# **Data Visualisation**

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Chapter 3 – Visual Encoding

## **Contents**

- Visual Encoding
- Deconstructions



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# **Chapter 3.1 – Visual Encoding**

#### **Contents**

- Visual Variables
- Visual Marks
- Visual Encoding Examples



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## **The Data Visualisation Process**

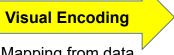
#### **An Overview**

• Data visualisation is the transformation of data into an image using an appropriate visual encoding strategies.

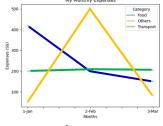
Month	Category	Amount
1-Jan	Food	420.30
1-Jan	Transport	200.20
1-Jan	Others	50.00
2-Feb	Food	200.45
2-Feb	Transport	210.00
2-Feb	Others	500.23
3-Mar	Food	150.60
3-Mar	Transport	205.70
3-Mar	Others	80.30

Data

Data & Conceptual Models



Mapping from data to image



Image

Visual Marks and Channels (Variables)

Lines

Colour

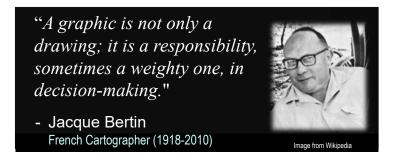
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#### **Visual Variables**

#### **Bertin's Visual Variables**

• Jacque Bertin, a French cartographer proposed set of **visual variables** in 1967 that can be used to construct map symbols and other graphical techniques.





[1] Bertin's quote from: J. Bertin, graphics and graphic information processing (1981), https://books.google.com.sg/books?id=csqX\_xnm4tcC&printsec=copyright&redir\_esc=y#v=onepage&q&f=false

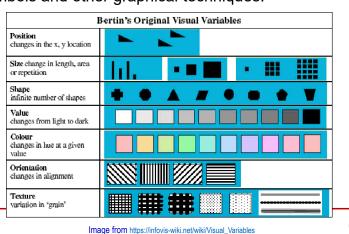
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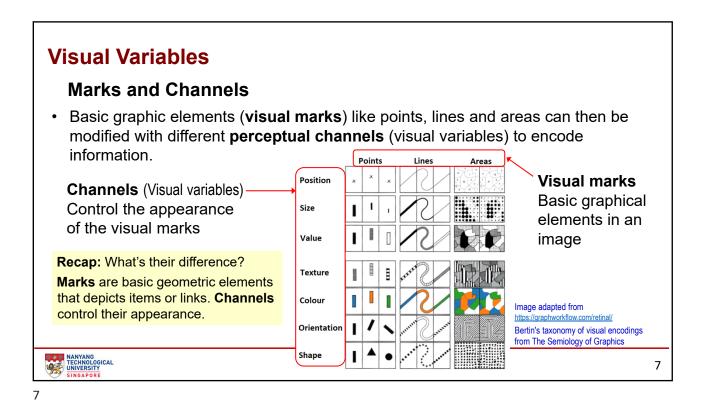
#### **Visual Variables**

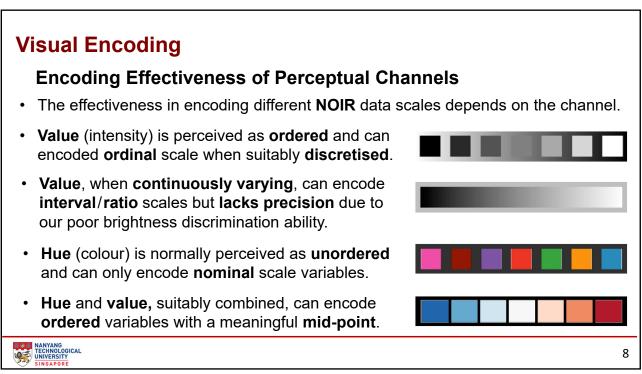
#### **Bertin's Visual Variables**

- Jacque Bertin, a French cartographer proposed set of visual variables in 1967 that can be used to construct map symbols and other graphical techniques.
- Bertin identified seven main categories of visual variables:
  - position
  - size
  - shape
  - value (intensity)
  - colour
  - orientation
  - texture



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## **Visual Variables**

## **Bertin's Levels of Organisation**

• Bertin's levels of organisation illustrates the degree of effectiveness in encoding the various **NOIR** data scales.

