Alexander Lex http://vdl.sci.utah.edu

A Hodgepodge of Visualization Research: **Provenance, User Studies, Misinformation**



VISUAIZATION design lab

THE UNIVERSITY OF UTAH







visualization design lab





http://vdl.sci.utah.edu/



Visualization =



Human Data Interaction

not numbers. pictures

visualization The purpose of computing is insight,

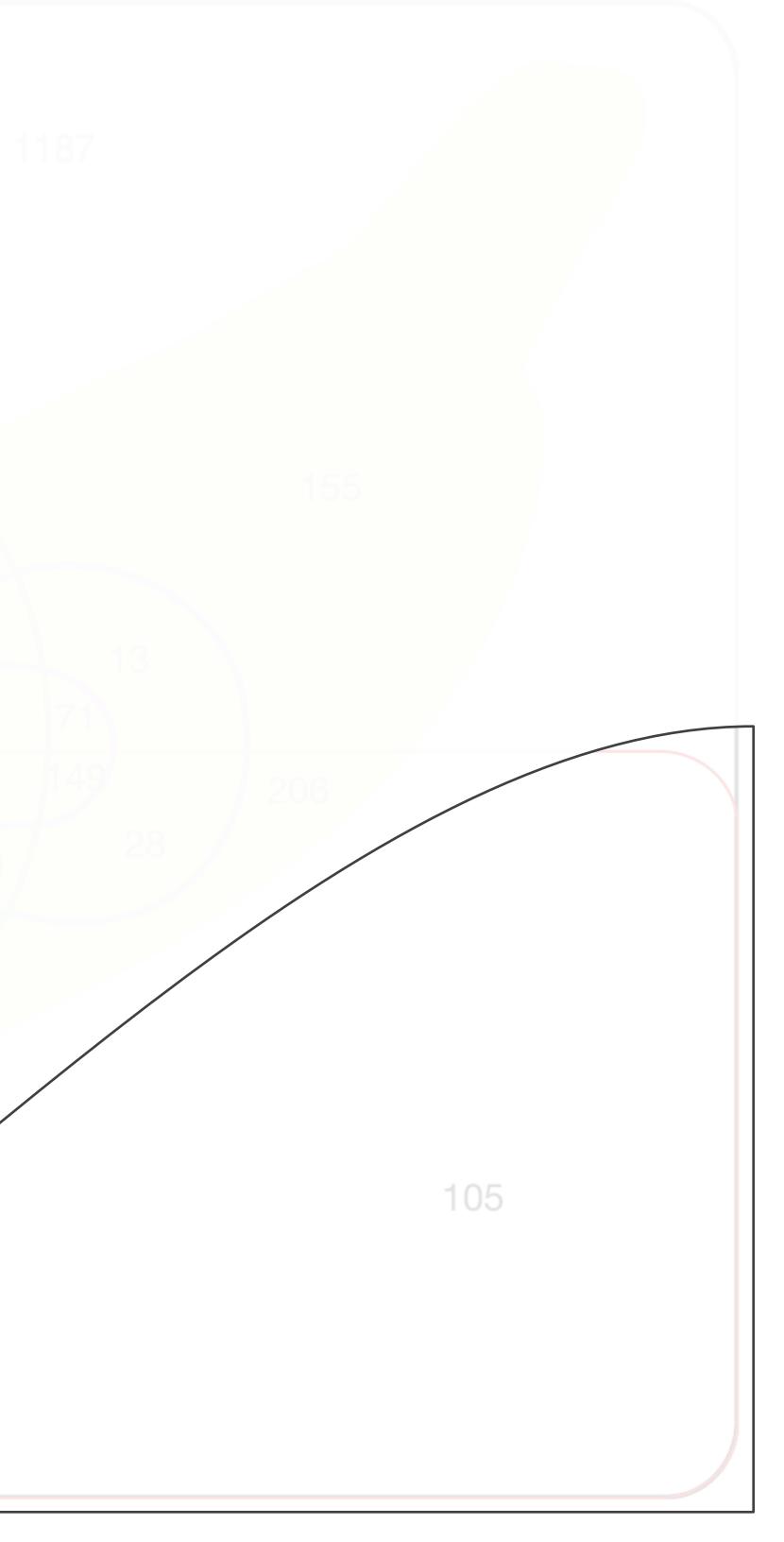
[Card, Mackinlay, Shneiderman] [Richard Wesley Hamming]



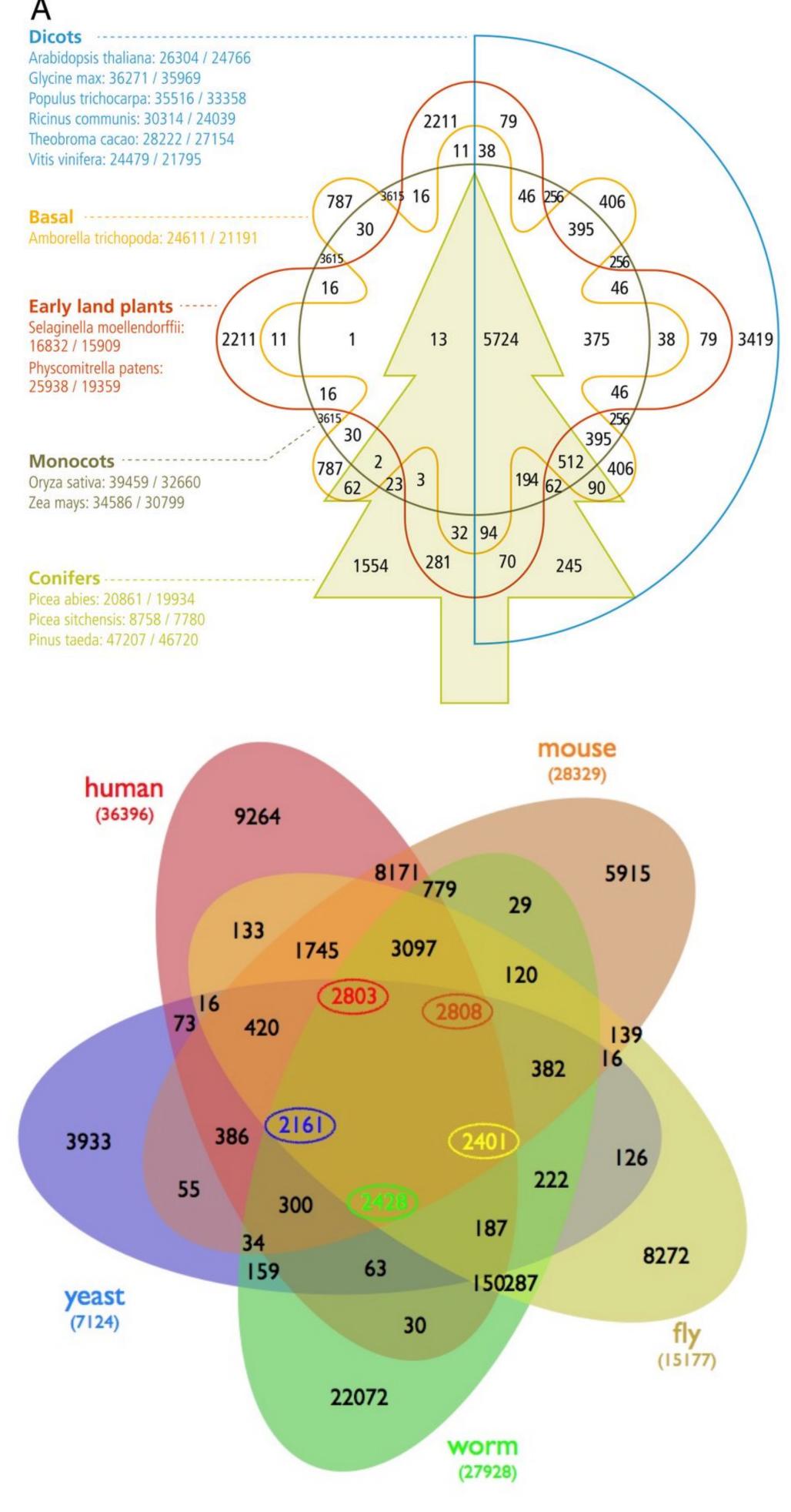
Banana M. acuminata Date P. dactylifera **Cress** Arabidopsis thaliana Rice Oryza sativa Sorghum Sorghum bicolor Brome Brachypodium distachyon



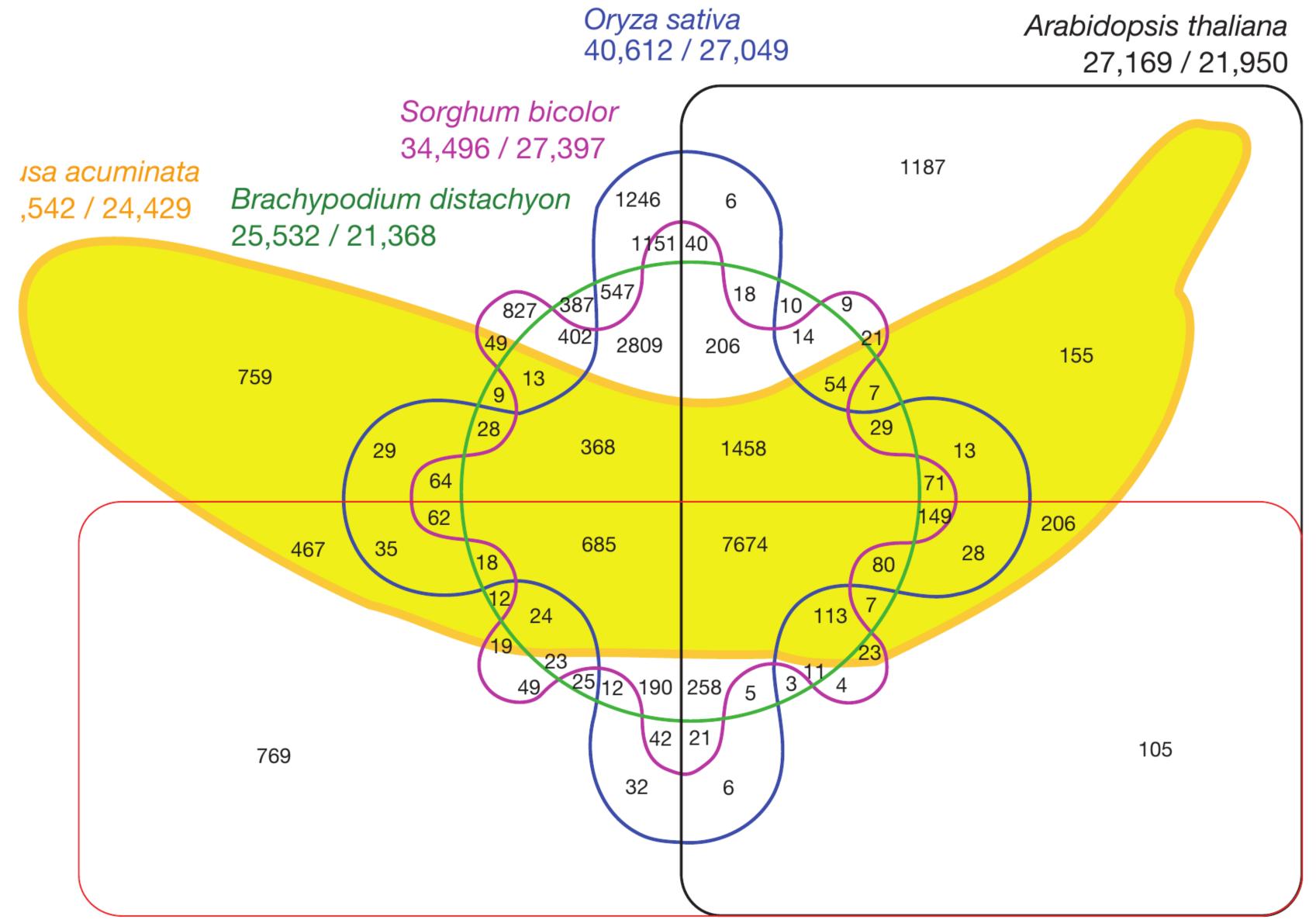
Brachypodium distachyon / 1246



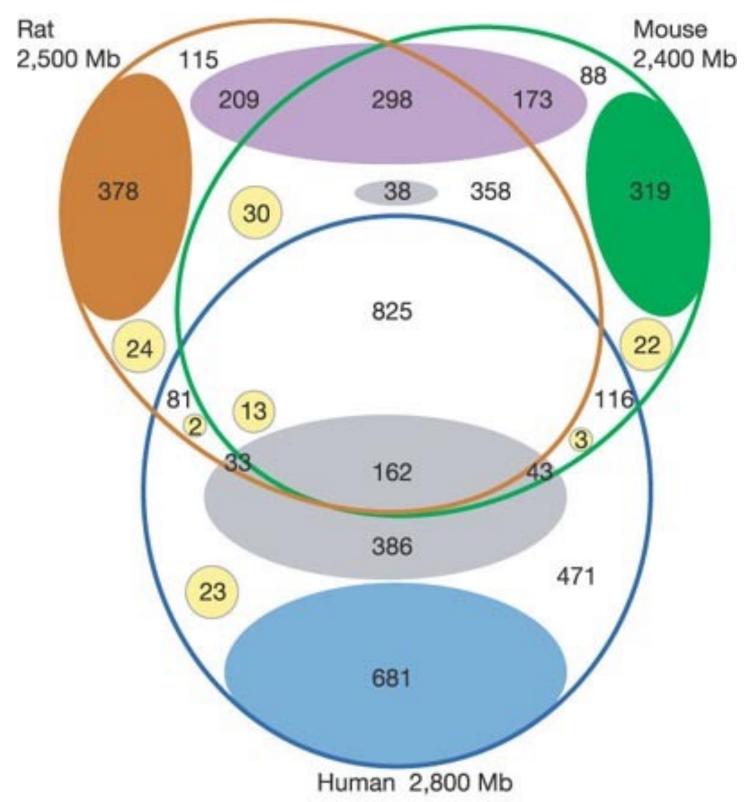
[Neale et al., BMC Genome Biology, 2014]

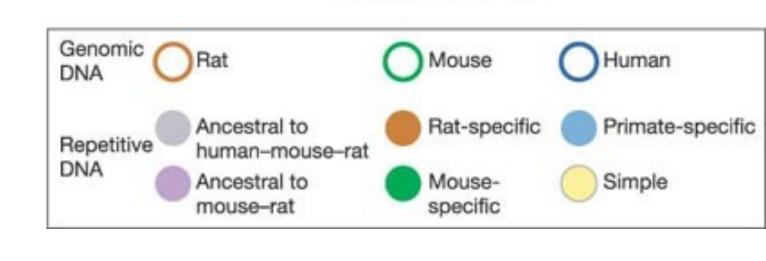


[Wiles et al., BMC Systems Biology]



Phoenix dactylifera 28,889 / 19,027





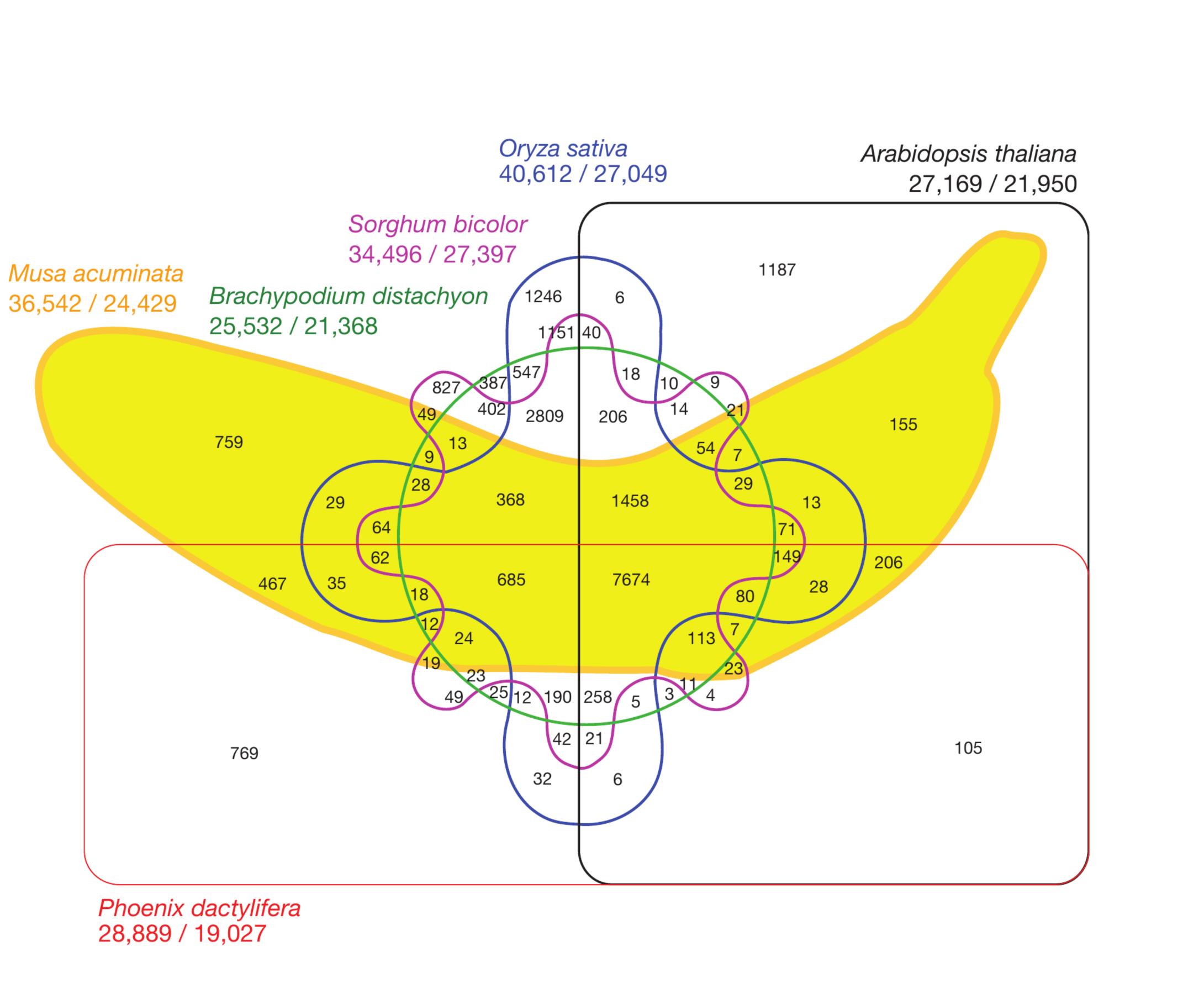
[Gibbs et al., Nature, 2004]

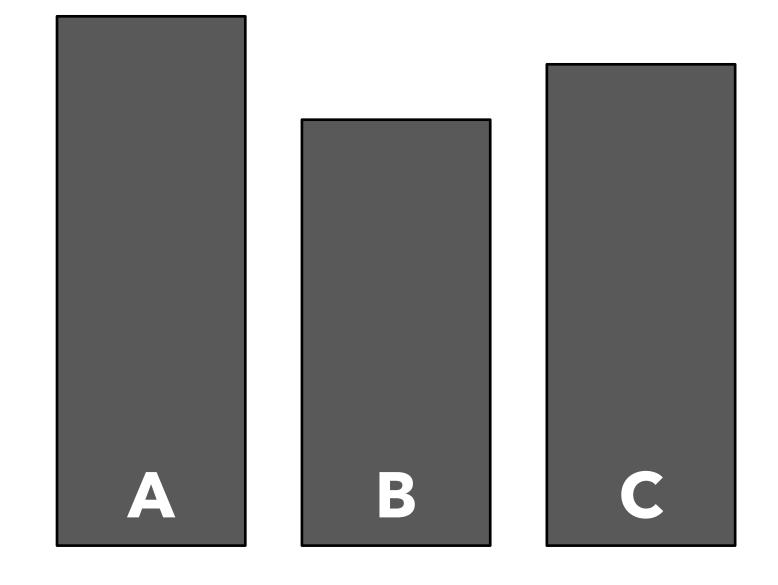
[D'Hont et al., Nature, 2012]



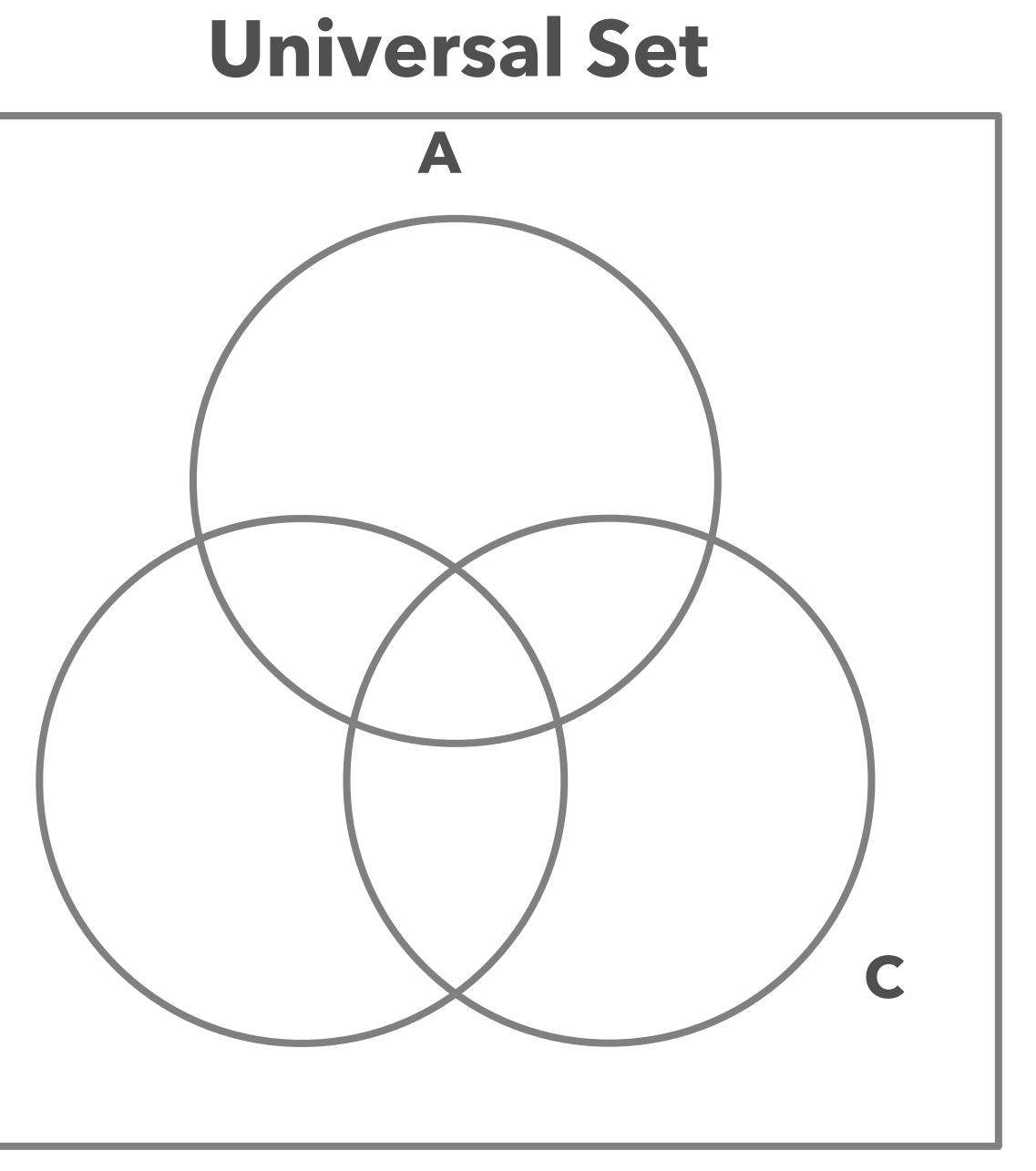
SO CAN WE DO BETTER?

