

Visualizations in Clinical Trials using Power BI and Python



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Disclaimer

The information and opinions expressed in this presentation are solely my own and do not express the views or opinions of my employer, Bayer.

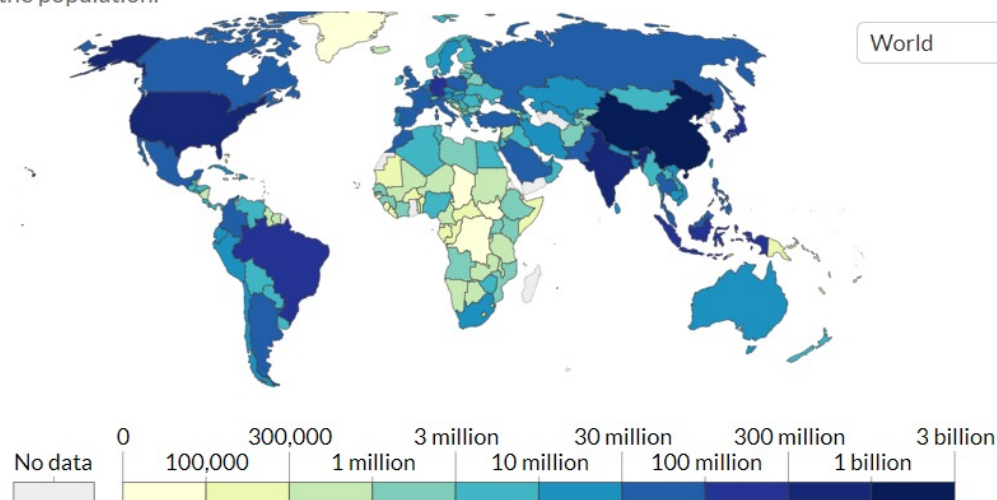


Let's start with some visualizations

COVID-19 vaccine doses administered, Sep 3, 2021

For vaccines that require multiple doses, each individual dose is counted. As the same person may receive more than one dose, the number of doses can be higher than the number of people in the population.

Our World in Data



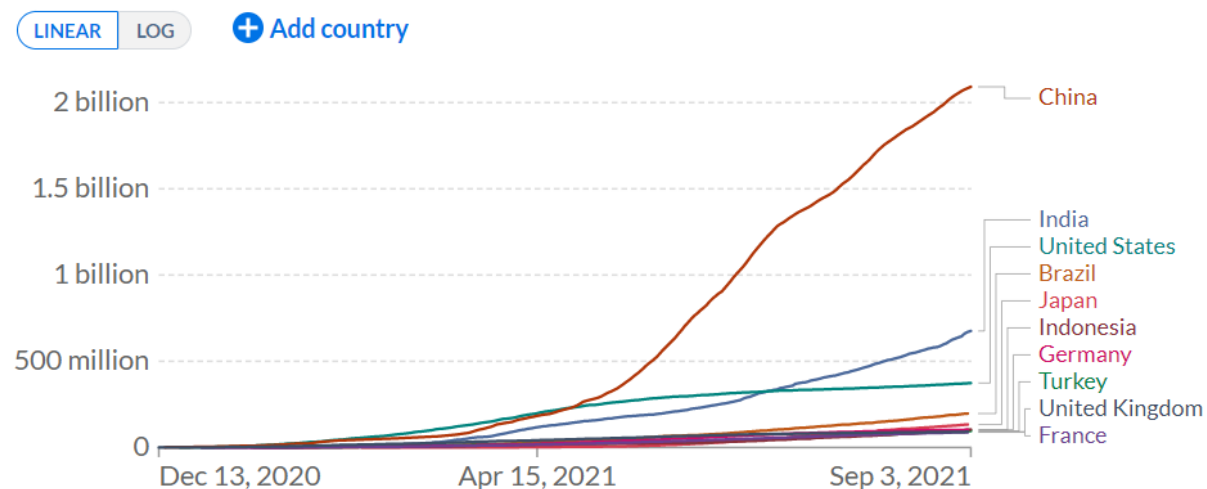
Source: Official data collated by Our World in Data - Last updated 4 September 2021, 09:00 (London time)
OurWorldInData.org/coronavirus • CC BY

▶ Dec 2, 2020 ◯ Sep 3, 2021

COVID-19 vaccine doses administered

For vaccines that require multiple doses, each individual dose is counted. As the same person may receive more than one dose, the number of doses can be higher than the number of people in the population.

Our World in Data



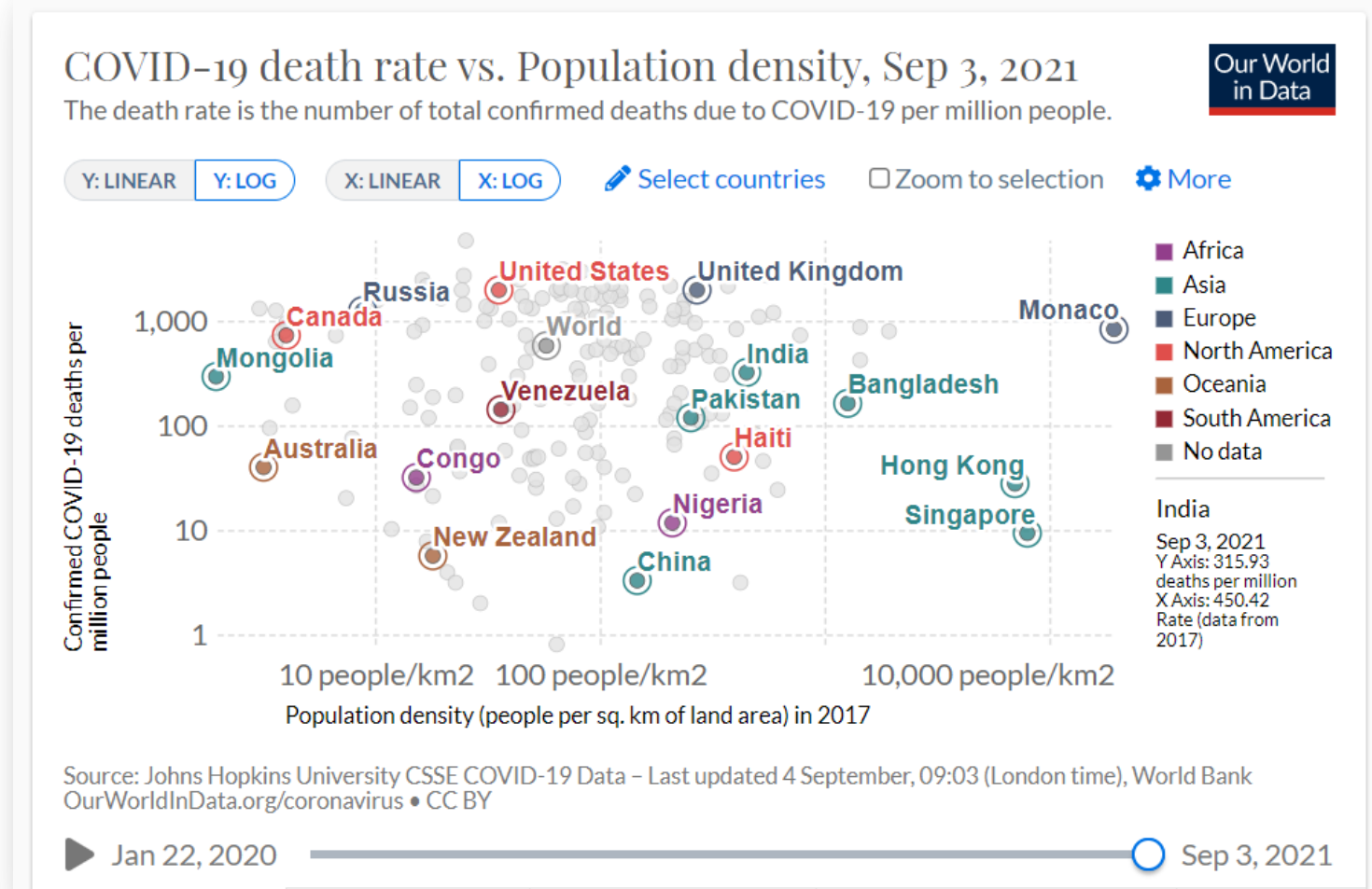
Source: Official data collated by Our World in Data - Last updated 4 September 2021, 09:00 (London time)
OurWorldInData.org/coronavirus • CC BY

▶ Dec 13, 2020 ◯ Sep 3, 2021



Why visualizations?

- Quick and easy way of understanding data
- Most efficient holistic way of looking at data
- Enable us to draw quick insights and faster decision making
- Makes the data more natural for the human mind to comprehend
- Makes it easier to identify trends, patterns, and outliers
- Communicates information in a fast and effective way
- Sometimes answers the questions that we don't know to ask
- Promotes storytelling





Popular Data Visualization Tools in Clinical Research Domain

No-code/low-code



Power BI



Power BI

Google



Data Studio

Programming Languages





Power BI

Pros

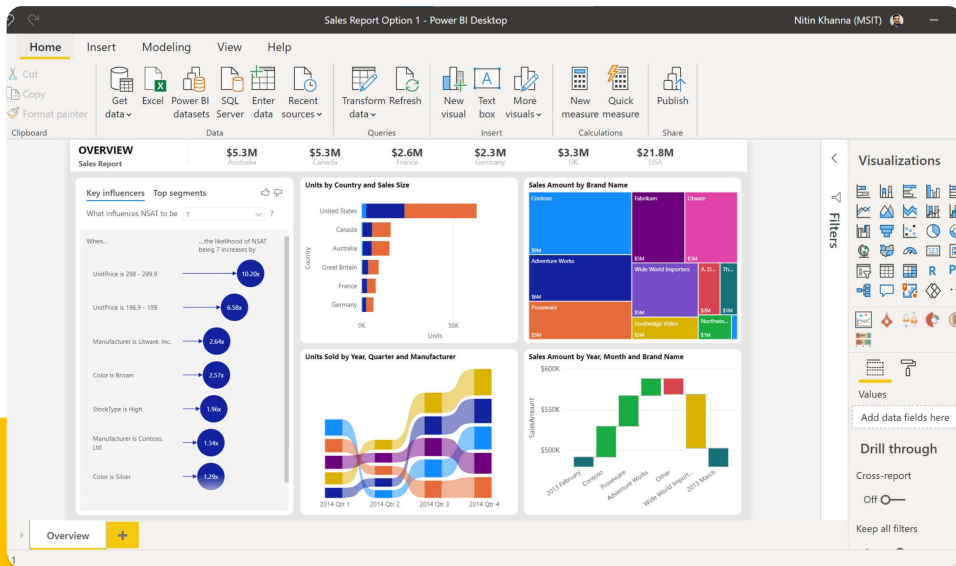
- Free to use, easy to learn
- No code/Low code solution
- Interactive Visualizations
- Intuitive
- Less Error Prone
- Distribution is easy (Requires license)
- Easy data wrangling
- Carries data
- Seamless integration with Microsoft platforms

