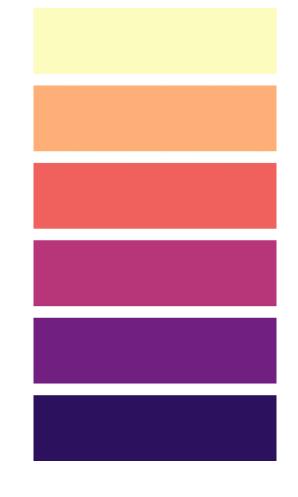
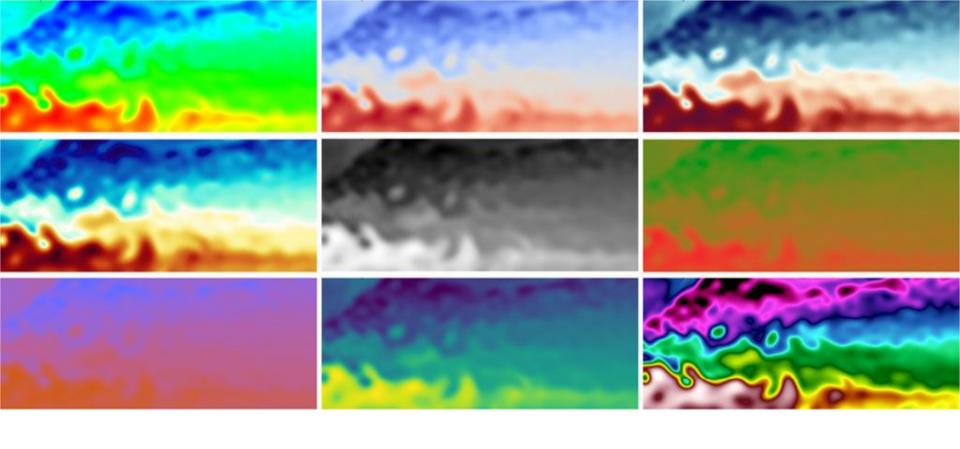
IEEE Trans. Visualization and Computer Graphics 2019

Colin Ware, Terece L. Turton, Roxana Bujack, Francesca Samsel, Piyush Shrivastava, David H. Rogers

# Measuring and Modeling the Feature Detection Threshold Functions of Colormaps

Presented by Jerry Yin





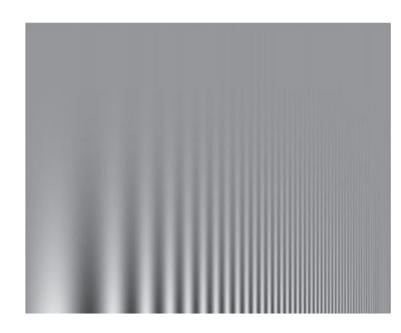
Which colourmap is the best at visualizing the data?

### Paper contributions

- Paper type: evaluation
- Describes way to measure frequency-dependent discriminative power function of a colourmap
  - Discriminative power: ability to distinguish different colours
  - Frequency-dependent: more later
- Defines metric for "overall discriminative power" across entire range of a colourmap

## **Spatial frequency**

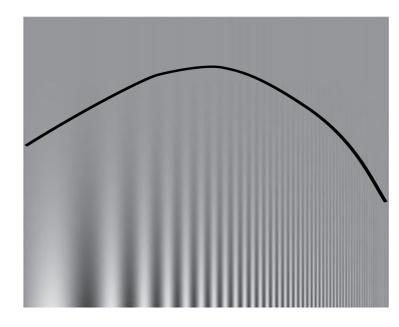
- Discriminative power depends on spatial frequency
- Uniform colour spaces (UCS) intended to be *visually* uniform
  - Based on measurements between large patches of uniform colour
- Thus, uniform colour spaces may not actually appear uniform in highfrequency datavis contexts!



The bands visually disappear at different heights along the image.

## **Spatial frequency**

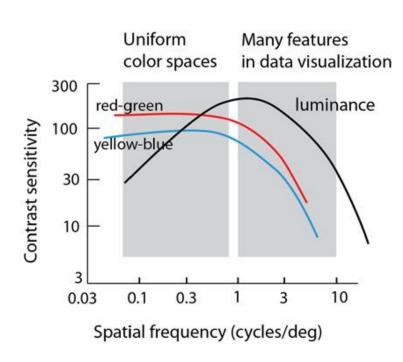
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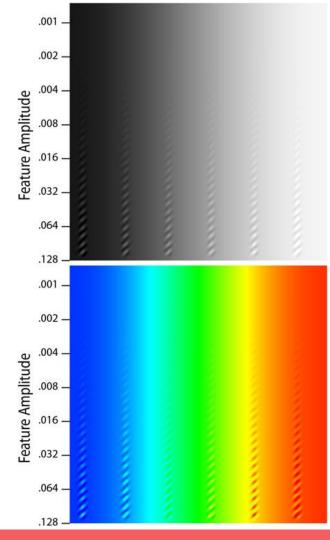
### Spatial frequency

- Discriminative power depends on spatial frequency
- Uniform colour spaces (UCS) intended to be *visually* uniform
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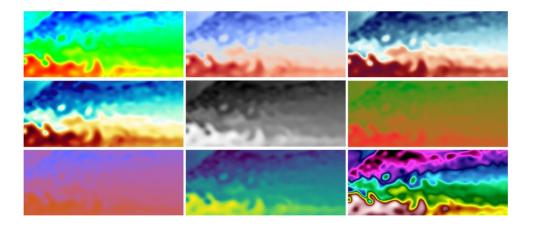