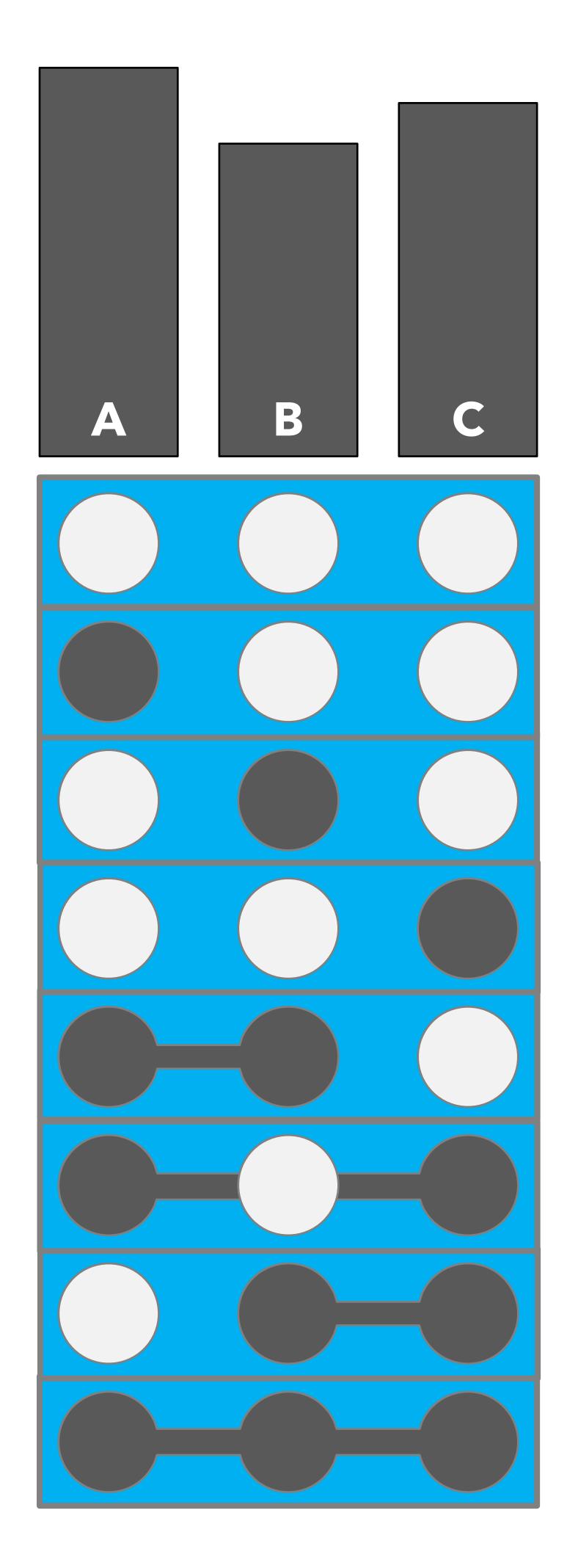






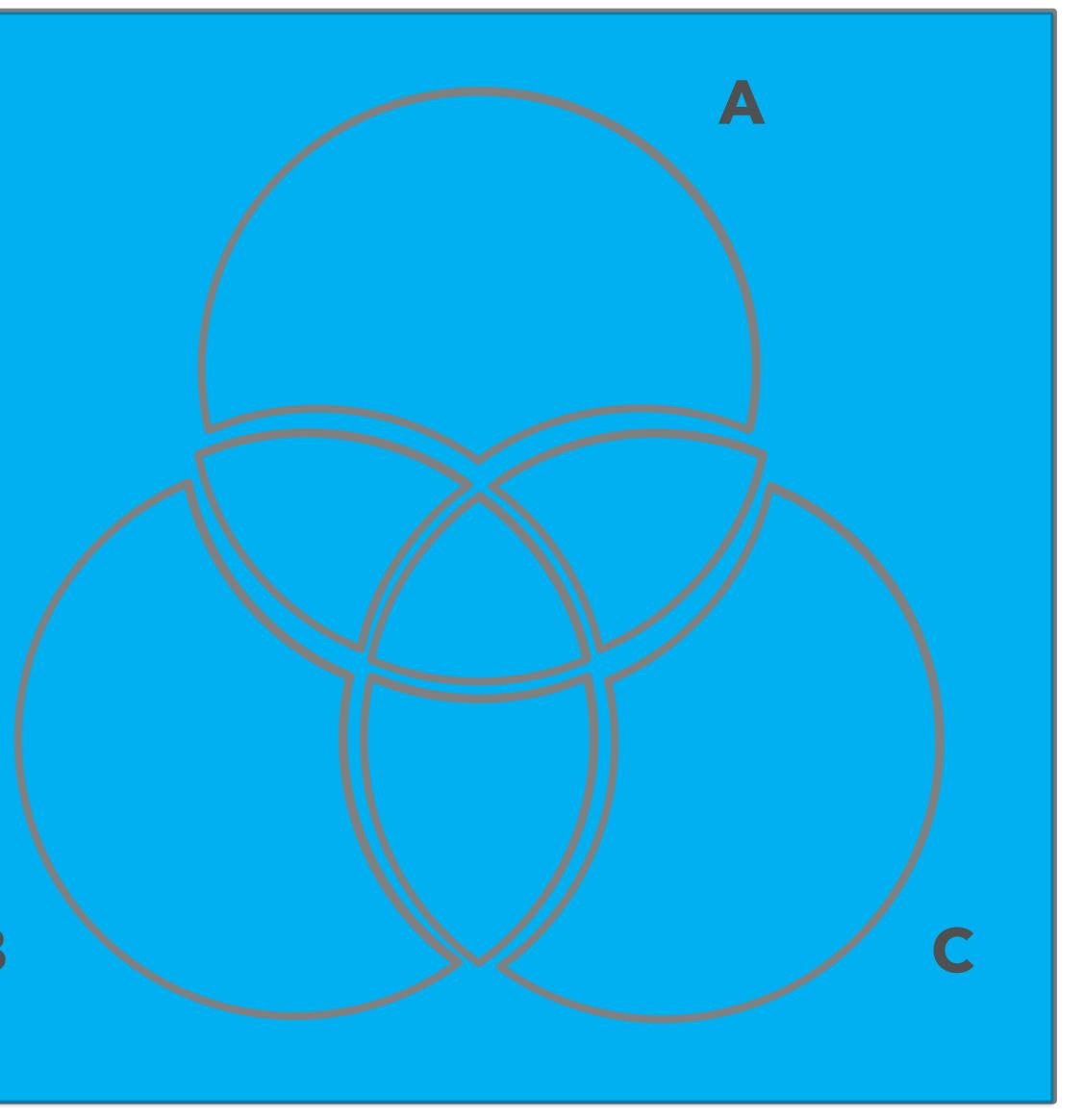
B

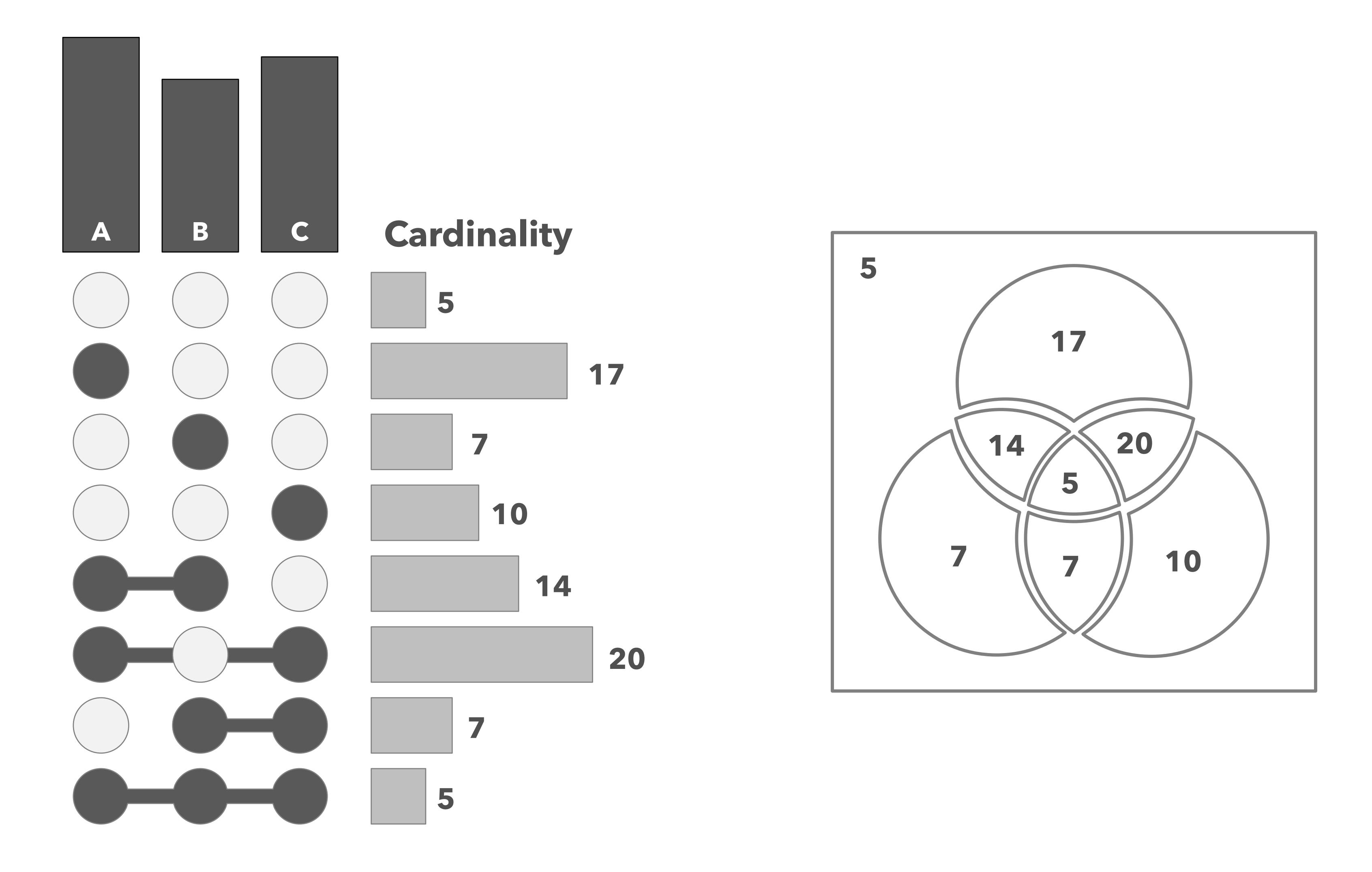






Universal Set

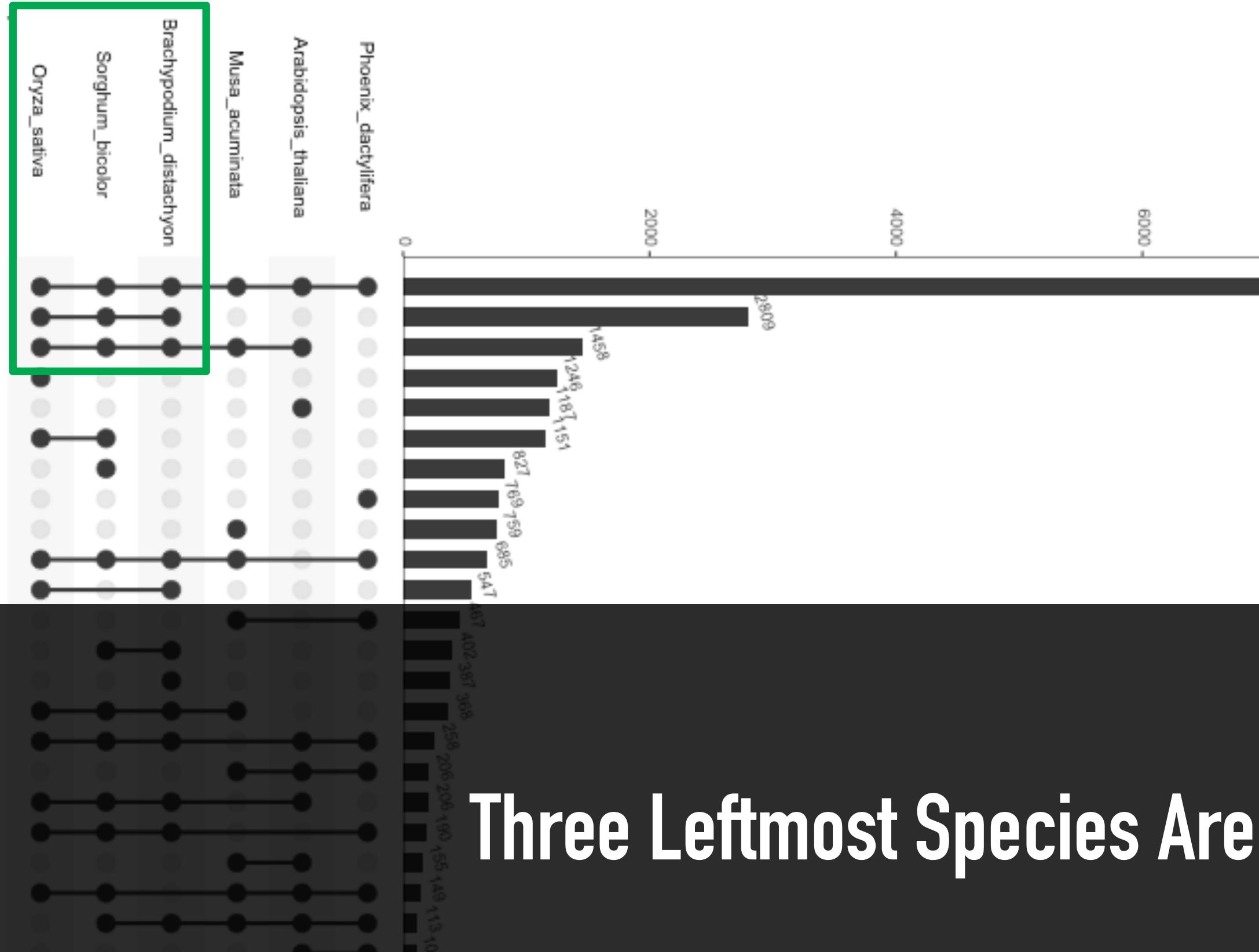




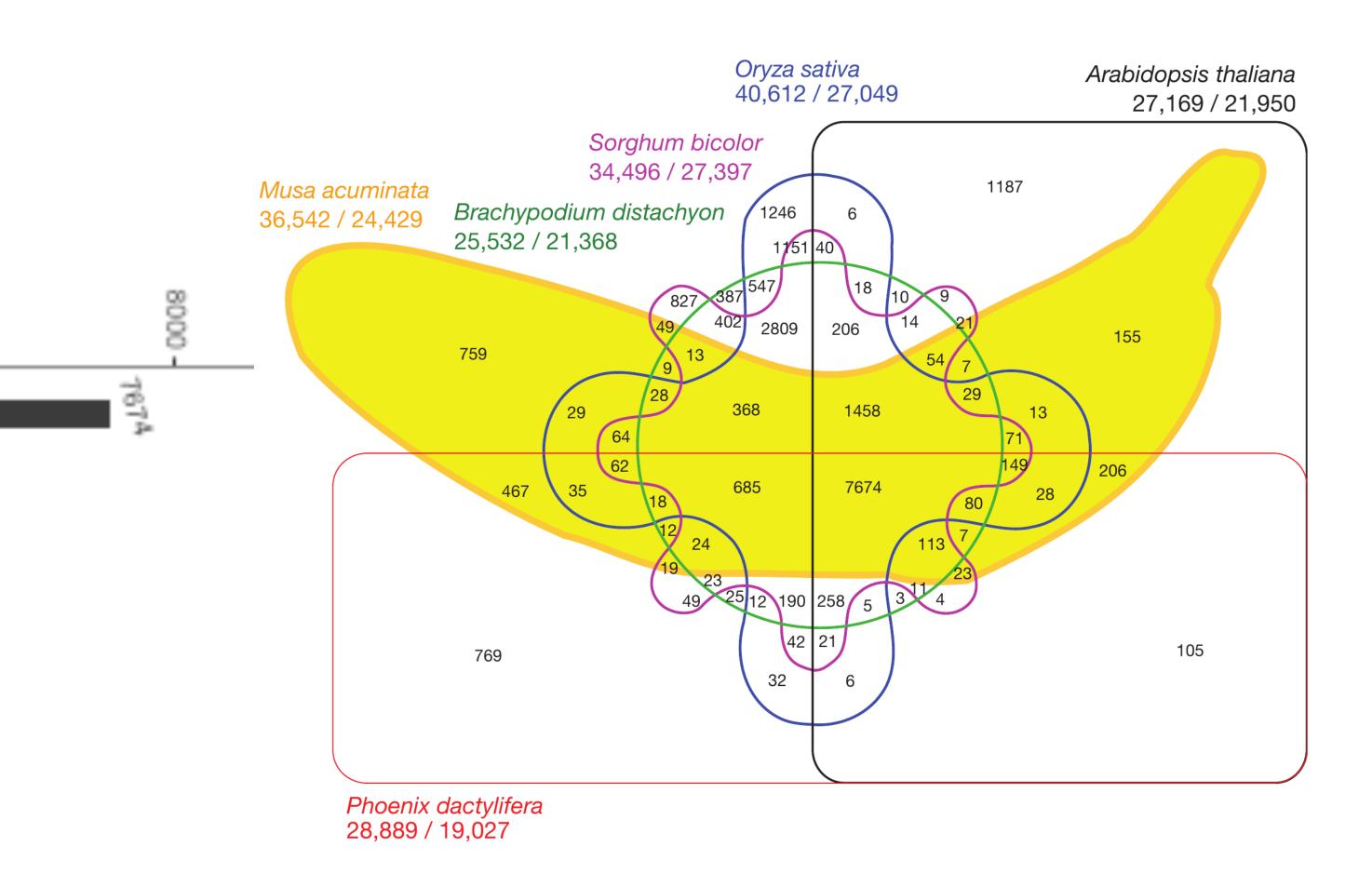
THE BANANA CHART REDESIGNED: UPSET



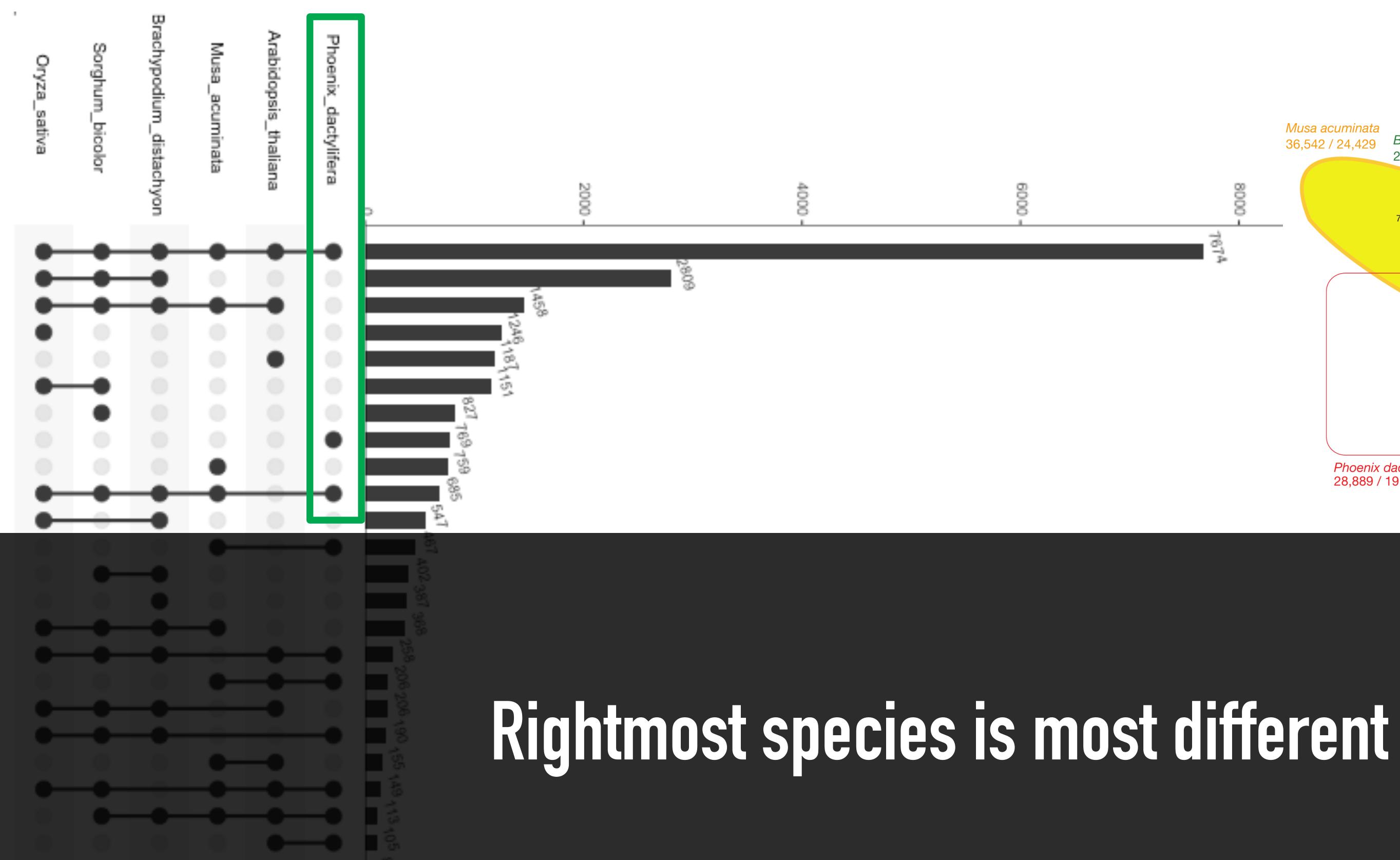
THE BANANA CHART REDESIGNED: UPSET

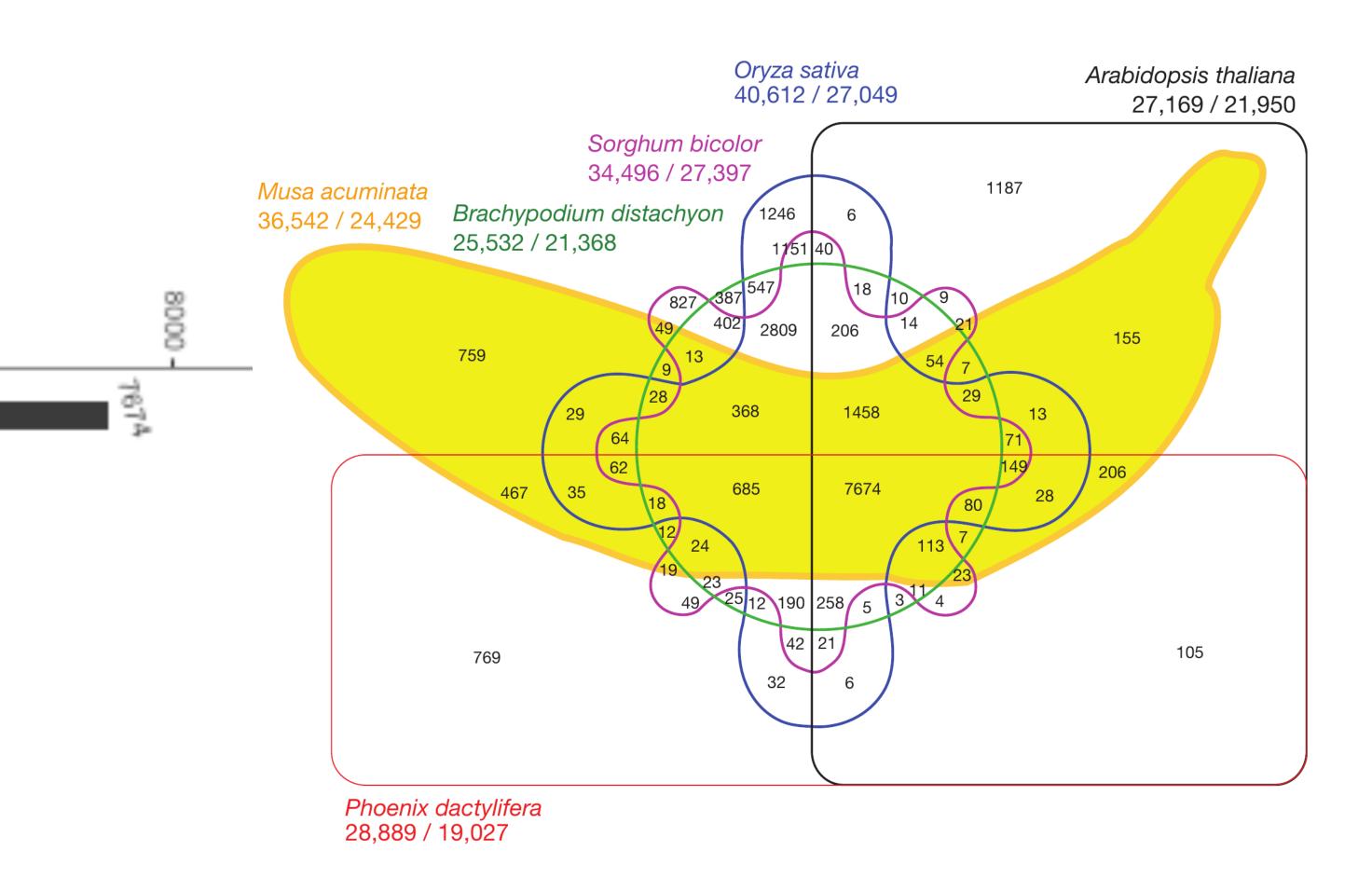


Three Leftmost Species Are Most Similar



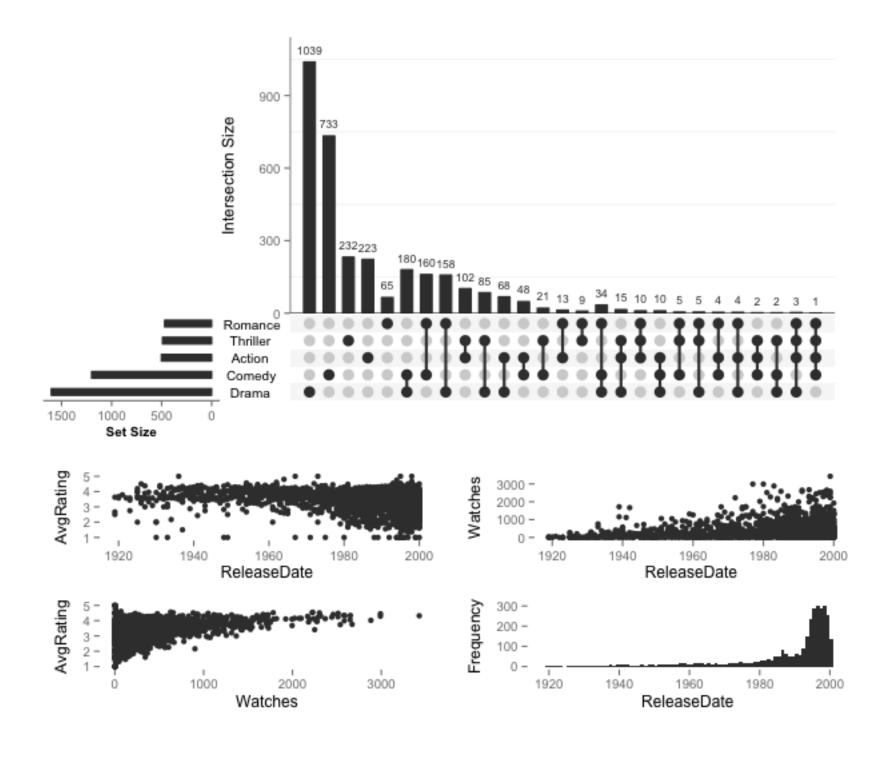
THE BANANA CHART REDESIGNED: UPSET



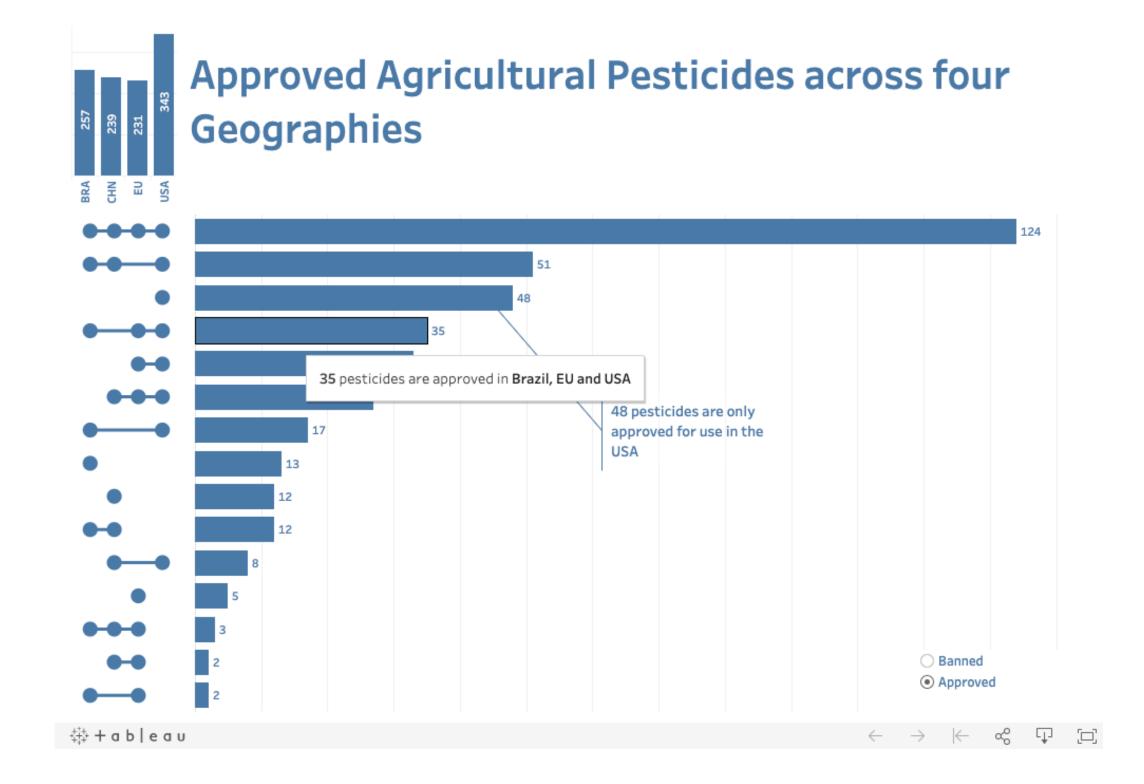




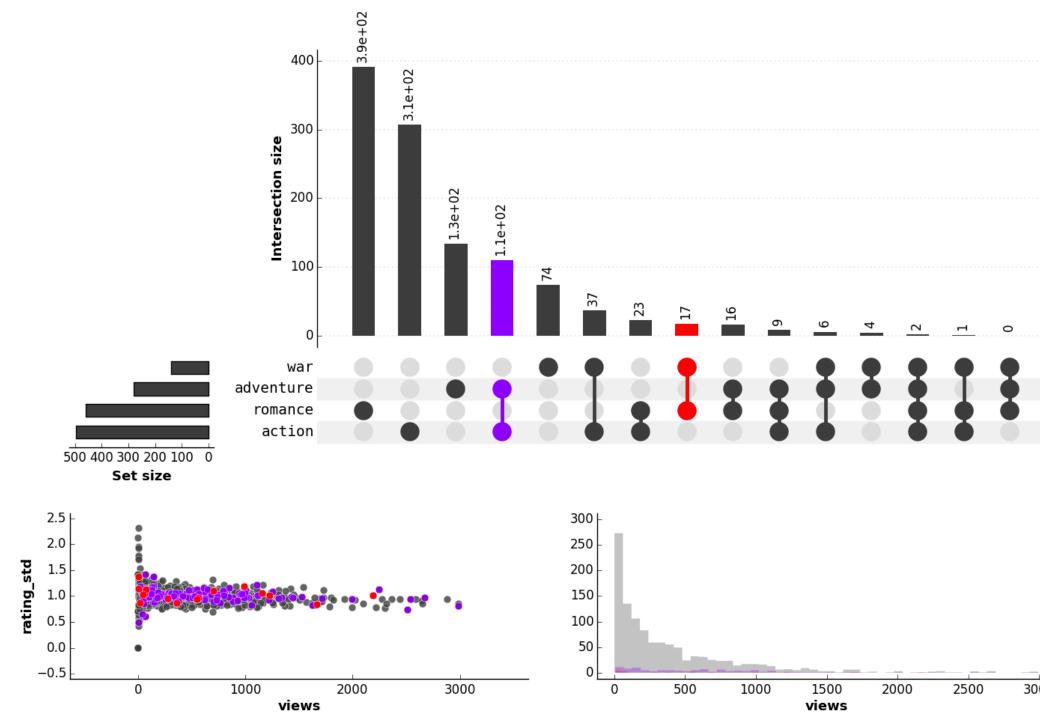
The canonical way to show set data with > 3 sets Second-most cited VIS paper of the last decade Available in various languages: <u>http://upset.app/</u>



R



Tableau

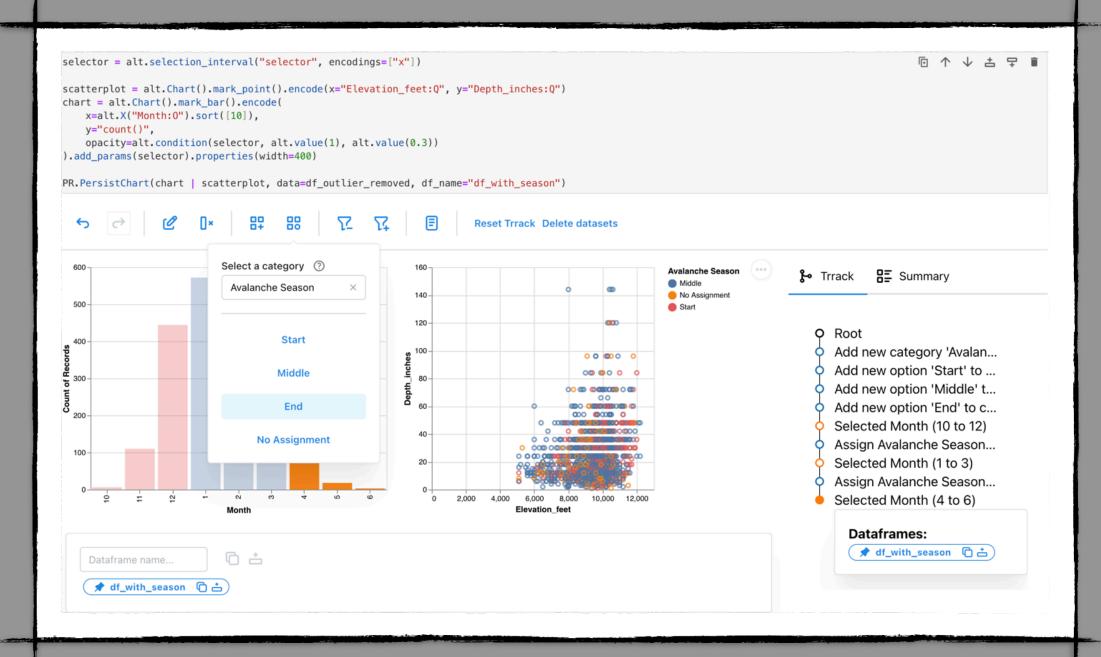


Python

000

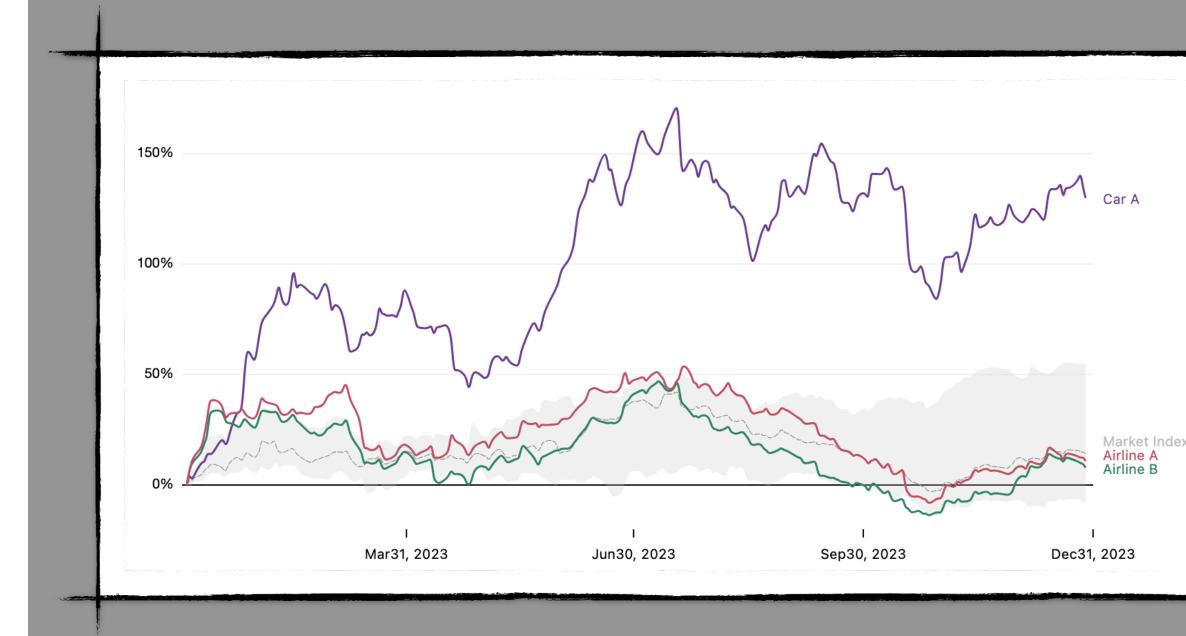
TECHNICAL CONTRIBUTIONS

Interaction in Notebooks



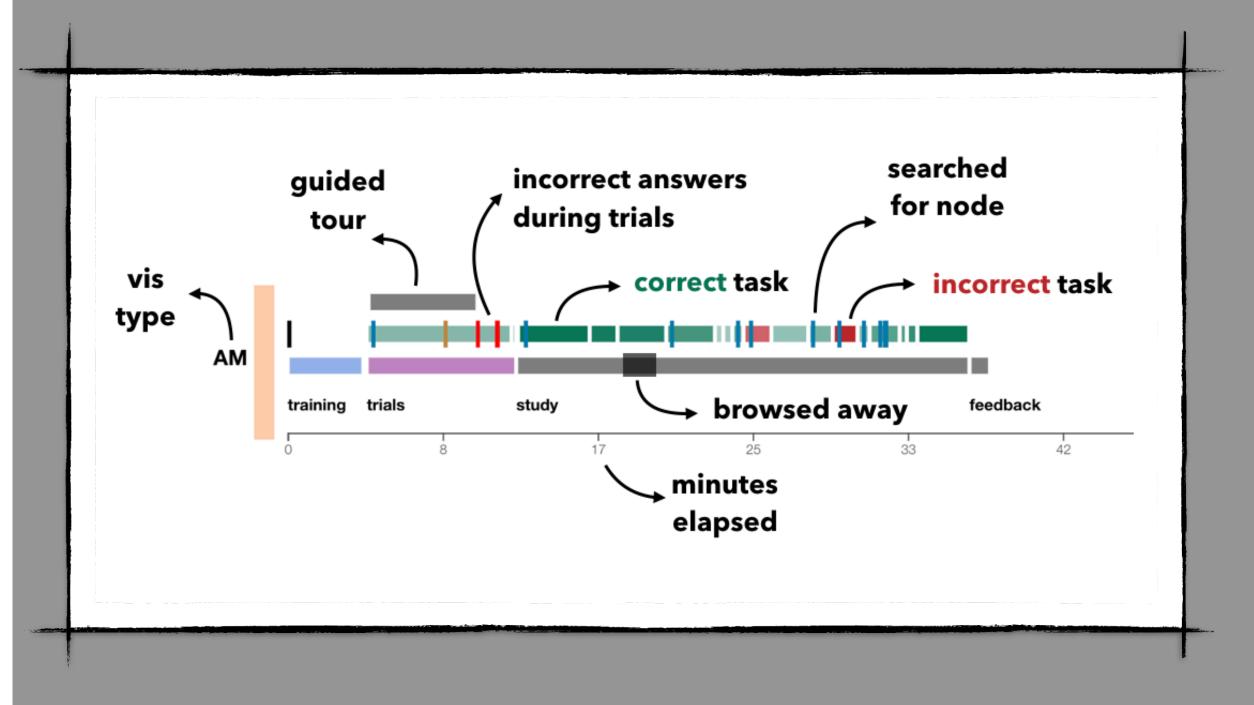
EMPIRICAL & THEORETICAL WORK

Visual Misinformation



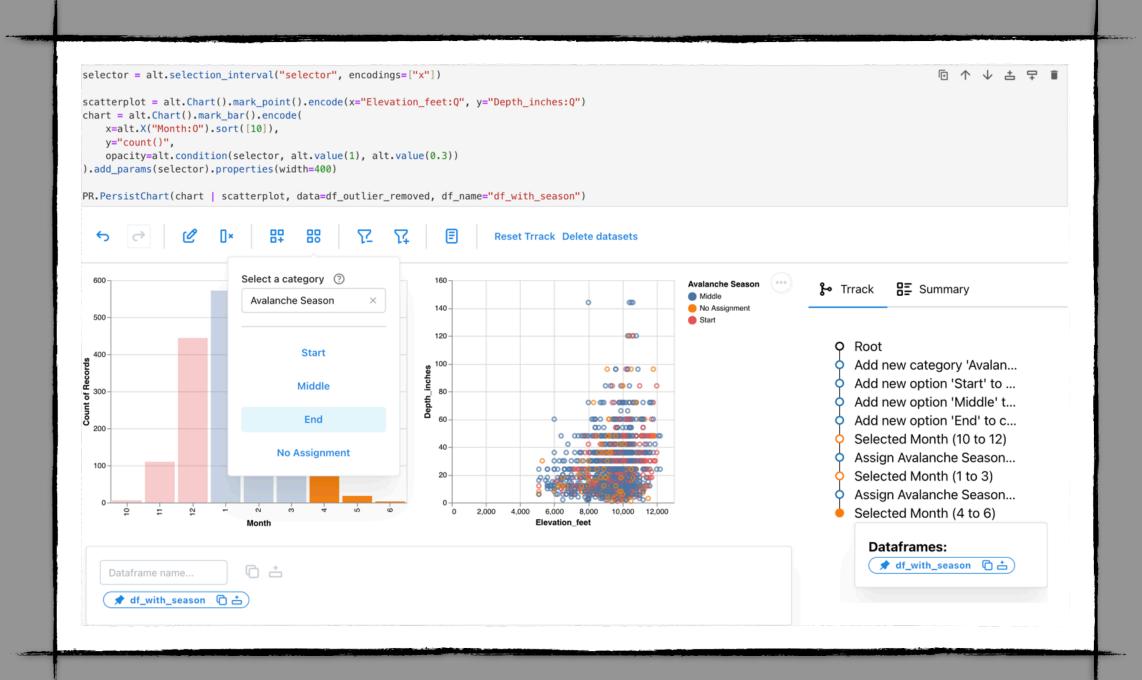


Evaluating Complex Systems





TECHNICAL CONTRIBUTIONS



PERSISTENT AND REUSABLE INTERACTIONS ... IN COMPUTATIONAL NOTEBOOKS



Kiran Gadhave, Zach Cutler







PERSISTENT AND REUSABLE INTERACTIONS IN **COMPUTATIONAL NOTEBOOKS**

€	₡ 0× 8₽ 8	₽ \7_ \7_ ■	Reset Trrack Delete da	itasets			
Q Search							Trrack 🗄 Summary
- #	Date ↑↓ ፡፡:	: Place ↑↓ ፡፡፡	Trigger	weak Layer	s ↑↓ ፡፡፡ : ;Aspect ↑↓	Elevation_feet 斗 :	Root
2338	3-23-2023	Dry Creek	Natural	Change column 'Trigger' data type 🕨	West	8000	 Rename column ;Weak Lay Rename column ;Trigger to
955	1-19-2014	Whitney Basin	Snowmobiler	前 Drop column 'Trigger'	East	10500	Drop column Comment
1028	2-21-2014	Chalk Creek	Natural	🖉 Rename column 'Trigger'	Northeast	10600	O Updated column 'Depth_i O Updated column 'Depth_i
1024	2-17-2014	Upper Weber Canyon	Natural	Ξ↑ Sort by Trigger ascending	Northeast	10400	Updated column 'Depth_i Finished removing null values from Changed column 'Dept
998	2-12-2014	Upper Weber Canyon	Natural	Ξ↓ Sort by Trigger descending	Northeast	10400	• Sort (descending) by 'Dep
938	1-14-2014	Upper Weber Canyon	Explosive	-🛱 Pin to left	East	10300	O Drop column Coordinates
2 1299	1-26-2016	Currant Creek Peak	Snowmobiler	➡ Pin to right	Southwest	9500	<pre>selector = alt.selection_interval("selector", encodings=["x"])</pre>
1044	2-28-2014	Chalk Creek	Natural	🕅 Unpin	Northeast	10600	<pre>scatterplot = alt.Chart().mark_point().encode(x="Elevation_feet:Q", y="Depth_inches:Q") chart = alt.Chart().mark_bar().encode(</pre>
2348	3-30-2023	Bunnels	Natural	Reset column size	Northeast	10800	<pre>x=alt.X("Month:0").sort([10]), y="count()",</pre>
1977	4-6-2021	Blue Ice	Natural	X Hide Trigger column	Northeast	10400	<pre>opacity=alt.condition(selector, alt.value(1), alt.value(0.3))</pre>
3 of 2392 ro	w(s) selected			≣≣ Show all columns	ws per page 10	of 2,392 K X).add_params(selector).properties(width=400)
Dataframe nam		s_df □ ≛ I ₪	nates_df (미급 亿 🗐	final_df ☐ ➡ ☑ 册			PR.PersistChart(chart scatterplot, data=df_outlier_removed, df_name="df_with_season") ←
							600 Select a category ⑦ 160
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ttps://vdl.sci.utah.edu/persist/					200 100 100 100 100 100 100 100		
in	incta	all me	rcic	t ovt			0 + + + + + + + + + + + + + + + + + + +

pip install persist ext

Demo Notebook: <u>https://tinyurl.com/5db7nynn</u>

G 📥 Dataframe name. 🖈 df_with_season 🛛 🖨

E Reset Trrack Delete datasets Avalanche Season ⊱ Trrack 🛛 🗄 Summary Middle œ 😑 No Assignment 140 Start 120 00 **Q** Root Add new category 'Avalan... 000000 Add new option 'Start' to ... 000000 80 -Add new option 'Middle' t... 00 Add new option 'End' to c... 000 0000 0 Selected Month (10 to 12) Assign Avalanche Season... Selected Month (1 to 3) Assign Avalanche Season... Selected Month (4 to 6) 2,000 4.000 6,000 8,000 10,000 12,000 Elevation feet Dataframes: 🗭 df_with_season 🛱 📥



DATASET EXAMPLE: HISTORICAL AVALANCHES IN UTAH

Avalanches are a major hazard in Utah Utah Avalanche Center collects data about avalanches, including where it occurred (location, elevation), how it occurred, how big it was, etc.



WHAT IS THIS TALK ABOUT?

Supposed you're doing data analysis in Python

What's the pandas code... • ...to drop a column?

• ...to change the order of columns? • ...to change the label of a column?

Nothing here is hard, but it's annoying.







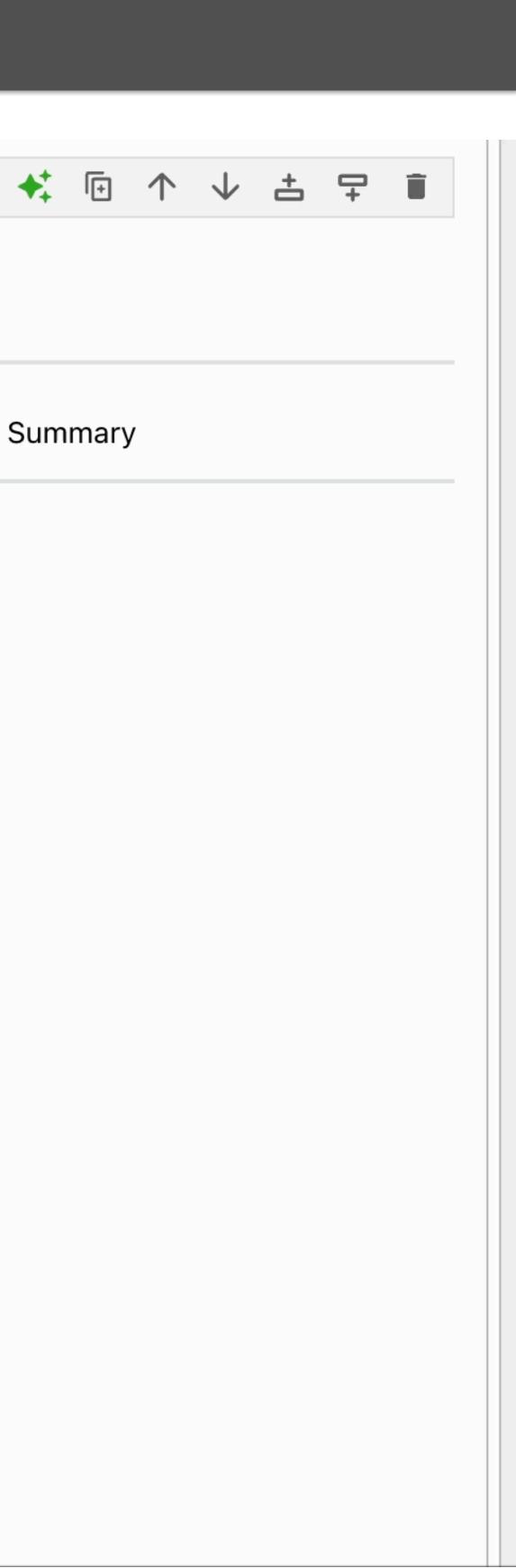


PERSIST MAKES THIS EASY

[4]: PR.PersistTable(avalanches, df_name="avalanches")

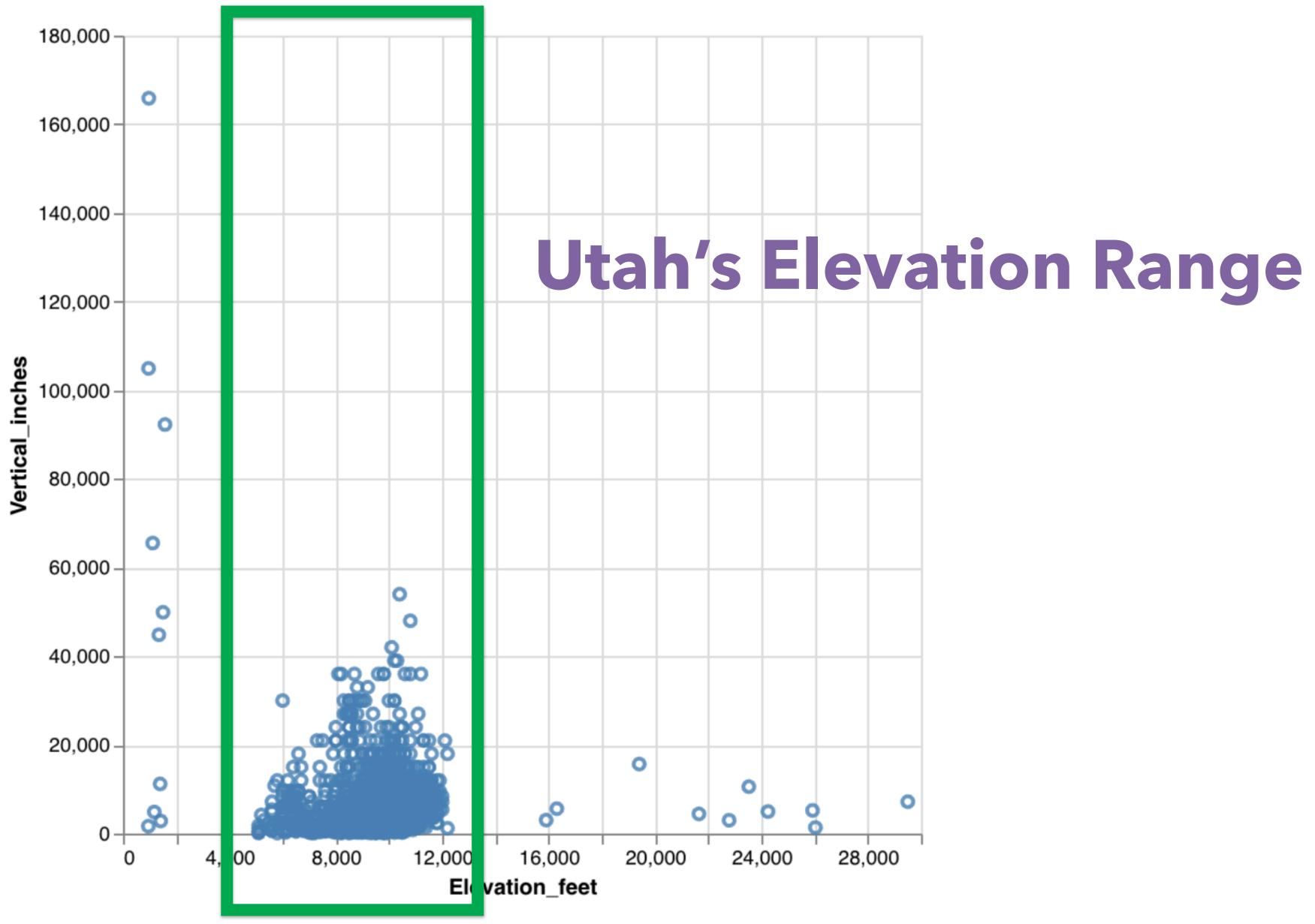
Search		•					မှိုး Trrack ခြံ Summary
#	;Region ↑↓ ፡፡፡	: Month	↑↓ ። : Day	↑↓ ። : Year	↑↓ ፡፡፡ : ;Trigger ↑↓ ፡፡፡	: ;Weak Layer ↑↓ ::	Root
) 1	Salt Lake	11	9	2012	Snowboarder	New Snow/Old Snow	
2	Salt Lake	11	11	2012	Skier	New Snow/Old Snow	
3	Salt Lake	11	11	2012	Skier	Facets	
4	Salt Lake	11	11	2012	Skier	New Snow	
5	Salt Lake	11	11	2012	Skier	Facets	
6	Salt Lake	11	10	2012	Skier	New Snow/Old Snow	
7	Salt Lake	11	12	2012	Skier	Facets	
8	Salt Lake	12	8	2012	Skier	Facets	
9	Salt Lake	12	9	2012	Skier	Facets	
10	Salt Lake	12	10	2012	Skier	Facets	
				Rows per pag	ge 10	2 < < > >	

Dataframe name	
🖈 avalanches 🗖 📩	



WHAT IS THIS TALK **ABOUT?**

Have you ever plotted something and wished you could just "fix" things as you spot them?

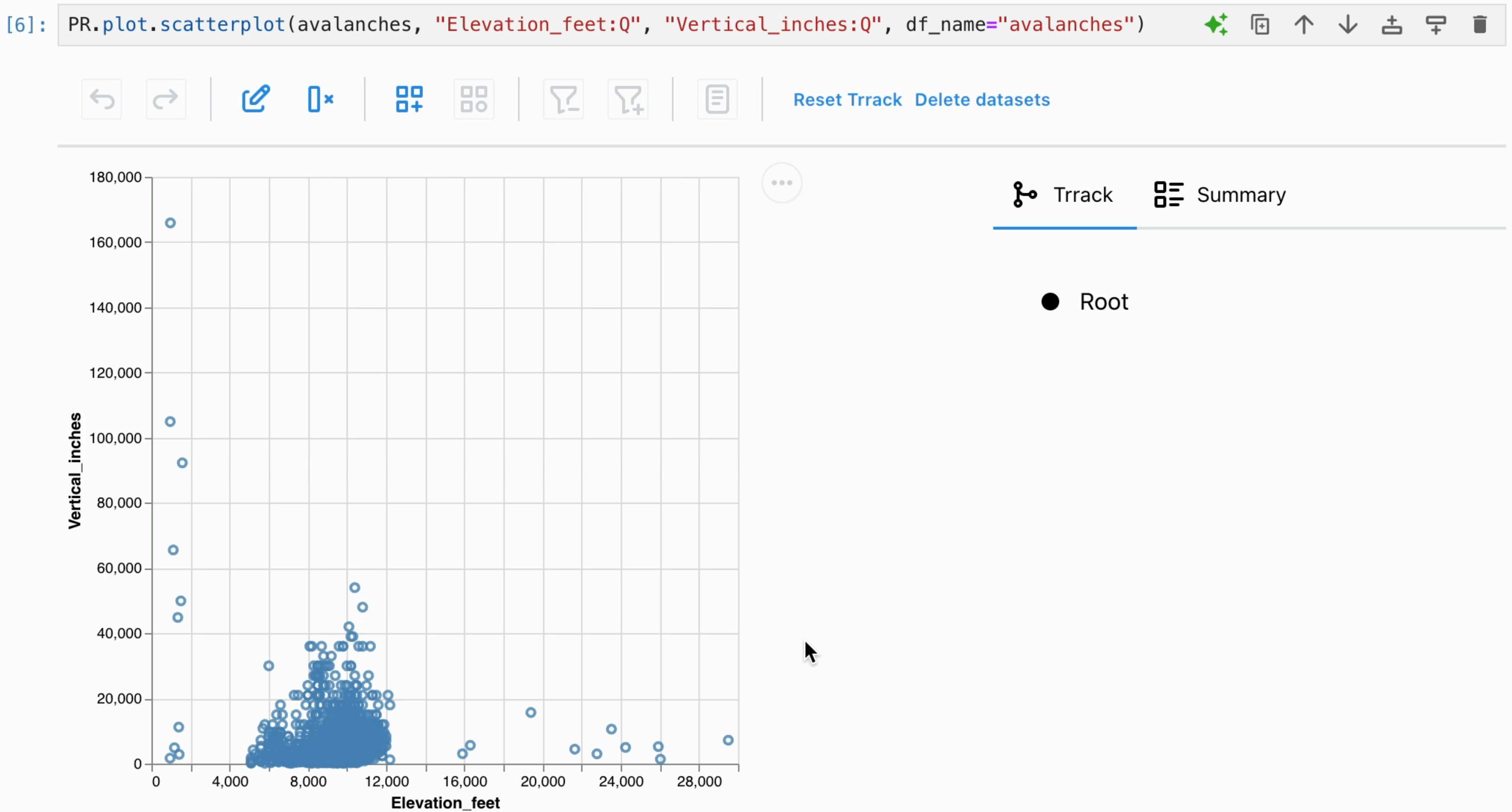


vas valanche th (big) dee

Elevation where the avalanche occurred



PERSIST MAKES THIS EASY





SO WHAT'S SPECIAL HERE?

Lots of vis tools support these operations Most data wrangling happens in code: it's just more powerful **Opportunity: bring interactive operations to code!**

Persist works **INSIDE your Jupyter Notebook**

BRIDGING BETWEEN DATA ANALYSIS MODALITIES

BRIDGING BETWEEN MODALTES

What are Modalities?

