

# Astroparticle physics online masterclass built on the KASCADE Cosmic Ray Data Centre

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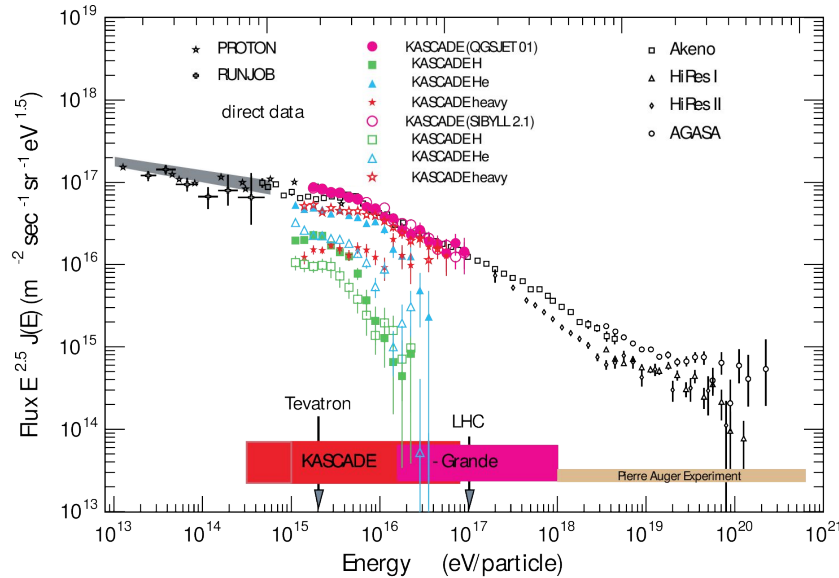
INSTITUT FÜR ASTROTEILCHENPHYSIK (IAP)

18 March 2021, DPG Spring meeting



# Motivation

- Online education: bringing an equal access to education worldwide
- Corona situation
- GRADLCI project and IAP BigData Infrastructure



Primary cosmic ray flux and primary energy range covered by KASCADE-Grande and KASCADE experiments

# IAP BigData Infrastructure timeline

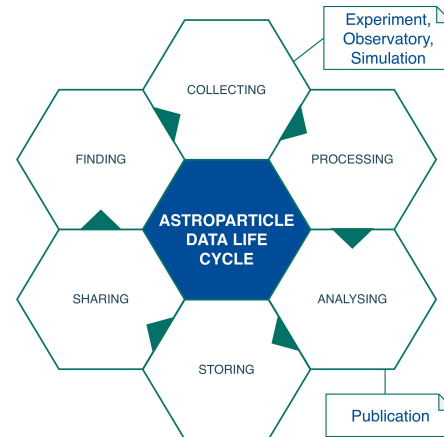
1989 - 2013: Data engineering for KASCADE and other experiments

2013 - now: KASCADE Cosmic-Ray Data Center (KCDC)

- only open-source technologies
- all-in-one: data center, archive, information and educational platform
- <http://kcdc.i kp.kit.edu>

2018 - now: German-Russian Astroparticle Data Life Cycle Initiative (GRADLCI):

- KCDC update
- online analysis server
- machine learning methods
- outreach and education
- <https://gradlc-dc.i kp.kit.edu/>

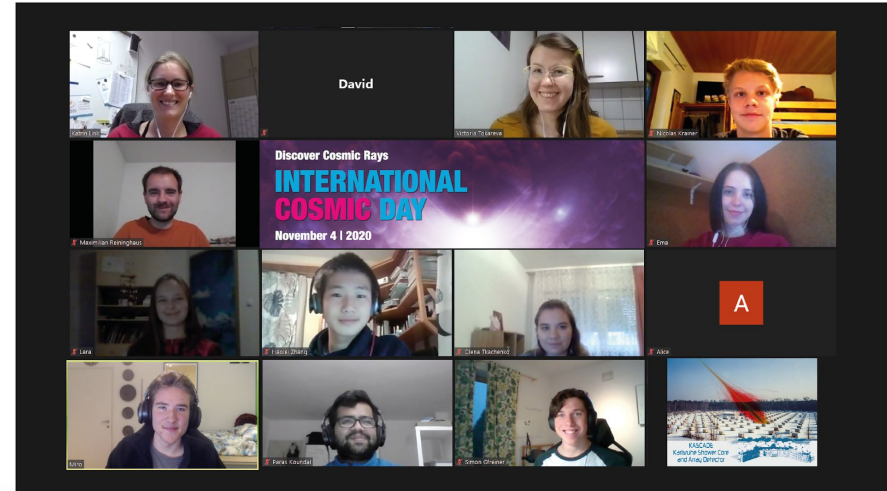


# Educational program

## Lecturers and tutors:

- Maximilian Reininghaus
- Paras Koundal
- Victoria Tokareva
- Katrin Link
- Olena Tkachenko
- Miro Joensuu

