

Lecture 1, InfoVis MiniCourse

Perception, Frameworks, Color

LaBRI, University of Bordeaux

14 June 2004

Tamara Munzner

Information visualization

interactive visual representation of abstract data
· help human perform some task more effectively

external representation
· reduces load on working memory

2

External representation example

book topic relationships

· [Godel, Escher, Bach, Hofstadter 1979]

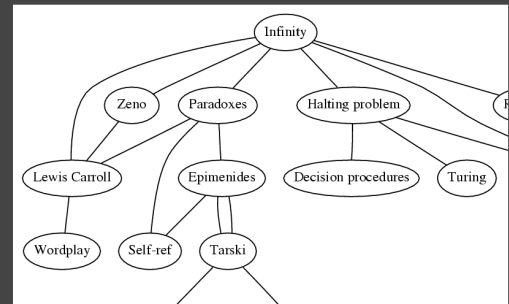
Paradoxes – Lewis Carroll
Turing – Halting problem
Halting problem – Infinity
Paradoxes – Infinity
Infinity – Lewis Carroll
Infinity – Unpredictably long searches
Infinity – Recursion
Infinity – Zeno
Infinity – Paradoxes
Lewis Carroll – Zeno
Lewis Carroll – Wordplay

Halting problem – Decision procedures
BlooP and FlooP – AI
Halting problem – Unpredictably long searches
BlooP and FlooP – Unpredictably long searches
BlooP and FlooP – Recursion
Tarski – Truth vs. provability
Tarski – Epimenides
Tarski – Undecidability
Paradoxes – Self-ref
[...]

3

External representation example

offload cognition to visual systems
read off answer



4

Mini-Course Outline

Perception

Frameworks

Color

Space/Order

Depth/Occlusion

High Dimensionality

Interaction

Navigation/Zooming

Focus+Context

Graphs/Trees

Scalability

Task-Centered Design

5

Human Perception

sensors/transducers

· psychophysics: determine characteristics

relative judgements: strong

absolute judgements: weak

· continuing theme

different optimizations than most machines

· eyes are not cameras

· perceptual dimensions not nD array

· (brains are not hard disks)

6

Psychophysical Measurement

JND: just noticeable difference

increment where human detects change

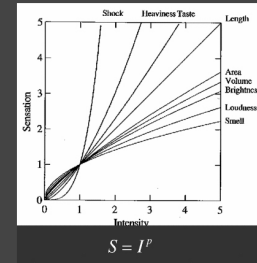
average to create "subjective" scale

7

Nonlinear Perception of Magnitudes

sensory modalities **not** equally discriminable

· Stevens power law

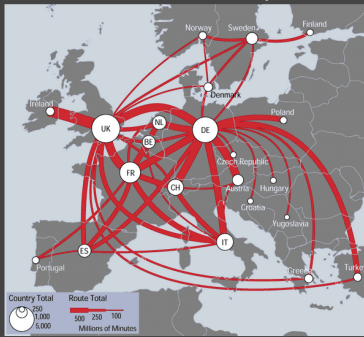


[Stevens, On the Theory of Scales of Measurement, Science 103:206-217, 1946]

8

Dimensional Dynamic Range

linewidth: limited discriminability



9

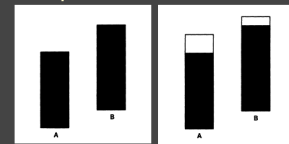
Weber's Law

ratio of increment threshold to background intensity is constant

· relative judgements within modality

$$\frac{\Delta I}{I} = K$$

Cleveland example: frame increases accuracy



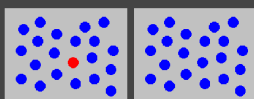
[Graphical Perception: Theory, Experimentation and the Application to the Development of Graphical Models]

10

Preattentive Visual Dimensions

color (hue) alone: preattentive

- attentional system not invoked
- search speed independent of distractor count



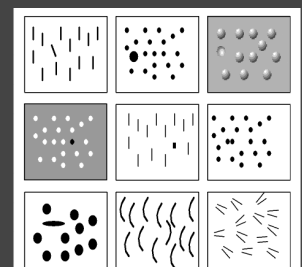
[Chris Healey, Preattentive Processing, www.csc.ncsu.edu/faculty/healey/PP/PP.html]

