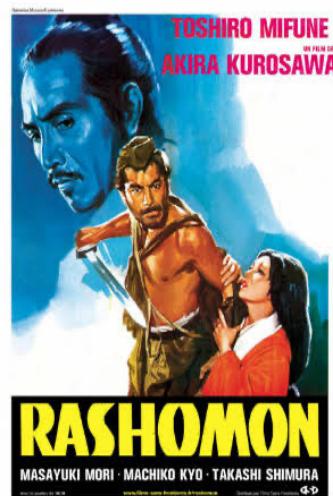


Interacting with your visualizations

Laura Garrison, University of Bergen

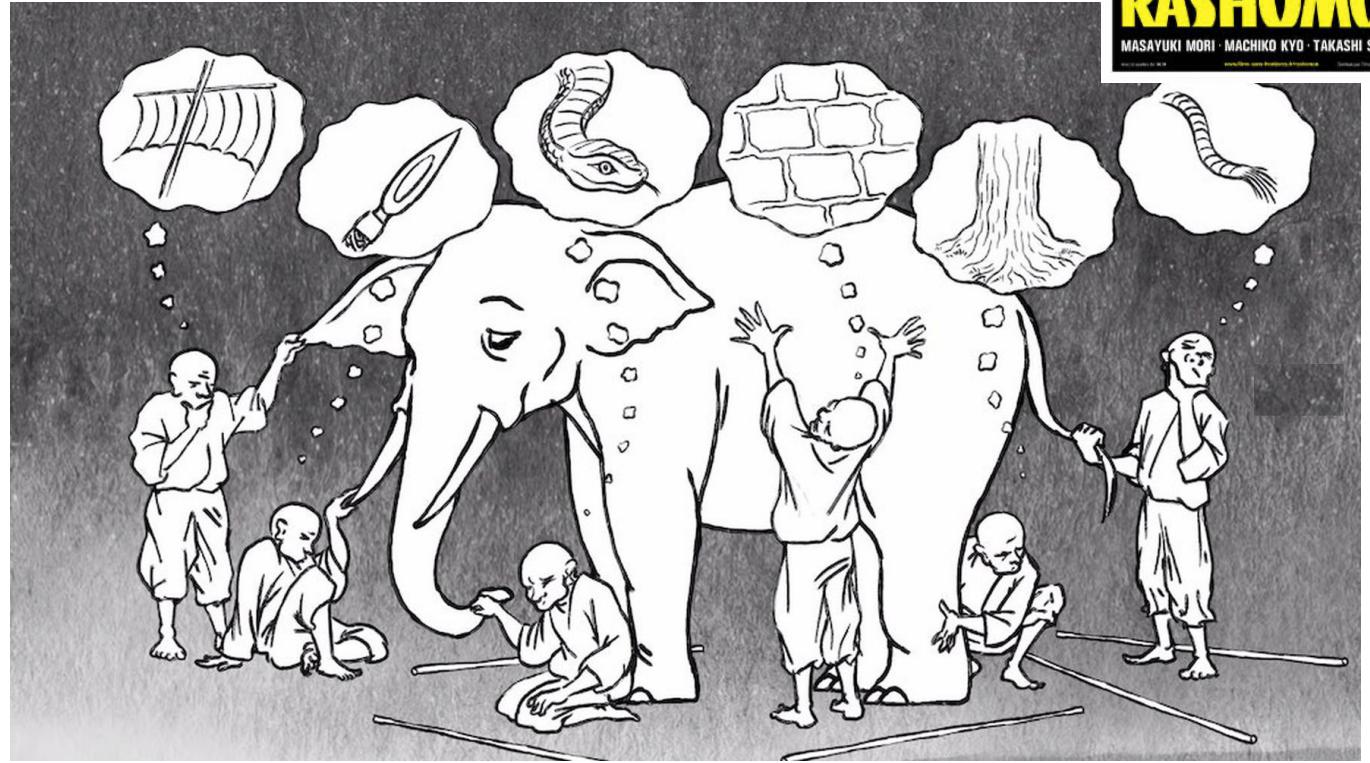
laura.garrison@uib.no

ICTP Workshop 2023



Why visualize?

- Rashomon Effect/»Elephant in the Dark» (Rumi)
 - Different models, parameters, representations, etc. can tell different stories
- Visualization can help us spot and understand reasons for these differences

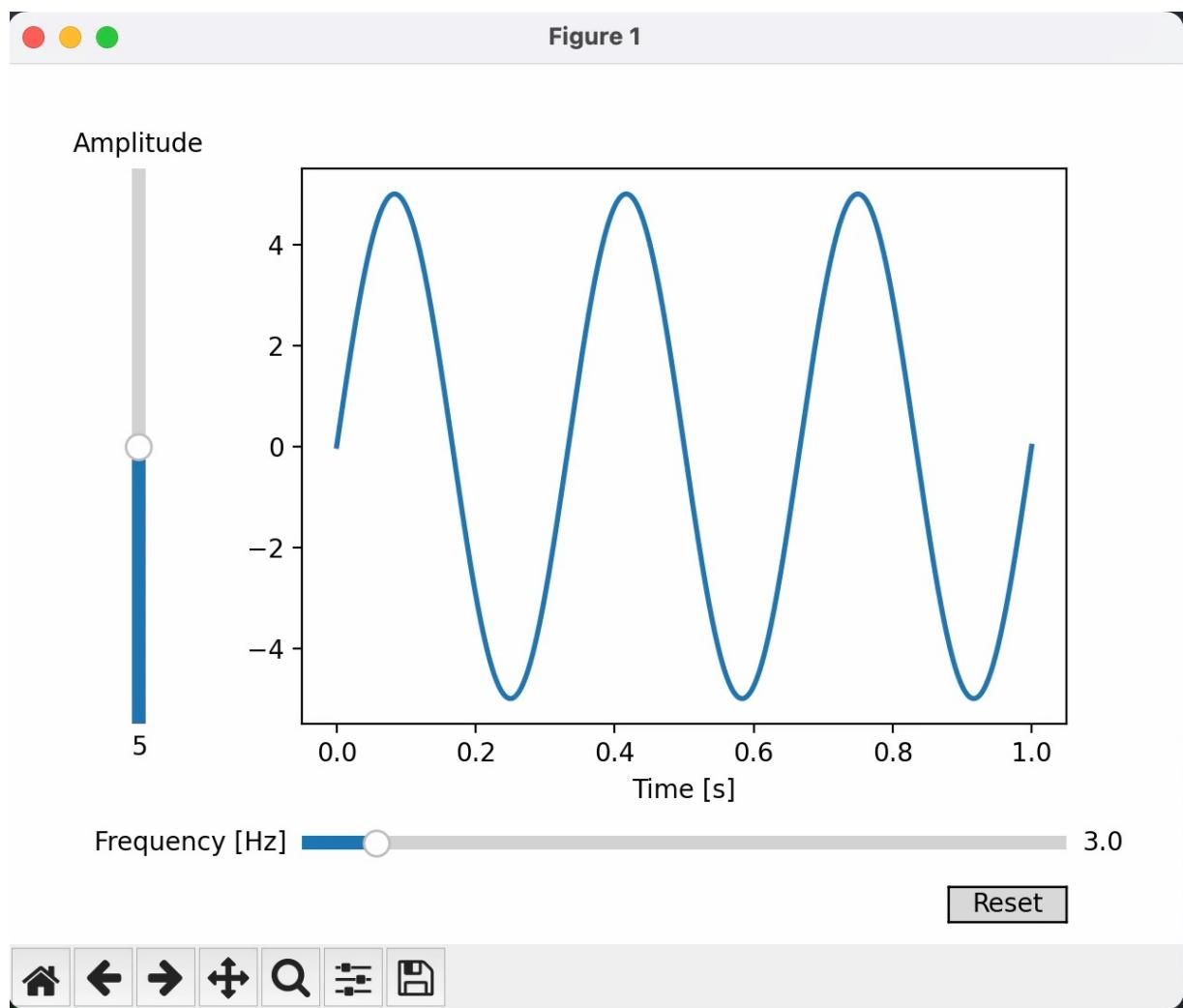


Why interact?

