



Doing Data Visualisation in Python with Altair, A Primer

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Associate Professor @ CIM at University of Warwick

Senior Lecturer in at giCentre, City, University of London (2014 - 2019)

PhD in Visualisation, University of Bergen, NO

MSc & BSc in Computer Science (Sabanci University, TR & METU, TR)

Research on ...

Methods: Interactive visual analytics techniques

Integrating **Computational Tools** and **Visualisation**

New **interaction** forms (e.g., natural lang. + vis)

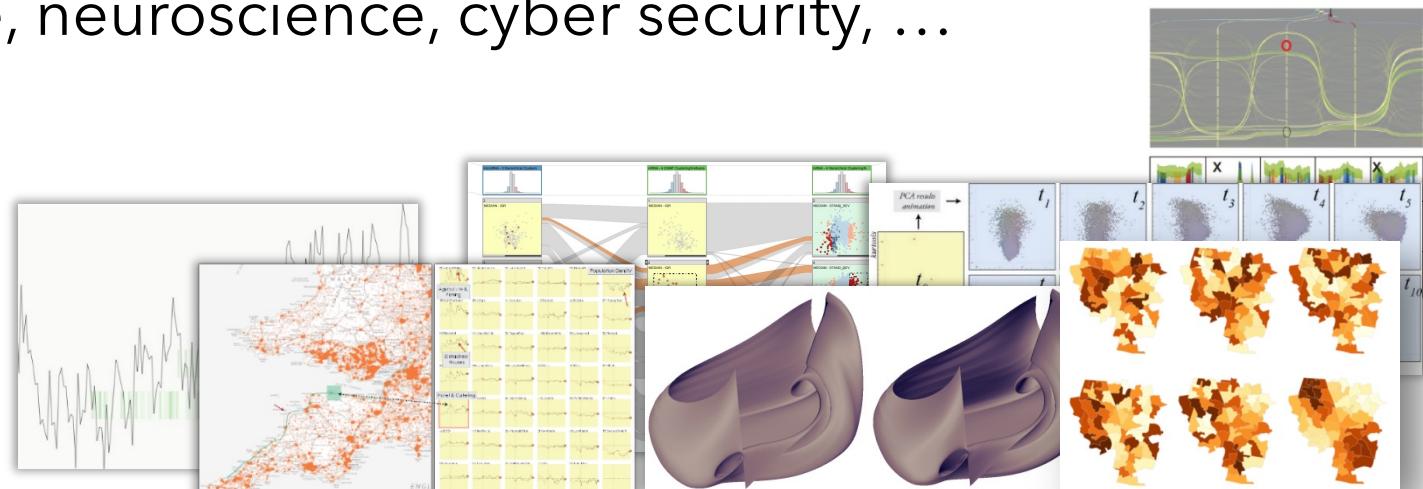
Theoretical Foundations: Perceptual/Cognitive mechanisms of Visualisation

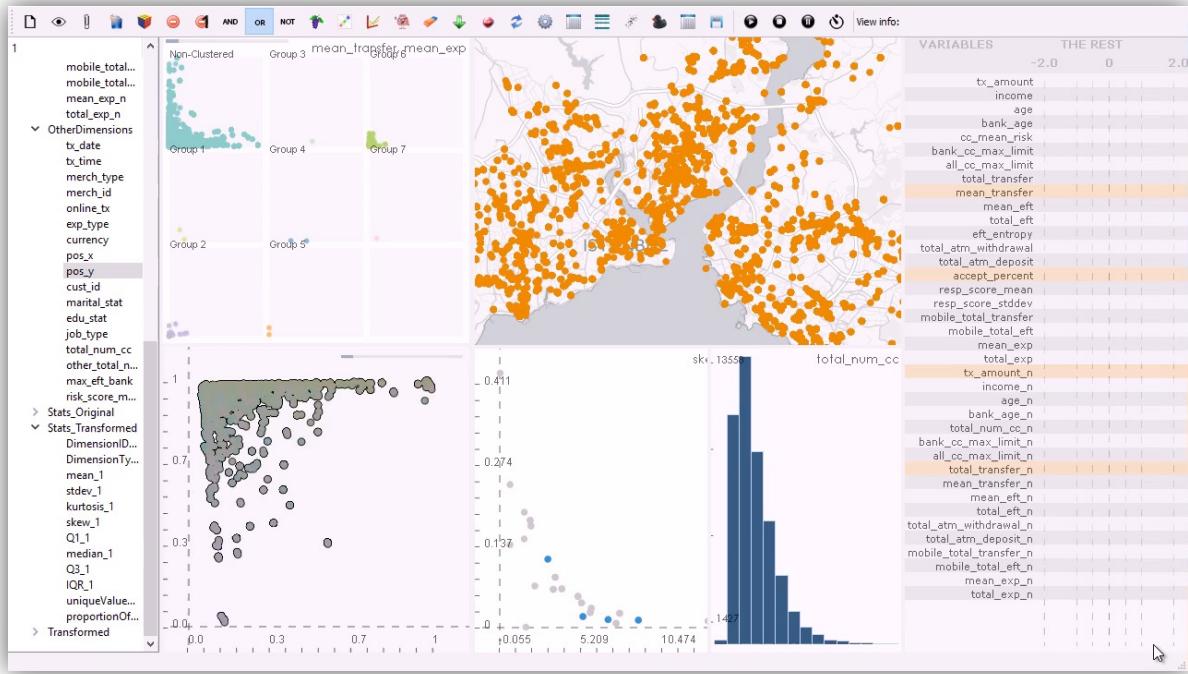
Applications in ...

Biology, transport, intelligence, neuroscience, cyber security, ...

Teaching on ...

(Visual) Data Science





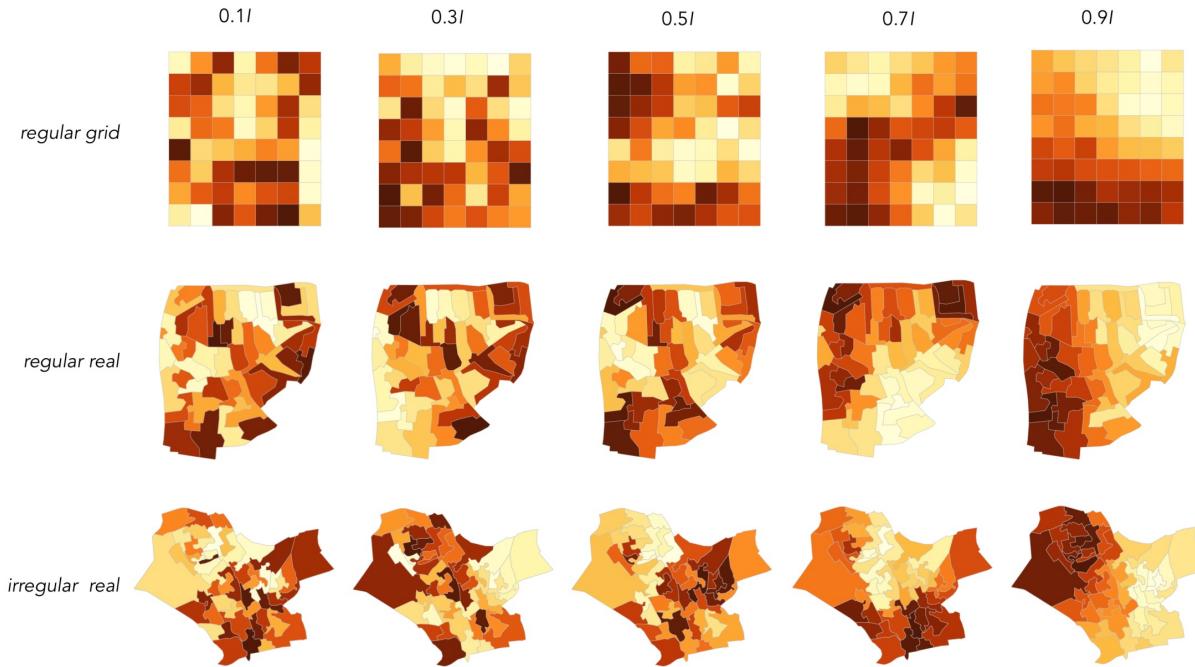
Enhancing a Social Science Model-building Workflow with Interactive Visualisation

