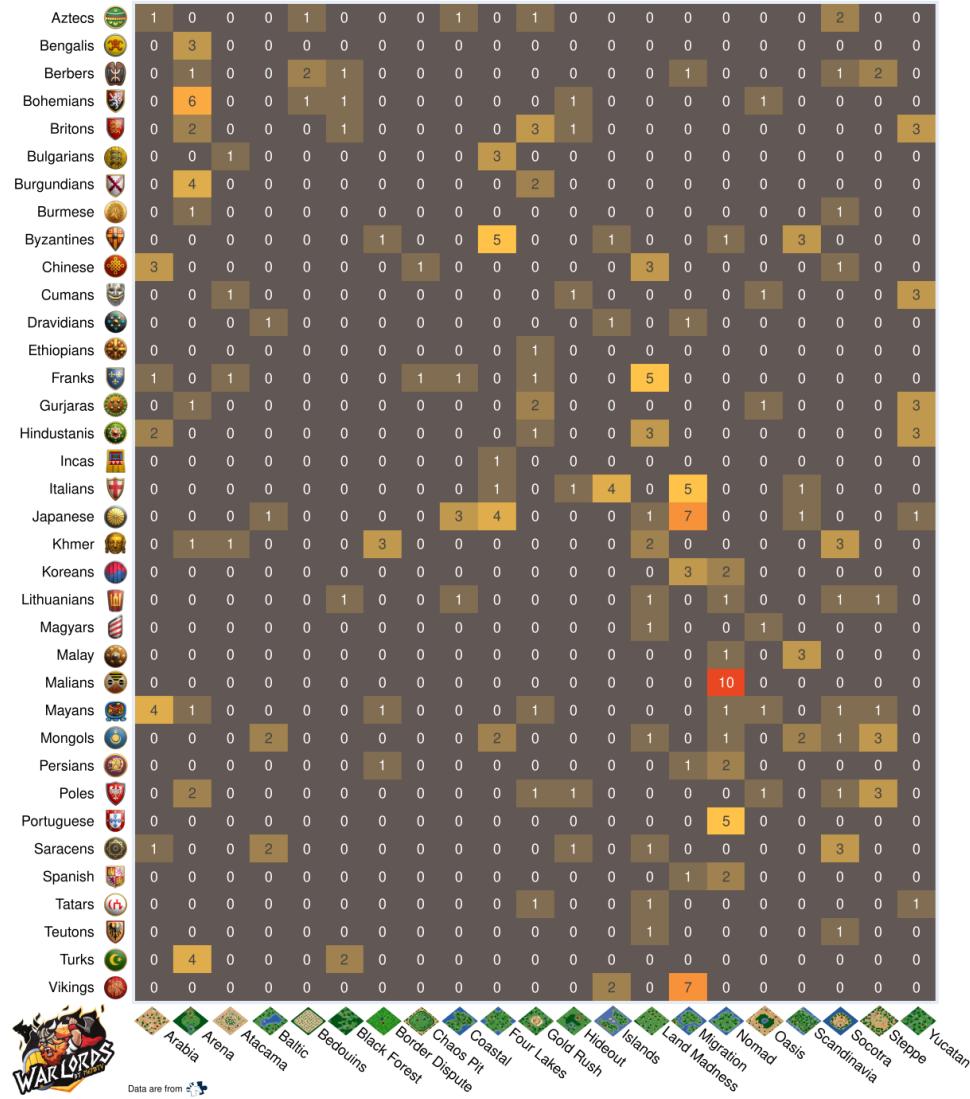
Pros

- Great defaults that reduce my codes.
- Great documentation.
- Supported by MLFlow.

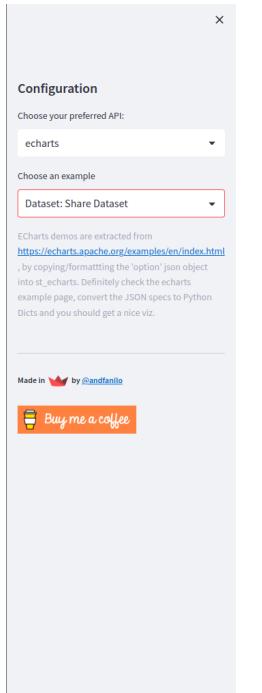
Cons

- A lot of plotting in VSCode notebook's causes slowdowns.
- Having images as ticks is a mess to implement (if you know a simple way, don't hesitate).

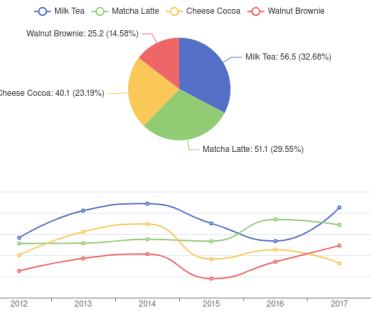
Some work of mine



Bonus dashboard/webapp with Streamlit



Streamlit ECharts Demo



Source Code

Credit: <https://echarts.apache.org/examples/en/editor.html?c=dataset-link>

Made with Streamlit

Thank you for listening!