- To engage!
- Manage complexity of data
- On-the-fly hypothesis testing

Why interact?

- To engage!
- Manage complexity of data
- On-the-fly hypothesis testing
- **See effects of parameter(s) changes on model**

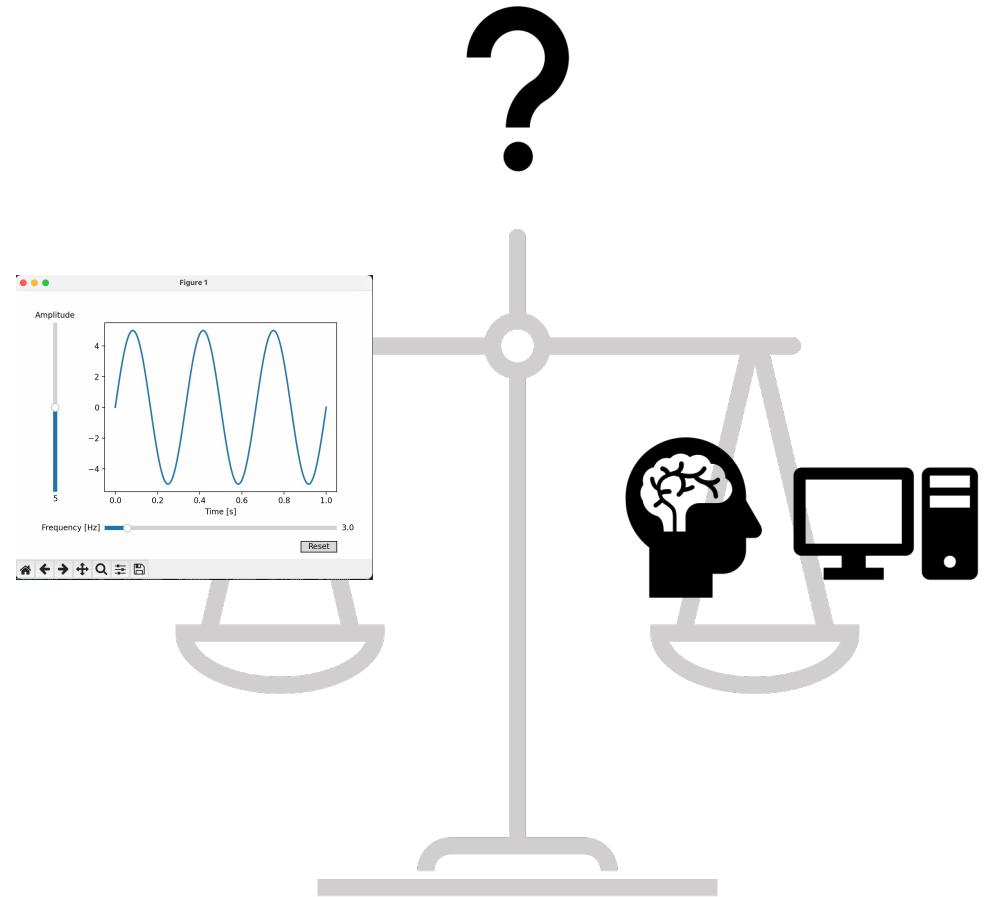


Efficiency Principle

“The gains from using an **interactive** visual approach should **outweigh** the **computational resources** and **human effort** necessary to carry out the analysis.”

[Interactive Visual Data Analysis](#)

