
Data Visualisation

(CMP020L013S)

Week 1: Introduction

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Agenda

- ▶ Introduction to the module and logistics
- ▶ Data, information, history of visualisation
- ▶ The need for visualisations, definitions

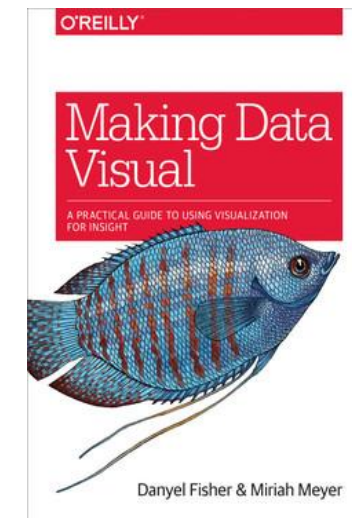
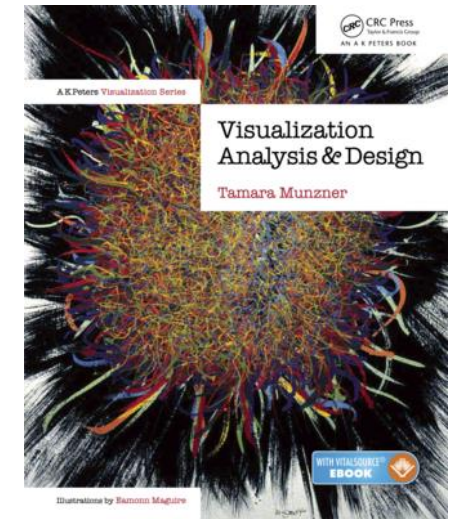
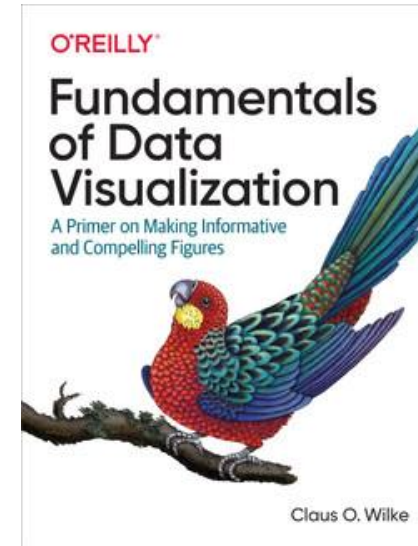
Learning Outcomes

- ▶ Critically analyse the effectiveness of a given visualisation technique for a particular task.
- ▶ Critically evaluate the appropriate visualisation technique for specific problems.
- ▶ Develop visualisations to report on studies of both quantitative and qualitative evaluations.
- ▶ Python skill are pre-requisite for this module

- ▶ Delivery approach
- ▶ Expectations, preparation, lab sessions
- ▶ Attendance monitoring

Module Material/Reading List

- ▶ Matplotlib, NumPy, and Seaborn etc.
- ▶ We will use **only libraries** which are **available/can be imported** on the uni machines



► We are Data Rich but Information Poor (DRIP syndrome)



Is data enough?

Data vs Information

- ▶ Data != Information
- ▶ Data Vis: data → information
 - ▶ Using power of human visual processing
 - ▶ Making visible the patterns and structures in the data
 - ▶ Using graphs, tables, diagrams, maps...

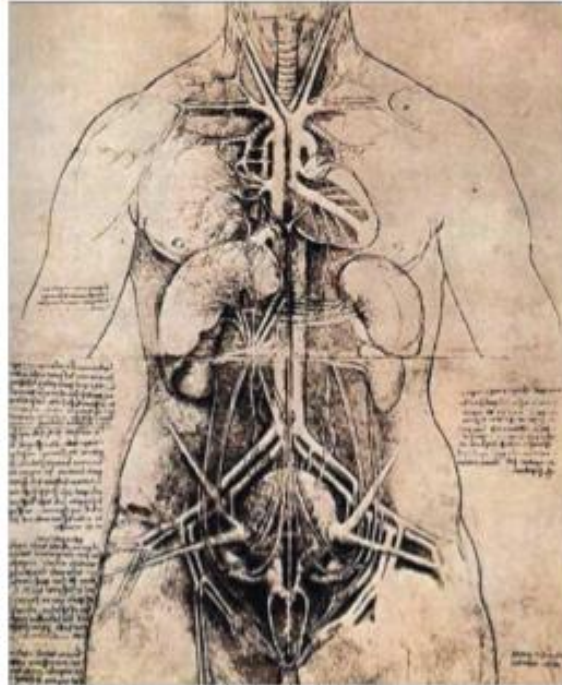
History of Visualisation

- ▶ Dr. John Snow
- ▶ Location of deaths in the 1854 London Cholera Epidemic
- ▶ X marks the locations of the water pumps

