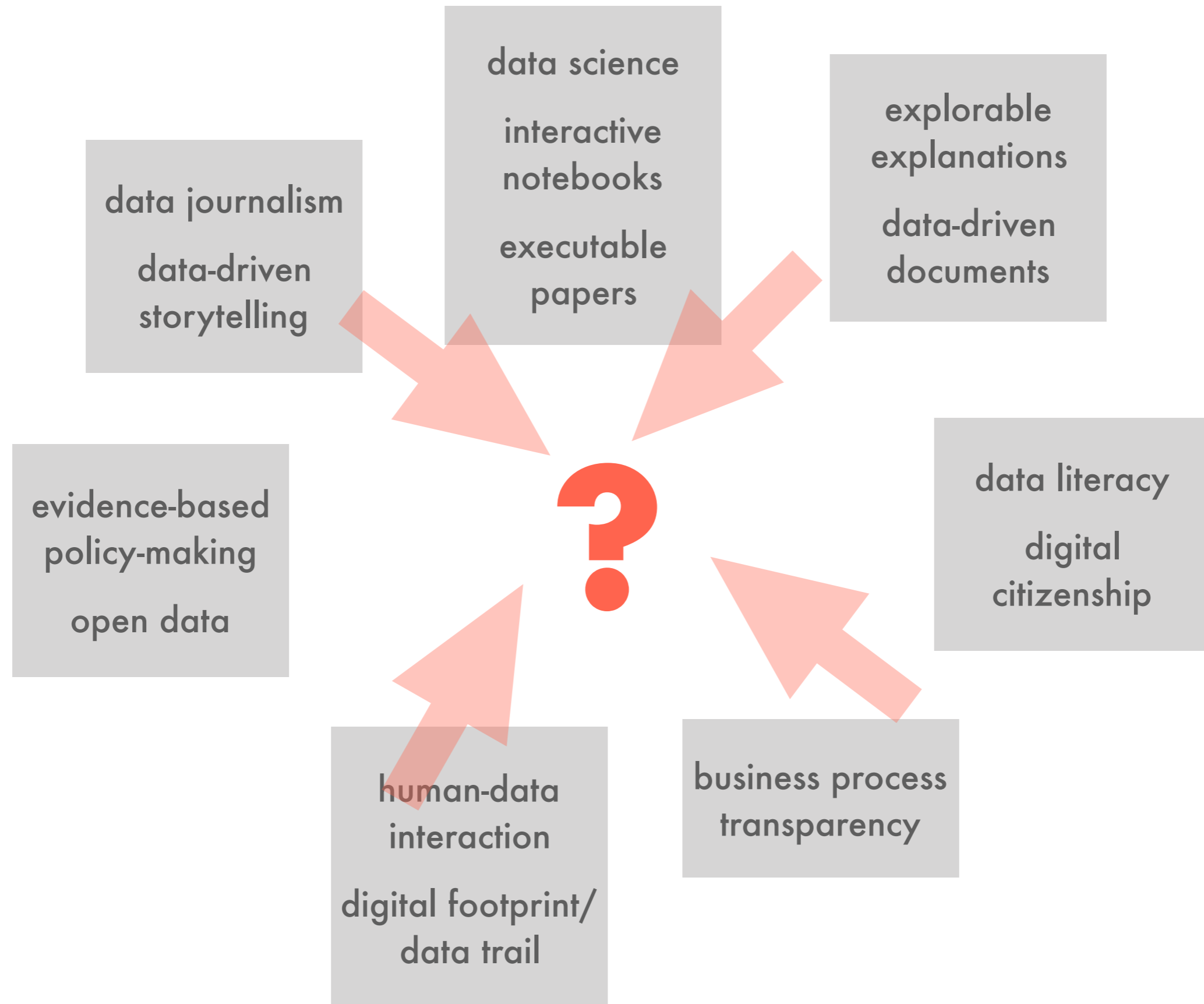


# Self-Explaining Computation with Explicit Change

Roly Perera  
University of Glasgow

# trends towards a data-centric society



# overview

computation growing into a new role

a literate medium for expressing arguments,  
narratives, workflows and ideas

# overview

some desiderata of these new apps:

- human-readable
- transparent, reproducible
- access to *process*, not just outcome
- multiple stakeholders, different concerns
- end-user empowerment

(some tensions in there..)

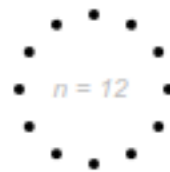
# plan for what follows

1. examples of explorable explanations, interactive notebooks, data journalism
2. some present limitations
3. proposal: *self-explaining computation with explicit change*
4. existing work we will build on

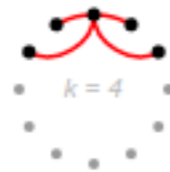
# explorable explanations

**ALGORITHM** To interpolate between regular and random networks, we consider the following random rewiring procedure.

We start with a ring of  $n$  vertices



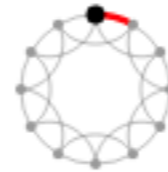
where each vertex is connected to its  $k$  nearest neighbors



like so.



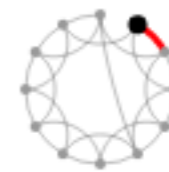
We choose a vertex, and the edge to its nearest clockwise neighbour.



With probability  $p$ , we reconnect this edge to a vertex chosen uniformly at random over the entire ring, with duplicate edges forbidden. Otherwise, we leave the edge in place.