Tominski & Schumann 2020

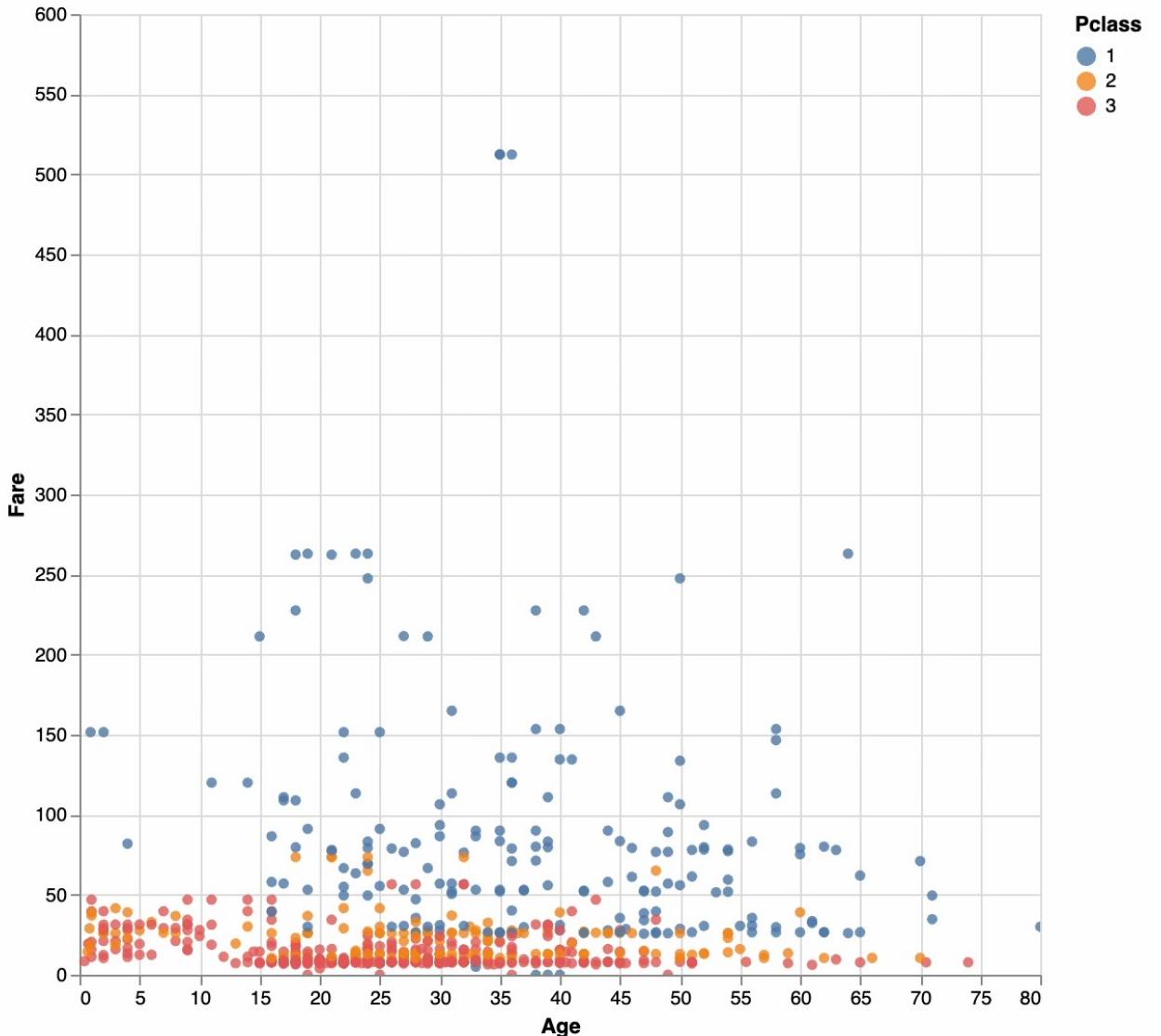


**Overview first,
zoom and filter,
details on demand**

Shneiderman's Visual Information-Seeking Mantra

Visual information-seeking mantra in action

Overview first,
zoom and filter,
details on demand

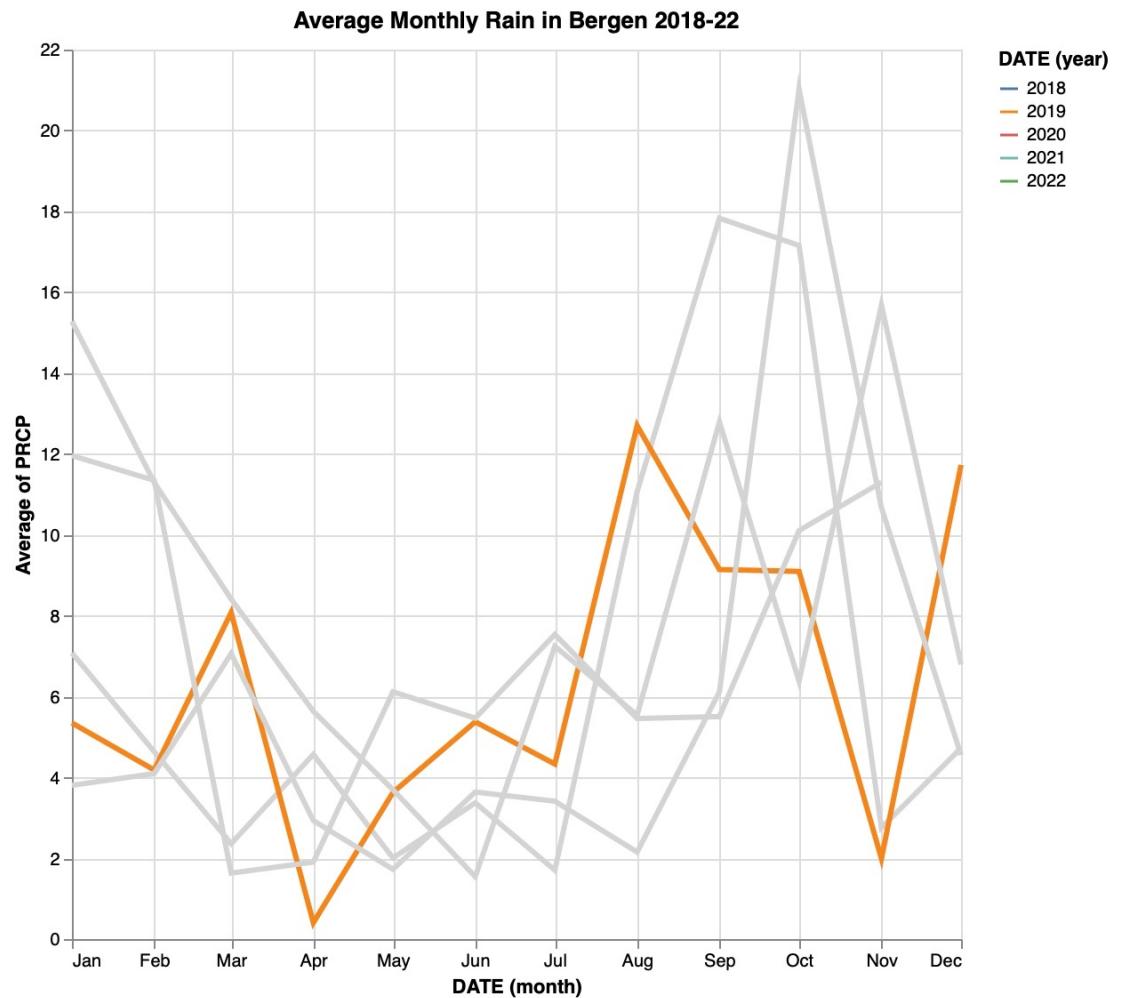


Interaction approaches to explore today

- **Manipulate view**
 - Adjust view
- **Coordinate faceted views**
 - Coordinate manipulations across multiple views
- **Reduce view(s)**
 - Filter
 - aggregate
 - embed information

Manipulate View

- Select
 - e.g., highlighting
- Change over time
 - e.g., animated transitions



Manipulate View

- Navigate
 - e.g., pan within a view, scrolltelling
-



[View this gorgeously impactful visualization](#)

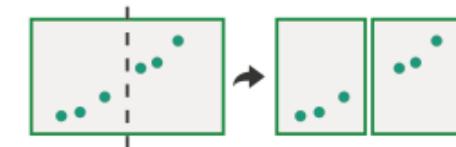
Coordinate Faceted Views

- Faceting → a separation of information into different views
- Why facet?
 - Eyes > memory
- Approaches
 - Side by side (juxtapose/partition)
 - On top of (superimpose)

Juxtapose



Partition

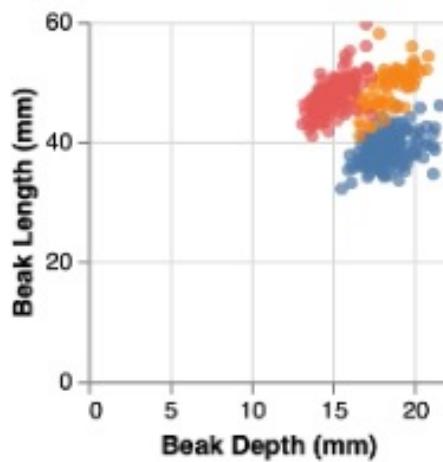


Superimpose



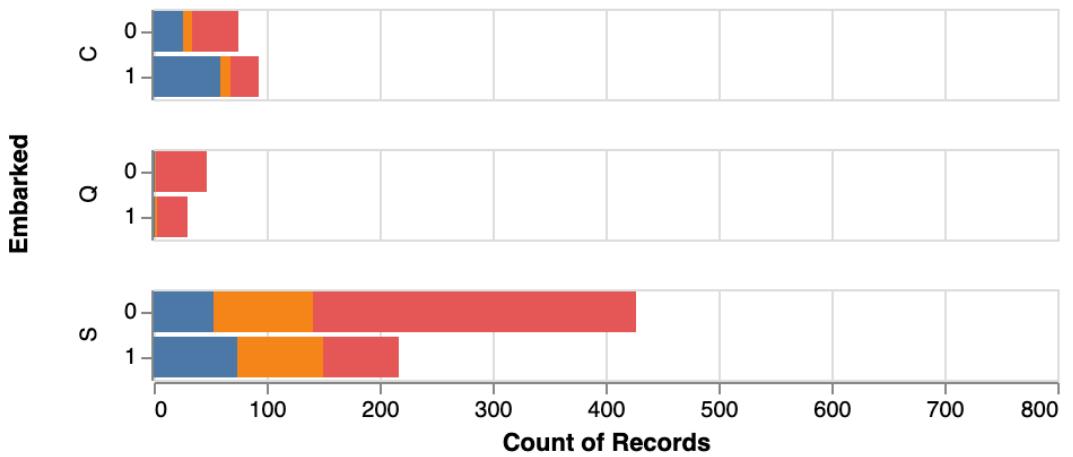
Juxtaposed views

Side by side (can be same/subset/totally different data)