14:00	<b>International Cosmic Day: Welcome</b>	
	Online	14:00 - 14:05
	<b>Astroteilchenphysik</b>	
	Online	14:05 - 15:00
15:00	<b>Luftschauerphysik</b>	
	Online	15:00 - 15:30
	<b>Pause</b>	
	Online	15:30 - 15:45
	<b>KASCADE Grande</b>	
	Online	15:45 - 16:00
16:00	<b>KCDC Introduction</b>	
	Online	16:15 - 16:30
	<b>Break</b>	
	Online	16:30 - 16:45
	<b>Python /Data Analysis Introduction</b>	
17:00	Online	16:45 - 17:05
	<b>Data analysis</b>	
	Online	17:05 - 17:55
18:00	<b>Video call</b>	
	Online	17:55 - 18:15



# Requirements to the tutorial environment

- Identical workspace for all participants and tutors
- Save as much time as possible on installation and setup of the working environment
- Fast processing of rather significant amount of data
- Data format, well-known for students
- Interactivity
- Minimal programming knowledge required

# Requirements to the tutorial environment

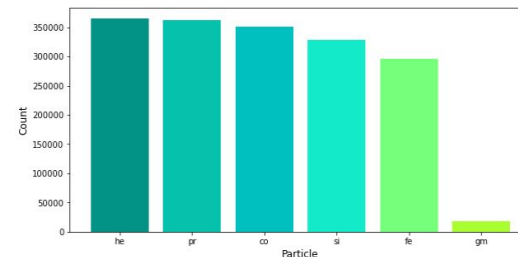
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Solution: JupyterHub, Python3, Pandas+matplotlib, \*.csv, \*.txt.

# Data samples

- Sample 1:
  - KASCADE data shortened
  - .txt
  - 4102999 events

	E	Ne	Nmu	Ze
0	14.9335	4.55575	4.02881	18.2060
1	15.1466	4.83307	4.15094	23.3173
2	14.8551	4.56024	3.86447	21.4443
3	14.6315	4.22959	3.68777	21.1421
4	14.5349	4.54745	3.37969	18.2577



- Sample 2:
  - KASCADE simulations
  - QGS-Jet4 + FLUKA
  - .csv
  - 1720319 events

	Particle	IgE	X	Y	CoreDist	Ze	Az	IgNe	IgNmu
0	co	14.1173	59.4186	-39.391200	71.2898	17.5974	63.79230	3.53951	3.26399
1	co	14.1492	-11.8334	55.367400	56.6179	33.5999	205.23600	3.43951	3.08027
2	co	14.3944	-28.4504	-0.732422	28.4599	15.7591	60.74060	3.99656	3.49747
3	co	14.2358	61.4100	-28.061300	67.5176	26.3883	282.38000	2.98808	3.49411
4	co	14.4836	-45.8457	-35.637500	58.0677	25.4860	6.27402	4.13502	3.43070