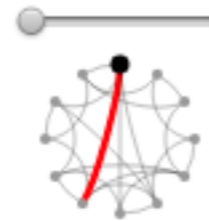
We repeat this process by moving clockwise around the ring, considering each vertex in turn until one lap is completed.



Next, we consider the edges that connect vertices to their second-nearest neighbours clockwise.



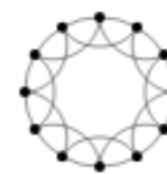
As before, we randomly rewire each of these edges with probability  $p$ .



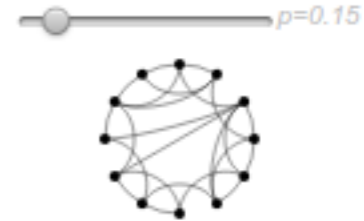
We continue this process, circulating around the ring and proceeding outward to more distant neighbours after each lap, until each original edge has been considered once.

As there are  $nk/2$  edges in the entire graph, the rewiring process stops after  $k/2$  laps.

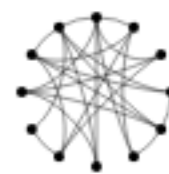
For  $p = 0$ , the ring is unchanged.



As  $p$  increases, the graph becomes increasingly disordered.



At  $p = 1$ , all edges are re-wired randomly.



This construction allows us to 'tune' the graph between regularity ( $p = 0$ ) and disorder ( $p = 1$ ), and thereby to probe the intermediate region  $0 < p < 1$ , about which little is known.

# explorable explanations

Let's look at an example where a machine learning model makes a new type of interface possible. To understand the interface, imagine you're a type designer, working on creating a new font <sup>1</sup>. After sketching some initial designs, you wish to experiment with bold, italic, and condensed variations. Let's examine a tool to generate and explore such variations, from any initial design. For reasons that will soon be explained the quality of results is quite crude; please bear with us.

The image displays a user interface for font exploration, divided into three main sections:

- STARTING FONT:** A vertical list of five rows, each showing the letters 'A B C D E F' followed by an ellipsis. The rows represent different initial font styles: a simple sans-serif, a slightly condensed sans-serif, a bold sans-serif, a serif font, and an italicized serif font. A vertical orange bar highlights the third row (bold sans-serif).
- MODIFICATIONS:** Three horizontal sliders control the font's appearance:
  - BOLD:** The slider is positioned at approximately 25% of the range.
  - ITALIC:** The slider is positioned at approximately 75% of the range.
  - CONDENSED:** The slider is positioned at approximately 25% of the range.
- RESULT:** A grid of the resulting font characters, showing 'A B C D E F G' on the first line, 'H I J K L M N' on the second, 'O P O R S T U' on the third, and 'V W X Y Z' on the fourth. The characters are rendered in a single, bold, italicized, and condensed style.

# explorable explanations

Imagine if Blinder's proposal in the New York Times were written like this:

Say we allocate **\$3.0 billion** for the following program: Car-owners who trade in an old car that gets less than **17 MPG**, and purchase a new car that gets better than **24 MPG**, will receive a **\$3,500** rebate.

We estimate that this will get **828,571** old cars off the road. It will save **1,068 million gallons** of gas (or **68 hours** worth of U.S. gas consumption.) It will avoid **9.97 million tons** CO<sub>2</sub>e, or **0.14%** of annual U.S. greenhouse gas emissions.

The abatement cost is **\$301** per ton CO<sub>2</sub>e of federal spending, although it's **-\$20** per ton CO<sub>2</sub>e on balance if you account for the money saved by consumers buying less gas.

calculations for "cars traded" (you can change assumptions in green)

budget = \$3.0 billion

rebate = \$3,500

overhead = \$100 million

Assume that the program "sells out", and all available rebates are collected. Given the demand for new cars, this will be true for any reasonable rebate amount.

$$\text{cars traded} = (\text{budget} - \text{overhead}) / \text{rebate}$$
$$= 828,571$$

drag the number  
left or right

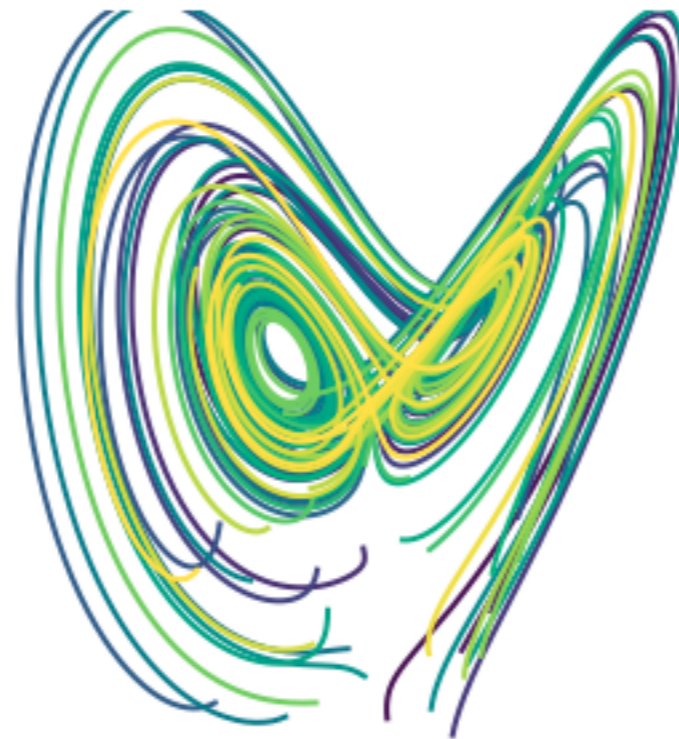


# interactive notebooks

Let's change  $(\sigma, \beta, \rho)$  with ipywidgets and examine the trajectories.

```
In [10]: from lorenz import solve_lorenz  
w=interactive(solve_lorenz,sigma=(0.0,30.0),rho=(0.0,50.0))  
w
```