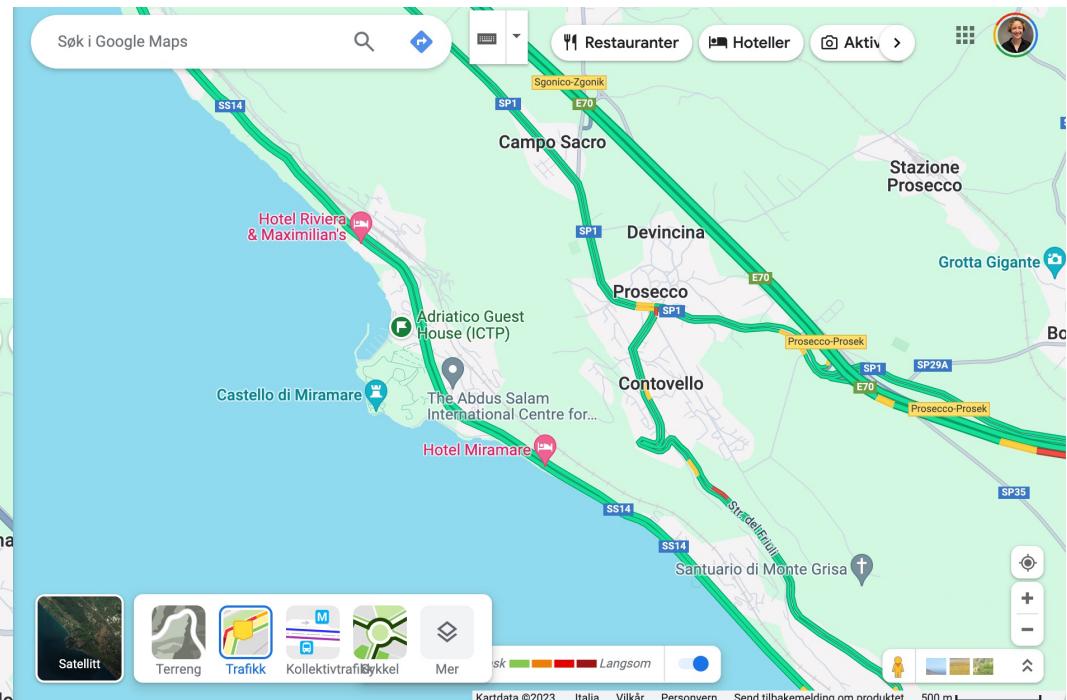
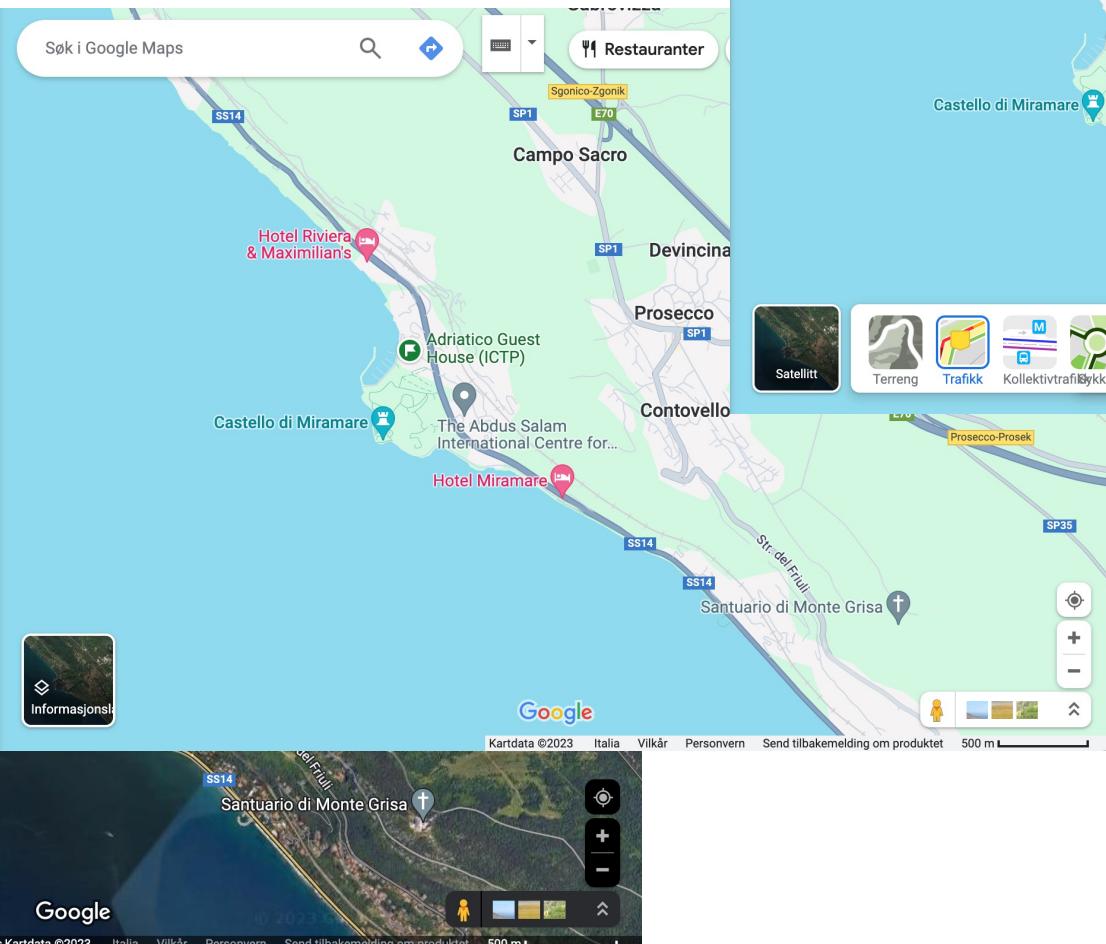
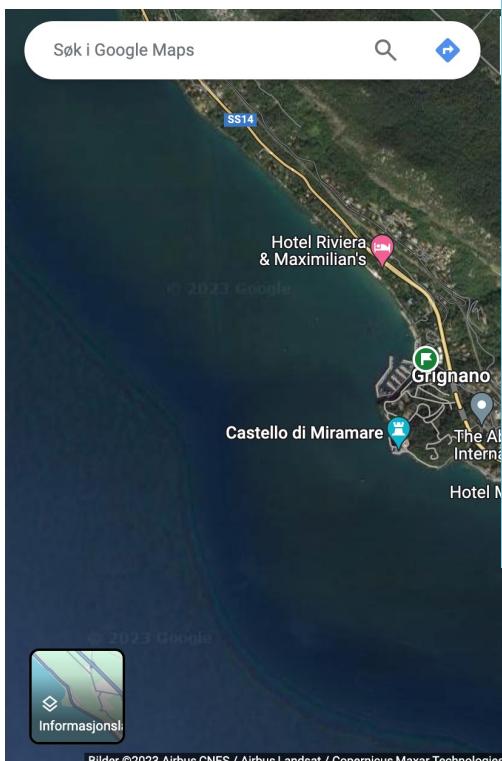
(Palmer penguins dataset)

Partitioned into different categories of one variable in a dataset



(titanic dataset)