DATA ANALYSIS 1. Interactive Vis 2. Code

INTERACTIVE VISUALIZATION: BENEFITS

Intuitive Easy to use Uses human perceptual capabilities

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	1		Dmitri Poloz			Zenit St. Peters		W
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	3	_	Sebastián Driussi			Zenit St. Peters		CF
	4	_	Aleksandr Kokorin			Zenit St. Peters		CF
	5	_	Anton Zabolotnyi			Zenit St. Peters		CF
	6	_	Quincy Promes			Spartak Mosco		W
	7	_	Pedro Rocha			Spartak Mosco		W
	8	_	Lorenzo Melgarejo			Spartak Mosco		W
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	10		Luiz Adriano		RUS - Pre	Spartak Mosco		CF
	11		Zé Luís		RUS - Pre	Spartak Mosco		CF
	12		Ahmed Musa		RUS - Pre	CSKA Moscow		CF
	13		Fedor Chalov		RUS - Pre	CSKA Moscow		CF
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	21		lvan Ignatjev		RUS - Pre	FK Krasnodar		CF
	22		Alan Kasaev		RUS - Pre	Lokomotiv Mos		W
	23		Jefferson Farfán		RUS - Pre	Lokomotiv Mos		W
	24		Arshak Koryan		RUS - Pre	Lokomotiv Mos		W
	25		Éder		RUS - Pre	Lokomotiv Mos		CF
	26		Ari		RUS - Pre	Lokomotiv Mos		CF
	27		Gökdeniz Karadeniz		RUS - Pre	Rubin Kazan		W
	28		Rifat Zhemaletdinov		RUS - Pre	Rubin Kazan		W
	29		Sardar Azmoun		RUS - Pre	Rubin Kazan		CF
	30		Léo Jabá		RUS - Pre	Akhmat Grozny		W
	31		Bernard Berisha		RUS - Pre	Akhmat Grozny		W
	32		Magomed Mitrishev		RUS - Pre	Akhmat Grozny		W
	33		Odise Roshi		RUS - Pre	Akhmat Grozny		W
	34		Khalid Kadyrov		RUS - Pre	Akhmat Grozny		W
	35		Bekim Balaj		RUS - Pre	Akhmat Grozny		CF
	36		Ablaye Mbengue		RUS - Pre	Akhmat Grozny		CF
	37		Zaur Sadaev		RUS - Pre	Akhmat Groznv	I	CF



INTERACTIVE VISUALIZATION: DOWNSIDES

Limited Expressivity Some operations are difficult

e.g., conditional queries. Not reusable

need to redo analysis when data changes Not reproducible

Agg	Rank	Sele	$ extsf{T}$ Player	Current	op Current		• Po
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	1		Dmitri Poloz	RUS - Pre	Zenit St. Peters		W
	2	_	Emiliano Rigoni		Zenit St. Peters		W
	3	_	Sebastián Driussi		Zenit St. Peters		CF
	4	_	Aleksandr Kokorin		Zenit St. Peters		CF
	5	_	Anton Zabolotnyi		Zenit St. Peters		CF
	6	_	Quincy Promes		Spartak Moscov		W
	7	_	Pedro Rocha	_	Spartak Moscov		W
	8		Lorenzo Melgarejo		Spartak Moscov		W
	9		Zelimkhan Bakaev		Spartak 2 Mosc		W
	10		Luiz Adriano	_	Spartak Moscov		CF
	11		Zé Luís	=	Spartak Moscov		CF
	12	_	Ahmed Musa	_	CSKA Moscow		CF
	13	_	Fedor Chalov		CSKA Moscow		CF
	14	_	Timur Zhamaletdinov		CSKA Moscow		CF
	15		Wanderson	RUS - Pre	FK Krasnodar		W
	16		Joãozinho	RUS - Pre	FK Krasnodar		W
	17		Andrei Ivan	RUS - Pre	FK Krasnodar		W
	18		Ricardo Laborde	RUS - Pre	FK Krasnodar		W
	19		Magomed-Shapi Suleyr	RUS - Pre	FK Krasnodar		W
	20		Fedor Smolov	RUS - Pre	FK Krasnodar		CF
	21		lvan Ignatjev	RUS - Pre	FK Krasnodar		CF
	22		Alan Kasaev	RUS - Pre	Lokomotiv Mos		W
	23		Jefferson Farfán	RUS - Pre	Lokomotiv Mos		W
	24		Arshak Koryan	RUS - Pre	Lokomotiv Mos		W
	25		Éder	RUS - Pre	Lokomotiv Mos		CF
	26		Ari	RUS - Pre	Lokomotiv Mos		CF
	27		Gökdeniz Karadeniz	RUS - Pre	Rubin Kazan		W
	28		Rifat Zhemaletdinov	RUS - Pre	Rubin Kazan		W
	29		Sardar Azmoun	RUS - Pre	Rubin Kazan		CF
	30		Léo Jabá	RUS - Pre	Akhmat Grozny		W
	31		Bernard Berisha	RUS - Pre	Akhmat Grozny		W
	32		Magomed Mitrishev	RUS - Pre	Akhmat Grozny		W
	33		Odise Roshi	RUS - Pre	Akhmat Grozny		W
	34		Khalid Kadyrov	RUS - Pre	Akhmat Grozny		W
	35		Bekim Balaj	RUS - Pre	Akhmat Grozny		CF
	36		Ablaye Mbengue	RUS - Pre	Akhmat Grozny		CF
	37		Zaur Sadaev	RUS - Pre	Akhmat Groznv		CF

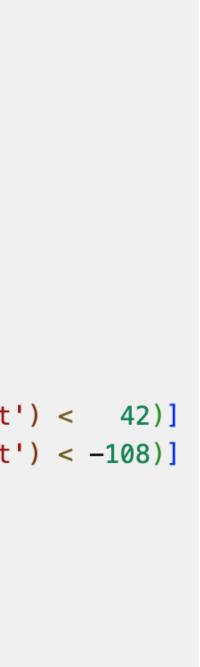


CODE BENEFIS

Flexible and powerful you basically can do anything Reusable if your data changes, re-run Reproducible everything is documented

1	# Keep
2	avy_df
3	
4	# Remov
5	avy_df
6	
7	<pre># Split</pre>
8	avy_df[
9	
10	# Remov
11	avy_df
12	avy_df
13	
14	# Keep
15	avy_df
16	avy_df.

```
this cell
= pd.read_csv('./avalanches.csv')
e NaN coordinates
= avy_df[avy_df['Coordinates']==avy_df['Coordinates']]
into latitude & longitude
['lat', 'lon']] = avy_df['Coordinates'].str.split(',', expand=True)
ve values outside of Utah bounds
= avy_df[ (36 < avy_df['lat'].astype('float')) & (avy_df['lat'].astype('float') < 42)]</pre>
= avy_df[(-114 < avy_df['lon'].astype('float')) & (avy_df['lon'].astype('float') < -108)]</pre>
columns we need
= avy_df[['Date', 'Region', 'Trigger', 'lat', 'lon']]
head (
```

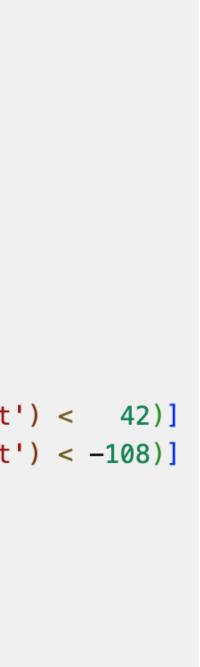


CODE / SCRIPTING: DOWNSIDES

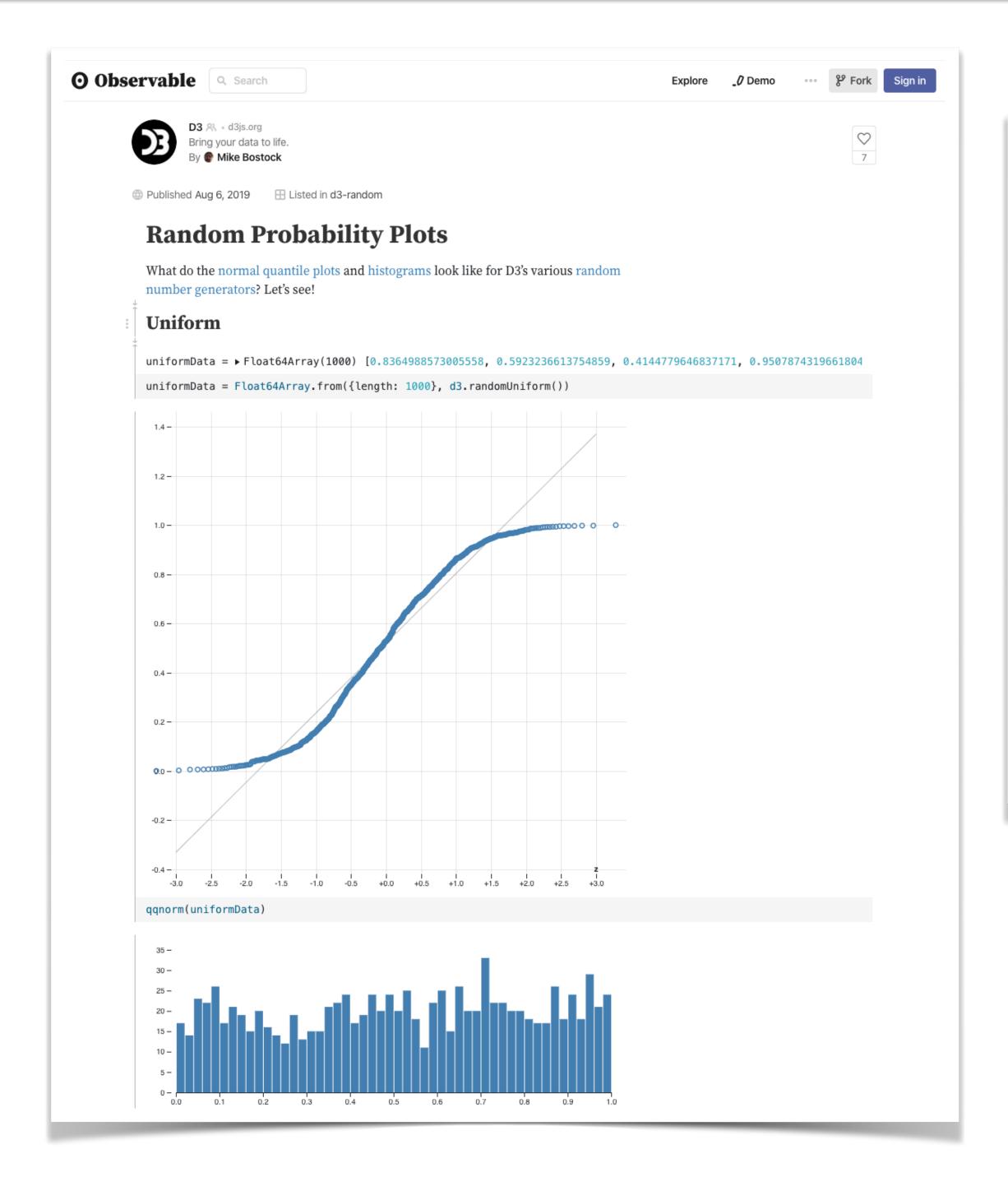
It's hard requires extensive training reading documentation not discoverable It's time consuming even simple things require effort Some operations are difficult e.g., labeling data points

1	# Кеер
2	avy_df
3	
4	# Remov
5	avy_df
6	
7	# Split
8	avy_df[
9	
10	# Remov
11	avy_df
12	avy_df
13	
14	# Keep
15	avy_df
16	avy_df.

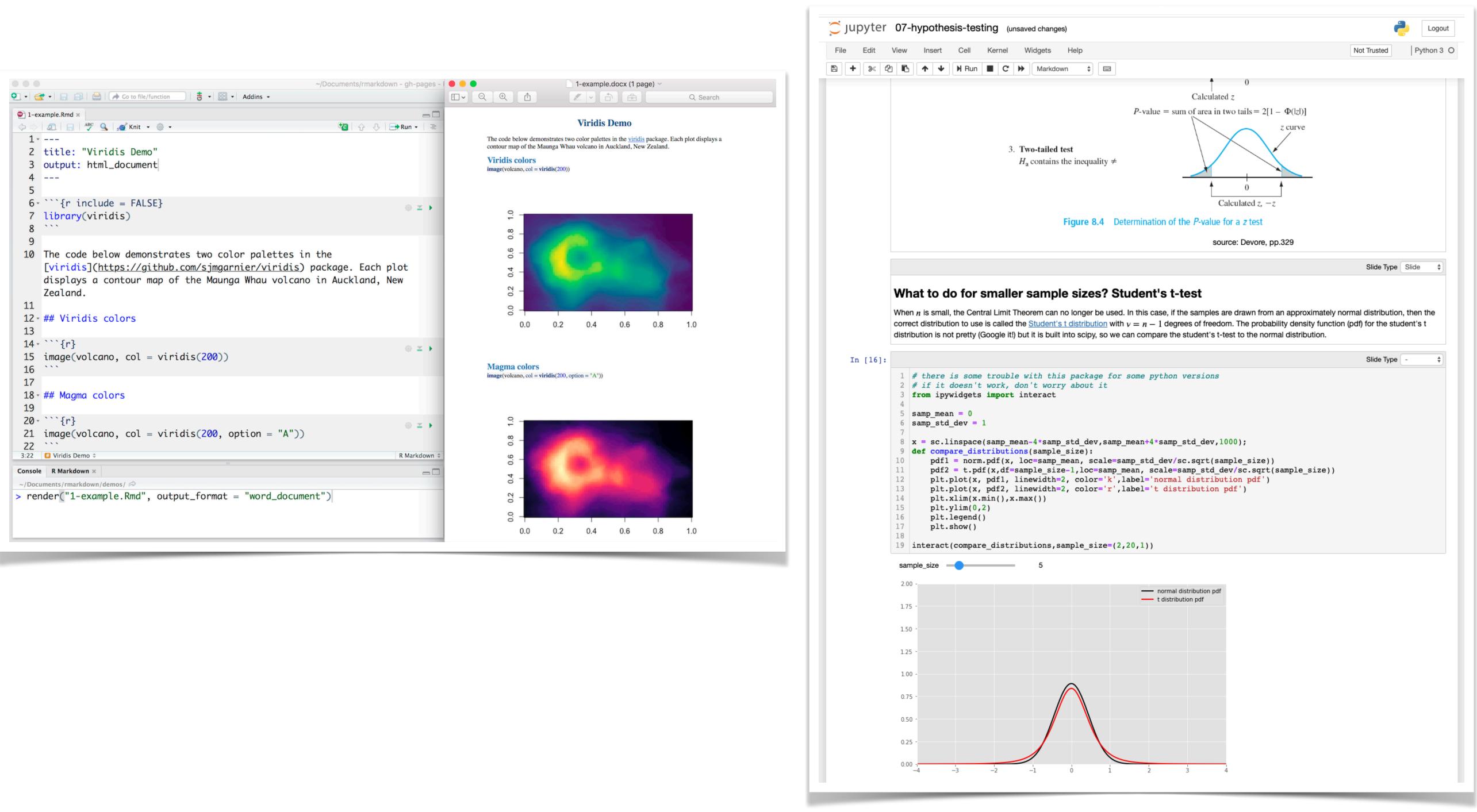
```
this cell
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= avy_df[avy_df['Coordinates']==avy_df['Coordinates']]
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= avy_df[(-114 < avy_df['lon'].astype('float')) & (avy_df['lon'].astype('float') < -108)]</pre>
columns we need
= avy_df[['Date', 'Region', 'Trigger', 'lat', 'lon']]
head ()
```



COMPUTATIONAL NOTEBOOKS: A MIDDLE GROUND?



Observable



R Markdown

Jupyter Notebooks

COMPUTATIONAL NOTEBOOKS: A MIDDLE GROUND?

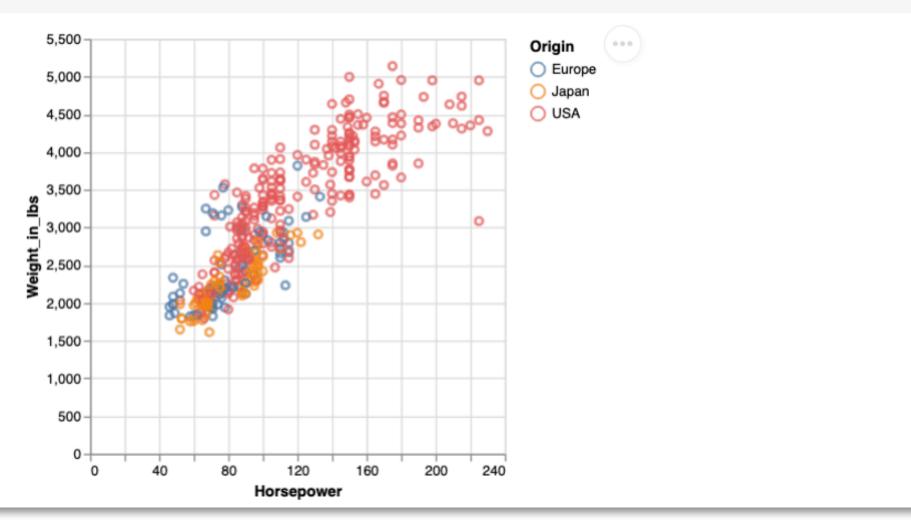
Yes Afford both scripting and interactive visualization But visualizations are a dead end can't "use" interaction in code e.g., changing a label, or filtering a value

Cars Dataset Analysis

Here we will see the relationship between weight of the car and horsepower

```
[6] selection = alt.selection_interval()
    chart = alt.Chart(cars_df).mark_point().encode(
          x="Horsepower:Q",
          y="Weight_in_lbs:Q",
           color=alt.condition(
              selection, "Origin:N", alt.value("gray")
         ).properties
           width=300
           height=300
          .add_params
             selection
```

chart

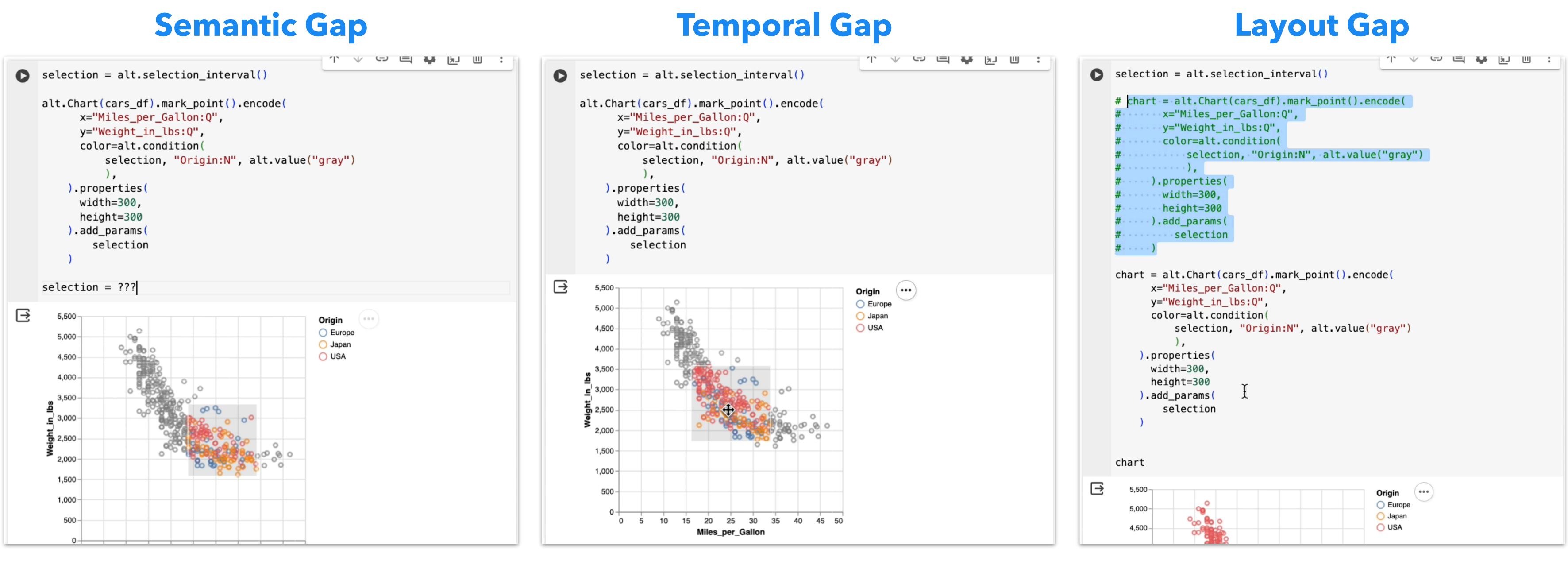


Code

Visualizations



GAPS BETWEEN CODE AND INTERACTIVE OUTPUTS*



Information only flows from code to visualization

Changes made to code are preserved Changes made to vis are lost

Changes in code are messy

*[Wu, Hellerstein, Satyanarayan, UIST 2020]



THESIS: BRIDGING BETWEEN CODEAND INTERACTIVE VIS IS USEFULEasy handoffs are important!



THE PERSIT APPROACH

PRINCIPLE

Track events in interactive visualizations

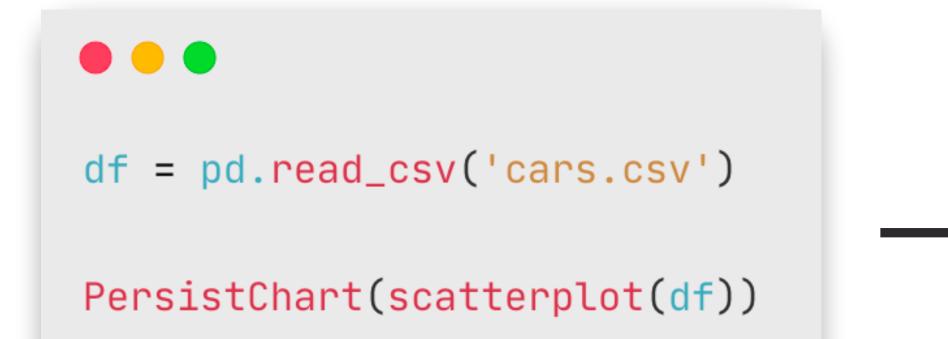
Map them to data frame operations

Operations then applied to data frame

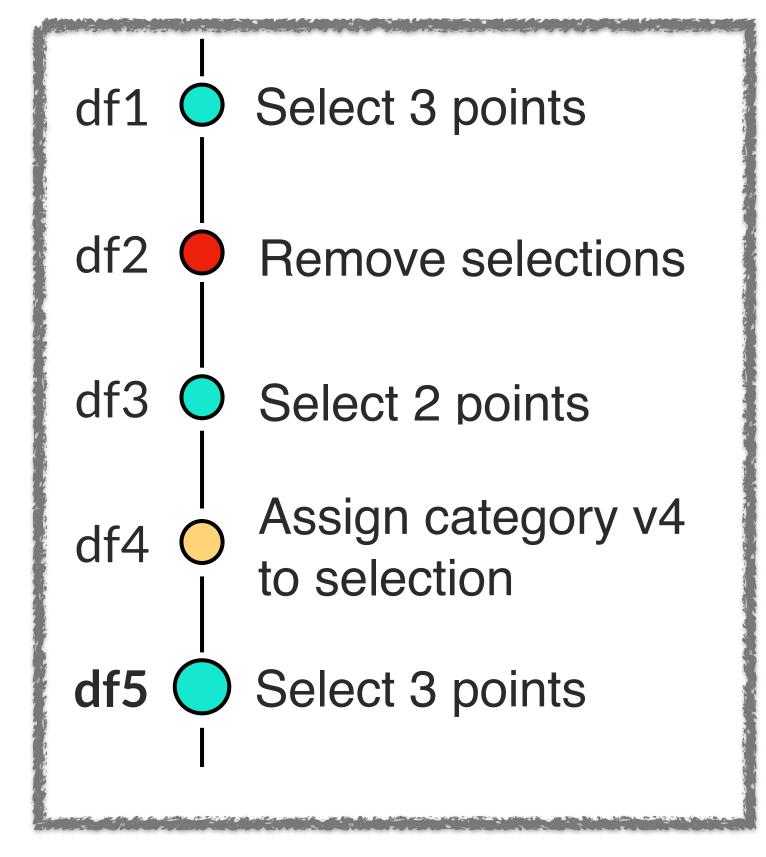


HOW IT WORKS

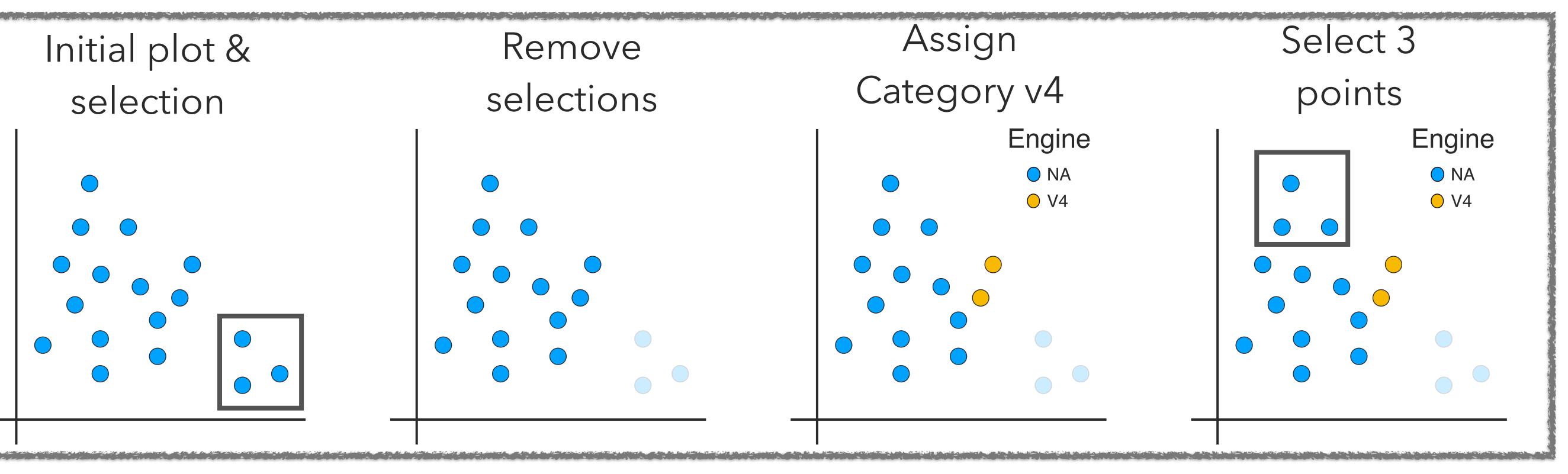
Code to create chart

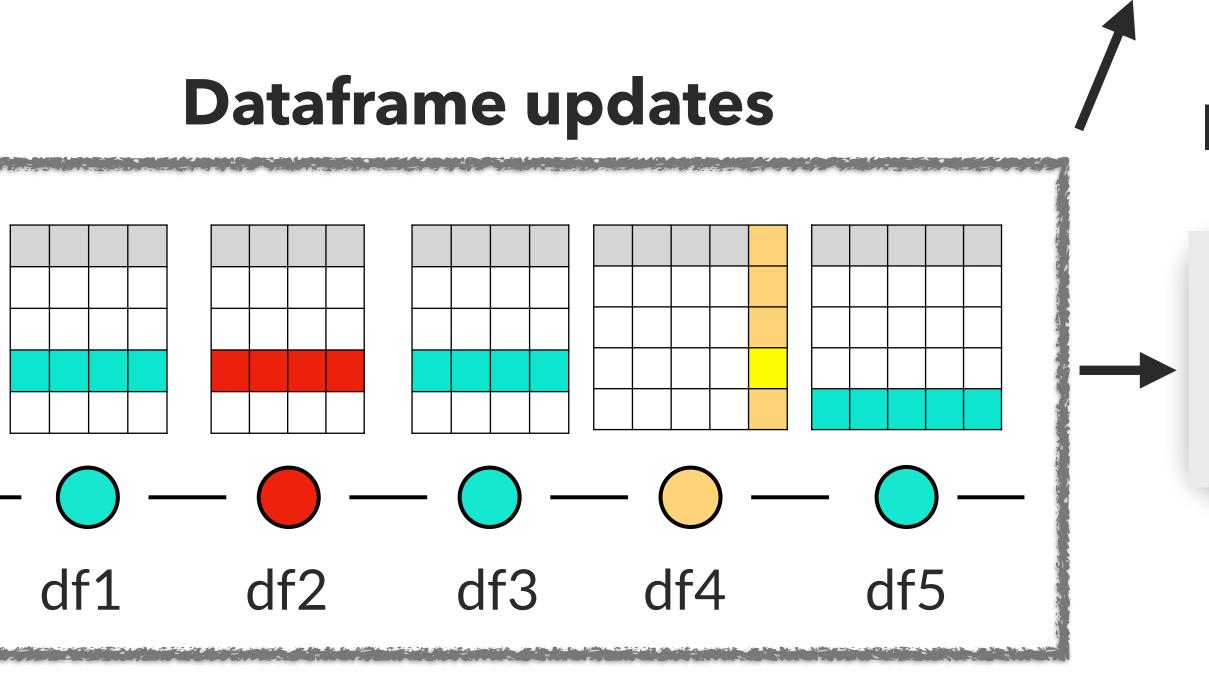


Provenance

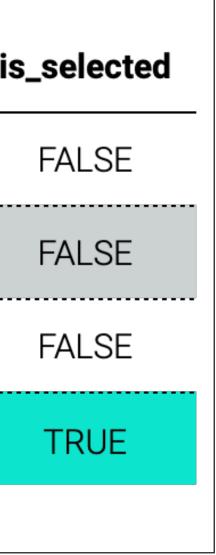


Interactive Visualization





In code	Name	MPG	Cylinders	origin	Engine	is
	ford	18	8	USA	NA	
df5.head()	dodge	15	8	USA	NA	
	volkswagen	22	4	Europe	V4	
	amc	16	8	USA	NA	
	27 rows x 8 c	olumn	S			

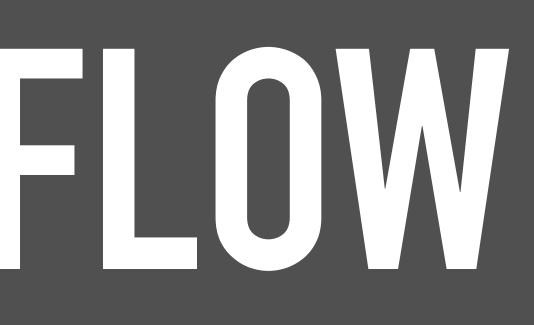


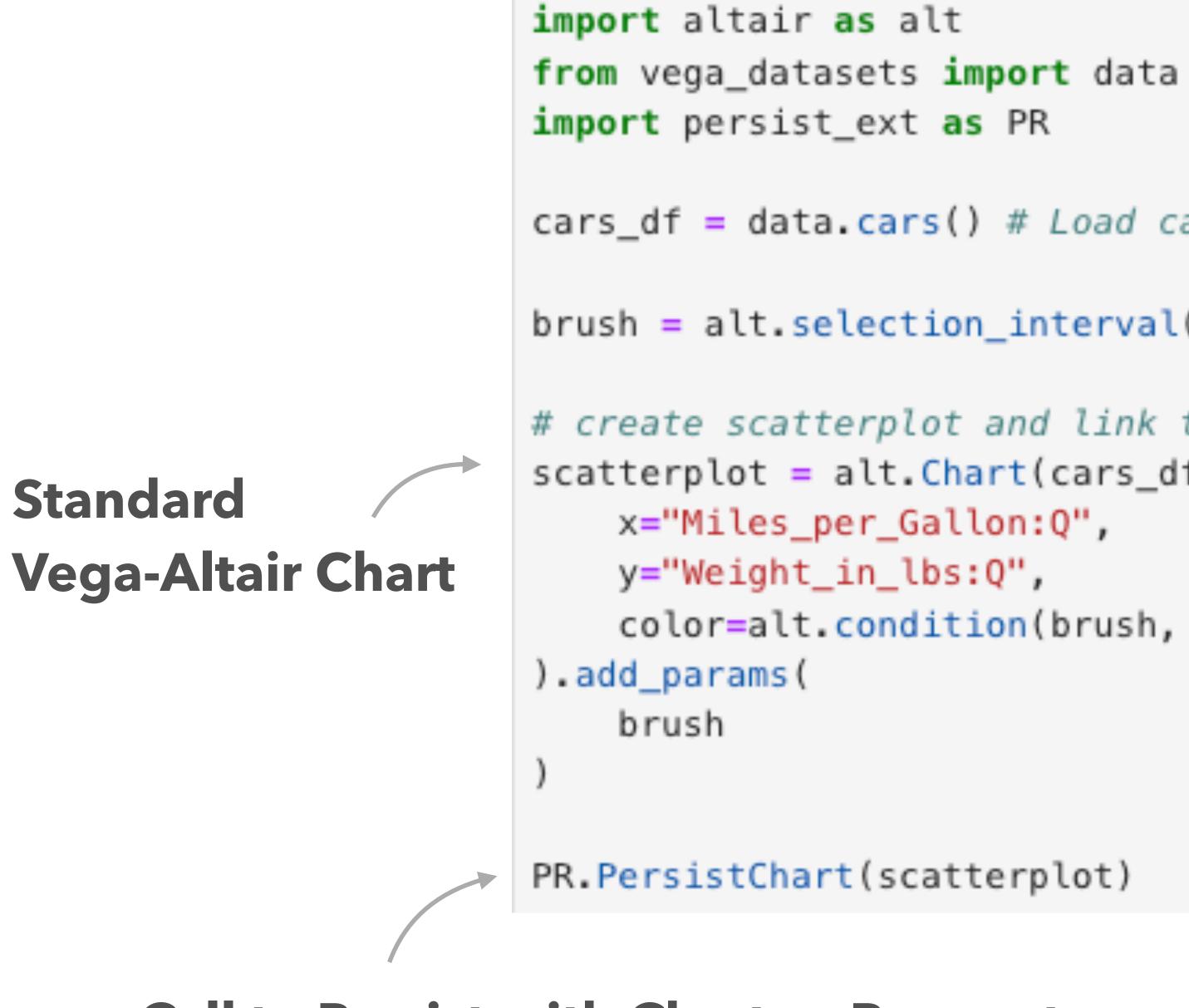
OPERATIONS

Selection **Edit column names, edit cells EV** Sort rows/columns **Drop columns Filter (in/out) items E** Label items **Categorize items Change data types**



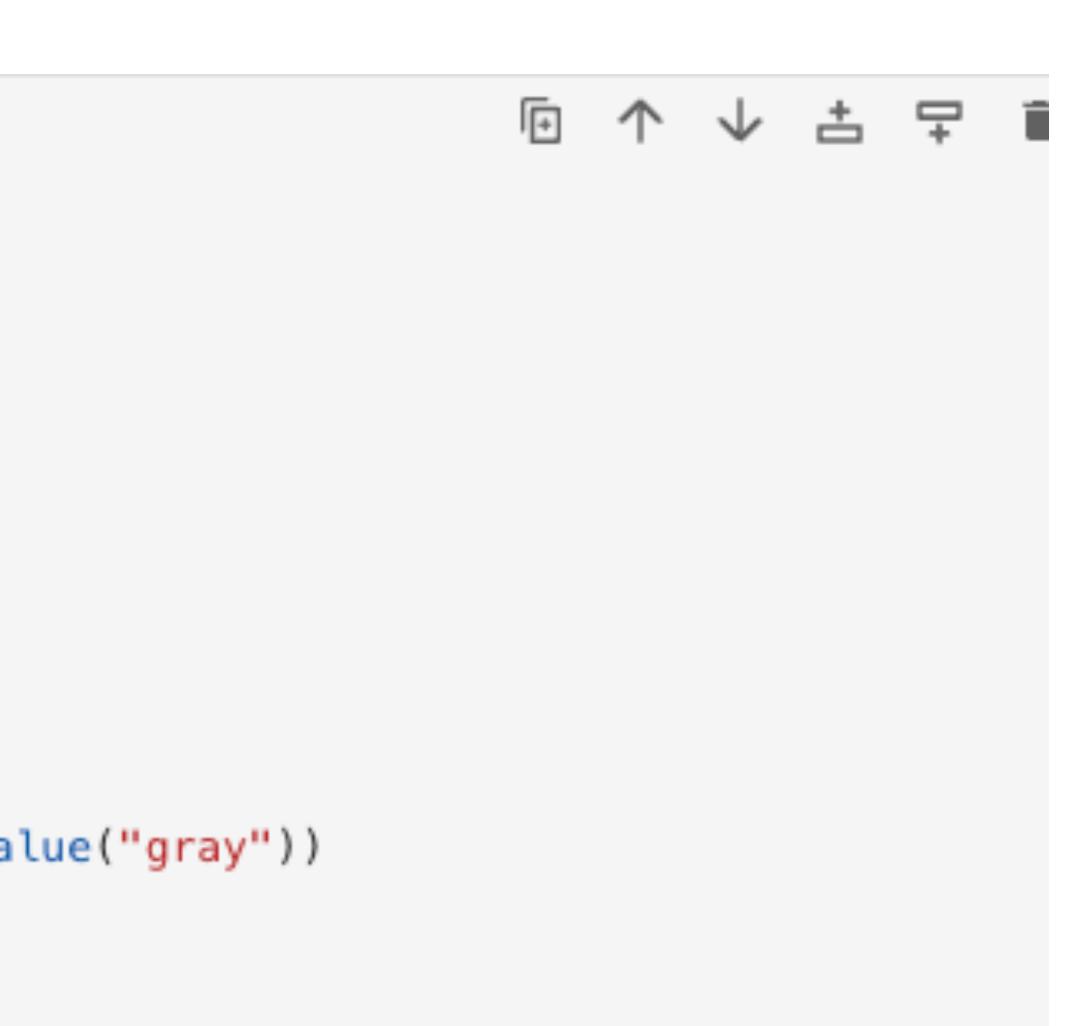
PERSIST WORKFLOW





```
cars_df = data.cars() # Load cars dataset
brush = alt.selection_interval() # Create a 2d brush
# create scatterplot and link to brush
scatterplot = alt.Chart(cars_df).mark_point().encode(
   x="Miles_per_Gallon:Q",
    color=alt.condition(brush, alt.value("steelblue"), alt.value("gray"))
```

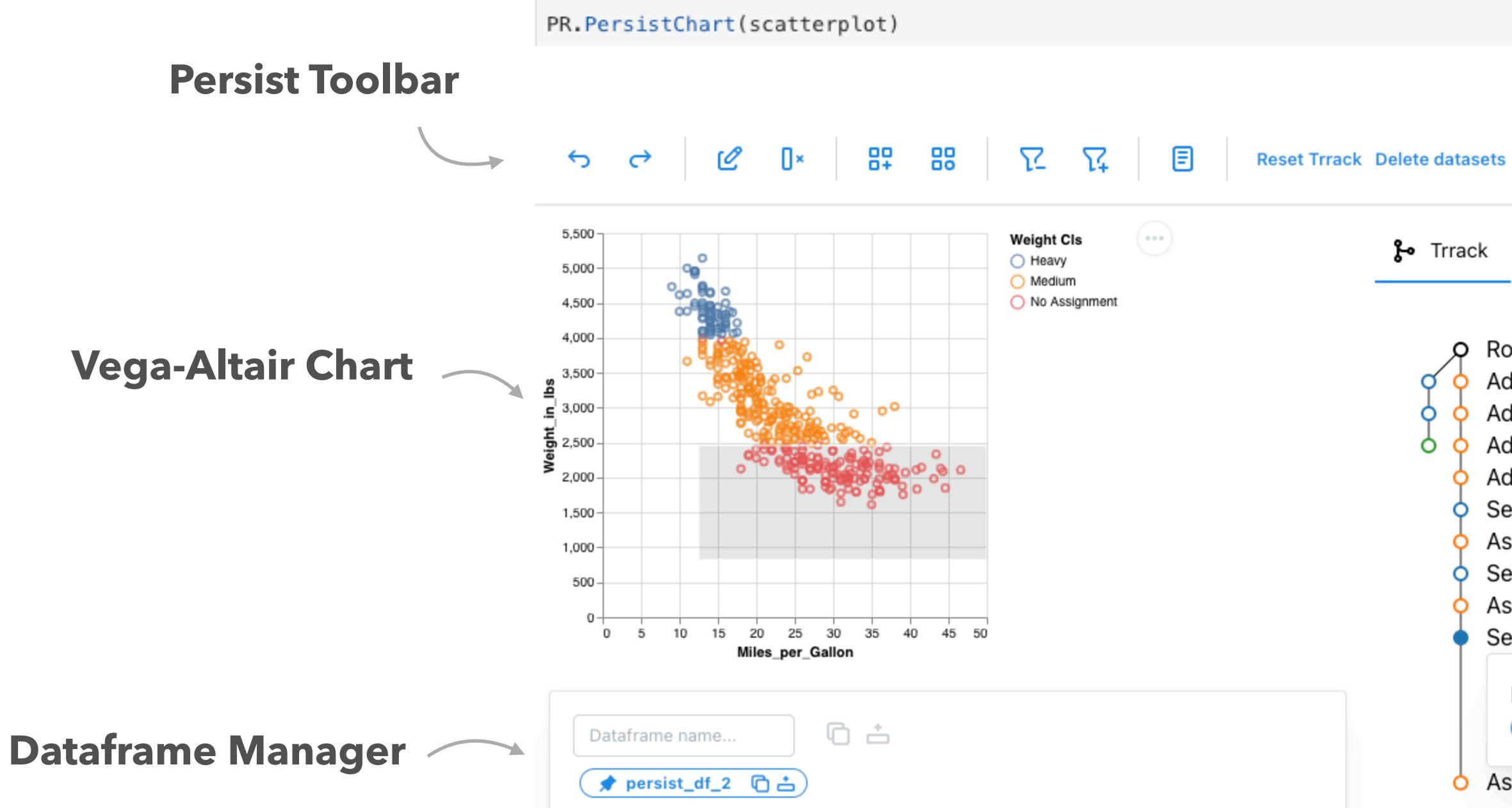
Call to Persist with Chart as Parameter



```
import altair as alt
         from vega_datasets import data
         import persist_ext as PR
         cars_df = data.cars() # Load cars dataset
         brush = alt.selection_interval() # Create a 2d brush
         # create scatterplot and link to brush
         scatterplot = alt.Chart(cars_df).mark_point().encode(
             x="Miles_per_Gallon:Q",
             y="Weight_in_lbs:Q",
             color=alt.condition(brush, alt.value("steelblue"), alt.value("gray"))
         ).add_params(
             brush
         PR.PersistChart(scatterplot)
5,500 -
                                             Weight Cls
                                                        ....

    Heavy

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                                               ) Medium
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                             0
          3,500-
         .⊆ 3,000-
         ີ່ຫຼັ 2,500 -
                                                                              0
          2,000-
```



⊱ Trrack 🛛 🗄 Summary

Provenance History

Root

Add new category 'We...

- Add new option 'Heav...
- Add new option 'Medi...
- Add new option 'Light'...
- Selected Miles_per_G...
- Assign Weight Cls (He...
- Selected Miles_per_G...
- Assign Weight Cls (Me...
- Selected Miles_per_G...