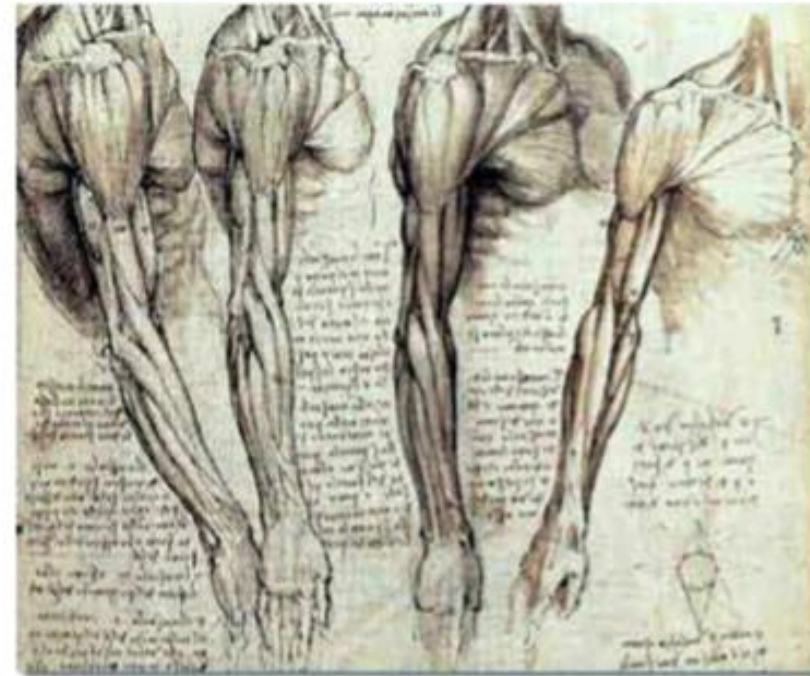


History of Visualization

- Visualization = rather old



L. da Vinci (1452-1519)



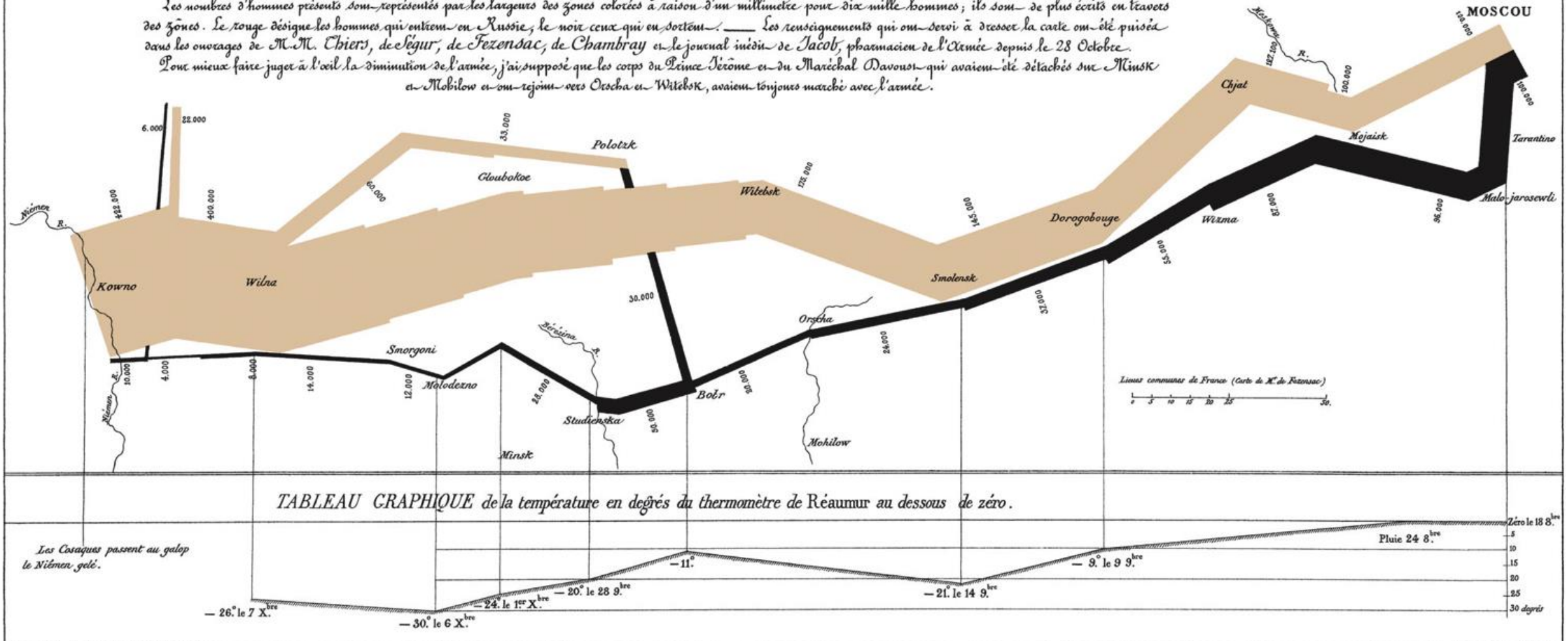
- Often an intuitive step: graphical illustration

Charles Joseph Minard 1869 Napoleon's March

Carte Figurative des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.
Dressée par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite. Paris, le 20 Novembre 1869.