<b>N</b> ominal
<b>O</b> rdinal
Quantitative

Note 1: **Interval** and **Ratio** is classified as a **Quantitative** scale.

Note 2: **Size** indicates **degree** of encoding effectiveness

Position	Ν	0	Q
Size	N	0	Q
Value	N	0	Q
Texture	N	0	
Colour	N		
Orientation	N		
Shape	7		



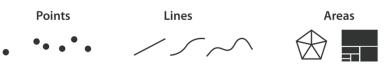
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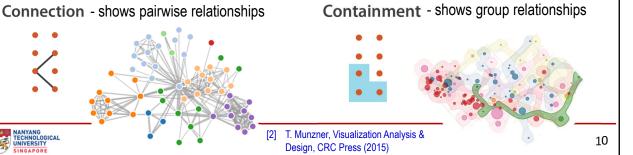
#### **Visual Marks**

### **Types of Marks**

• Item marks – are basic geometric objects that depict items on a graph<sup>[2]</sup>.



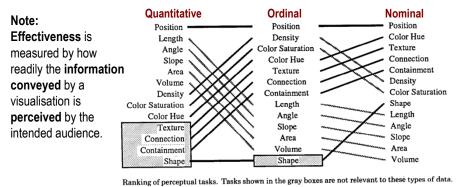
• Link marks – show relationship between items<sup>[2]</sup>.



# **Visual Encoding**

## Mackinlay's Visual Variables

• **Jock Mackinlay**<sup>[3]</sup> expanded Bertin's variables and proposed an **effectiveness ranking** based on the NOIR data scales.



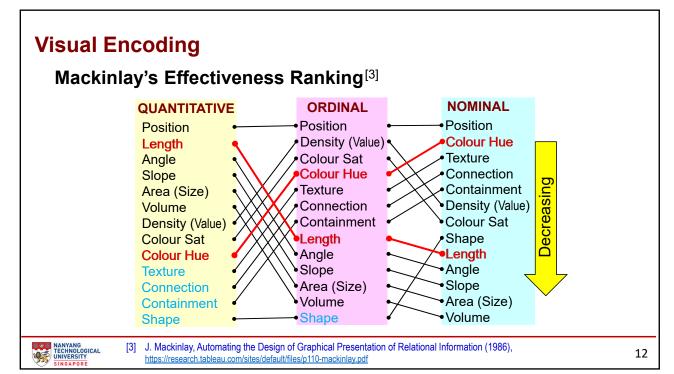


Jock D. Mackinlay Information Visualisation Expert and Former Vice President of R&D at Tableau Software (Image from Twitter)

NANYANG TECHNOLOGICAL UNIVERSITY SINGAPORE  J. Mackinlay, Automating the Design of Graphical Presentation of Relational Information (1986), https://research.tableau.com/sites/default/files/p110-mackinlay.pdf

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# **Visual Encoding Examples Univariate Data** In visual encoding of univariate datasets, common Variables ABCDE visual marks used are lines (e.g. line and bar charts) Values and points (e.g. dot plot). Values В Α С D Bar Chart (Lines) Dot Plot (Points) 13

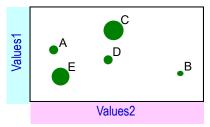
**Visual Encoding Examples Bivariate Data** · In visual encoding of bivariate datasets, the most BCDE Variables common visual mark used are points in 2D scatter 1 Values1 plot. Line mark may be used, such as with a dual 2 Y-axis bar charts. Values2 В Values2 Ε Α В 2D Scatter Plot (Points) Dual-axis Bar Chart (Lines) 14

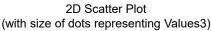
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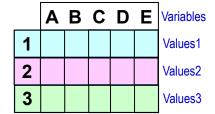
# **Visual Encoding Examples**

#### **Trivariate Data**

Trivariate datasets are commonly encoded using points. The two axes of a 2D scatter plot can encode two quantitative (Q) data type and area of the dots can encode the third ordinal or Q data type.









