Interactive Information Visualization

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UBC CPSC 414 Week 11, Wed 12 November 2003

Outline

information visualization motivation

designing for humans

information visualization techniques

future directions

Information visualization

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

Interactivity

static images

10,000 years art, graphic design

moving images cinematography

interactive graphics

20 years

computer graphics, human-computer interaction

Information visualization

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

external representation

reduces load on working memory

- bridging many fields
 graphics: interacting in realtime
 cognitive psych: finding appropriate representation
 HCI: using task to guide design and evaluation

Visualization Tasks

overview zoom filter details-on-demand

relate history

[The Eyes Have It: A Task by Data Type Taxonomy for Information Visualizations. Ben Shneiderman citeseer.nj.nec.com/shneiderman96eyes.html]

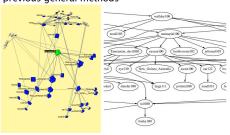
Task-oriented design

custom design for checking semantic networks reading definition subgraph labels



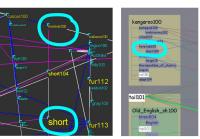
Task-oriented design

previous general methods



Design tradeoffs

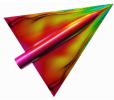
information density vs. visual salience



Scientific vs. information visualization

scivis: inherently spatial data fluid flow over airplane wing

infovis: abstract data, choice of spatialization FilmFinder

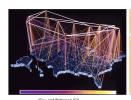




Example: node-link graphs

powerful abstraction

common in many domains





Why visualize graphs?

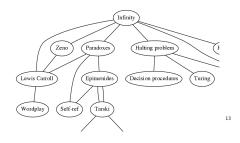
Example: book topic relationships [Godel, Escher, Bach. Hofstadter 1979]

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 12

Why visualize graphs?

offload cognition to visual systems minimal attention to read answer



Why draw graphs automatically?





automatic: seconds

[Godel, Escher, Bach. Hofstader 79]

dot, [Gansner et al 93]

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Human Perception

sensors/transducers psychophysics: determine characteristics

relative judgements: strong absolute judgements: weak

different optimizations than most machines eyes are not cameras perceptual dimensions not nD array (brains are not hard disks)

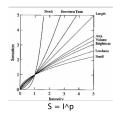
limits of intuition

thoughts, goals, plans: accurate vision, hearing, attention, memory: inaccurate

15

Nonlinear perception of magnitudes

sensory dimensions not equally discriminable JND: Just Noticeable Differences Stevens power law



Eyes

foveal vision

- high resolution
- thumbnail at arm's length

saccades [video]

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



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

Fovea

low-res periphery, high-res sensor general concept, not just for eyes foveal touch!: star-nosed mole

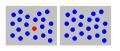




[www.nature.com/nsu/010329/010329-6.html brain.nips.ac.jp/event/work131030/Catania_and_Kaas,_1997.pdf]

Preattentive visual dimensions

color (hue) alone: preattentive attentional system not invoked search speed independent of distractor count

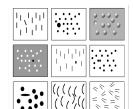


19

Preattentive visual dimensions

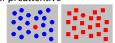
many preattentive dimensions of visual modality

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

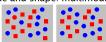


Preattentive visual dimensions

color alone: preattentive shape alone: preattentive



combined hue and shape: multimodal



- requires attention
- search speed linear with distractor count

20

Well, actually...

sometimes works (motion + color)

but need both preattentive and cognitive for, say, designing visualizations

Integral vs. separable dimensions



red-green yellow-blue

x-size y-size

size orientation

color shape

color motion

color location

Data types

continuous (quantitative)
10 inches, 17 inches, 23 inches

ordered (ordinal) small, medium, large

categorical (nominal) apples, oranges, bananas



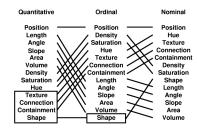


[graphics.stanford.edu/papers/polaris]

25

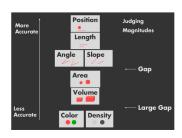
Dimensional ranking varies by data type

spatial position best for all types



architecture Automation the Davison of Countries Documentations of Relational Information, ACM TOC 5-2, 1986