Turkay, C., Slingsby, A., Lahtinen, K., Butt, S., & Dykes, J., ESANN 2016 (& Neurocomputing 2017)

Designing Progressive and Interactive Analytics Processes for High-Dimensional Data Analysis

Turkay, C., Kaya, E., Balcisoy, S., Hauser, H., IEEE TVCG 2017



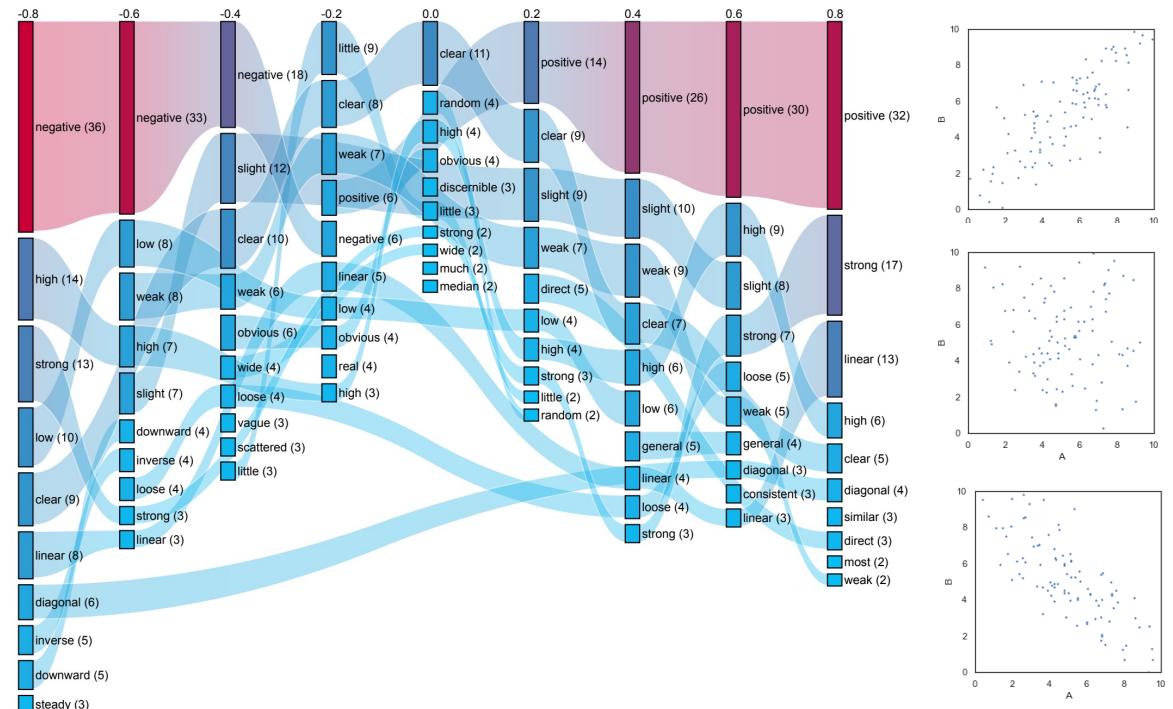


Map LineUps: effects of spatial structure on graphical inference

Beecham, R., Dykes, J., Meulemans, W., Slingsby, A., Turkay, C. & Wood, J. (2016).
IEEE Transactions on Visualization and Computer Graphics.

Words of Estimative Correlation: Studying Verbalizations of Scatterplots

Henkin, R., Turkay, C.(2021).
IEEE Transactions on Visualization and Computer Graphics.

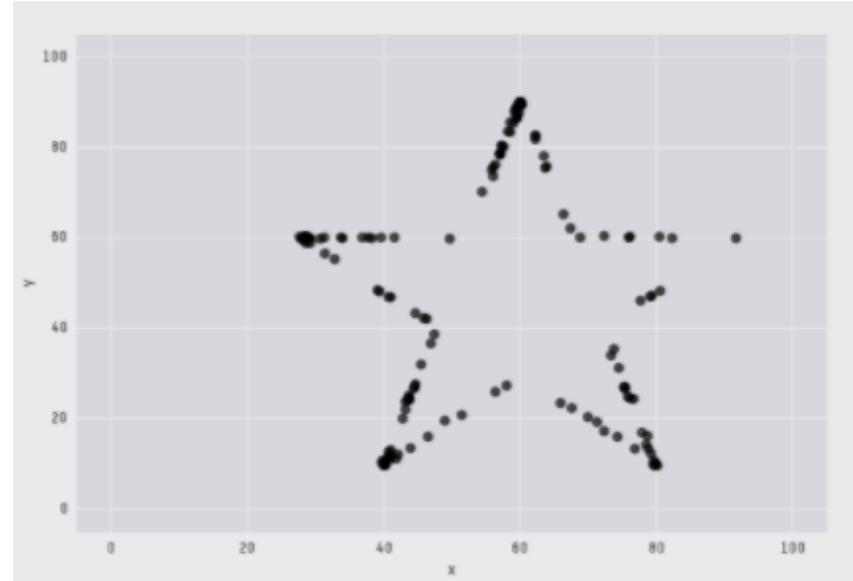


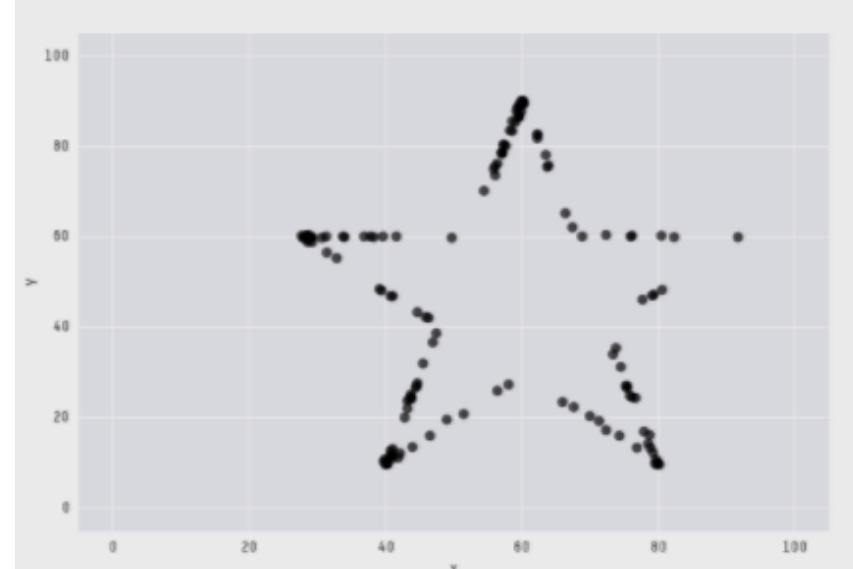
Today?

