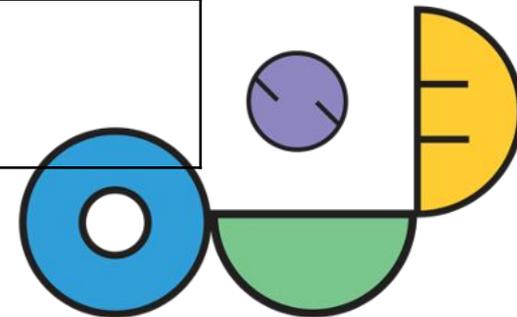
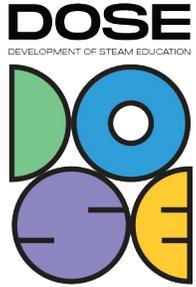


# Epistemic Programming - Exploring your own environment

Description of the activity:	The activity introduces the didactic programming concept of Epistemic Programming as well as an exemplary teaching project about collecting and analyzing environmental data with Arduinos and Jupyter Notebooks
Target group(s):	(Future) Secondary school teachers in STEAM-subjects
Keywords:	Insight-driven Programming, Data Exploration, Local Environment, Self-Expression, Jupyter Notebooks
Duration of activity:	5 hours
Description of activity environment and materials needed:	<ul style="list-style-type: none"> <li>• Powerpoint</li> <li>• Jupyter Notebook from the Material-Folder</li> <li>• Anaconda as a software to open the Jupyter-Notebook-Environment (<a href="https://anaconda.org">https://anaconda.org</a>)</li> </ul>
Contact	Sven Hüsing, Paderborn University <a href="mailto:sven.huesing@upb.de">sven.huesing@upb.de</a>

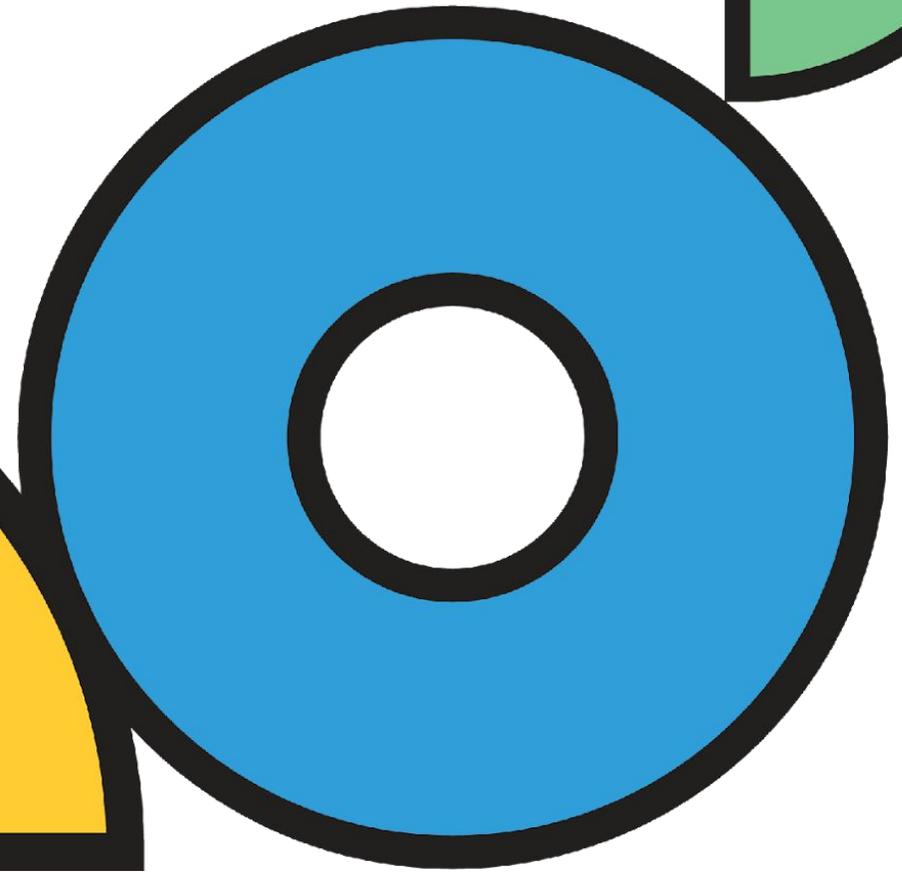
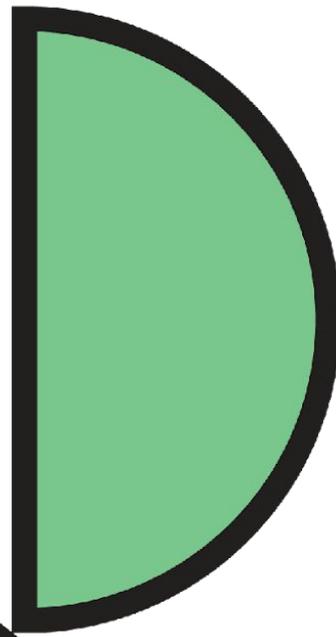
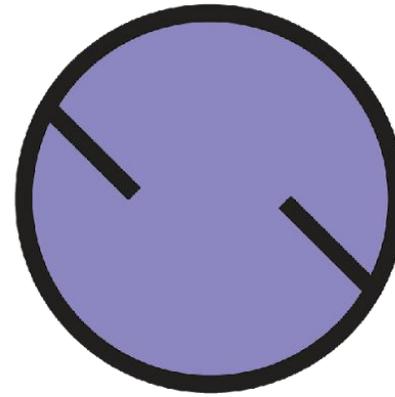
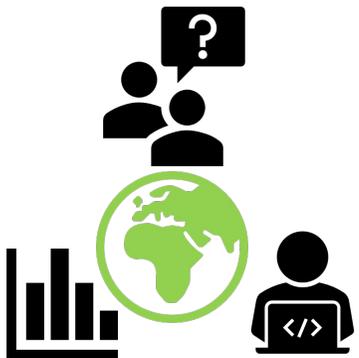