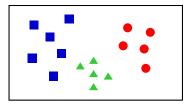
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# Visual Encoding Examples

# **Redundant Encoding**

- Redundant encoding is the use of **more than one graphical** or visual variable (e.g. color + shape) to encode/represent **one variable of data**.
- Redundancy allows **more perceptual channels** to get the same information into our brain. This helps information to be perceived **faster**, **easier** and more **accurately**<sup>[4]</sup>. It should be considered if there are leftover unused visual variables.



2D Scatter Plot Example (Redundant **Colour + Shape** )



Bar Chart Example (Redundant **Length + Value**)



[4] C. Ware, Information Visualization: Perception for Design, 3<sup>rd</sup> Edition, Morgan Kaufmann (2013),

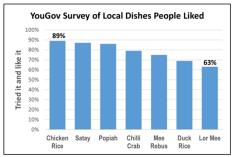
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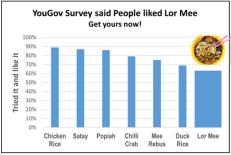
# **Ethical Visualisation**

Use fair and consistent visual encoding

 Your stall specialises in Lor Mee and you saw the results of the recent YouGov local food survey. You decided to share these results with your potential customers.



The YouGov survey results of selected local dishes



Your visualisation of the YouGov food survey

Has ethical visualisation been violated here?

**Note:** Height and width are integral perceptual dimensions. People tend to see the size of the bar as representing the data value [5].



Note: Food survey figures from: <a href="https://sg.yougov.com/en-sg/news/2019/08/13/classic-singaporean-cuisine-ranked-singaporeans/">https://sg.yougov.com/en-sg/news/2019/08/13/classic-singaporean-cuisine-ranked-singaporeans/</a>
[5] Stephen M. Kosslyn, Graph Design for the Eye and Mind, Oxford University Press (2006)

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# **Think and Apply**

## **Visual Encoding Exercise**

• Sketch a possible chart that uses an appropriate visual encoding design for each of the data tables. Consider the given data attributes carefully in your encoding design.

Living Members from Class of '49					
	Names	Gender	Age		
1					
2					
3					
4					
5					
6					

Visual mark – lines

	Orientation 1-shirts - Comparing Hails					
	No.	Gender	Weight	Height	Hall (A,B,C,D)	T-shirt Size
1						
2						
3						
:						
:						
700						

Visual mark - points

	Dates	Total Cases	Warded	ICU
1				
2				
3				
:				
:				
365				

COVID-19 Infection in 2020

Visual mark - areas



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## Summary

# **Visual Encoding**

- Data visualisation is the transformation of data into an image that is informative and conveys useful insights embedded within the data.
- Visual encoding is the process of mapping these data into that image.
- Based on the attributes of the various dimensions of the dataset, appropriate combinations of visual marks and variables (channels) are used to encode the data.
- Effective visual encoding requires a good understanding of the characteristics and limitations of **human visual perception**.



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# **Chapter 3.2 – Deconstruction**

#### **Contents**

- Playfair's Import/Export (Line Charts)
- Wattenberg's Map of the Market (Tree Map)
- Fry's Basketball Salary versus Performance (Slope Chart)



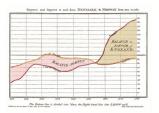
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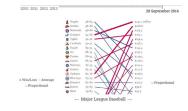
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#### **Deconstruction**

### Seeing How Others Do It

- The process of deconstruction allows us to **analyse** how others have visually encoded the information they wanted to convey in their dataset.
- Deconstruction is a great way to learn how to construct your own visualisation, especially if the examples analysed have been effective in telling their visual story.





Playfair's Import/Export (1786) A good old pre-digital chart

Wattenberg's Map of the Market (1998) A visualisation still current in use

Fry's Salary vs Performance (2014) Visualisation for US Basketball fans

checkout: https://fathom.info/salaryper/

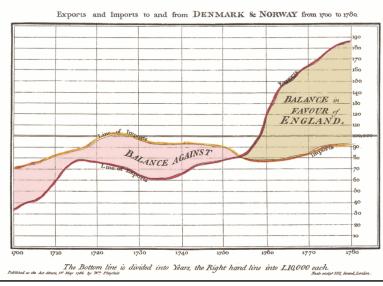
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NANYANG TECHNOLOGICAL UNIVERSITY SINGAPORE checkout: https://finviz.com/map.ashx