Cons (Vs Python)

- Limitations in customization
- Handling very large datasets can be challenging
- Easy to learn, complex to master

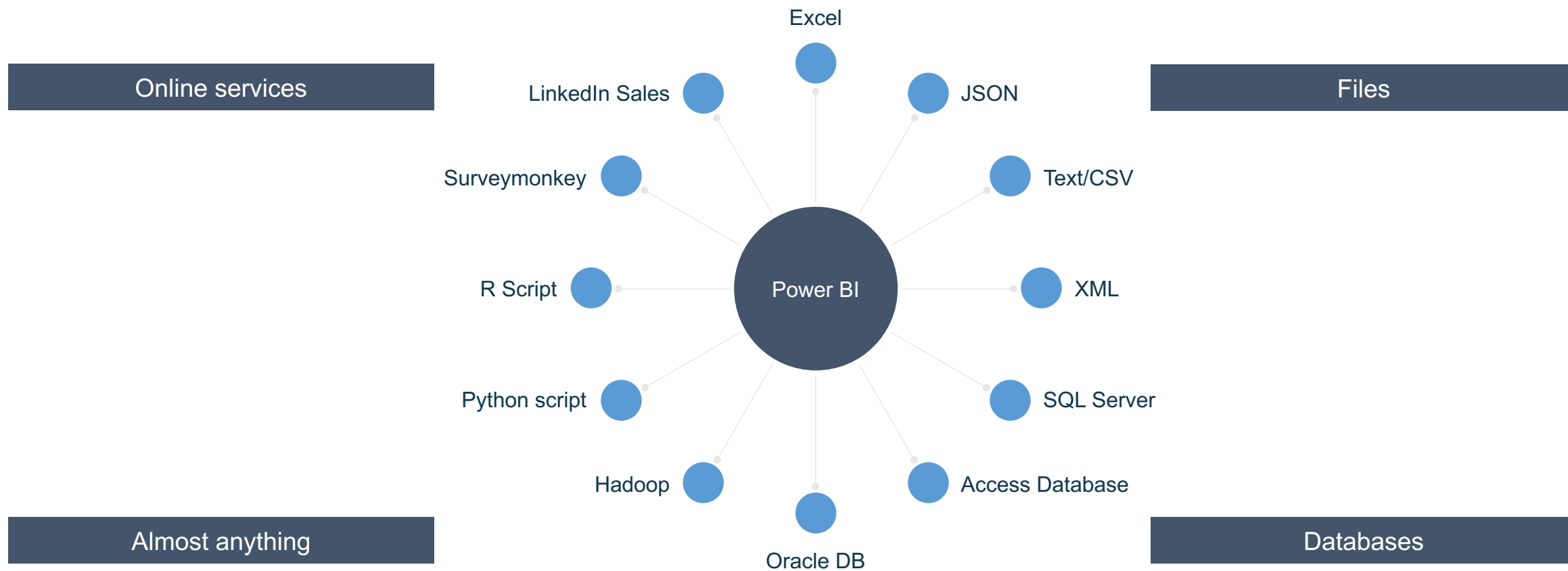
Resources

- Download from this [link](#)





Data Connections





Use case

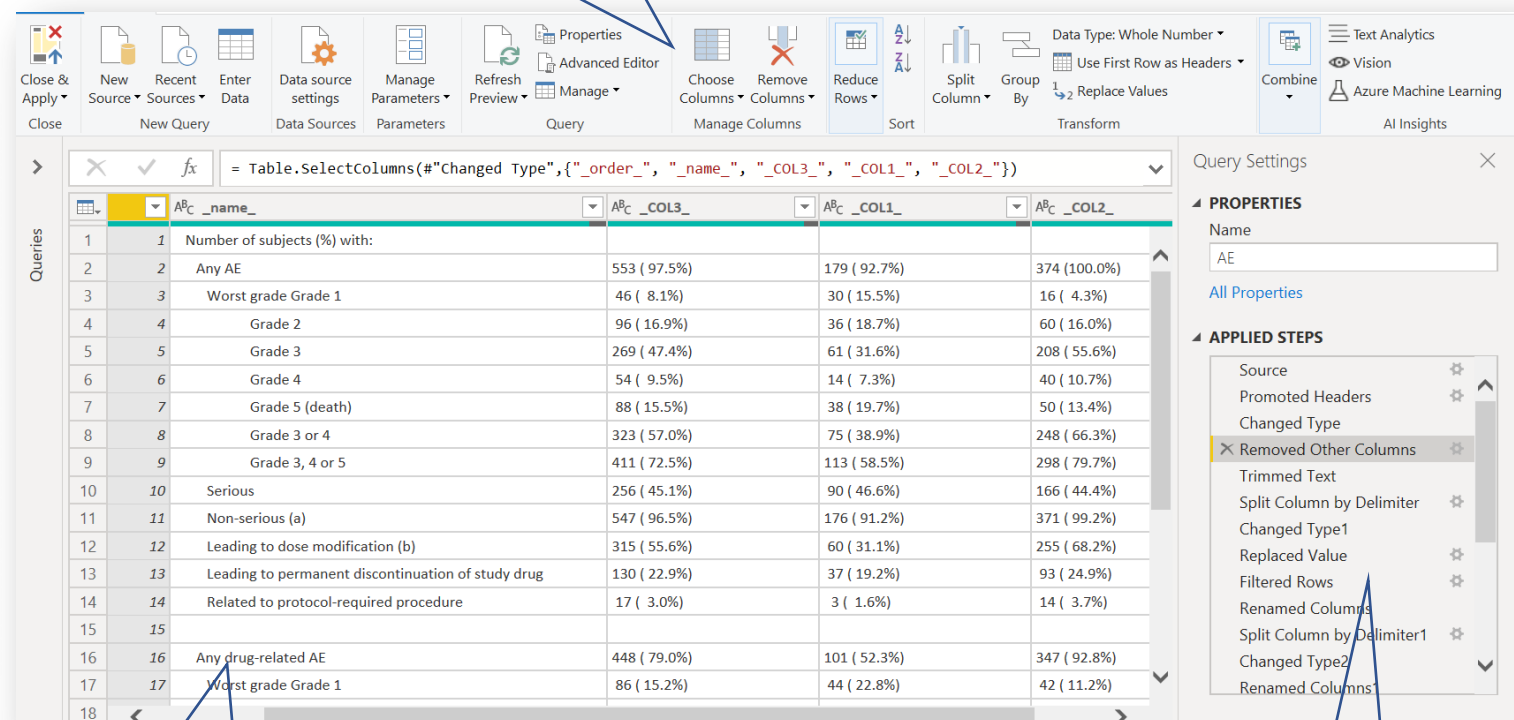
From **Adverse events summary table** to **visualization** in less than 30 minutes with Power BI

- Convert the SAS dataset to .csv file and import data in Power BI
- Transform data
- Create visualizations



Power BI Data Transformation

Data Transformation Tools



The screenshot shows the Power BI Data Transformation Tools ribbon and the Query Editor. The ribbon includes tabs for Query, Transform, and Load. The Query tab is active, showing options like Close & Apply, New Source, Recent Sources, Enter Data, Data source settings, Manage Parameters, Refresh Preview, Properties, Advanced Editor, Choose Columns, Remove Columns, Reduce Rows, Sort, Split Column, Group By, Data Type, Use First Row as Headers, and Replace Values. The Query Editor shows a table with columns: ABc_name_, ABc_COL3_, ABc_COL1_, and ABc_COL2_. The table contains 18 rows of data. The 'Applied Steps' pane on the right shows a list of steps: Source, Promoted Headers, Changed Type, Removed Other Columns, Trimmed Text, Split Column by Delimiter, Changed Type1, Replaced Value, Filtered Rows, Renamed Columns, Split Column by Delimiter1, Changed Type2, and Renamed Columns1. The 'Removed Other Columns' step is highlighted.