# Epistemic Programming

## Exploring your own environment

Collection and Analysis of Environmental Data with  
Arduinos and Jupyter Notebooks

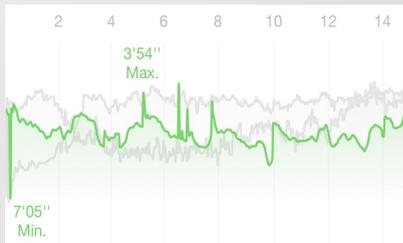


# Data Collection and Data Analysis in our everyday life

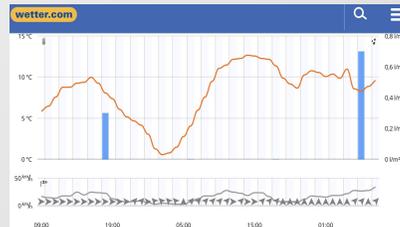
## Sources:

- NikeRunClub app (iPhone)
- wetter.com
- stocks-app (iPhone)
- JupyterNotebooks

## sports



## weather



## stocks



## medicine





# “Epistemic Programming”

- Perceive programming as a **means to follow one’s own interests and ideas**
- Reduce the “prejudices” regarding the “nerdiness” of programming by showing how programming can help to **discover one’s environment/one’s own world**
  - Foster an understanding of both one’s own (local) environment (=external insights), as well as the competent usage of digital artefacts for this purpose (=internal insights) (see Schulte, 2013)
  - Communicate the **intertwined process of programming and knowledge acquisition**
- Focus on:
  - Insights one can gain through the programming process
  - Making the (programming) process itself perceivable, comprehensible and reproducible



# Idea: Gaining knowledge through “Epistemic Programming”

Why should one make the programming process perceivable??