- A brief intro to visualization
 - Elements of a visualization
 - Channels and marks
 - Grammar of visualization (which motivated Altair et al)
- Altair primer
 - Hands on with Altair
 - Basic plotting with examples for further exploration

WHY VISUALISE?

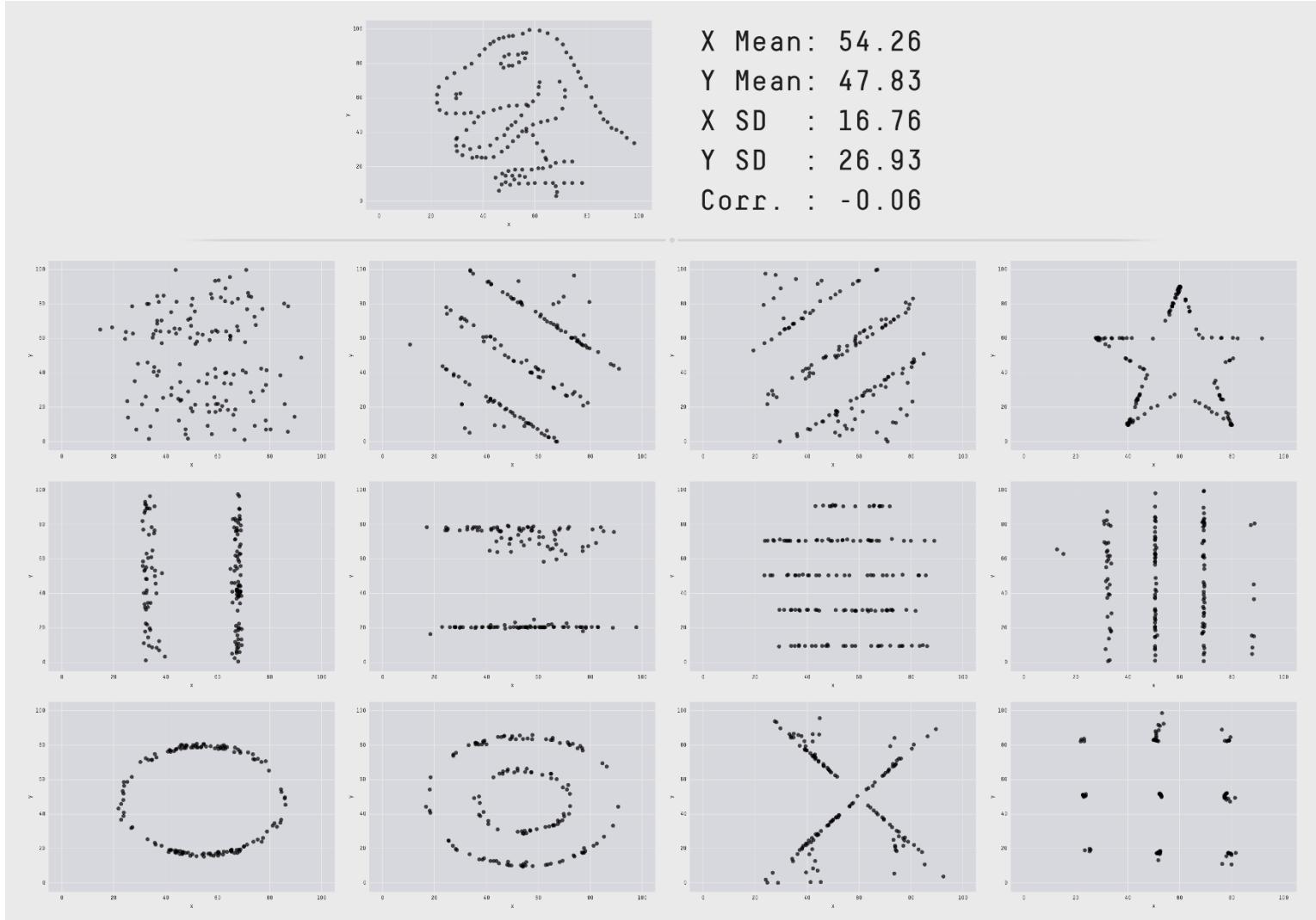
Do you see any similarities?





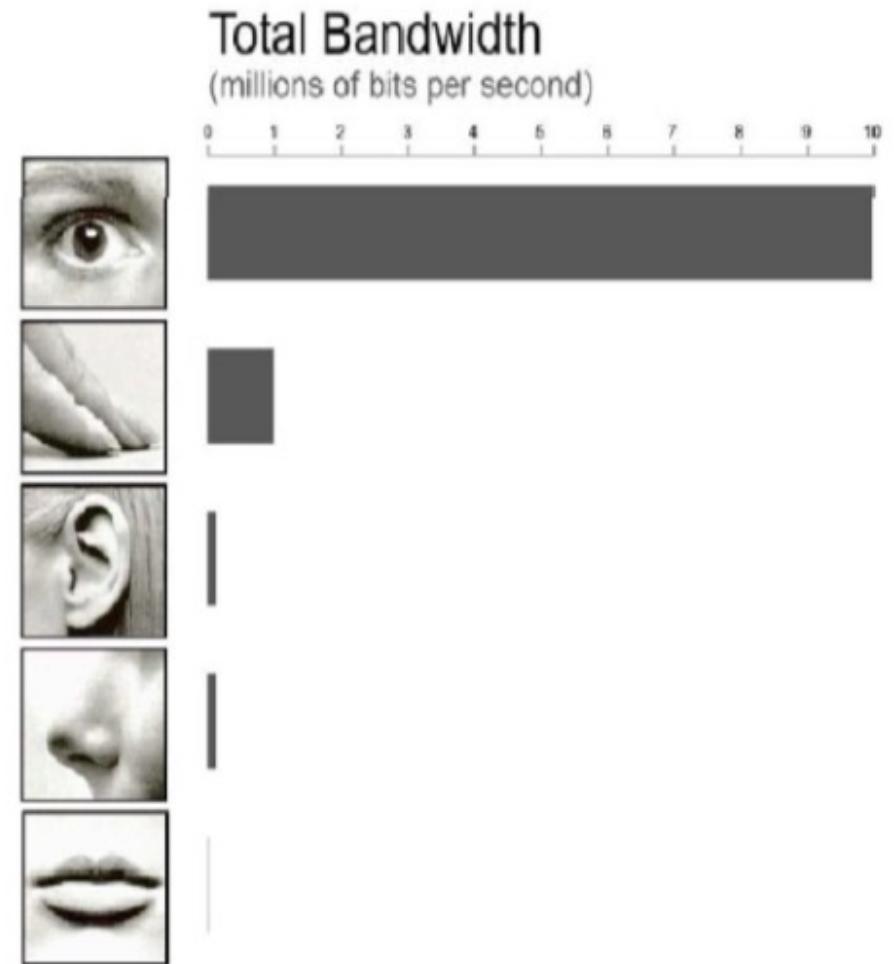
X Mean: 54.26
Y Mean: 47.83
X SD : 16.76
Y SD : 26.93
Corr. : -0.06

The Datasaurus Dozen



Why Visual?

- Figures are **richer**; provide more information with less clutter and in less space.
- Figures provide the *gestalt* effect: they give an overview; **make structure more visible**.
- Figures are more **accessible**, easier to understand, **faster to grasp**, more **comprehensible**, more **memorable**, more **fun**, and **less formal**.



