

# AI UX & Data Visualisation Design Principles (CA6002)

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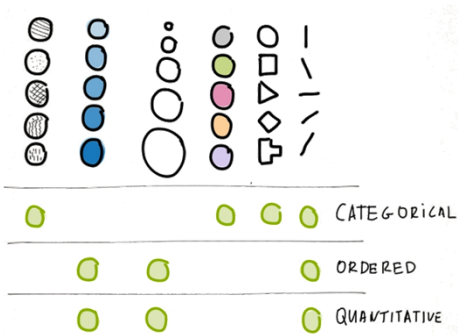
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## Chapter 1.3 – Visual Encoding

### Contents

- Visual Encoding
  - Visual Variables
  - Visual Marks
  - Visual Encoding Examples
- Deconstructions



# The Data Visualisation Process

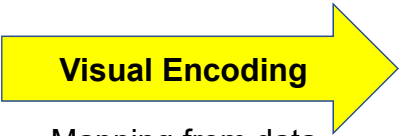
## An Overview

- 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



# 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.

*"A graphic is not only a drawing; it is a responsibility, sometimes a weighty one, in decision-making."*

- Jacque Bertin  
French Cartographer (1918-2010)



Image from Wikipedia



[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](https://books.google.com.sg/books?id=csqX_xnm4tcC&printsec=copyright&redir_esc=y#v=onepage&q&f=false)

## 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

Bertin's Original Visual Variables	
<b>Position</b> changes in the x, y location	
<b>Size</b> change in length, area or repetition	
<b>Shape</b> infinite number of shapes	
<b>Value</b> changes from light to dark	
<b>Colour</b> changes in hue at a given value	
<b>Orientation</b> changes in alignment	
<b>Texture</b> variation in 'grain'	



Image from [https://infovis-wiki.net/wiki/Visual\\_Variables](https://infovis-wiki.net/wiki/Visual_Variables)

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## Visual Variables

### Marks and Channels

- Basic graphic elements (**visual marks**) like points, lines and areas can then be modified with different **perceptual channels** (visual variables) to encode information.

**Channels** (Visual variables)  
Control the appearance of the visual marks

**Recap:** What's their difference?  
**Marks** are basic geometric elements that depicts items or links. **Channels** control their appearance.

	Points	Lines	Areas
Position			
Size			
Value			
Texture			
Colour			
Orientation			
Shape			

**Visual marks**  
Basic graphical elements in an image

Image adapted from <https://graphworkflow.com/retinal/>  
Bertin's taxonomy of visual encodings from The Semiology of Graphics



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# Visual Encoding

## Encoding Effectiveness of Perceptual Channels

- The effectiveness in encoding different **NOIR** data scales depends on the channel.
- **Value** (intensity) is perceived as **ordered** and can encoded **ordinal** scale when suitably **discretised**.
- **Value**, when **continuously varying**, can encode **interval/ratio** scales but **lacks precision** due to our poor brightness discrimination ability.
- **Hue** (colour) is normally perceived as **unordered** and can only encode **nominal** scale variables.
- **Hue** and **value**, suitably combined, can encode **ordered** variables with a meaningful **mid-point**.



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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.

Nominal  
Ordinal  
Quantitative

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

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

Position	N	O	Q
Size	N	O	Q
Value	N	O	Q
Texture	N	o	
Colour	N		
Orientation	N		
Shape	N		



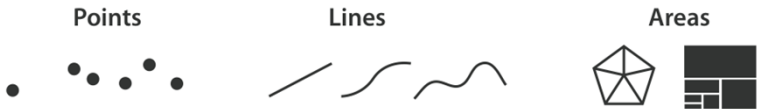
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Visual Marks