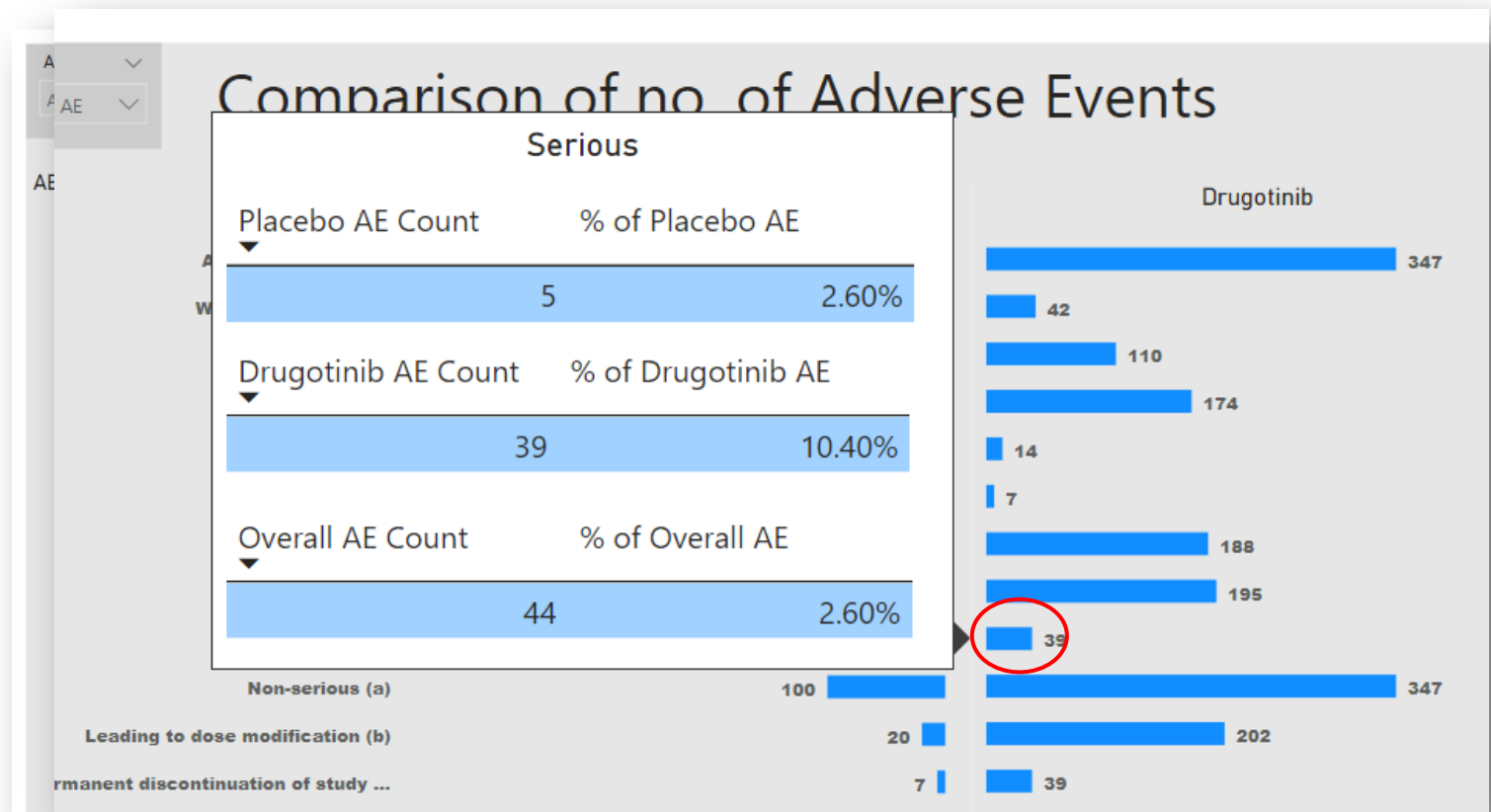
	ABc_name_	ABc_COL3_	ABc_COL1_	ABc_COL2_
1	1	Number of subjects (%) with:		
2	2	Any AE	553 (97.5%)	374 (100.0%)
3	3	Worst grade Grade 1	46 (8.1%)	16 (4.3%)
4	4	Grade 2	96 (16.9%)	60 (16.0%)
5	5	Grade 3	269 (47.4%)	208 (55.6%)
6	6	Grade 4	54 (9.5%)	40 (10.7%)
7	7	Grade 5 (death)	88 (15.5%)	50 (13.4%)
8	8	Grade 3 or 4	323 (57.0%)	248 (66.3%)
9	9	Grade 3, 4 or 5	411 (72.5%)	298 (79.7%)
10	10	Serious	256 (45.1%)	166 (44.4%)
11	11	Non-serious (a)	547 (96.5%)	371 (99.2%)
12	12	Leading to dose modification (b)	315 (55.6%)	255 (68.2%)
13	13	Leading to permanent discontinuation of study drug	130 (22.9%)	93 (24.9%)
14	14	Related to protocol-required procedure	17 (3.0%)	14 (3.7%)
15	15			
16	16	Any drug-related AE	448 (79.0%)	347 (92.8%)
17	17	Worst grade Grade 1	86 (15.2%)	42 (11.2%)
18	18			

Untidy input data

Data Transformation Steps



Adverse Events Visualization



Hover mouse pointer to see more details



What else can Power BI do?

- Drill through the visualizations
- Automated data refresh
- Send automated mails
- Accessible through smartphone app
- Embed the dashboard in apps, SharePoint, Teams
- Export the dashboard report in ppt or pdf
- Custom Visualizations
- Power BI analytics (Automated) and narratives
- Export data in excel



Learn from



Challenges, projects, prizes, datasets



Questions?





Pros

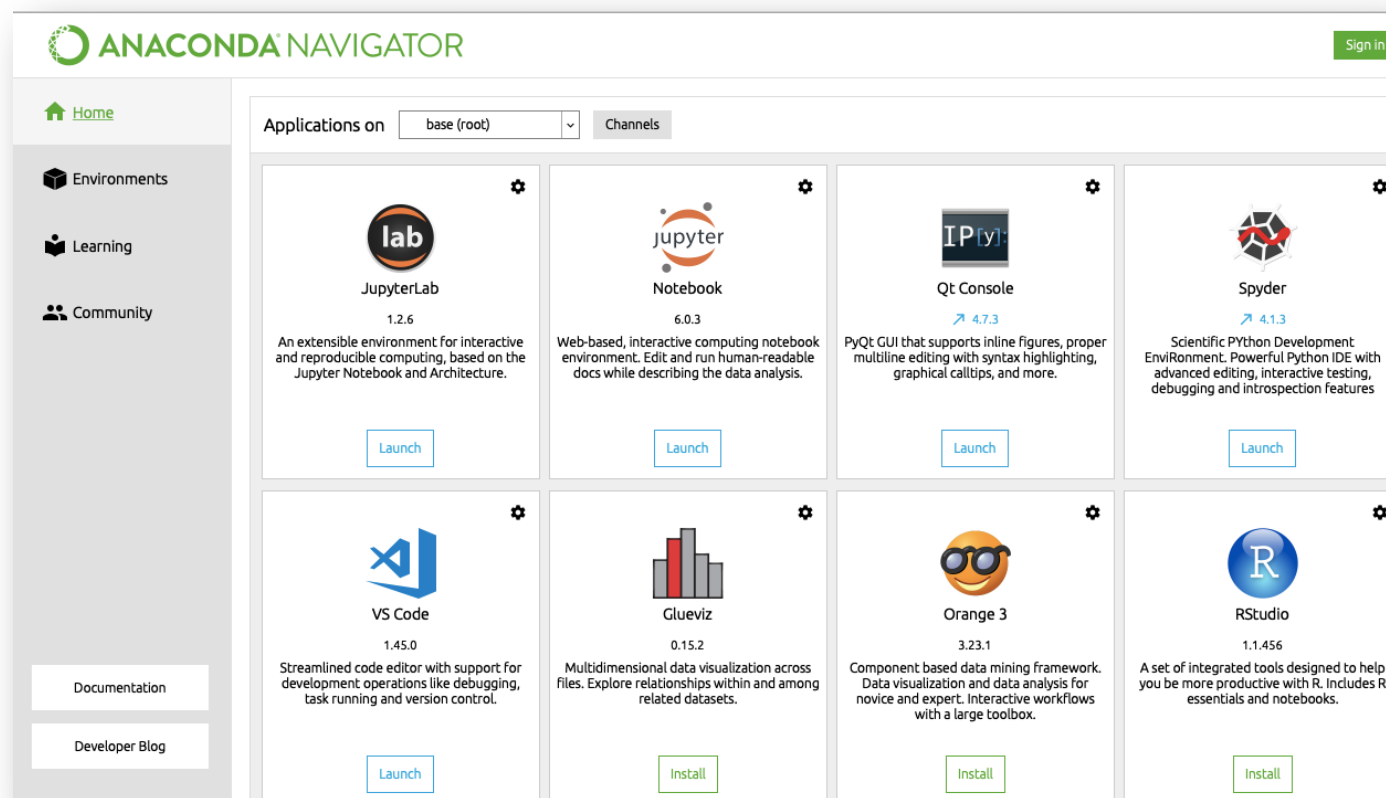
- “The” tool for data science
- Free and open source
- Simple and easy to use
- Static as well as Interactive visualizations
- It’s a complete programming language
- Scalability
- Various libraries/packages
- Latest machine learning libraries

Cons (Vs Power BI)

- Quite a lot of coding
- More Error Prone
- Distribution and deployment is not easy
- Data is stored separately

Resources

- Download from this [link](#)





Use case

Adverse Events Visualization in python in an hour

- Read data from SAS dataset
- Transform the data to make them tidy
- Use “*plotly*” library to create interactive visualizations
- Use “*matplotlib*” or “*seaborn*” library to create static visualizations



Coding in python using Spyder

```
141 #####
142 |
143 #https://plotly.com/python/subplots/
144
145 #Interactive visualizations
146 #####
147
148 import os
149 import plotly.offline as pyo
150 import plotly.graph_objs as go
151 from plotly.subplots import make_subplots
152
153 #####
154
155 from ToplineResults_Preprocessing import df_anyAE
156
157 Placebo = df_anyAE['Placebo_1']
158 Placebo_size = df_anyAE['Placebo_2'].apply(lambda x: 10 if x <= 10 else 10+x/2)
159 AE = df_anyAE['AE']
160
161 fig = make_subplots(rows=1,
162                     cols=2,
163                     shared_yaxes=True,
164                     subplot_titles=("Any AE",
165                                   "Any drug-related AE"))
166
167 #Adding traces
168 fig.add_trace(go.Scatter(
169     x=Placebo,
170     y=AE,
171     marker=dict(color="crimson", size=Placebo_size),
172     mode="markers",
173     name="Placebo"
174 ), row=1, col=1)
```