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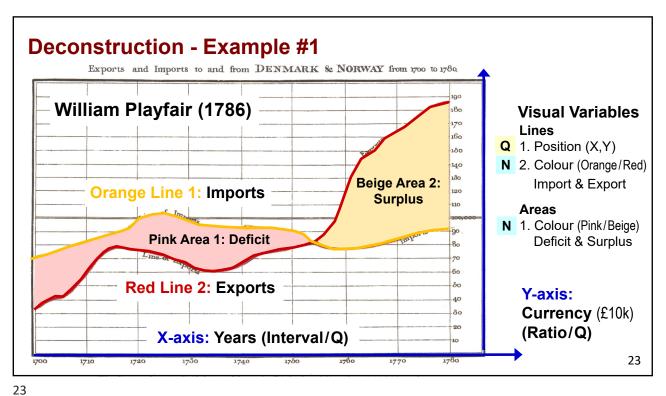
# **Deconstruction - Example #1**

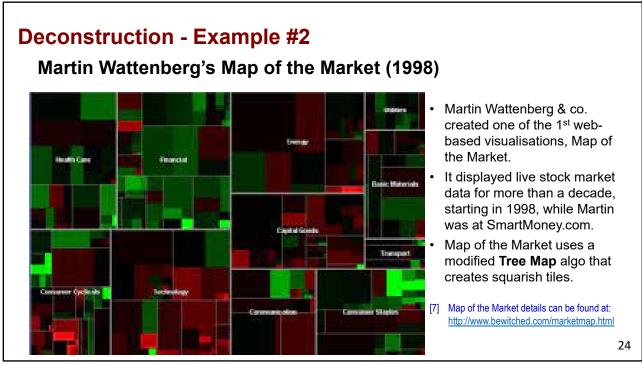
# William Playfair's Line Charts (1786)

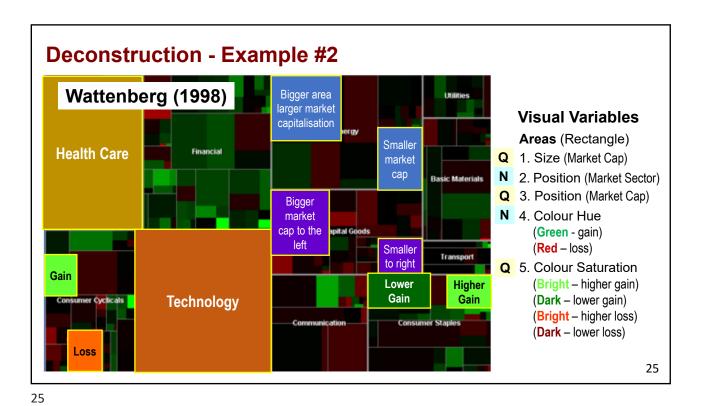


- William Playfair's Commercial and Political Atlas was published in 1786
- Playfair examined the imports and exports between Britain and various countries.
- To illustrate these trade relationships, Playfair created the first ever line charts that show the change in trade over time.
  - [6] Image can be found at: https://upload.wikimedia.org/wikipedia/com mons/5/52/Playfair TimeSeries-2.png

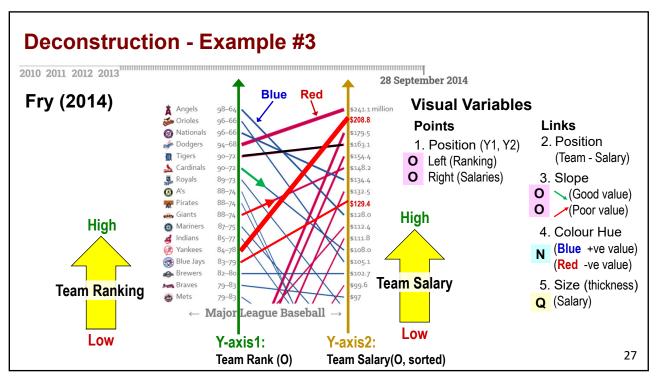
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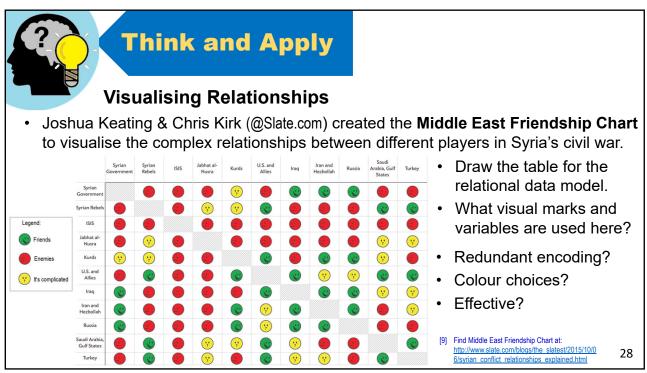




