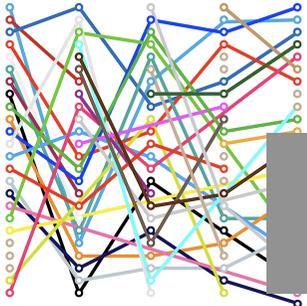


VU 188.305 – VO: 2 hrs. / 3 ECTS



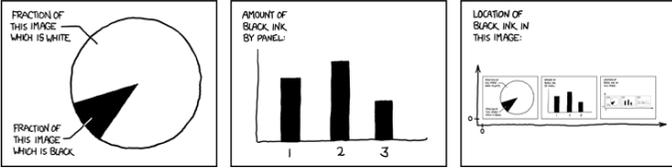
Information Visualization

Visual Perception, Cognition & Visual Encoding

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TU WIEN FAKULTÄT FÜR INFORMATIK Faculty of Informatics
QVAST Vienna Research Center for Analytics Science & Technology
isis.
ieg information engineering group



FRACTION OF THIS IMAGE WHICH IS WHITE.

FRACTION OF THIS IMAGE WHICH IS BLACK.

AMOUNT OF BLACK INK BY PANEL:

LOCATION OF BLACK INK IN THIS IMAGE:

[<http://xkcd.com/688/>]

WOLFGANG AIGNER perception, cognition and visual encoding **TU WIEN**

Contents

- What is perception
- Anatomy of the visual system
- Visual processing
- Color perception
- Preattentive processing
- Gestalt principles
- Cognitive models
- Visual encoding & graphical excellence

WOLFGANG AIGNER

perception, cognition and visual encoding



Perception

Perception deals with the human senses that generate signals from the environment through sight, hearing, touch, smell and taste.

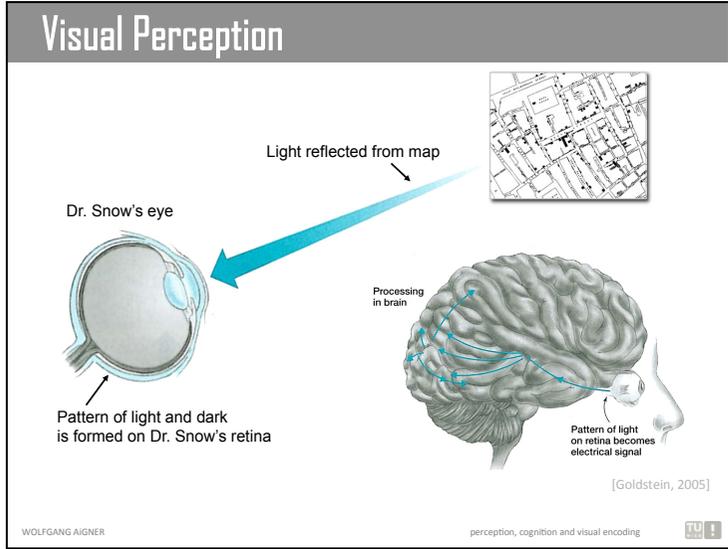
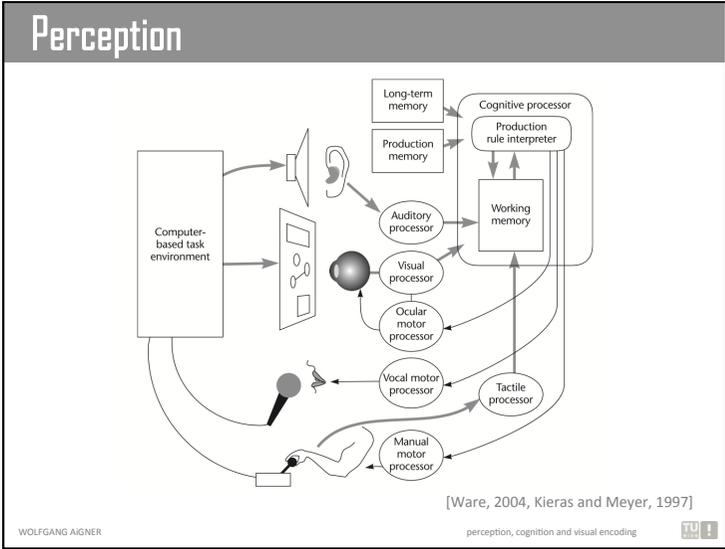
[...] the process of recognizing (being aware of), organizing (gathering and storing), and interpreting (binding to knowledge) sensory information

[Ward et al., 2010]

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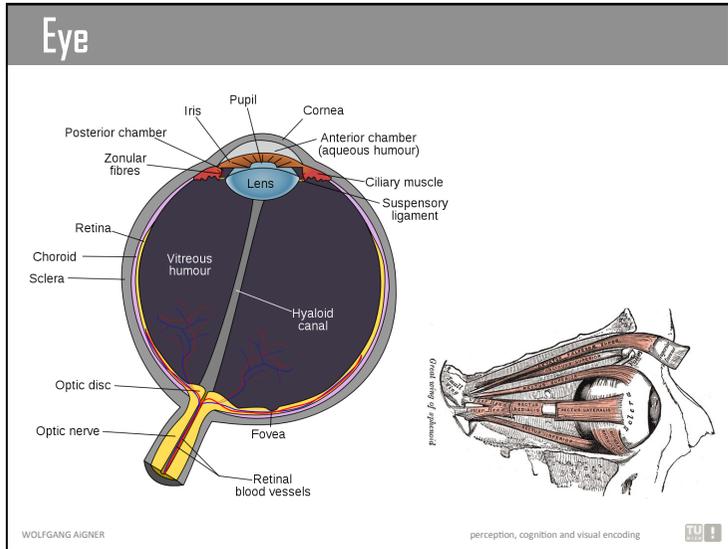
perception, cognition and visual encoding





ANATOMY OF THE VISUAL SYSTEM

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Retina

The diagram shows a cross-section of the eye with labels: Sclera, Choroid, Retina, Cornea, Pupil, Lens, Iris, Ciliary body, and Optic nerve. An inset shows the layered structure of the retina with rods and cones. A graph plots Normalized absorbance (0-100) against Wavelength (nm) from 400 to 700. It shows three curves: S (Short wavelength, peak at 420 nm), M (Medium wavelength, peak at 498 nm), and L (Long wavelength, peak at 564 nm). A dashed line represents the combined response.

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Brain Pixel

A circular arrangement of pixels, where the size of each pixel increases as it moves away from the center, illustrating the concept of a 'Brain Pixel' or receptive field size.

[Ware, 2008]

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VISUAL PROCESSING

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Visual Perception

An axial MRI scan of the brain. The optic nerves are highlighted in orange and labeled 'Optic Nerve'. The primary visual cortex is also highlighted in orange and labeled 'Primary Visual Cortex'.

