# Not Just Pretty Pictures

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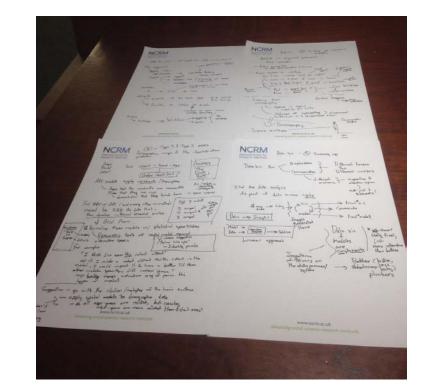
University of Glasgow

### Contents

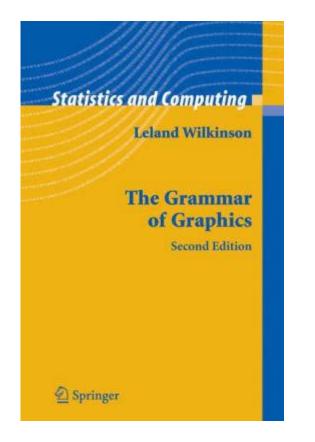
- What are data visualisations?
- How do data visualisations build on human strengths and intuitions?
- Data Visualisation and Statistical Modelling
- Summing up

What are data visualisations?

- Various definitions
- I'm partisan
- Grammar of Graphics Definition
- (Excluding my notes, right)



# Grammar of Graphics





# Grammar of Graphics

#### Layers

- 1. Support Layer
- 2. Data Layer
- 3. Annotation Layer
- Data Layer, defined by consistently applied mapping rules, is a necessary condition for an image being a data visualisation rather than something else
- Not (usually) a sufficient condition for a data visualisation to be a good data visualisation

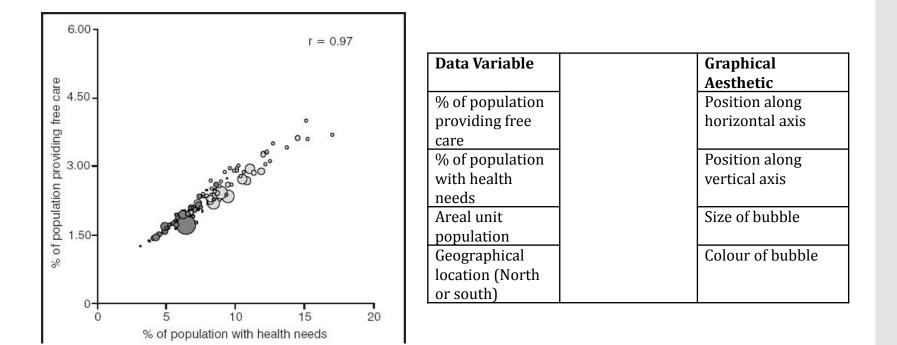
#### Mapping Rules

- Consistently applied
- Defined a priori
- From variables in data
- To graphical elements on an image

## Simple Mapping Rules Example

#### Graphic

#### **Mapping Rules**

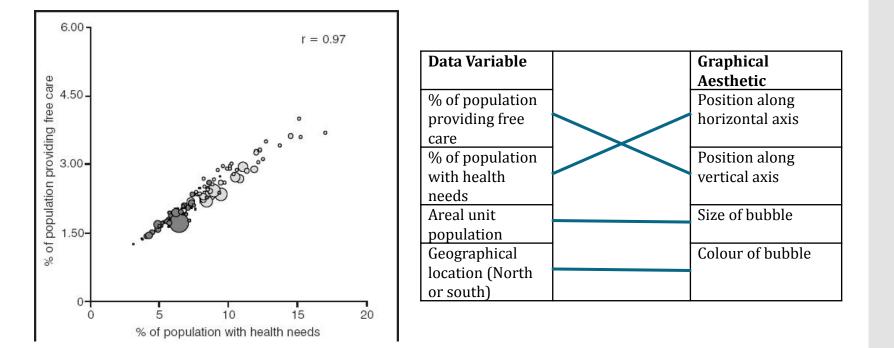


**Source**: Shaw M & Dorling D (2004) "Who cares in England and Wales? The Positive Care Law: cross-sectional study" *British Journal of General Practice*, 54 (509): 899-903

## Simple Mapping Rules Example

#### Graphic

#### **Mapping Rules**



**Source**: Shaw M & Dorling D (2004) "Who cares in England and Wales? The Positive Care Law: cross-sectional study" *British Journal of General Practice*, 54 (509): 899-903

# Simple Mapping Rules Example

#### Graphic Layers Annotation 6.00 r = 0.97 % of population providing free care 4.50-Data 3.00-1.50-Support 15 10 20 5 % of population with health needs

**Source**: Shaw M & Dorling D (2004) "Who cares in England and Wales? The Positive Care Law: cross-sectional study" *British Journal of General Practice*, 54 (509): 899-903

### Benefits of Data Visualisations

#### **Sequential Processing**

"Can you do Addition?" the White Queen asked. "What's one and one?"