Lines of code required

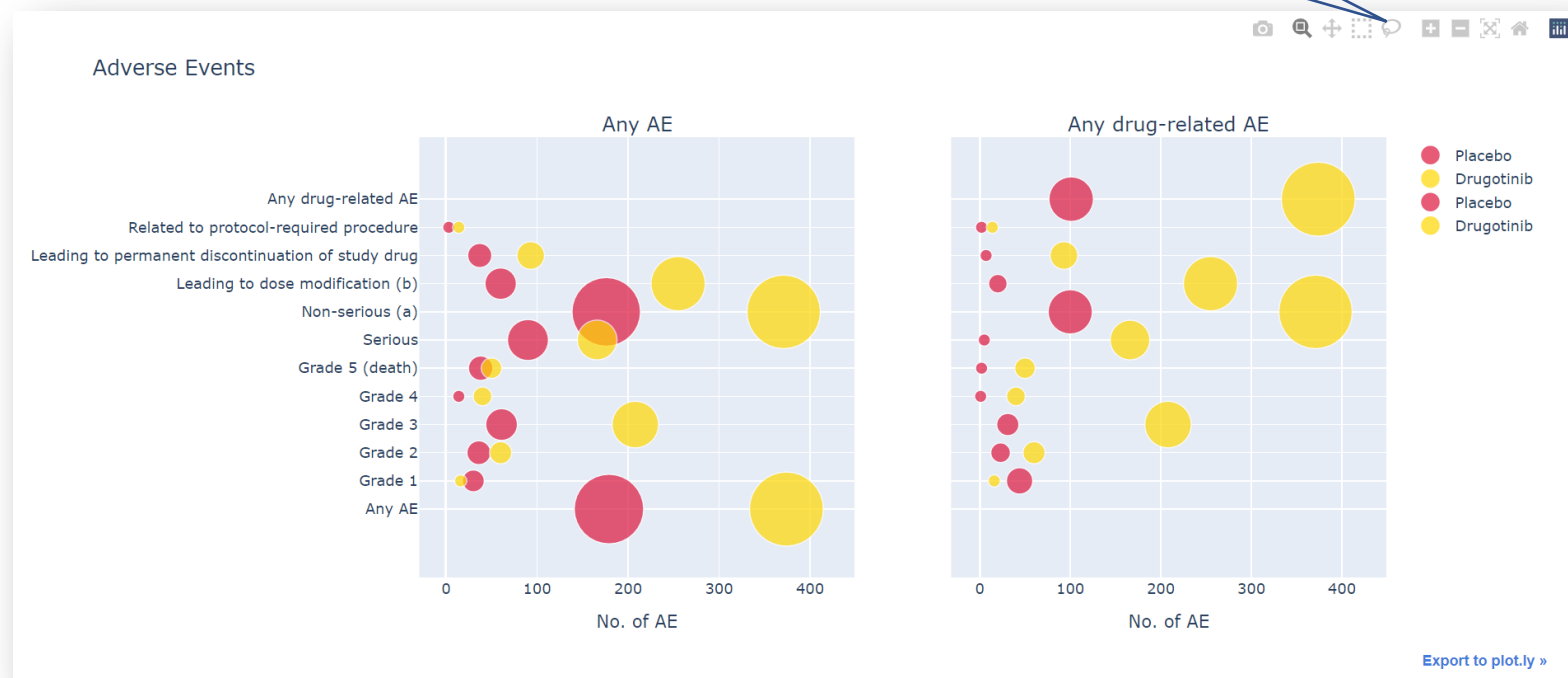
- Around 100 lines of code
- 70 lines for data transformation
- 30 lines for interactive visualization

Write data transformation code in editor



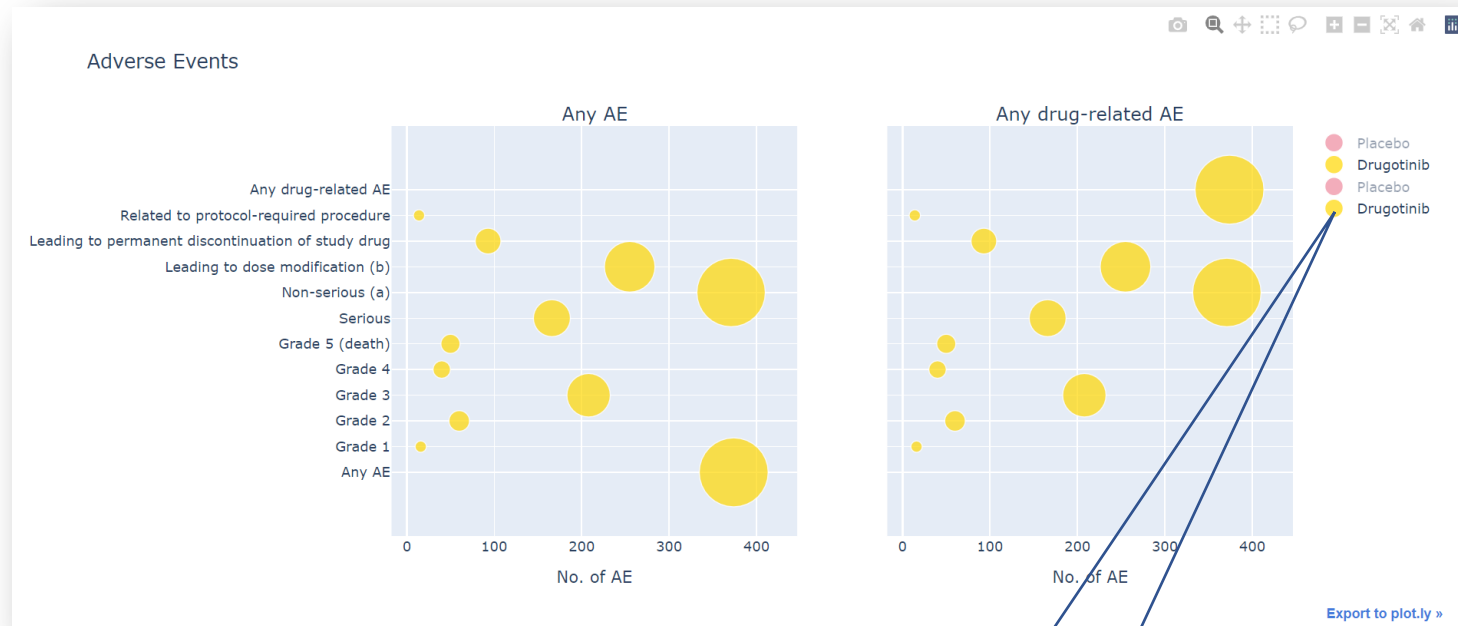
Export static image,
zoom in and zoom
out, focus on specific
area etc.

Adverse Events Visualization in .html file





Interactivity of plotly visualizations



Select one arm to
see only the chart
for that arm



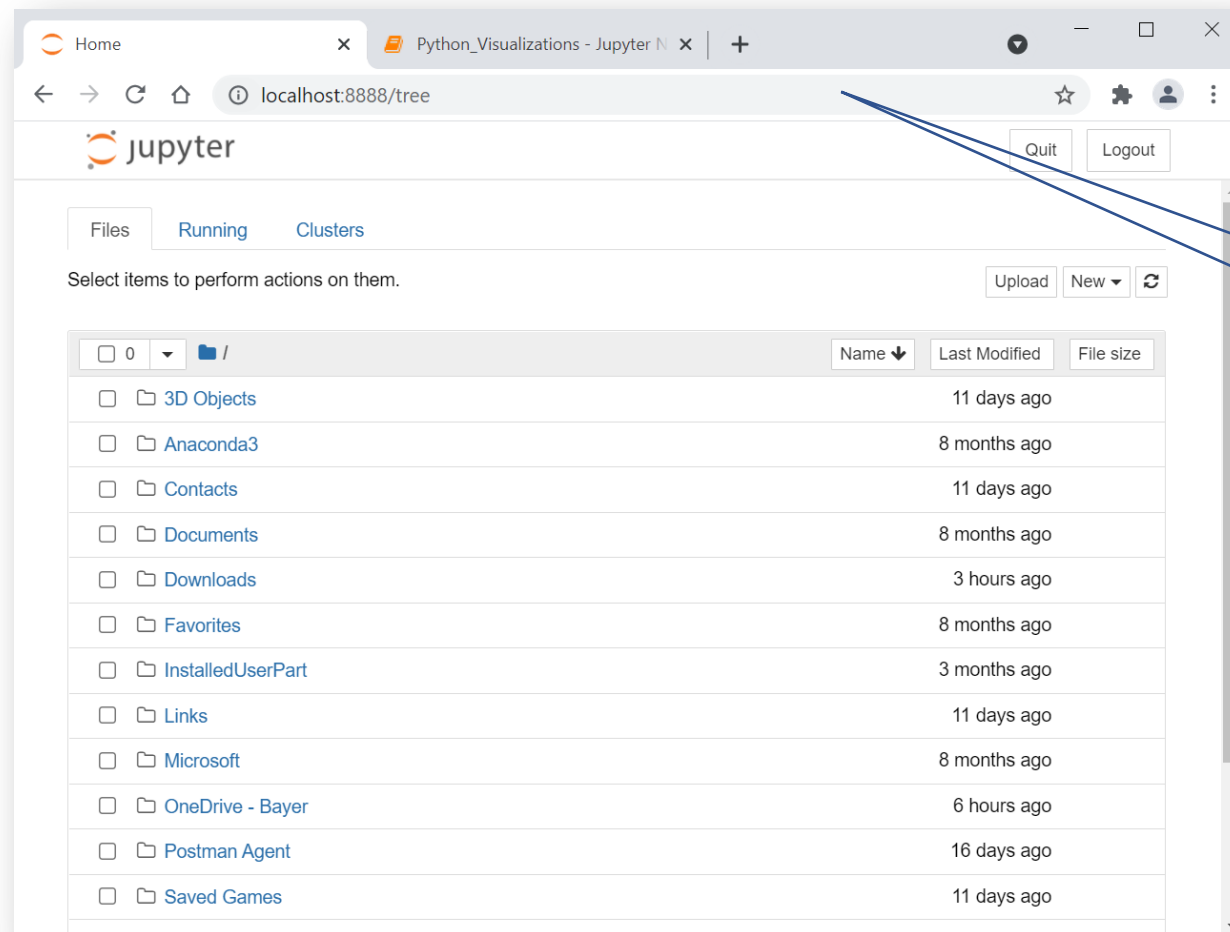
Tooltips in plotly visualizations



Hover mouse
pointer on points
to see more details.