Les nombres d'hommes présents sont représentés par les largeurs des zones colorées à raison d'un millimètre pour dix mille hommes; ils sont de plus écrits en travers des zones. Le rouge désigne les hommes qui ont péri en Russie, le noir ceux qui en sont sortis. Les renseignements qui ont servi à dresser la carte ont été puisés dans les ouvrages de M. M. Thiers, de Ségur, de Fezensac, de Chambray et le journal inédit de Jacob, pharmacien de l'Armée depuis le 28 Octobre.

Pour mieux faire juger à l'œil la diminution de l'armée, j'ai supposé que les corps du Prince Jérôme et du Maréchal Davout qui avaient été détachés sur Minsk et Mohilow et qui rejoignent vers Orscha et Witebsk, avaient toujours marché avec l'armée.



Autog. par Rognier, 8. Par. 5^{de} Maria 5^{de} 0^{de} à Paris.

Defining Visualisation

- ▶ The National Science Foundation (U.S.):
 - ▶ The use of **computer graphics** for the analysis and presentation of **computed** or **measured** scientific data.
- ▶ Oxford Engl. Dict., 1989
 - ▶ To form a mental vision, image, or picture of (something not visible or present to the sight, or of an abstraction); to make visible to the mind or imagination
- ▶ Visualization **transforms data into images** that effectively and accurately represent information about the data.

Defining Visualisation

- ▶ “Transformation of the symbolic into the geometric”
[McCormick et al. 1987]
- ▶ “... finding the artificial memory that best supports our natural means of perception.” [Bertin 1967]
- ▶ “The use of computer-generated, interactive, visual representations of data to amplify cognition.”
[Card, Mackinlay, & Shneiderman 1999]

Defining Visualisation

- ▶ Visualisation is the process that transforms (abstract) data into interactive graphical representations for the purpose of exploration, confirmation, or presentation.
- ▶ **Tool to enable a user insight into Data**

Defining Visualisation

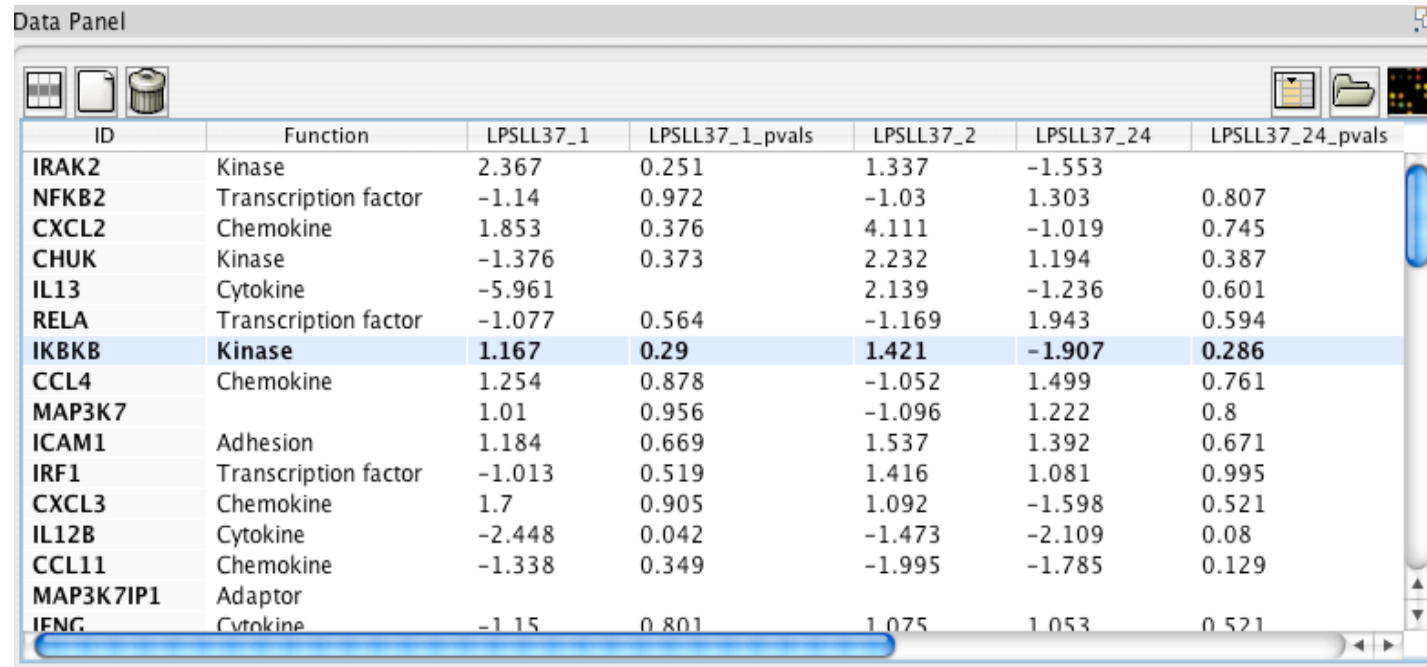
- ▶ Computer-based visualisation systems provide **visual** representations of datasets designed to help **people** carry out tasks more effectively.
- ▶ Why we need Vis?
- ▶ Why have a human in the loop?