sigma  3.10  
beta  1.13  
rho  36.30



# interactive notebooks

## > Number of harmonics



Sine waves are added to create a sawtooth wave

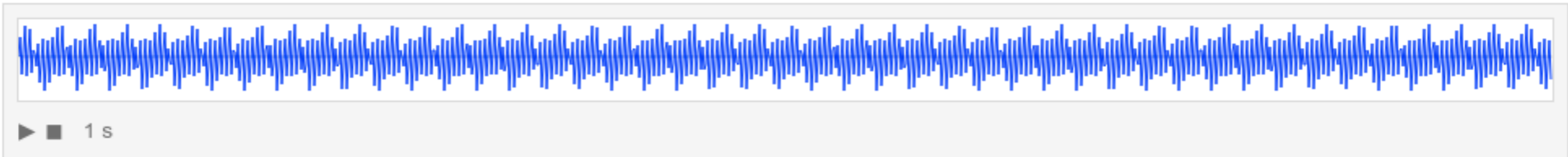
$$f(t) = -\frac{2}{\pi} \left( -\frac{\sin(1 \cdot \text{freq} \cdot t \cdot 2\pi)}{1} + \frac{\sin(2 \cdot \text{freq} \cdot t \cdot 2\pi)}{2} - \frac{\sin(3 \cdot \text{freq} \cdot t \cdot 2\pi)}{3} + \frac{\sin(4 \cdot \text{freq} \cdot t \cdot 2\pi)}{4} - \frac{\sin(5 \cdot \text{freq} \cdot t \cdot 2\pi)}{5} \right)$$

```
GenerateTex('f(t) = -\\frac{2}{\\pi}(', (i) => {  
  const sign = i % 2 == 1 ? '+' : '-';  
  return `${sign} \\frac{\\sin(${i + 1} \\cdot \\textit{freq} \\cdot t \\cdot 2 \\pi)}{${i + 1}}`;  
}, sawtoothHarmonics, ')')
```



> Let's hear how the sawtooth wave sounds as it is being formed from sine waves (drag the slider above to generate the waveform with more harmonics):