11

Preattentive Visual Dimensions

many preattentive dimensions of visual modality

- hue
- shape
- texture
- length
- width
- size
- orientation
- curvature
- intersection
- intensity
- flicker
- direction of motion
- stereoscopic depth
- lighting direction

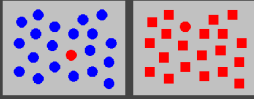


[Chris Healey, Preattentive Processing, www.csc.ncsu.edu/faculty/healey/PP/PP.html]

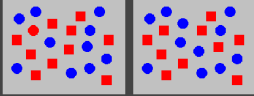
12

Preattentive Visual Dimensions

color alone: preattentive
 shape alone: preattentive



combined hue and shape: multimodal



- requires attention
- search speed linear with distractor count

13

Integral vs. Separable Dimensions



red-green
yellow-blue

x-size
y-size

size
orientation

color
shape

color
motion

color
location

[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]₁₄

Gestalt Laws

principles of pattern perception

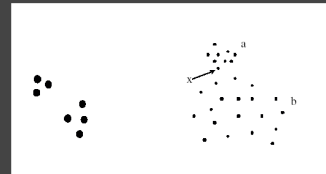
- "gestalt": German for "pattern"
- original proposed mechanisms wrong
- rules themselves still useful
- "Pragnatz": simplest possibility wins

principles

- proximity
- similarity
- continuity/connectedness/good continuation
- closure
- symmetry
- common fate (things moving together)
- relative sizes

15

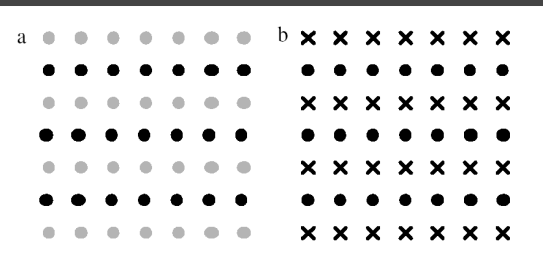
Proximity



[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

16

Similarity

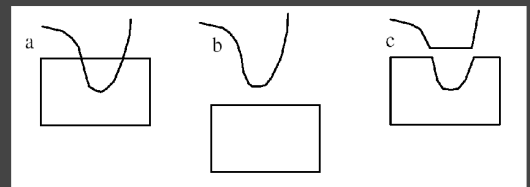


[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

17

Continuity

smooth not abrupt change
 overrules proximity

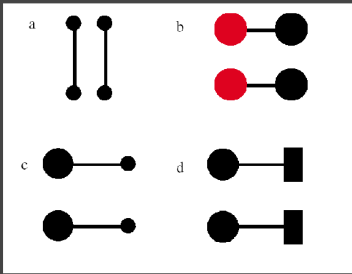


[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

18

Connectedness

can overrule size, shape

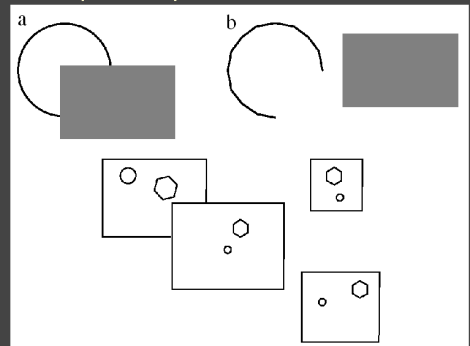


[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

19

Closure

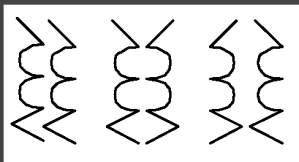
overrules proximity



20

Symmetry

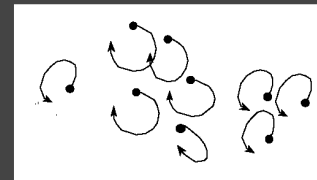
emphasizes relationships



[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

21

Common Fate

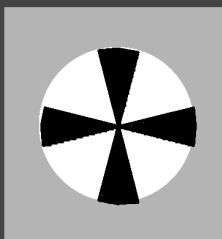


[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

22

Relative Size

smaller components perceived as objects



[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999.]