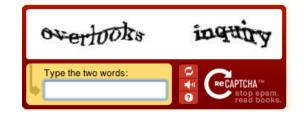
"I don't know," said Alice. "I lost count."

"She can't do Addition," the Red Queen interrupted.

#### Sources:

http://www.shmoop.com/alice-in-wonderland-looking-glass/education-quotes-4.html http://www.captcha.net/ https://upload.wikimedia.org/wikipedia/en/b/bg/MagrittePipe.jpg

#### **Gestalt Processing**





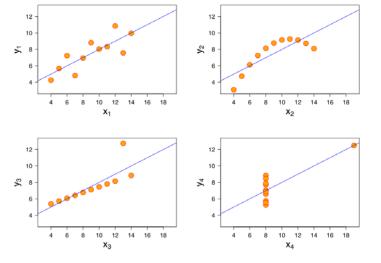
### Benefits of Data Visualisations

#### Scale and Surprise

#### Anscombe Quartet



Data visualisations
 Surprise



**Sources**: <u>http://r4ds.had.co.nz/intro.html</u> <u>https://en.wikipedia.org/wiki/Anscombe%27s\_quartet</u> Intuitions that data visualisations build on

# Mental models and metaphors

#### Use of colour

- Time 'runs from left to right' (position)
- More is 'higher' and less is 'lower'. (position)
- More is 'bigger' and less is 'smaller' (size)
- Red is 'bad' & blue is 'good'?



Accessibility

Cultural

variation

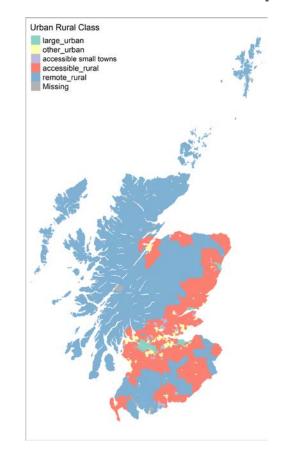


Data Variable		Graphical Aesthetic
Variable 1		
Variable 2		Colour
Variable 3		Something else
Variable 4		

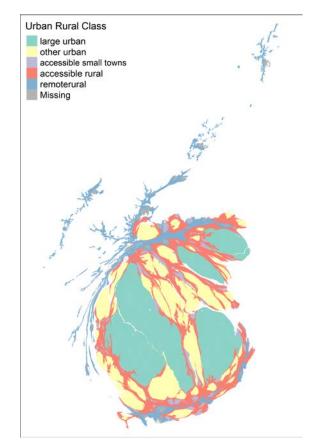
#### Source:

http://www.ebay.co.uk/itm/like/261727816290?lpid=122&chn=ps&googleloc=9046939&poi=&campaignid=620865095&device=c&adgroupid=273787608 66&rlsatarget=pla-181484341026&adtype=pla&crdt=0