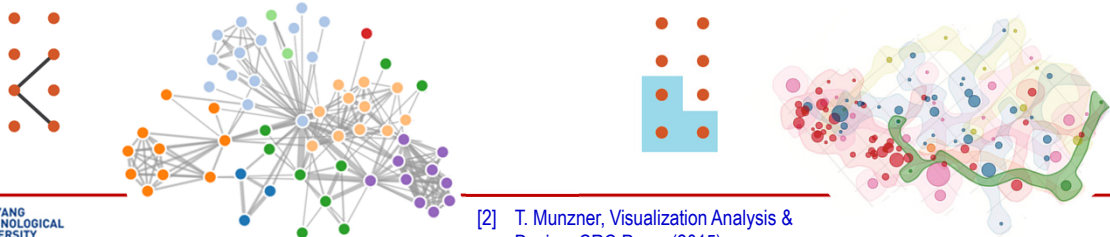
Types of Marks

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



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

Connection - shows pairwise relationships      Containment - shows group relationships



[2] T. Munzner, Visualization Analysis & Design, CRC Press (2015)

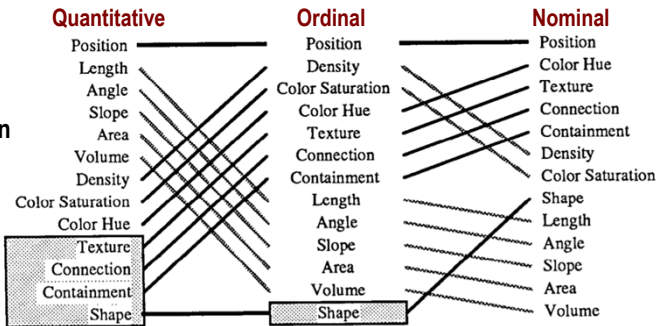
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Visual Encoding

Mackinlay's Visual Variables

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

**Note:**  
Effectiveness is measured by how readily the information conveyed by a visualisation is perceived by the intended audience.



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

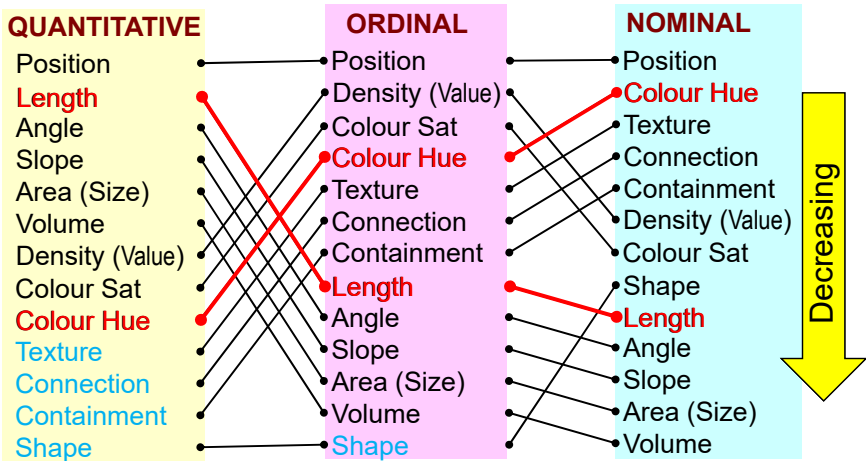


[3] J. Mackinlay, Automating the Design of Graphical Presentation of Relational Information (1986), <https://research.tableau.com/sites/default/files/p110-mackinlay.pdf>

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Visual Encoding

Mackinlay’s Effectiveness Ranking<sup>[3]</sup>

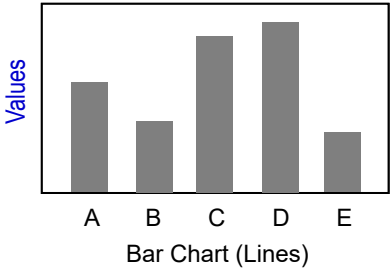
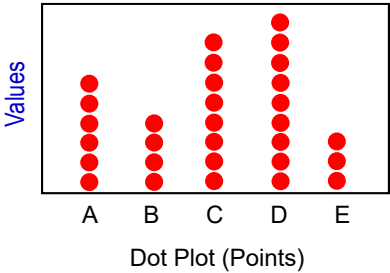


Visual Encoding Examples

Univariate Data

- In visual encoding of univariate datasets, common visual marks used are **lines** (e.g. line and bar charts) and **points** (e.g. dot plot).