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Perceptual processing

intrinsic and uncontrolled (preattentive)

fast and is performed in parallel, often within 250ms

limited set of visual properties that are detected very rapidly and accurately by the low-level visual system

controlled (attentive)

slower and uses short-term memory

[Ward et al., 2010]

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TU

Model of visual processing

The diagram illustrates a model of visual processing. It starts with a light source on the left illuminating a scene. A camera lens captures the scene, which is then processed by a series of layers. The first layer is labeled 'Features' and shows a grid of small circles representing parallel processing. The second layer is labeled 'Patterns' and shows a grid of larger circles representing pattern building. The final layer is labeled 'Objects' and shows a dog and a person sitting on a bench, with arrows pointing to them from the pattern layer. Below the diagram, two arrows indicate the flow of information: a blue arrow pointing right labeled 'Bottom-up information drives pattern building' and a red arrow pointing left labeled 'Top-down attentional processes reinforce relevant information'.

Features are processed in parallel from every part of the visual field. Millions of features are processed simultaneously.

Patterns are built out of features depending on attentional demands. Attentional tuning reinforces those most relevant.

Objects most relevant to the task at hand are held in Visual Working Memory. Only between one and three are held at any instant. Objects have both non-visual and visual attributes.

→ Bottom-up information drives pattern building
← Top-down attentional processes reinforce relevant information

[Ware, 2008]

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Eye movements

- Saccadic movements
- Smooth-pursuit movements
- Convergent movements

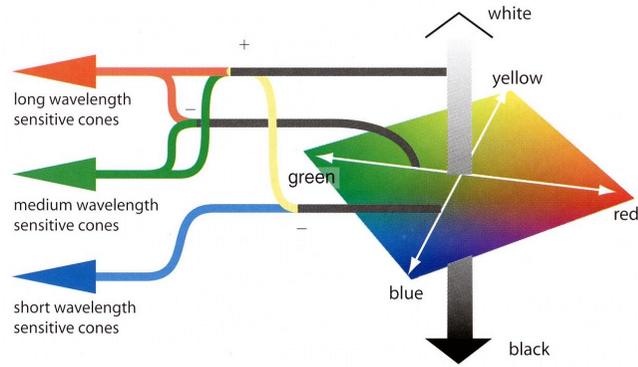
An anatomical diagram of the human eye, showing the various muscles that control its movement. The muscles are labeled with their names: Rectus Superior, Rectus Medialis, Rectus Lateralis, Rectus Inferior, Obliquus Superior, Obliquus Inferior, and Ciliary Muscles. The diagram also shows the optic nerve and the ciliary body. The text 'Tread suture of sphenoid' is written vertically on the left side of the diagram.

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COLOR PERCEPTION

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Opponent process theory



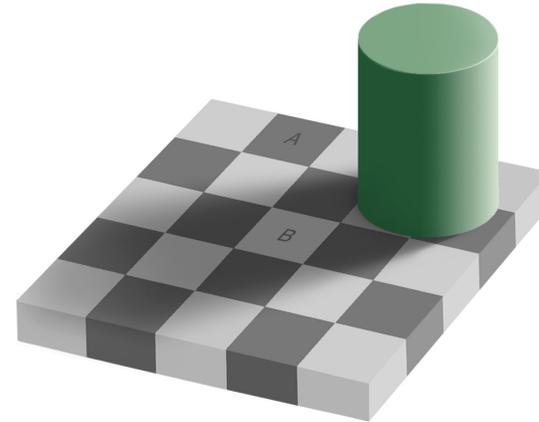
[Ware, 2008]

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Contrast

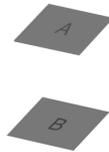


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Contrast



WOLFGANG AIGNER

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Contrast

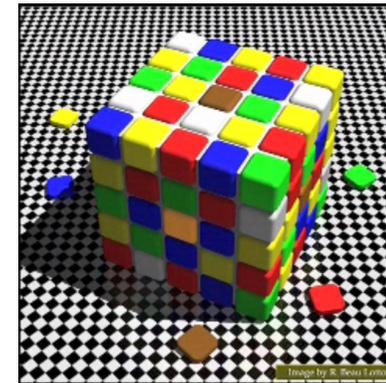


Image by S. Bowler

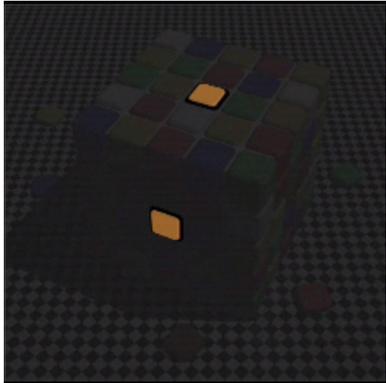
[Beau Lotto]

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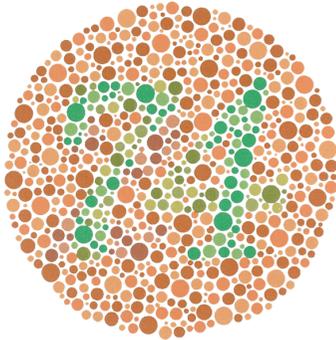
Contrast



[Beau Lotto]

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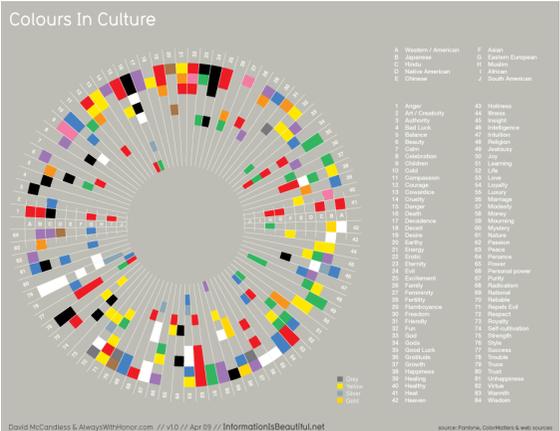
Color blindness



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Semantics of color

Colours In Culture



Category	Color
1. Anger	Red
2. Art / Creativity	Yellow
3. Authority	Black
4. Bad Luck	White
5. Balance	Grey
6. Beauty	Blue
7. Calm	Green
8. Celebration	Purple
9. Danger	Orange
10. Debt	Brown
11. Compassion	Pink
12. Courage	Light Blue
13. Cowardice	Light Green
14. Cruelty	Light Purple
15. Danger	Light Orange
16. Death	Light Brown
17. Deceit	Light Yellow
18. Denial	Light Grey
19. Dislike	Light Blue-Grey
20. Dislike	Light Green-Grey
21. Dislike	Light Purple-Grey
22. Dislike	Light Orange-Grey
23. Dislike	Light Brown-Grey
24. Dislike	Light Yellow-Grey
25. Dislike	Light Grey-Grey
26. Dislike	Light Blue-Grey
27. Dislike	Light Green-Grey
28. Dislike	Light Purple-Grey
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81. Dislike	Light Grey-Grey
82. Dislike	Light Blue-Grey
83. Dislike	Light Green-Grey
84. Dislike	Light Purple-Grey