- For oneself:
  - “Strengthening” of specific programming and project processes
  - Comprehensible expression of own thoughts and ideas
  - Later: Being able to reproduce the programming process (maybe slightly adapted)
- For others:
  - Communicate aspects of programming in an understandable way
  - Adapt or expand the programmed artefact based on the representation and documentation of the programming process according to own interests
    - Educate others so they can themselves engage in programming projects (maybe with a slightly different goal)



# Idea: Gaining knowledge through “Epistemic Programming”

„How can you record insights and the (programming)  
process at the same time??”



# Reproducible Research in the Context of Data Analysis

Goal: Comprehensible and reproducible programming projects

Here: Focus on (data-)analyses

- Document data analyses in such a way, they can be repeated with the same or similar data in order to get comparable results
- Connecting results and explanatory text
- Making the process comprehensible (by documenting the individual steps)

McNamara (2019) and Sandve et al. (2013)



# Computational Essays – A possibility to implement reproducible research

Idea:

- „digital essay“ which contains small visualisations of the process and the results

Goal: Communication of an Idea or of Insights, Support for a hypothesis etc.

- Computational Essays report about the knowledge acquisition process in a kind of “dialogue” between its author and the digital artefact
- Viewers can interact with programs/visualisations
- Through the interactions with computational essays, its “readers” can explore the process of knowledge acquisition and might change their role from reader to programmer by tinkering with the code in the Computational Essays

B Odden & Malthe-Sørensen (2021), DiSessa (2000) and Wolfram (2017)

# Computational Essays with Jupyter Notebooks

Jupyter Notebooks – a possibility to prepare a programming environment in which students can perform data analyses (also see the concept of Worked Examples (Atkinson et al., 2000; Caspersen & Bennedsen, 2007; Merrienboer & Krammer, 1987))



```
In [27]: #Mittelwert
print("Mittelwert: " + str(round(df['value'].mean(),2)) + " °C")
#Standardabweichung
print("Standardabweichung: " + str(round(df['value'].std(),2)) + " °C")
#unteres Quartil
print("unteres Quartil: " + str(round(df['value'].quantile(q=0.25),2)) + " °C")
#oberes Quartil
print("oberes Quartil: " + str(round(df['value'].quantile(q=0.75),2)) + " °C")

Mittelwert: 18.5 °C
Standardabweichung: 0.53 °C
unteres Quartil: 18.47 °C
oberes Quartil: 18.74 °C
```



Granger & Perez (2021)



# Computational Essays with Jupyter Notebooks

Computational Essays in Jupyter Notebooks have four different cell types:

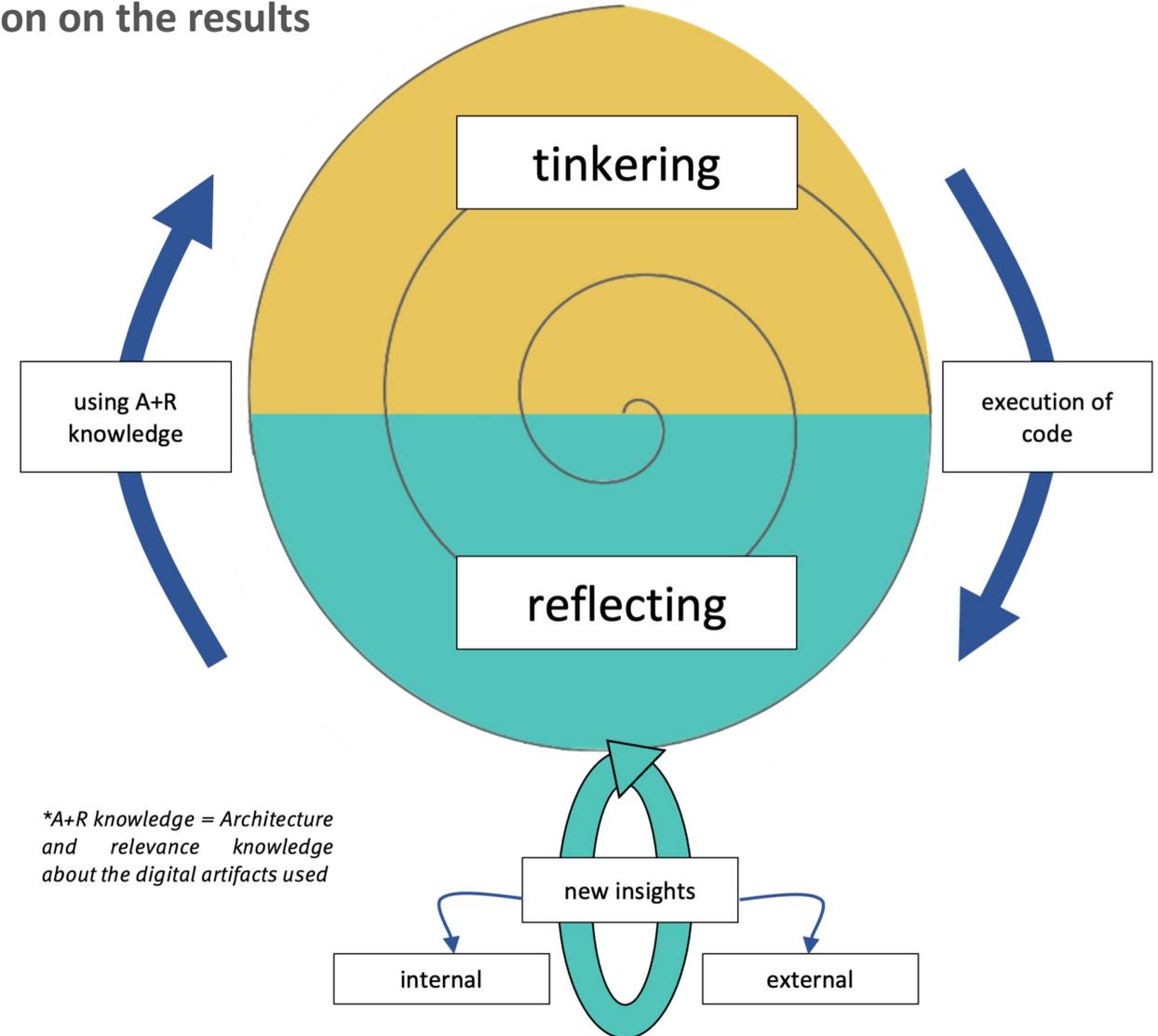
- Technical-Code-Cells (provide the “invisible” prerequisites for the data analyses, e.g. read the dataset)
- Output-Code-Cells (generate output (e.g. visualisations), readers/viewers can interact with)
- Explaining-Markdown-Cells (explain the code and the process)
- Interpreting-Markdown-Cells (interpret/comment the results)