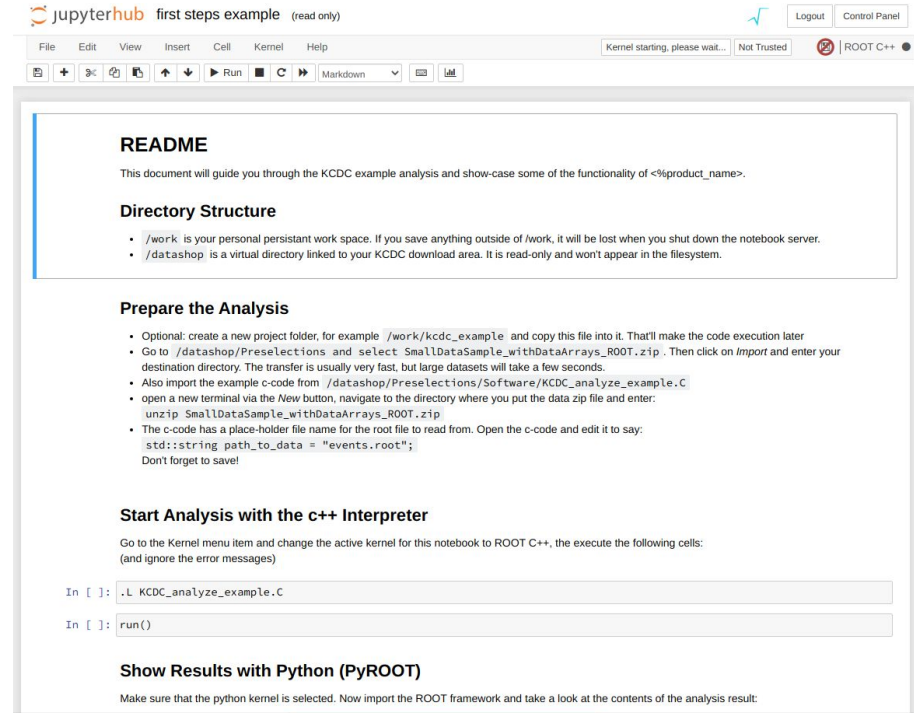
- Sample 3:
  - KASCADE data
  - Sept - Oct 1999
  - .CSV
  - 1500317 events

	datetime	particle_id	energy	zenith	azimuth	core_x	core_y	electron_number	muon_number	shower_age	temperature	pressure
0	1998-05-11 07:54:03	0	14.619700	17.664600	117.292000	-22.430800	13.229600	4.145430	3.748440	1.041510	18.420200	1003.600000
1	1998-05-09 06:15:11	0	14.775900	27.404800	78.212900	68.574100	-23.827000	4.432480	3.697480	1.113110	19.080500	1011.330000
2	1998-05-14 07:50:46	0	14.730100	8.570720	159.879000	-65.301000	-60.631700	4.487310	3.829930	1.010680	13.820400	1007.990000
3	1998-05-16 20:59:44	0	14.576800	24.471400	260.102000	-1.075740	-59.533100	4.400930	3.442650	1.139100	17.870000	1005.960000
4	1998-05-19 01:52:24	0	14.946300	45.756200	3.883100	3.685060	-68.299400	3.947760	3.574800	0.854729	17.809900	1012.410000

# Jupyter Interactive environment



- Well-known
- Different kernels, pyRoot and C++ ROOT support
- Interactive code, easy to test and modify
- Visualizations, arbitrary output



jupyterhub first steps example (read only) Logout Control Panel

File Edit View Insert Cell Kernel Help Kernel starting, please wait... Not Trusted ROOT C++

## README

This document will guide you through the KCDC example analysis and show-case some of the functionality of `<%product_name>`.

### Directory Structure

- `/work` is your personal persistent work space. If you save anything outside of `/work`, it will be lost when you shut down the notebook server.
- `/datashop` is a virtual directory linked to your KCDC download area. It is read-only and won't appear in the filesystem.

### Prepare the Analysis

- Optional: create a new project folder, for example `/work/kcdc_example` and copy this file into it. That'll make the code execution later
- Go to `/datashop/Preselections` and select `SmallDataSample_withDataArrays_ROOT.zip`. Then click on `Import` and enter your destination directory. The transfer is usually very fast, but large datasets will take a few seconds.
- Also import the example c-code from `/datashop/Preselections/Software/KCDC_analyze_example.C`
- open a new terminal via the `New` button, navigate to the directory where you put the data zip file and enter:  

```
unzip SmallDataSample_withDataArrays_ROOT.zip
```
- The c-code has a place-holder file name for the root file to read from. Open the c-code and edit it to say:  

```
std::string path_to_data = "events.root";
```

  
Don't forget to save!