Visualization in jupyter Notebook



Opens in a
browser



Code in Jupyter notebook

The screenshot shows a Jupyter Notebook titled "Python_Visualizations - Illustrations" running on a local host. The interface includes a menu bar (File, Edit, View, Insert, Cell, Kernel, Widgets, Help) and a toolbar with icons for saving, running, and other actions. The notebook content is as follows:

```
Different useful python modules:
```

- Pandas - <https://pandas.pydata.org/>
- Time - <https://docs.python.org/3/library/time.html>
- SAS7BDAT - <https://pypi.org/project/sas7bdat/>
- Pyreadstat - <https://pypi.org/project/pyreadstat/>
- Matplotlib - <https://matplotlib.org/>
- Numpy - <https://numpy.org/>

```
In [1]: #Importing necessary modules
import pandas as pd
from time import process_time, perf_counter
from sas7bdat import SAS7BDAT
import pyreadstat
import matplotlib.pyplot as plt
import numpy as np

In [2]: #SAS files

#SAS file on local drive 11 MB
sas_file_local = r'C:\Users\Bhavini Solanki\Desktop\Workspace\07.Meta data projec

#SAS file on server 11 MB
```

Equivalent
to
R Markdown
file

Write code
in cells



Output in Jupyter notebook

In [16]: #Cleaning and making the SAS dataset tidy

#Adverse events

ae_sas_file = r'C:\Users\Bhavin Solanki\Desktop\Workspace\01.Topline Results\03.SAS Datasets\t_ae.sas7bdat'

df = pd.read_sas(ae_sas_file, encoding="latin1")

#checking the dataframe

df

Out[16]:

	order	_name_	_COL3_	_COL1_	_COL2_	fline	error	_orderN	_obsid_	_1_b1	_1_b2	_1_b3
0	1.0	Number of subjects (%) with:	NaN	NaN	NaN	1.0	0.0	1.0	1.0	1.0	1.0	1.0
1	2.0	Any AE	553 (97.5%)	179 (92.7%)	374 (100.0%)	2.0	0.0	2.0	2.0	1.0	1.0	1.0
2	3.0	Worst grade	Grade 1	46 (8.1%)	30 (15.5%)	16 (4.3%)	3.0	0.0	3.0	3.0	1.0	1.0
3	4.0	Grade 2	96 (16.9%)	36 (18.7%)	60 (16.0%)	4.0	0.0	4.0	4.0	1.0	1.0	1.0
4	5.0	Grade 3	269 (47.4%)	61 (31.6%)	208 (55.6%)	5.0	0.0	5.0	5.0	1.0	1.0	1.0
5	6.0	Grade 4	54 (9.5%)	14 (7.3%)	40 (10.7%)	6.0	0.0	6.0	6.0	1.0	1.0	1.0
6	7.0	Grade 5 (death)	88 (15.5%)	38 (19.7%)	50 (13.4%)	7.0	0.0	7.0	7.0	1.0	1.0	1.0
7	8.0	Grade 3 or 4	323 (57.0%)	75 (38.9%)	248 (66.3%)	8.0	0.0	8.0	8.0	1.0	1.0	1.0
8	9.0	Grade 3, 4 or 5	411 (72.5%)	113 (58.5%)	298 (79.7%)	9.0	0.0	9.0	9.0	1.0	1.0	1.0
9	10.0	Serious	256 (45.1%)	90 (46.6%)	166 (44.4%)	10.0	0.0	10.0	10.0	1.0	1.0	1.0

Prints the result just below the "code" cell



Output in Jupyter notebook

```
In [17]: #List of treatments in the clinical study
trt = ['Placebo', 'Drugotinib']

#Columns to keep
cols_to_keep = ['_name_', '_COL1_', '_COL2_']

df1 = df[cols_to_keep].dropna() #Method chaining
df1['_name_'] = df1._name_.str.strip()
df1.columns = ["AE"] + trt

#Checking first few lines
df1.head()
```

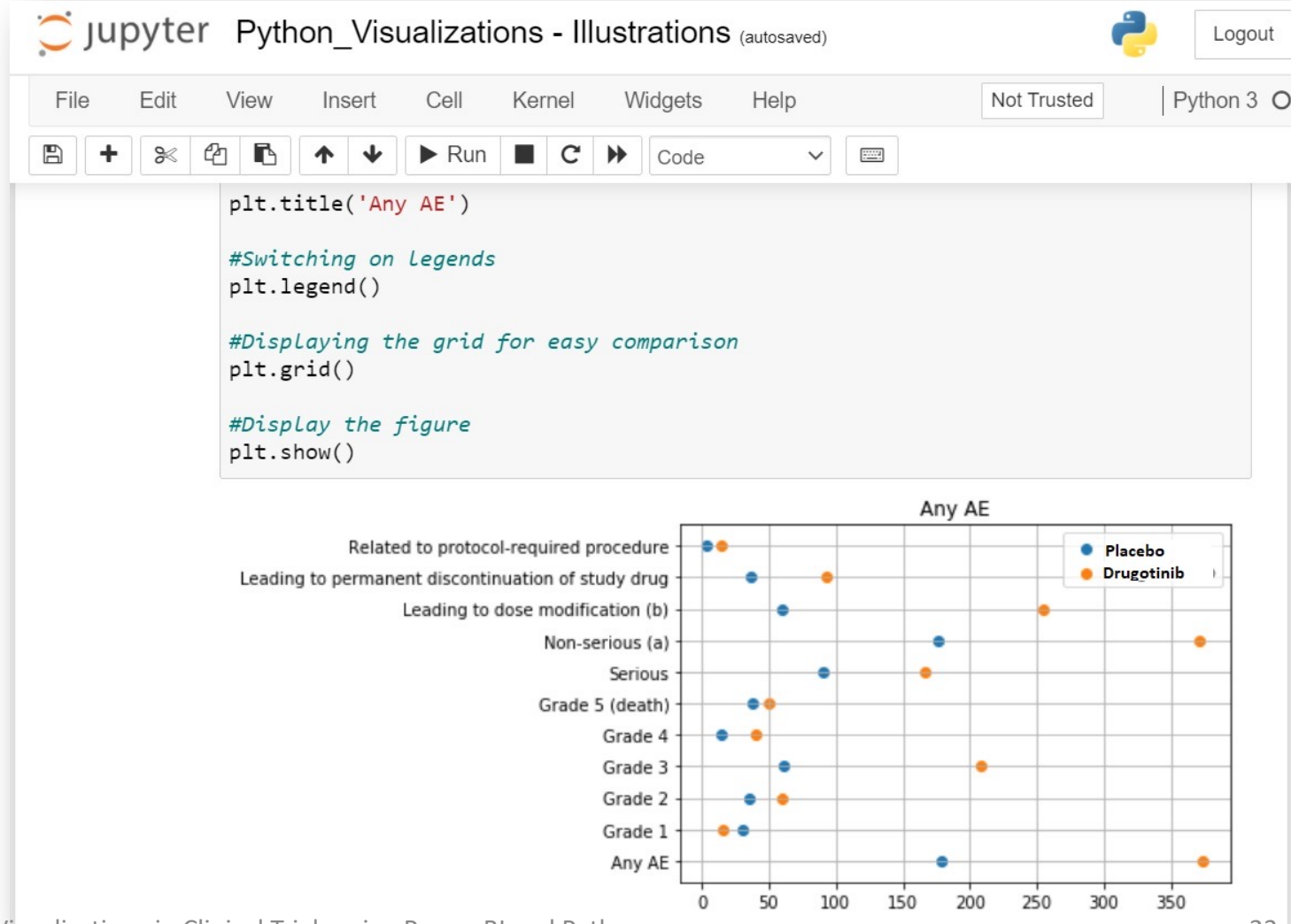
Out[17]:

	AE	Placebo	Drugotinib
1	Any AE	179 (92.7%)	374 (100.0%)
2	Worst grade		
	Grade 1	30 (15.5%)	16 (4.3%)
3	Grade 2	36 (18.7%)	60 (16.0%)
4	Grade 3	61 (31.6%)	208 (55.6%)
5	Grade 4	14 (7.3%)	40 (10.7%)

Quick
summary
table



Static Visualization in Jupyter notebook





What else can python do?

- Data Science
 - Handling Data
 - Visualizations
 - Machine learning
 - Scrap websites
- Web applications
- Desktop applications
- Websites
- Mobile applications
- Just about anything that excludes low-level machine functions and writing operating systems



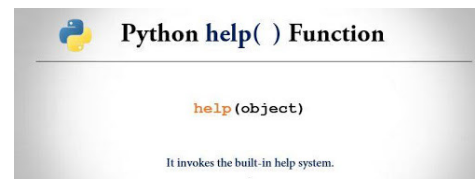
Learn from



Challenges, projects, prizes, datasets



Questions?