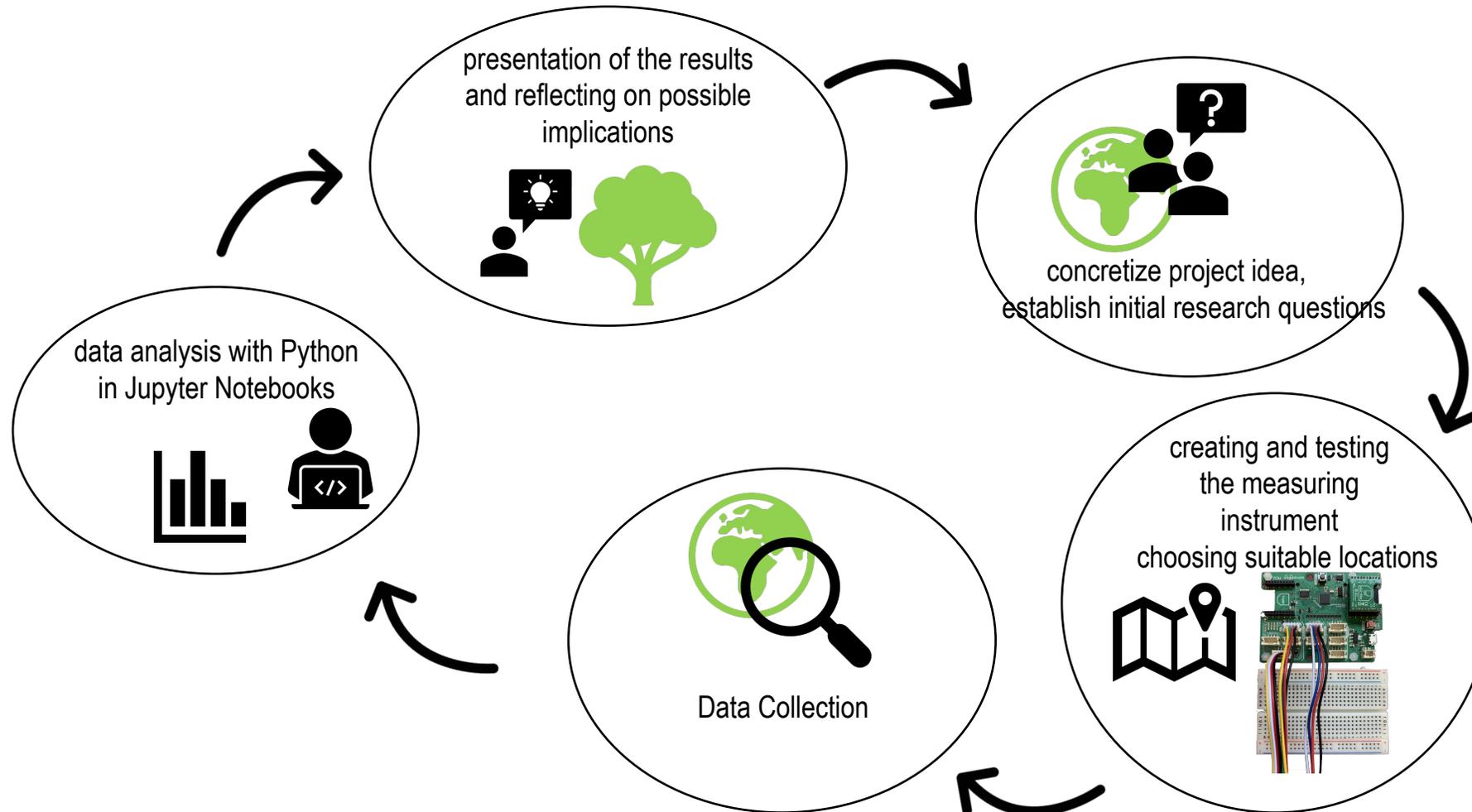
# Intertwined Knowledge- and Programming-Process

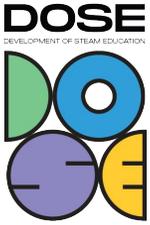
spiral-like interplay of tinkering and reflection on the results

1. Tinkering (with different programming snippets)
2. Execution of the Code
3. Reflection on the programming results
  - new internal and external insights can emerge:
    - *external*:  
about the explored fields of interest
    - *internal*:  
the adequate use/combination/adaptation of digital artefacts for this exploration and knowledge processes (Schulte, 2013)
4. using these insights & architecture- and relevance-knowledge about the interaction of actors within the cognitive system (de Ridder, 2007; Kroes & Meijers, 2006; Schulte & Budde, 2018; Hollan et al., 2000)
5. identify sections in the code, where it can be changed in order to improve the programming results regarding the personal interests
6. Based on this: tinkering with the code, which means trying different methods, libraries or programming snippets, before executing the adapted code again