### Start Analysis with the c++ Interpreter

Go to the Kernel menu item and change the active kernel for this notebook to ROOT C++, the execute the following cells:  
(and ignore the error messages)

```
In [ ]: .L KCDC_analyze_example.C
```

```
In [ ]: run()
```

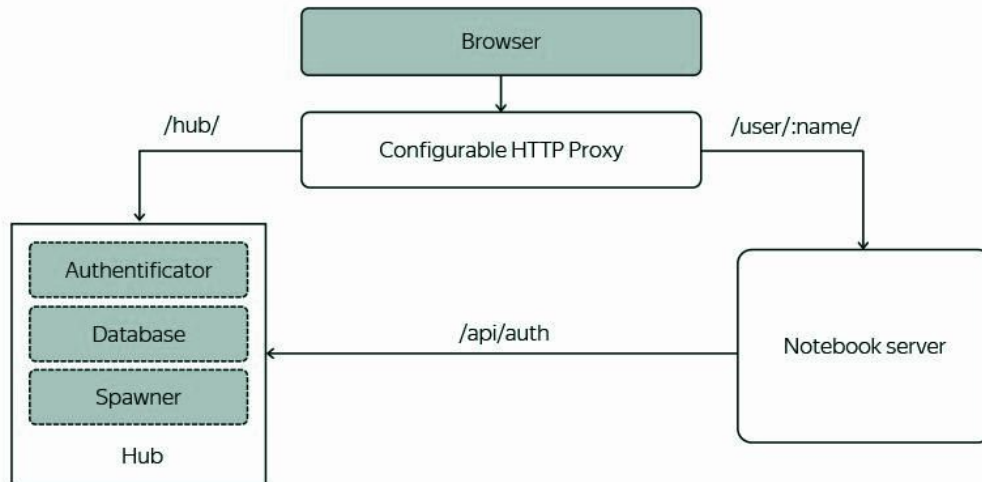
### Show Results with Python (PyROOT)



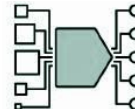

Make sure that the python kernel is selected. Now import the ROOT framework and take a look at the contents of the analysis result:



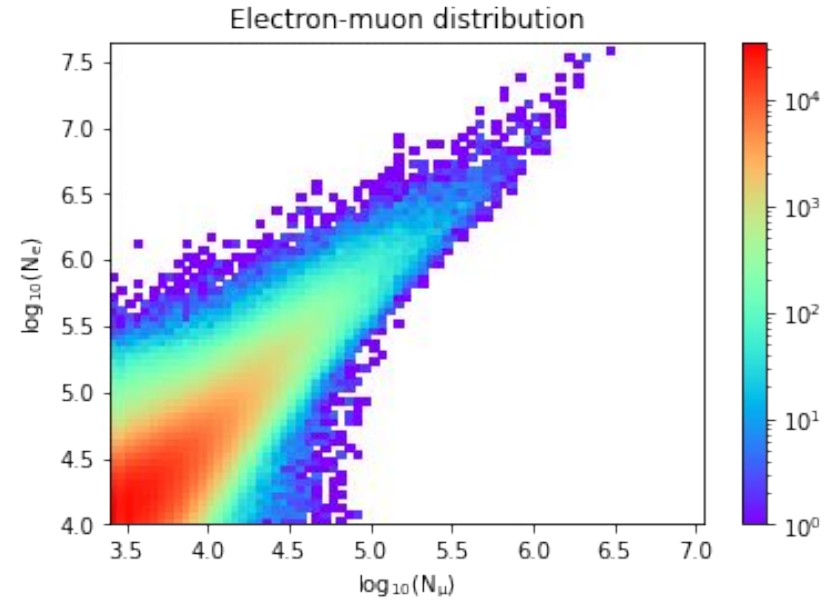
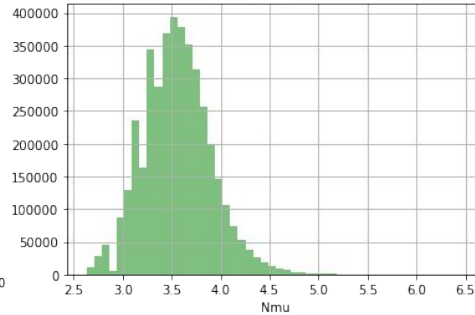
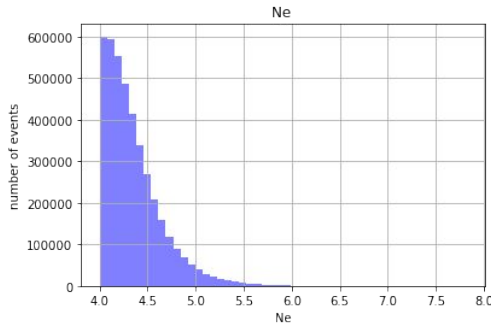
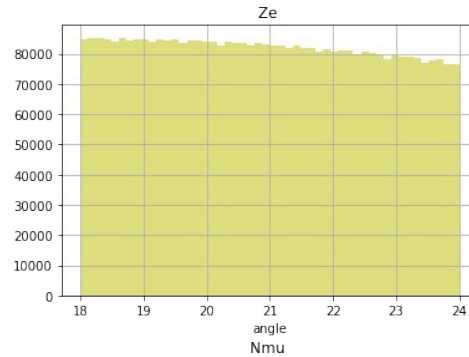
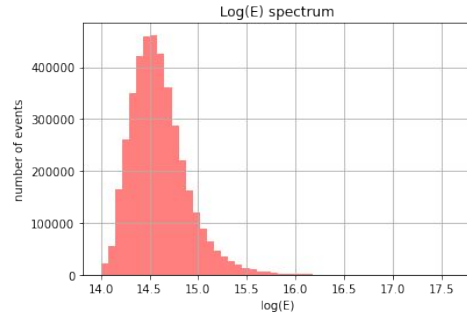
# JupyterHub @ IAP