Superimposed views



Coordinating your views

→ Share Encoding: Same/Different

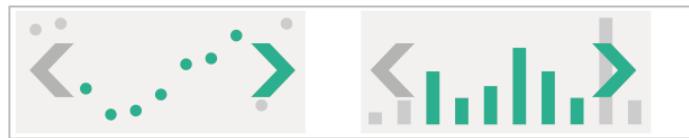
→ *Linked Highlighting*



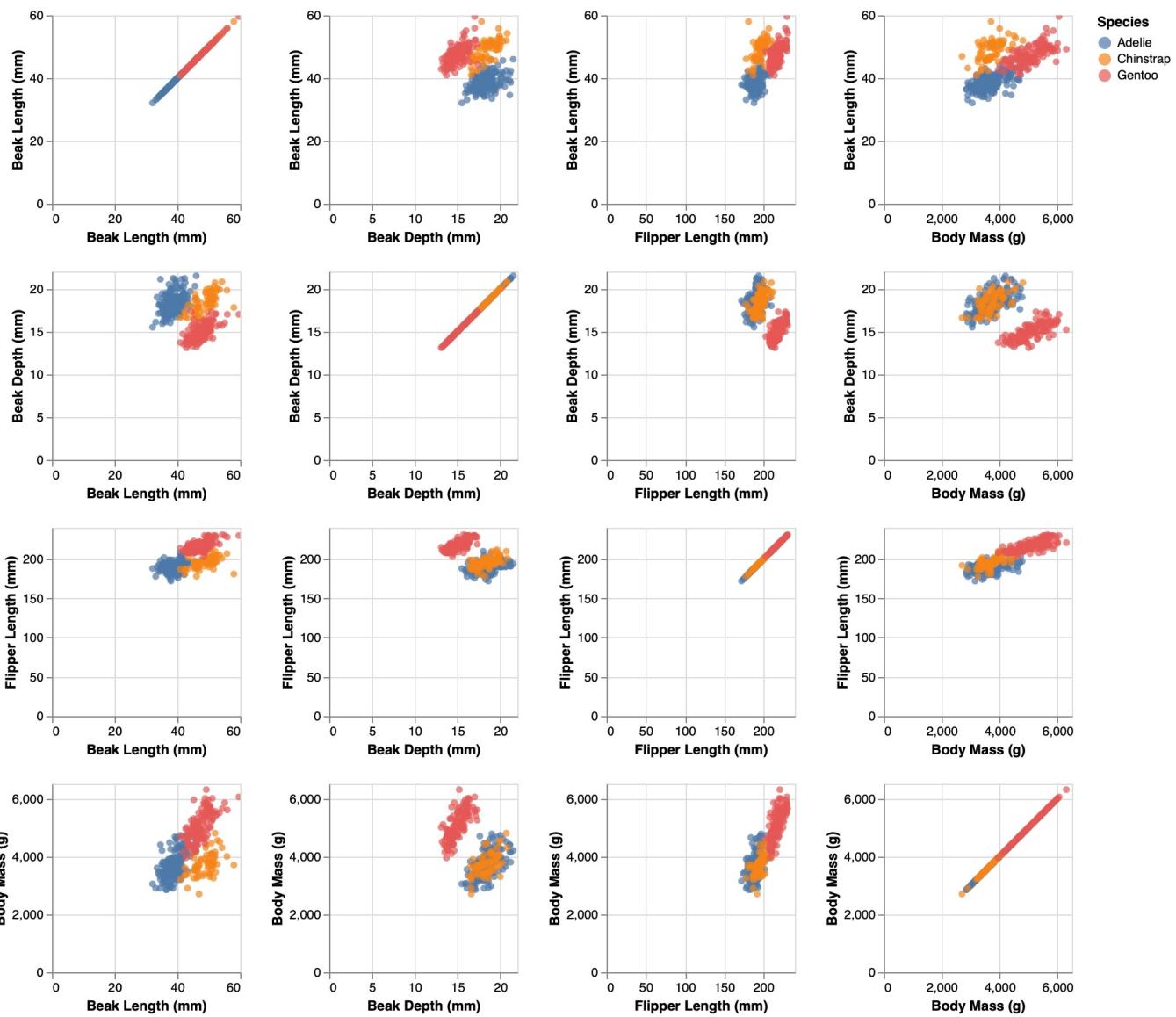
→ Share Data: All/Subset/None



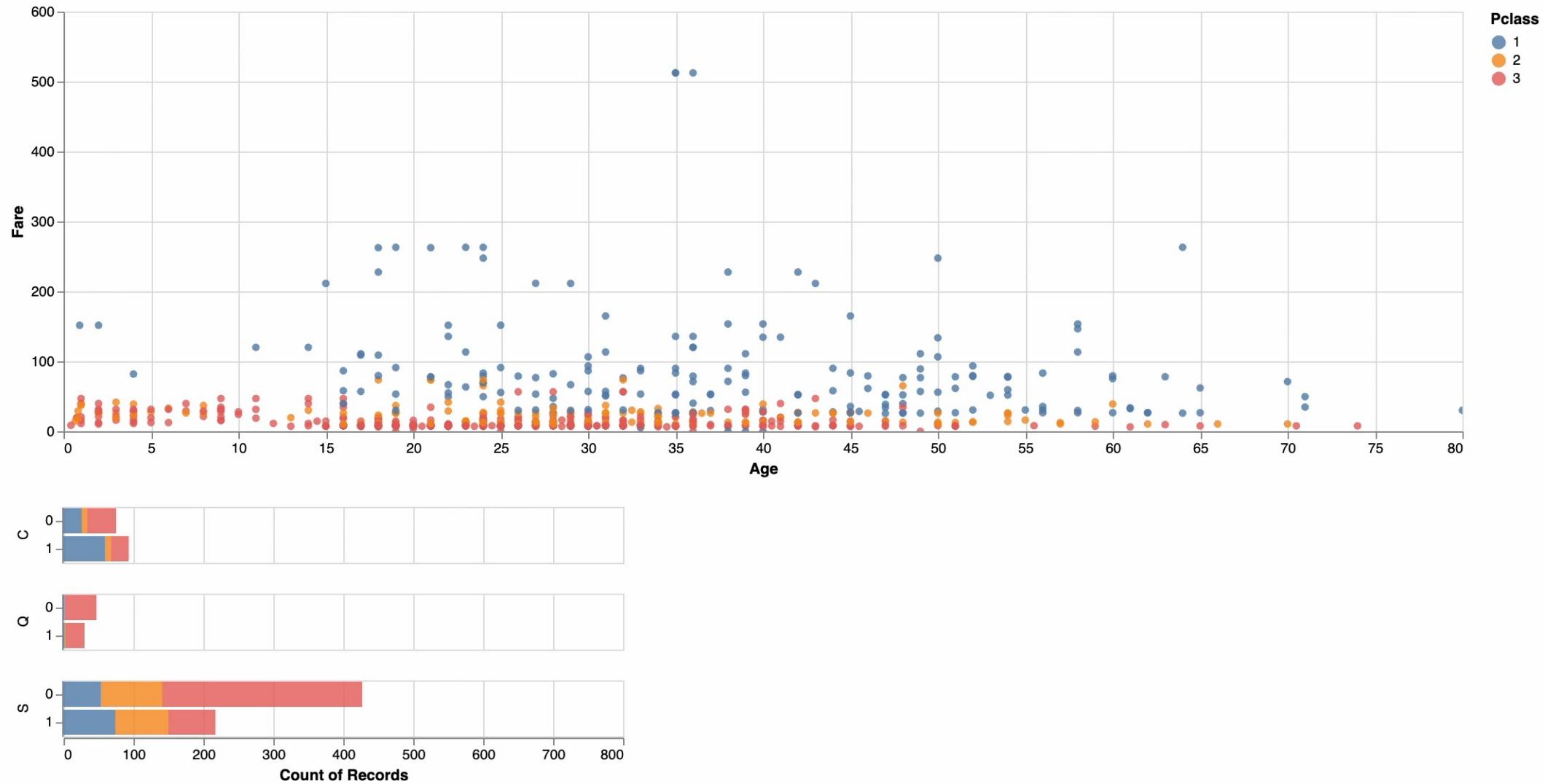
→ Share Navigation