23

Graph Drawing Tension

node placement

close

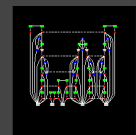
· proximity

far

· visual popout of long edge

either

· connectedness



[www.research.att.com/sw/tools/graphviz]

tradeoffs abound in infovis!

24

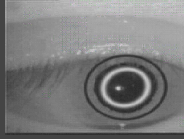
Eyes

fovea

- thumbnail at arm's length
- small high resolution area

saccades [video]

- high-resolution samples
- brain makes collage
- vision perceived as entire simultaneous field
- fixation points: dwell 200–600ms
- moving: 20–100ms



[vision.arc.nasa.gov/personnel/jbm/home/projects/osa98/osa98.html/]

25

Ears

perceived as temporal stream

- but also samples over time
 - hard to filter out when not important
- visual vs auditory attention

implications

- harder to create overview?
- hard to use as separable dimension?

'sonification' still very niche area

- alternative: supporting sound enhances immersion

26

More Reading

Information Visualization: Perception for Design. Colin Ware. Morgan Kaufmann 1999.
Chapter 5: Visual Attention and Information That Pops Out

Information Visualization: Perception for Design. Colin Ware. Morgan Kaufmann 1999.
Chapter 6: Static and Moving Patterns

The Psychophysics of Sensory Function, S. S. Stevens, Sensory Communication, MIT Press, 1961, pp 1–33.
<http://www.cs.ubc.ca/~tmm/courses/cpsc533c-03-spr/readings/ss.pdf>

Graphical Perception: Theory, Experimentation and the Application to the Development of Graphical Models William S. Cleveland, Robert McGill, J. Am. Stat. Assoc. 79:387, pp. 531–554, 1984.

<http://www.jstor.org/cgi-bin/jstor/printpage/01621459/di985961/98p1201a/0.pdf?userID=8e670917@ubc.ca/01cc99333c0050103e7f6backcontext=citation&config=jstor&dowhat=Acrobat60.pdf>

Perception in Visualization, Christopher G. Healey
<http://www.csc.ncsu.edu/faculty/healey/PP/index.html>

27

Mini-Course Outline

Perception

Frameworks

Color

Space/Order

Depth/Occlusion

High Dimensionality

Interaction

Navigation/Zooming

Focus+Context

Graphs/Trees

Scalability

Task-Centered Design

28

Data Types

categorical (nominal)

- apples, oranges, bananas



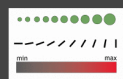
ordered (ordinal)

- small, medium, large
- days: Sun, Mon, Tue, Wed, Thu, Fri, Sat



continuous (quantitative)

- 10 inches, 17 inches, 23 inches



[graphics.stanford.edu/papers/polzig3]

29

Mackinlay, Card Framework

Data Types

- nominal, ordered, quantitative

Marks

- point, line, area, surface, volume
- geometric primitives

Retinal Properties / Perceptual Dimensions

- size, brightness, color, texture, orientation, shape,...
- parameters that control the appearance of geometric primitives
- separable channels of information flowing from retina to brain

Data Variables

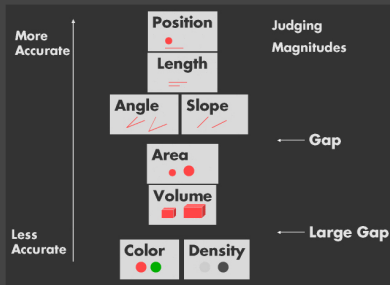
- 1D, 2D, 3D, 4D, 5D, etc

Bertin; Wilkinson; Stolte et al

- closest thing to central dogma we've got

30

Ranking Perceptual Dimensions

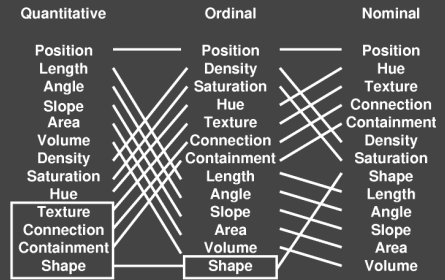


[graphics.stanford.edu/courses/cs448b-02-spring/lectures/encoding/walk015.html]

