

Perception & Information Visualization

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CS533C Topic Presentation
November 25, 2009

- Face Perception & Colour

- Kindlmann, G. , Reinhard, E. , & Creem, S. (2002). Face-based Luminance Matching for Perceptual Colormap Generation. *Proc. Vis 2002* .

- Motion

- Huber, D. E. , & Healey, C. G. (2005). Visualizing Data with Motion. *Proc. IEEE Visualization 2005* , pp. 527-534.

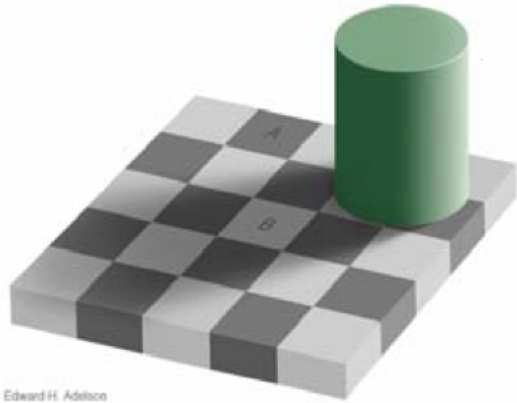
- Haptic Perception

- Yannier, N. , Basdogan, C. , Tasiran, S. , & Sen, O. L. (2008). Using Haptics to Convey Cause and Effect Relations in Climate Visualization. *IEEE Transactions on Haptics 1 (2)* , pp. 130-141.

Face-based Luminance Matching for Perceptual Colormap Generation

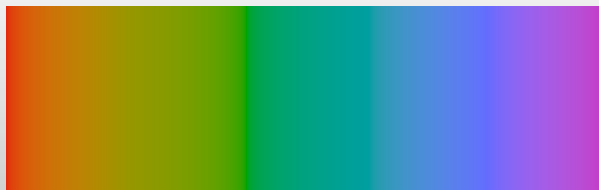
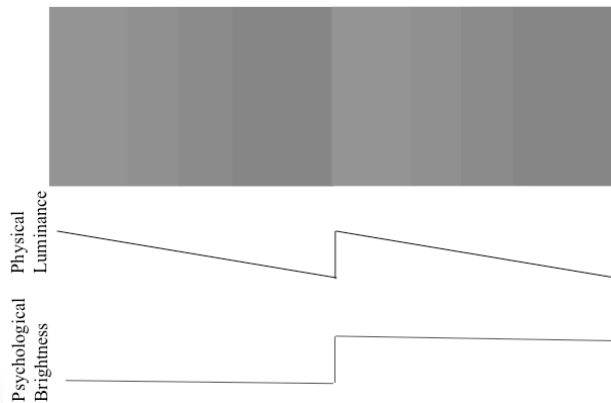
Kindlmann, G. , Reinhard, E. , &
Creem, S. (2002). *Proc. Vis 2002* .

Adelson 'illusion'



Edward H. Adelson

Craik-O'Brien-Cornsweet Illusion



- Problem

- creating isoluminant colormaps
- unknown display / room conditions

- Luminance

- intensity of light / unit area reaching eye
- photometric quantity
- display of image structure, surface shape