List & slide adapted from: [Stasko et al. 1998 & Alexander Lex]

E N C O D I N G

D E C O D I N G

A C T I O N

$x_1, y_1,$
 $x_2, y_2 \dots$



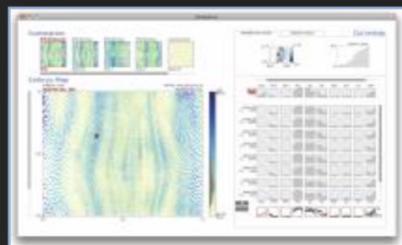
Engage



“Wow, X & Y
looks
amazing!”

“I need to
find out
more!”

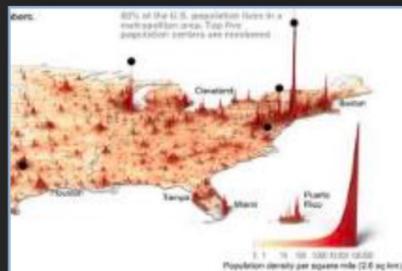
Explore



“I wonder how
x relates to y”

“Maybe z is
important?”

Explain

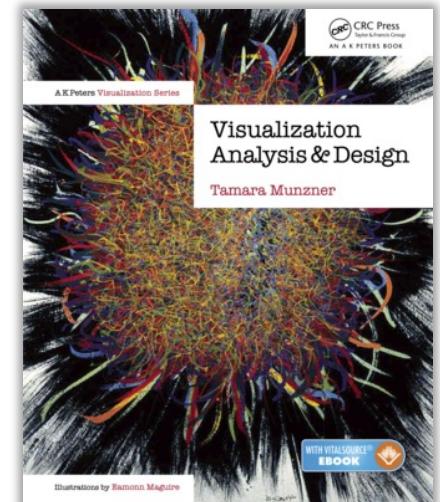


“X does y”

“if I do x
then...”

Visualisation?

*“Computer-based visualization systems provide visual representations of **datasets** designed to help **people** carry out **tasks** more effectively.”* [Tamara Munzner, 2014]



BUILDING BLOCKS OF VISUALISATIONS

VISUAL ELEMENTS

Marks & Channels

Visual Marks: represent **items** (or **links**)

Channels: change **appearance** based on **attribute**

Channel = Visual Variable

Basic visual elements

VISUAL MARKS
i.e., primitives

→ Points



→ Lines



→ Areas



→ Position

→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



VISUAL CHANNELS
i.e., how they look

→ Size

→ Length



→ Area

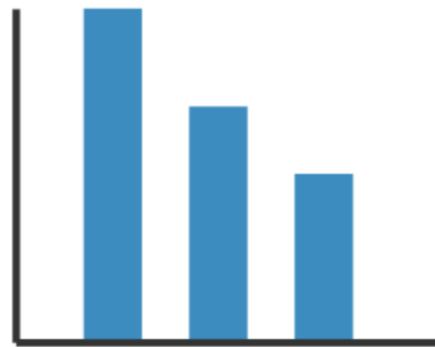


→ Volume



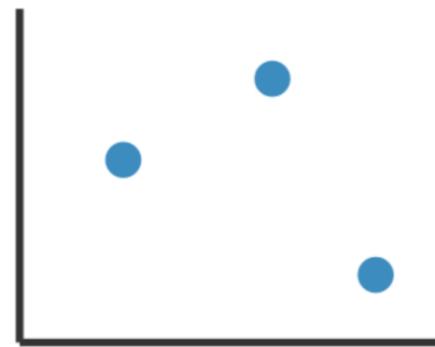
Visual encoding

Def. Representing information visually as combination of marks and channels



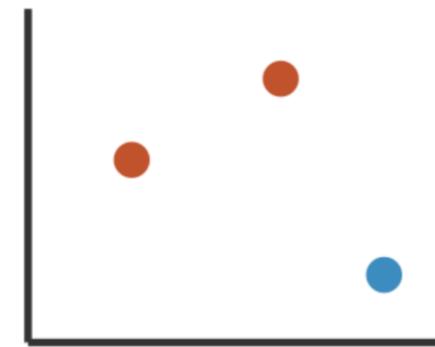
1:
vertical position

mark: line



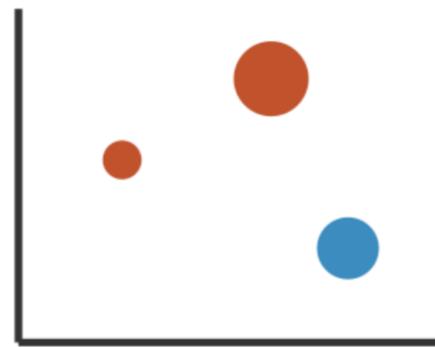
2:
vertical position
horizontal position

mark: point



3:
vertical position
horizontal position
color hue

mark: point



4:
vertical position
horizontal position
color hue
size (area)

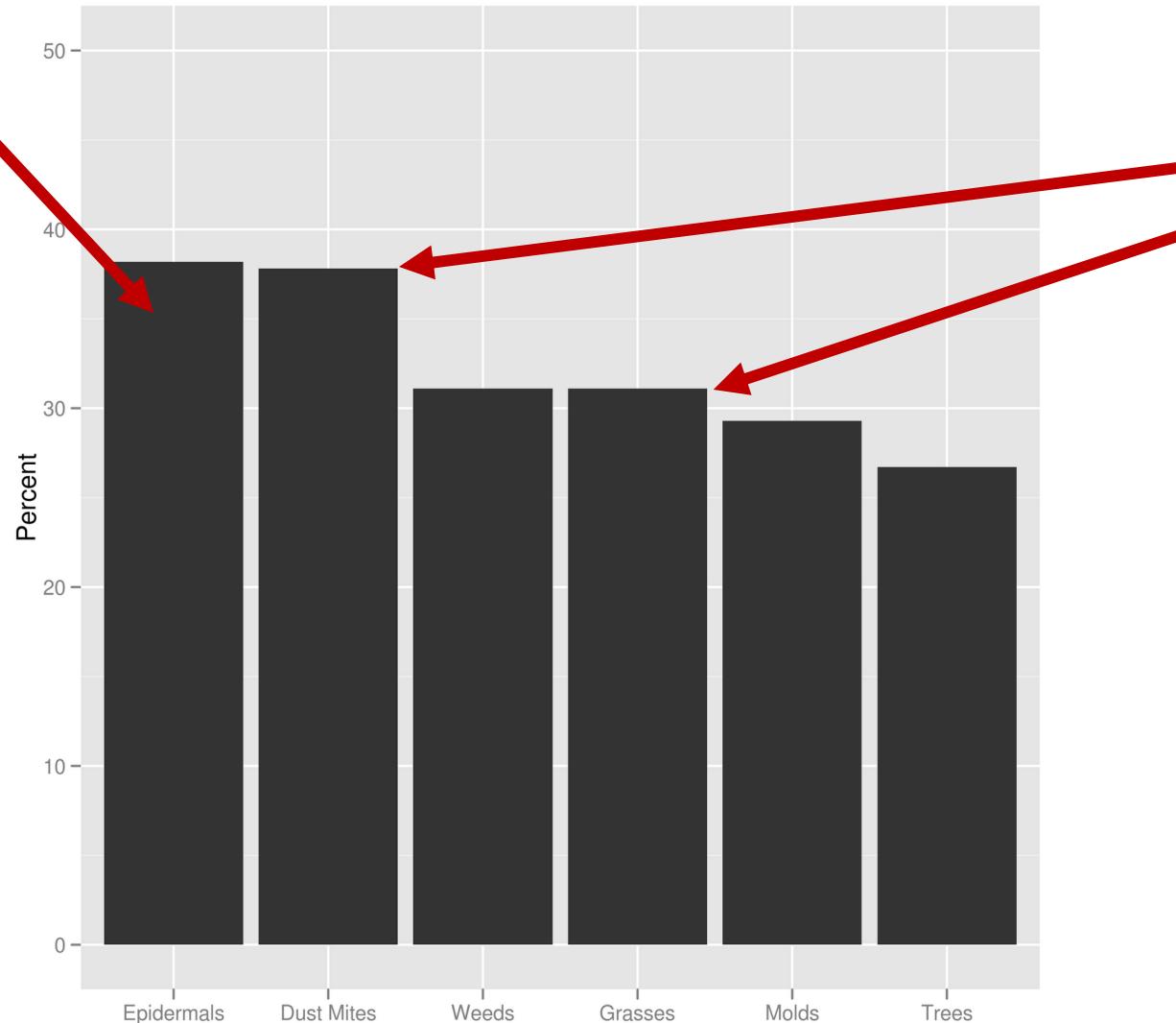
mark: point

Basic visual elements in action

VISUAL MARK

lines (thick ones)