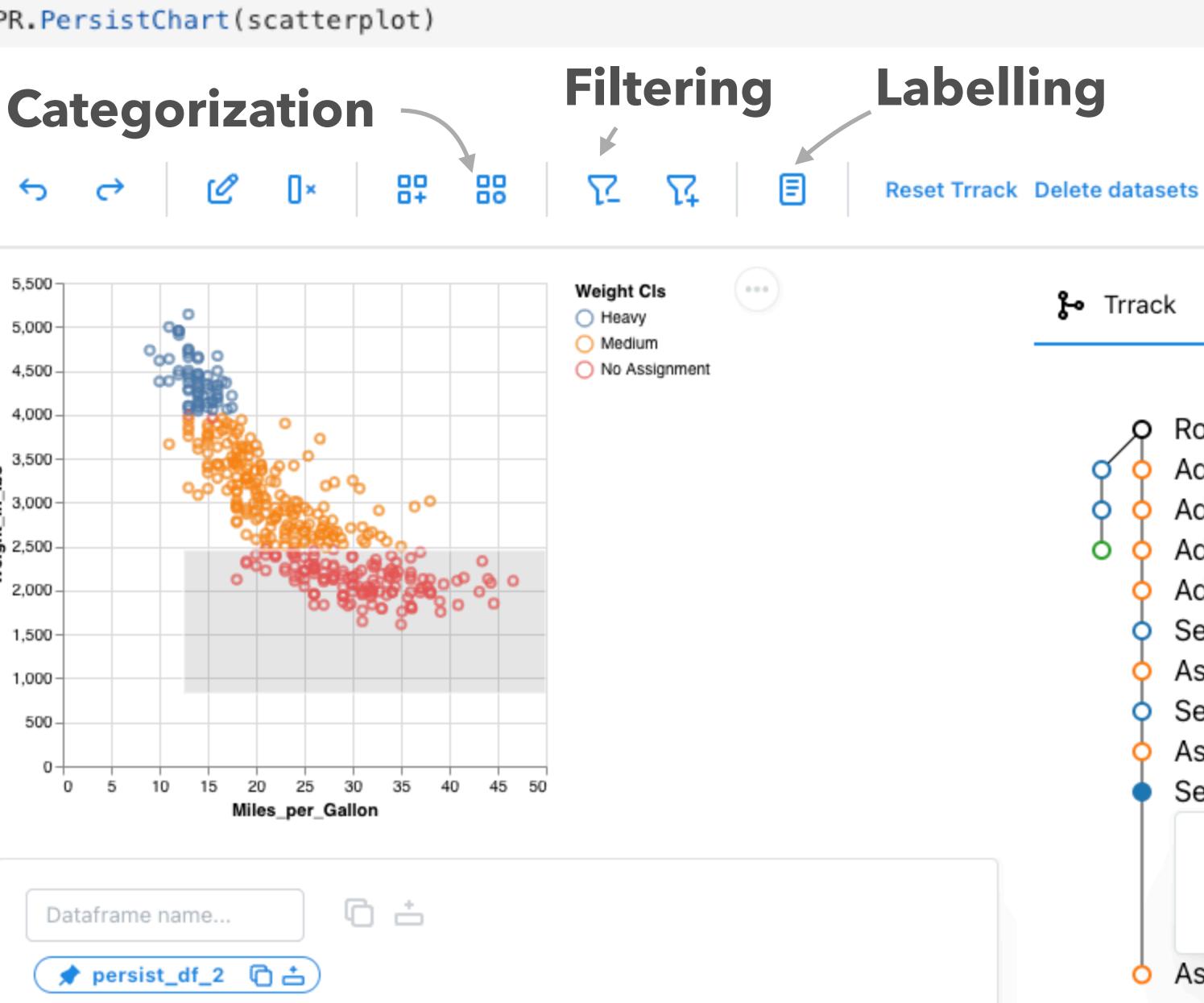
Dataframes:

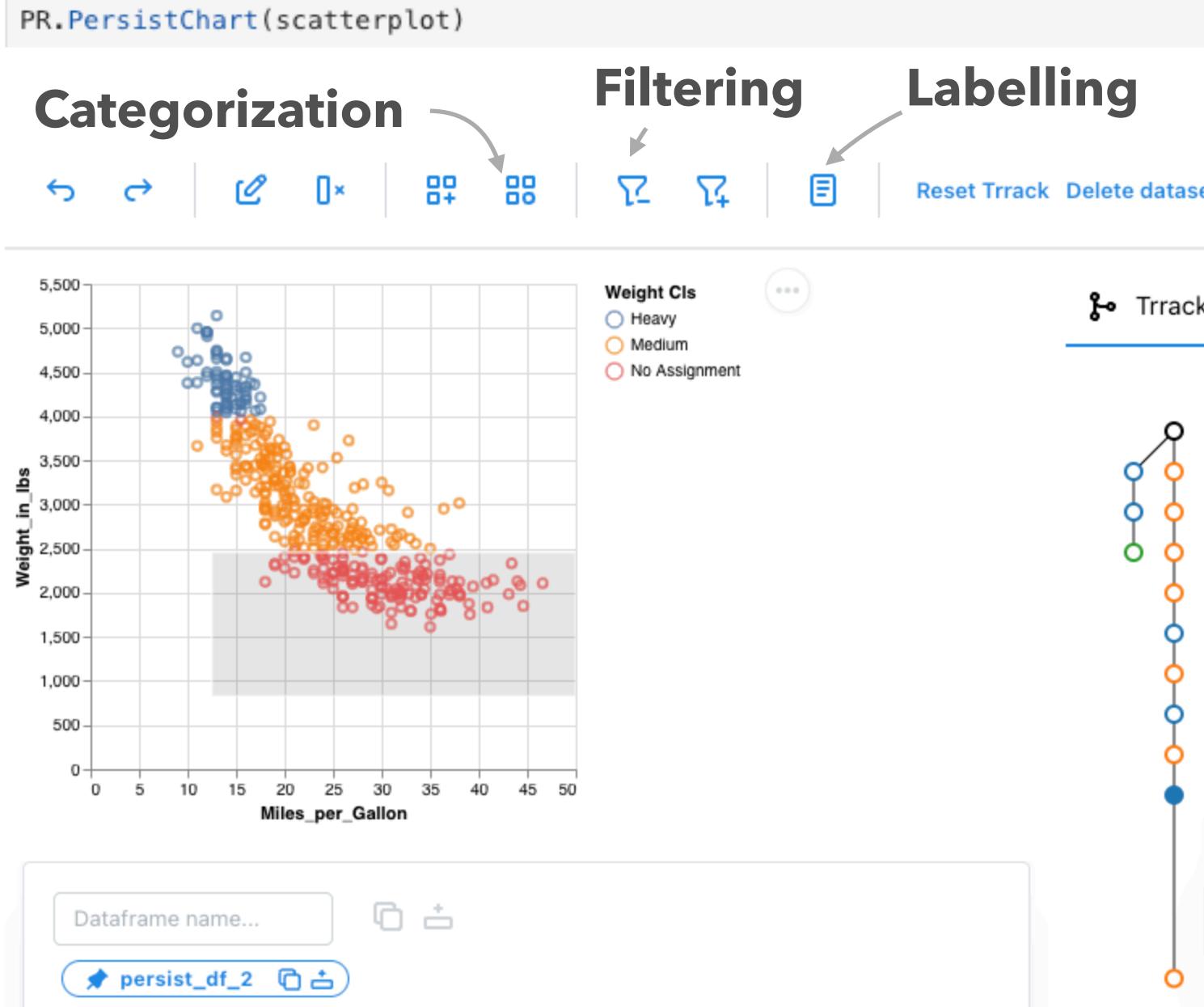
🖈 persist_df_2 🛛 📥

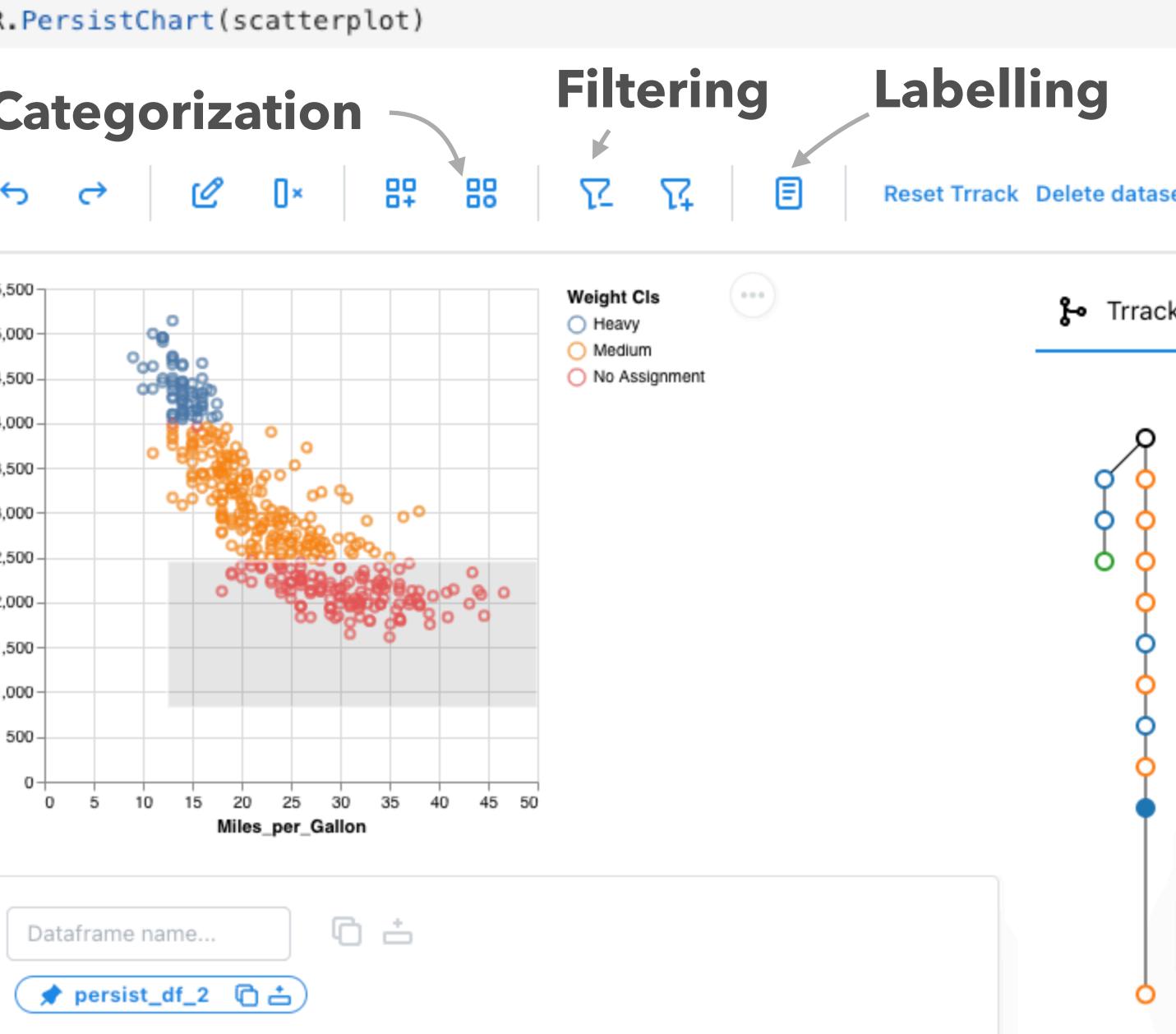
Assign Weight Cls (Lig...



```
import altair as alt
from vega_datasets import data
import persist_ext as PR
cars_df = data.cars() # Load cars dataset
brush = alt.selection_interval() # Create a 2d brush
# create scatterplot and link to brush
scatterplot = alt.Chart(cars_df).mark_point().encode(
   x="Miles_per_Gallon:Q",
   y="Weight_in_lbs:Q",
    color=alt.condition(brush, alt.value("steelblue"), alt.value("gray"))
).add_params(
    brush
```







⊱ Trrack 🛛 🖶 Summary

Root

Add new category 'We...

- Add new option 'Heav...
- Add new option 'Medi...
- Add new option 'Light'...
- Selected Miles_per_G...
- Assign Weight Cls (He...
- Selected Miles_per_G...
- Assign Weight Cls (Me...
- Selected Miles_per_G...

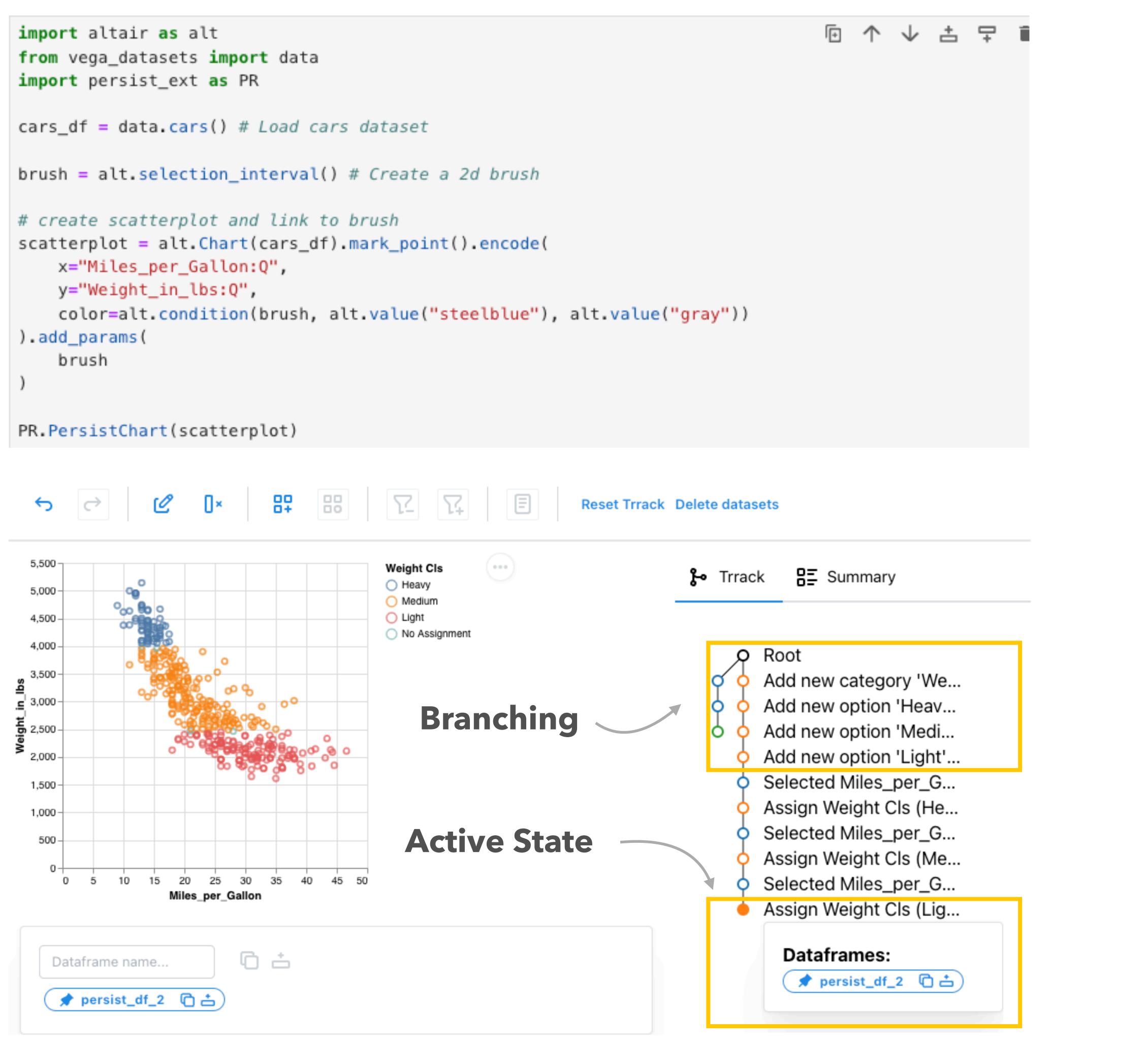
Dataframes:



Assign Weight Cls (Lig...

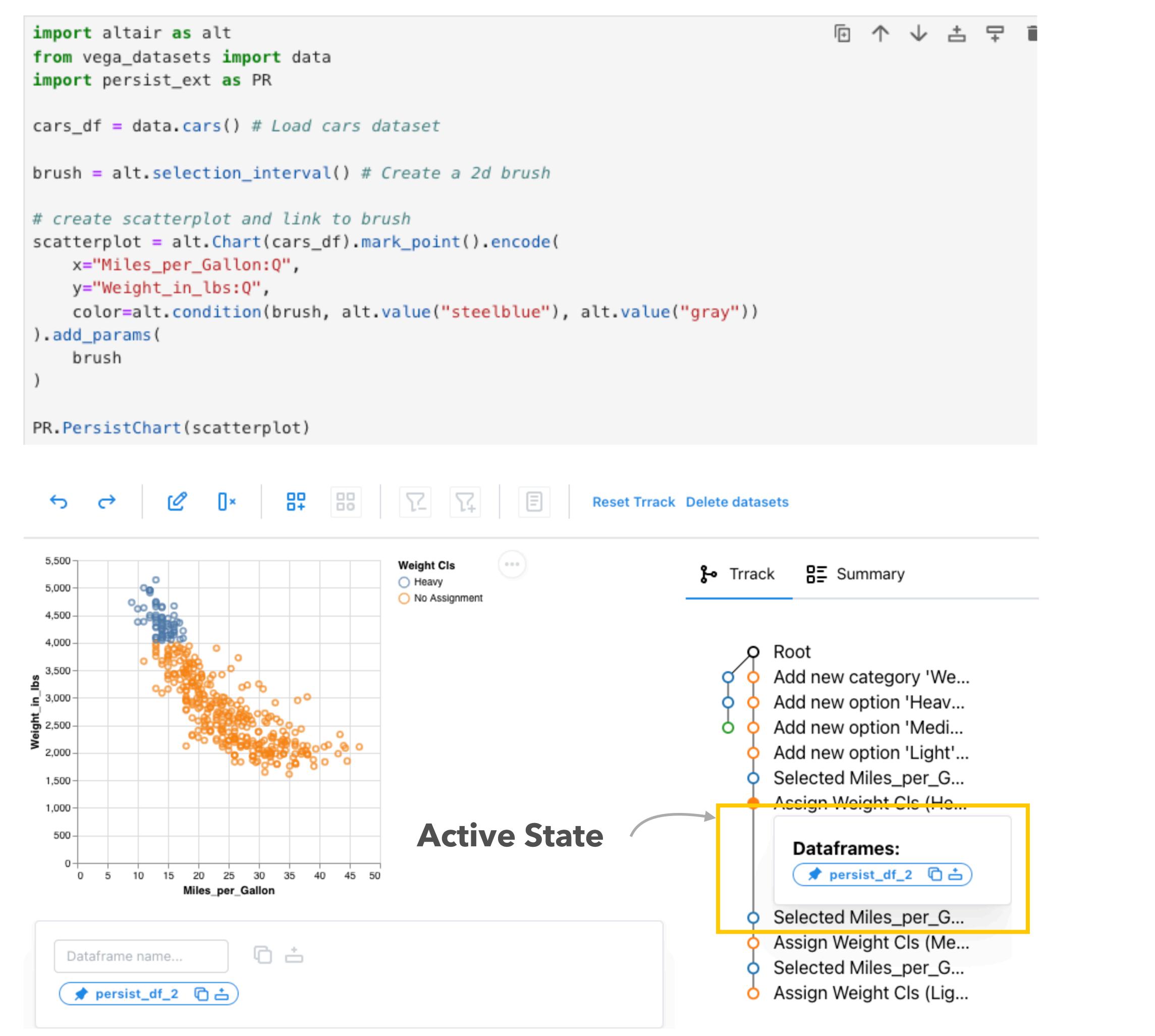
Branches and Choosing a State in provenance support non-linear analysis, addressing the layout gap

```
import altair as alt
from vega_datasets import data
import persist_ext as PR
cars_df = data.cars() # Load cars dataset
brush = alt.selection_interval() # Create a 2d brush
# create scatterplot and link to brush
scatterplot = alt.Chart(cars_df).mark_point().encode(
   x="Miles_per_Gallon:Q",
   y="Weight_in_lbs:Q",
    color=alt.condition(brush, alt.value("steelblue"), alt.value("gray"))
).add_params(
    brush
```

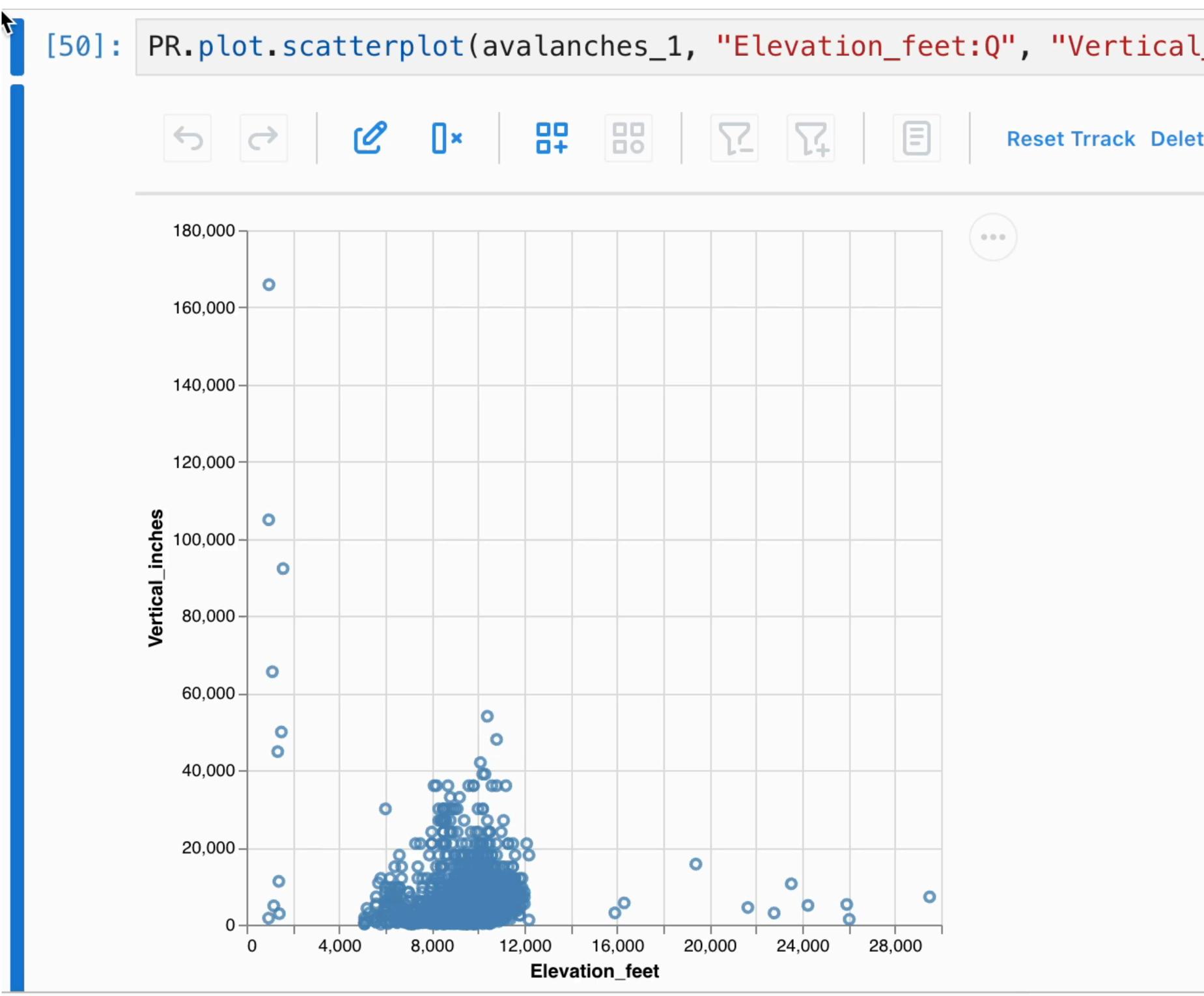


Branches and Choosing a State in provenance support non-linear analysis, addressing the layout gap

```
import altair as alt
                                                                               ⊡ ↑ ↓
                                                                                          ±
                                                                                             7
from vega_datasets import data
import persist_ext as PR
cars_df = data.cars() # Load cars dataset
brush = alt.selection_interval() # Create a 2d brush
# create scatterplot and link to brush
scatterplot = alt.Chart(cars_df).mark_point().encode(
   x="Miles_per_Gallon:Q",
   y="Weight_in_lbs:Q",
   color=alt.condition(brush, alt.value("steelblue"), alt.value("gray"))
).add_params(
   brush
```

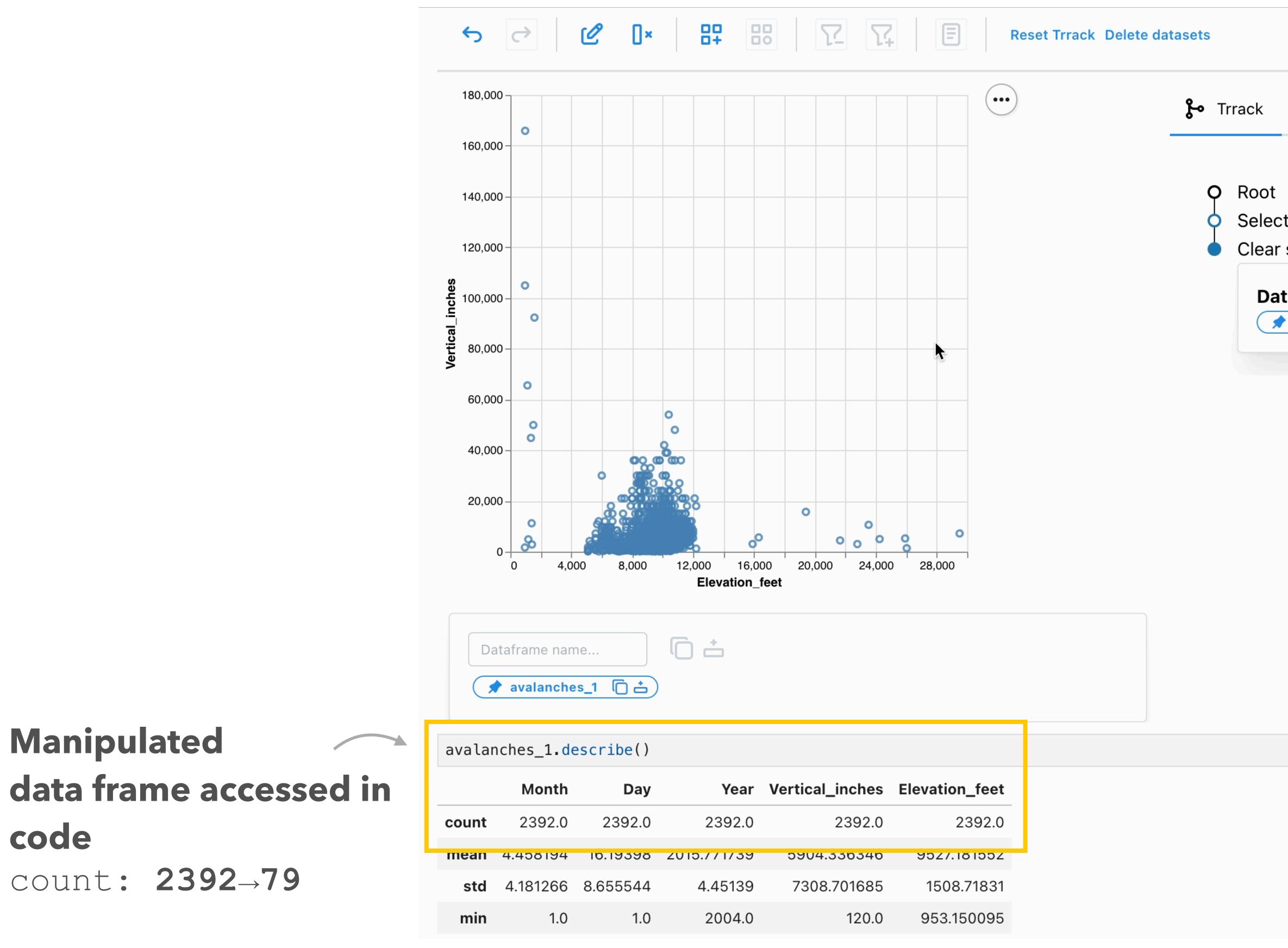


Persist **re-runs the interactions** in the output, addressing **the temporal gap**



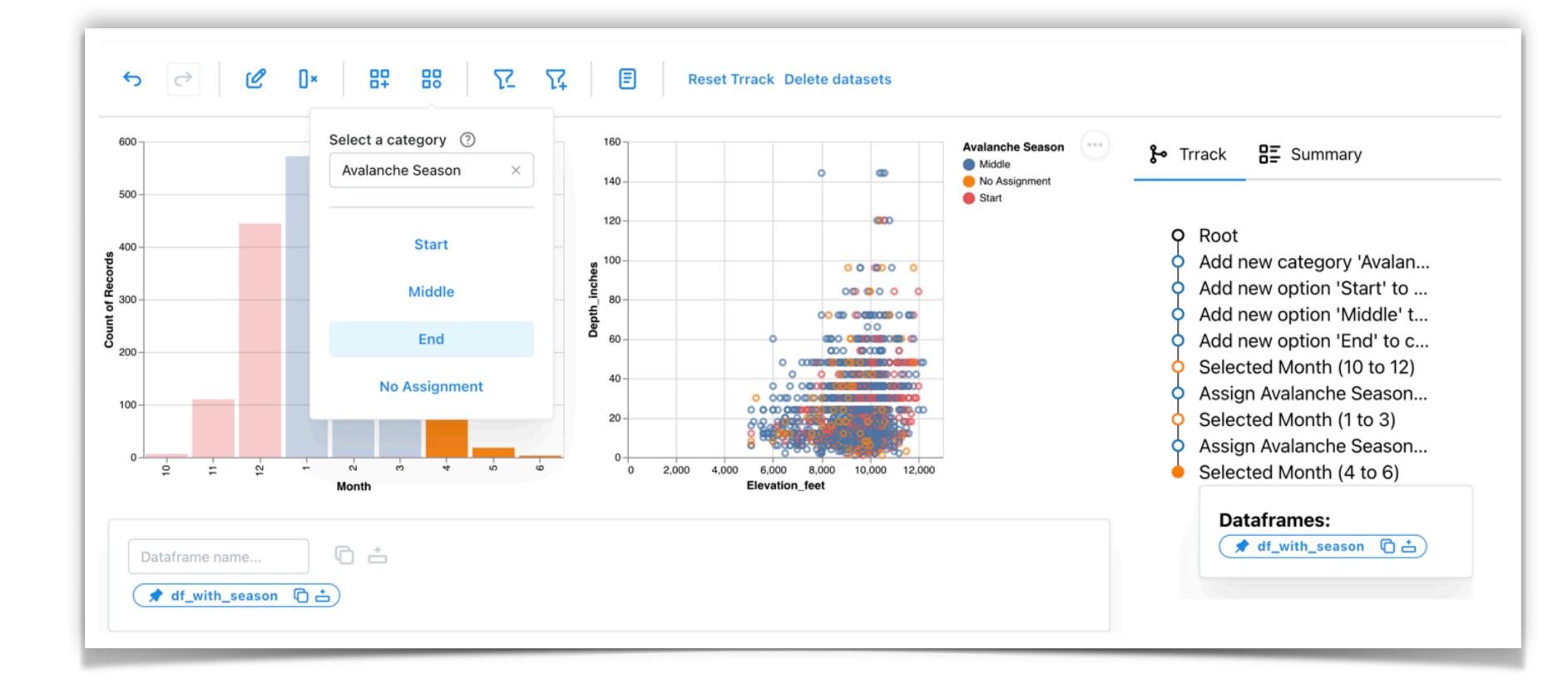
incl	hes	:Q",	df_na	ame="	aval	anche	es_2'	")
e datas	sets							
ዮ	Trra	ck	8= Sur	mmary	,			
	D R	oot						
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			aframe avalanch)		

Persist *applies interactions to data frames* that can be accessed in code, addressing the **semantic gap**



🗄 Summary
cted Elevation_feet (10252 t selections
taframes:
🕈 avalanches_1 🗋

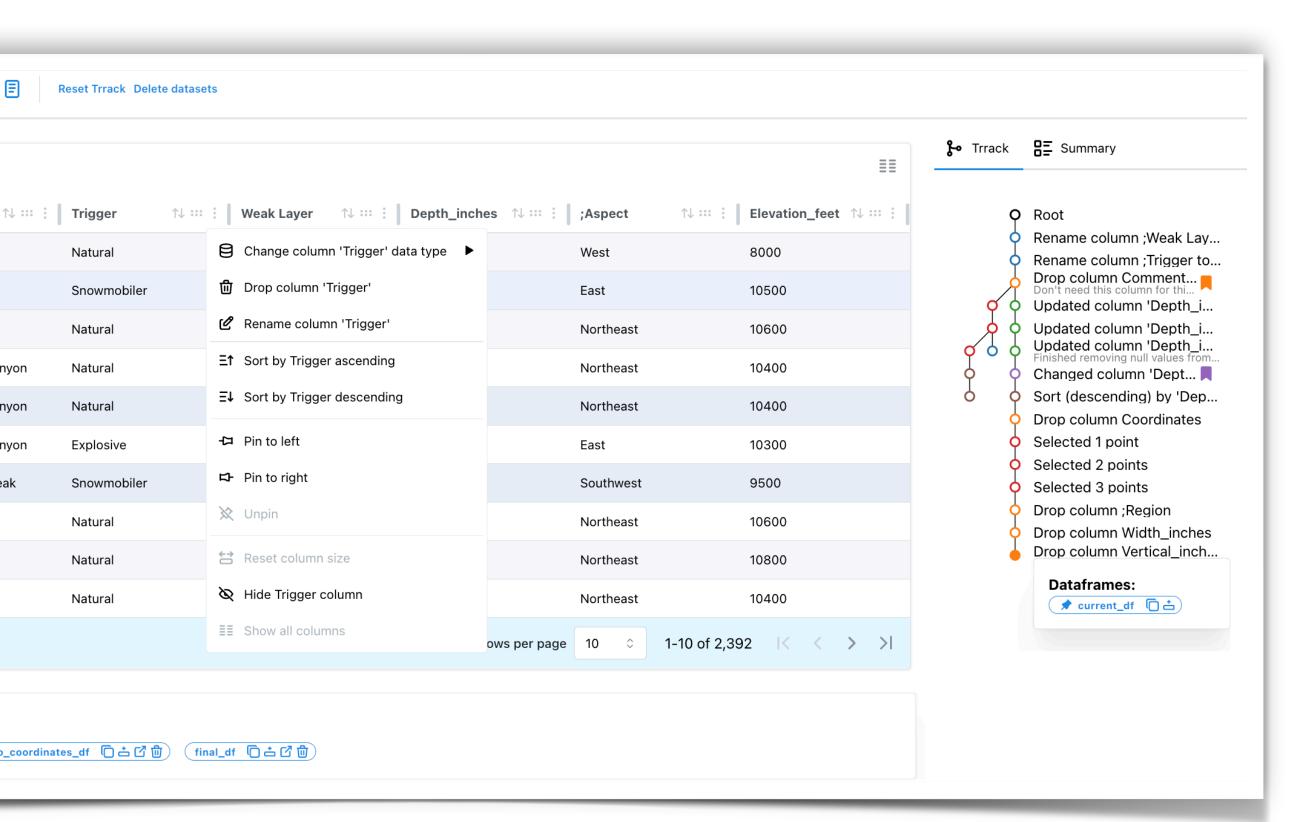
VISUALIZATION OPTIONS



Arbitrary Vega-Altair Charts

Q	Search				
	#	Date ↑↓ ፡፡፡ :	Place 1		
	2338	3-23-2023	Dry Creek		
	955	1-19-2014	Whitney Basin		
	1028	2-21-2014	Chalk Creek		
	1024	2-17-2014	Upper Weber Cany		
	998	2-12-2014	Upper Weber Cany		
	938	1-14-2014	Upper Weber Cany		
	1299	1-26-2016	Currant Creek Pea		
	1044	2-28-2014	Chalk Creek		
	2348	3-30-2023	Bunnels		
	1977	4-6-2021	Blue Ice		
3 of 2392 row(s) selected					
	aframe name		「☐ ் Ґ ⑪ (no_		





An Interactive Data Table

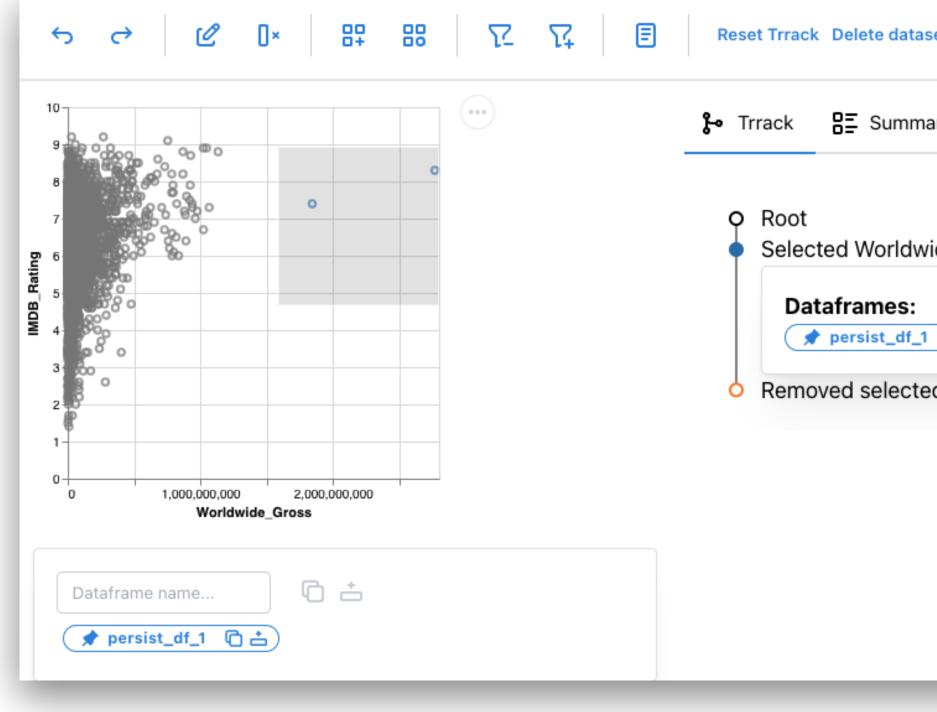
VEGA-ALTAIR

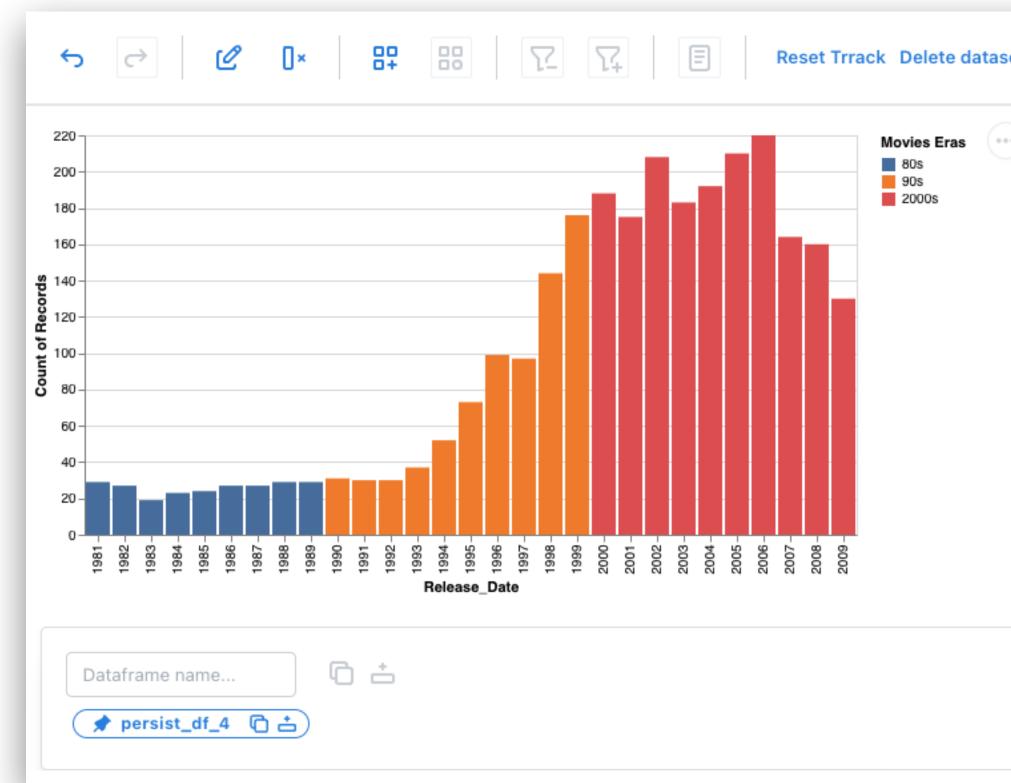
Persist works with most Vega-Altair charts Updates Vega charts:

Use original chart spec when possible (e.g., filters) Update spec when necessary (categories, labels)

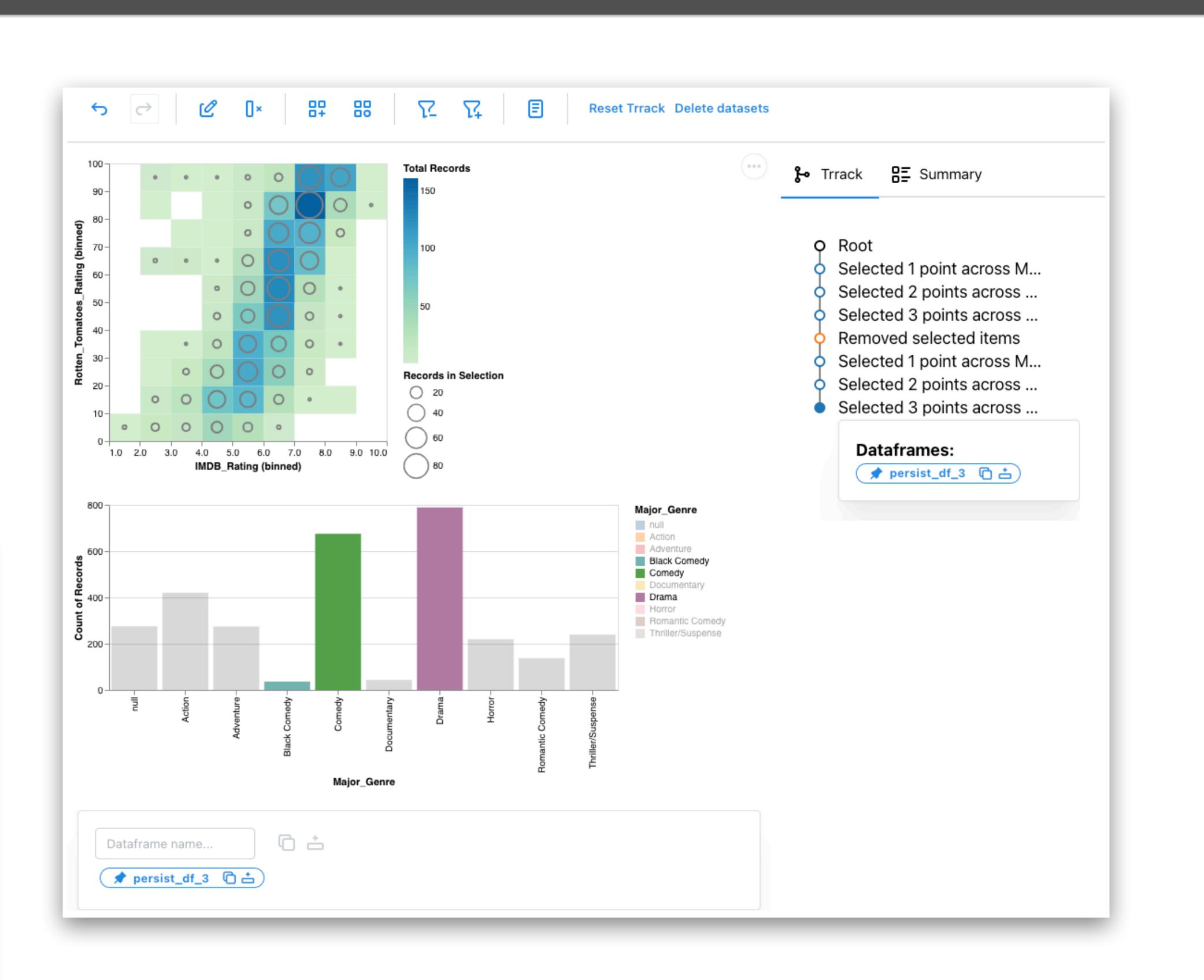
"Listens" to native operations (selections)

EXAMPLE CHARTS





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ummary		
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les: t_df_1 ☐ ≟)		
elected items		
lected items		
datasets		
s	⊱ Trrack 📲 Summary	
	O Root	
	 Add new category 'Movie Add new option '80s' to c 	
	 Add new option '80s' to c Add new option '90s' to c 	
	Add new option '2000s' t	
	Selected Release_Date (1	
	 Assign Movies Eras (80s) Selected Release_Date (1 	
	Assign Movies Eras (90s)	
	Selected Release_Date (2	
	 Assign Movies Eras (200 	
	Dataframes:	
	📌 persist_df_4 🔂 📩	



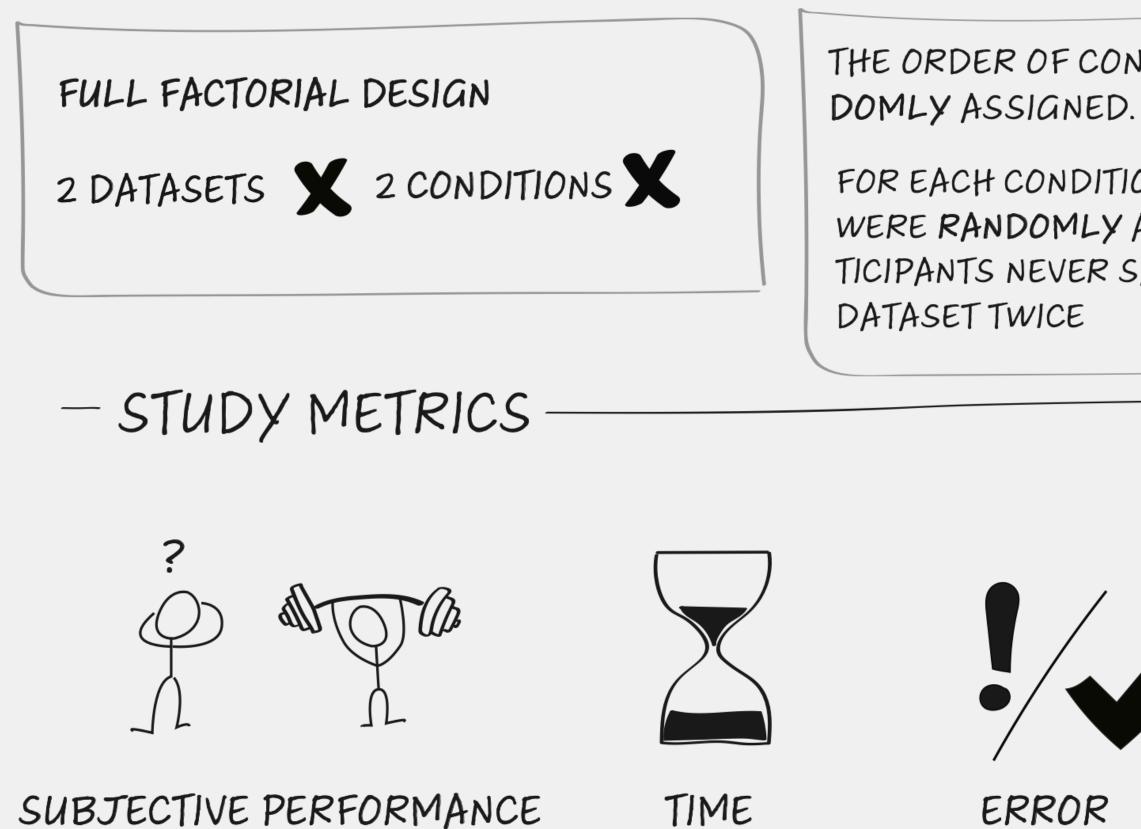
EVALUATION



IN-LAB STUDY

III STUDY DESIGN

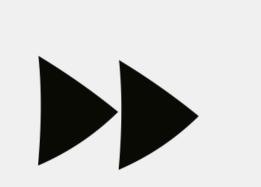
WE RECRUITED ELEVEN PARTICIPANTS FOR THE STUDY. PARTICIPANTS ALL HAD PRIOR EXPERIENCE WITH PYTHON AND PANDAS.