David McCandless & AlwaysWithAStonor.com // v1.0 // Apr 09 // InformationIsBeautiful.net

WOLFGANG AIGNER [McCandless, 2009] perception, cognition and visual encoding 

PREATTENTIVE PROCESSING

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Preattentive Processing

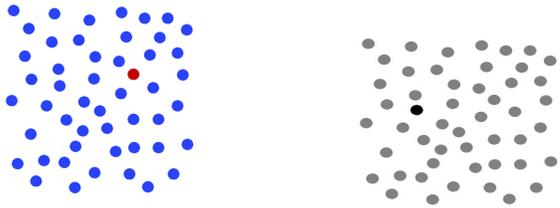
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 03707774179527931749270973401
 9743217909370945179279417

15613212036584130765103746274
 17312752732759273299070974217
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Color

Hue Intensity

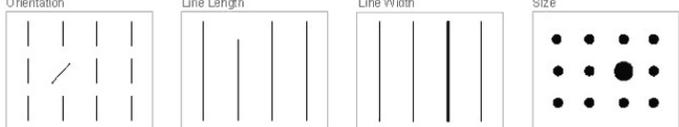


[Dürsteler, 2006]

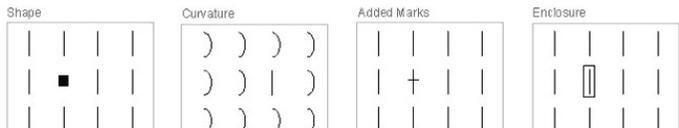
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Form

Orientation Line Length Line Width Size



Shape Curvature Added Marks Enclosure

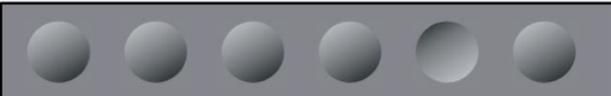


Applet:
<http://www.csc.ncsu.edu/faculty/healey/PP/index.html> [Few, 2004]

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Spatial position

2D position
 Stereoscopic depth
 Concavity convexity



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Movement

(Direction of) motion

Flicker

[Dürsteler, 2006]

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ELEMENTARY GRAPHICAL PERCEPTION TASKS

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Visual variables

More accurate

Less accurate

[Cleveland & McGill, 1984]

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Visual variables & data types

Quantitative	Ordinal	Nominal
Position	Position	Position
Length	Density	Color Hue
Angle	Color Saturation	Texture
Slope	Color Hue	Connection
Area	Texture	Containment
Volume	Connection	Density
Density	Containment	Color Saturation
Color Saturation	Length	Shape
Color Hue	Angle	Length
Texture	Slope	Angle
Connection	Area	Slope
Containment	Volume	Area
Shape	Shape	Volume

[Mackinlay, 1987]

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LAWS

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Relative judgements

Which of the two bars is longer?



[Cleveland and McGill, 1984]

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Weber's Law

the likelihood of detecting a change is proportional to the **relative change**, not the absolute change, of a graphical attribute

[Ward et al., 2010]

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Steven's Law



As the **dimension** of an attribute **increases**, the degree at which we **underestimate it increases**

[Ward et al., 2010]

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GESTALT PRINCIPLES

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Proximity



[Winkler, 2005]

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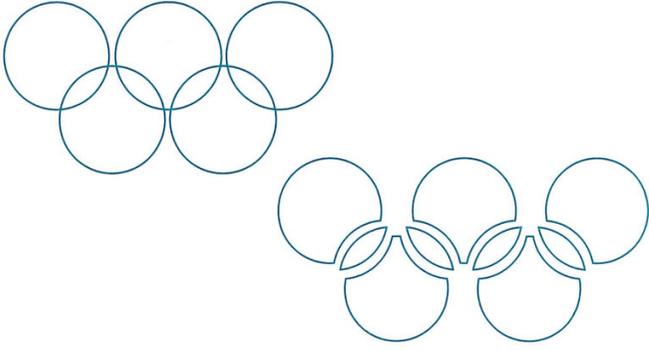
Similarity



[Winkler, 2005]

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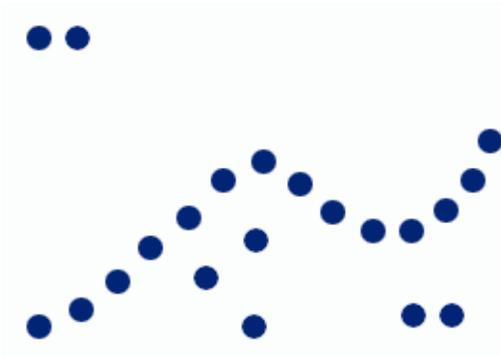
Prägnanz



[Goldstein, 2005]

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Good Continuation



[Winkler, 2005]

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Common Fate



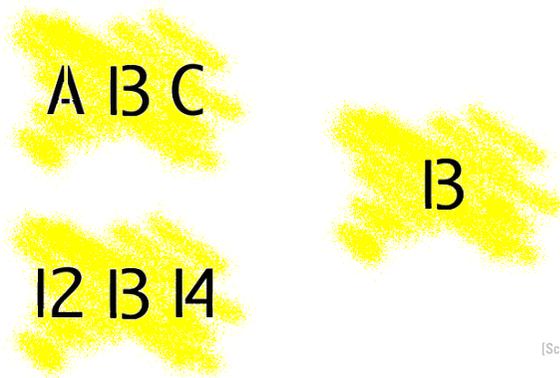
[Pedroza, 2005]

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Familiarity



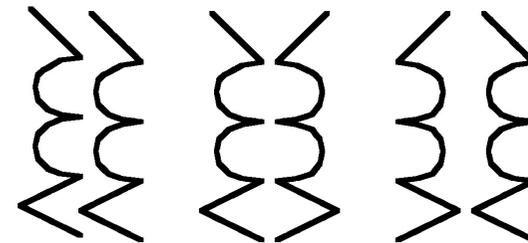
[Schmidt, 2005]

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Symmetry



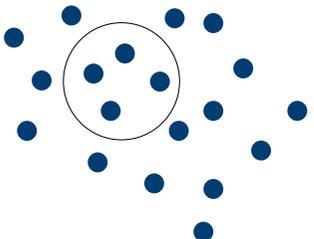
[Ware, 2004]

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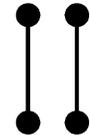


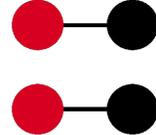
Enclosure

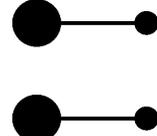


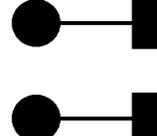
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Connection

a 

b 

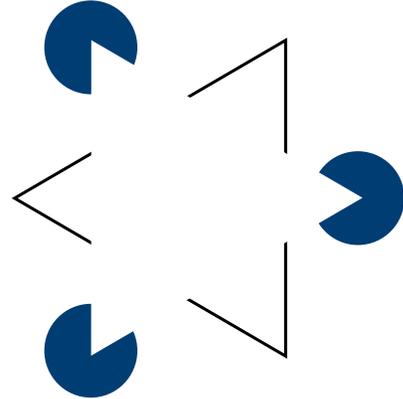
c 

d 

[Ware, 2004]