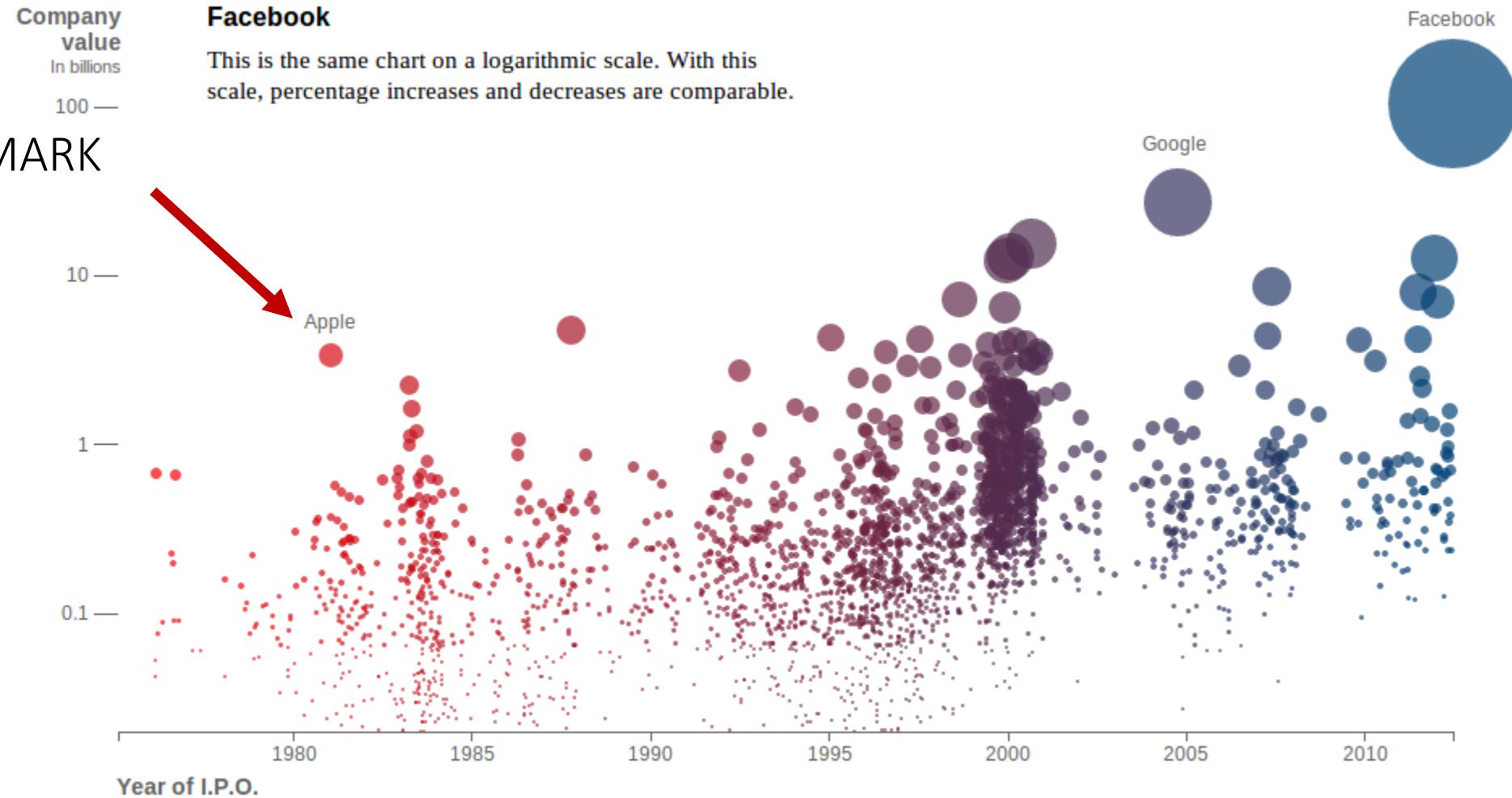
VISUAL CHANNEL
length (of bars)



VISUAL CHANNELS
position x and y
colour
Area (redundant)

Basic visual elements in action

VISUAL MARK
points



FIND THE ODD ONE OUT!