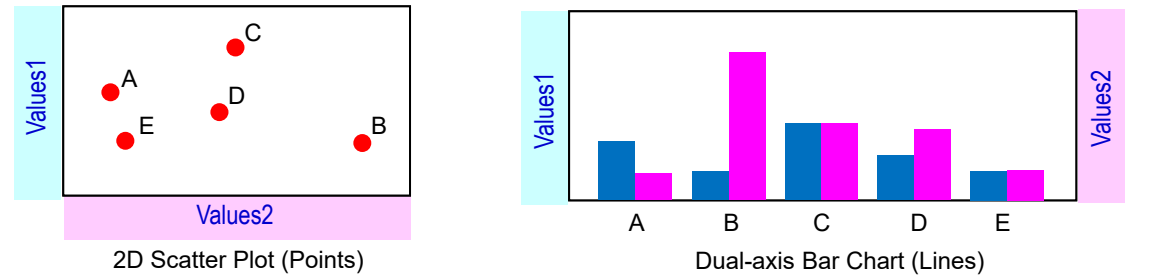
	A	B	C	D	E
Variables					
Values	1				



## Visual Encoding Examples

### Bivariate Data

- In visual encoding of bivariate datasets, the most common visual mark used are **points** in 2D **scatter plot**. **Line** mark may be used, such as with a **dual Y-axis bar charts**.



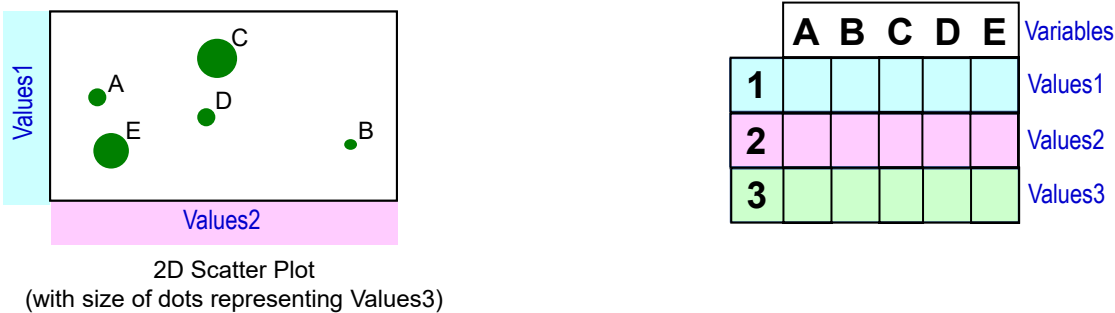
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## Visual Encoding Examples

### 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.



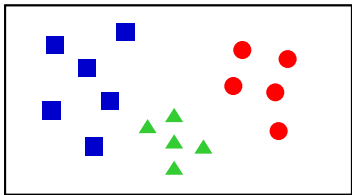
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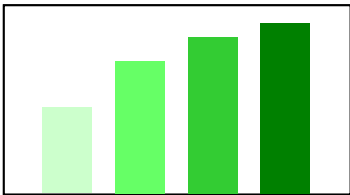
## Visual Encoding Examples

### 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),

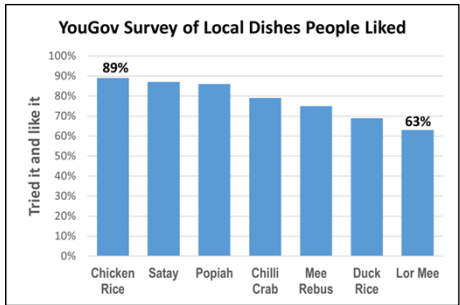
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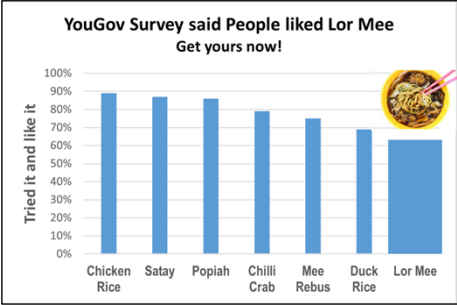
## Ethical Visualisation

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<sup>[5]</sup>.