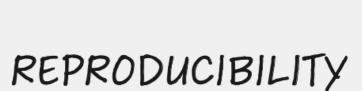
SUBJECTIVE PERFORMANCE

THE ORDER OF CONDITIONS WAS RAN-

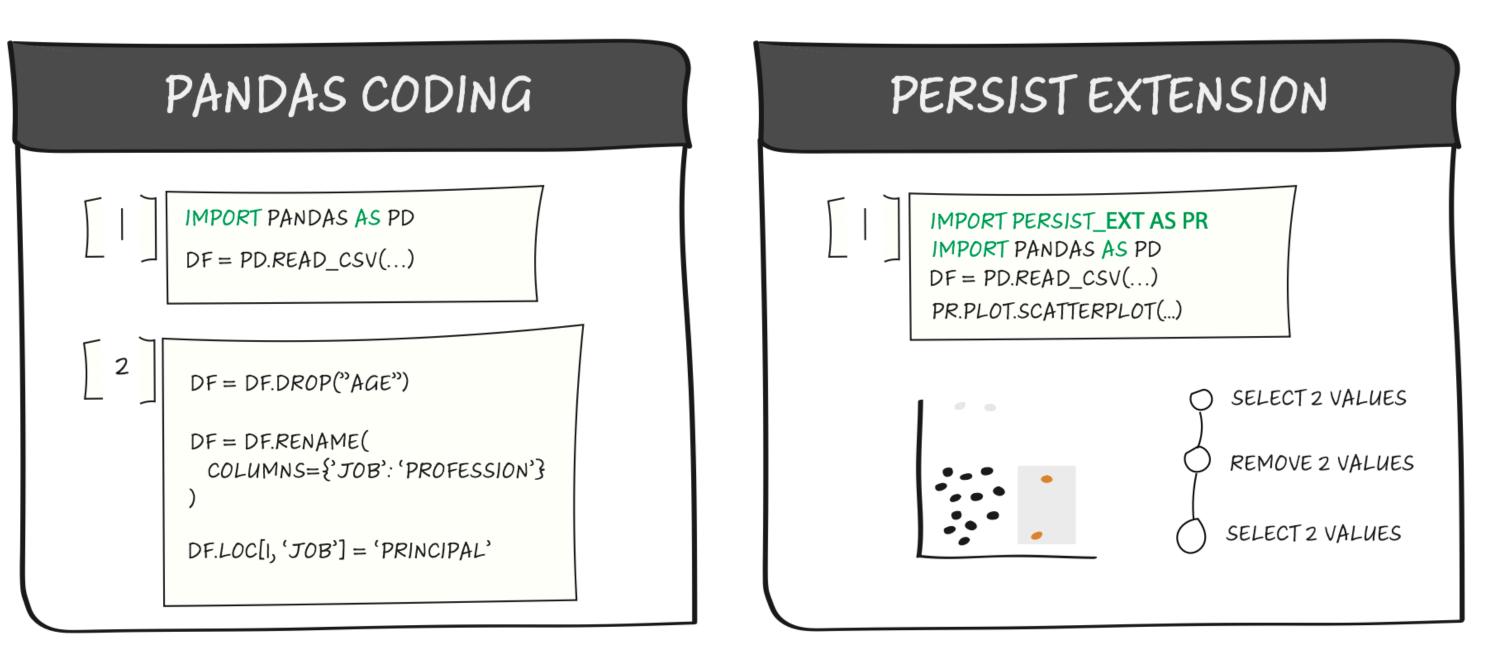
FOR EACH CONDITION, DATASETS WERE RANDOMLY ASSIGNED. PAR-TICIPANTS NEVER SAW THE SAME

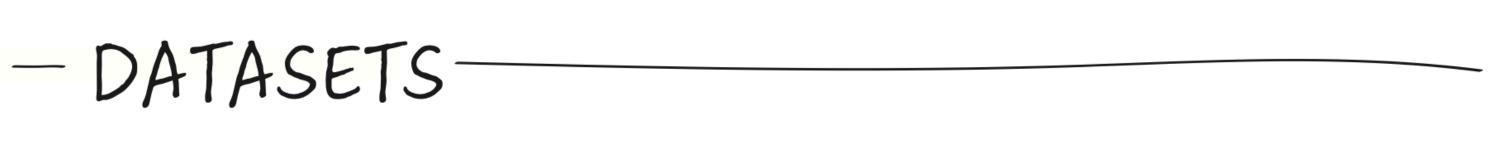


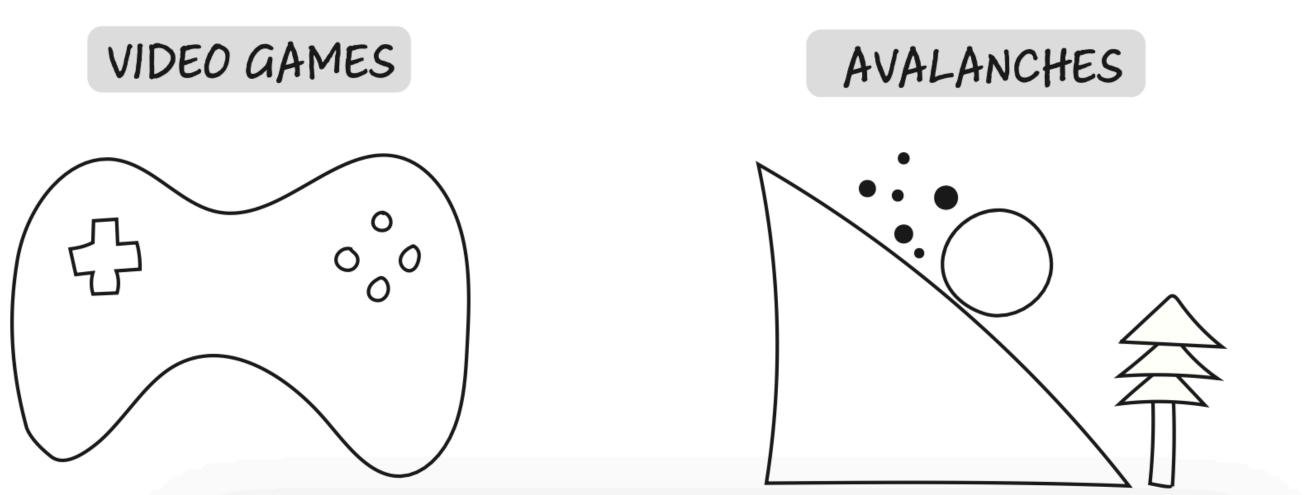
ERROR



- CONDITIONS



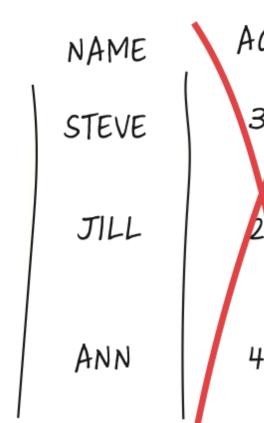






IN-LAB STUDY

- TASKS PARTICIPANTS MADE THE FOLLOWING CHANGES TO A DATASET

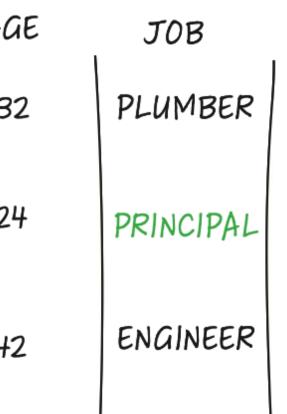


NAME	AC
STEVE	3:
JILL	24
ANN	4:

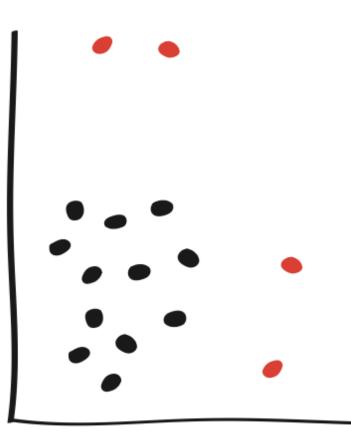
AGE	JOB	NAME	AGE	PROFESSION	I
32	PLUMBER	STEVE	32	PLUMBER	
24	TEACHER	JILL	24	TEACHER	
42	ENGINEER	Ann	42	ENGINEER	
N.				1	

REMOVE COLUMNS

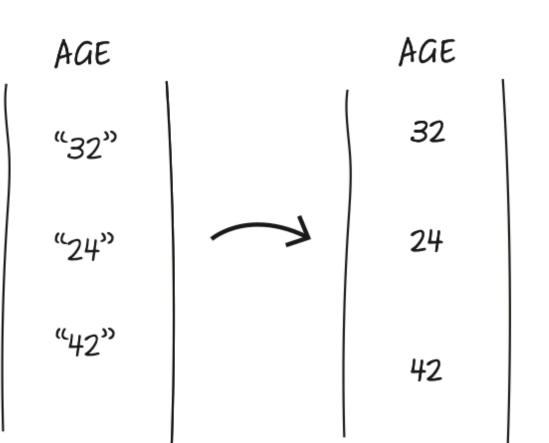
RENAME COLUMNS



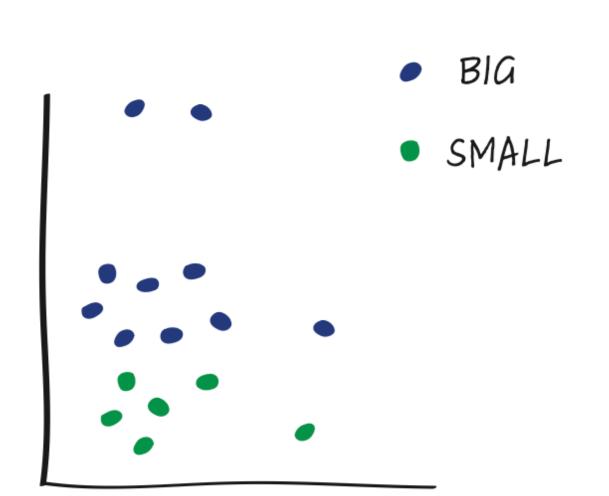
EDIT VALUES



FILTER DATA



CHANGE DATA TYPE



ADD CATEGORICAL COLUMAN

RESULTS



tasks correctly using Persist, compared to 85% for Pandas



times faster with Persist

1111

notebooks using Persist were **reproducible**

only 7/11 using pandas were



"easier as compared to the code and everything was visible [...] and it didn't take much time."

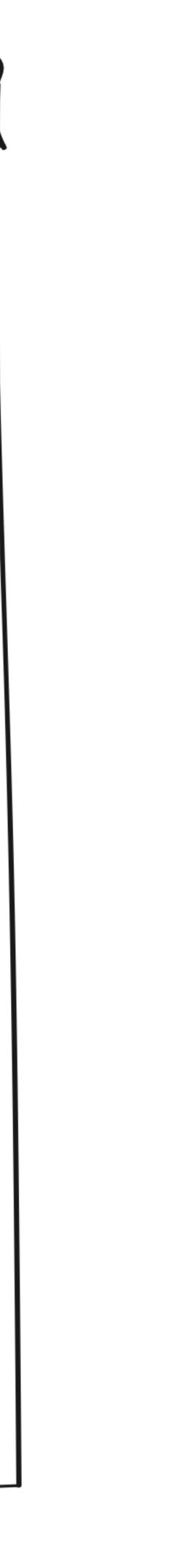
"Changing the category type, or adding new categories or removing anomalies from data, they were very much easier in [Persist] than coding." - M7

"The thing I really liked about is version control, which shows the history of all operations [...] and also saves the changes [...] into a data frame."

"so much easier than manually coding." - M4

- M14

- M2



DISCUSSION



GENERATING CODE VS **PROVENANCE TRACKING**

Provenance better for most cases No code clutter Undo/redo Consistent semantics But code generation might be more robust works outside of notebooks works w/o the library

GENERATE CODE ON DEMAND [Beta]

Ś	\rightarrow	C	0 ×	87	72	74	E

Q Search						
#	Region ↑↓ ፡፡፡ : Yea	ar î↓ ∷ : <mark> </mark> ;Tri	igger ↑↓ ፡፡፡ : ¦;Wea	k Layer î↓ ःः : Der	Q	Root
3	Salt Lake	2012	Skier	Facets		Rename column ;Region to Re
4	Salt Lake	2012	Skier	New Snow	J J	Selected 1 point
5	Salt Lake	2012	Skier	Facets	7 7	Selected 2 points
6	Salt Lake	2012	Skier New	Snow/Old Snow Interface	🔶 🦕	Removed selected items
7	Salt Lake	2012	Skier	Facets		
8	Salt Lake	2012	Skier	Facets		Dataframes:
9	Salt Lake	2012	Skier	Facets		🗩 persist_df_1 🔂
10	Salt Lake	2012	Skier	Facets	6	
11	Salt Lake	2012	Skier	Facets		
12	Salt Lake	2012	Unknown	Ground Interface		
Dataframe na	ame					
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<pre> persist_c def creat df = # Add df.in df[" # Add </pre>	<pre>df_1 () : df_1 () : df.copy(deep=True) I "id_column" as the isert(0, "id_column", id_column"] = df["i I selection column</pre>	df.index + 1)	pe(str)			
<pre> persist_c def creat df = # Add df.in df[" # Add df.in df[" </pre>	<pre>df_1 (df): ce_persist_df_1(df): df.copy(deep=True) "id_column" as the sert(0, "id_column", id_column"] = df["i selection column selected"] = False</pre>	df.index + 1)	<pre>pe(str)</pre>			
<pre> persist_c def creat df = # Add df.in df[" # Add df.in df[" # Add df[" # Add </pre>	<pre>df_1 () : df_1 () : df.copy(deep=True) I "id_column" as the isert(0, "id_column", id_column"] = df["i I selection column</pre>	df.index + 1) d_column"].asty				
<pre> persist_c def creat df = # Add df.in df[" # Add df.in df[" # Add df[" # Add df[" # Ren df = </pre>	<pre>df_1 () : df_1 () : df.copy(deep=True) "id_column" as the sert(0, "id_column", id_column"] = df["i selection column selected"] = False ame column</pre>	df.index + 1) d_column"].asty	• })			
<pre> persist_c def creat df = # Add df.in df.in # Add df.in # Add df.in df["</pre>	<pre>df_1 (df): ce_persist_df_1(df): df.copy(deep=True) "id_column" as the sert(0, "id_column", id_column"] = df["i selection column selected"] = False name column df.rename(columns={';F</pre>	<pre>df.index + 1) d_column"].asty Region': 'Region sin(['2']), "s</pre>	<pre>'}) selected"] = True</pre>			



TRY OUT PERSIST!

Persist is available now! https://vdl.sci.utah.edu/persist/ Documentation & examples

Feedback / bug reports appreciated!

Getting Started

Requirements

- pandas >= 0.25
- altair >= 5
- ipywidgets
- anywidget

Install

To install the extension, execute:

pip install persist_ext

If the Jupyter server was already running, you might have to reload the browser page and restart the kernel.

Uninstall

To remove the extension, execute:

pip uninstall persist_ext

Usage

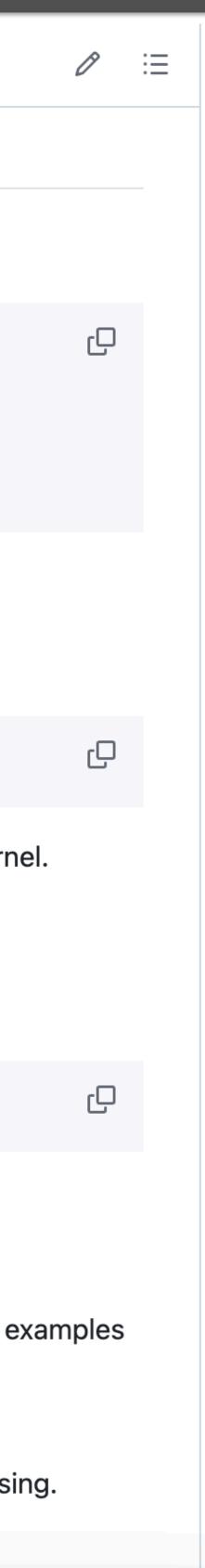
Persist supports two types of interactive outputs — a custom data table and Vega-Altair (>=5.0.0, see requirements and caveats) charts. The following examples will walk you through creating each one. The examples are also available as notebooks in the examples folder of the repository. Each section will link to the corresponding notebook as well as a binder link for the notebook.

Persist currently works with pandas dataframes, so load/convert the data to pandas dataframe before using.

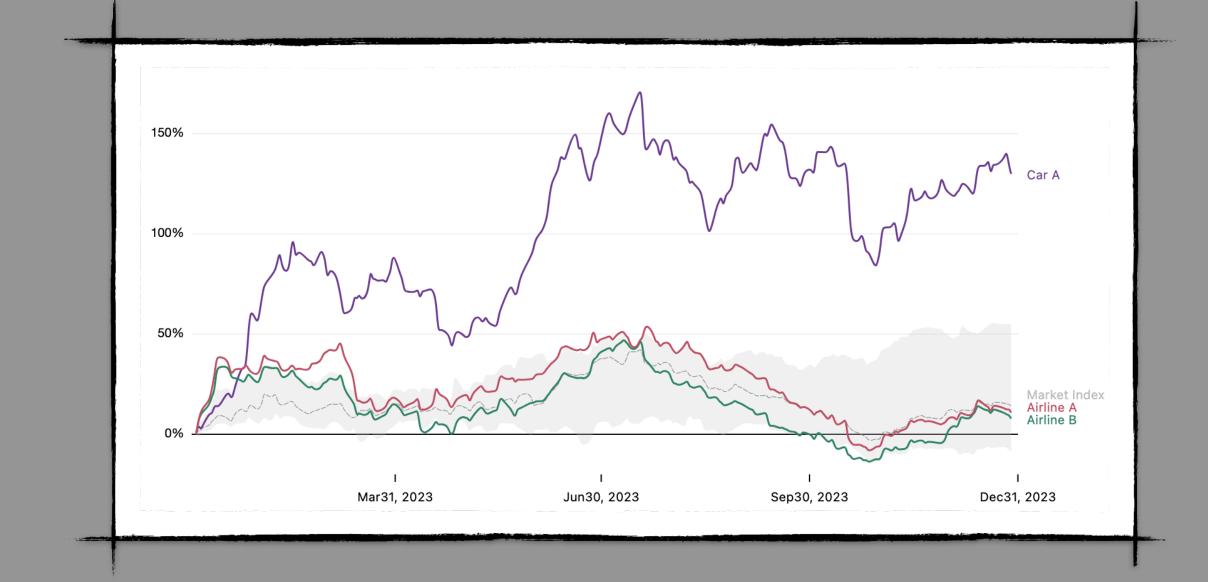
Examples

▲ BSD-3-Clause license

```
– JupyterLab >= 4.0.0 or Jupyter Notebook >= 7.0.0
```



EMPIRICAL & THEORETICAL WORK





HOW PEOPLE ACTUALLY LIE WITH CHARTS... ... AND WHAT WE CAN DO ABOUT IT



Max Lisnic, Marina Kogan

[CHI 2023 & CHI 2024]

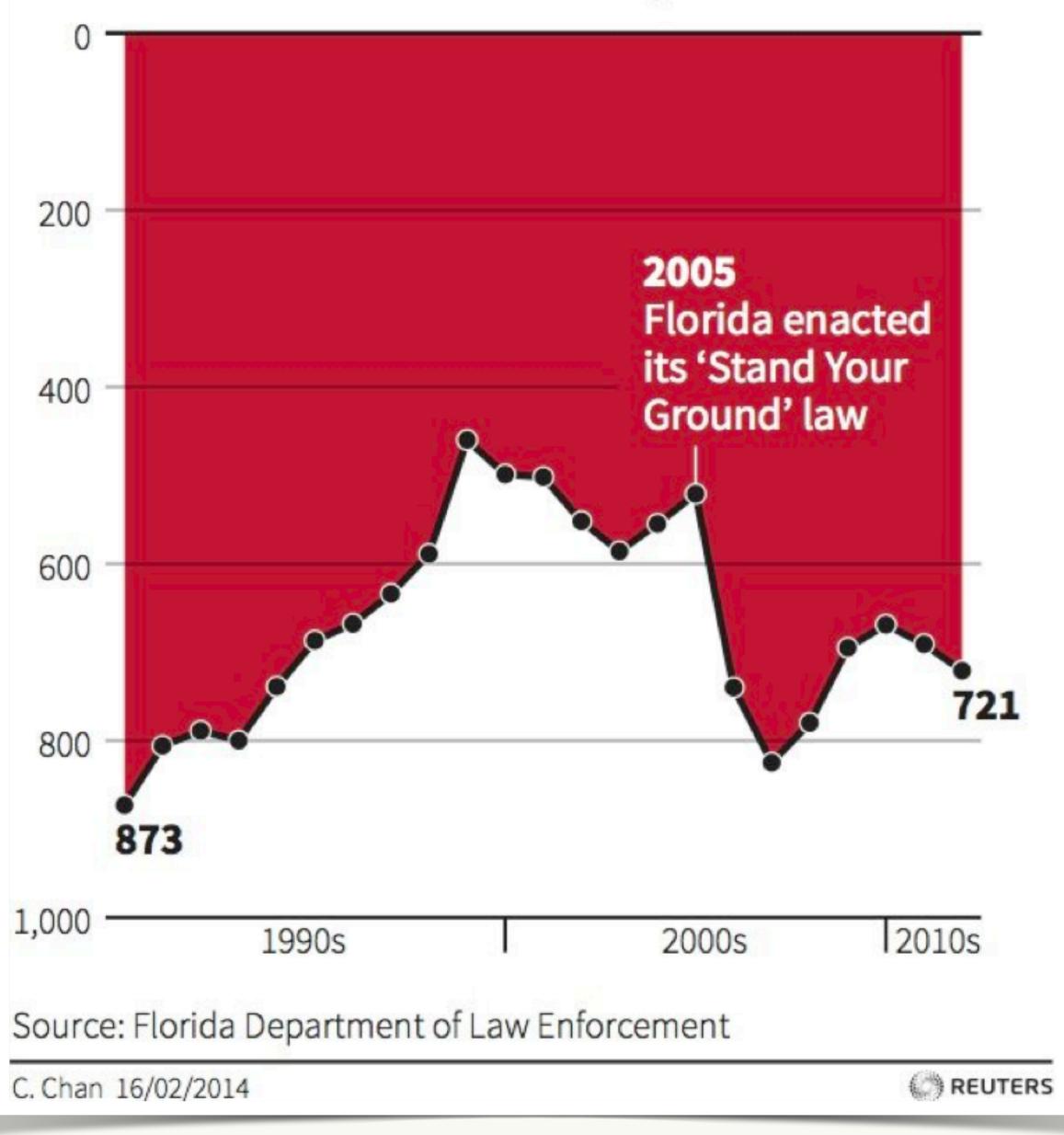




HOW WE THINK PPL LIE WITH CHARTS: VIOLATIONS OF VISUALIZATION DESIGN GUIDELINES

Gun deaths in Florida

Number of murders committed using firearms

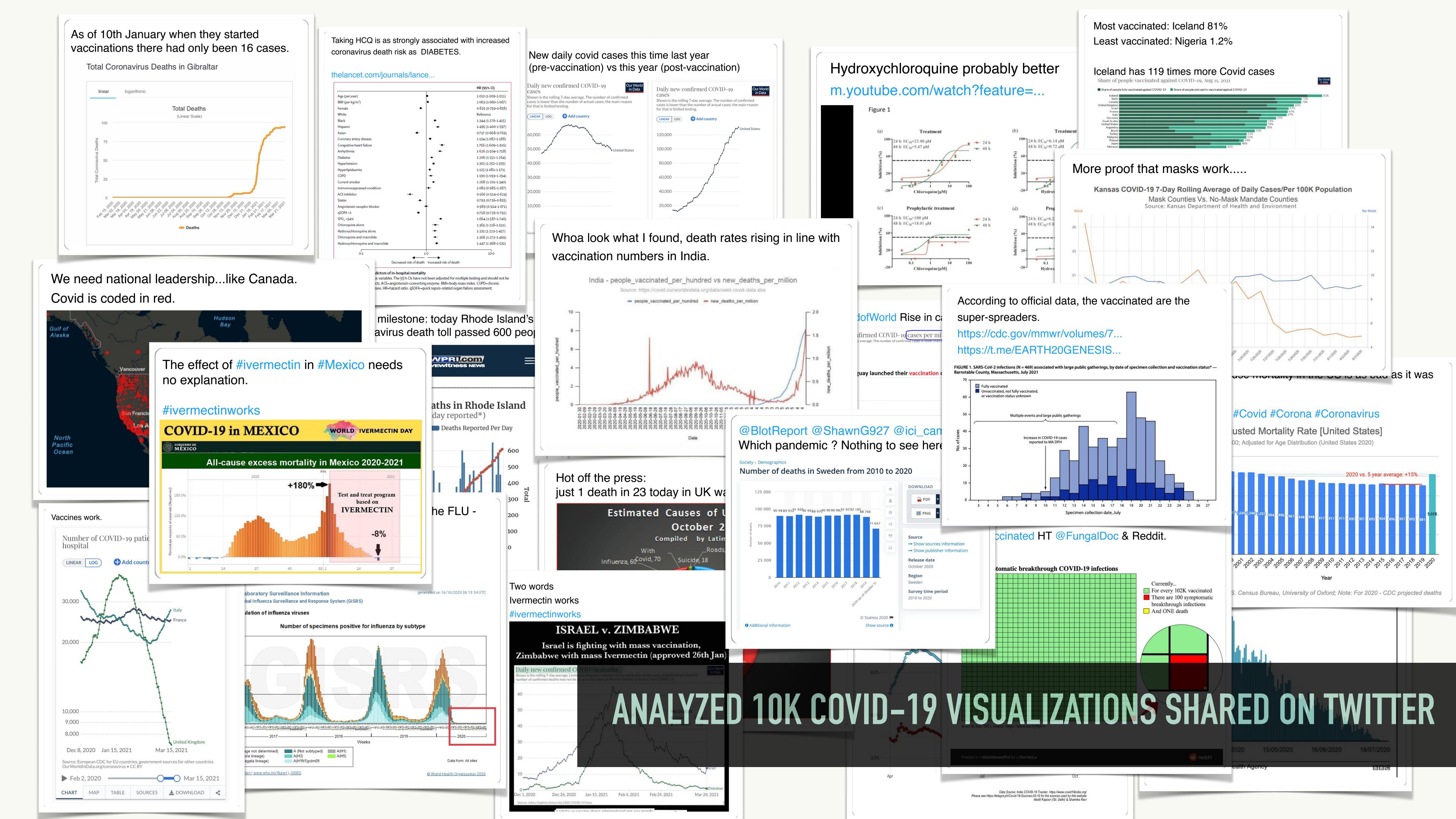


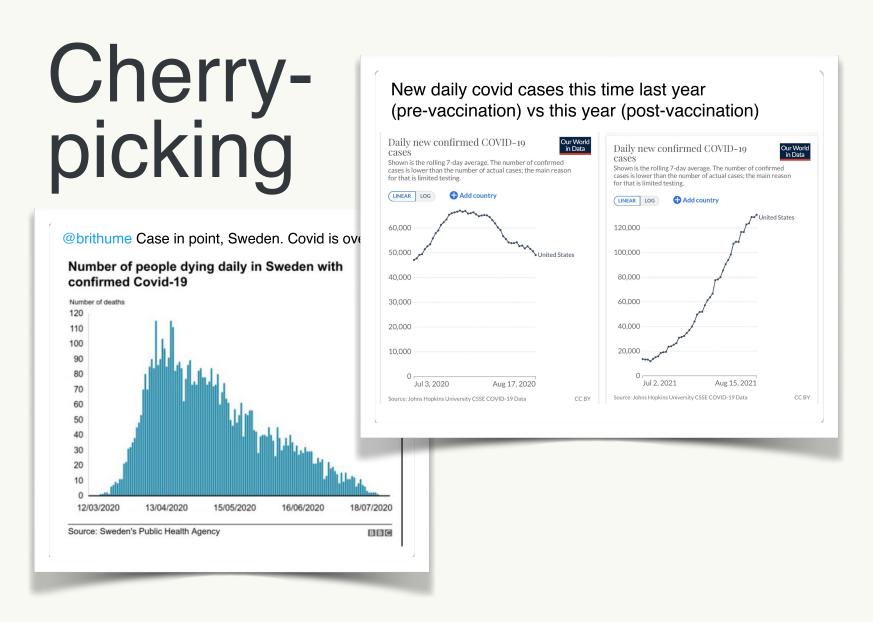
Inverted y axis



Truncated y axis

BUT IS THAT REALLY HOW PEOPLE LIE WITH CHARS?





We need national leadership...like Canada. Covid is coded in red.

Incorrect reading



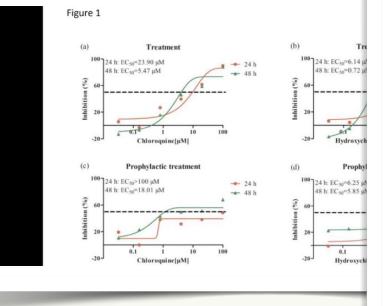
More proof that masks work..... Kansas COVID-19 7-Day Rolling Average of Daily Cases/Per 100K Population Mask Counties Vs. No-Mask Mandate Counties Source: Kansas Department of Health and Environmen 1107000 1107000 1107000 1107000 1107000 1107000 1107000 100700 107000

Arbi	itra	Y
thre	shc	bĺd

esnoid	The all-cause mortality in the US is as bad as it was in 2003!
t off the press: the 1 death in 23 today in UK was 'with Covid' Stimated Causes of UK Deaths on 7 October 2020 Diabetes, 70 United by Latimer Alder United Stroke, 100 Diabetes, 70 United Stroke, 100 Diabetes, 70 Stroke, 100 Diabetes, 70 Cancers, 400 As of 10th January when they stavaccinations there had only been Total Coronavirus Deaths in Gibraltar	
linear logarithmic Total Deaths (Linear Scale) 100 50 50 50 50	eden from 2010 to 2020

Misrepresentation of scientific results

Hydroxychloroquine probably better m.youtube.com/watch?feature=.



Taking HCQ is as strongly associated with increased coronavirus death risk as DIABETES.

thelancet.com/journals/lance

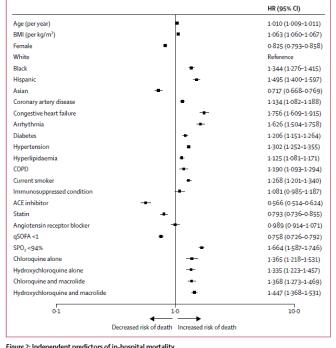
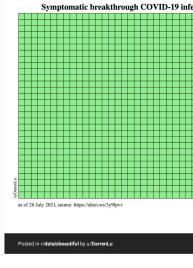
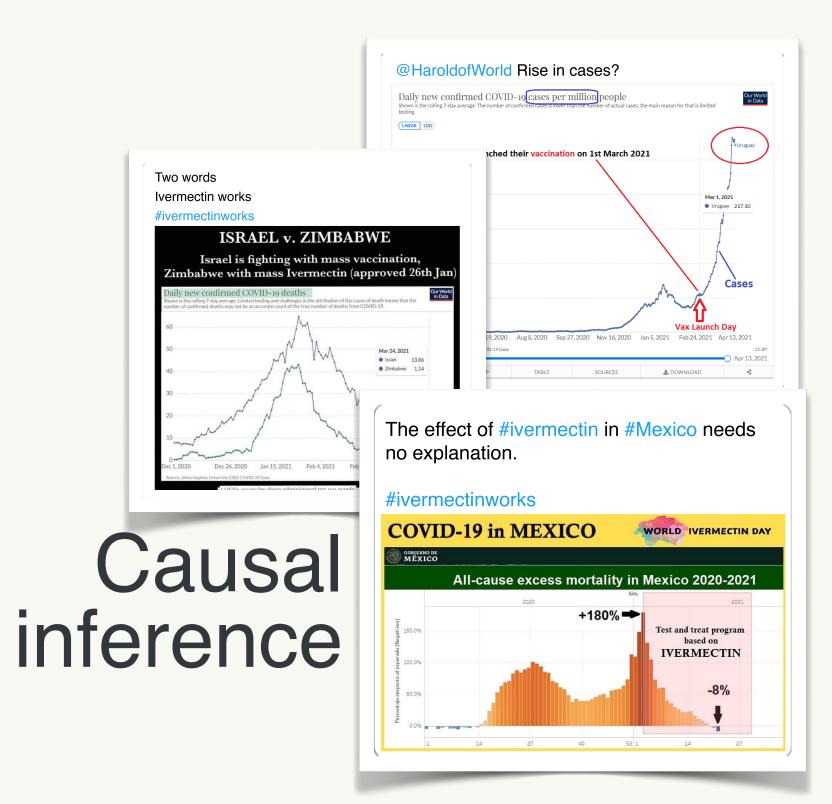


Figure 2: Independent predictors of in-hospital mortality Age and BMI are continuous variables. The 95% CIs have not been adjusted for Age and BMI are continuous variables. The 95% CIs have not been adjusted for multiple testing and should not be used to infer definitive effects. ACE=angiotensin-converting enzyme. BMI=body mass index. COPD=chronic obstructive pulmonary disease. HR=hazard ratio. qSOFA=quick sepsis-related organ failure assessment. PO₂=oxygen saturation

Statistical

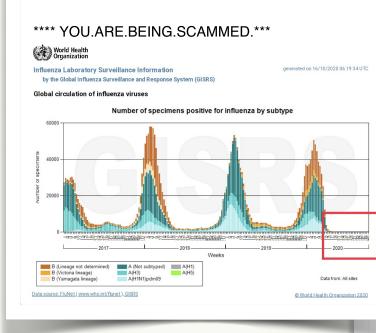
Make this go viral. Vaccines work. #GetVaccinated HT @FungalDoc & Reddit. Symptomatic breakthrough COVID-19 infections



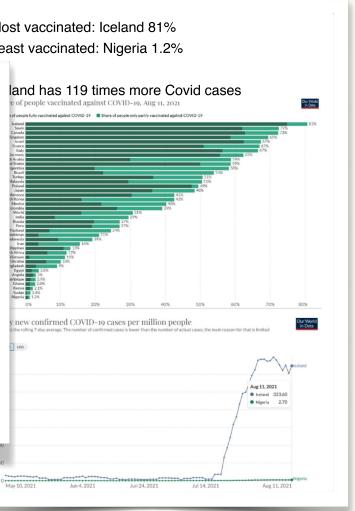


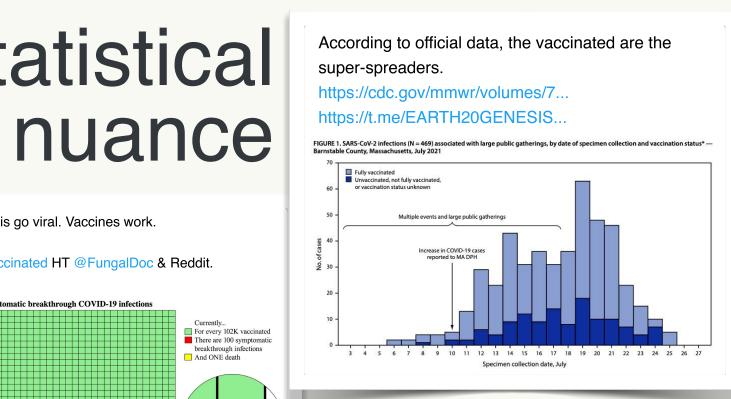
Issues with data validity

WOW, the CHINA VIRUS has eradicated the FLU -Amazing!!