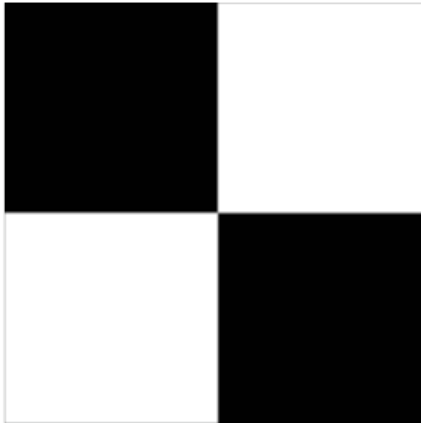
- Solution

- novel luminance matching technique
- use of ability to detect faces

- Background



- luminance measuring techniques based on matching paradigm



- for InfoVis, a predetermined pattern of luminance variation is often desired

- Current Practice / Previous Work

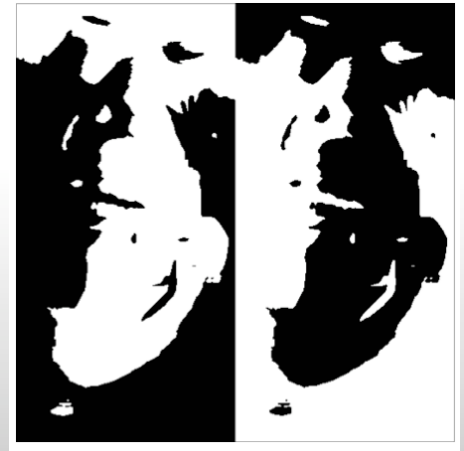
- minimally distinct border (MDB)
- challenging with different chromaticities

- threshold face images



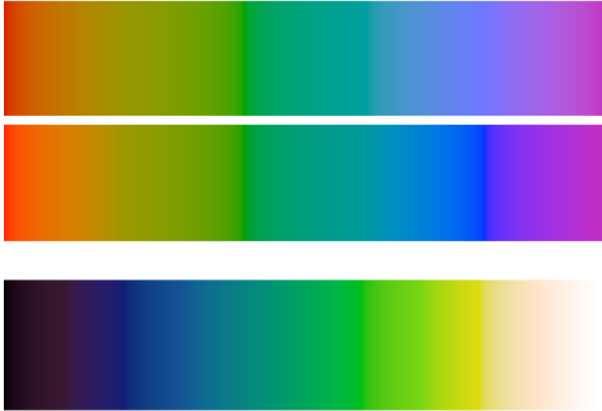


- Method
 - replace threshold image colours: black with a shade of grey, white with a colour
 - one face appears positive
- User Study
 - compare technique with adapted MDB
 - preserve border length
 - task
 - adjust HLS lightness
 - find cross-over point



- Results

- same accuracy and RT as MDB
- better precision than MDB

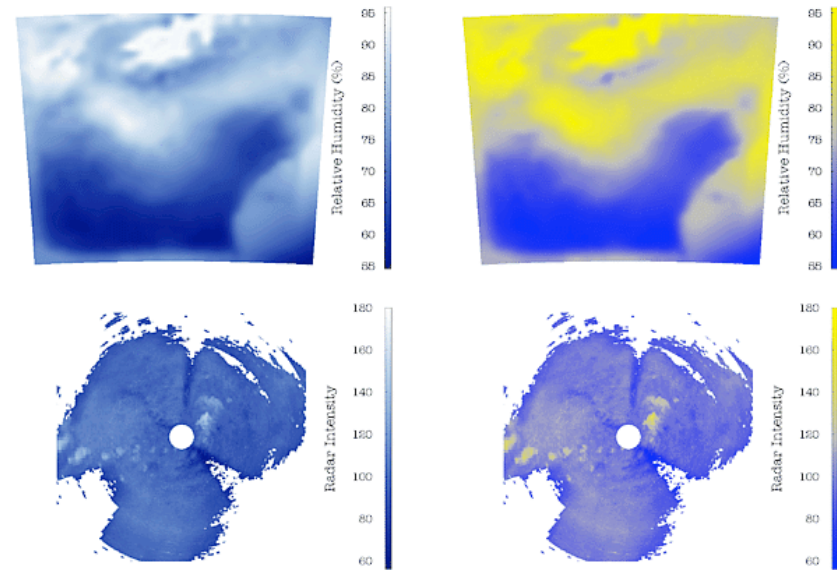


- Colour map generation

- user study: 1st step in creating isoluminant colormap
 - avg. control pts. across participants
 - interpolate colormap values in RGB space (with γ estimate)
-
- can also generate colormaps with monotonically increasing luminance