- Login via KCDC credentials
- Administration using Docker Swarm
- Up to 24 connections at the same time

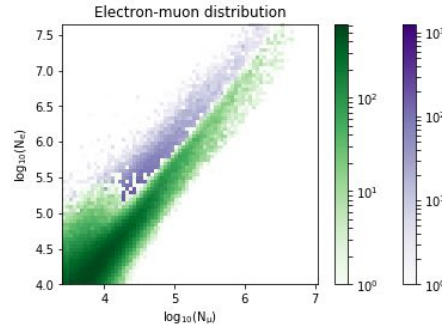
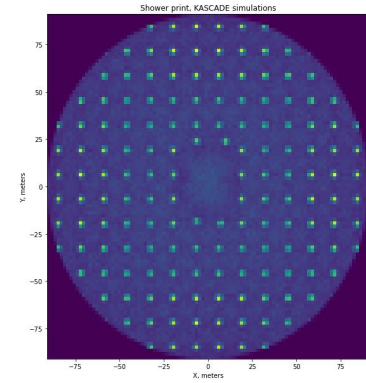
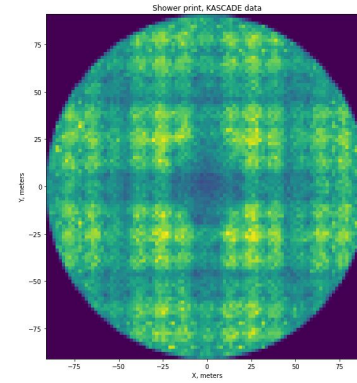
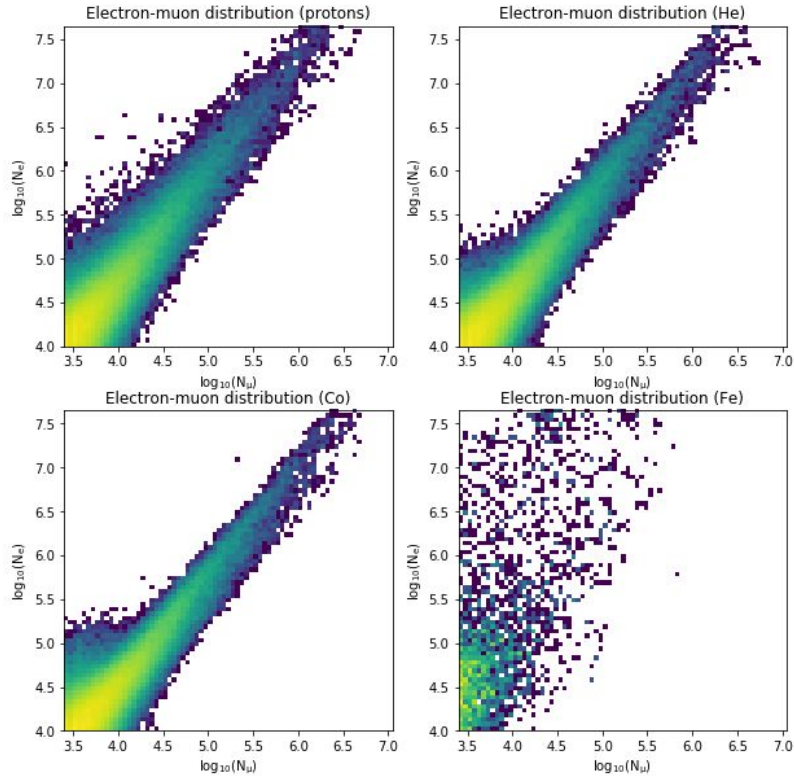


