

A Data Science and Visualization Primer

Specials – 3D Visualisation and Interactive Plots

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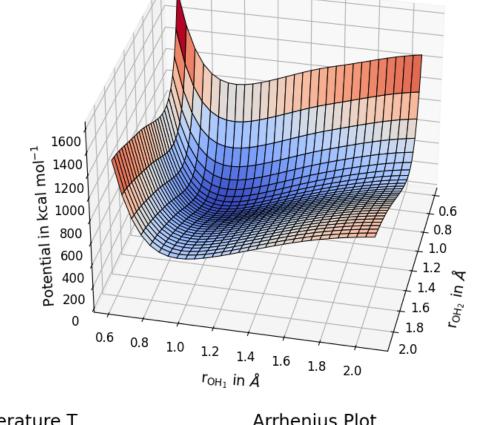
Welcome to the Fourth Exercise Block

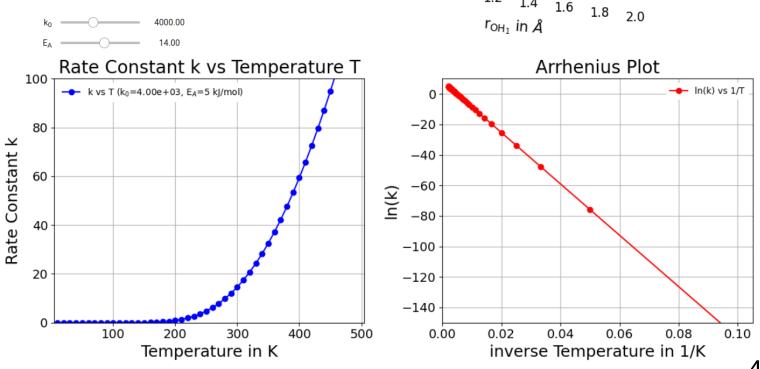
In this slide set, we will outline some special topics, that can be useful in some cases.

The first aspect we will discuss is the generation of 3D data, which is typically challenging in different software packages. Matplotlib is just one of the tools capable of generating compelling visualizations of 3D data.

Another useful tool are interactive diagrams, in which the influence of parameter changes can be visualized on the fly.

This can be useful in many situation, *e.g.* in the classroom, when presenting complex data in meetings and even at conference talks.





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3D Visualization (intermediate difficulty)

In this exercise we read the data *via* the package Pandas (short for **pan**el **da**ta). This is a package specialized for data management.

In large data set it is not uncommen that some entries are missing, which would then show as na (not available) or nan (not a number).

The pandas module data.dropna() is useful to eliminate these missing data entries.

Using data.to_numpy() provides a convenient way to transfer the data to Numpy arrays.

In order to generate the 3D grid for visualization, a 2D base grid has to be available:

- First we have to extract the unique x- and ypositions from the 3D data using np.unique().
- Using these the 2D base grid is created *via* np.meshgrid().

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from scipy.interpolate import griddata
data = pd.read csv('grid data.dat',
                   header=None,
                   skipinitialspace=True,
                   sep=' ')
data = data.dropna()
data = data.to numpy()
x=data[:,0]
y=data[:,1]
z=data[:,2]
x unique = np.unique(x)
y unique = np.unique(y)
X, Y = np.meshgrid(x unique, y unique)
Z = griddata((x, y), z, (X, Y), method='cubic')
fig = plt.figure(figsize = (8,8))
ax = fig.add subplot(111, projection='3d')
ax.view init(elev = 32.5, azim = 12.5)
ax.plot surface(X, Y, Z, cmap='coolwarm',
                linewidth = 0.75, edgecolor='k')
```

3D Visualization (intermediate difficulty)

Once the 2D base grid has been constructed, the associated z-values are generated using SciPy's griddata() command. In case data points are missing, they are added *via* cubic interpolation.

To plot the newly generated data a 3D plot has to be requested using plt.axes(projection='3d') when defining the figure.

The orientation of the camera relative to the axes is controlled *via*:

- the elevation angle relative to the xy-plane (top view: elev = 0 side view: elev = 90)
- the azimuth angle controlling the z-rotation (x-aligned: azim = 0 y-aligned: azim = 90)

The data is finally plotted via ax.plot_surface() with cmap selecting the colormap and edgecolor the color of the grid lines.

```
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import numpy as np
import pandas as pd
from scipy.interpolate import griddata
data = pd.read csv('grid data.dat',
                   header=None,
                   skipinitialspace=True,
                   sep=' ')
data = data.dropna()
data = data.to numpy()
x=data[:,0]
y=data[:,1]
z=data[:,2]
x unique = np.unique(x)
y unique = np.unique(y)
X, Y = np.meshgrid(x unique, y unique)
Z = griddata((x, y), z, (X, Y), method='cubic')
fig, ax = plt.figure(figsize=(8, 8)),
                     plt.axes(projection='3d')
ax.view init(elev = 32.5, azim = 12.5)
ax.plot surface(X, Y, Z, cmap='coolwarm',
                linewidth = 0.75, edgecolor='k')
```

3D Visualization (interm. difficulty)

Adjusting the axis labels requires some fine tuning.

First, the automatic rotation of the labels is turned off using ax.xaxis.set_rotate_label(False).

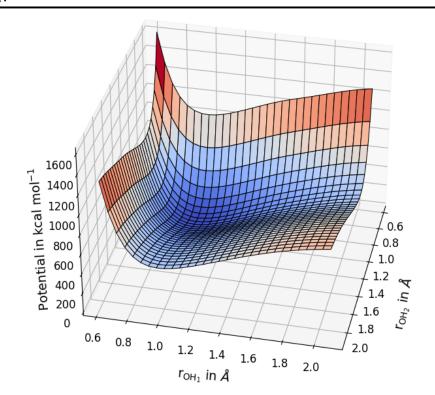
The orientation of the labels is then manualyl adjusted using:

- labelpad: distance between axis and label
- rotation: angle of the label

<u>Fun Fact:</u> axis='both' in ax.tick_params() controls the ticks of all three axis in the plot.

<u>ProTipp:</u> Using the LaTeX non-breaking space symbol '~' enables us to shift the z-label a little bit up along the axis.

To enter LaTeX math mode raw strings r'...' is used together with dollar symbols, *i.e.* \$~~~\$



Interactive Plots (interm. difficulty)

Sometimes it may prove useful to generate plots that have an interactive element, to visualize concepts to the audience.

This can be achieved with the interact() module in the package ipywidgets.

To do this, the target function and the associate have to be defined in a custom function using the def command.

Using np.linspace() an interval with a given number of points is created for the x-axis.

The magic then happens inside the function interact(). With the help of FloatSlider() interactive sliders for all parameters can be defined.

That way any function can be transformed into a compelling, interactive presentation.

```
from ipywidgets import interact, FloatSlider
def van deemter(A=1.0, B=1.0, C=0.1):
    u = np.linspace(0.01, 10, 500)
    H = A + (B / u) + (C * u)
    H \min = np.min(H)
    plt.axhline(H min, color='r', linestyle='--',
                linewidth=0.8)
    plt.plot(u, H, color='k',
             label=f'Van Deemter (A={A}, B={B}, C={C})')
    [\ldots]
interact(van deemter,
         A = FloatSlider(min=0.1, max=10.0, step=0.5,
                       value=1.0, description='A'),
         B = FloatSlider(min=0.1, max=5.0, step=0.2,
                       value=4.0, description='B'),
         C = FloatSlider(min=0.01, max=5.0, step=0.1,
                       value=2.0, description='C'))
```