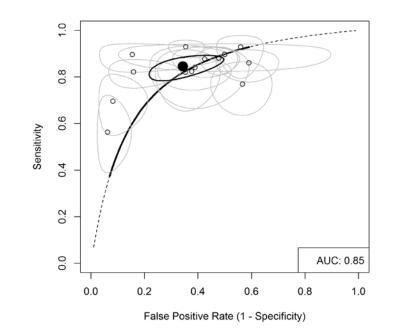
#### **Standard Urban Rural Map**



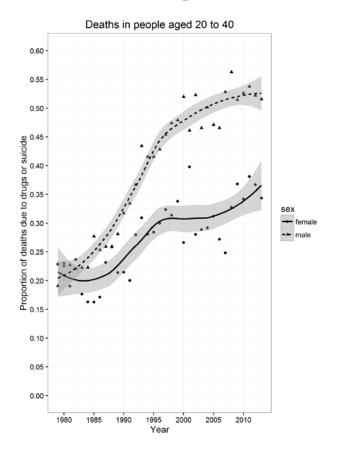
#### Cartogram by population



#### **Diagnostic Meta-Analysis**

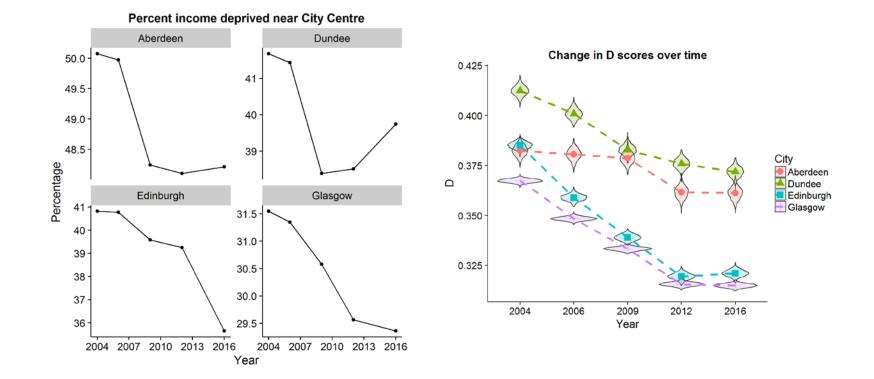


#### Suicide and Drug Deaths



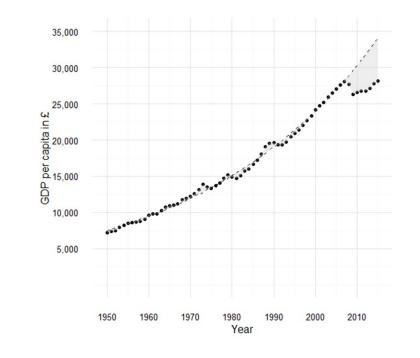
#### Decentralisation of poverty

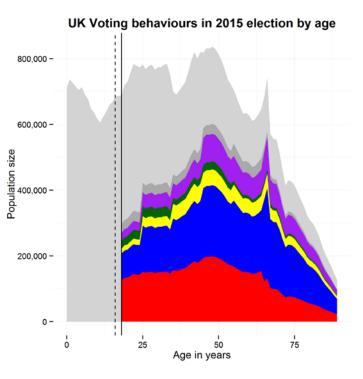
#### Index of Dissimilarity



#### Per Capita GDP trends in UK

#### (Non)voting by age in 2015





# My special interest

#### Lexis surface

- Demography as cartography
- Latitude, Longitude, Height
- Age, Year, something else
- Age: relative time
- Year: absolute time

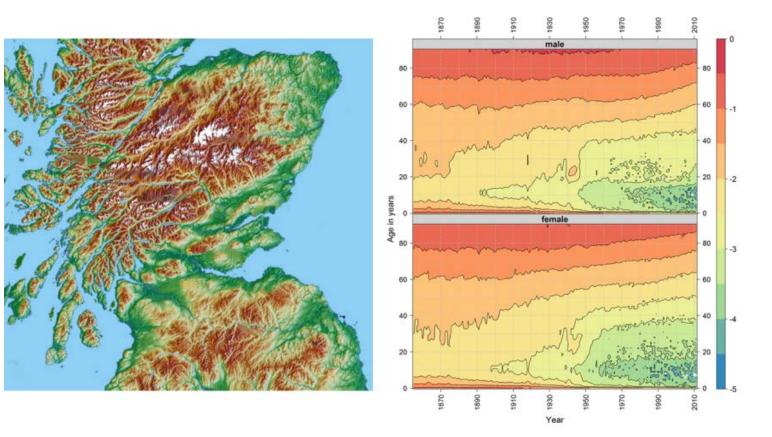
# Challenge of representing 3<sup>rd</sup> variable

- Two dimensional 'canvas'
- Three dimensional relationship
- Double-mapping of 3<sup>rd</sup> attribute
  - Colour/shade
  - Contour lines
- Contour lines link places on surface w/ same 3<sup>rd</sup> value

### Maps of space and time

#### Topographic map of Scotland

#### Log mortality surface map

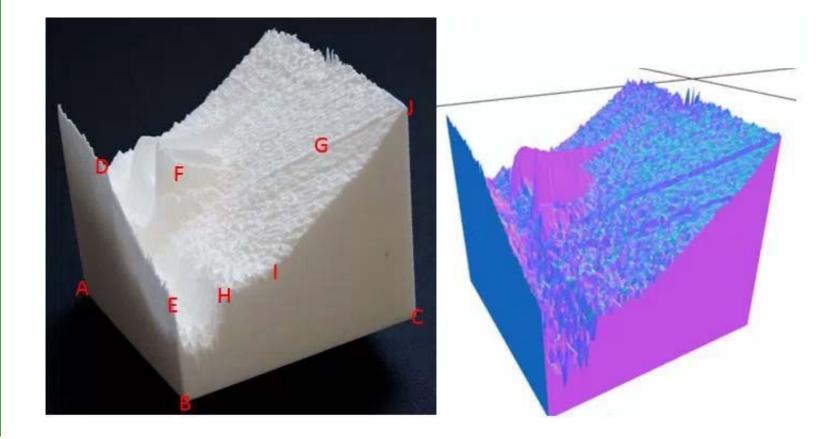


Source: https://ije-blog.com/2016/06/27/lexis-cubes-1-from-maps-of-space-to-maps-of-time/