31

Ranking Varies by Data Type

spatial position best for all types



[Mackinlay, Automating the Design of Graphical Presentations of Relational Information, ACM TOG 5:2, 1986]

32

More Reading

Chapter 1, Readings in Information Visualization: Using Vision to Think. Stuart Card, Jock Mackinlay, and Ben Shneiderman, Morgan Kaufmann 1999.

The Eyes Have It: A Task by Data Type Taxonomy for Information Visualizations. Ben Shneiderman, Proc. 1996 IEEE Visual Languages, also Maryland HCIL TR 96-13 <ftp://ftp.cs.umd.edu/pub/hcil/Reports-Abstracts-Bibliography/96-13html/96-13.html>

The Structure of the Information Visualization Design Space. Stuart Card and Jock Mackinlay, Proc. InfoVis 97 <http://citeseer.nj.nec.com/card96structure.html>

The Grammar of Graphics, Leland Wilkinson, Springer 1999

Semiology of Graphics: Diagrams, Networks, Maps. Jaques Bertin. University of Wisconsin Press, Madison (WI), 1983. W. J. Berg (Translator).

33

Mini-Course Outline

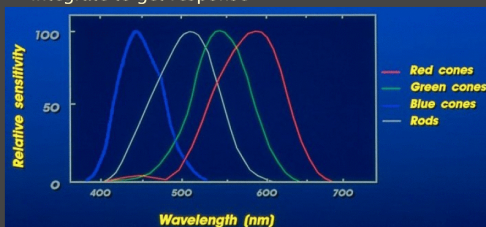
Perception
 Frameworks
 Color
 Space/Order
 Depth/Occlusion
 High Dimensionality
 Interaction
 Navigation/Zooming
 Focus+Context
 Graphs/Trees
 Scalability
 Task-Centered Design

34

Trichromacy

cone response is a function of wavelength for a given spectrum

- multiple by response curve
- integrate to get response



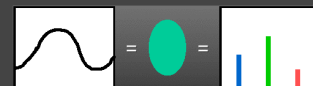
[Stone, SIGGRAPH 2001 course notes, graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

35

Metamerism

brain sees only cone response

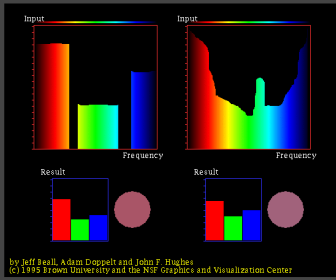
different spectra appear the same



[Stone, SIGGRAPH 2001 course notes, graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

36

Metamerism



by Jeff Beall, Adam Doppelt and John F. Hughes
© 1999 Brown University and the NSF Graphics and Visualization Center

[www.cs.brown.edu/exploratories/freeSoftware/repository/edu/brown/cs/exploratories/applets/spectrum/metamers_java_browser.html] 37

Color Rules of Thumb

nominal

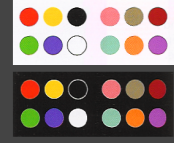
- bad: > 12 hues
- good: use <= ~12 hues

ordinal

- bad: using hue
- good: saturation/brightness

quantitative

- bad: rainbow colormaps
- good: interpolate between two hues

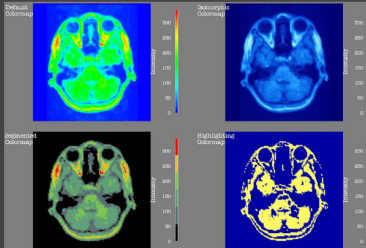


[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999, Figure 4.21]

Colormaps

rainbow colormaps usually bad idea

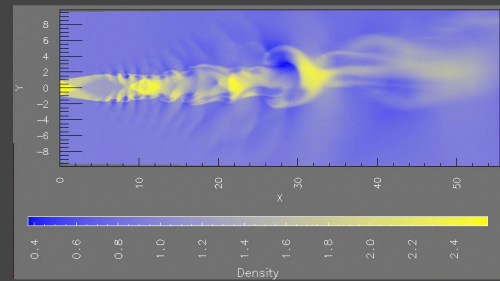
- hue is mediocre for showing order
- not perceptually linear!