Most vaccinated: Iceland 81% Least vaccinated: Nigeria 1.2%





🌀 reddit

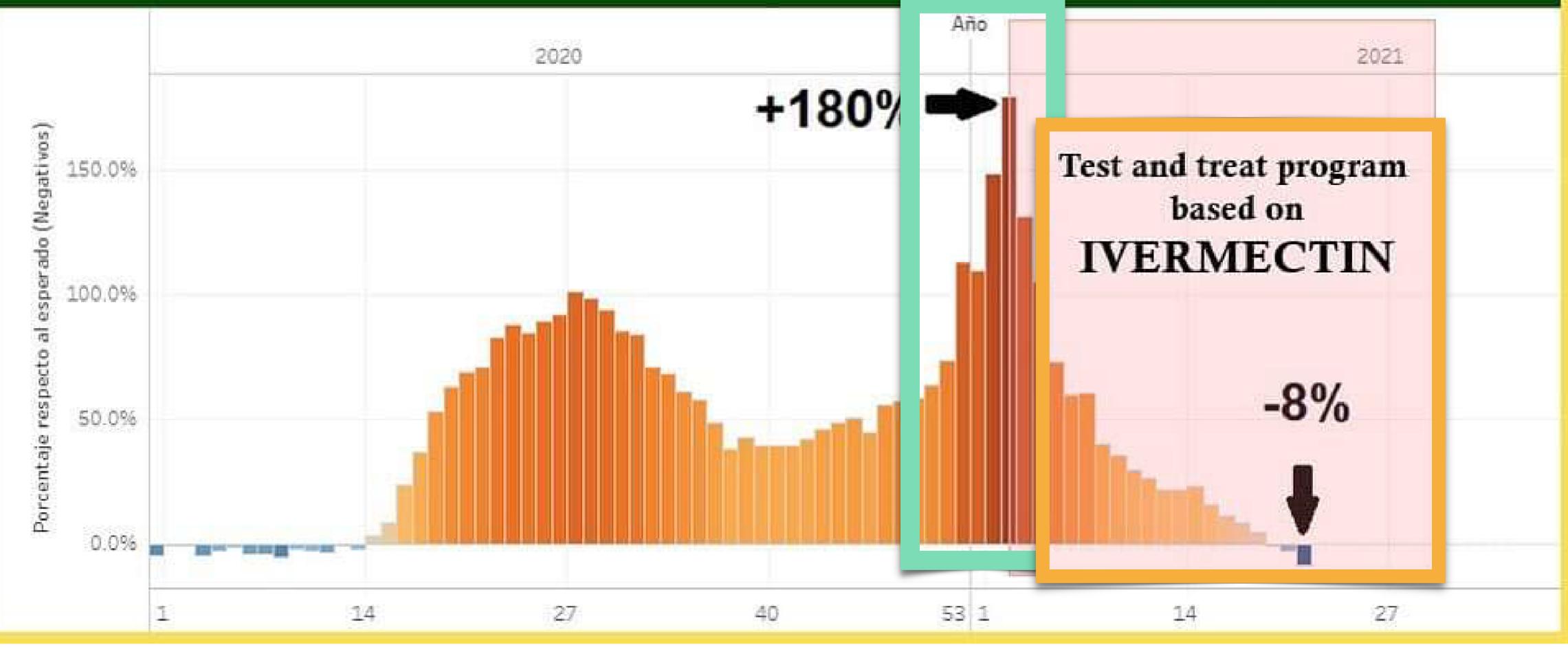
The effect of *#ivermectin* in *#Mexico* needs no explanation.

#ivermectinworks

gobierno de MÉXICO

COVID-19 in MEXICO

All-cause excess mortality in Mexico 2020-2021





WORLD IVERMECTIN DAY

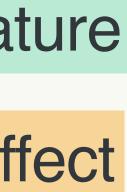
Causal inference

Screenshot of a government dashboard

Sharp mortality drop is a salient visual feature

Added annotation assigns a cause-and-effect

Tweet text further explains the argument

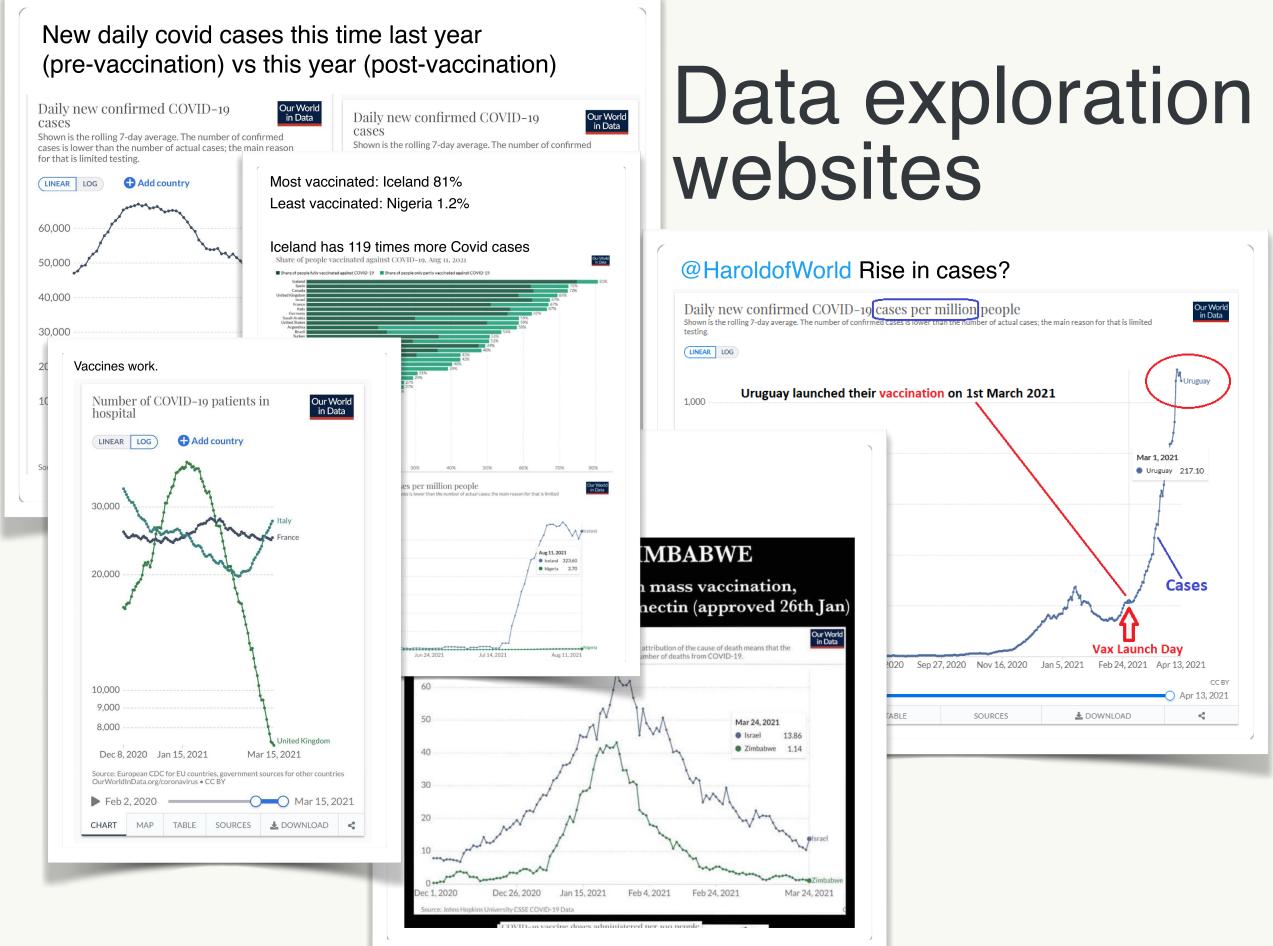


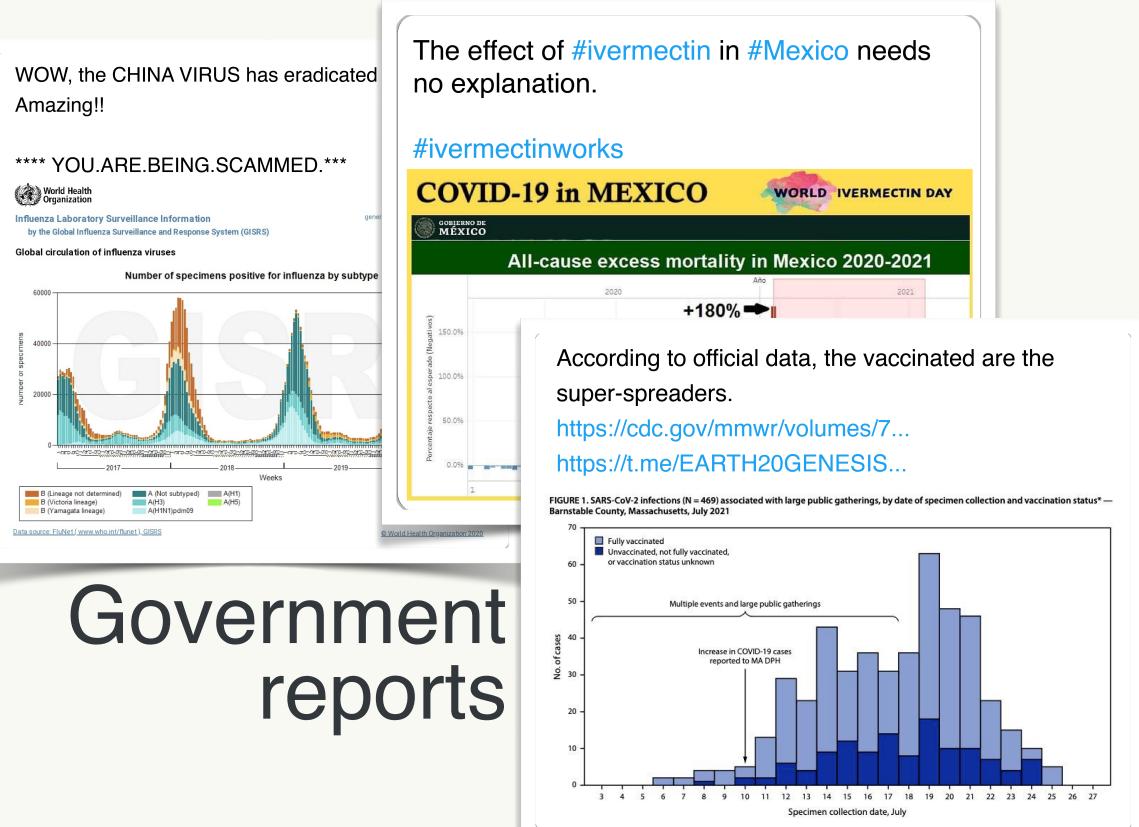
TAKEAWAYS

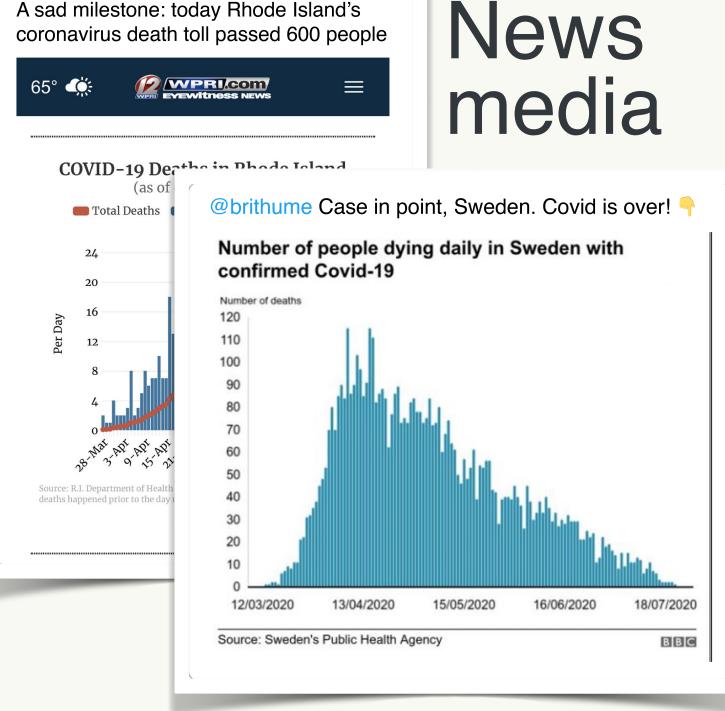
89% of charts with reasoning errors do not violate any visualization guidelines

The majority of **misleading charts** are screenshots from **reputable sources**

Sources of Deceptive Visualizations

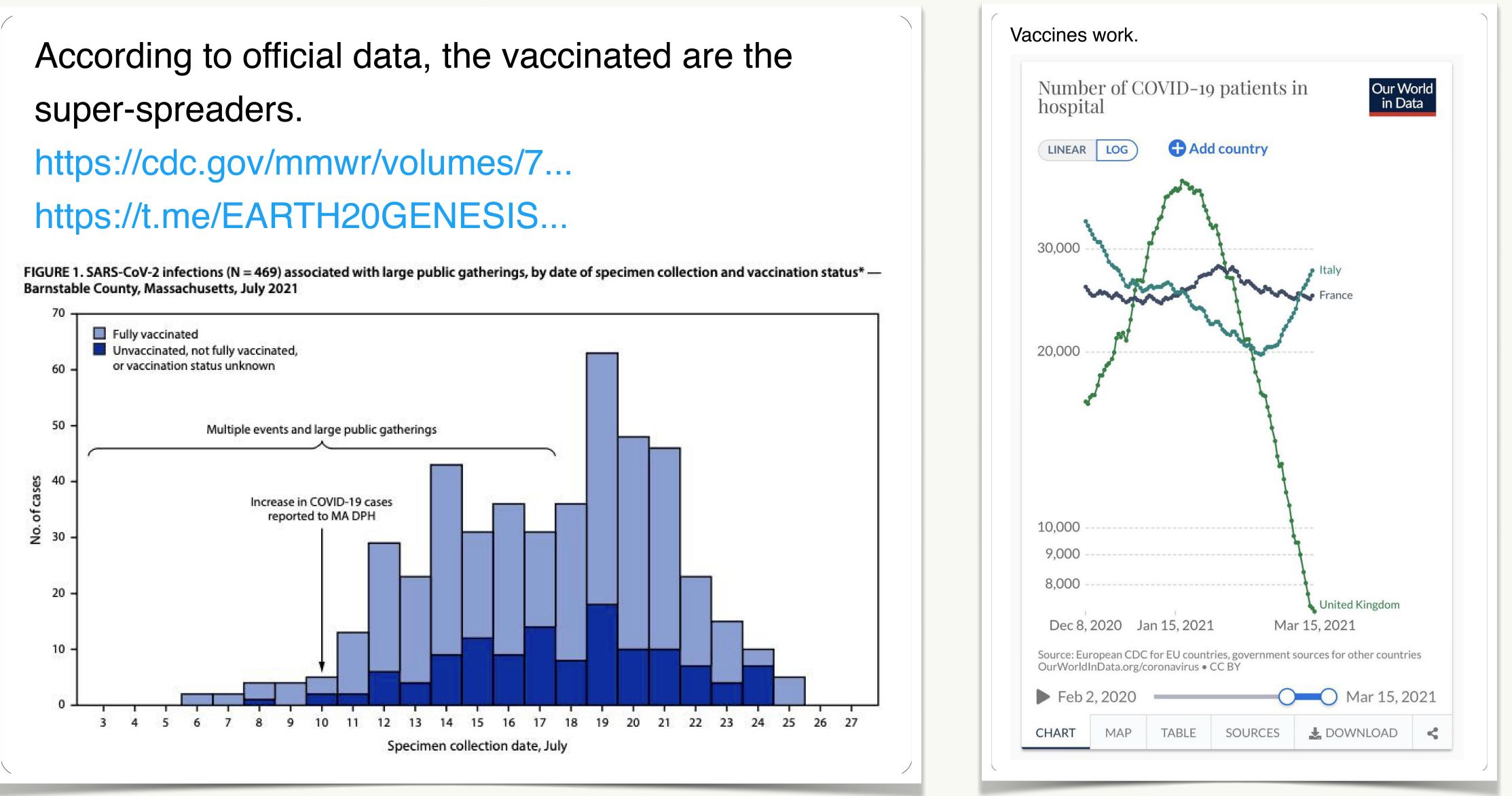






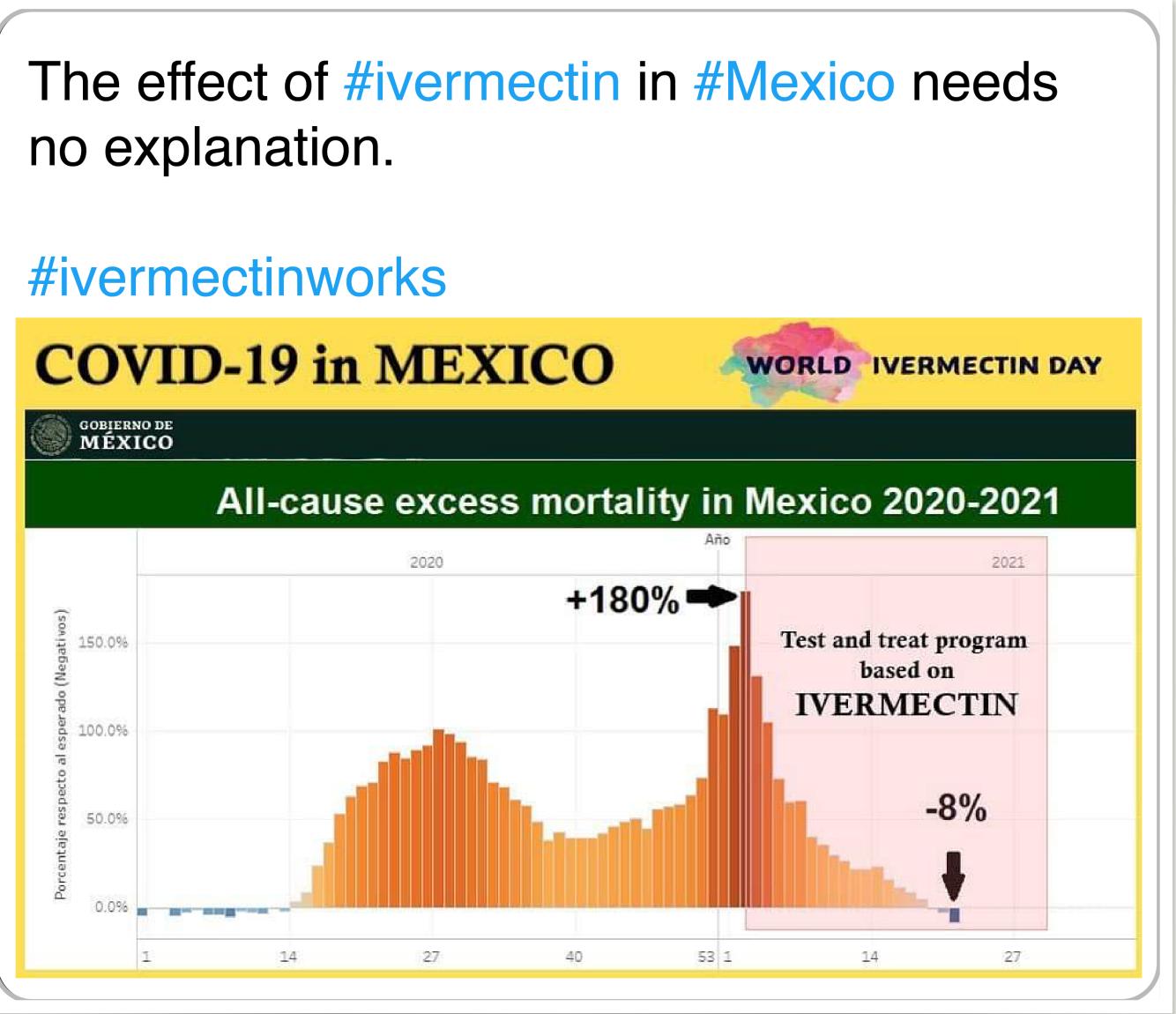
A sad milestone: today Rhode Island's

VULNERABLE VISUALIZATIONS



Notes outside the chart

Unrestricted interaction



Added annotations

PROTECTING VISUALIZATIONS

Can we design data exploration tools that prevent cherry-picking while maintaining freedom of exploration?

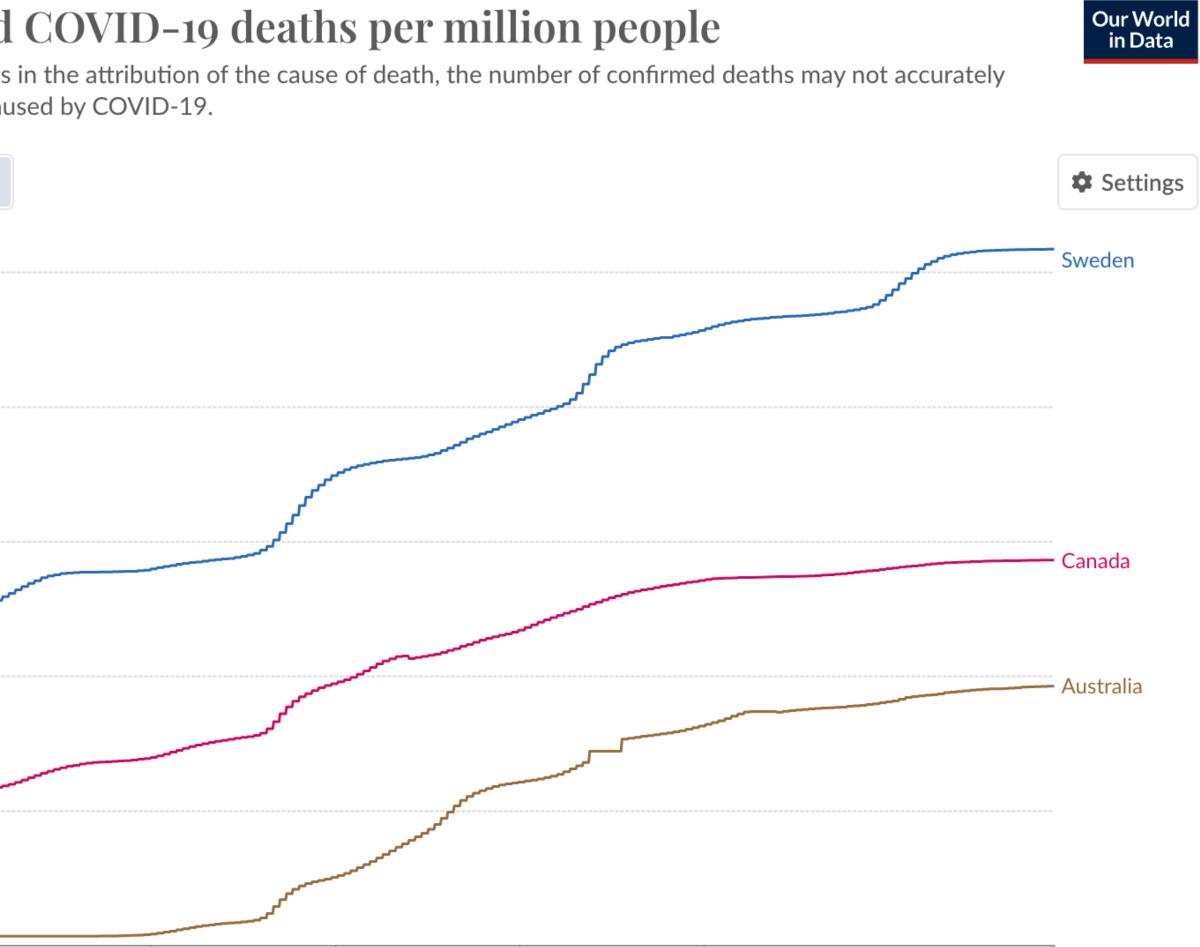
COVID-19 Data Explorer	METRIC
Explore global data on COVID-19. Download this dataset	Confirmed deaths v
Q Type to add a country or reg Sort by Country name V	Cumulative confirmed Due to varying protocols and challenges represent the true number of deaths cau
Australia	🖽 Table 🚯 Map 🗠 Chart
Canada	0.500
Sweden	2,500
Afghanistan	2,000
	2,000
🗆 Albania	1,500
□ Algeria	
American Samoa	1,000
□ Andorra	کر ک
Angola	500
Anguilla	d and a second design of the s
Antigua and Barbuda	0 Mar 8, 2020 Feb 24, 202
□ Argentina	Play time-lapse Mar 8, 2020
	Data source: WHO COVID-19 Dashboard
Aruba	CC BY
× Clear selection	

ITERVAL

Cumulative

Relative to Population

Color by test positivity





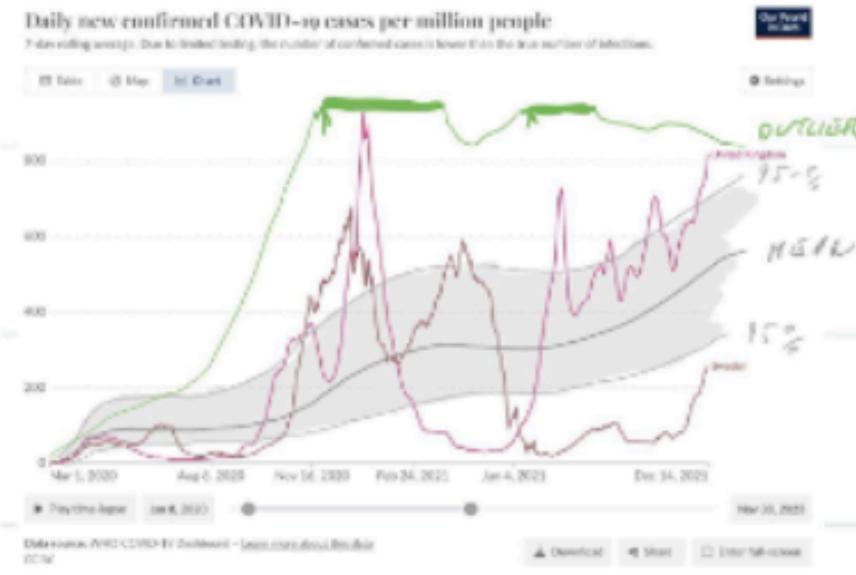
Related: Since 8 March, we rely on data from the WHO for confirmed cases and deaths



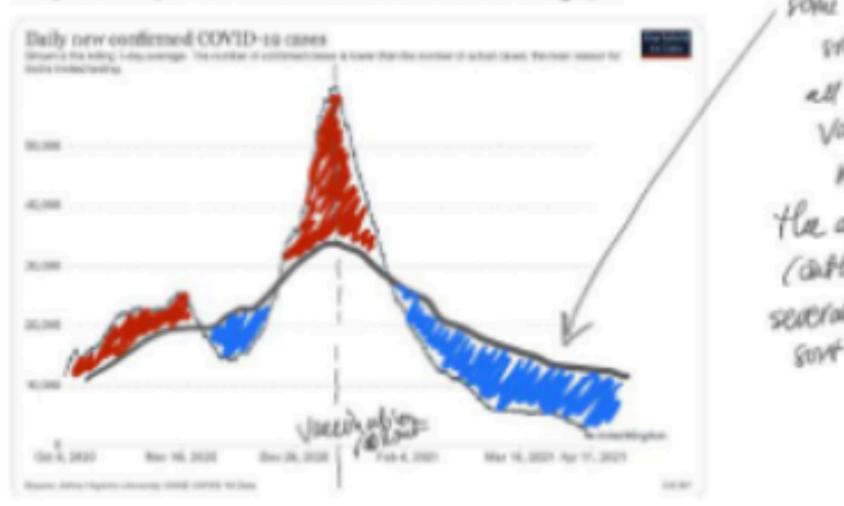


DESIGNING GUARDRAILS

Parallel Design by 4 Authors **Cheery-Picking in Temporal Charts** (Stocks, COVID)



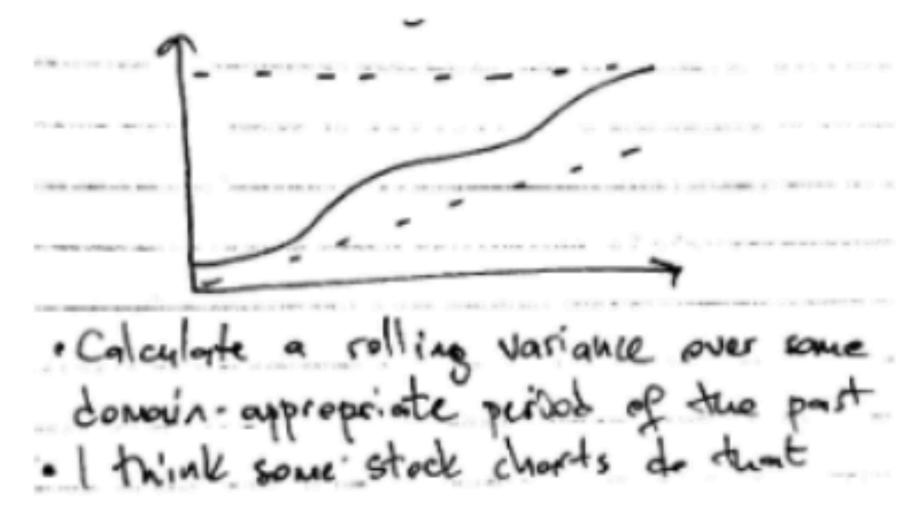
The positive impact of the UK's vaccination efforts in one graph



and the second s background (peint drifty?) · Denselines (Moritz & Fisher) perrint

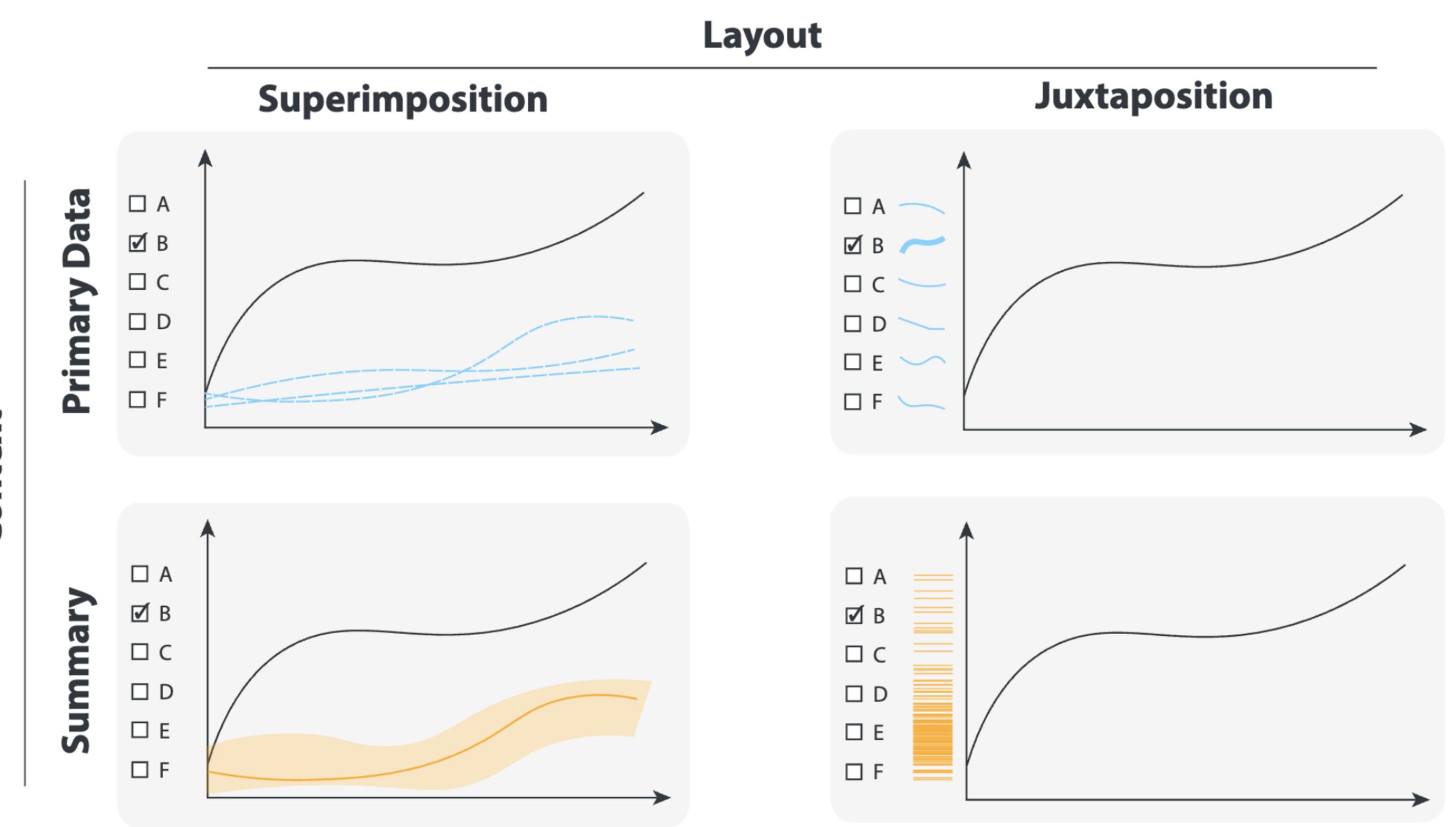
some sort of weighted all aligned around their Vaccination rollant date nobel UK's in relative terus the ang is also very problematic

sout of all except



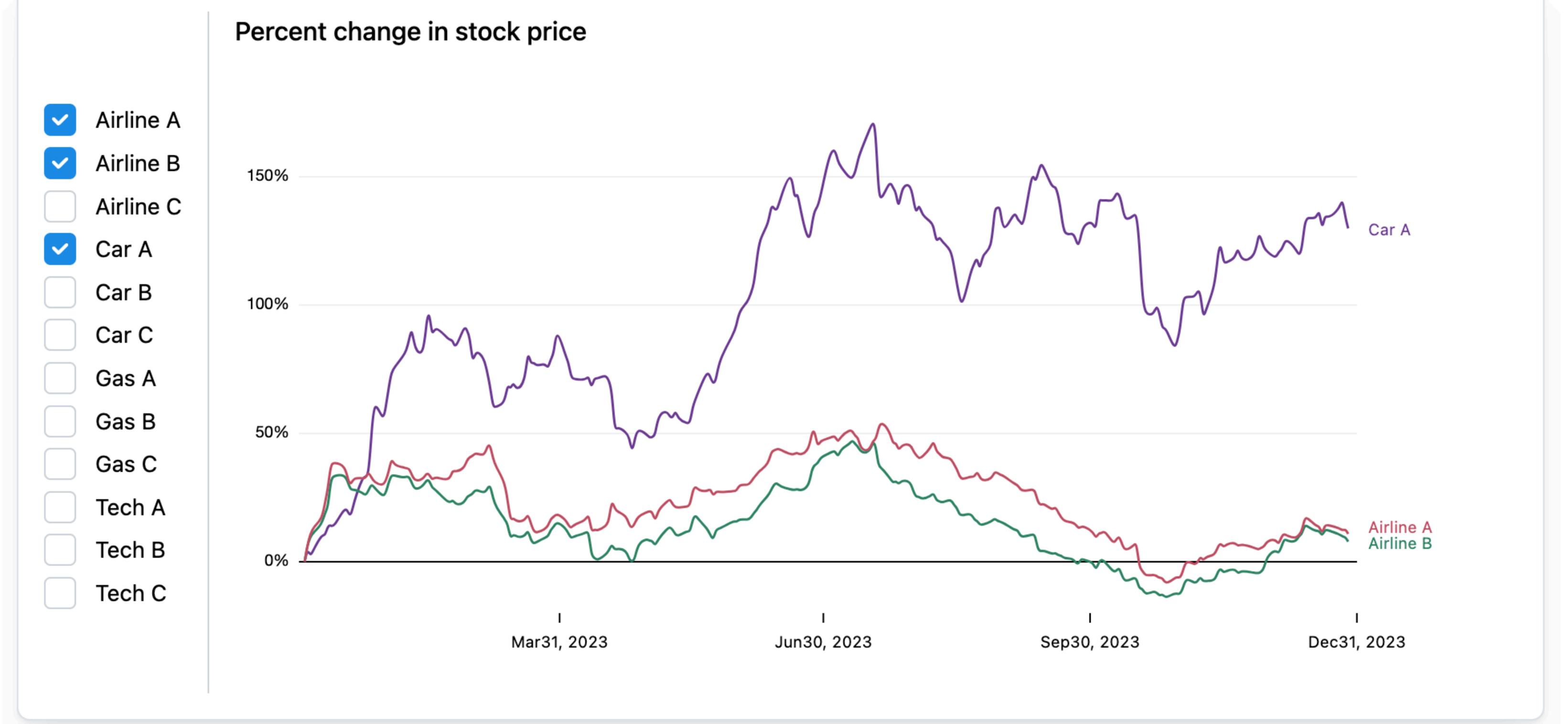


DESIGN SPACE

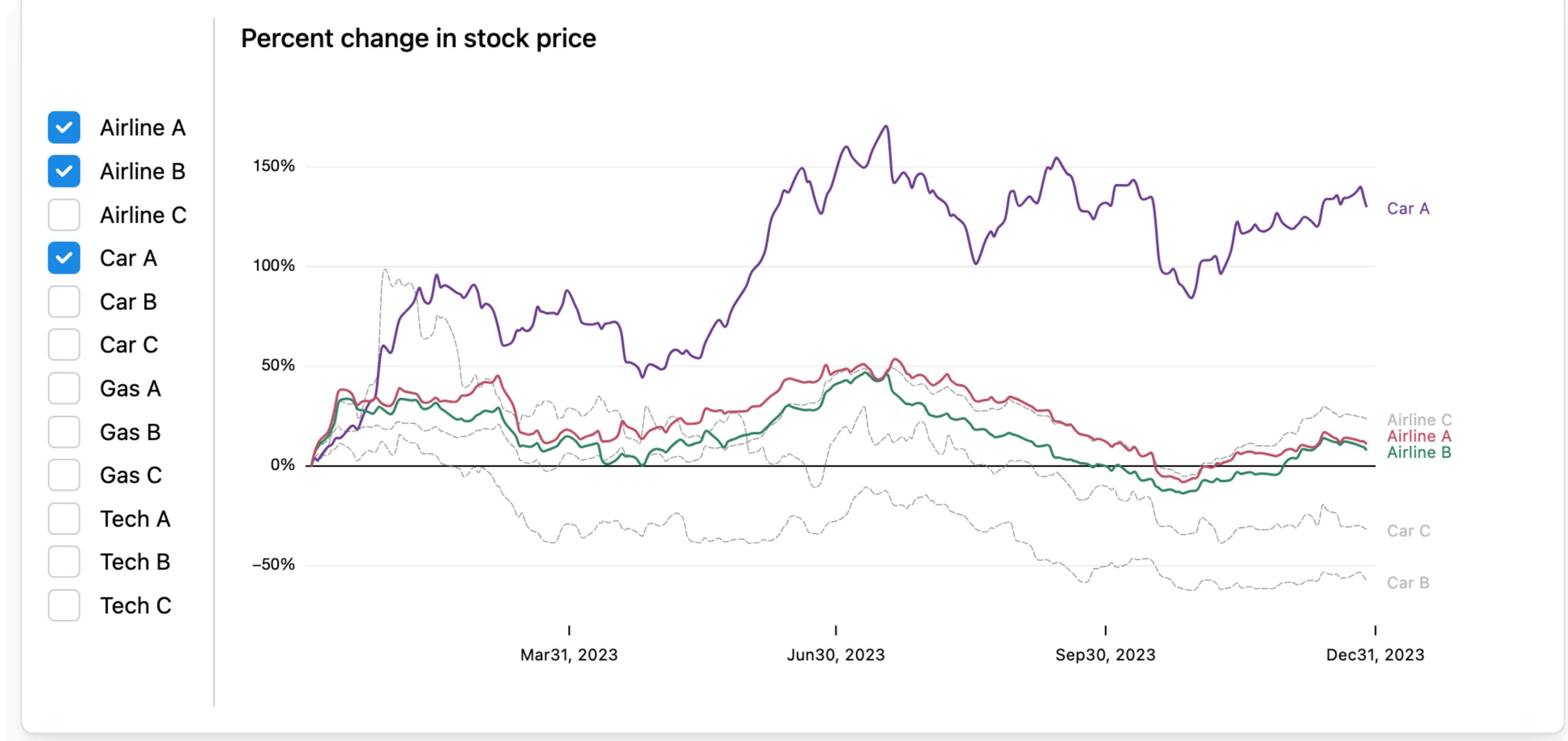


Context

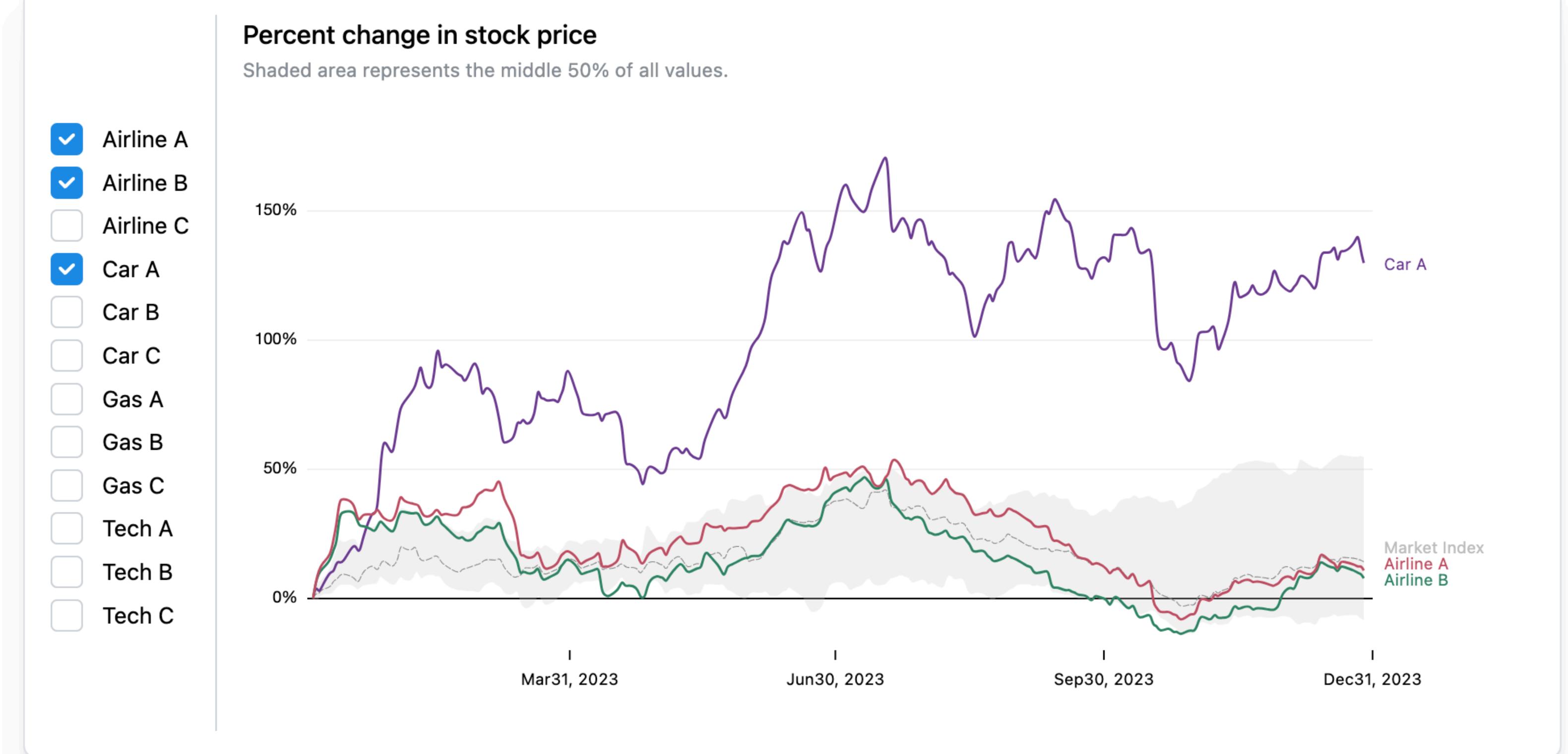
CHERRYPICKING: NO GUARDRAILS



ENFORCING CONTEXT - SUPERIMPOSITION / PRIMARY DATA



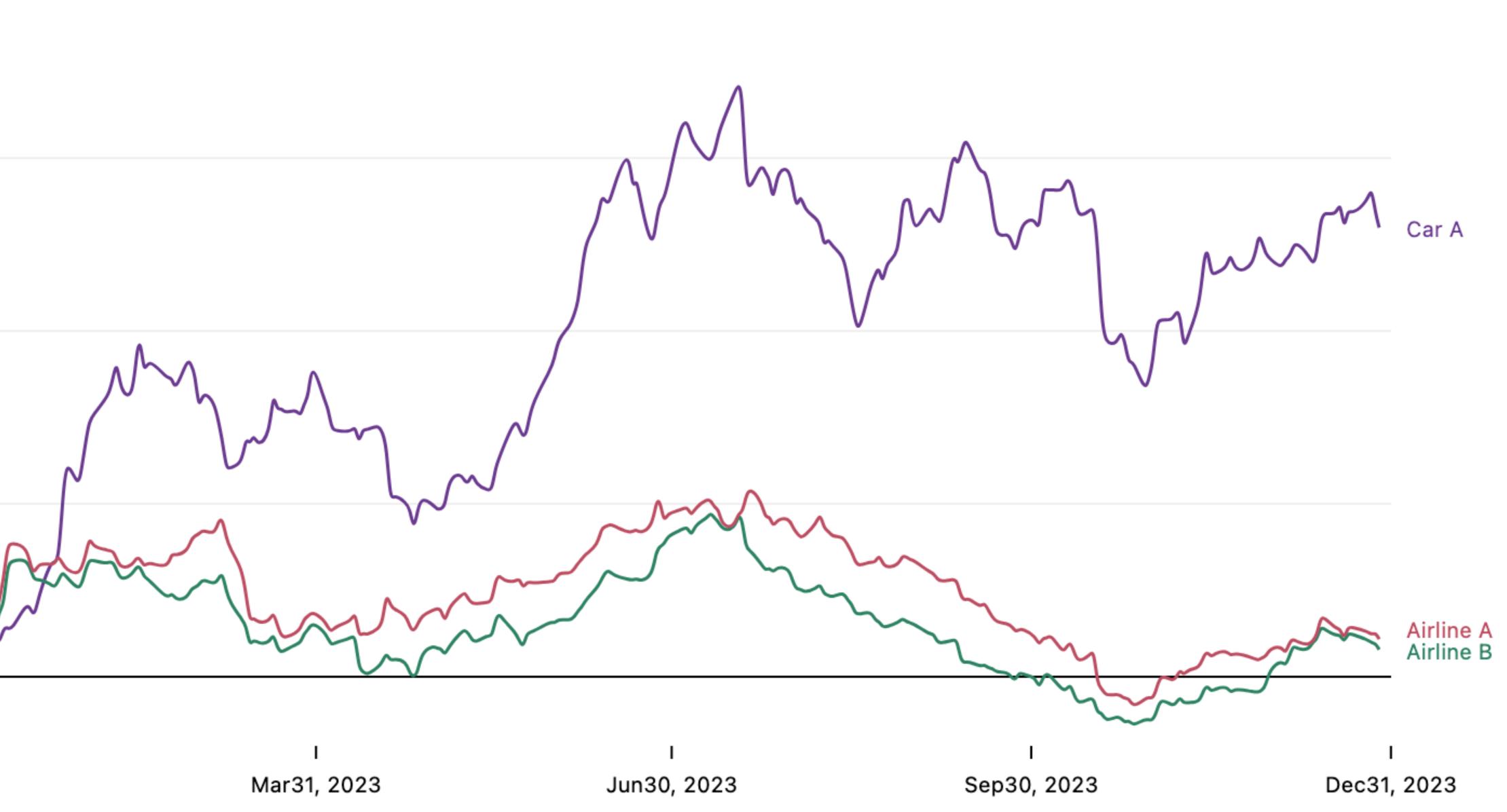
SHOWING STATISTICAL CONTEXT – SUPERIMPOSITION / SUMMARY

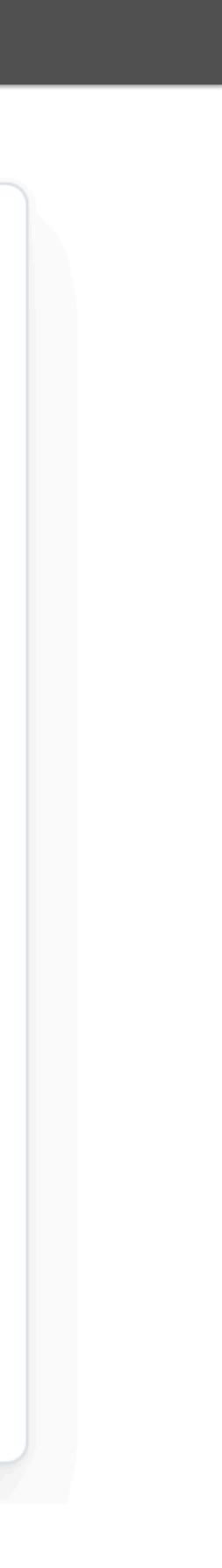


	Percent
Airline A Airline B	150%
Airline C Car A	
Car B	 100%
Car C Gas A	
Gas B Gas C	 50%
Tech A Tech B	 0%
Tech C	