# Escaping the Canvas

#### 3D Printed 'Data Cube'

#### **Github rendering**

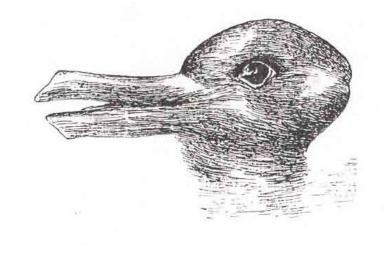


Source: <u>https://ije-blog.com/2016/06/27/lexis-cubes-2-case-study-log-mortality-for-males-in-finland-1878-to-2012/</u> https://github.com/JonMinton/Statistical\_Sculpture/blob/master/stl/individual/Imorts/fin\_male\_(1878-2012).stl Modelling of Age-Period-Cohort Effects

#### **The Identification Problem**

- Age matters
- Period matters
- Cohort matters
- But Cohort = Period Age
  An accounting identity
- Under-Identified

#### **Visual Analogy**



Two ways of resolving the data

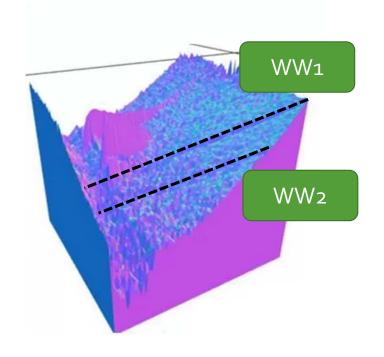
**Source**: https://commons.wikimedia.org/wiki/File:Duck-Rabbit\_illusion.jpg

# **APC** Debates

Argument Wars (Good for citations)

- Yang et al:
  - Intrinsic Estimators
  - Hierarchical APCs
  - Assumptions reasonable
- Bell et al
  - Assumptions still don't defeat logic
  - Simulation modelling shows assumptions not always reasonable

# But (some) cohort effects are real



Data Vis Driven Approaches

#### Modelling as formalisation

- See the data
- Build intuitions
  - Important Age-relationships
  - Important Period relationships
  - Important Cohorts
- Also
  - Trends in age-relationships
  - Sex differences
  - Age-localised period effects
- Formalise intuitions
- Test intuitions

# Lexis surfaces as a modelling metaphor

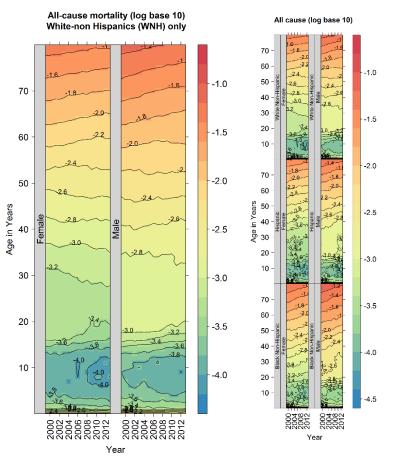
- Tobler's First Law:
  - Everything related
  - Near things more related
- Applies to space
- Can it apply to age-time?
- BYM models
  - Spatial Stats and Image Processing
- CARBayes package