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

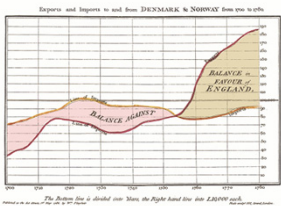
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# Deconstruction

## Seeing How Others Do It

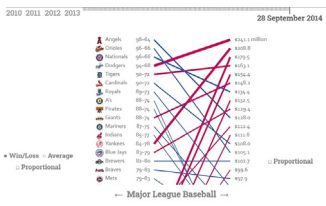
- The process of deconstruction allows us to **analyse** how others have visually encoded the information they wanted to convey in their dataset.
- Deconstruction is a great way to **learn how to construct** your own visualisation, especially if the examples analysed have been effective in telling their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

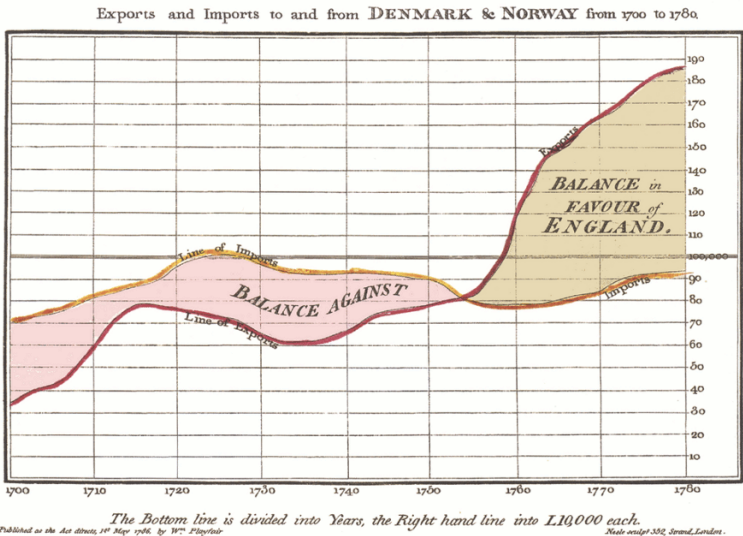


checkout: <https://finviz.com/map.ashx>

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

# Deconstruction - Example #1

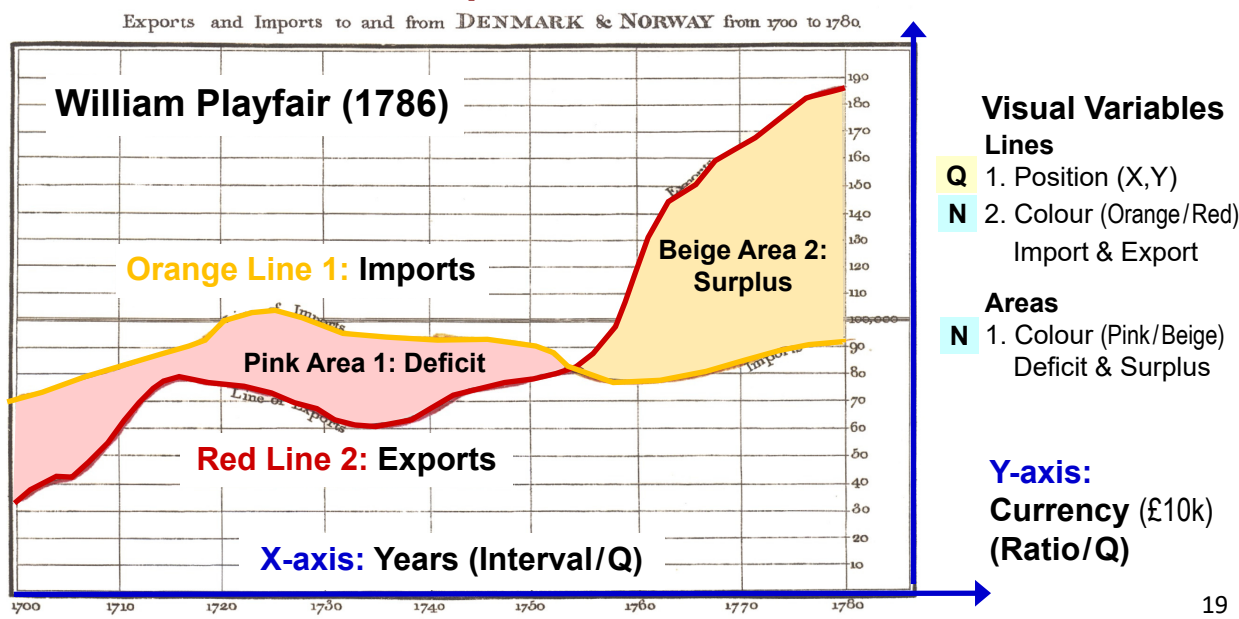
## William Playfair's Line Charts (1786)