CONTEXT OUTSIDE - JUXTAPOSITION / PRIMARY DATA

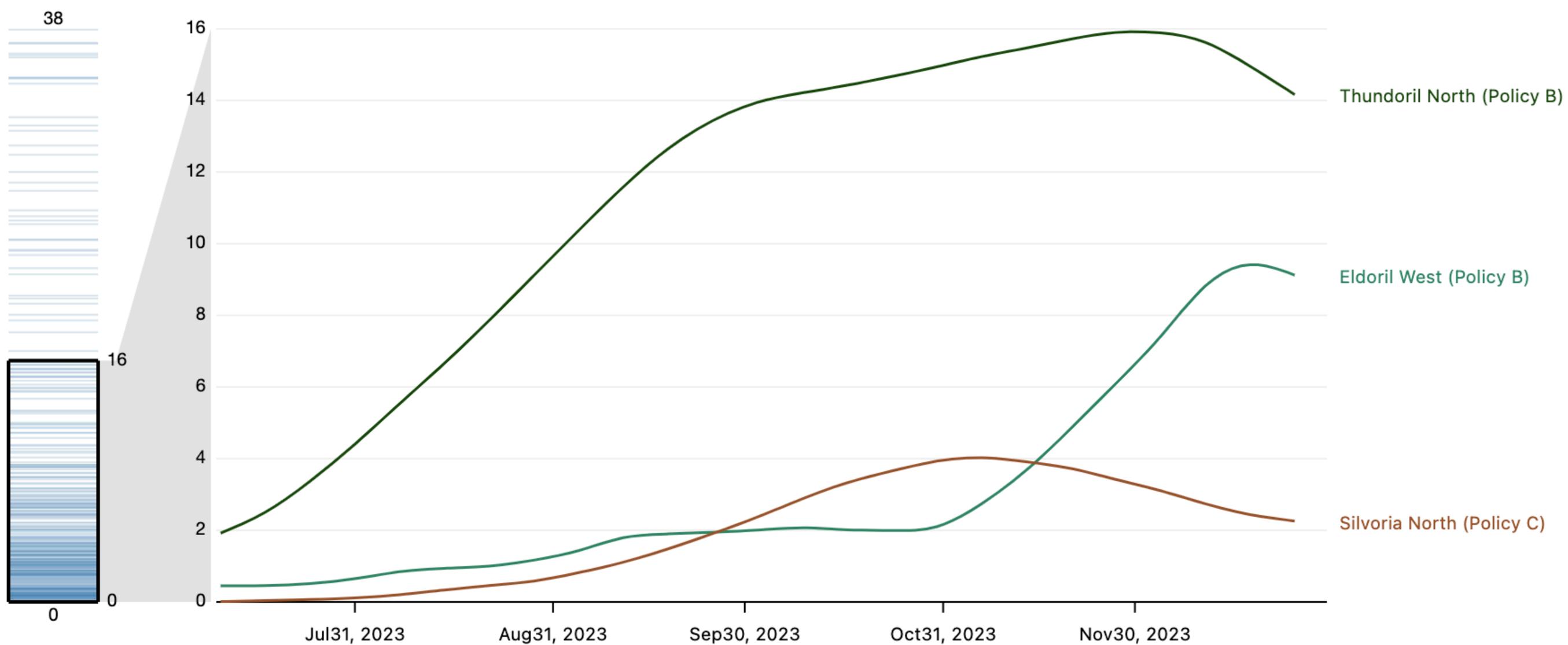
change in stock price





Infections per million people

Bar on the left highlights the range of selection among all data.



SUMMARY CONTEXT – JUXTAPOSITION / SUMMARY

EVALUATION

Red-Team Study (Production): Challenge participants to deceive w and w/o guardrails

Validation Study:

Measure how deceptive charts are

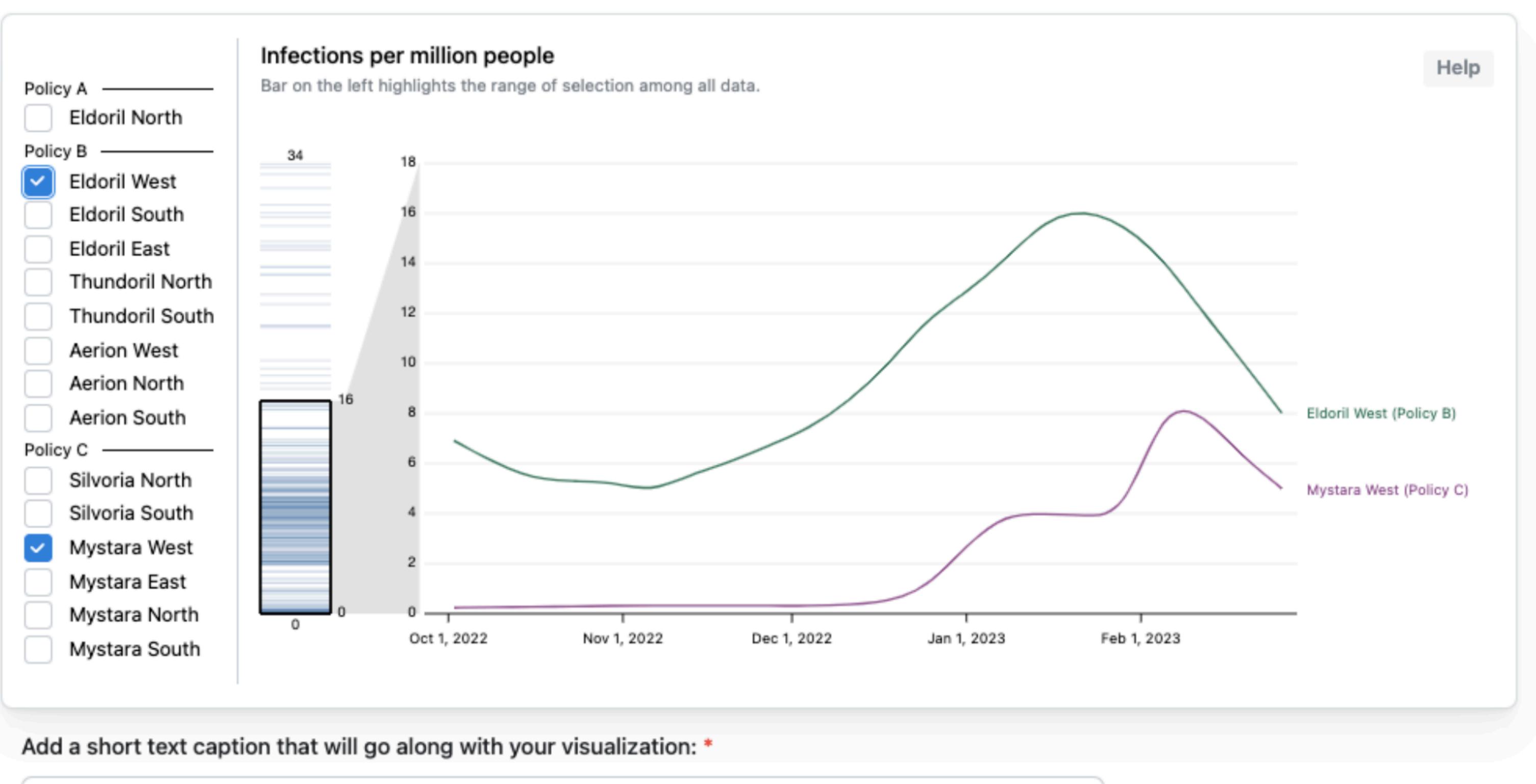
RED-TEAM SCENARIOS Viral Disease & Stocks

Viral Scenario: the most effective policy.

imagine that you work for a public health agency and are in charge of promoting a health policy campaign against a viral disease that, judging by the data alone, does not appear to be

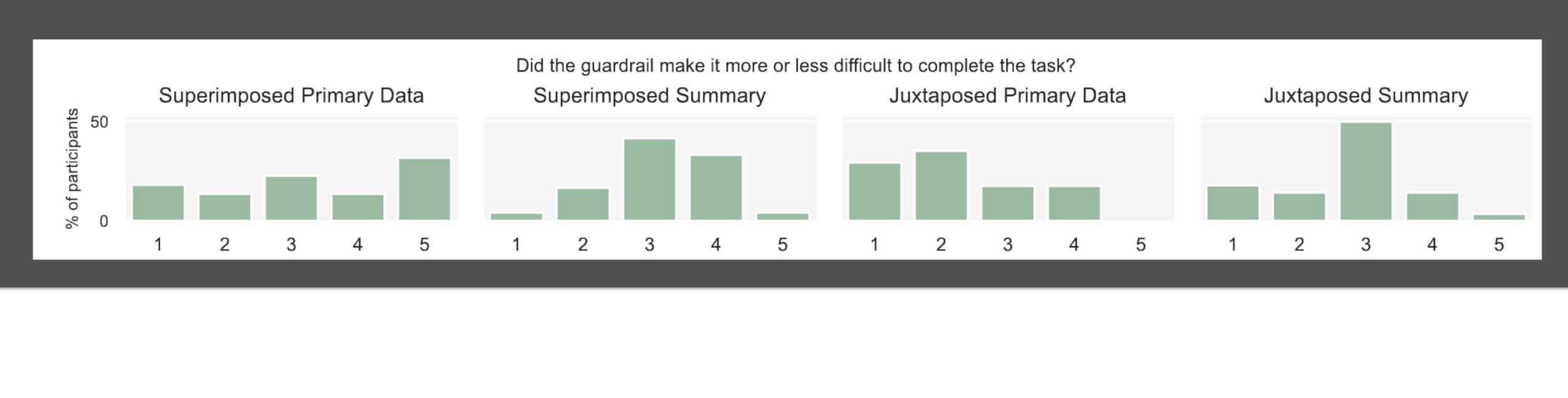


STIMULUS

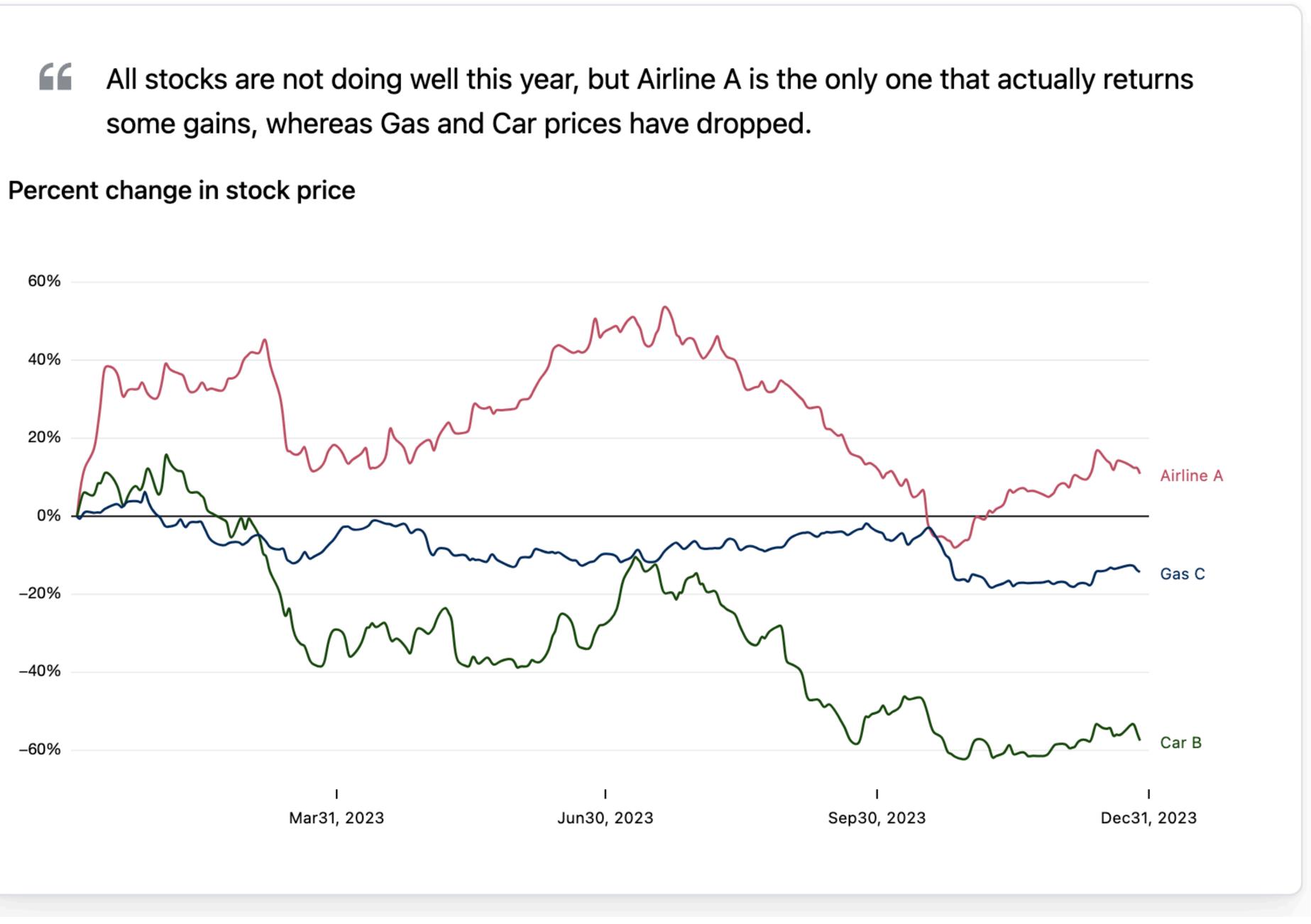


RESULTS

Juxtaposed Primary is a **Cherry-Picking Machine** Superimposed Primary makes it harder to cherrypick Others show little effect didn't get it?



some gains, whereas Gas and Car prices have dropped.

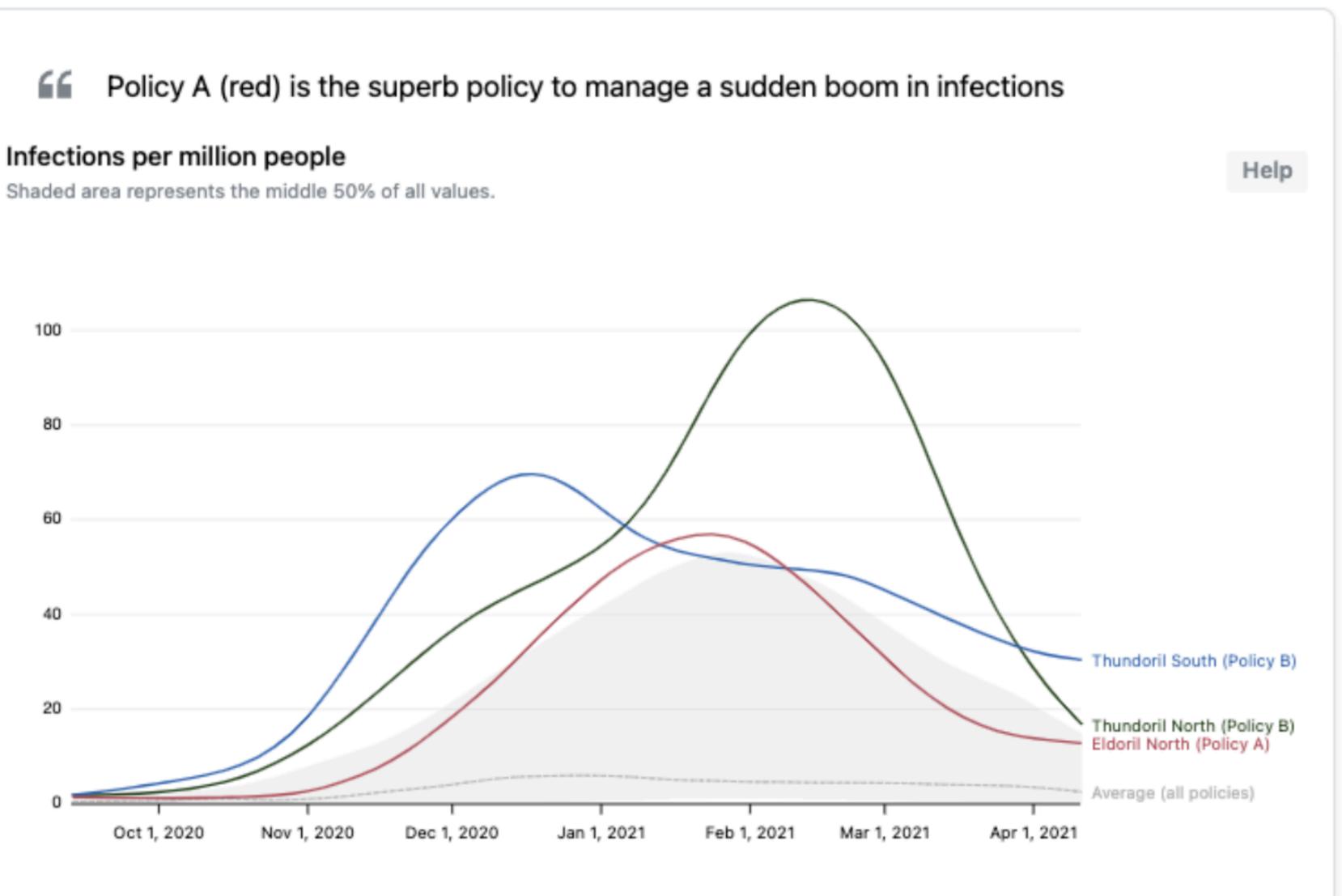


VALIDATION STUDY **Do Guardrails Work?**

"Investment" task: make monetary decision based on information

You will need to travel to Eldoril North (Policy A) for work. You've come across this visualization and the accompanying caption. Please review the visualization and the caption, and then answer the questions below based solely on this information. (Please try to not rely on other visualizations you've seen.)

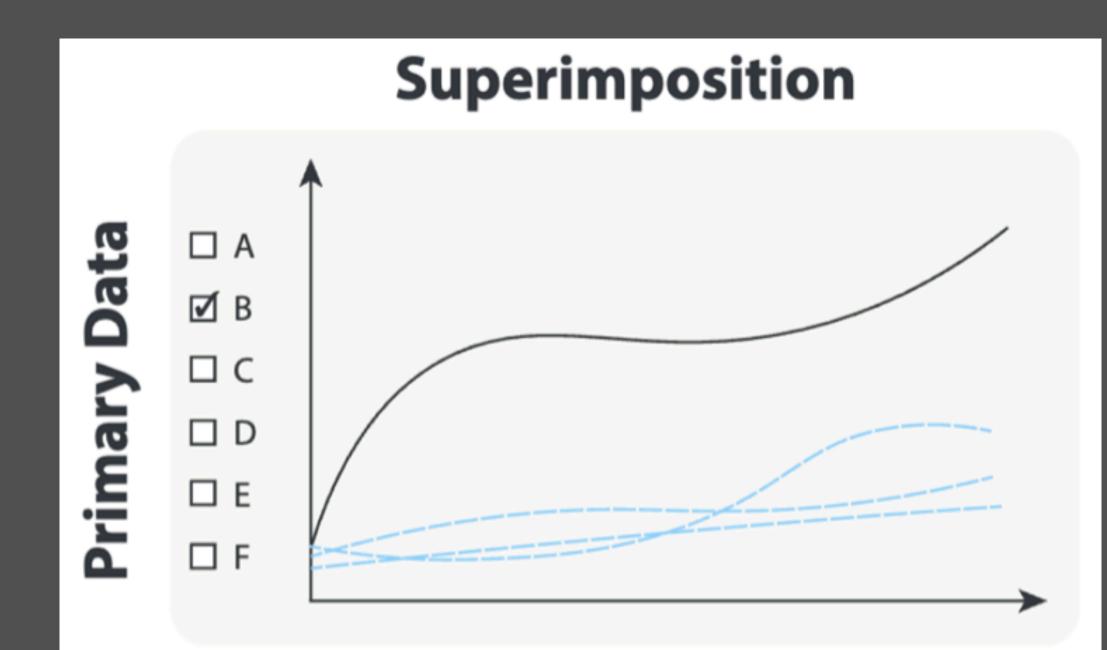
Shaded area represents the middle 50% of all values.



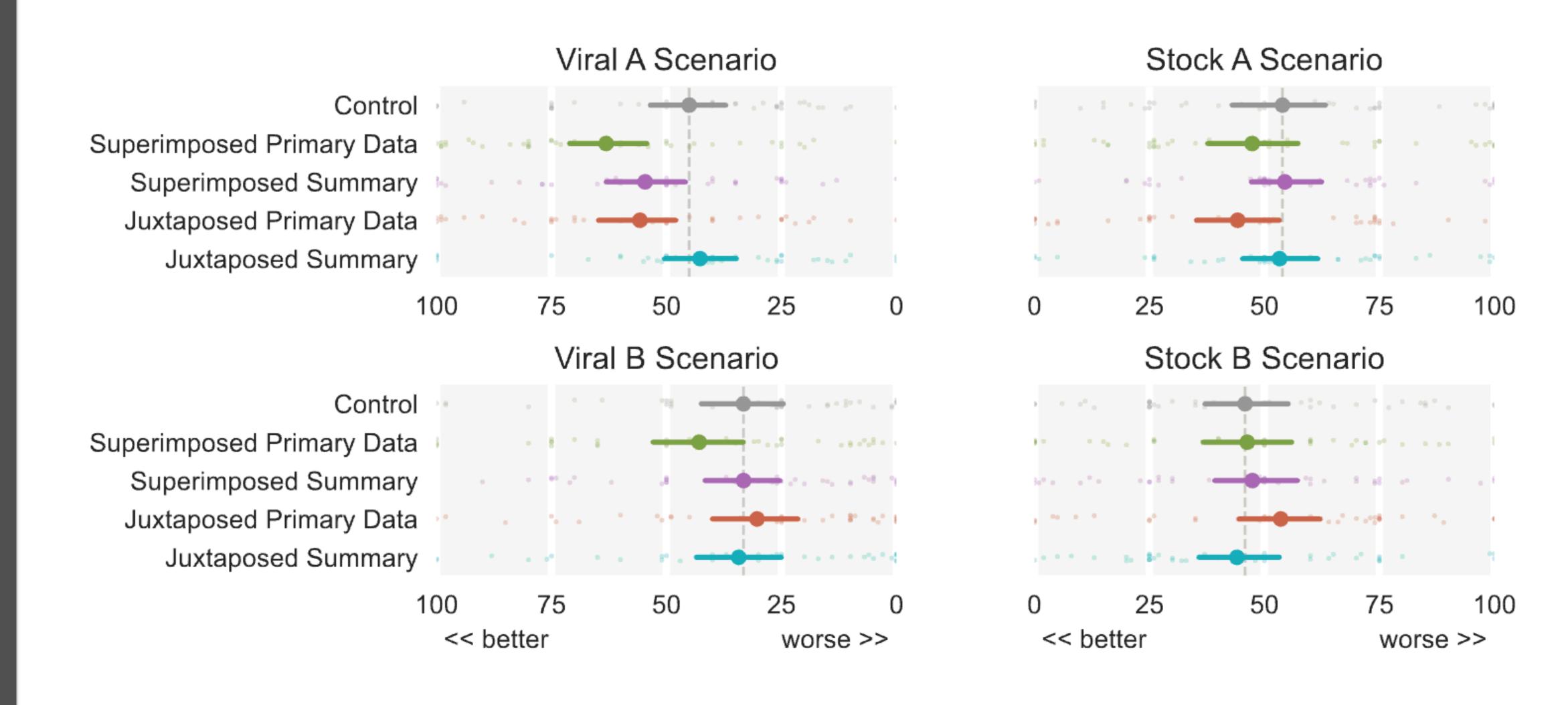
Based on this information, before traveling to Eldoril North (Policy A) I would buy this much insurance: (\$0 = no risk of getting sick, \$100 = very high risk of getting sick)

\$0 \$25 \$50 \$75	\$100

RESULTS



Differences by Scenario: More effective in "extreme" scenarios Superimposed Primary Data most promising Juxtaposed Summary no better than baseline





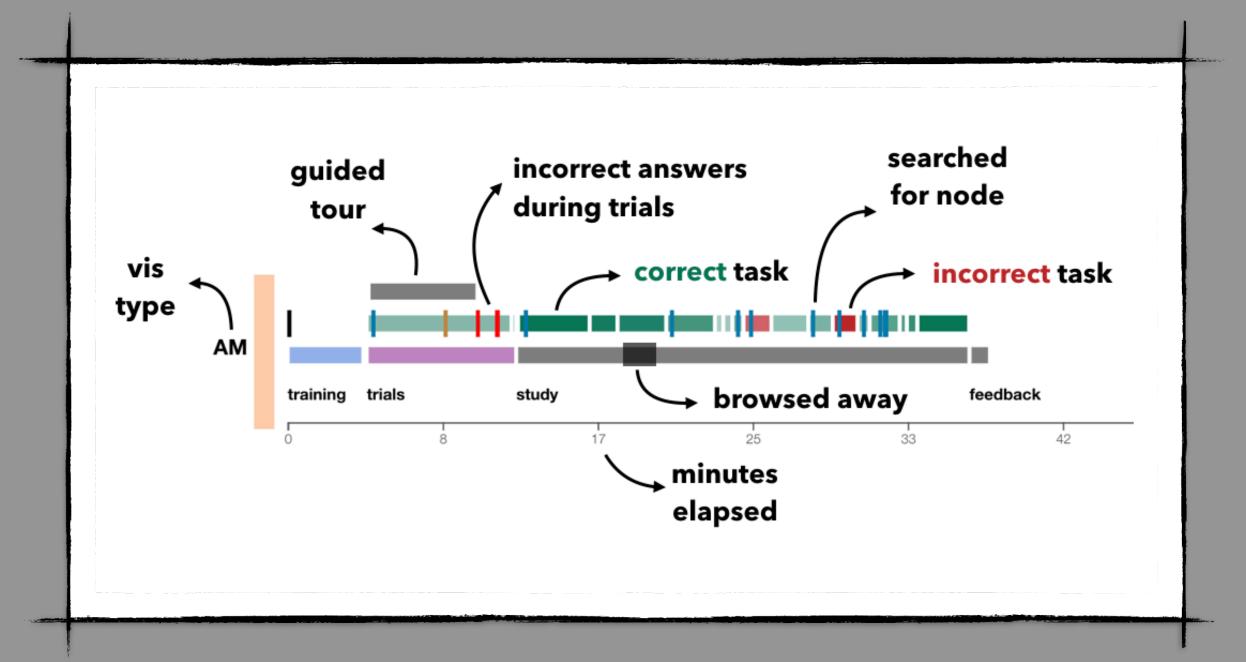
DISCUSSION

Superimposed Primary seems like a promising design against cherrypicking easy to implement easy to understand Juxtaposed Primary design useful as a way to make data explorers more usable

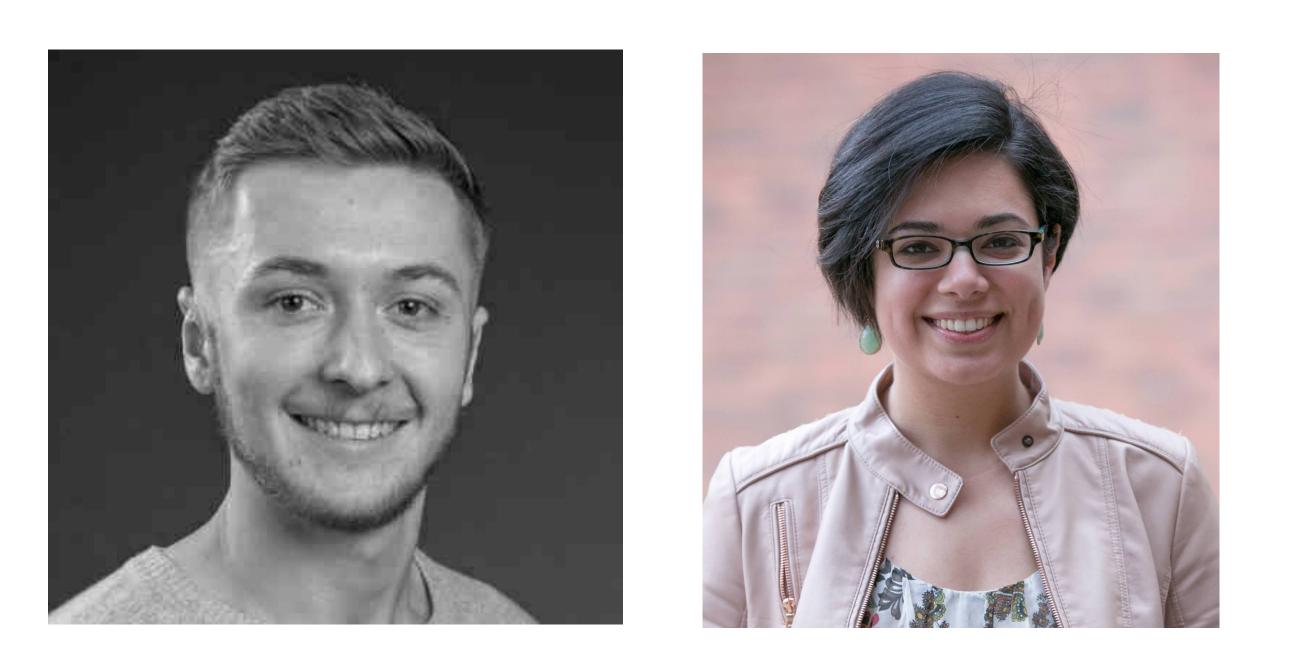


EMPIRICAL & THEORETICAL WORK

Evaluating Complex Systems







Zach Cutler, Jack Wilburn, Carolina Nobre, Lane Harrison et al.

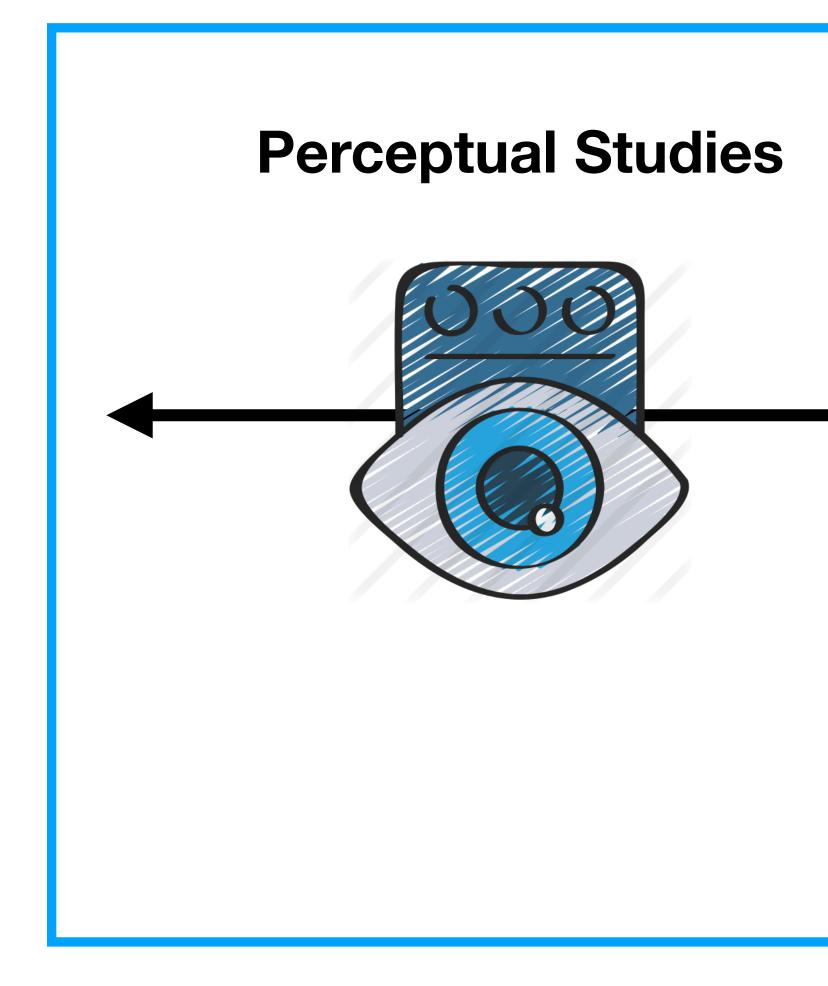
reVISit: Empirically Evaluating Complex **Interactive Visualization Techniques**



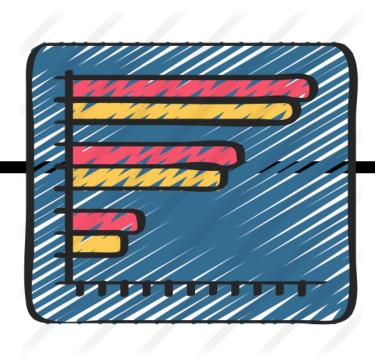
[CHI 2020, CHI 2021 & VIS 2023]

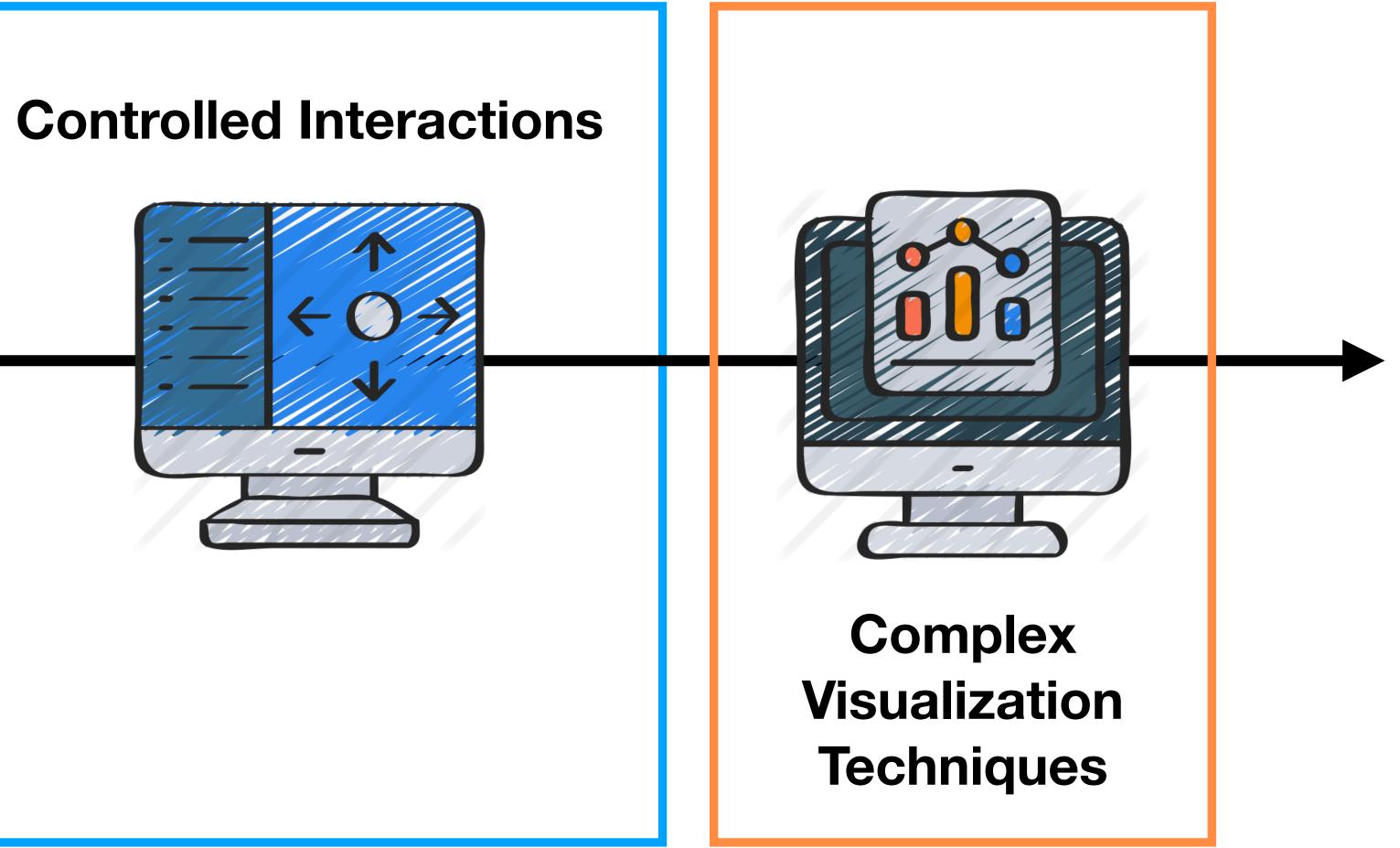






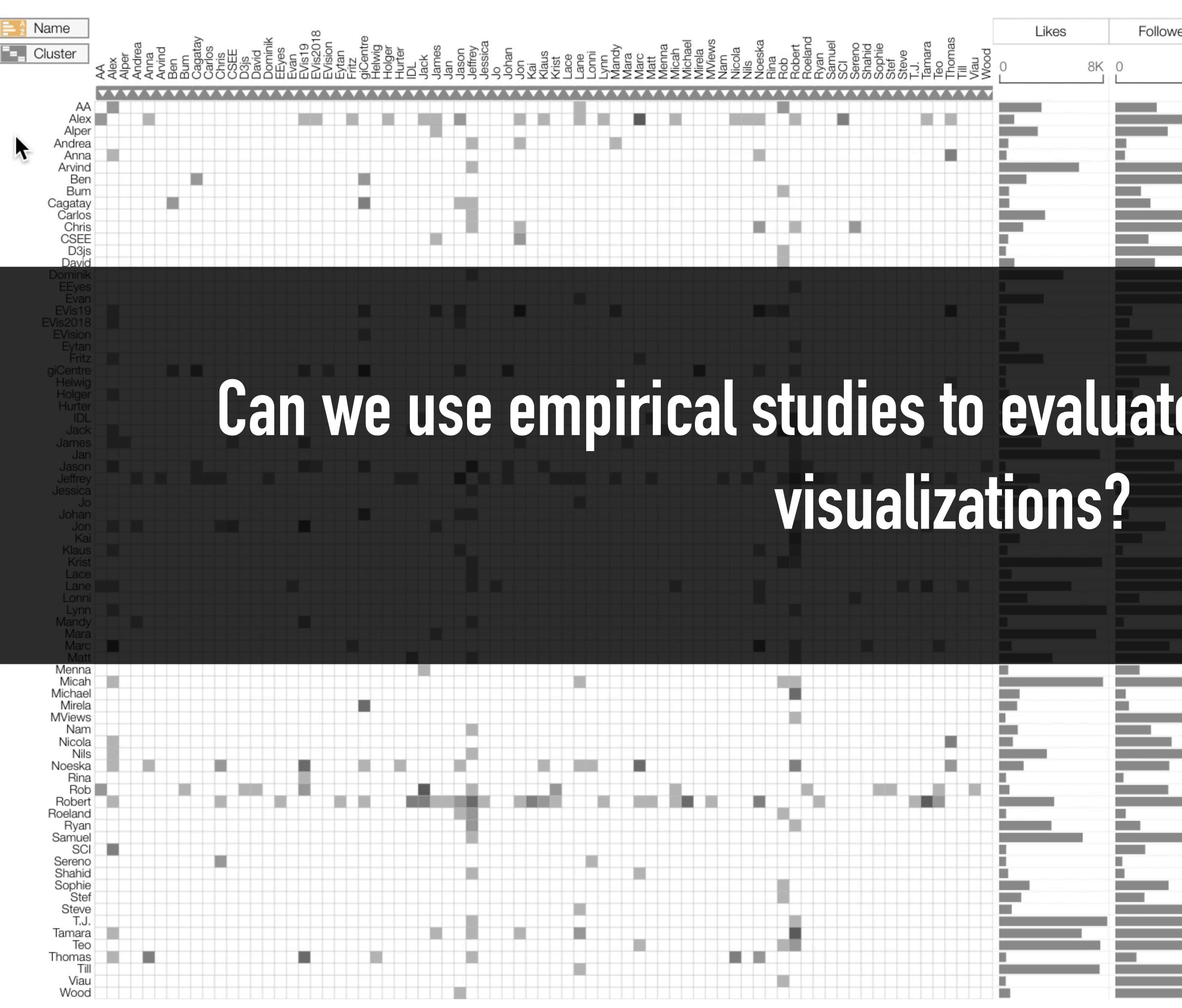
Commonly Evaluated Using Crowdsourcing





Static Visualizations

Considered not Amenable to Crowdsourced Evaluation



wers	Tweets	Acct. Age	Туре	Continer	nt
1K	0 6K	175 4K	Person Inst.	NA SA E	EU AS
			iteract		

CAN WE DO QUANTITATIVE **EVALUATION WITH** COMPLEX SYSTEMS?