**Deconstruction - Example #3** Ben Fry's Slope Chart (2010-2014) 2010 2011 2012 2013 28 September 2014 Ben Fry is a visualisation expert and principal of Fathom, Angels
Orioles \$241.1 million 96-66 a design and software Nationals 96-66 consultancy in Boston, USA. 94-68 Dodgers \$163.1 **Tigers** \$154.4 He designed Salary vs. 🚣 Cardinals \$148.2 Royals 89-73 \$134.4 Performance to visualised if (A's 88-74 \$132.5 baseball teams are spending 88-74 Pirates \$129.4 88-74 \$128.0 their money well over the Giants Mariners 87-75 \$112.4 season. Indians
Yankees 85-77 \$111.8 The visualisation uses a slope Blue Jays 83-79 \$105.1 • Win/Loss o Average chart. Brewers □ Proportional □ Proportional Braves Fry's website for Salary vs Performance: Mets https://benfry.com/salaryper/ ← Major League Baseball → 26



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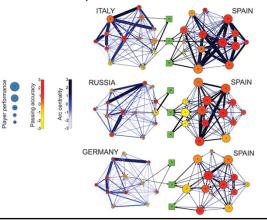




# **Think and Apply**

### **Visualising Performance in Team Games** (optional)

 Duch et al. analysed three Euro 2008 matches of the Spanish team using a visualisation technique from social network analysis.



- What types of data variables are visually represented in the image.
- What perceptual channels were used to encode their values?
- Colour choices? Can be better?

[10] Visualisation described in this paper: https://journals.plos.org/plosone/articl e?id=10.1371/journal.pone.0010937

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#### **Summary**

## **Deconstruction**

- Data visualisation is both an art and a science.
- Designing **informative** and **visually appealing** visualisations requires us to learn from many other visualisation experts.
- We can do this by deconstructing their good pieces of work.
- We observe how they make use of the various visual marks and variables to encode useful information embedded in the datasets.
- We need to do these observation with a critical eye, to see what has worked well and what has not and can be improved.



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# **References for Visual Encoding and Deconstruction**

- [1] Bertin's quote from: J. Bertin, graphics and graphic information processing (1981)  $\frac{1}{2} \frac{1}{2} \frac{1$
- [2] T. Munzner, Visualization Analysis & Design, CRC Press (2015)
- [3] J. Mackinlay, Automating the Design of Graphical Presentation of Relational Information (1986), https://research.tableau.com/sites/default/files/p110-mackinlay.pdf
- [4] C. Ware, Information Visualization: Perception for Design, 3rd Edition, Morgan Kaufmann (2013),
- [5] William Playfair's image can be found at: https://upload.wikimedia.org/wikipedia/commons/5/52/Playfair\_TimeSeries-2.png
- [6] William Playfair's image can be found at: https://upload.wikimedia.org/wikipedia/commons/5/52/Playfair\_TimeSeries-2.png
- [7] Martin Wattenberg's Map of the Market details can be found at: http://www.bewitched.com/marketmap.html
- [8] Ben Fry's website for Salary vs Performance: <a href="https://benfry.com/salaryper">https://benfry.com/salaryper</a>
- [9] Keating & Kirk's Middle East Friendship Chart at: http://www.slate.com/blogs/the\_slatest/2015/10/06/syrian\_conflict\_relationships\_explained.html
- [10] Duch et al.'s visualisation described in this paper: https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0010937



Note: All online articles were accessed between May to June 2021

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