Use case

Dashboard created in Power BI for Clinical Trial Scientists for Medical Review

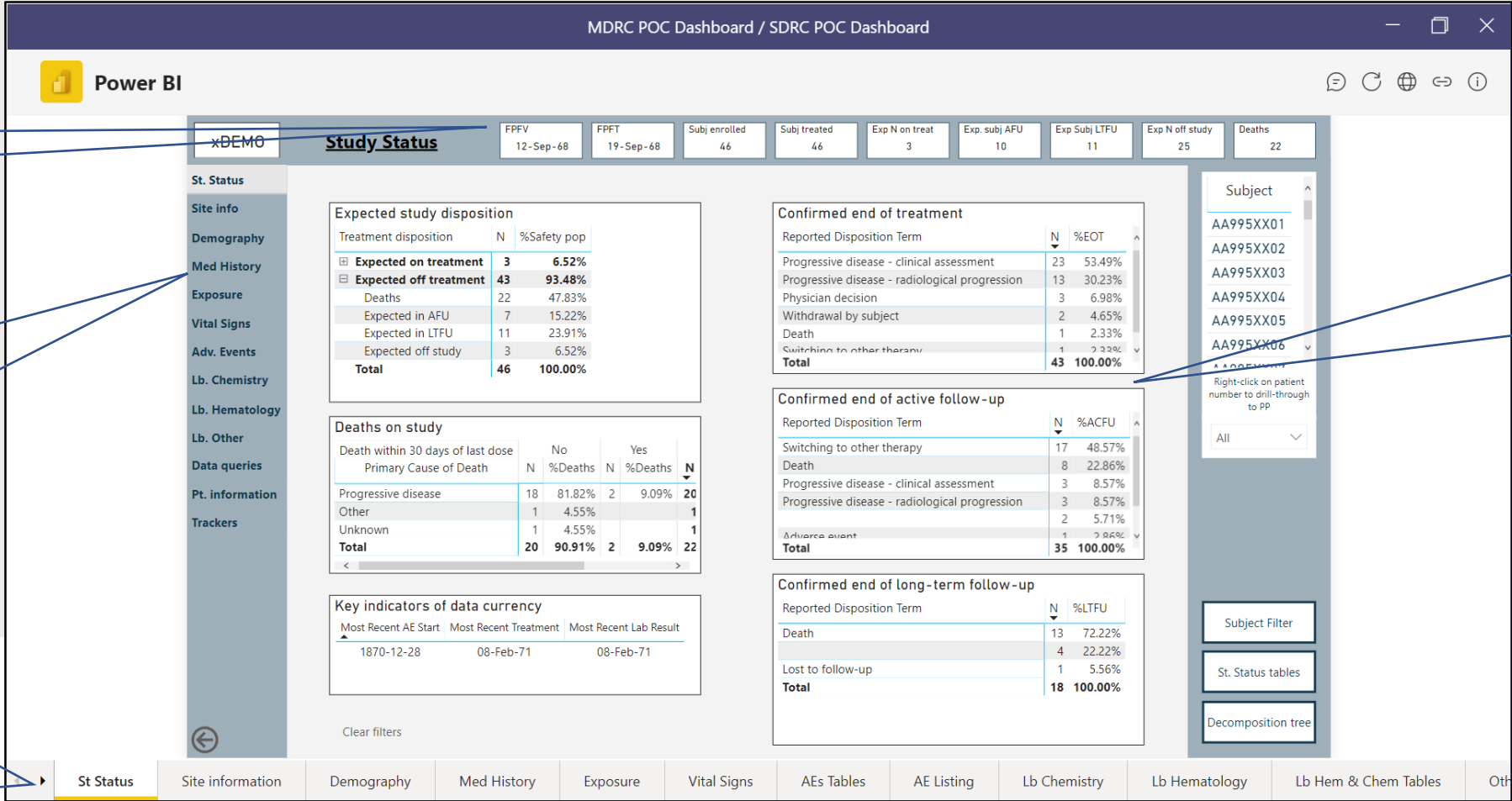


Power BI Dashboard embedded in Teams Channel

Information Cards

Navigation links created for easy navigation

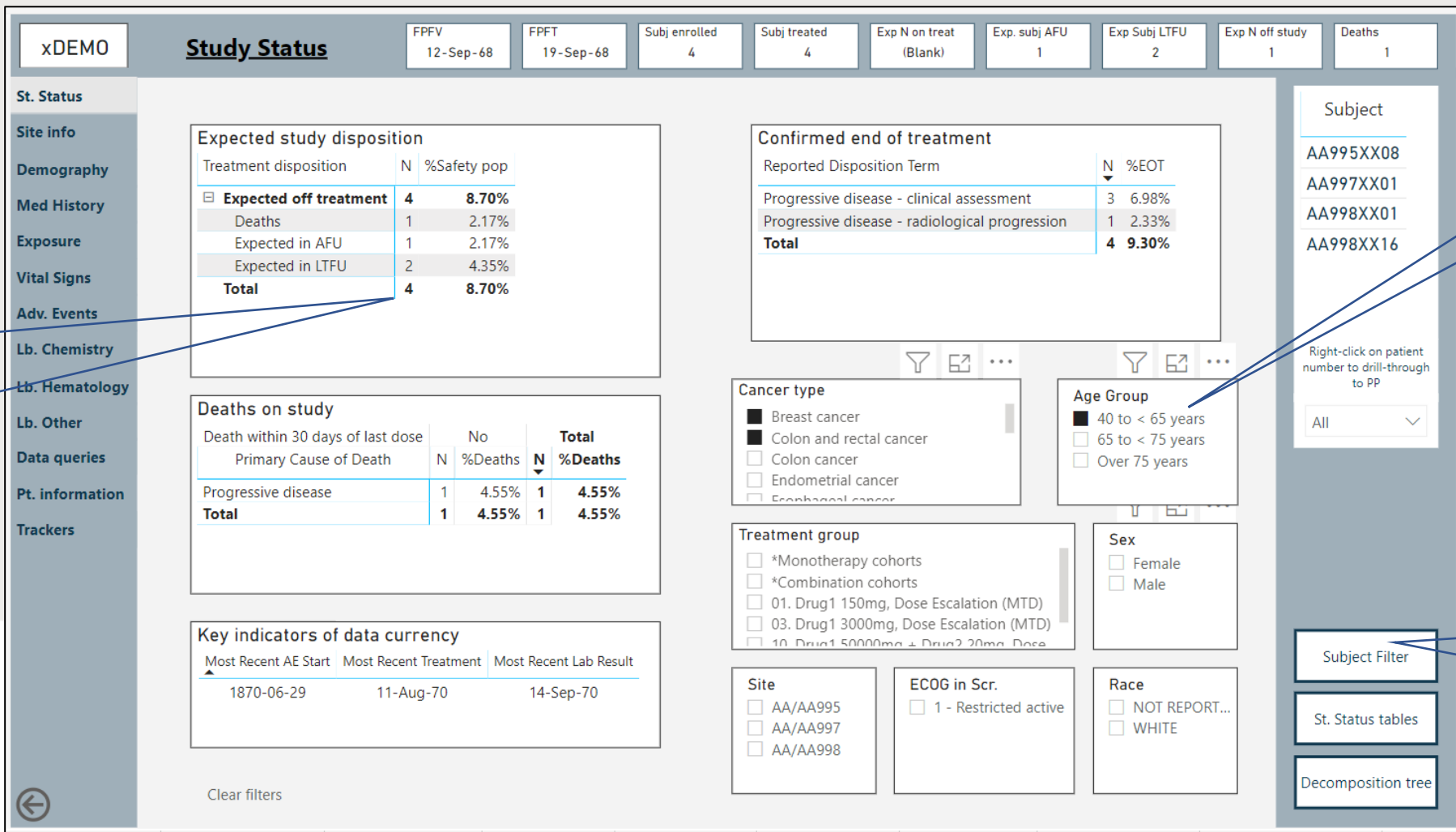
Different report pages



Interactive and expandable Table visualizations



Filtering data using slicers



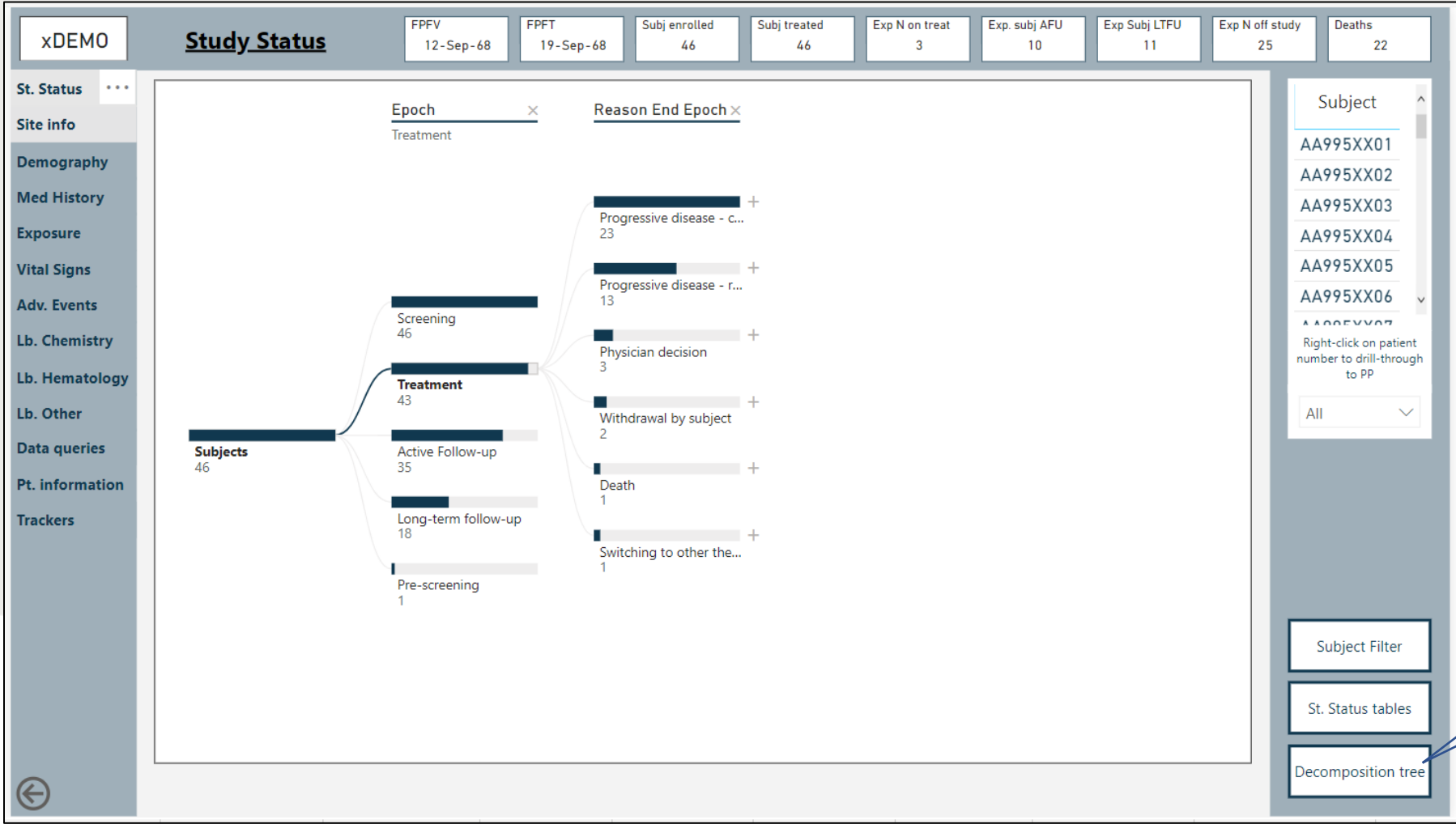
All the data tables and cards are automatically updated