Defining Visualisation

- ▶ Computer-based visualisation systems provide **visual representations** of **datasets** designed to help **people** carry out tasks more **effectively**.
- ▶ Visualization is suitable when there is a need to **augment human capabilities** rather than replace people with computational **decision-making** methods.

Why use an external representation?

- **visual representations as** external representation: replace cognition with perception



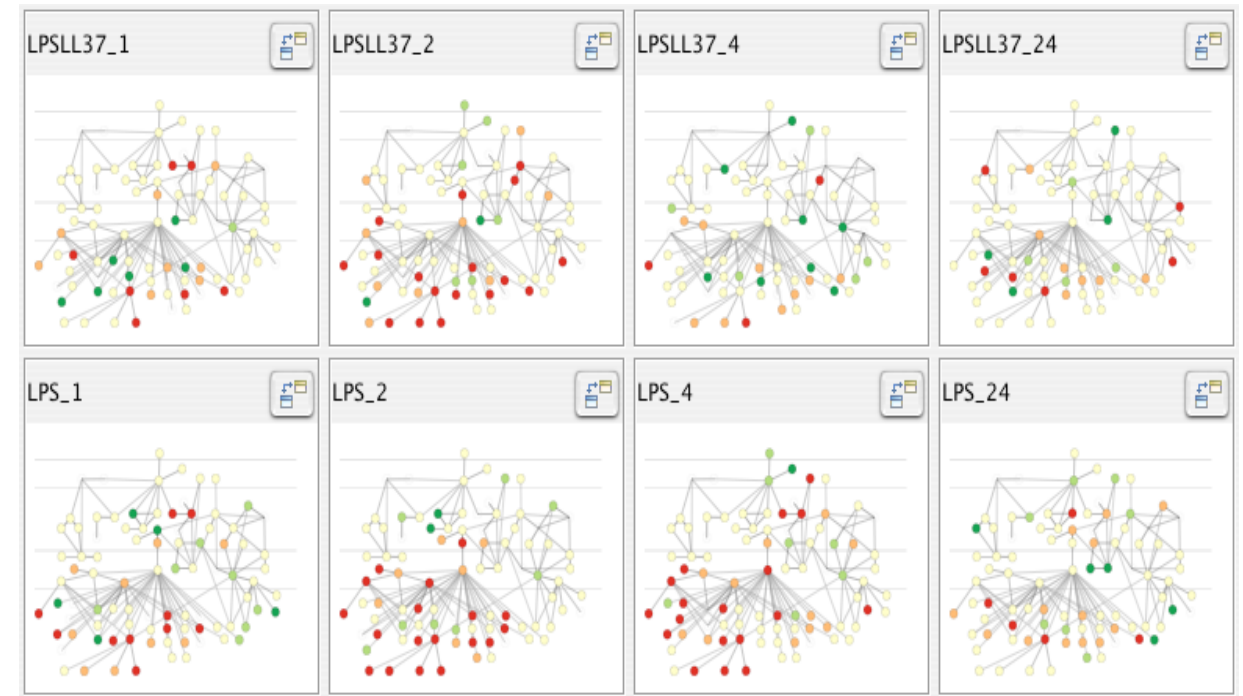
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IRAK2	Kinase	2.367	0.251	1.337	-1.553	
NFKB2	Transcription factor	-1.14	0.972	-1.03	1.303	0.807
CXCL2	Chemokine	1.853	0.376	4.111	-1.019	0.745
CHUK	Kinase	-1.376	0.373	2.232	1.194	0.387
IL13	Cytokine	-5.961		2.139	-1.236	0.601
RELA	Transcription factor	-1.077	0.564	-1.169	1.943	0.594
IKKB	Kinase	1.167	0.29	1.421	-1.907	0.286
CCL4	Chemokine	1.254	0.878	-1.052	1.499	0.761
MAP3K7		1.01	0.956	-1.096	1.222	0.8
ICAM1	Adhesion	1.184	0.669	1.537	1.392	0.671
IRF1	Transcription factor	-1.013	0.519	1.416	1.081	0.995
CXCL3	Chemokine	1.7	0.905	1.092	-1.598	0.521
IL12B	Cytokine	-2.448	0.042	-1.473	-2.109	0.08
CCL11	Chemokine	-1.338	0.349	-1.995	-1.785	0.129
MAP3K7IP1	Adaptor					
IFNG	Cytokine	-1.15	0.801	1.075	1.053	0.521

[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. IEEE TVCG (Proc. InfoVis) 14(6):1253-1260, 2008.]

Why use an external representation?