- 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/commons/5/52/Playfair\\_TimeSeries-2.png](https://upload.wikimedia.org/wikipedia/commons/5/52/Playfair_TimeSeries-2.png)

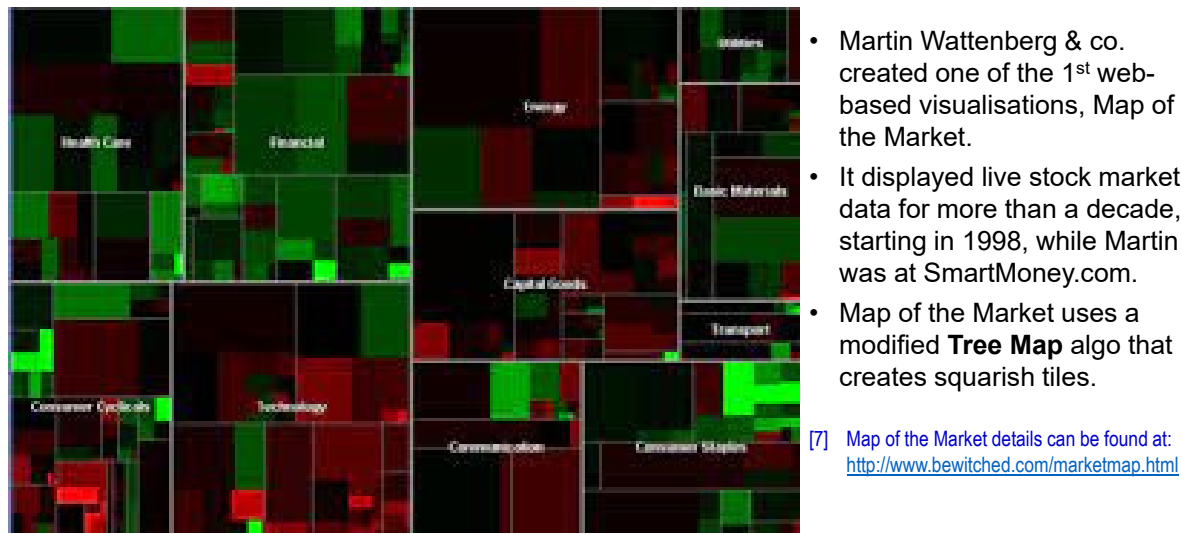
Deconstruction - Example #1



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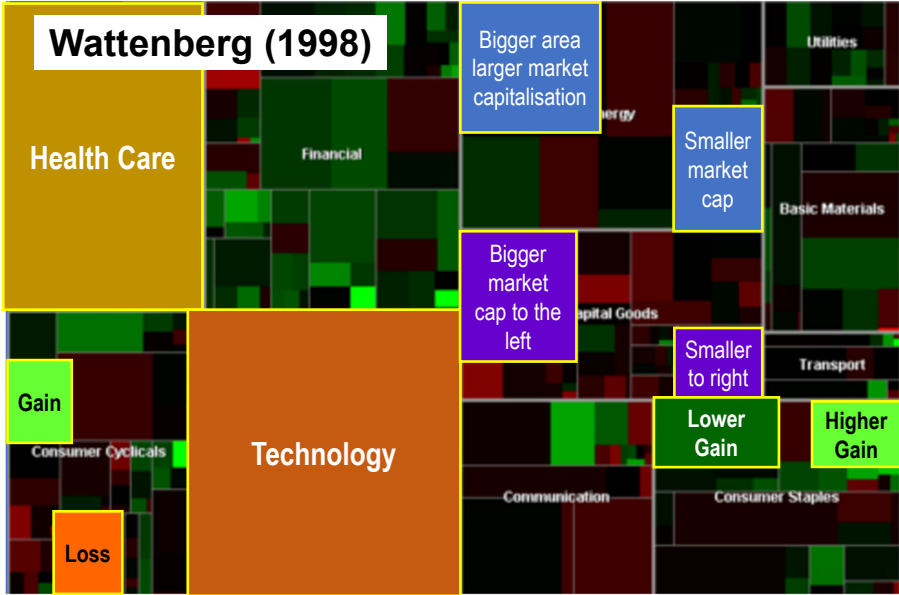
Deconstruction - Example #2

Martin Wattenberg's Map of the Market (1998)



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Deconstruction - Example #2