- Critique

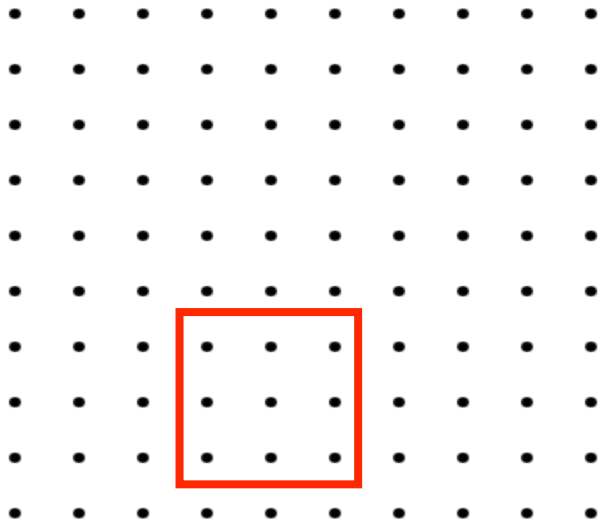
- creating colormaps with monotonically increasing luminance not evaluated
- spatial frequency of data not considered



How Not to Lie with Visualization, Bernice E. Rogowitz and Lloyd A. Treinish, *Computers In Physics* 10(3) May/June 1996 , pp 268-273.

Visualizing Data with Motion

Huber, D. E. , & Healey, C. G.
(2005). *Proc. IEEE Visualization 2005* ,
pp. 527-534.



- Problem
 - guidelines for use of motion as visualization cues with MD data
- Experiments
 - dot array stimuli to evaluate motion cues:
 - flicker
 - direction
 - velocity

- Flicker experiment

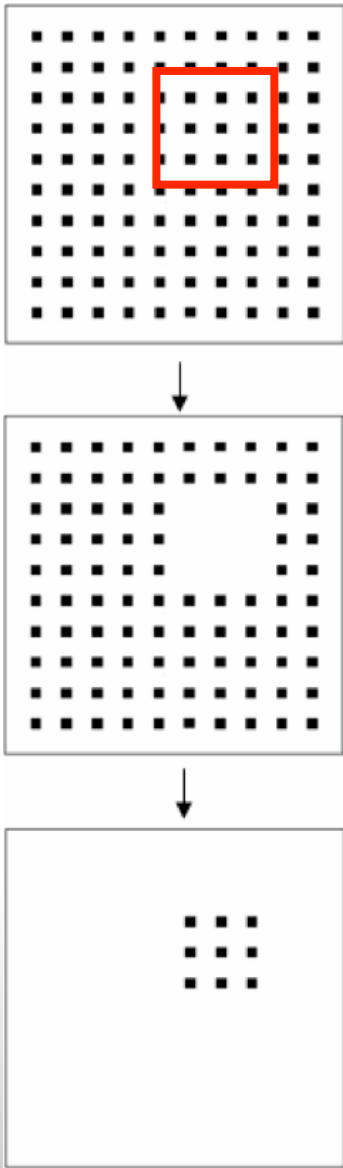
- target elements flicker at different rate than background elements

- evaluated:

- cycle length f_t
- cycle difference ΔF
- coherency

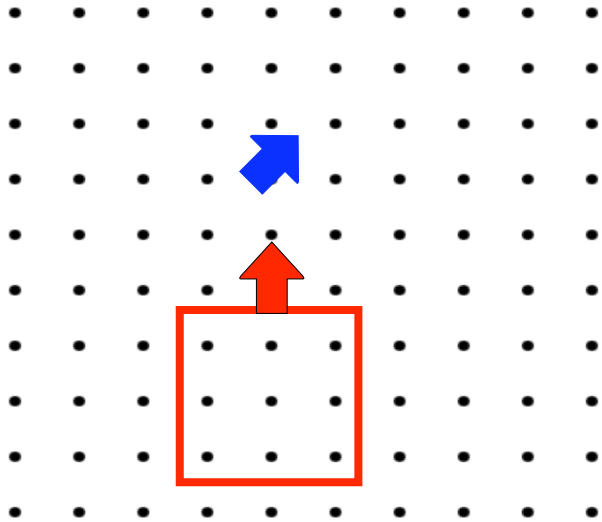
- results (based on error rates, RT):

- non-coherent error rates at chance
- coherent trials: ΔF of 120 ms easy to detect