Data Panel

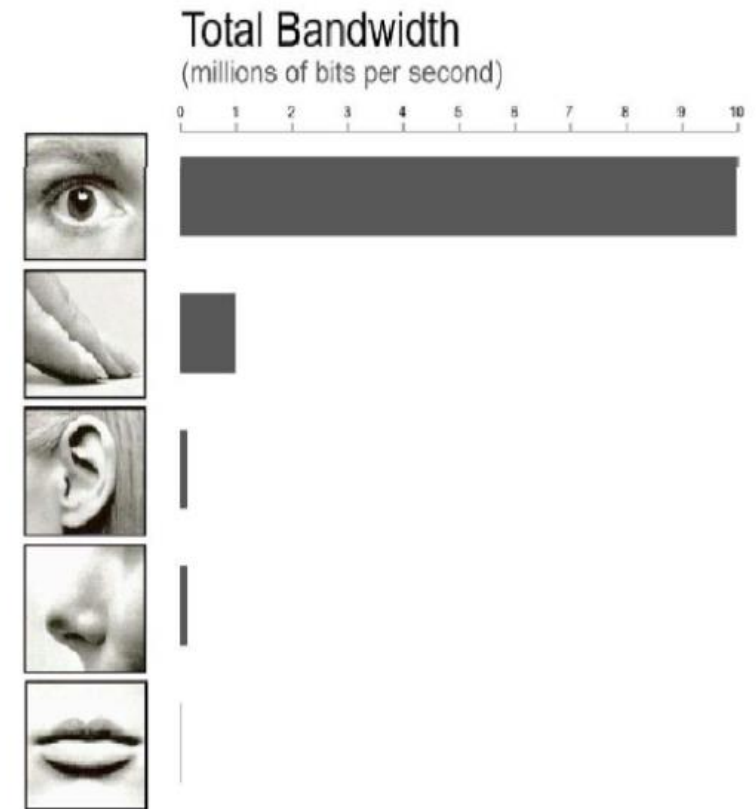
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Why depend on vision?

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

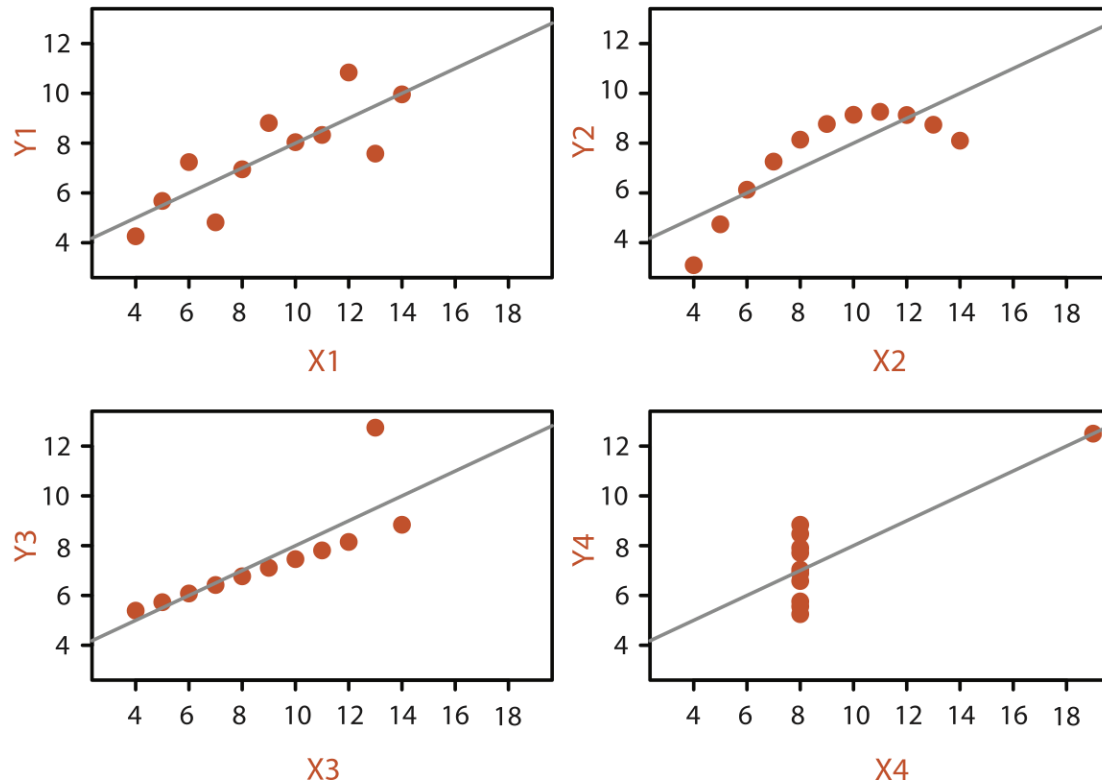


Why depend on vision?

- ▶ Human visual system is a high-bandwidth channel to the brain
 - ▶ overview possible due to background processing
 - ▶ subjective experience of seeing everything simultaneously
 - ▶ significant processing occurs in parallel and pre-attentively
- ▶ Sound: lower bandwidth and different semantics
 - ▶ overview not supported
 - ▶ subjective experience of sequential stream
- ▶ Touch/haptics: impoverished record/replay capacity
 - ▶ only very low-bandwidth communication thus far
- ▶ Taste/smell: no viable record/replay devices

Why represent all the data?