Shared Navigation



Linked Highlighting



Reduce Items

④ Filter

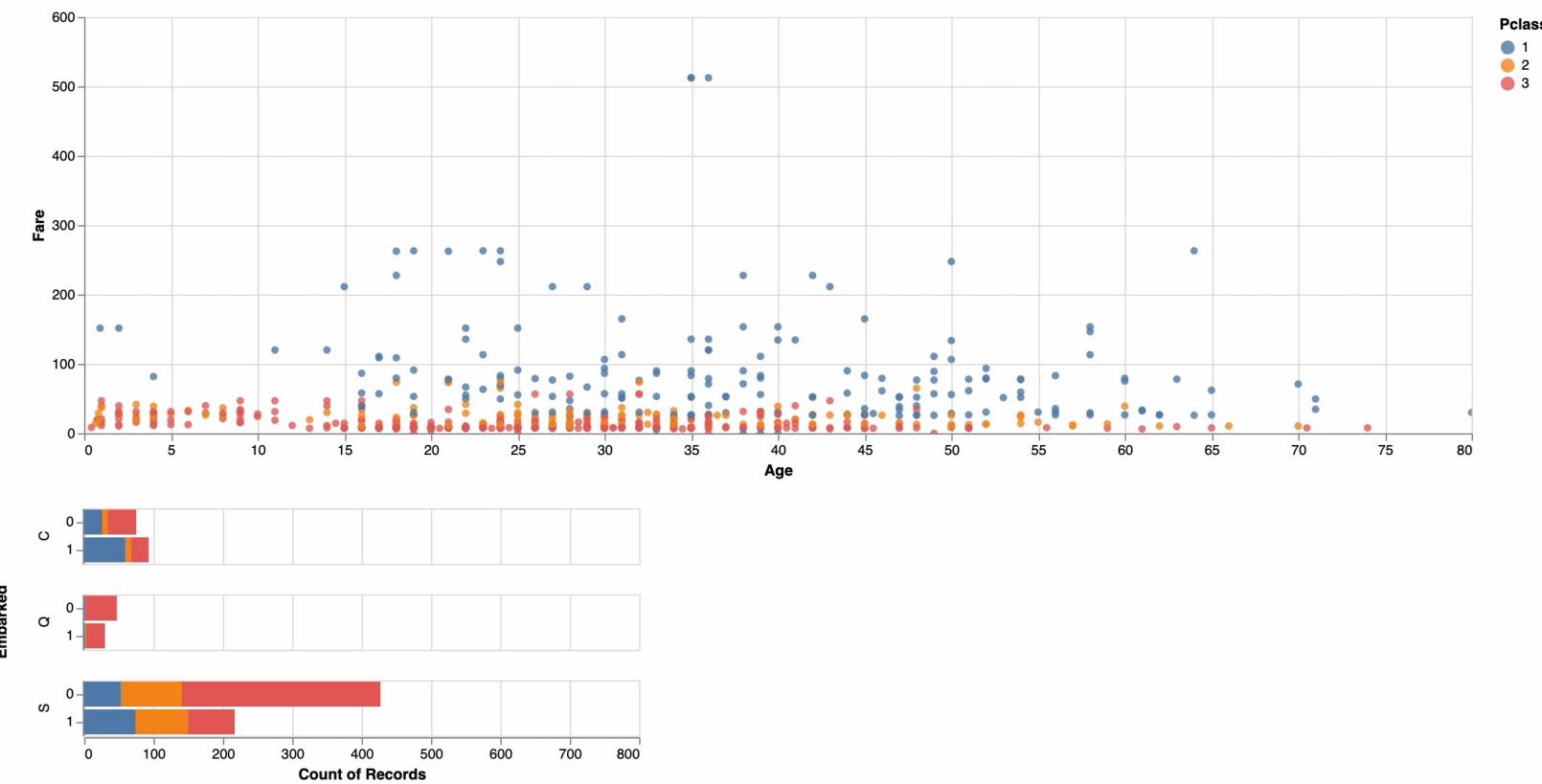
→ Items



→ Attributes

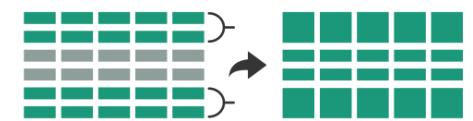


e.g., an option in interface to remove “Embarked” or “Survived” variables from display in this visualization