Selected Filters

Switch to filter mode



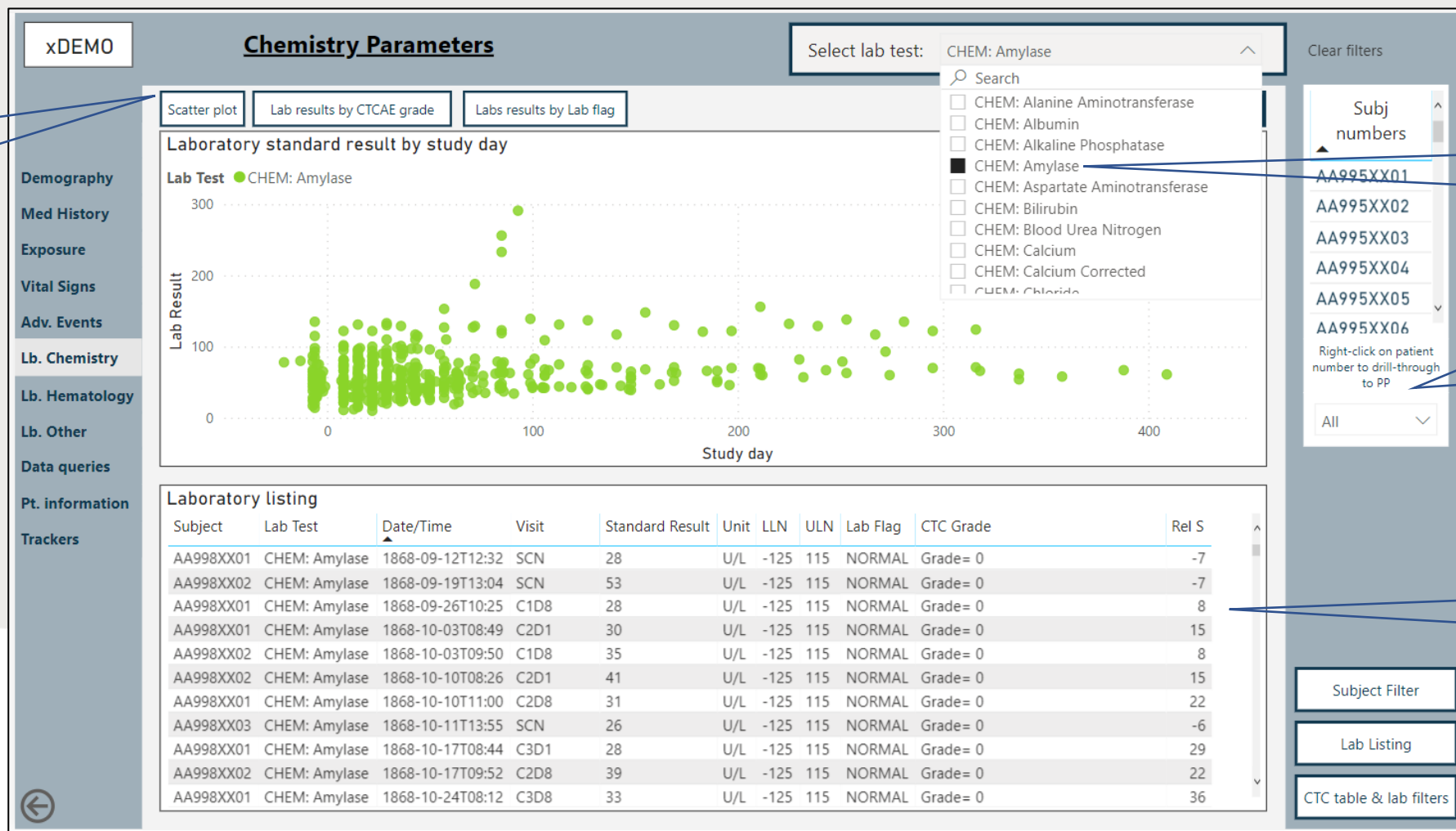
Decomposition tree visual



Decomposition tree



Lab Data Visuals



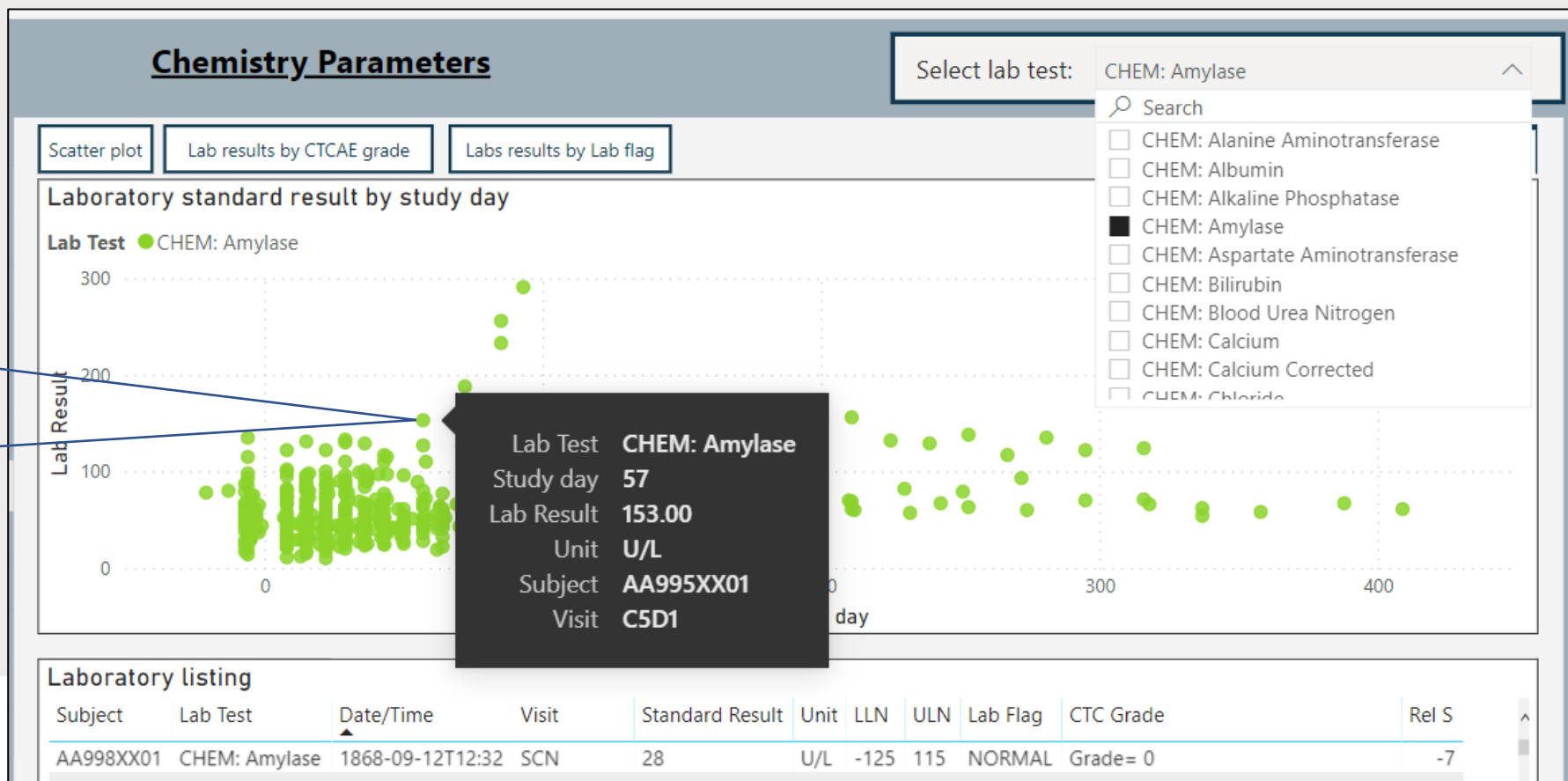
Select one, multiple or all tests

Drill-through to see complete patient profile

Subject listing



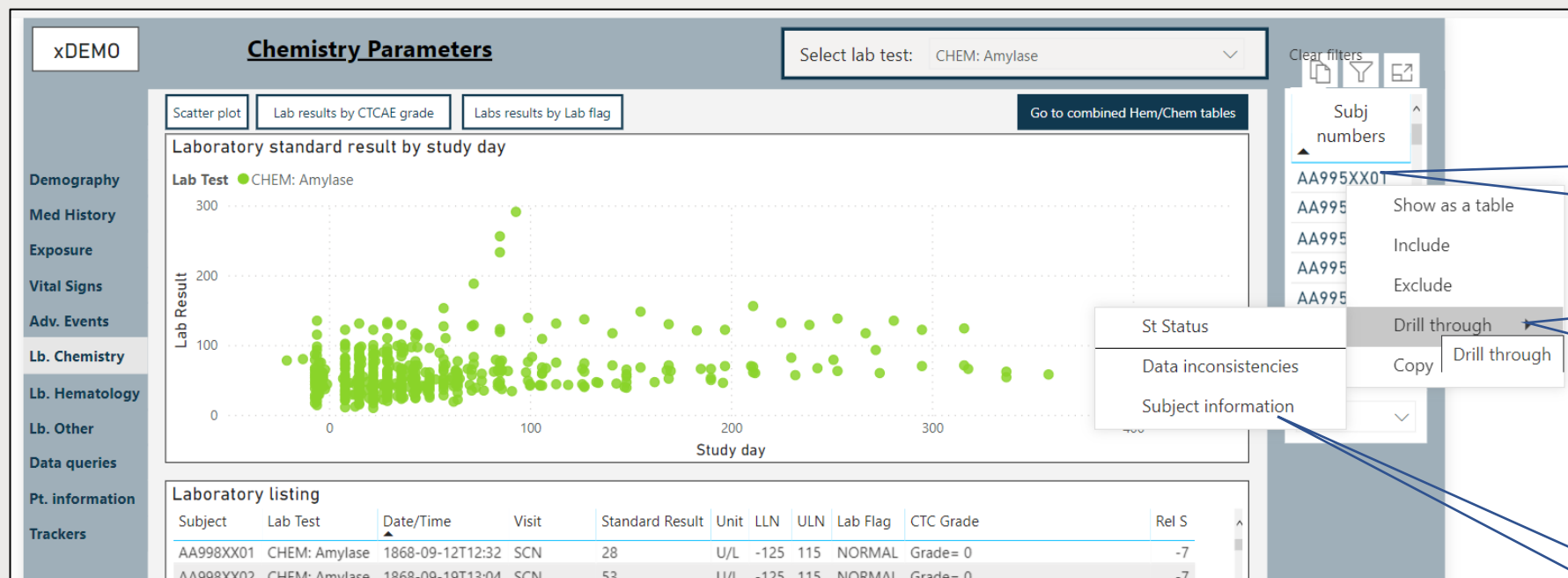
Identifying the subjects



Hover mouse pointer on individual points to see more details in the tooltip



Drill-through a subject



Right click on a subject

Select Drill through

Click on Subject Information



Subject Information (1/3)

←

Subject information

AA995XX01

Reset filters

Summary

25-Feb-69

1869-05-07

Expected off treatment

Progressive disease - radiological progression

1869-05-28

C1D1

Latest AE

Treatment disposition

EOT reason

EOT date

Demography

Drug1 3000mg, Dose Escalatio...