# Example: Environmental Analysis through Epistemic Programming

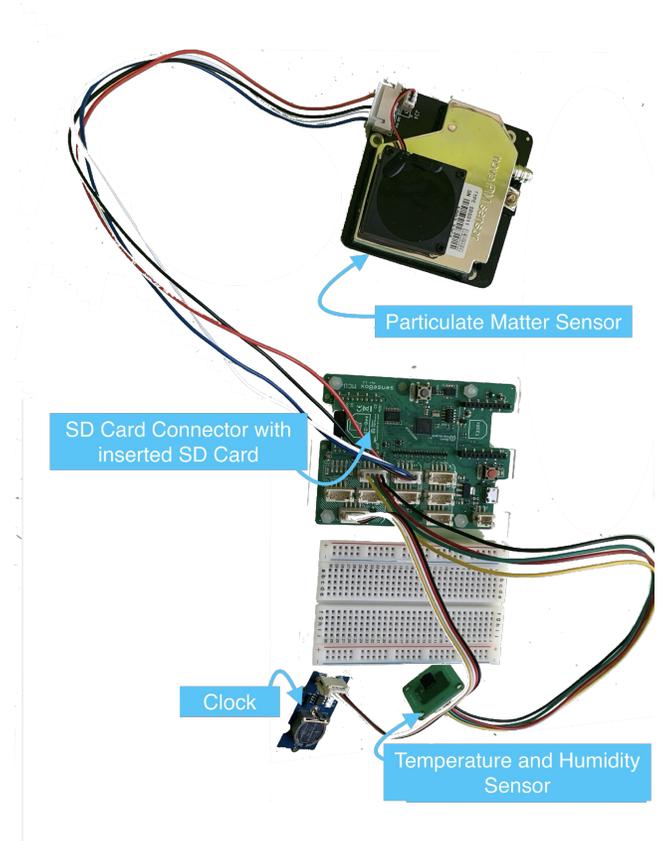




# Project Example: Fine dust pollution

(Measuring Instrument: Sensebox)

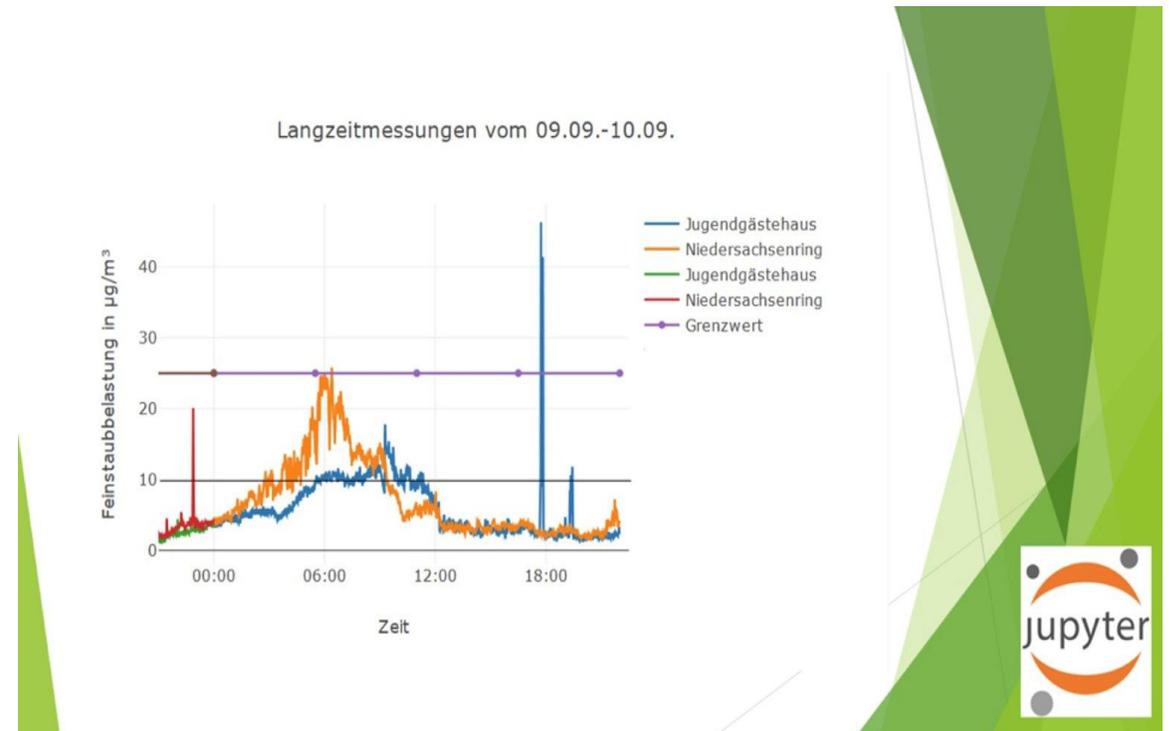
- Programming the SenseBox:
- Arduino-Software
- With Blockly for the Sensebox
  - block-based programming tool
  - Especially for students without programming-experience
  - Link: [>click me<](#)