# Dangers of not looking enough first

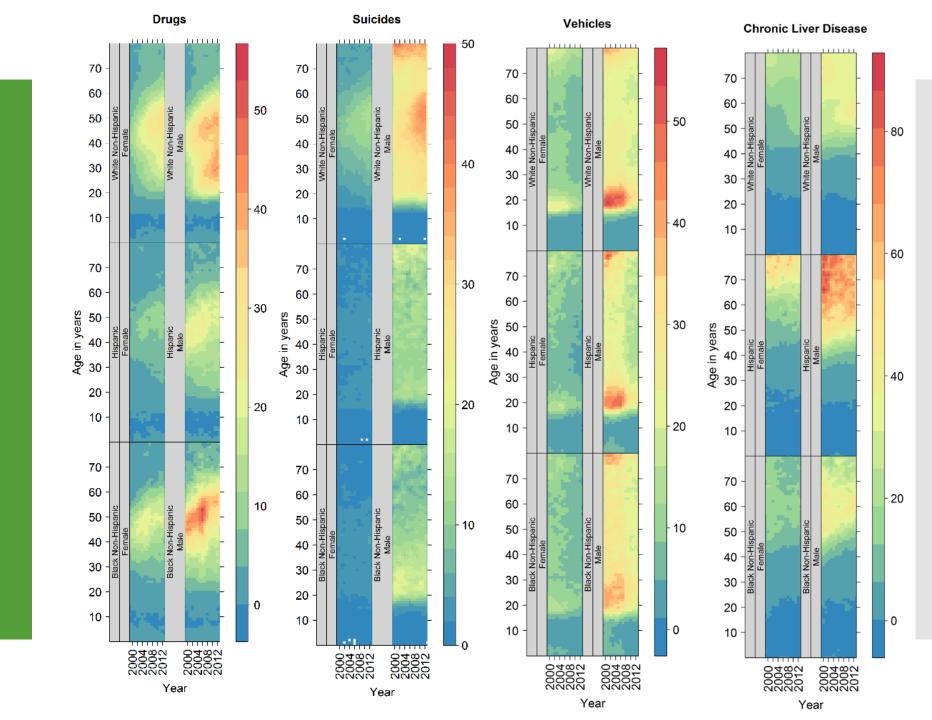
#### **Aggregation Bias**

- Case & Deaton
- Collapsing two genders and 10 years of age into one number
- Changing age-composition
   over time
- IJE Commentary

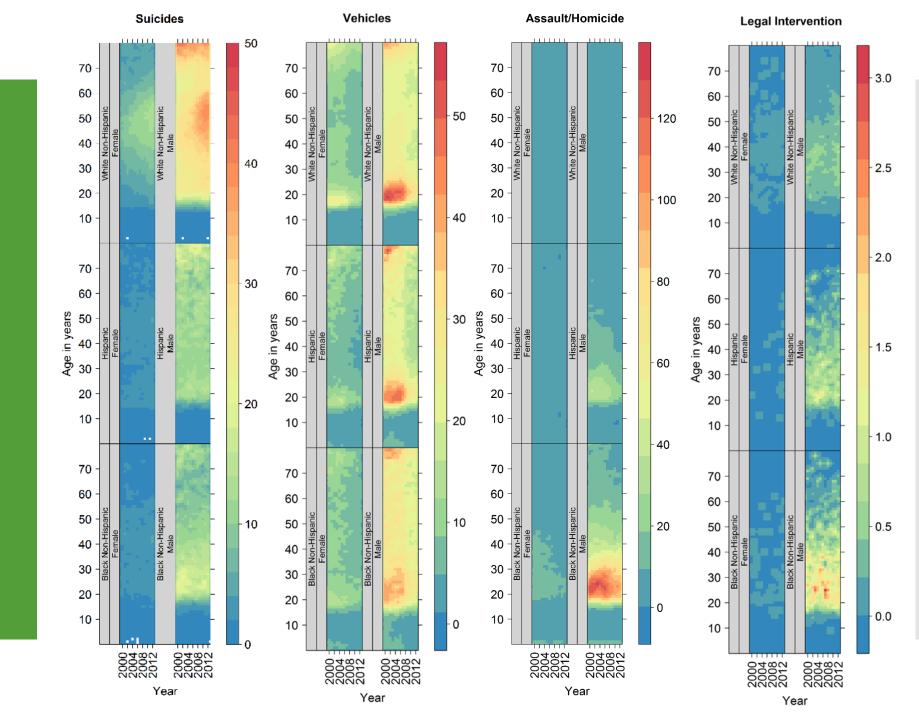
#### **Disaggregated Data**



# Missed Opportunities



# Missed Opportunities

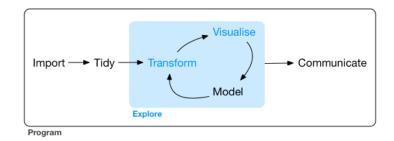


# **Final Thoughts**

#### **Types of Data Visualisation**

- Communicatory
  - General Audience
  - Specialist Audience
- Exploratory
- "Different horses for different courses"
- The 'three second rule'

#### Data Vis in Data Science



- Data to knowledge/insight
- Data Vis vital `in the loop' as surprise is vital for learning from data
- Models and vis are complements:
  - Develop intuitions: Data Vis
  - Test intuitions: Modelling
  - Type 1/Type 2 error

Source: http://r4ds.had.co.nz/explore-intro.html