- Direction experiment

- target elements move in different direction than background elements



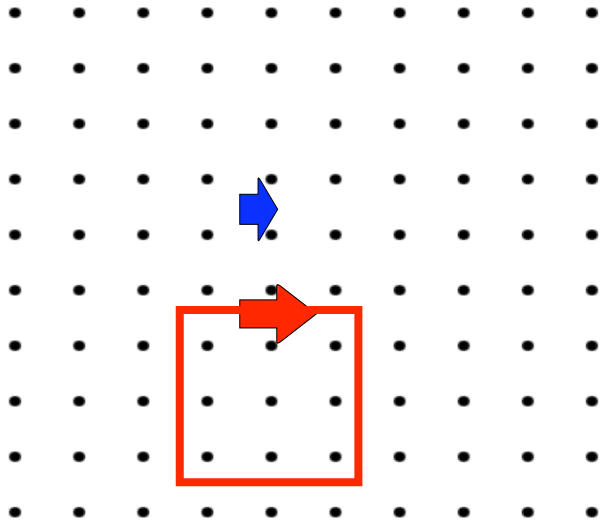
- evaluated:

- absolute target motion direction d_t
- direction difference ΔD

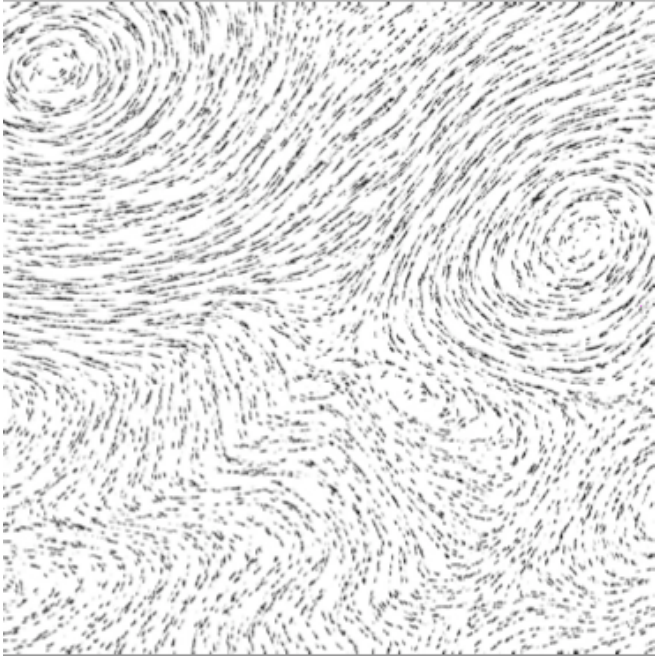
- results (based on error rates, RT):

- d_t doesn't matter
- ΔD more than 20 degrees easy to detect

- Velocity experiment
 - target elements move at different velocity than background elements



- evaluated:
 - absolute target velocity v_t
 - velocity difference ΔV
- results (based on error rates, RT):
 - v_t doesn't matter
 - ΔV more than 10px/s easy to detect (0.43 degrees)



- Implications + Applications
- flow visualizations
 - highlight changes in a data set over time or space
 - temperature and pressure gradients in meteorological datasets

- Critique

- interaction of motion cues not evaluated
 - possible interaction with non-motion cues
- representative behaviour of real-world data?
 - grid layout of stimuli appropriate?
- increased cognitive load for processing motion