Visual Variables

Areas (Rectangle)

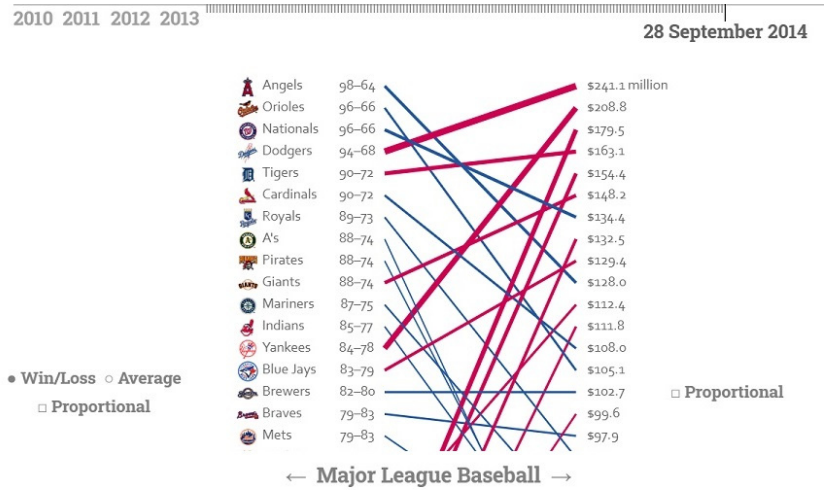
- Q 1. Size (Market Cap)
- N 2. Position (Market Sector)
- Q 3. Position (Market Cap)
- N 4. Colour Hue (Green - gain) (Red - loss)
- Q 5. Colour Saturation (Bright - higher gain) (Dark - lower gain) (Bright - higher loss) (Dark - lower loss)

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Deconstruction - Example #3

Ben Fry's Slope Chart (2010-2014)