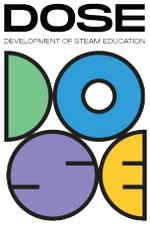
# Data analysis with Python in Jupyter Notebooks

The following steps are executed by the students ...

- importing libraries, reading in data and displaying data
- data cleansing
- evaluation and visualizing data
  - developing research questions
  - finding suitable visualizations
  - trying things out
  - “code-tuning”
  - Documenting
    - in the Jupyter Notebook itself
    - creation of a Computational Essay showing the code, the programming results, the interpretation and description of both



possible result (fine dust pollution and legal limit)



# Data analysis with Python in Jupyter Notebooks

... with the help of already prepared Jupyter Notebooks (= Worked Example)

- to support the individual programming steps
- by prescribing code that students can use/combine/adapt
- to show the students how a Computational Essay can be structured and what needs to be included

(Atkinson et al., 2000; Caspersen & Bennedsen, 2007; Merrienboer & Krammer, 1987)

- see `PDP_Epistemic Programming_Worked_Example.ipynb`

## Visualization of the filtered data sets

Line-Graph for April 12 to April 13 by creating a figure-environment and adding two Scatter-Plots for the filtered data-sets:

```
In [6]: fig = go.Figure()
# Add traces
fig.add_trace(go.Scatter(x=df_hum_filter.index, y=df_hum_filter['value'], mode='lines', name='Humidity'))
fig.add_trace(go.Scatter(x=df_pm_filter.index, y=df_pm_filter['value'], mode='lines', name='Particulate Matter'))
fig.show()
```



## Interpretation of the graphs

As it can be seen here, there is a rise in the humidity-data, once the particulate-matter-value increases. Also, after the particulate-matter-value decreases to its "normal" value, the humidity-curve drops as well.

However, when the humidity-value rises, the particulate-matter-value does not always increase in accordance.



# Hands-On-Activity: Environmental Analysis

**... with the help of the prepared Jupyter Notebook (Worked Example) in the material-folder**

- Build groups of 5 and examine the Jupyter Notebook from the material-folder.
- Create an own Computational Essay for your Environmental Analysis by adapting and combining Code from the Worked Example and by adding Markdown-Cells, in which you should document your code and interpret your results.
- You do not have to use all code cells from the examined Jupyter Notebook, but you can come up with your own evaluation including your own visualizations. Your Computational Essay should contain at least two visualizations including documentation/interpretation.



# In summary: “Epistemic Programming”

- Goal: Programming in order to express oneself, get a differentiated perspective on the world and thus be able to act in a self-determined way.

Regarding Data-Driven Programming:

- Data science with the goal of deriving new insights by analyzing data;
- Gaining knowledge through data as a digital skill or digital competency
  - Highlights the role of digital artefacts and their ubiquity in data mining, collection, and storage, as well as analysis and interpretation
  - clarifies the role of digital data and the ability to gain insights on your own with new approaches and tools

This approach emphasizes:

- Digitization changes the world - If I program epistemically, then I can change the world (in small ways)



# A few more project ideas

- Air pressure and the correlations with the weather
- Development of national/regional Covid-19 cases (or any other disease)
- Success from sports team and its money spent
- ...

**What Ideas do you have?**



# Reflection

## **Discuss the following questions:**

- What have you learnt during the activity?
- Did you use mathematics? When? Examples?
- What did you do very well? Why?
- What went wrong? Why?
- What will you do differently next time?



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