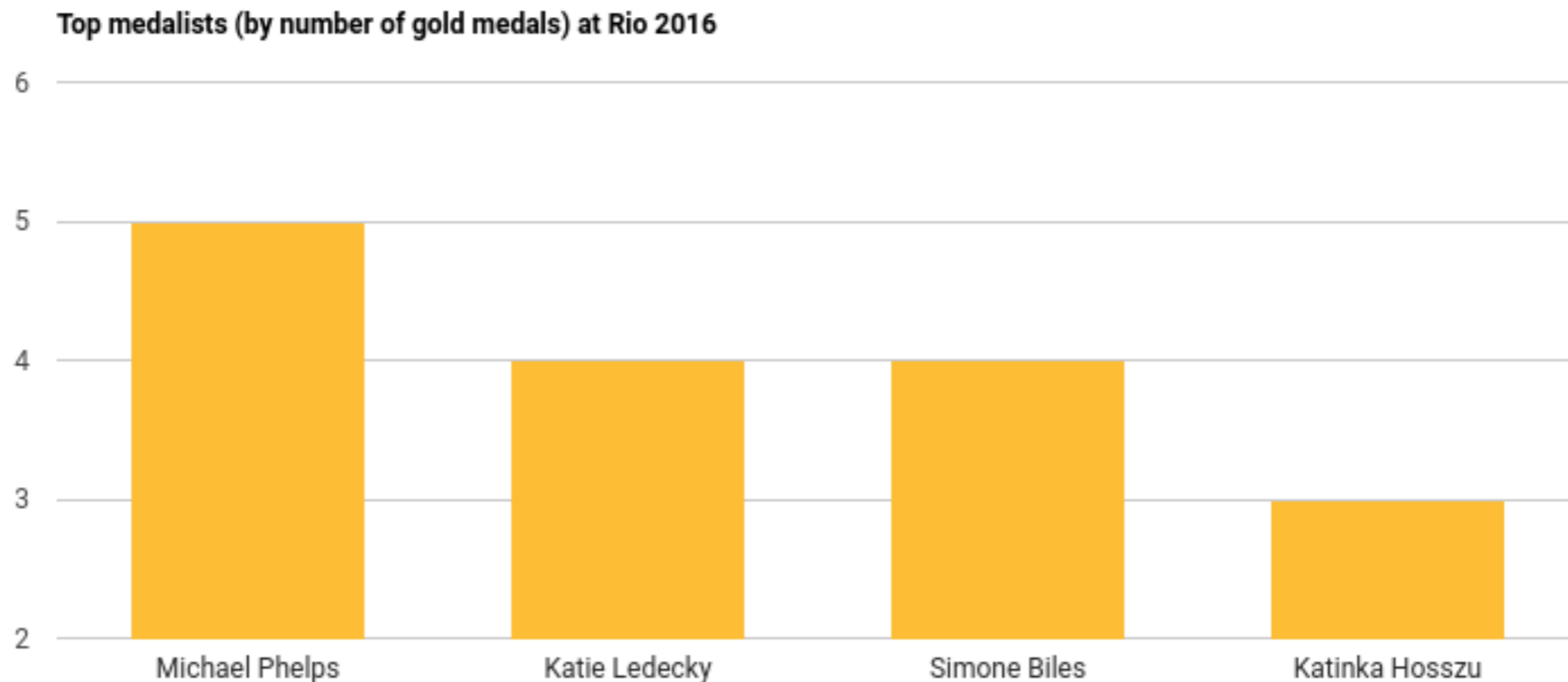
> *Sawtooth formed from 5 sine waves*



# data-driven storytelling

```
let data =  
  olympics  
  . 'filter data' . 'Games is' . 'Rio (2016)' . then . 'group data' . 'by Athlete' . 'sum Gold' . then  
  . 'sort data' . 'by Gold descending' . then . paging . take(4)  
  . 'get series' . 'with key Athlete' . 'and value Gold'  
  
chart.column(data).legend(position="none")  
  .set(fontName="Roboto", fontSize=12, colors=["#F4C300"], title="Top medalists (by number of gold medals) at Rio 2016")
```

preview



# some limitations

An exciting new direction in content creation, dissemination and pedagogy

But still quite *ad hoc*:

- transparency only partial or pre-set
- coarse-grained data/view relationship

# some limitations

Distill-style essays and explorable explanations typically fix in advance what can be explored

- not fully transparent/open

# some limitations

Notebooks are more open: code is inline in the document

- not easy to see all intermediate results
- not clear how code relates to data/views

# proposed research

## Self-Explaining Computation with Explicit Change

build on techniques from self-explaining and incremental computation to make explicit:

- how *parts* relate to *other parts*
- how *changes* cause *other changes*

# proposed research

## Three main use-cases:

#1

how?

easily view or hide  
any subcomputation  
or intermediate  
result

#2

whence?

understand how  
views relate to  
data in a fine-  
grained way

#3

what if?