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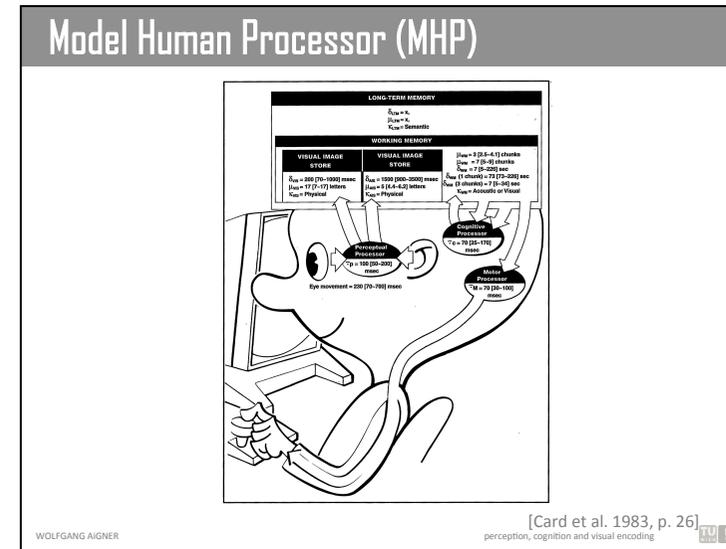
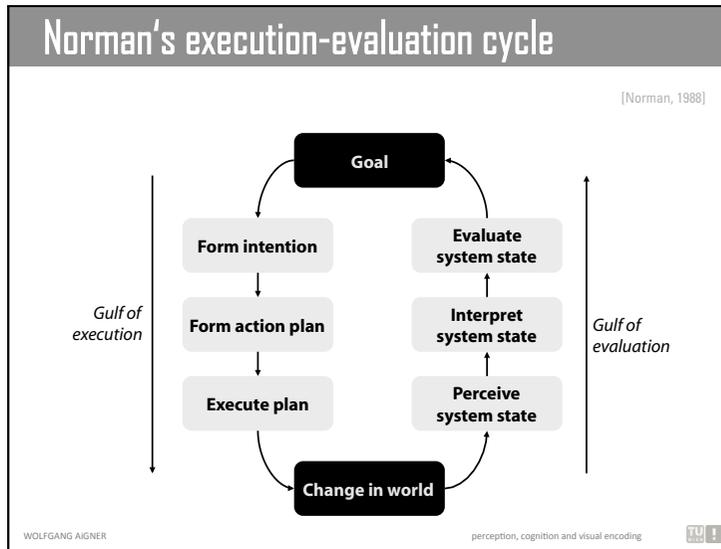
Closure



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COGNITIVE MODELS

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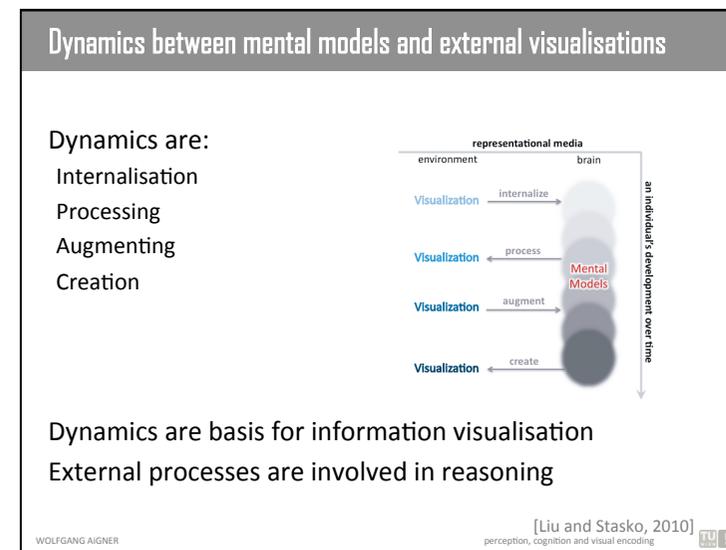
Mental models & visual reasoning

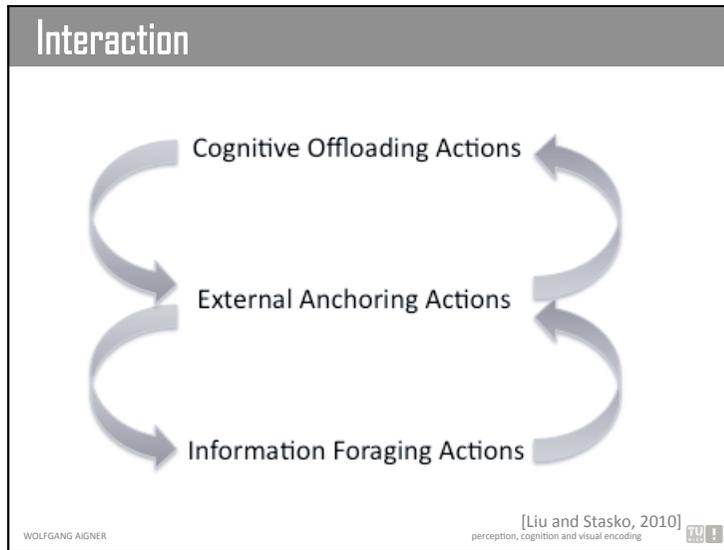
What is a mental model?
 Internal representation of real-world phenomenon
 Logical reasoning based on formal logic
 Functional
 Structural
 → functional, analogue representation to an external interactive visualisation system

Where are mental models situated?
 created in working memory (limited capacity)
 High relationship between mental models and external visualisation

What are mental models used for?
 Reasoning
 Constructing, simulating and manipulating mental models and external visualisations
 Theories for design, evaluation and theory of development