- ▶ summaries lose information, details matter
 - ▶ confirm expected and find unexpected patterns
 - ▶ assess the validity of the statistical model



Identical statistics

x mean	9
x variance	10
y mean	7.5
y variance	3.75
x/y correlation	0.816

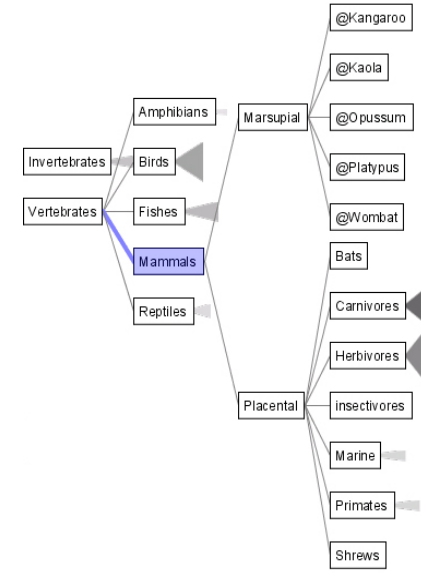
What resource limitations are we faced with?

- ▶ Vis designers must take into account three very different kinds of resource limitations:
- ▶ Display limits
 - ▶ pixels are precious and the most constrained resource
 - ▶ **information density:**
 - ▶ ratio of space used to encode info vs unused whitespace
 - ▶ tradeoff between clutter and wasting space
 - ▶ find a sweet spot between dense and sparse
- ▶ human limits
 - ▶ human time, human memory, human attention

Why analyse Visualition?

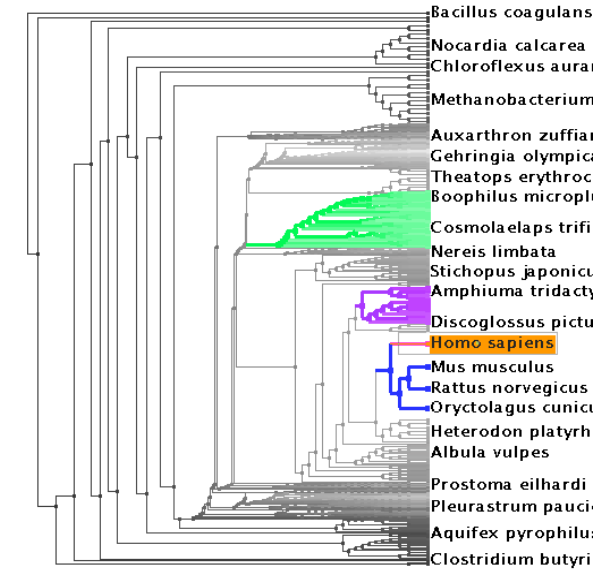
- ▶ Helps to think **systematically** about choices of visualisation
- ▶ Analysing **existing visualisations** as a stepping stone **to designing new ones**
- ▶ Vis design is full of **trade-offs**,
- ▶ most possibilities in the design space are ineffective for a particular task
- ▶ Validating the effectiveness of a design is both **necessary** and **difficult**

SpaceTree



[SpaceTree: Supporting Exploration in Large Node Link Tree, Design Evolution and Empirical Evaluation. Grosjean, Plaisant, and Bederson. Proc. InfoVis 2002, p 57–64.]