change data or  
code and see  
fine-grained  
consequences

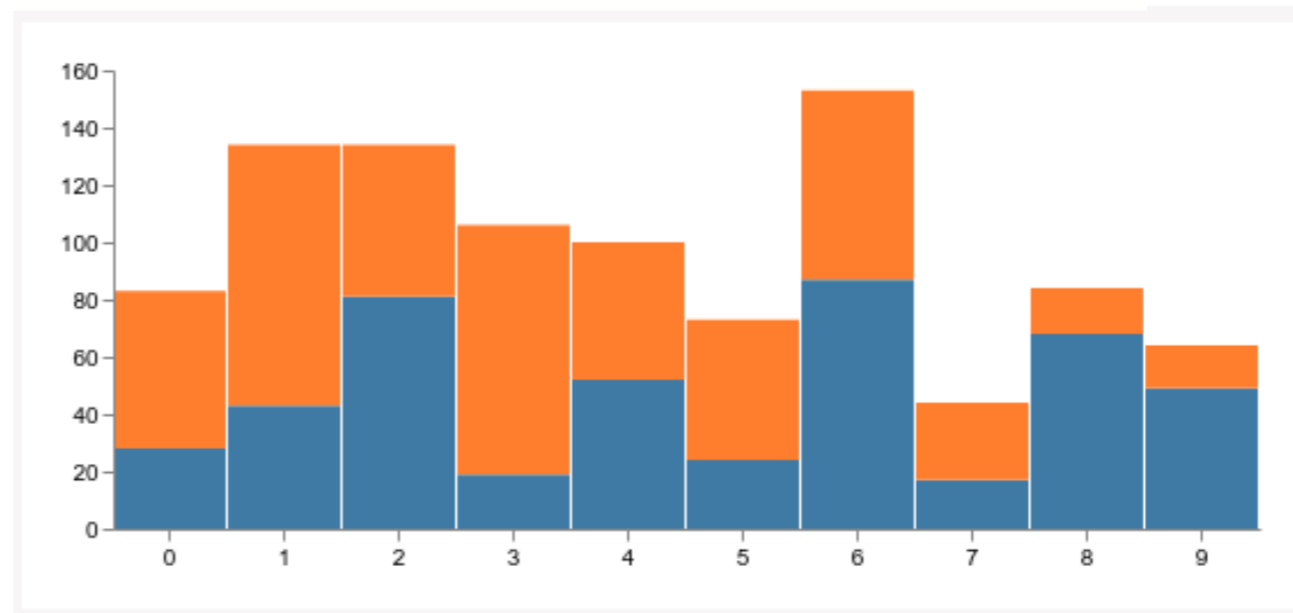


# Use cases

# #2

whence?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

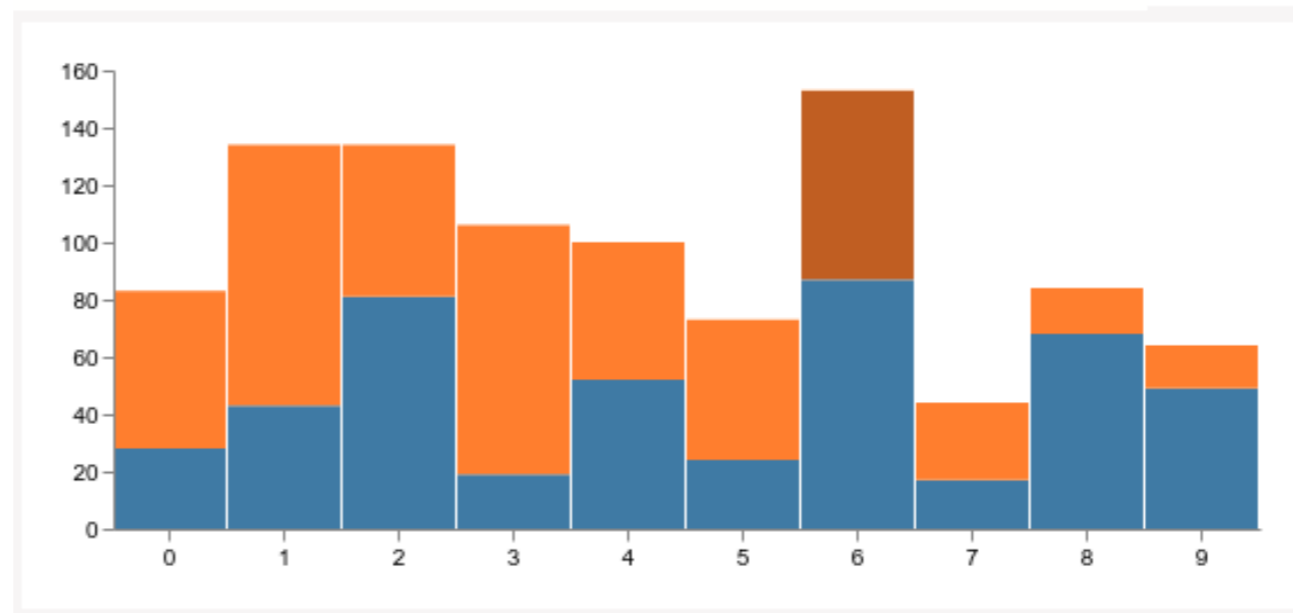


# Use cases

# #2

whence?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

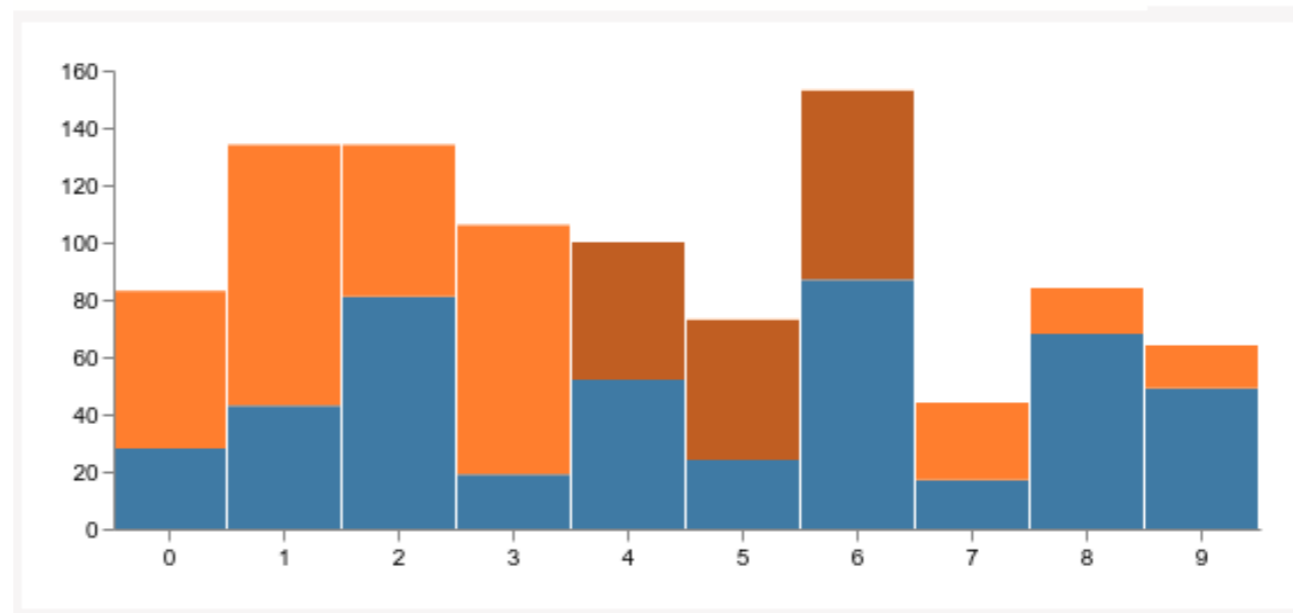


# Use cases

# #2

whence?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

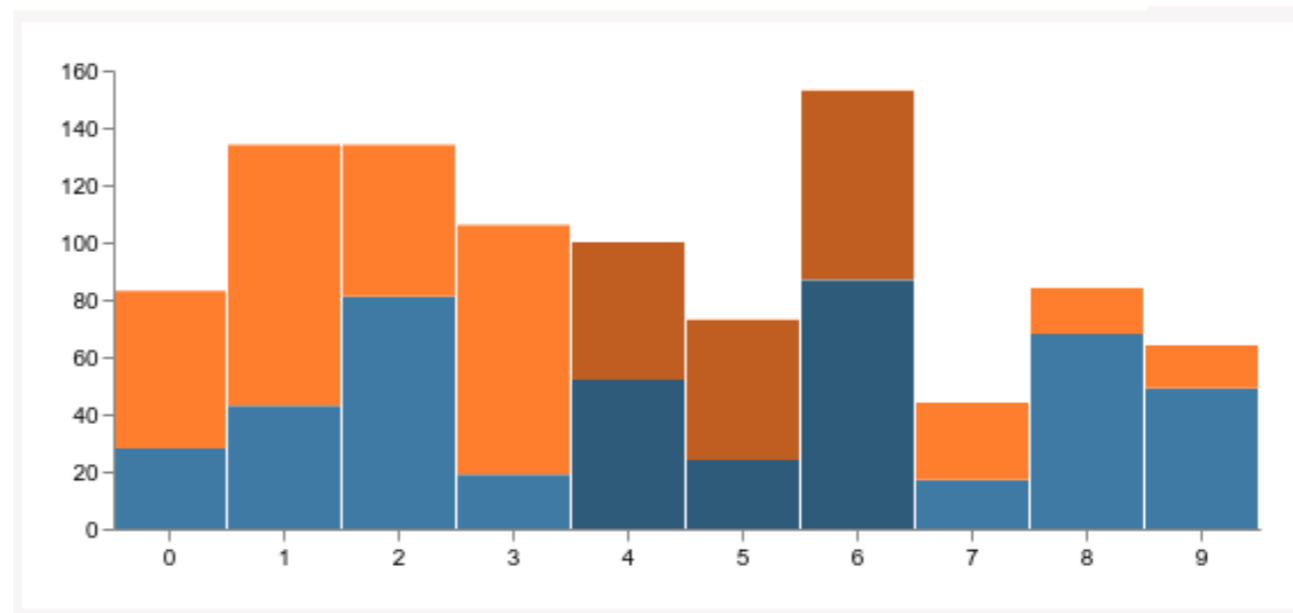


# Use cases

# #2

whence?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

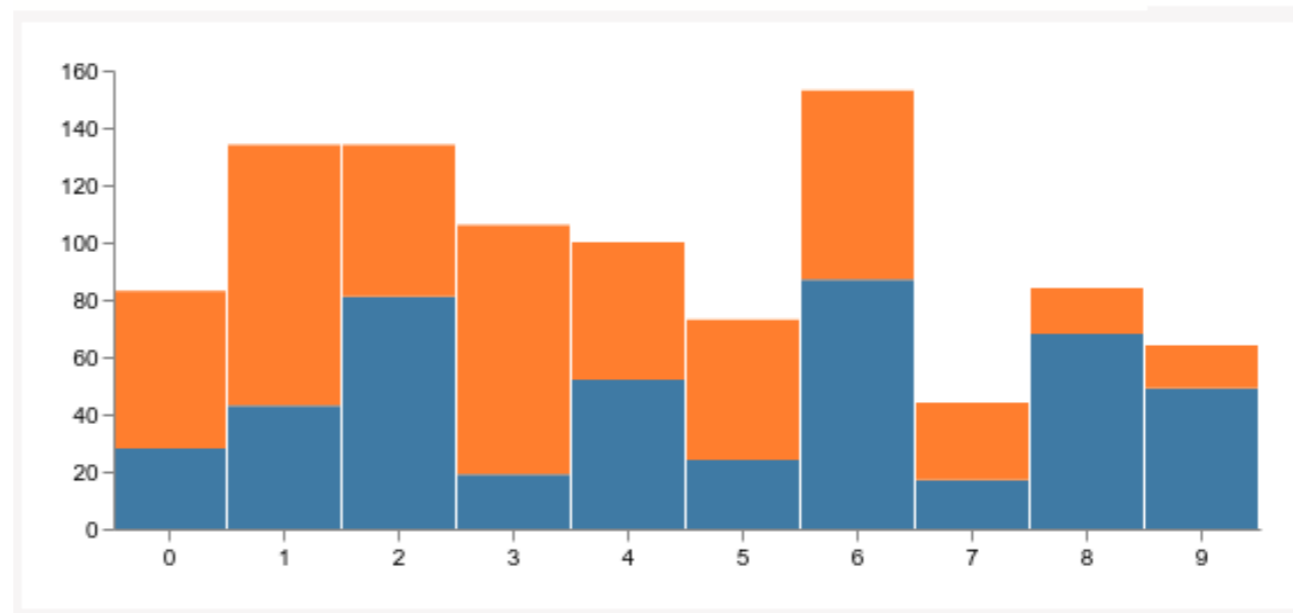


# Use cases

# #2

whence?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c": 0}, {"x": 0, "y": 55, "c": 1},  
      {"x": 1, "y": 43, "c": 0}, {"x": 1, "y": 91, "c": 1},  
      {"x": 2, "y": 81, "c": 0}, {"x": 2, "y": 53, "c": 1},  
      {"x": 3, "y": 19, "c": 0}, {"x": 3, "y": 87, "c": 1},  
      {"x": 4, "y": 52, "c": 0}, {"x": 4, "y": 48, "c": 1},  
      {"x": 5, "y": 24, "c": 0}, {"x": 5, "y": 49, "c": 1},  
      {"x": 6, "y": 87, "c": 0}, {"x": 6, "y": 66, "c": 1},  
      {"x": 7, "y": 17, "c": 0}, {"x": 7, "y": 27, "c": 1},  
      {"x": 8, "y": 68, "c": 0}, {"x": 8, "y": 16, "c": 1},  
      {"x": 9, "y": 49, "c": 0}, {"x": 9, "y": 15, "c": 1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