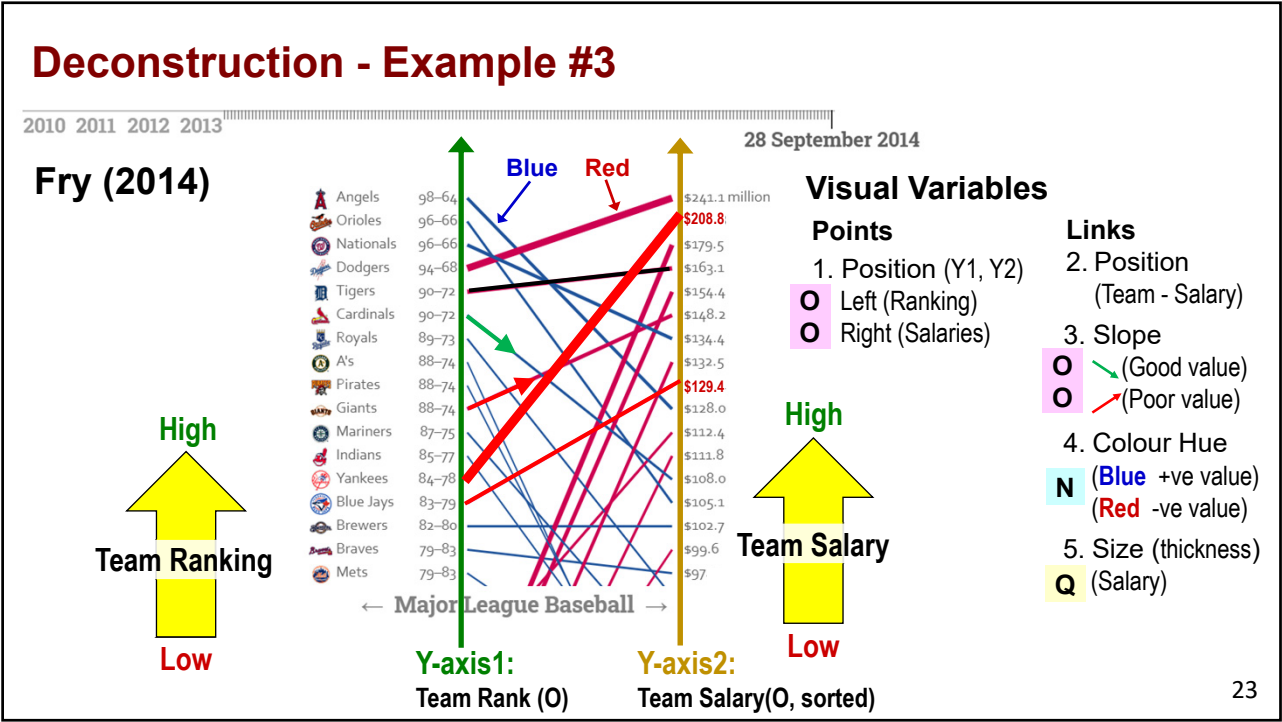


- Ben Fry is a visualisation expert and principal of Fathom, a design and software consultancy in Boston, USA.
- He designed Salary vs. Performance to visualised if baseball teams are spending their money well over the season.
- The visualisation uses a slope chart.

[8] Fry's website for Salary vs Performance: <https://benfry.com/salaryper/>

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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.
- The attributes of the various data dimensions can be appropriately encoded using a combinations of visual **marks** and **variables**.
- Effective visual encoding requires a good understanding of the characteristics and limitations of **human visual perception**.
- **Deconstructing** good visualisation examples can teach us how to do effective visual encoding using various visual marks and variables.

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## References for Visual Encoding and Deconstruction

- [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](https://books.google.com.sg/books?id=csqX_xnm4tcC&printsec=copyright&redir_esc=y#v=onepage&q&f=false)
- [2] T. Munzner, Visualization Analysis & Design, CRC Press (2015)
- [3] J. Mackinlay, Automating the Design of Graphical Presentation of Relational Information (1986), <https://dl.acm.org/doi/10.1145/22949.22950>
- [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](https://upload.wikimedia.org/wikipedia/commons/5/52/Playfair_TimeSeries-2.png)
- [6] Martin Wattenberg's Map of the Market details can be found at: <http://www.bewitched.com/marketmap.html>
- [7] Ben Fry's website for Salary vs Performance: <https://benfry.com/salaryper>



Note: All online articles were accessible on 5 Nov 2025

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