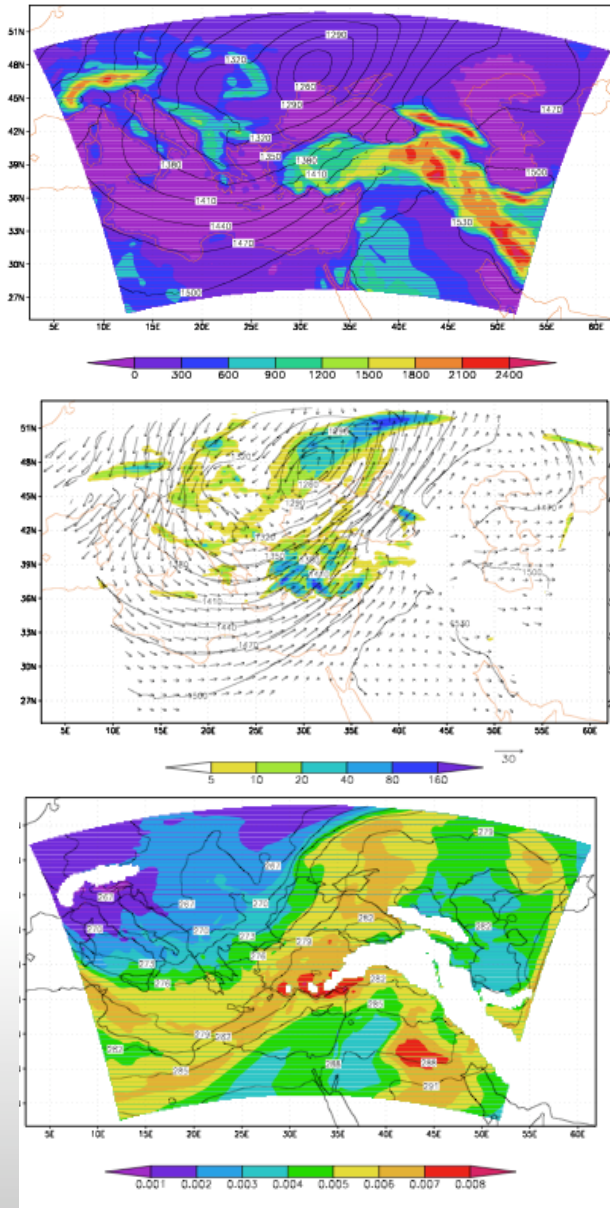
Using Haptics to Convey Cause and Effect Relations in Climate Visualization

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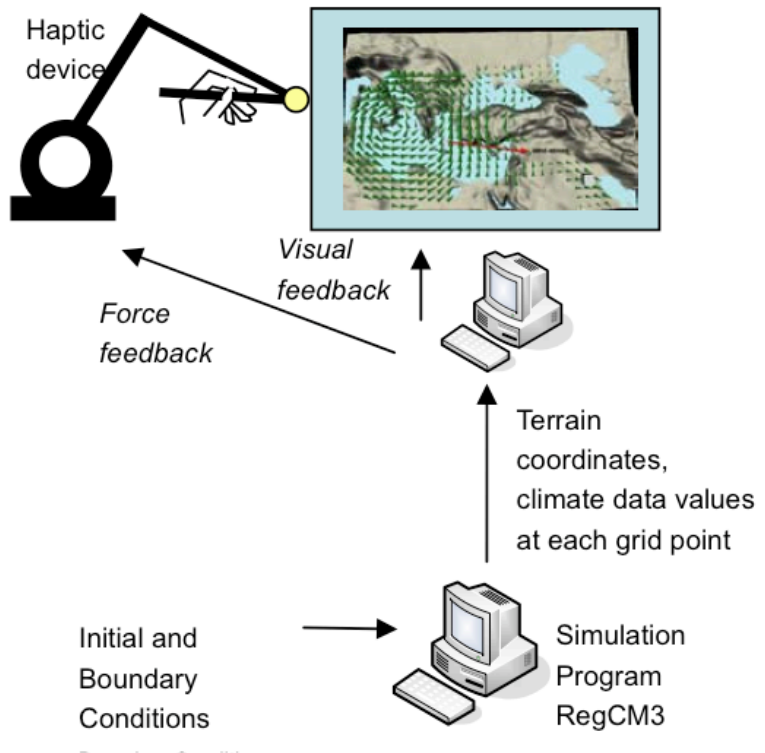
- Background & Problem

- climate data highly multidimensional
- multiple contour plots used to interpret data
- difficult for learners / non-experts
- typical displays do not target exploration, discovery, understanding cause + effect patterns