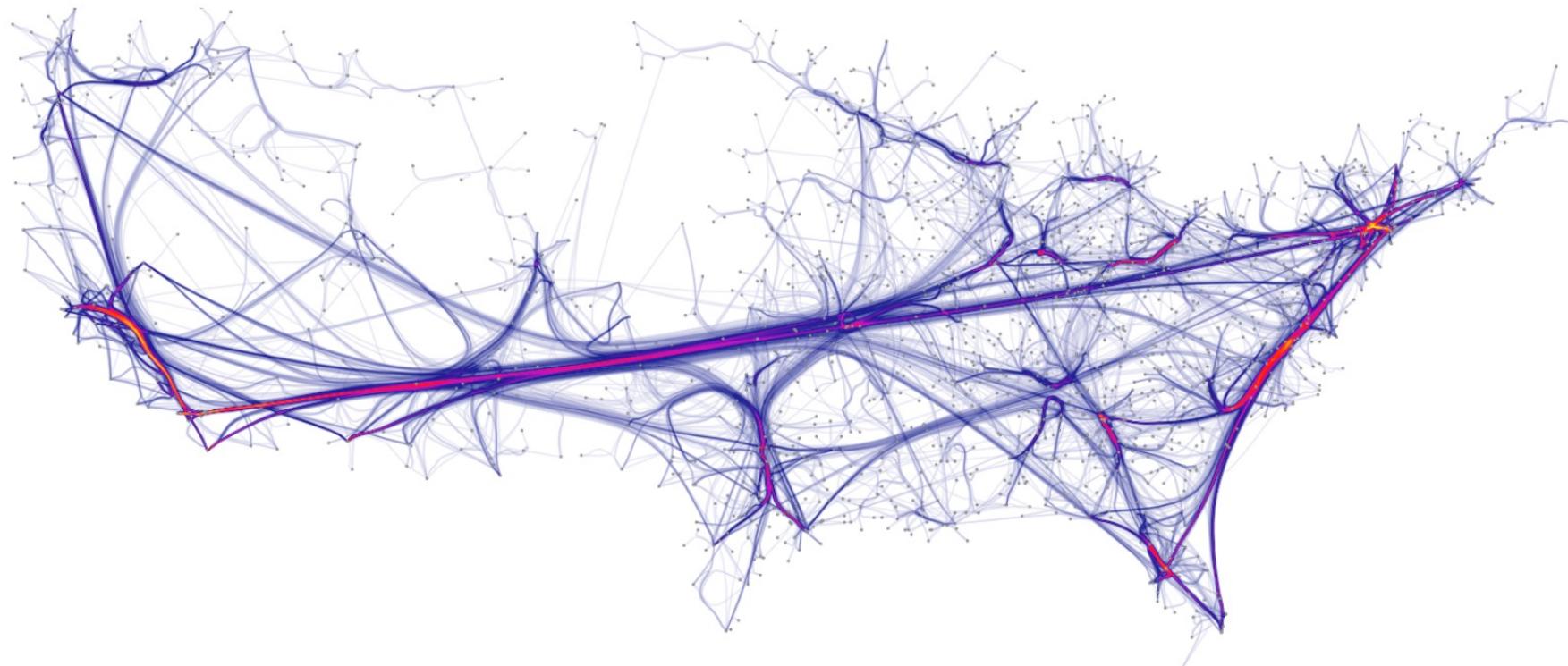


⟳ Aggregate

→ Items



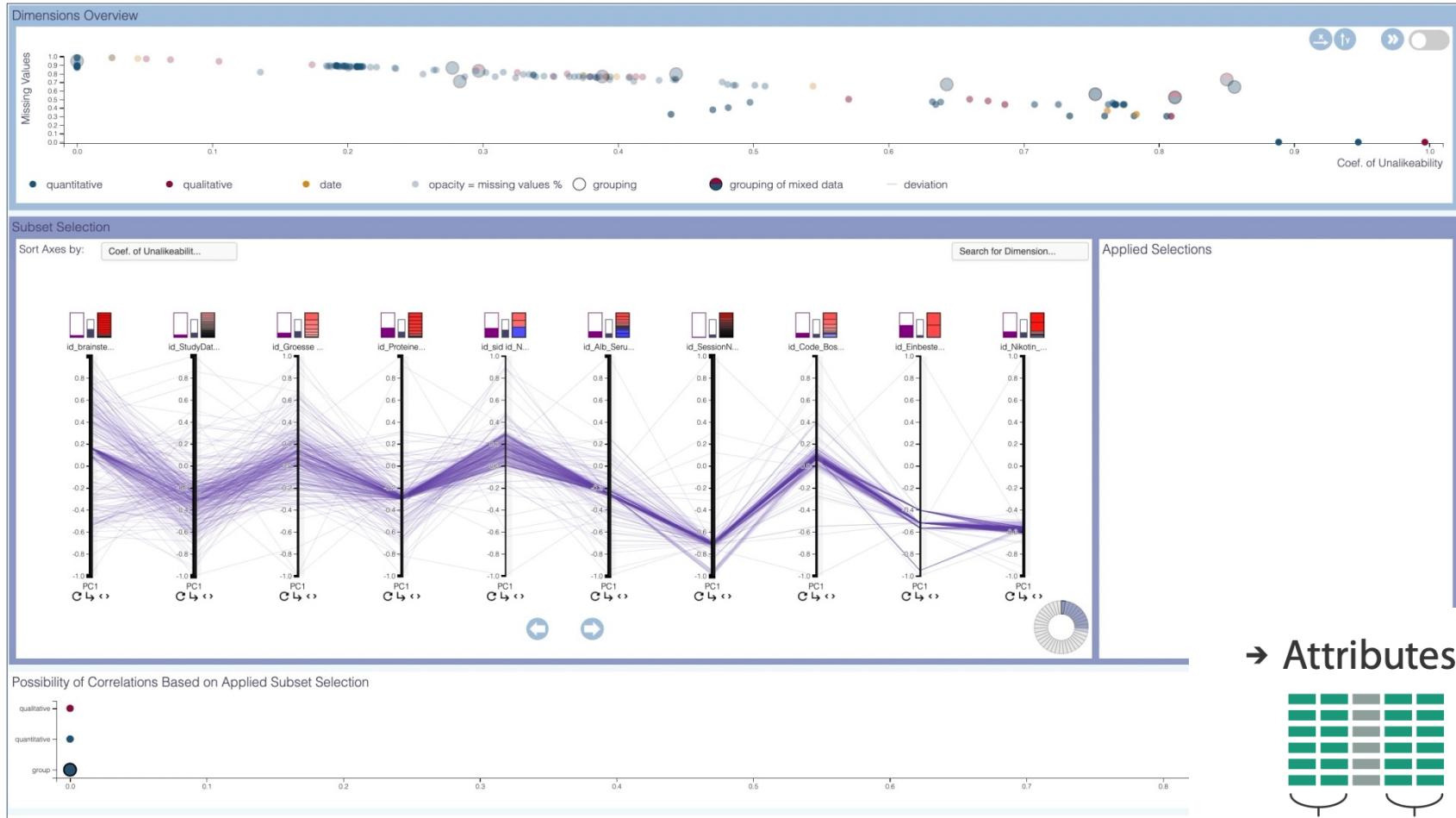
Aggregate: Items



Application of edge bundling on maps

Source: Holten et al. 2009. [Force-Directed Edge Bundling for Graph Visualization](#).

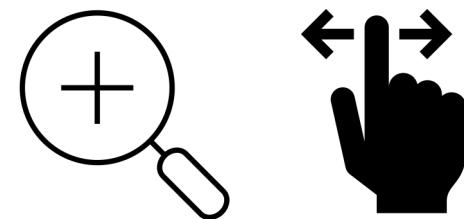
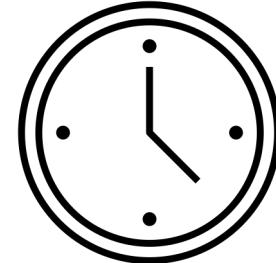
Aggregate Variables (Dimensions), Embed



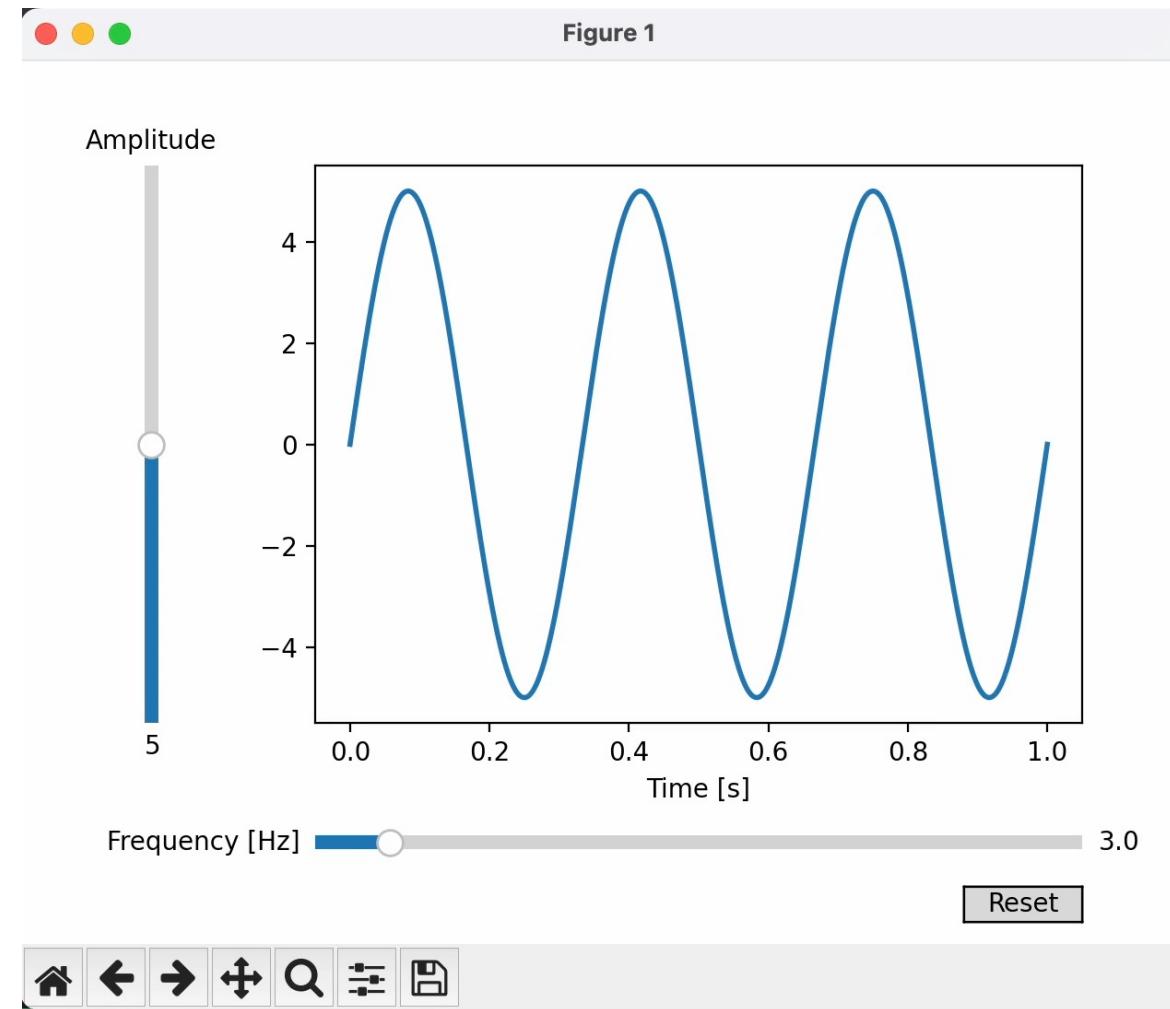
Garrison, L., Müller, J., Schreiber, S., Oeltze-Jafra, S., Hauser, H., & Bruckner, S. (2021). Dimlift: Interactive hierarchical data exploration through dimensional bundling. *IEEE Transactions on Visualization and Computer Graphics*, 27(6), 2908-2922.