[Liu and Stasko, 2010]
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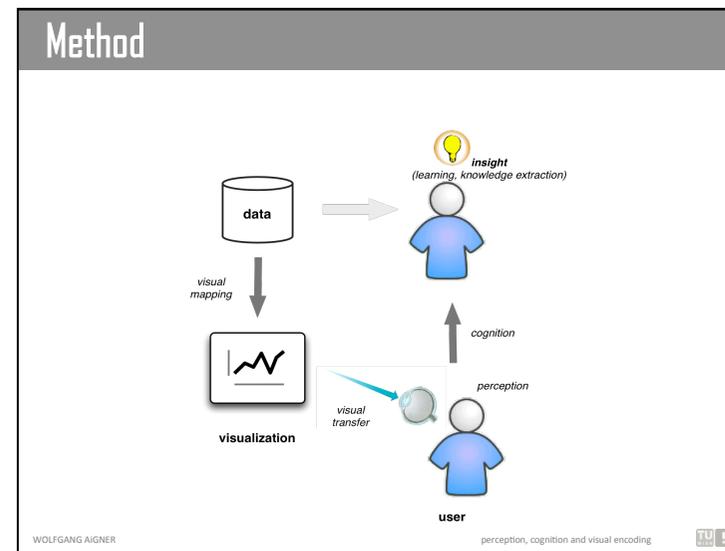
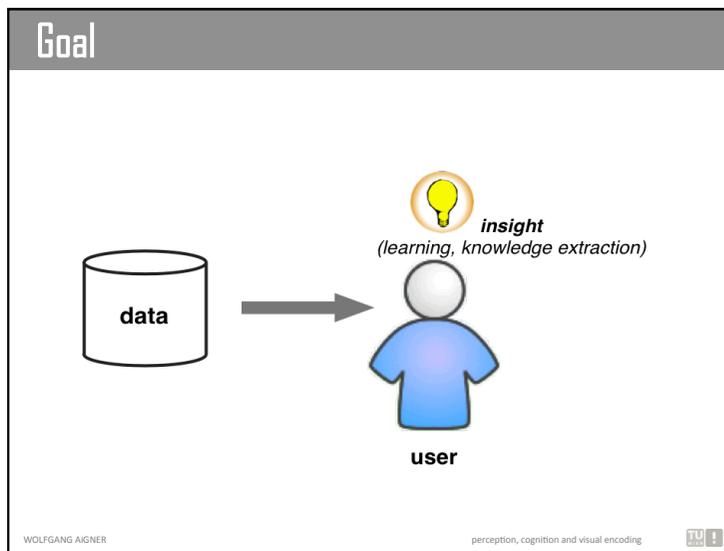


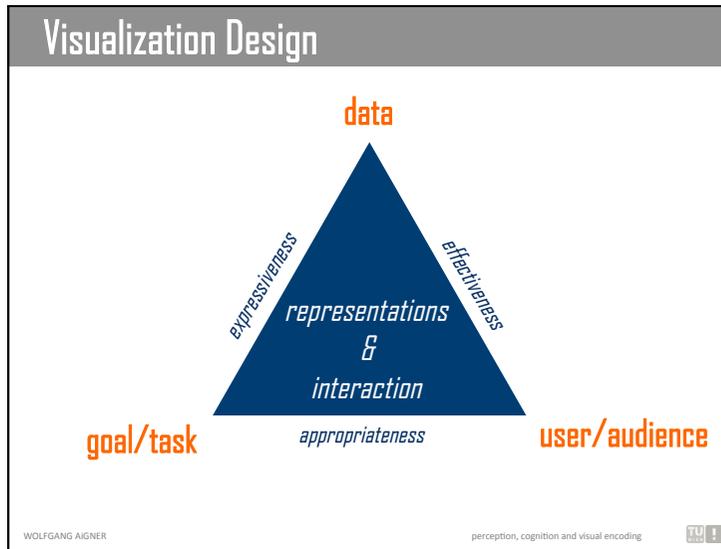


VISUAL ENCODING & GRAPHICAL EXCELLENCE

WOLFGANG AIGNER

perception, cognition and visual encoding





Data

cmp. [Shneiderman, 1996]

Variables –
Scale of measurement

- nominal
- ordinal
- quantitative
 - discrete
 - continuous
- (binary)

Data Type -
Structure

- univariate
- multivariate
- hierarchy / tree
- graph / network
- time-oriented
- text/document

(2D map)
(3D world)

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Data: Example

FilmID	230	105	540	...
Title	Goldfinger	Ben Hur	Ben Hur	...
Director	Hamilton	Wyler	Niblo	...
Actor	Connery	Heston	Novarro	...
Actress	Blackman	Harareet	McAvoy	...
Year	1964	1959	1926	...
Length	112	212	133	...
Popularity	7.7	8.2	7.4	...
Rating	PG	G	G	...
Film Type	Action	Action	Drama	...

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Data: Example

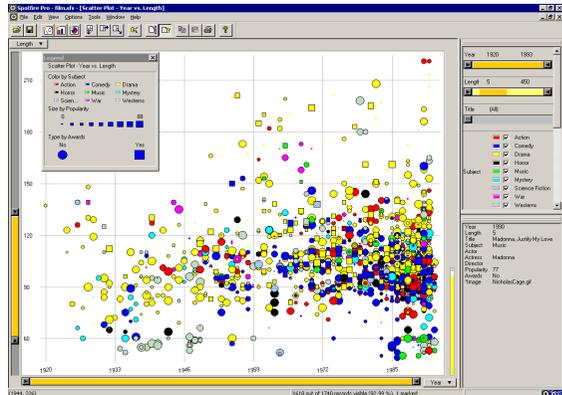
Nominal
Ordinal
Quantitative

FilmID	230	105	540	...
Title	Goldfinger	Ben Hur	Ben Hur	...
Director	Hamilton	Wyler	Niblo	...
Actor	Connery	Heston	Novarro	...
Actress	Blackman	Harareet	McAvoy	...
Year	1964	1959	1926	...
Length	112	212	133	...
Popularity	7.7	8.2	7.4	...
Rating	PG	G	G	...
Film Type	Action	Action	Drama	...

Data type / structure: multivariate

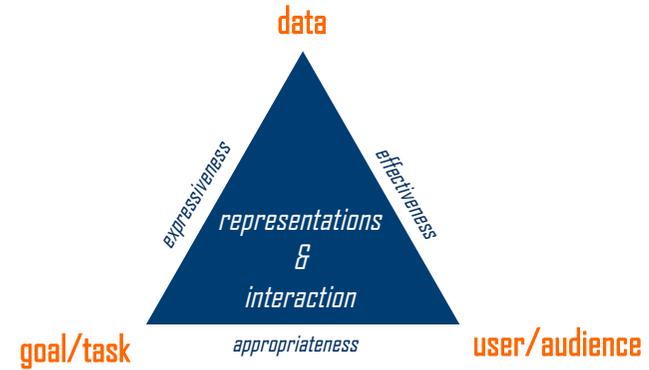
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Visual Mapping



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Visualization Design



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Expressiveness

A visualization is considered to be **expressive** if the **relevant information of a dataset (and only this) is expressed by the visualization**. The term "relevant" implies that expressiveness of a visualization can only be assessed regarding a **particular user** working with the visual representation to achieve **certain goals**.

„A visualization is said to be expressive if and only if it encodes all the data relations intended and no other data relations.“ [Card, 2008, p. 523]

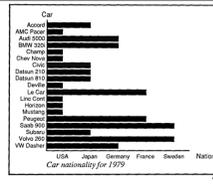


Fig. 11. Incorrect use of a bar chart for the Nation relation. The lengths of the bars express an ordering on the vertical axis, as if the USA cars were bigger or better than the other cars, which is not true for the Nation relation.