05-Feb-69

Female

WHITE

62 YEARS

Treatment Group

Initial Informed Consent Date

Sex

Race

Age at Enrollment

Cancer Classification

Head and neck cancer

Other

Carcinoma, NOS

(Blank)

Cancer Type

Location of primary cancer

Histology

Grading (AJCC) at initial diagnosis

1868-09-02

-176

Stage IV C

Most recent progression date

Study Day of Most recent progression

Stage at study entry

Disposition

E.C

Epoch

Reported Disposition Term

Rel. AE Terms

If Other

Date of E.

ode

Discontinuation

100

Screening

Completed

1869-02-18

300

Treatment

Progressive disease - radiological progression

1869-05-28



Subject Information (2/3)

Death information

1870-01-01

Date/Time

Progressive disease

Primary Cause of Death

(Blank)

If Other, specify

Screening Characteristics

Test	Standard Result	Standard Unit
Performance Status	1 - Restricted active	
Heart rate	75	beats/min
Respiration rate	16	breaths/min
Temperature	36.9	C
Weight	82.6	kg
Diastolic blood pressure	88	mmHg
Systolic blood pressure	152	mmHg

Medical History

MH No	Reported Term	Start Date/Time	End Date/Time	Ongoing?	CTCAE grade at baseline	CTCAE not gradable
011	Heart Murmur	1806		Ongoing	1	
008	History of Shingles	1859	1859			Yes
006	Obesity	1865		Ongoing	1	
007	Low back pain	1865		Ongoing	2	
014	Facial Edema	1865-03		Ongoing	2	
013	Left facial pain	1865-03-11		Ongoing		Yes
003	Acquired hypothyroidism	1865-04		Ongoing	2	
004	Arthritis	1865-04		Ongoing	2	
005	Hyperlipidemia	1865-04		Ongoing	2	

Exposure

Drug1 3000mg. Dose Escalation (MTD)

Visit	SD Taken?	Actual Treatment	Dose	Regimen	Route	Start Date	Start Time	End Date	End Time	D. modification?	Reason	Rel. AE 1	Rel. AE 2	Rel. AE 3	Reason, Oth
Cycle 01 Day 01	Yes	Drugotinib 1	3000	Once every 2 weeks	Intravenous	25-Feb-69	11:30	25-Feb-69	12:44	None					
Cycle 02 Day 01	Yes	Drugotinib 1	3000	Once every 2 weeks	Intravenous	11-Mar-69	12:45	11-Mar-69	13:45	None					
Cycle 03 Day 01	Yes	Drugotinib 1	3000	Once every 2 weeks	Intravenous	25-Mar-69	12:47	25-Mar-69	13:55	None					
Cycle 04 Day 01	Yes	Drugotinib 1	3000	Once every 2 weeks	Intravenous	08-Apr-69	11:36	08-Apr-69	12:36	None					
Cycle 05 Day 01	Yes	Drugotinib 1	3000	Once every 2 weeks	Intravenous	22-Apr-69	11:21	22-Apr-69	12:28	None					

Subject Information (3/3)



Adverse events

AE No	Reported Term	MedDRA PT	TEAE?	Start Date/Time	End Date/Time	Rel S/E	Rel to SD?	Rel
1	Right Shoulder Pain	Arthralgia	N	1869-02-18		-7/-91	No	
2	decreased Lymphocyte count	Lymphocyte count decreased	Y	1869-03-11		15/-70	No	
3	Intermittent left eye neuropathic pain	Neuralgia	Y	1869-03-19		23/-62	No	
4	Discomfort to the left posterior roof of mouth	Oral discomfort	Y	1869-03-25		29/-56	No	
5	Left cheek erythema close to left buccal wound	Erythema	Y	1869-03-26		30/-55	No	
7	Sore throat	Oropharyngeal pain	Y	1869-03-29	1869-05-07	33/-52	No	
10	Left cheek swelling	Swelling face	Y	1869-04-07		42/-43	No	
11	Left cheek redness	Erythema	Y	1869-04-07		42/-43	No	
8	Left facial pus drainage	Purulent discharge	Y	1869-04-22	1869-05-07	57/-28	No	
9	Cancer related pain	Cancer pain	Y	1869-04-22		57/-28	No	
12	Bilateral Leg Edema	Oedema peripheral	Y	1869-05-07		72/-13	No	

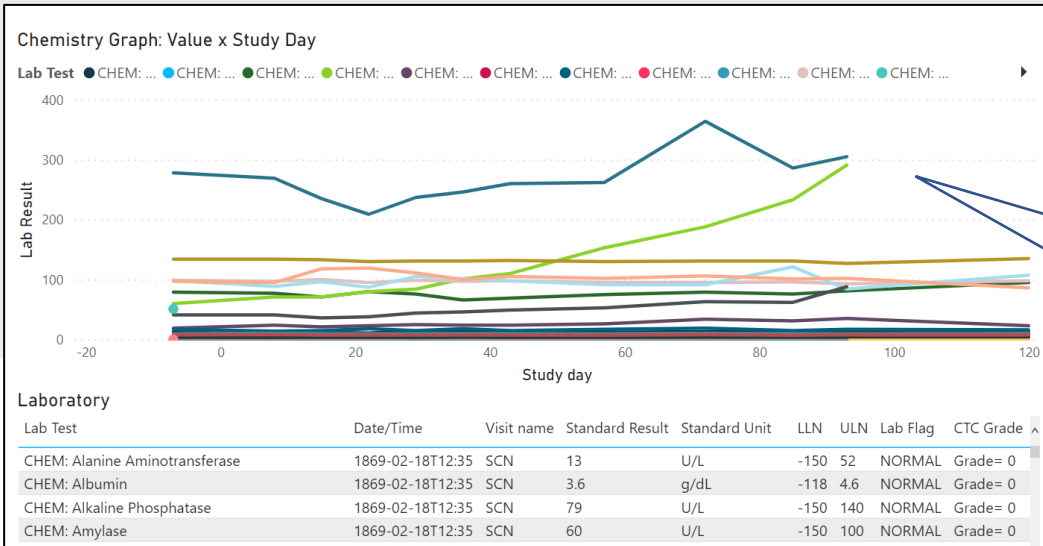
Date/Time	Visit Name	Test	Standard Result	Standard Unit	Systolic Blood Pressure	Diastolic Blood Pressure	Heart Rate	Respiration Rate	Temperature	Oxygen Saturation
1869-02-18	SCN	Performance Status	1 - Restricted active		93	160				
1869-02-18T11:21	SCN	Diastolic blood pressure	88	mmHg	Min	Max				
1869-02-18T11:21	SCN	Heart rate	75	beats/min						
1869-02-18T11:21	SCN	Respiration rate	16	breaths/min						
1869-02-18T11:21	SCN	Systolic blood pressure	152	mmHg	Min	Max				
1869-02-18T11:21	SCN	Temperature	36.9	C						
1869-02-18T11:21	SCN	Weight	82.6	kg						
1869-02-25	C1D1	Performance Status	1 - Restricted active							
1869-02-25T11:23	C1D1	Diastolic blood pressure	78	mmHg						
1869-02-25T11:23	C1D1	Heart rate	74	beats/min						
1869-02-25T11:23	C1D1	Oxygen saturation	97	%						
1869-02-25T11:23	C1D1	Respiration rate	16	breaths/min						
1869-02-25T11:23	C1D1	Systolic blood pressure	148	mmHg						
1869-02-25T11:23	C1D1	Temperature	36.9	C						
1869-02-25T11:23	C1D1	Weight	81.6	kg						

Vital Signs

ECG Findings and Cardiac Function

Date/Time	Visit Name	Result/Finding	If Abnormal, Clinically Significant?	Link to Rave
1869-02-25T10:22	C1D1	NORMAL OR NORMAL VARIANT		www.google.com
1869-02-25T10:24	C1D1	NORMAL OR NORMAL VARIANT		www.google.com
1869-02-25T10:26	C1D1	NORMAL OR NORMAL VARIANT		www.google.com
1869-02-25T13:09	C1D1	NORMAL OR NORMAL VARIANT		www.google.com
1869-02-25T13:11	C1D1	NORMAL OR NORMAL VARIANT		www.google.com

Date	Visit name	Method	Results	Unit	Interpretation	If Abnormal, are findings clinically significant?
1869-02-18	SCN	Echocardiography	65.5	%	Normal or normal variant	
1869-04-01	C4D1	Echocardiography	70.4	%	Normal or normal variant	
1869-05-17	EOT	Echocardiography	65	%	Normal or normal variant	



Chemistry Graph

And a lot more



Conclusion

- Quickly visualize clinical trials data
- Draw quick insights
- Create Data Diagnostics System

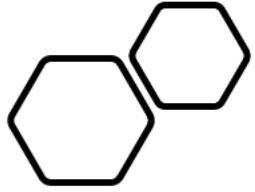


- Create data transformation pipelines
- Create highly customized visualizations
- Process, summarize and visualize large datasets
- Create web applications with dashboards
- An alternative to R Shiny apps
- jupyter notebook - an alternative to R markdown file
- Create Clinical-Trial-At-a-Glance report after database lock in jupyter notebook

References



- Most popular programming language video - <https://www.youtube.com/watch?v=ijjgPCDgILQ>
- Covid-19 data visualizations - https://ourworldindata.org/grapher/covid-19-death-rate-vs-population-density?country=USA~GBR~CHN~RUS~BGD~CAN~AUS~HKG~HTI~OWID_WRL~IND~MCO~SGP~NZL~PAK~COG~NGA~MNG~VEN
- Power BI desktop download link - <https://aka.ms/pbidesktopstore>
- Microsoft documentation on Power BI - <https://powerbi.microsoft.com/en-us/learning/>
- Anaconda (python) download link - https://repo.anaconda.com/archive/Anaconda3-2021.05-Windows-x86_64.exe
- Plotly library for interactive graphs - <https://plotly.com/python/>
- PHUSE Open Data Repository - https://github.com/phuse-org/PODR/tree/master/sample_code



Thank you!

Questions/comments?