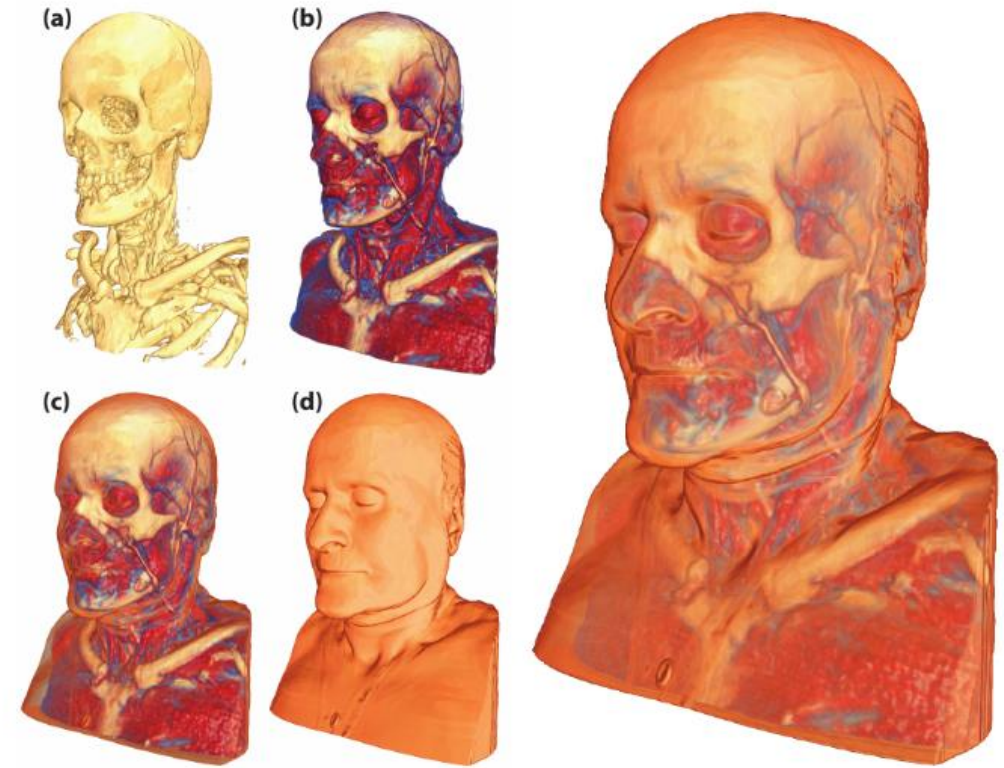
TreeJuxtaposer



[TreeJuxtaposer: Scalable Tree Comparison Using Focus+Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453–462, 2003.]

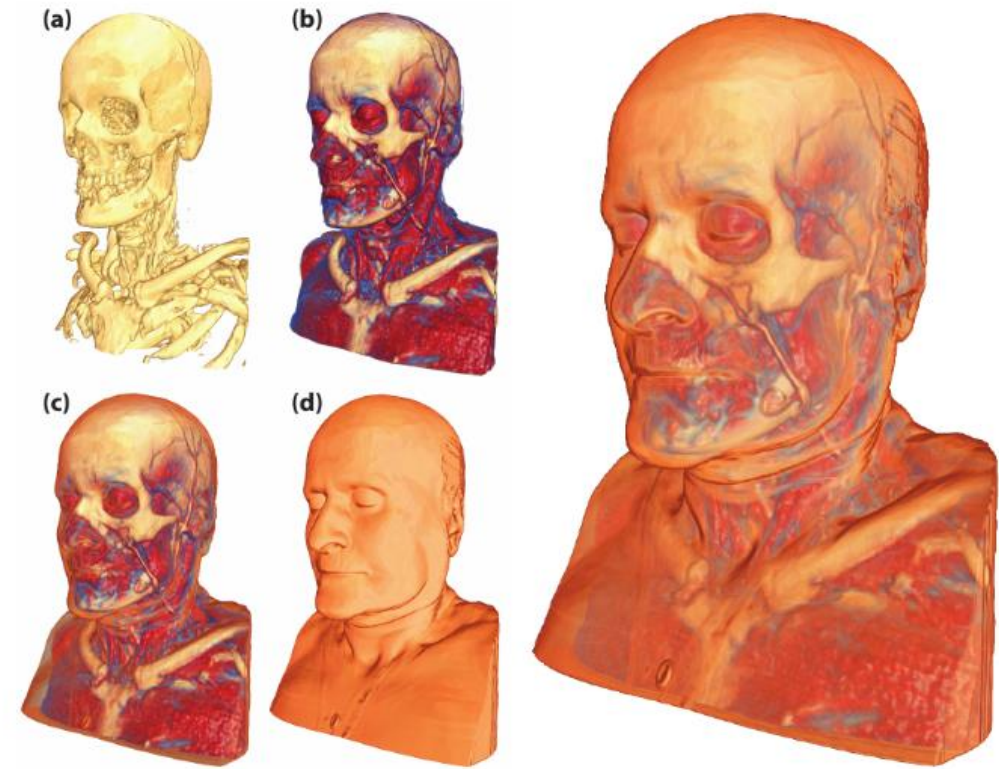
Why Use Computers?

- ▶ Scale
- ▶ Drawing by hand(Illustrator)
- ▶ Infeasible
- ▶ inflexible (updates!)
- ▶ How to draw an MRI scan?



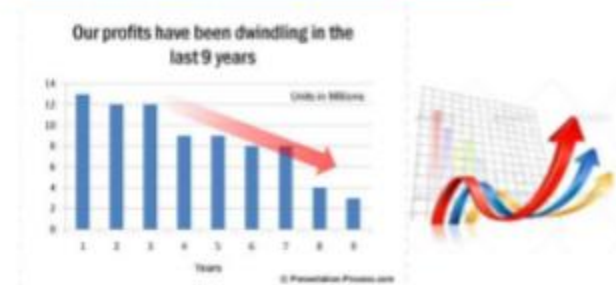
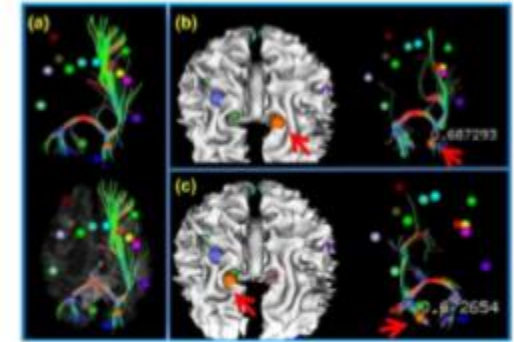
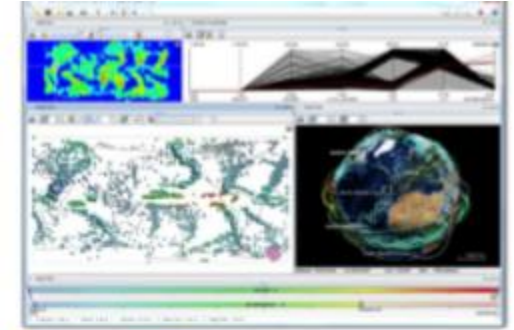
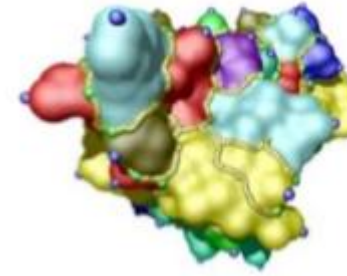
Why Use Computers?