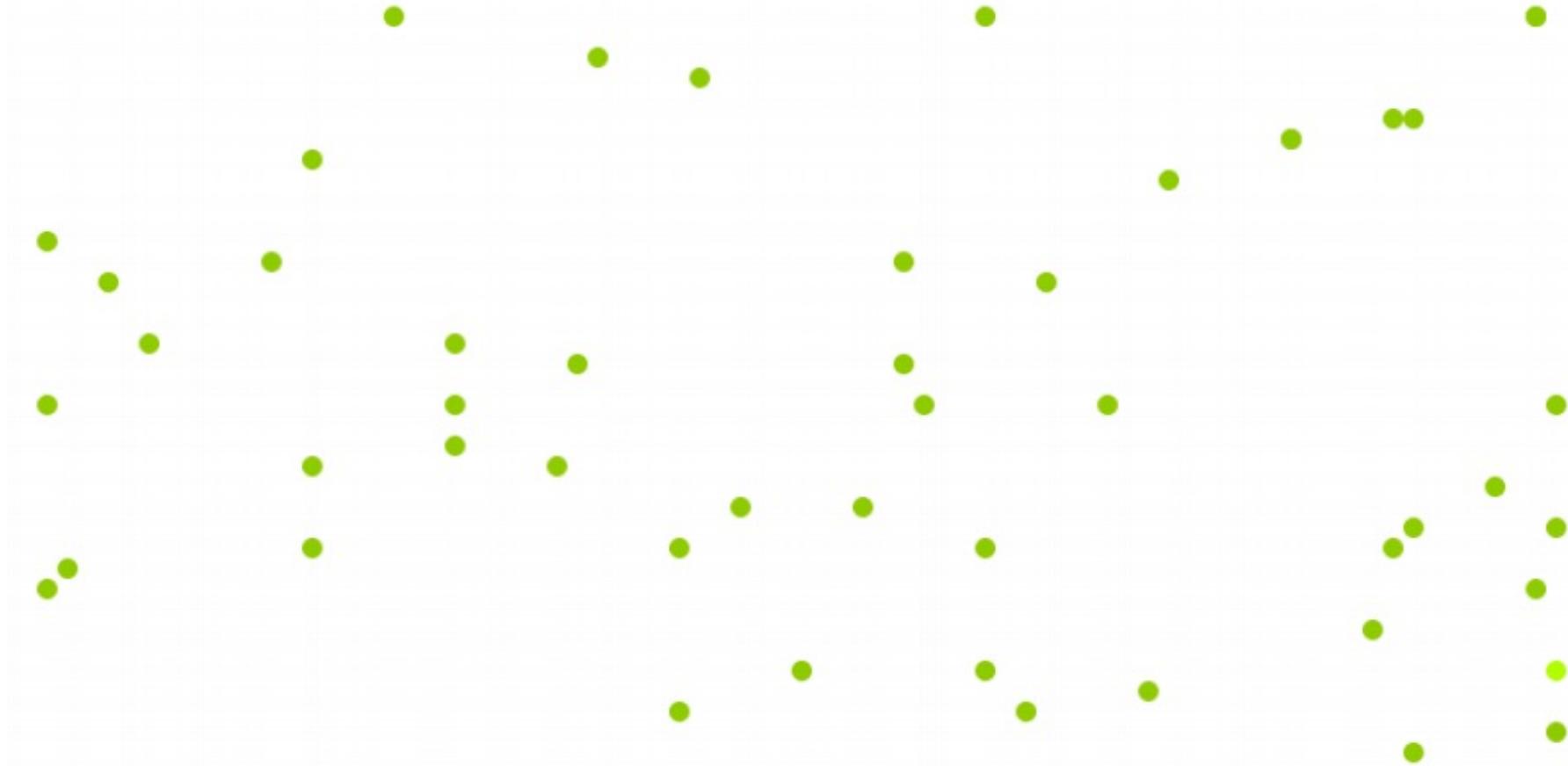
Length





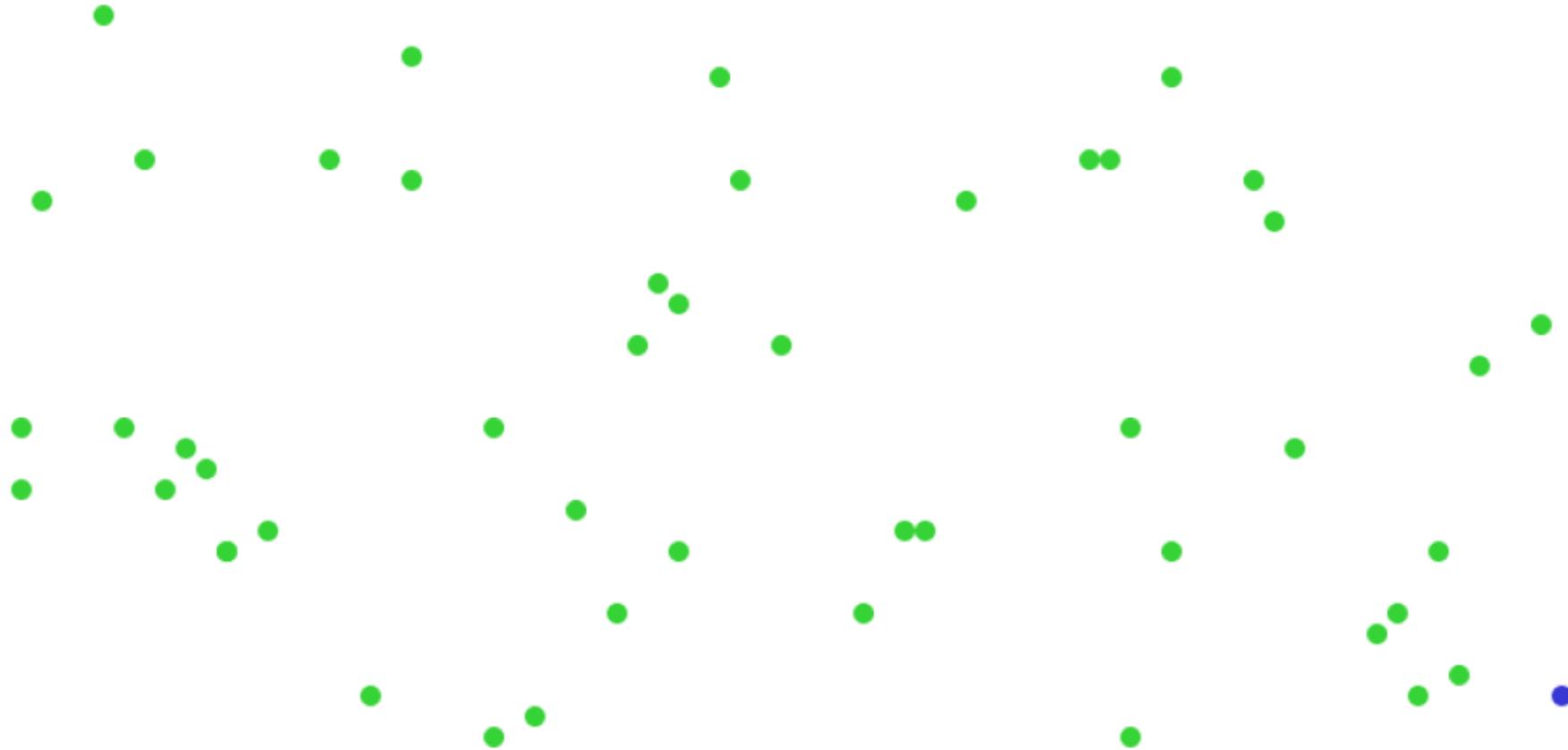
Size

Orientation



Lightness

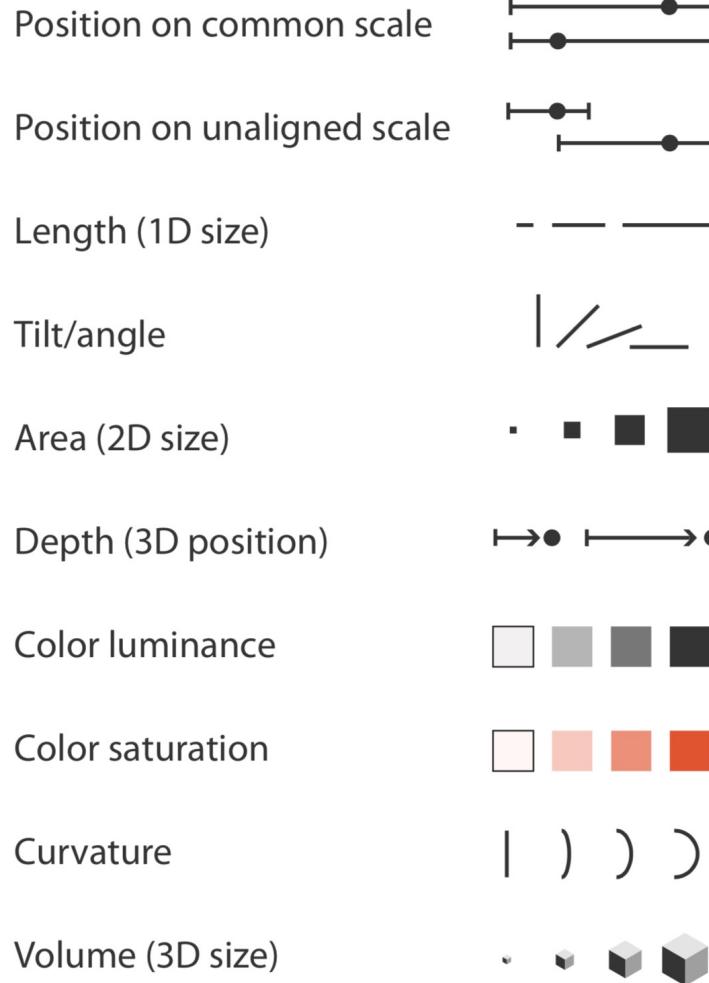
Hue



Visual channels don't work equally well!