[Rogowitz and Treinish, How NOT to Lie with Visualization, www.research.ibm.com/dx/proceedings/pravda/truevis.htm]

Colormaps

interpolating between two hues usually safe



[Rogowitz and Treinish, How NOT to Lie with Visualization, www.research.ibm.com/dx/proceedings/pravda/truevis.htm]

Colormaps, Tufte

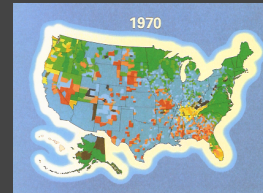


[Tufte, Envisioning Information, p. 91]

Color In Large Areas

Ware and Tufte agree: desaturate!

"excessively exuberant"



pastels for text bg

```
import java.applet.Applet;
import java.awt.Graphics;
import java.awt.Color;

public class ColorText extends Applet
{
    public void init ()
    {
        red = 100;
        green = 225;
        blue = 20;
    }

    public void paint (Graphics g)
    {
        Color color (new Color (red, green, blue));
        Graphics g.setColor (color);
    }

    private int red;
    private int green;
    private int blue;
}
```

[Edward Tufte, Envisioning Information, p.82]

[Colin Ware, Information Visualization: Perception for Design, Morgan Kaufmann 1999, Figure 4.20]

Color Deficiency

deutanope
protanope

- has red/green deficit
- 10% of males!

tritanope

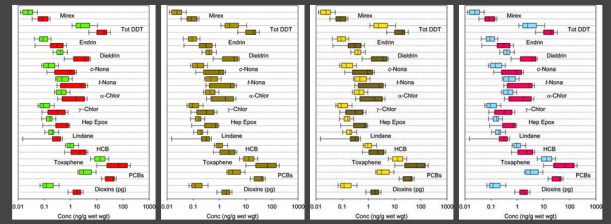
- has yellow/blue deficit

<http://www.vischeck.com/vischeck>
· test your images

43

Color Deficiency Examples: vischeck

original deutanope protanope tritanope



[www.cs.ubc.ca/~tmm/courses/cpsc533c-04-spr/a1/dmitry/533a1.html,
citing Global Assessment of Organic Contaminants in Farmed Salmon,
Ronald A. Hites, Jeffery A. Foran, David O. Carpenter, M. Coreen
Hamilton, Barbara A. Knuth, and Steven J. Schwager, Science 2004 303: 226-229.]

44

Designing Around Deficiencies

red/yellow/green could have domain meaning
then distinguish by more than hue alone
· saturation, brightness

original deutanope protanope tritanope

Qty	Limit	Class	Status	Ex Qty	Qty	Limit	Class	Status	Ex Qty	Qty	Limit	Class	Status	Ex Qty	Qty	Limit	Class	Status	Ex Qty
20,000	99.95			10,000	20,000	99.95			10,000	20,000	99.95			10,000	20,000	99.95			10,000
80,000	MKT			10,000	80,000	MKT			10,000	80,000	MKT			10,000	80,000	MKT			10,000
200,000	3P	Car/Trid		80,000	200,000	3P	Car/Trid		80,000	200,000	3P	Car/Trid		80,000	200,000	3P	Car/Trid		80,000
20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000
20,000	99.95	Port		10,000	20,000	99.95	Port		10,000	20,000	99.95	Port		10,000	20,000	99.95	Port		10,000
20,000	99.95	Joe O	Car/Trid	10,000	20,000	99.95	Joe O	Car/Trid	10,000	20,000	99.95	Joe O	Car/Trid	10,000	20,000	99.95	Joe O	Car/Trid	10,000
20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000	20,000	99.95	DOT		10,000
20,000	99.95	Port	Car/Trid	10,000	20,000	99.95	Port	Car/Trid	10,000	20,000	99.95	Port	Car/Trid	10,000	20,000	99.95	Port	Car/Trid	10,000
20,000	99.95	Joe O		10,000	20,000	99.95	Joe O		10,000	20,000	99.95	Joe O		10,000	20,000	99.95	Joe O		10,000
80,000	99.95	DOT		10,000	80,000	99.95	DOT		10,000	80,000	99.95	DOT		10,000	80,000	99.95	DOT		10,000
200,000	MKT			200,000	200,000	MKT			200,000	200,000	MKT			200,000	200,000	MKT			200,000
20,000	MKT	Joe O		25,000	20,000	MKT	Joe O		25,000	20,000	MKT	Joe O		25,000	20,000	MKT	Joe O		25,000