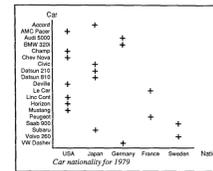


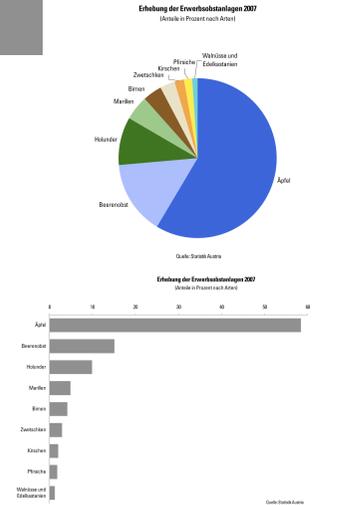
Fig. 12. Correct use of a plot chart for the Nation relation. Since bar charts encode ordered domain sets, plot charts are conventionally used to encode nominal domain sets. The ordering of the labels on the axes is ignored.

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Effectiveness

A visualization is effective if it **addresses the capabilities of the human visual system**. Since perception, and hence the mental image of a visual representation, varies among users, **effectiveness is user-dependent**. Nonetheless, some general rules for effective visualization have been established in the visualization community.

„Effectiveness criteria identify which of these graphical languages [that are expressive], in a given situation, is the most effective at exploiting the capabilities of the output medium and the human visual system.“ [Mackinlay, 1986]



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Appropriateness

(Schumann and Müller, 2000)

Appropriateness regards the **tradeoff** between efforts required for creating the visual representation and the **benefits** yielded by it. If this tradeoff is balanced, the visualization is considered to be appropriate.

Model of Van Wijk:

n users use visualization V to visualize a data set m times each where each session takes k exploratory steps and time T

- C_i ... Initial development costs
- C_u ... Initial costs per user (e.g., selection, acquisition, learning, tailoring)
- C_s ... Initial costs per session (e.g., data conversion, specification)
- C_e ... Perception and exploration costs (e.g., spend time to view and understand, modify, and tune)
- $W(\Delta K)$... Value of acquired knowledge $\Delta K = K(T) - K(0)$

Total costs:
 $C = C_i + n \cdot C_u + n \cdot m \cdot C_s + n \cdot m \cdot k \cdot C_e$
 Overall profit:
 $F = n \cdot m \cdot (W(\Delta K) - C_s - k \cdot C_e) - C_i - n \cdot C_u$

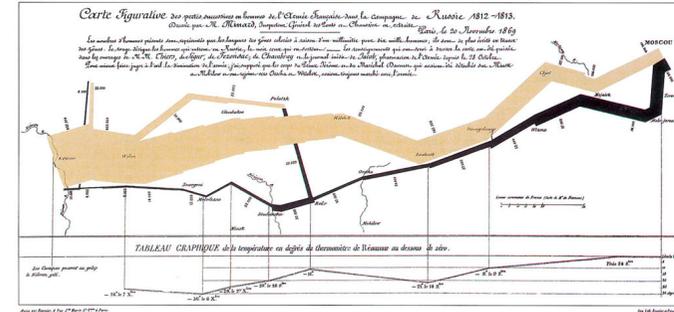
(Van Wijk, 2006)

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Graphical Excellence

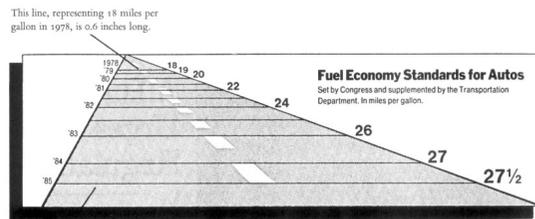


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Tell the truth about the data



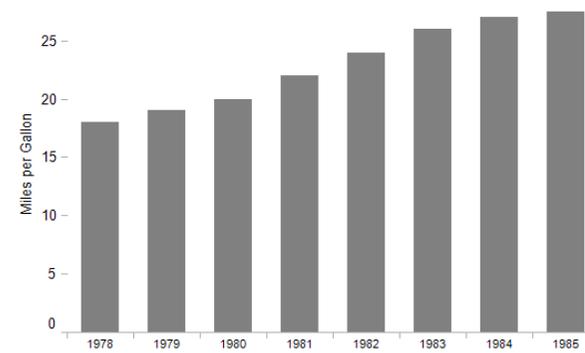
Data Effect = $\frac{27.5 - 18}{18} = 0.53$, Graph Effect = $\frac{5.3 - .6}{.6} = 7.83$,
 Lie Factor = 14.8 [Tufté, 1983]

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Fuel Economy Standard Redesign

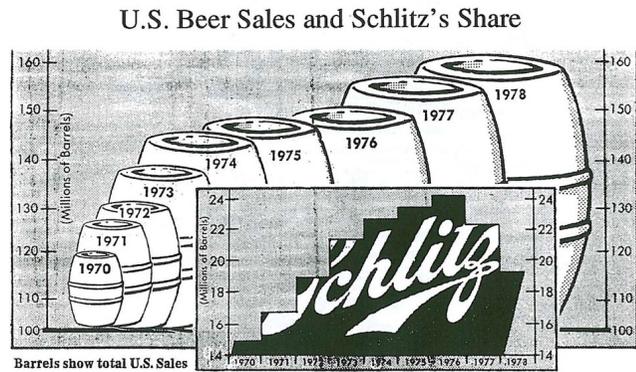


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Lie Factor

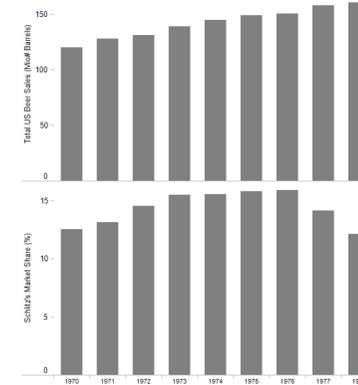


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Beer Sales Redesign

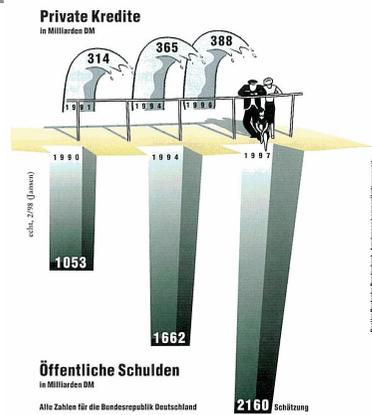


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Avoid Chartjunk



[Jansen & Scharfe, 1999]