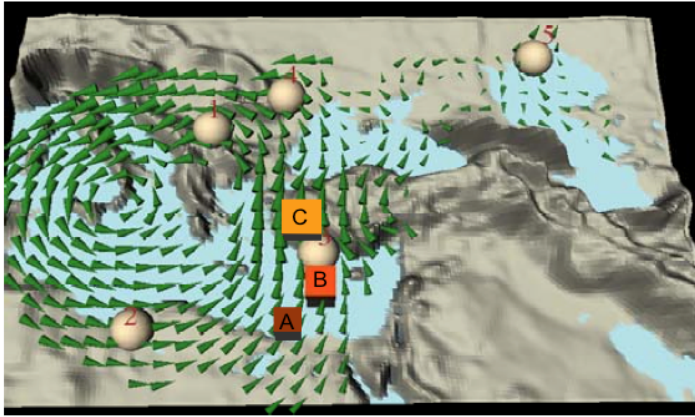
- Proposed Solution

- CEVIZ haptic + visual interface for interactive exploration of scalar, vector, tensor fields



- Hypotheses

- effective guided exploration of C+E patterns and relationships
- reduced load on visual system
- guide/confine exploration to interesting phenomena



less humid

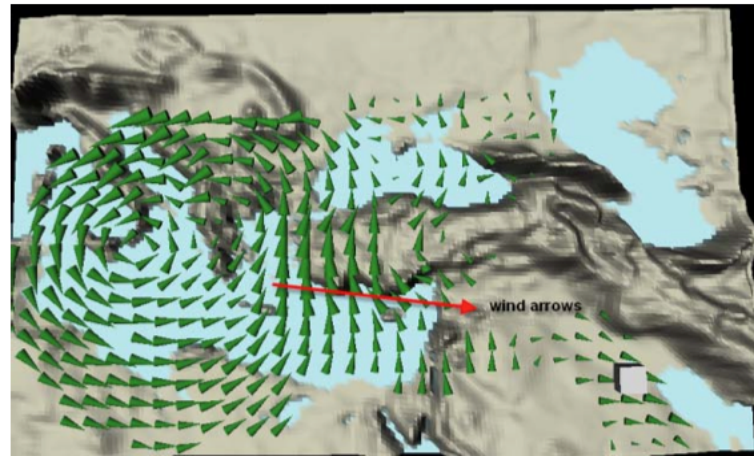
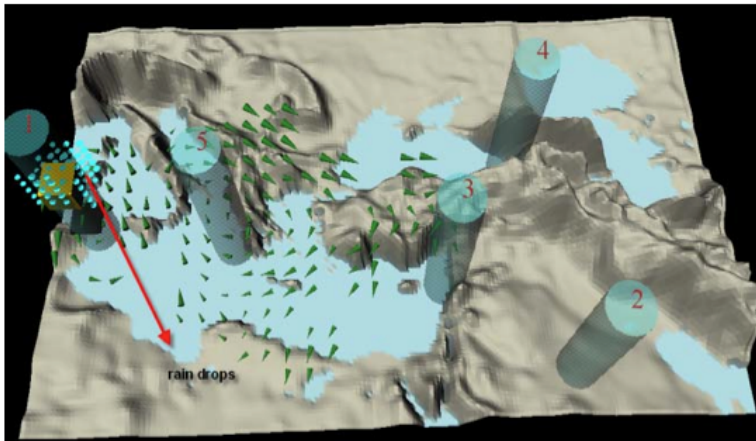
more humid



cooler

warmer

- Visual + Haptic Data Representation:
 - 3D mesh topography (V)
 - pressure surface (H)
 - wind (V + H: confine to path)
 - cloud water (H: bump patterns)
 - humidity + temp (V: InfoBox)
 - rain (V: cylinder)
 - vorticity and rain (wind curl) (H)



- User Study
 - V group, V + H group
 - free exploration of surface
- Evaluation
 - quizzes of C + E patterns and relationships
 - greater learning: V + H group
- Discussion
 - sig. dif. between groups
 - users naturally notice relationships without prior knowledge or interpretation

- Strength

- different haptic encodings allow users to perceive many additional cues in multidimensional data

- Critique

- representation of variables
- choice of evaluation tasks
- lacking a robust evaluation analysis

Questions?