Channels: Expressiveness Types and Effectiveness Ranks

④ **Magnitude Channels: Ordered Attributes**



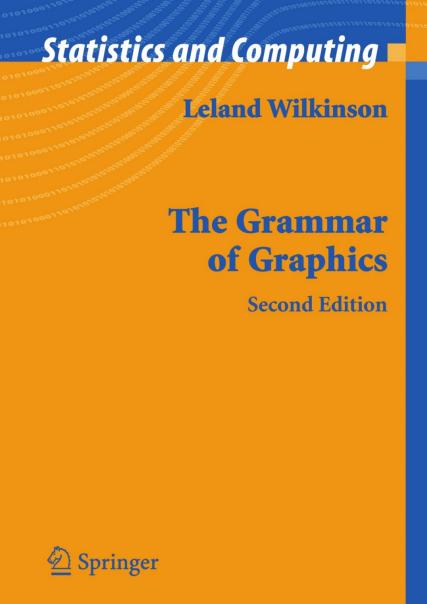
④ **Identity Channels: Categorical Attributes**



▲ Most
Effectiveness
▼ Least

MOST EFFECTIVE

LEAST EFFECTIVE



The Grammar of Graphics

Second Edition

Statistics and Computing

Leland Wilkinson

Springer

Statistical graphic specifications are expressed in six statements:

- 1) DATA: a set of data operations that create variables from datasets,
- 2) TRANS: variable transformations (*e.g., rank*),
- 3) SCALE: scale transformations (*e.g., log*),
- 4) COORD: a coordinate system (*e.g., polar*),
- 5) ELEMENT: graphs (*e.g., points*) and their aesthetic attributes (*e.g., color*),
- 6) GUIDE: one or more guides (*axes, legends, etc.*).

A Layered Grammar of Graphics

Hadley WICKHAM

GPL

ggplot2

DATA → Defaults

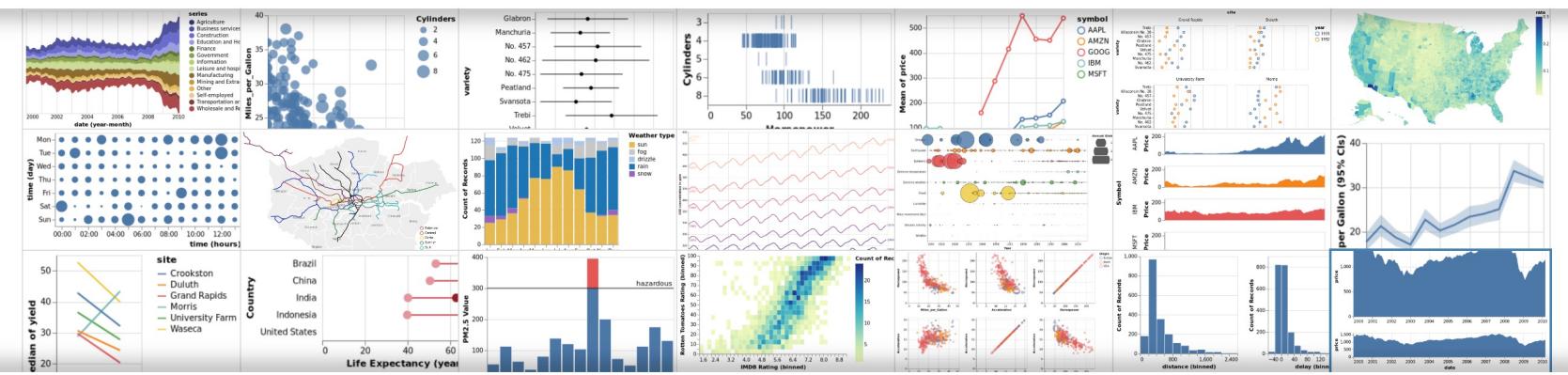
TRANS Data
Mapping

ELEMENT → Layer
Data
Mapping
Geom
Stat
Position

SCALE → Scale
GUIDE

COORD → Coord
Facet

Vega-Lite – A Grammar of Interactive Graphics



Vega-Lite is a high-level grammar of interactive graphics. It provides a concise, declarative JSON syntax to create an expressive range of visualizations for data analysis and presentation.

Vega-Lite specifications describe visualizations as encoding mappings from data to properties of graphical marks (e.g., points or bars). The Vega-Lite compiler automatically produces visualization components including axes, legends, and scales. It determines default properties of these components based on a set of carefully designed rules. This approach allows Vega-Lite specifications to be concise for quick visualization authoring, while giving user control to override defaults and customize various aspects. As we also designed Vega-Lite to support data analysis, Vega-Lite includes a set of transformations (e.g., aggregation, binning, filtering, sorting) and operations (e.g., stacking and faceting). Moreover, Vega-Lite specifications support layered and multi-view displays, and make it easy to create interactive with selection and brushing.

Compared to [Vega](#), Vega-Lite provides a more concise and convenient way to express visualizations. To convert Vega specifications to Vega-Lite specifications, users may use Vega-Lite's [Vega-to-Vega-Lite compiler](#). Vega-Lite also provides a subset of the features available in full-level Vega for advanced use cases.

For more information, read our [introduction article to Vega-Lite v1](#) or [Vega-Lite v2](#), see the [documentation](#) and take a look at our [example gallery](#).

Get started

Latest Version: 5.2.0

Vega-Lite: A Grammar of Interactive Graphics

Arvind Satyanarayan, Dominik Moritz, Kanit Wongsuphasawat, and Jeffrey Heer

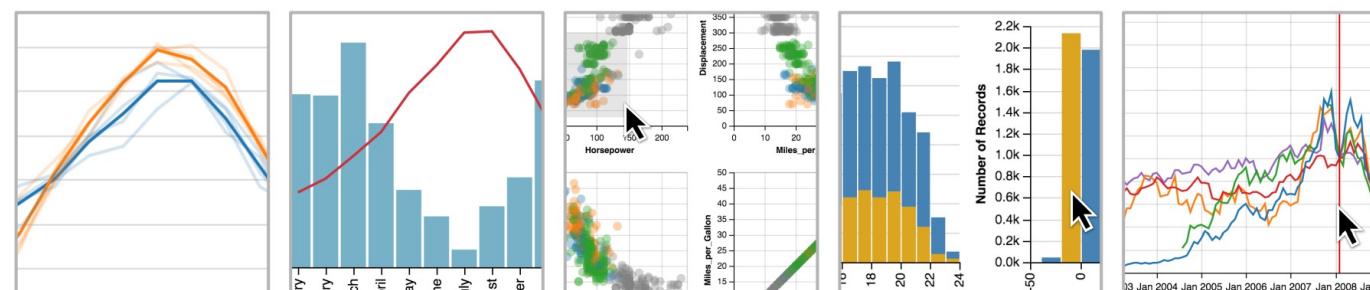


Fig. 1. Example visualizations authored with Vega-Lite. From left-to-right: layered line chart combining raw and average values, dual-axis layered bar and line chart, brushing and linking in a scatterplot matrix, layered cross-filtering, and an interactive index chart.