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Tufte Design Principles

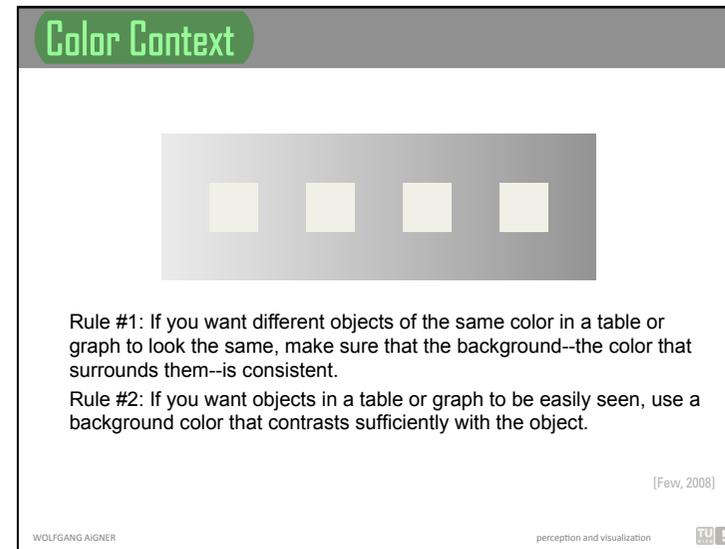
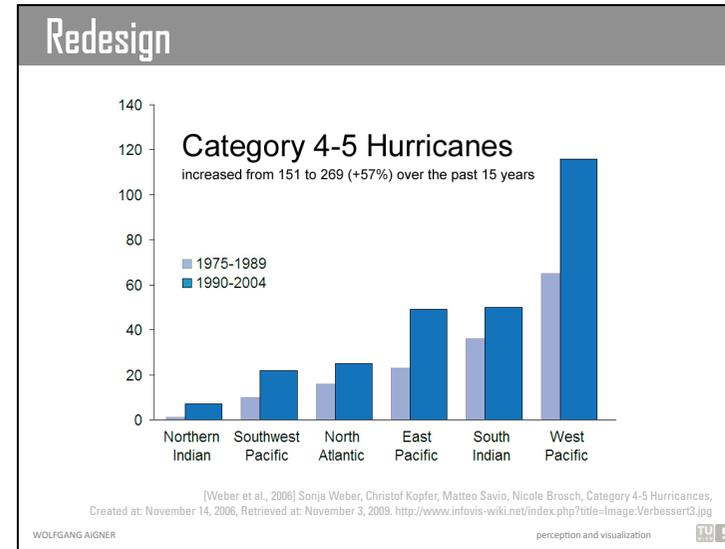
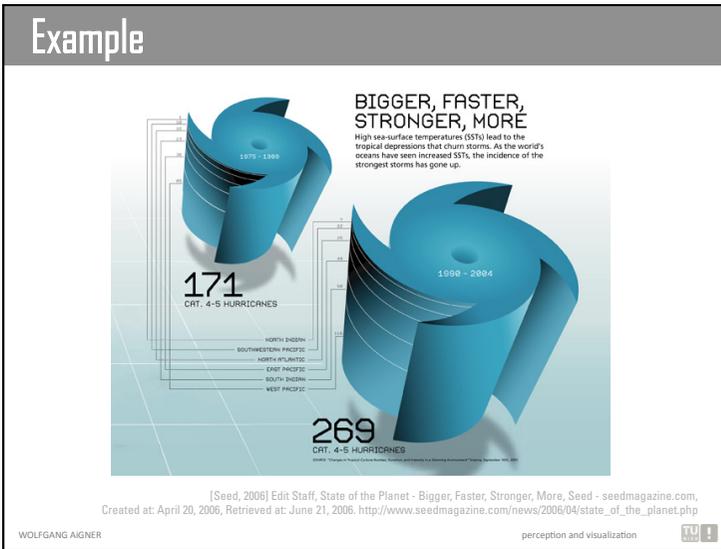
1. Above all else show the data.
2. Maximize the data-ink ratio.
3. Erase non-data-ink.
4. Erase redundant data-ink.
5. Revise and edit.

[Tufte, 1983]

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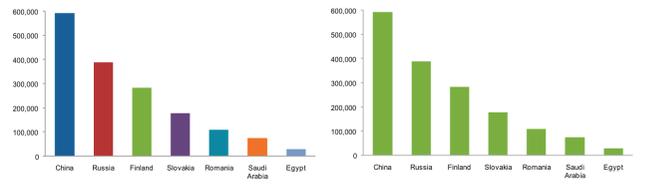
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Color Usage

- Rule #3: Use color only when needed to serve a particular communication goal.
- Rule #4: Use different colors only when they correspond to differences of meaning in the data.
 - To highlight particular data
 - To group items
 - To encode quantitative values



[Few, 2008]

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Color Usage

- Rule #5: Use soft, natural colors to display most information and bright and/or dark colors to highlight information that requires greater attention.



[Few, 2008]

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Palette Types

Categorical



Sequential



Diverging



- Rule #6: When using color to encode a sequential range of quantitative values, stick with a single hue (or a small set of closely related hues) and vary intensity from pale colors for low values to increasingly darker and brighter colors for high values.

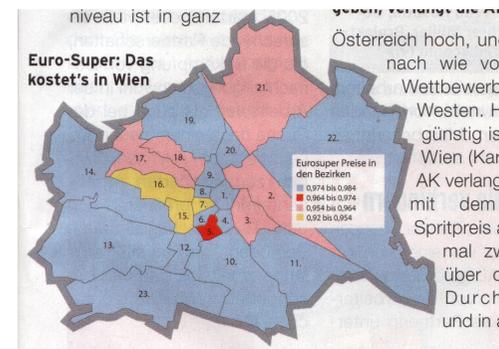
[Few, 2008]

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Example



Quelle: AK Für Sie, Mitgliederzeitschrift der AK Wien,

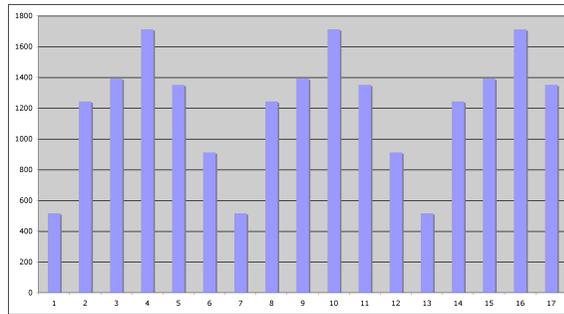
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De-emphasize non-data components

Rule #7: Non-data components of tables and graphs should be displayed just visibly enough to perform their role, but no more so, for excessive salience could cause them to distract attention from the data.