- ▶ Integration with algorithms
- ▶ Make visualization part of a data analysis pipeline
- ▶ Efficiency
 - ▶ Re-use charts / methods for different datasets
- ▶ Quality
 - ▶ Precise data driven rendering
- ▶ Storytelling
 - ▶ Use time



Goals of Visualisation

- ▶ Three goals for visualisation
- ▶ Explore
 - ▶ **Nothing is known** about the data
 - ▶ Vis is used for data exploration
- ▶ Analyse
 - ▶ There are hypotheses/assumptions
 - ▶ Vis is used for verification or falsification
- ▶ Present
 - ▶ **Everything is known** about the data
 - ▶ Vis is used to communicate the results



Further Applications of Visualisation

A: Support reasoning about information (analysis)

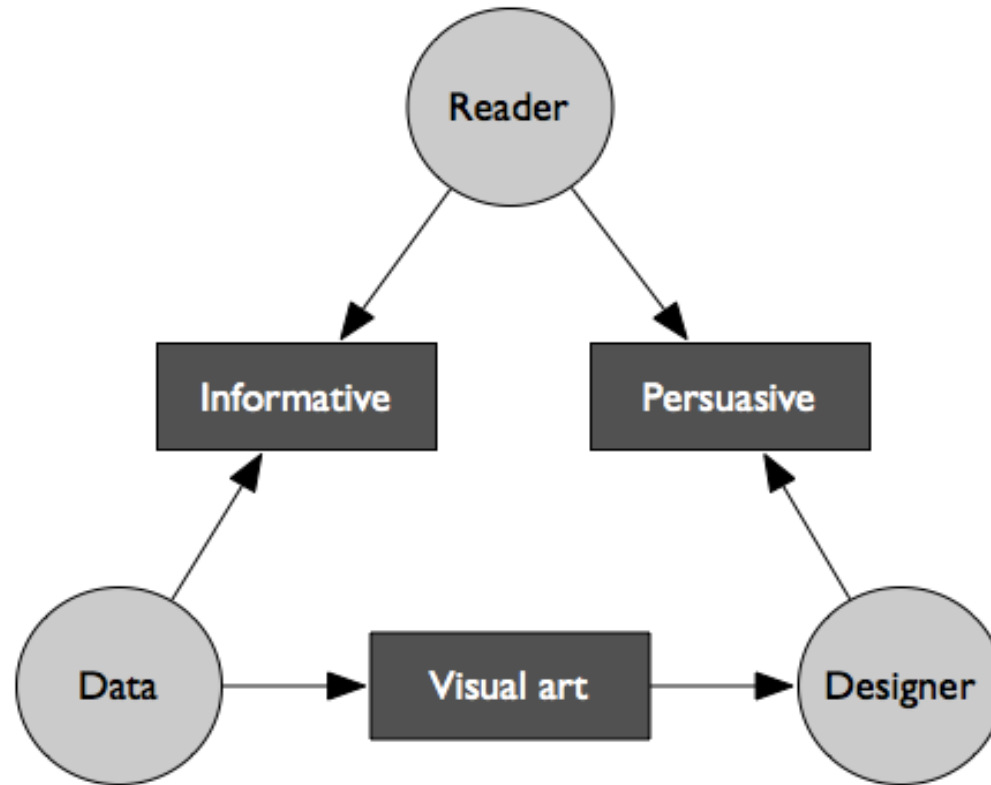
- ▶ Finding relationships
- ▶ Discover structure
- ▶ Quantifying values and influences
- ▶ Should be part of a query/analysis cycle

B: Inform and persuade others (communication)

- ▶ Capture attention, engage
- ▶ Tell a story visually
- ▶ Focus on certain aspects, and omit others

Data Presentation Trinity in Vis

► Designer-Reader-Data Trinity



“Designing Data Visualizations”, Iliinsky and Steele, O’Reilly, 2011

- ▶ Types of Data Visualisation
- ▶ Use Cases of Data Visualisation
- ▶ Value of Data Visualisation
- ▶ Good, bad and ugly visualiations

Further Material

- ▶ The famous Gapminder Video, Hans Rosling:
200 Countries, 200 Years, 4 Minutes

- ▶ https://www.youtube.com/watch?feature=player_embedded&v=jbkSRLYSojo

- ▶ Florence Nightingale was one of the pioneers of data visualisation

- ▶ https://youtu.be/yhX0OR1_Vfc