<p>Authorization</p> 	<p>Run individual Jupyter servers</p> 
<p>Proxy to Jupyter servers</p> 	<p>System and database management</p> 

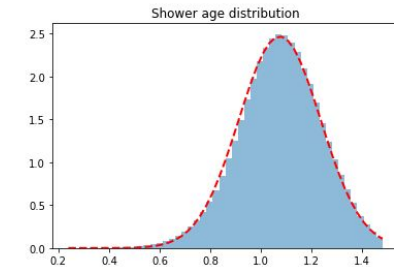
# Particle mass tutorial



# Particle mass tutorial



$\mu = 1.0742623777506533$ ,  $\sigma = 0.16167346289339393$



## What have you done? Part II

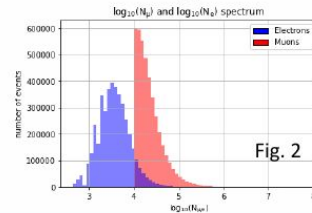
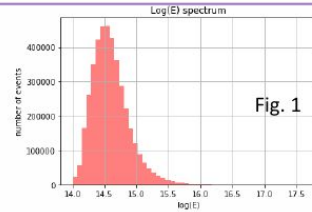
During the data analysis we split into smaller groups of two or three, each working together with one tutor from the KIT and went through problems using Jupyter Notebook. First, some easier examples were done, such as how to import and use libraries, as well as plotting simple histograms. Then data from the KCDC library of both real and simulated events was imported. In the last step, plots of muon to electron ratios were made and compared with the results from simulated events.

## What did you find out? Part I

The number of events decreases overall with energy, because higher energy events creating these types of particles are rarer (fig.1).

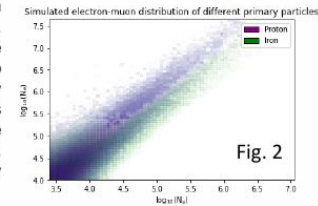
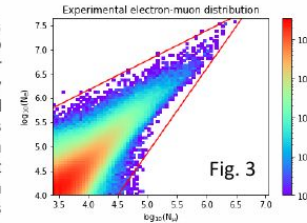
Another interesting observation was that the electron and muon distributions are shifted to one another (fig.2).

Cosmic ray events create electrons and muons unequally. In general, more electrons are created in total, than muons, but the former will scatter away, while the latter can reach the ground more easily.



## What did you find out? Part II

Events with heavier primary particles, such as iron cores, which tend to interact with the atmosphere earlier have more muons, than light primary particles, which tend to interact later and have an abundance of electrons. This can be seen in the 2d histograms from simulated events, which use different initial particles (as seen in fig. 3), such as iron or protons. These 2d histograms show the number of events with a specific number of electron and muon combinations through colour intensity. The results from the simulations can be used in experimental data to differentiate between different primary particles. Particles leading to events closer to the bottom to figure 4 are more likely to have heavy initial particles, while events closer to the top probably have light primary particles.



## What's your take-home message?

Apart from the extremely interesting theoretical input regarding cosmic rays, the event gave us an insight into how data analysis works and how meaningful conclusions can be drawn from a large amount of data. This enabled us a sneak peek into how science is conducted in real life, which we greatly appreciated.

# Outlook

- The BigData Infrastructure of IAP KIT and experience of our team in organisation of outreach events enabled us to organize an online data analysis masterclass
- In order to do so, we implemented an educational program and a masterclass program, organized preliminary data selection and preprocessing, instructed our students about work with KCDC and Jupyter Notebook in JupyterHub environment
- The master class was attended by 9 students aged 14-19 from Villach (Austria). Good beta testing for our outreach methodology and analysis environment at the same time
- Students acquired new knowledge and skills while working with open data and presented their results to other participants of ICD from Italy and Great Britain. Interesting discussion!
- Future plans include expanding our available materials with masterclasses that include machine learning, enriching our methodology with automated code validation capabilities, and further development of our BigData infrastructure

Thank you for your attention!

Contact: [Victoria.Tokareva@kit.edu](mailto:Victoria.Tokareva@kit.edu)