[Few, 2008]

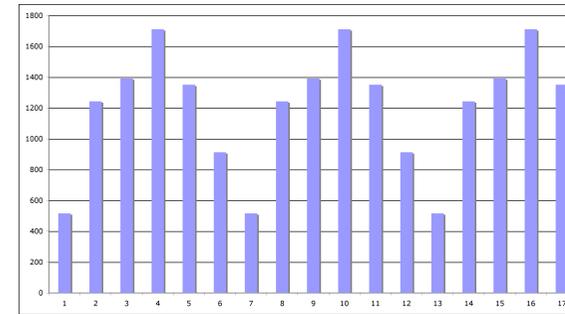
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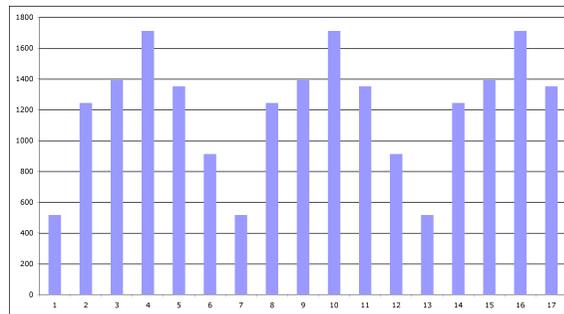
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[Few, 2008]

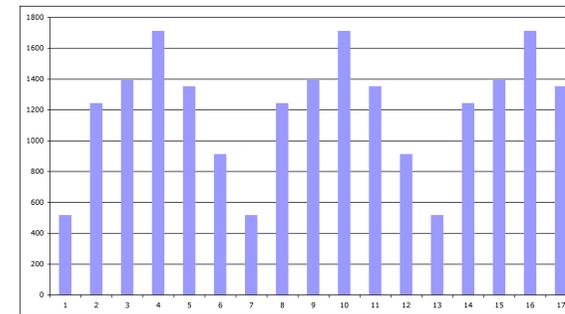
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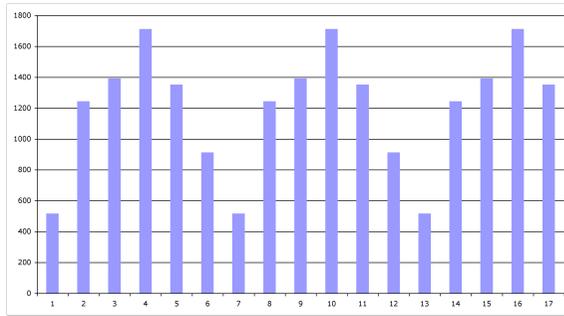
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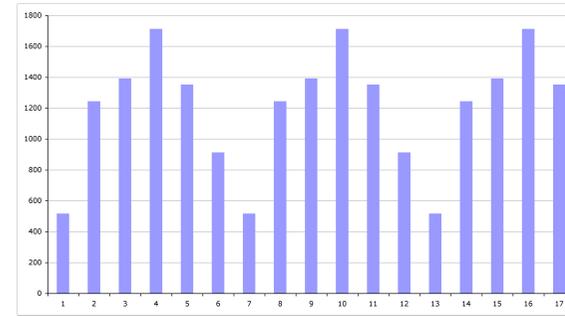
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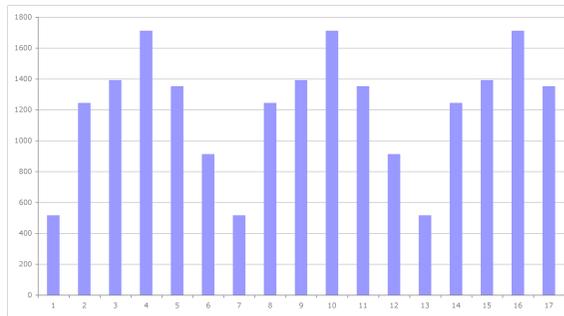
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[Few, 2008]

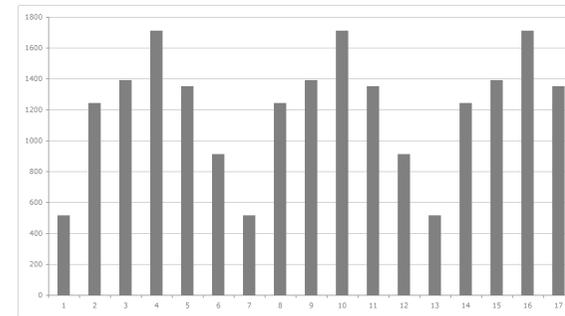
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[Few, 2008]

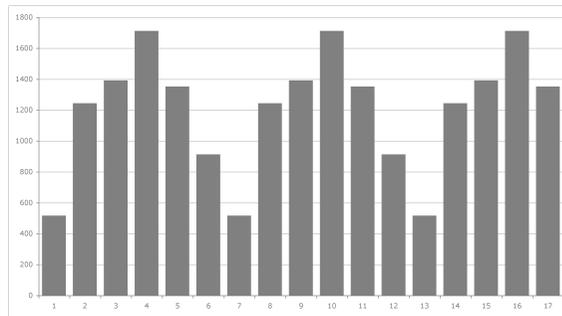
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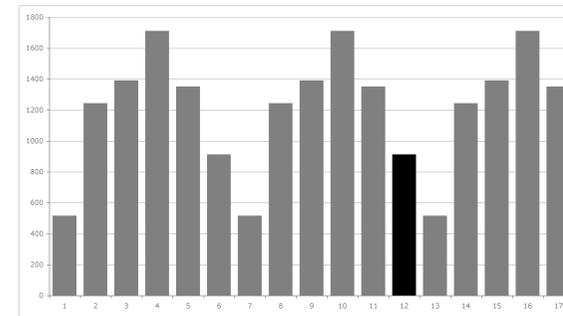
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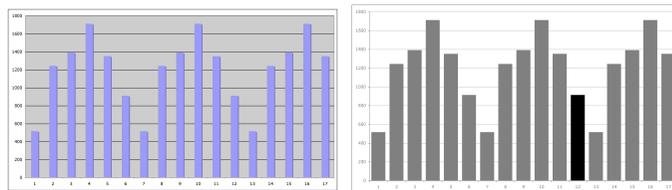
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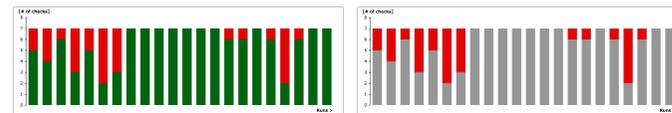
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Avoid Red-Green

Rule #8: To guarantee that most people who are colorblind can distinguish groups of data that are color coded, avoid using a combination of red and green in the same display.



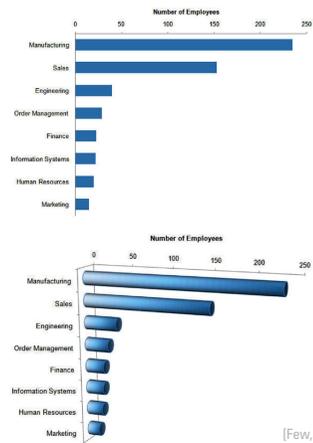
[Few, 2008]

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Rule #9: Avoid using visual effects in graphs.



[Few, 2008]

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Summary: InfoVis...

- ... is a very complex task
- ... can help to get insight into data more quickly
- ... requires preparation and sensible handling of the information
- ... should make use of the properties of human visual perception
- ... requires sensible handling, relative to the task
- ... is a big challenge, if you want to do it good

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»Die Umwelt, so wie wir sie wahrnehmen, ist unsere Erfindung.«

Heinz von Foerster

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