Dimensional ranking



graphics.stanford.edu/courses/cs448b-02-spring/lectures/encoding/walk015.htm

27

31

Dimensional dynamic range

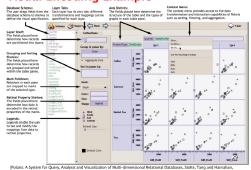
linewidth: limited discriminability



[mappa.mundi.net/maps/maps_014/telegeography.html]

28

Visual Encoding Example



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

30

32

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Color rules of thumb

nominal

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

.

ordinal

bad: using hue good: saturation/brightness

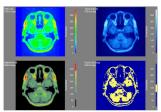


quantitative

bad: rainbow colormaps good: interpolate between two hues

Colormaps

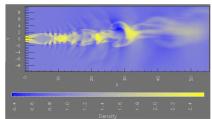
rainbow colormaps usually bad idea hue is mediocre for showing order not perceptually linear!



ogowitz and Treinish, How NOT to Lie with Visualization, w.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

Overview+detail

problem

avoid user disorientation when inspecting detail hard for big datasets

bad: one window, must remember position

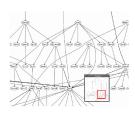




33

Overview and detail

better: add linked overview window(s)



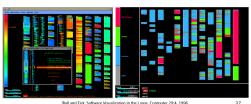
how to create overview?

Overview and detail

SeeSoft: software maintenance (colormaps: segmented vs. continuous)

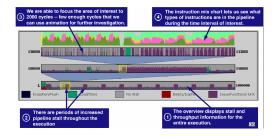






Overview+detail

Rivet: performance tuning level of detail



Overview to detail to sorting



Focus+context

linked windows still have cognitive load to correlate good solution: merge overview, detail into single window

fisheye views [Furnas 86], [Sarkar et al 94]



39

Focus+context

linked windows still have cognitive load to correlate good solution:

merge overview, detail into single window

fisheye views [Furnas 86], [Sarkar et al 94] nonlinear magnification [Keahey 96]



41

TableLens

focus+context power of sorting

www.tablelens.com

42

Focus+context

H3 [Munzner 97]

task: browsing large quasi-hierarchical graphs [demo]

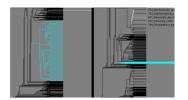


[Munzner 1997, 1998a, 1998b]

Global focus+context

TreeJuxtaposer: comparing trees

linked highlighting [demo]



44

Comparison

bad: temporal, if many items

intermediate ones "overload mental buffer" good: temporal blinking if two items

good: side by side array of small multiples creates overview



Minimizing occlusion

bad: Midwestern occlusion



[citeseer.nj.nec.com/becker95visualizing.html] [Becker, Eick, and Wilks. Visualizing Network Data, IEEE TVCG 1995]

46

Minimizing occlusion

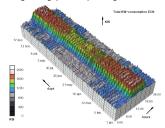
good: show only start and end of lines



[citeseer.nj.nec.com/becker95visualizing.html]
[Becker, Eick, and Wilks, Visualizing Network Data, IEEE TVCG 1995

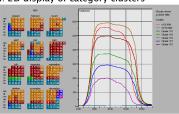
Minimizing occlusion: 3D vs. 2D

bad: 3D pretty but not useful metacognitive gap: lose by adding dimension



Minimizing occlusion: 3D vs. 2D

good: 2D display of category clusters



Motion: clarify structure

navigation

rotate/translate/zoom [demo: Geomview]



object recognition

moving lights at joints Johannson 1973

animated transitions

avoid change blindness jump increases cognitive load smooth transition from one state to next maintain object constancy

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51

Future: scaling to huge datasets

data explosion

sensors

Human Genome Project Sloan Digital Sky Survey

simulation

Accelerated Strategic Computing Initiative microprocessor design

long-distance telephony backbone Web traffic

52

Future: scaling display resolution

always pixel-bound in past

high-res displays now available 4K x 2K: 9Mpixels vs 1 Mpixel pixel rich

interactivity + resolution of paper add physical navigation (walk closer) to virtual navigation

More Information

UBC Term 2 course: 533C Visualization undergrads by consent of instructor

http://www.cs.ubc.ca/~tmm