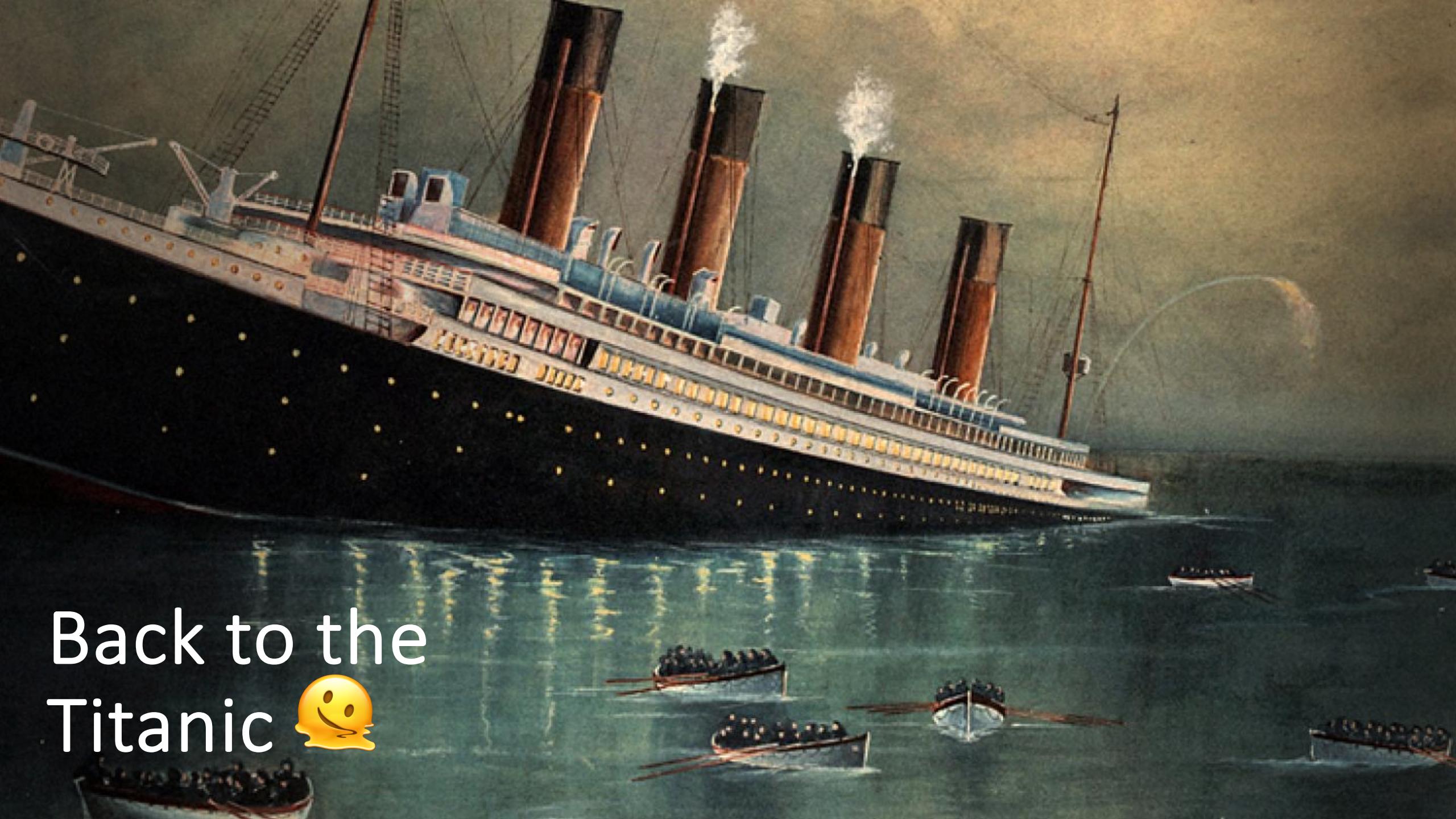
Interaction Caveats!

- Interaction costs **time**
- Cognitive load?
- Users may not interact with your visualization (90%?!?! – Aisch 2016)
- Controls vs “invisible” functionality



Matplotlib Code Study: Param Adjustment

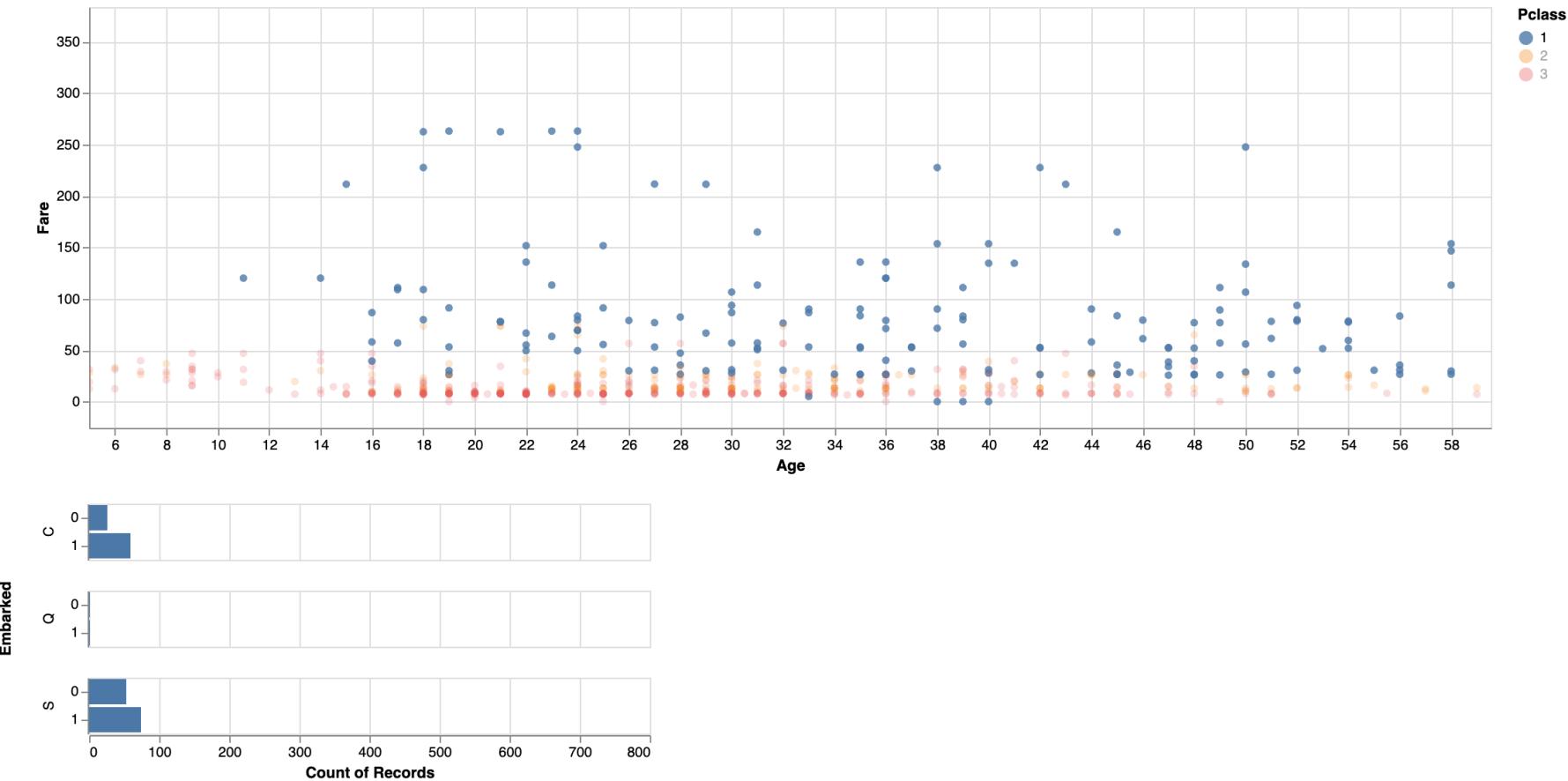




Back to the
Titanic



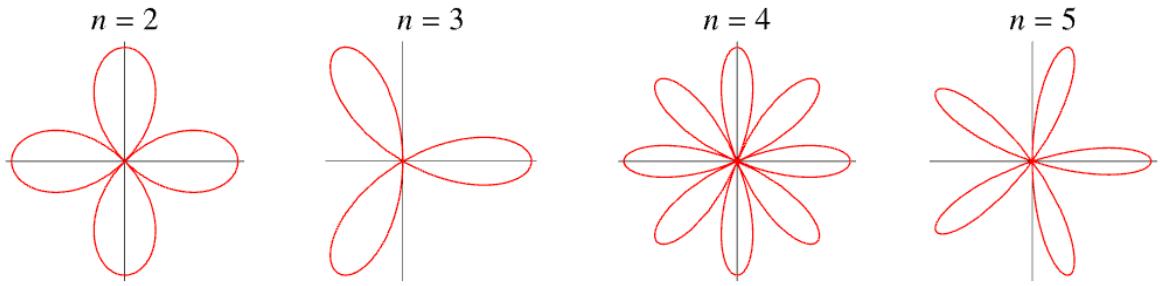
Altair Code Study: Select and Filter



To the lab!

Pick your poison:

1. Interactive parameter adjustment for Rose Curve (<https://mathworld.wolfram.com/RoseCurve.html>) in Matplotlib



2. Interactive visual exploration that continues your questions from earlier this week on the Titanic dataset