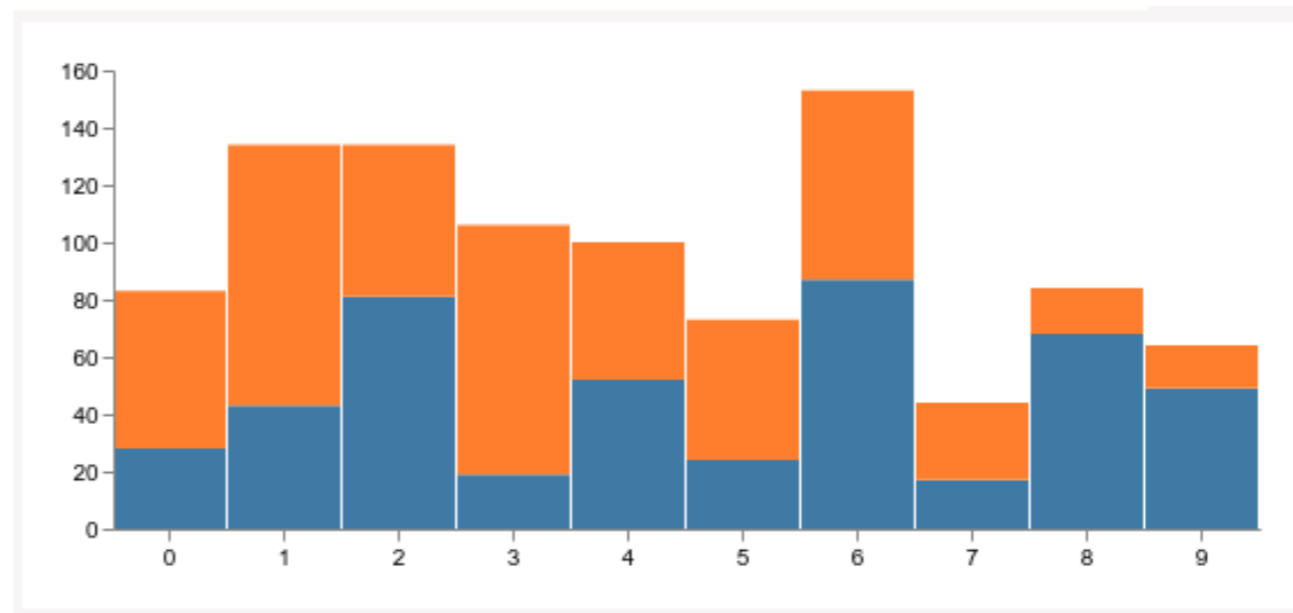


# Use cases

# #3

what if?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

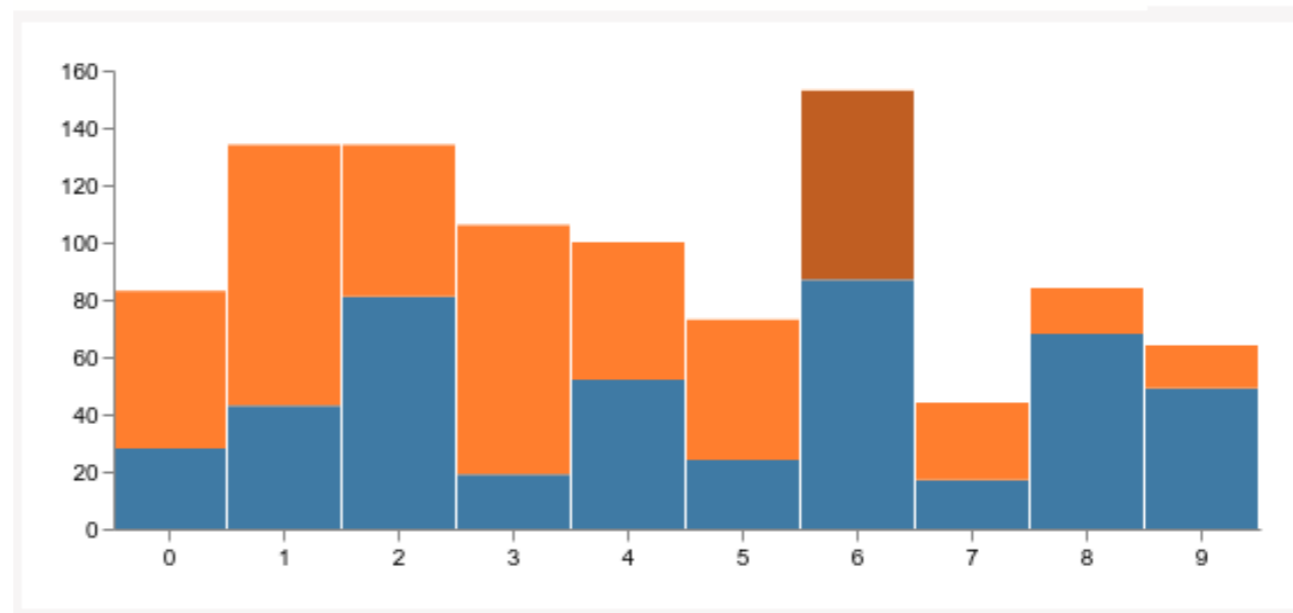


# Use cases

# #3

what if?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 27, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```