4.2.0

Search docs

GETTING STARTED

- [Overview](#)
- [Installation](#)
- [Dependencies](#)
- [Development Install](#)
- [Basic Statistical Visualization](#)

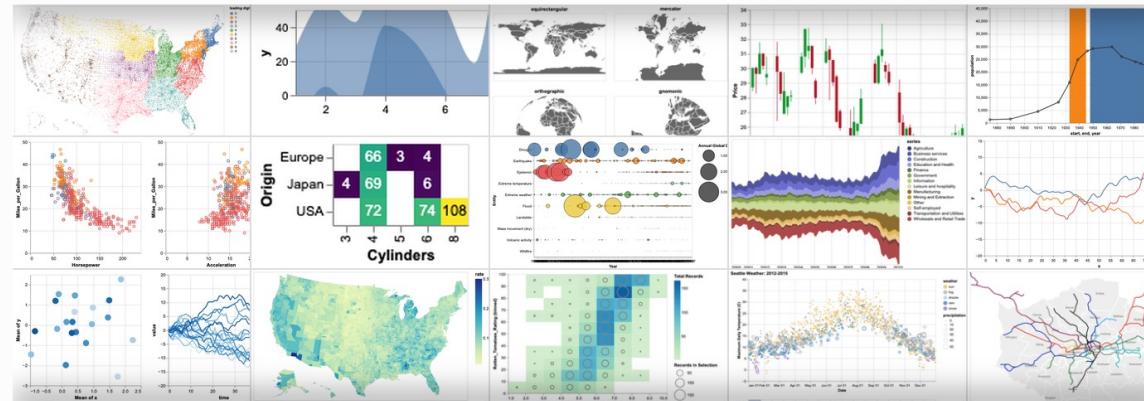
GALLERY

- [Example Gallery](#)

USER GUIDE

- [Specifying Data in Altair](#)
- [Encodings](#)
- [Marks](#)
- [Data Transformations](#)
- [Bindings, Selections, Conditions: Making Charts Interactive](#)
- [Top-Level Chart Configuration](#)
- [Compound Charts: Layer, HConcat, VConcat, Repeat, Facet](#)
- [Scale and Guide Resolution](#)
- [Saving Altair Charts](#)

Altair: Declarative Visualization in Python



Altair is a declarative statistical visualization library for Python, based on [Vega](#) and [Vega-Lite](#), and the source is available on [GitHub](#).

With Altair, you can spend more time understanding your data and its meaning. Altair's API is simple, friendly and consistent and built on top of the powerful [Vega-Lite](#) visualization grammar. This elegant simplicity produces beautiful and effective visualizations with a minimal amount of code.

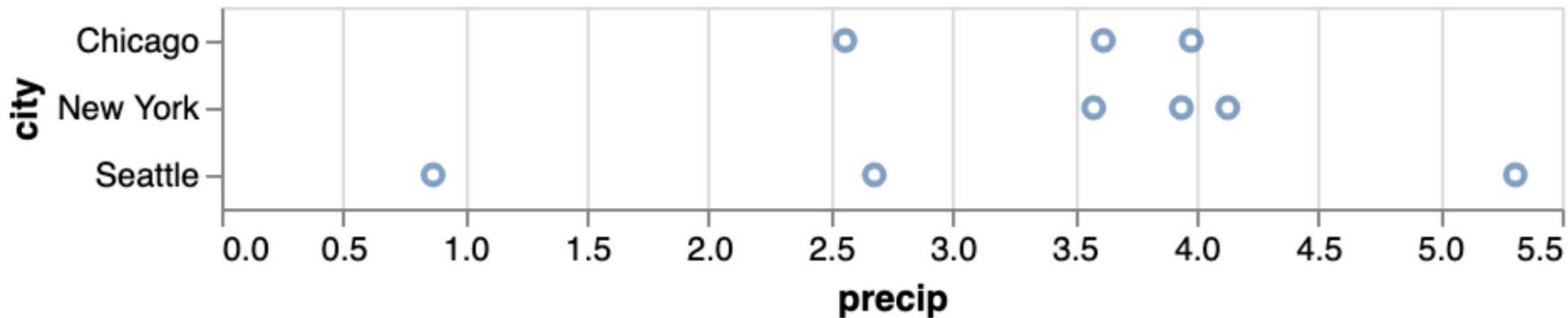
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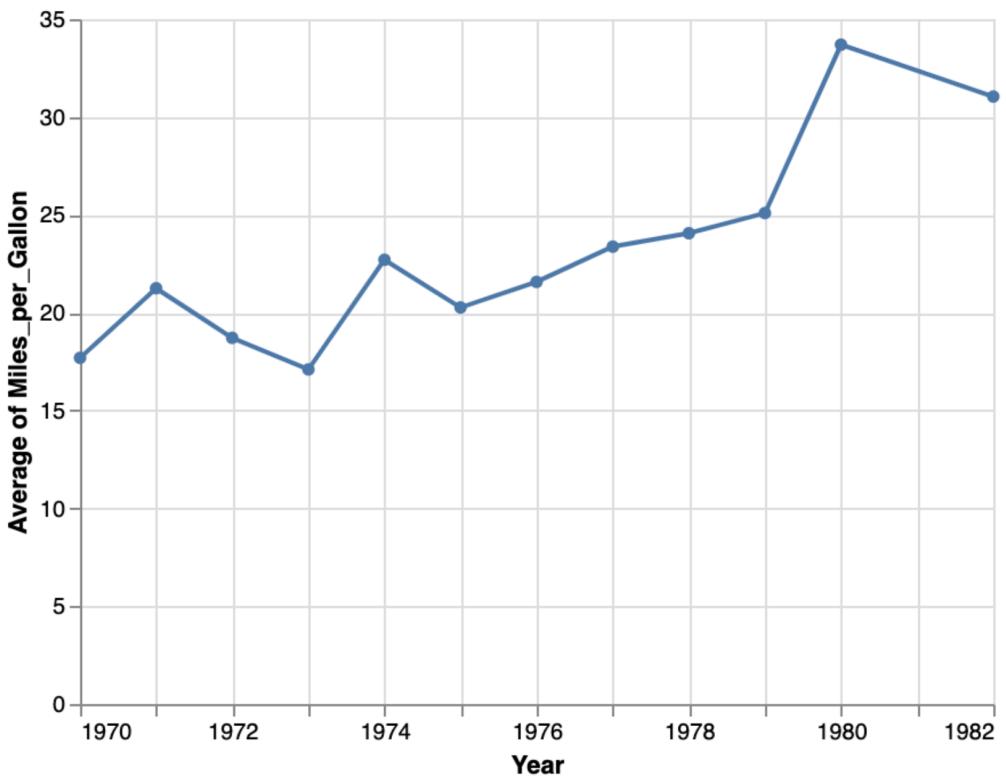
Gallery

- [Example Gallery](#)

```
alt.Chart(df).mark_point().encode(  
    alt.X('precip'),  
    alt.Y('city'))  
)
```



```
line = alt.Chart(cars).mark_line().encode(  
    alt.X('Year'),  
    alt.Y('average(Miles_per_Gallon)'))  
  
point = alt.Chart(cars).mark_circle().encode(  
    alt.X('Year'),  
    alt.Y('average(Miles_per_Gallon)'))  
  
line + point
```



Let's get our hands on some code!

Take-aways?

Altair as a declarative, powerful and versatile visualisation toolkit

- Rich specifications, faceting, interaction
- Working seamlessly with Pandas

Active and supportive community (see -- <https://github.com/altair-viz>)

A few key things to think about when designing a visualisation

- Think first what the objective is? Who is your audience? What message to deliver?
- What the data affords, what works best for the data?
- Which encodings, marks, layout work?
- How to activate the visualisation -- title, colour, style?



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