[Courtesy of Brad Paley]

45

Color/Brightness Constancy

relative judgements

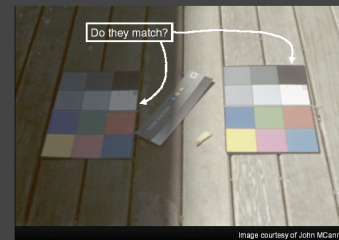


Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

46

Color Constancy

relative judgements

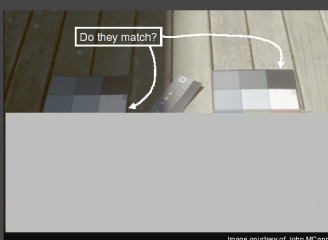


Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

47

Color Constancy

relative judgements



Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

48

Color Constancy

relative judgements

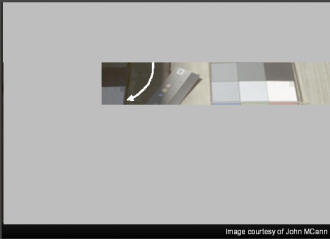


Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

49

Color Constancy

relative judgements

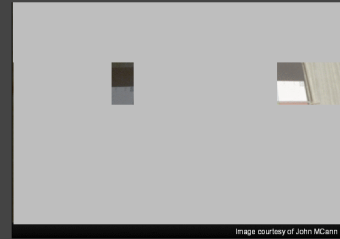


Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

50

Color Constancy

relative judgements

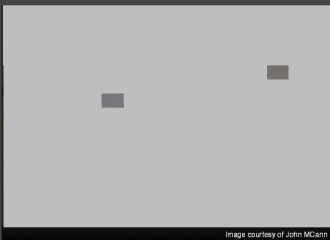


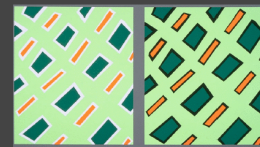
Image courtesy of John McCann

[courtesy of John McCann, from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

51

Context Matters

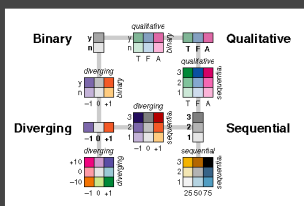
Bezold effect: outlines



[from Stone 2001 SIGGRAPH course
graphics.stanford.edu/courses/cs448b-02-spring/04cdrom.pdf]

52

Cartographic Color Advice



[Brewer, www.personal.psu.edu/faculty/c/a/cab38/ColorSch/Schemes.html]

53

More Reading

Information Visualization: Perception for Design. Colin Ware. Morgan Kaufmann 1999. Chapter 3: Lightness, Brightness, Contrast, and Constancy

Information Visualization: Perception for Design. Colin Ware. Morgan Kaufmann 1999. Chapter 4: Color

Envisioning Information. Edward Tufte. Graphics Press, 1990. Chapter 5: Color and Information

How Not to Lie with Visualization. Bernice E. Rogowitz and Lloyd A. Treinish, Computers In Physics 10(3) May/June 1996, pp 268-273.
<http://www.research.ibm.com/dx/proceedings/pravda/truevis.htm>

Color use guidelines for data representation. C. Brewer, 1999.
<http://www.personal.psu.edu/faculty/c/a/cab38/ColorSch/ASApaper.html>

A Field Guide To Digital Color, Maureen Stone, AK Peters 2003