Yes We Can!

- Picking the right techniques
- Evidence-based design
- Design validation
- Careful training
- Good compensation
- Interesting Tasks

Pushing the boundary of what can be evaluated using crowdsourcing



YET: COMPLEX STUDIES ARE HARD TO SET UP

PERILS OF ONLINE VISUALIZATION EXPERIMENTS

The support for types of stimuli in survey tools are limited

Lack of experiment **debugging** features

No built-in data collection and provenance tracking functions

There is no end-to-end platform for building interactive experiments



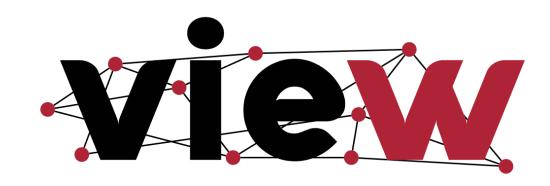




visualization design lab



REVISIT A TOOLKIT FOR DATA VISUALIZATION EXPERIMENTS







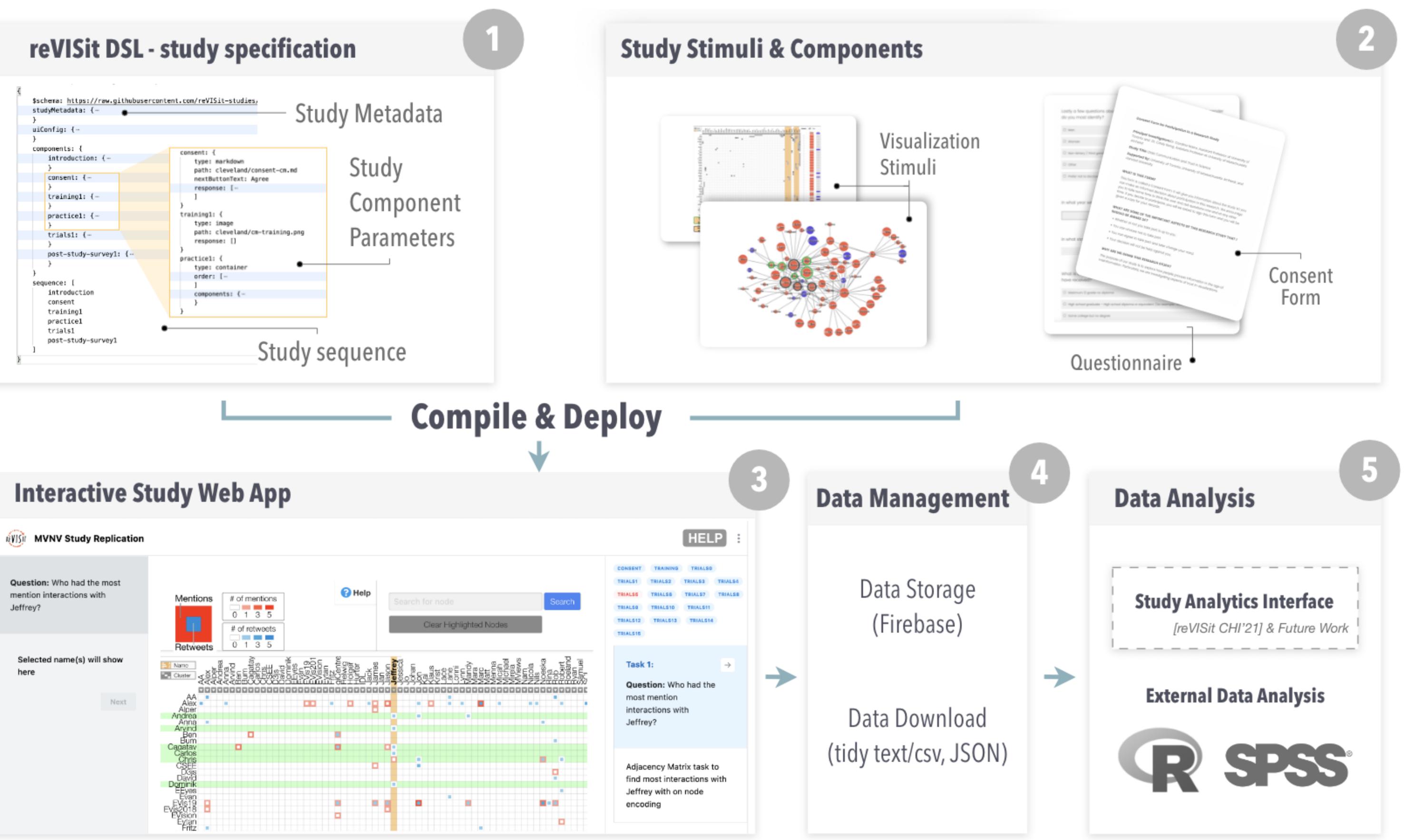


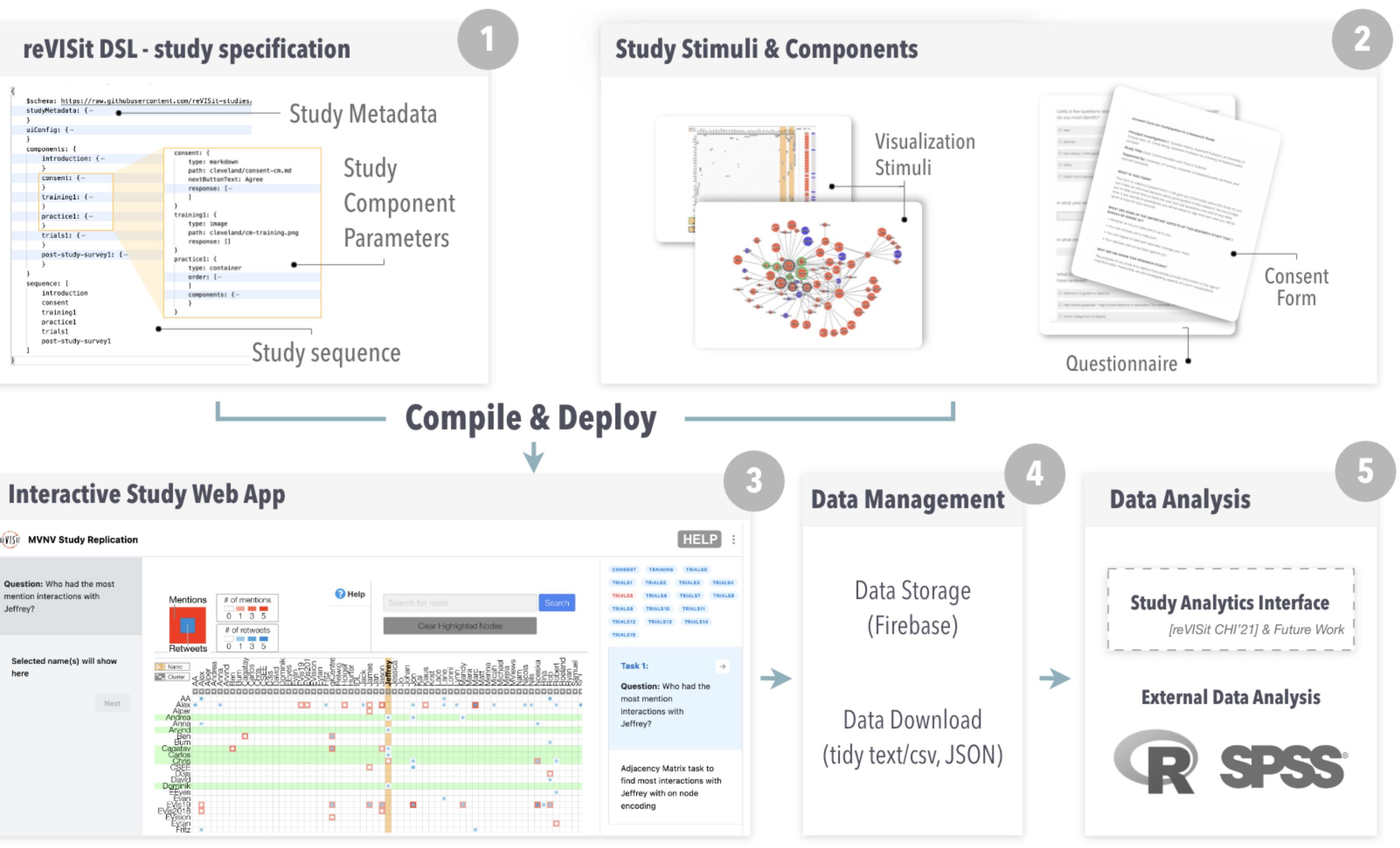






The Scope of reVISit





Task Instructions

Study Stimulus

Admin Panel

WHAT IS REVISIT: INFRASTRUCTURE FOR VIS STUDIES

Set up studies with all components (consent, training, trials, tasks, surveys, etc.) including sophisticated study designs (randomization) Simple data tracking and data export Simple to deploy and run studies Fully reproducible and open. You can share your whole study setup for anyone to re-run, without having to have a license for software (qualtrics).

REVISIT SPEC Components & Inheritance

```
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},
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```

REVISIT SPEC Study Design

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            "consent",
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                    "rectangleBrush_q1",
                    "axisBrush_q1",
                    "sliderBrush_q1"
            "post-study-survey",
            "survey"
```



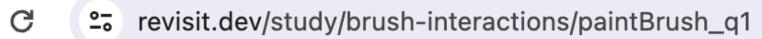


Task:

How many Adventure Movies have a production budget value greater than 100 million?

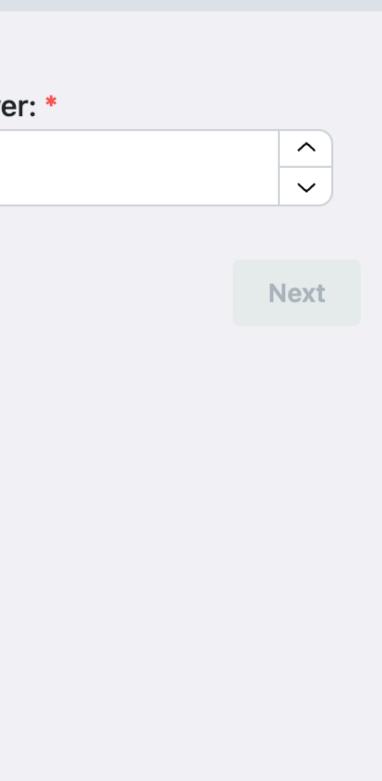
Answer: *

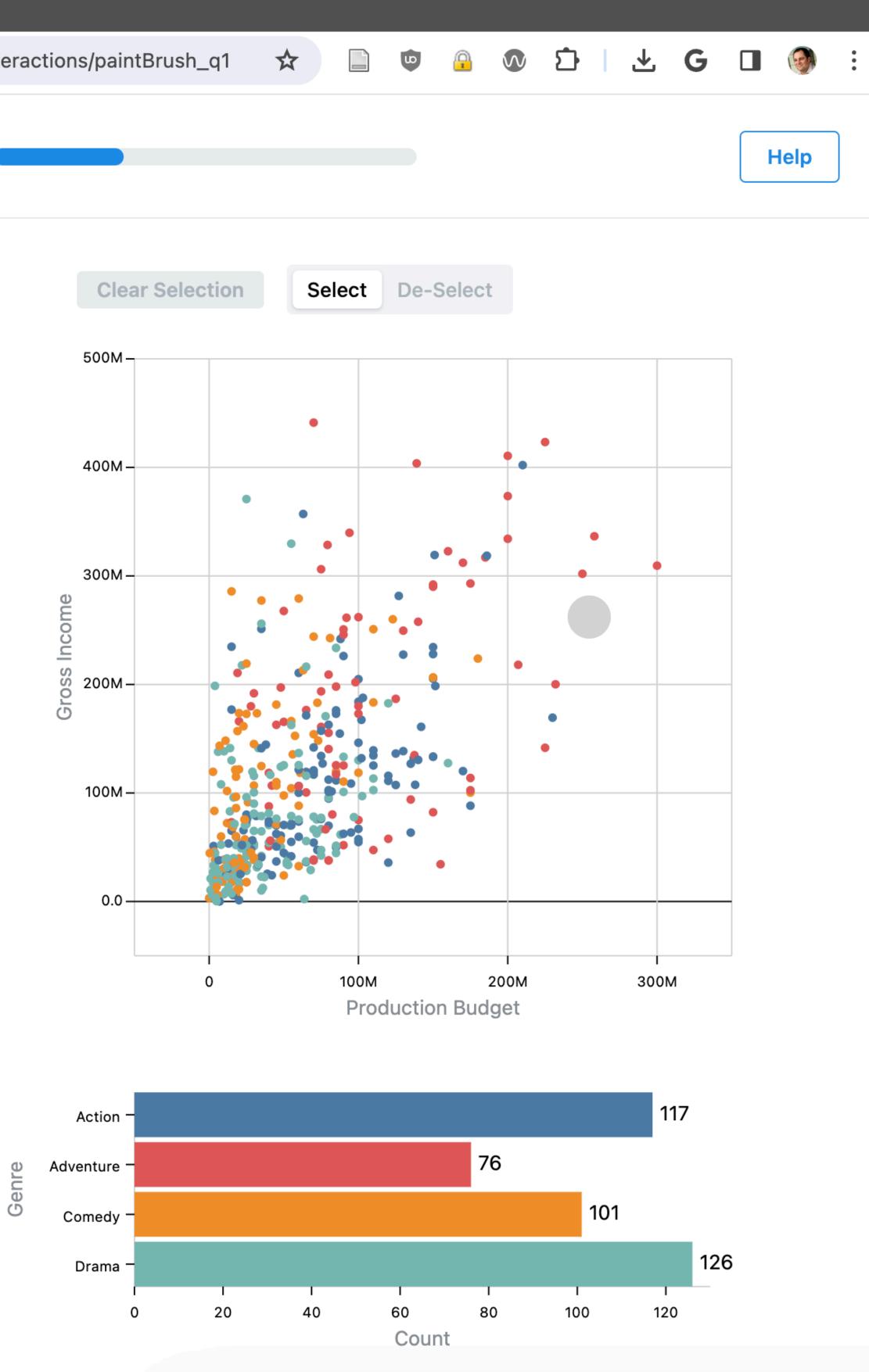




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Interactive Selections in Scatterplots

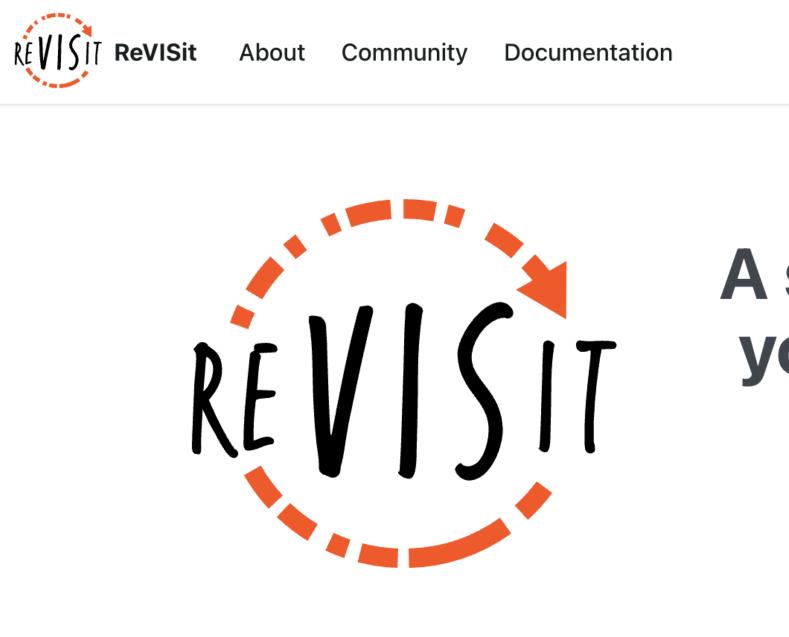




https://revisit.dev

DOCUMENTATION & COMMUNITY

Documentation & Tutorials on Website Community participation via slack etc. Tutorial at IEEE VIS 2024



About ReVISit

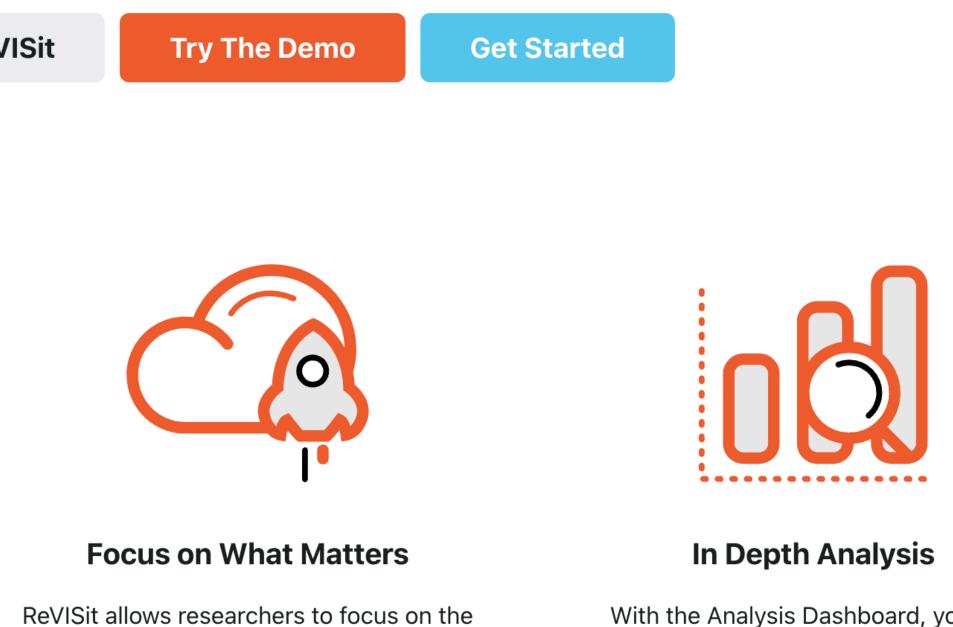


Flexible And Powerful

ReVISit is designed with researchers from all disciplines in mind. It is simple enough

v1.0.0-beta11 - GitHub 🖸 Demo 🖸 🔅 📿 Search 📧 🕅

A study creation platform allowing you to quickly create, publish, and disseminate your customized visualization study.

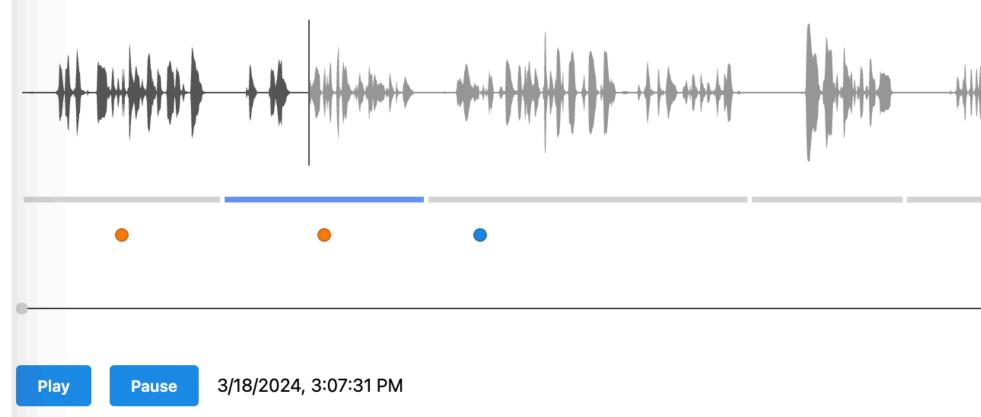


visual stimuli without the hassle of setting

With the Analysis Dashboard, you can investigate the results from your study with

CAN WE DO QUALITATIVE EVALUATION **ONLINE?**

Maybe? Think Aloud & Provenance



Transcript

So for this task, uh, I just need to consider penguins that weigh between 4 and a half and 5.

Okay kilograms, but all that I'll flip a length between 2 and 215.

All right. I'm going to have to be guite precise so I can

I'm I'm confident about the 4 to 5K but 2 to 15

Okay, so I'm just going to click on select.

All right, so actually 200 to 20.

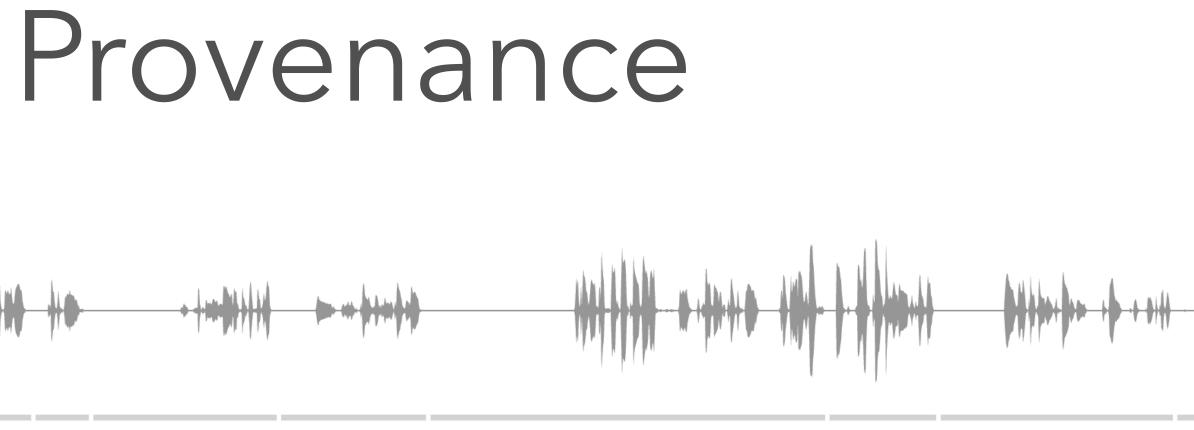
So I'm just going to have to have to estimate where.

2 15 is

which is approximately there about 75% of that.

Let me just clear that selection. Let me just try this again.

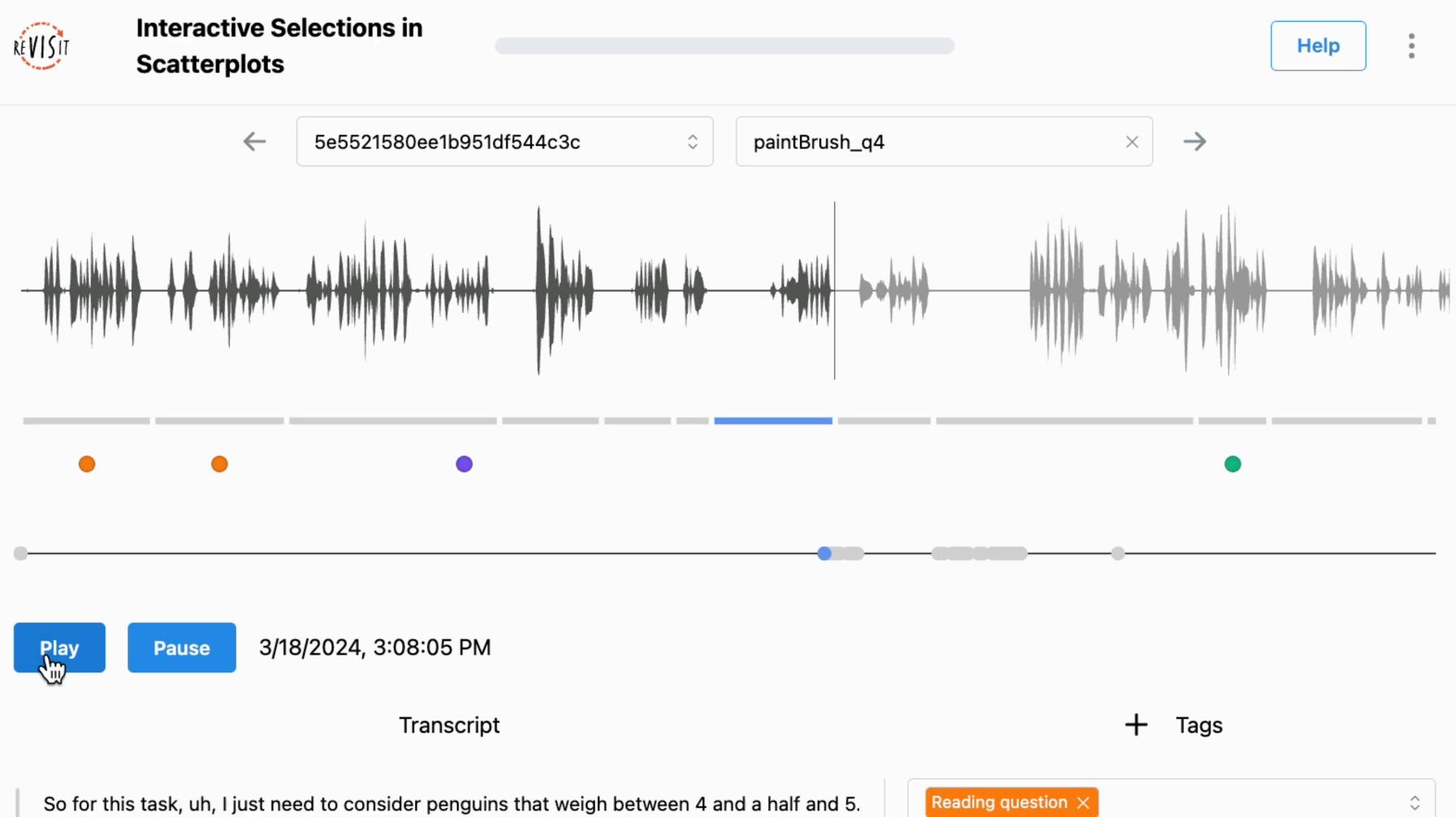
Pushing the boundary of what can be evaluated using crowdsourcing



+ Tags

Reading question \times	Ŷ
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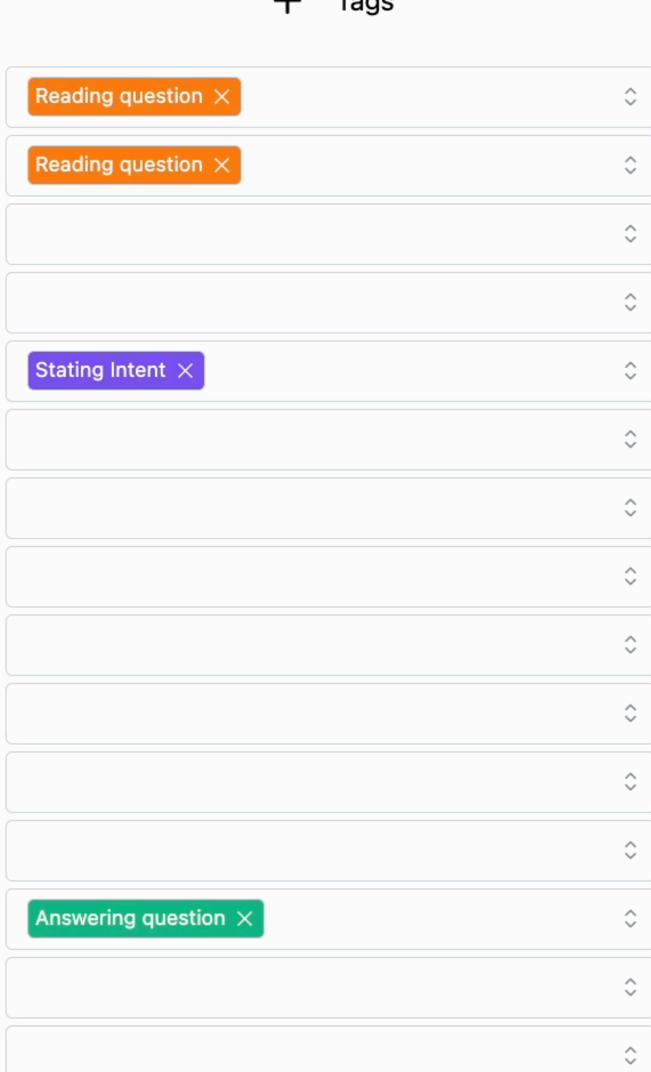
Okay. Alright, so that's the 4 to 5K body mass. And then the 2002215.

So that's giving me those results.

so the most common penguin There Is the Gen 2

and the least common penguin with 5 is the a deli if I've pronounced that correctly.

Okay clicking.





Interactive Selections in Scatterplots

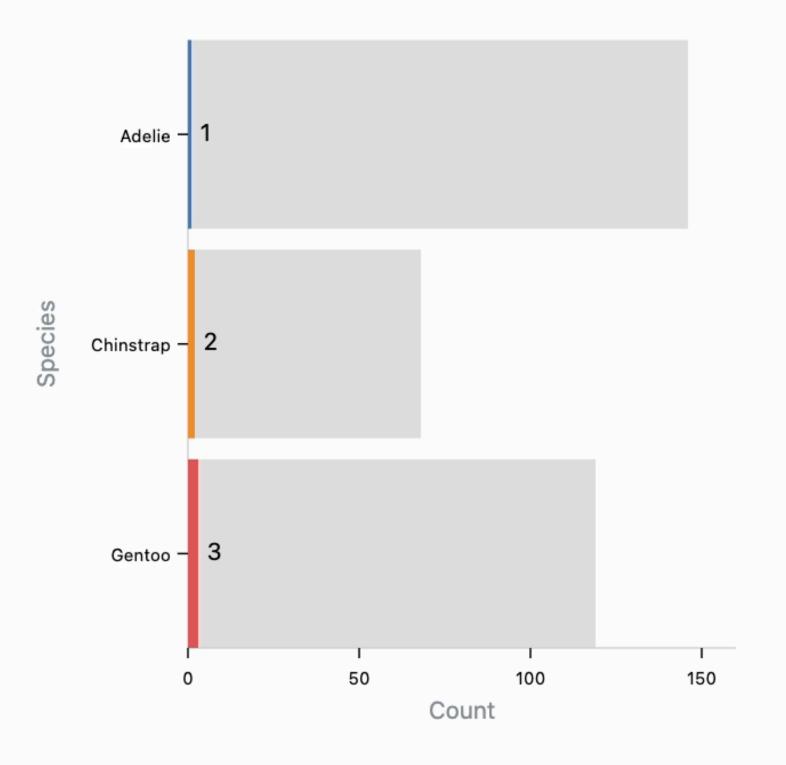
Task:

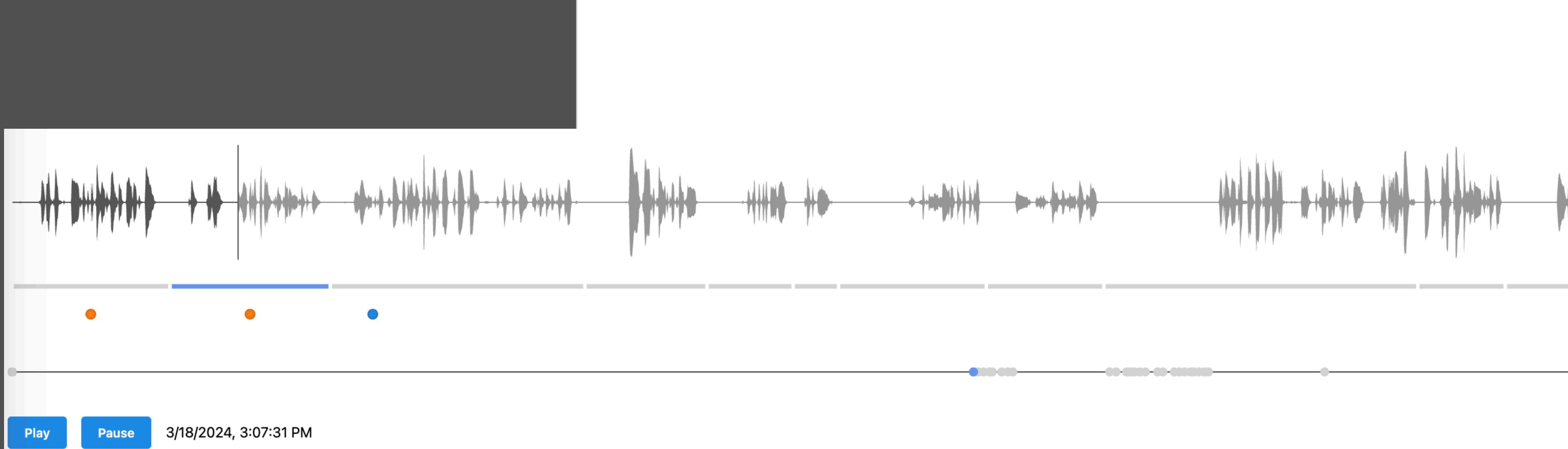
Consider only penguins that weigh between 4k and 5k grams (g) and have flipper lengths between 200 and 215mm. What is the most and least common type of penguin in this subset?

While answering this question, please verbalize your thoughts, especially any insights you have or problems you run into.

Most common Penguin: *	
	\$
Least common Penguin: *	
	¢
Next	







So for this task, uh, I just need to consider penguins that weigh between 4 and a half and 5.

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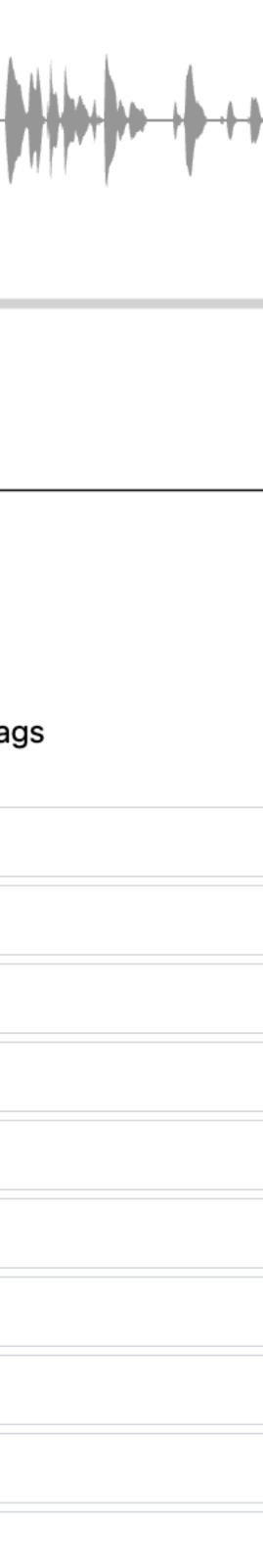
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Transcript

🕂 Tags

Reading question \times
Reading question \times
Stating intent \times



FUTURE WORK: BETTER SCAFFOLDING FOR STUDIES

reVISit: Scalable Empirical Evaluation of Interactive Visualizations

Community Input

Collaborators Core Community

Community Workshops

Core Infrastructure: Software Components

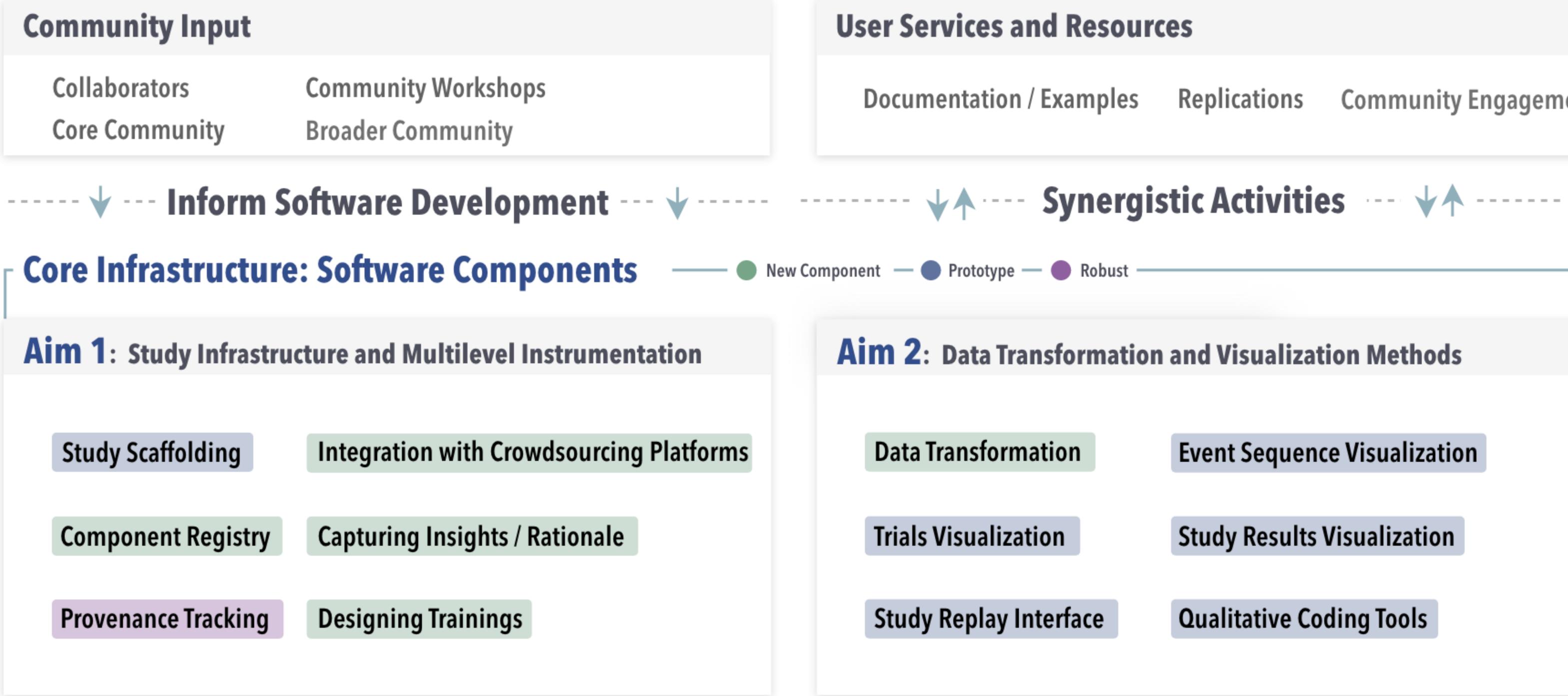
Aim 1: Study Infrastructure and Multilevel Instrumentation

Study Scaffolding

Component Registry

Provenance Tracking

Designing Trainings



Replications **Community Engagement**

Aim 2: Data Transformation and Visualization Methods

tion	Event Sequence Visualization	
on	Study Results Visualization	

Qualitative Coding Tools

REFLECTIONS



A STEW OF RESEARCH TOPICS

Doing different things is interesting! Keeps me engaged **Role of Engineering** Building / maintaining / documenting useful things increases impact reVISit, Persist, UpSet **Crossover Benefits** Provenance -> reVISit & Persist reVISit -> Misinformation Engineering infrastructure: benefits all



Alexander Lex http://vdl.sci.utah.edu

A Hodgepodge of Visualization Research: **Provenance, User Studies, Misinformation**



VISUAIZATION design lab

THE UNIVERSITY OF UTAH