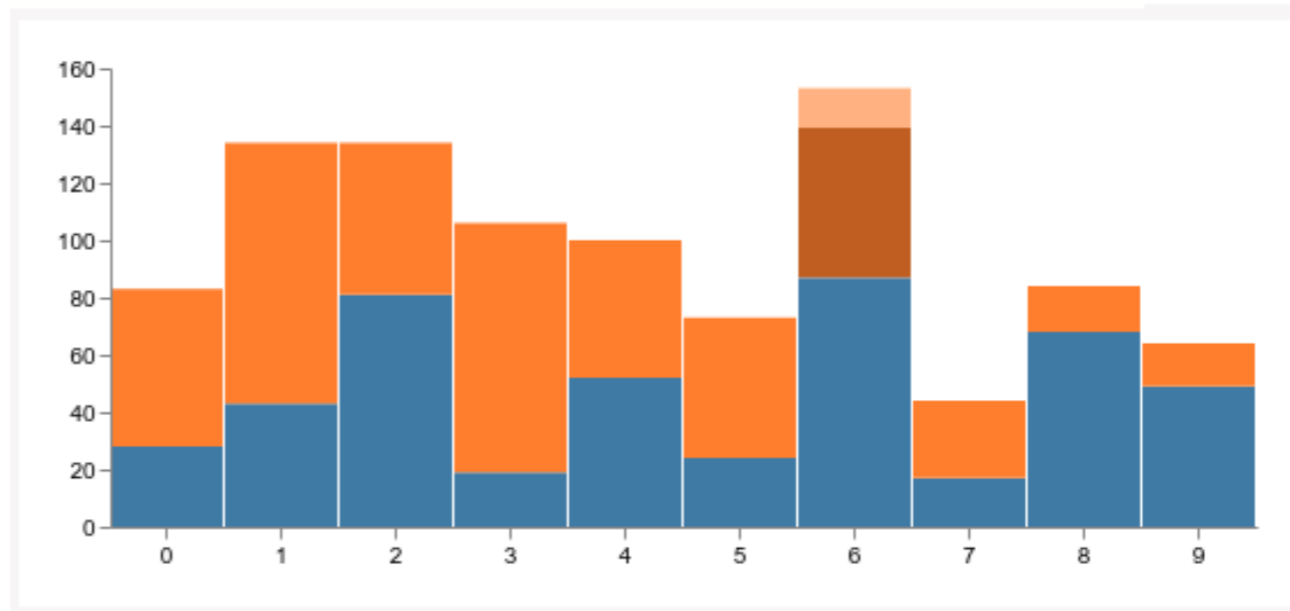


# Use cases

# #3

what if?

```
"data": [  
  {  
    "name": "table",  
    "values": [  
      {"x": 0, "y": 28, "c":0}, {"x": 0, "y": 55, "c":1},  
      {"x": 1, "y": 43, "c":0}, {"x": 1, "y": 91, "c":1},  
      {"x": 2, "y": 81, "c":0}, {"x": 2, "y": 53, "c":1},  
      {"x": 3, "y": 19, "c":0}, {"x": 3, "y": 87, "c":1},  
      {"x": 4, "y": 52, "c":0}, {"x": 4, "y": 48, "c":1},  
      {"x": 5, "y": 24, "c":0}, {"x": 5, "y": 49, "c":1},  
      {"x": 6, "y": 87, "c":0}, {"x": 6, "y": 66, "c":1},  
      {"x": 7, "y": 17, "c":0}, {"x": 7, "y": 53, "c":1},  
      {"x": 8, "y": 68, "c":0}, {"x": 8, "y": 16, "c":1},  
      {"x": 9, "y": 49, "c":0}, {"x": 9, "y": 15, "c":1}  
    ],  
    "transform": [  
      {  
        "type": "stack",  
        "groupby": ["x"],  
        "sort": {"field": "c"},  
        "field": "y"  
      }  
    ]  
  }  
],
```





# technical goals/innovations

## Theoretical

extend work on self-explaining computation to accommodate code and data changes

*parts of explanations explain parts of data*

*changes to explanations explain changes to data*

## Practical

extend these partial and incremental computation ideas to user interfaces

viz components which support slicing and delta-visualisation

make explanations accessible directly from data views

# prior work

## incremental computation

incremental relational algebra  
self-adjusting computation  
incremental lambda calculus

Griffin, Libkin, Trickey (1997)  
Horn, Perera, Cheney (2018)

Acar (2005)  
Hammer (2012)

Cai *et al* (2013)

## provenance & program slicing

data provenance  
self-explaining computation

Buneman, Khanna, Tan (2001)  
Cheney, Chiticariu, Tan (2009)

Perera *et al* (2012)  
Cheney, Acar, Perera (2013)  
Cheney, Ahmed, Acar (2014)

## notebooks & data-driven storytelling

Wrattler, The Gamma

Petricek (2017)  
Petricek, Geddes, Sutton (2018)

# target audience

easier/earlier



educators, students

authors of Distill-style  
exposition papers

data journalists,  
science journalists

R or Python based  
data scientists

science 2.0 publishers

harder/later



# summary

emerging new role of computation

a literate medium for expressing arguments,  
narratives, workflows and ideas

# summary

static documents superceded by interactive  
views of real-world processes

James Somers. *The Scientific Paper Is  
Obsolete*. The Atlantic, April, 2018

for computations to *explain* it needs to be clear  
how parts relate to other parts

and how changes cause other changes