# **Empirical study**

- Paper devises empirical study for measuring discriminative power across multiple spatial frequencies
- Used 600×600px images
- For each column, participants click the area where the sinusoidal pattern disappears
- Tested nine colour sequences and three frequencies (10px, 15px, 45px)
  - For each sequence, tested 30 locations



## **Tested colourmaps**



RA: Rainbow colormap. This version comes from Paraview software

CW: The Moreland cool-warm colormap

ECW: Extended cool-warm colormap from Samsel

BOD: Blue-orange-red divergent colormap from Samsel

GP: A uniform grey colormap

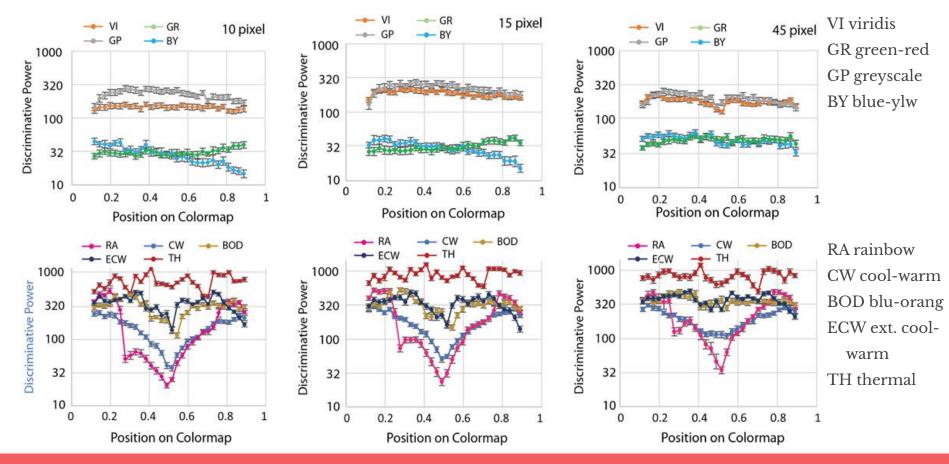
GR: Approximately equiluminous green-red colormap

BY: An approximately equiluminous yellow-blue colormap

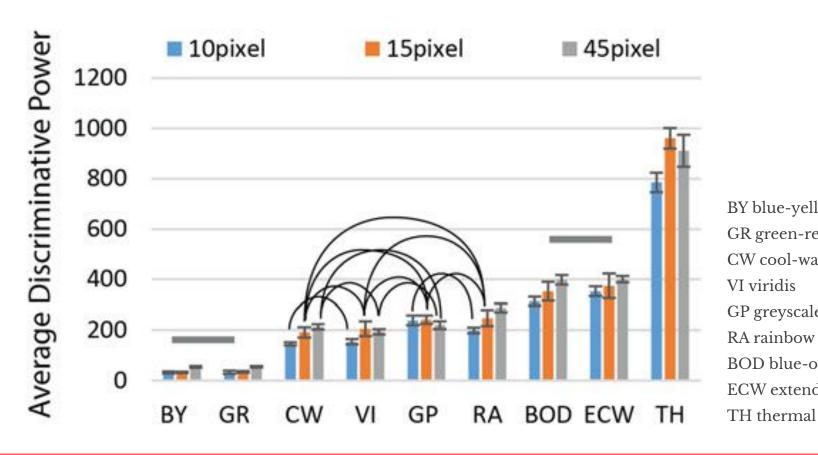
VI: Viridis colormap prized for its uniformity

TH: A colormap sometimes used in thermal imaging

#### Results



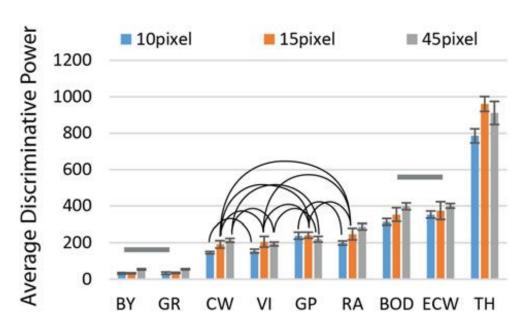
#### Results



BY blue-yellow
GR green-red
CW cool-warm
VI viridis
GP greyscale
RA rainbow
BOD blue-orange
ECW extended cool-warm

#### Results

- Ran 2-way ANOVA
- Arcs indicate where differences not statistically significant
- Ran Tukey HSD test (another significance test),
   horizontal bars show cases
   where colourmaps were not significantly different

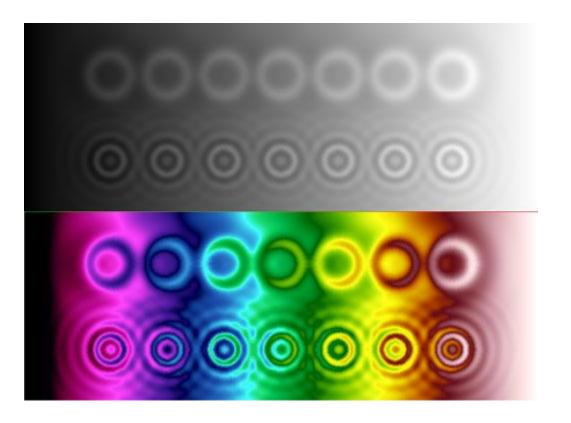


BY blue-yellow
GR green-red
CW cool-warm
VI viridis
GP greyscale

RA rainbow
BOD blue-orange
ECW extended cool-warm
TH thermal

## Which colourmap should I use?

 Despite having the highest discriminative power, the thermal colourmap is *confusing*.



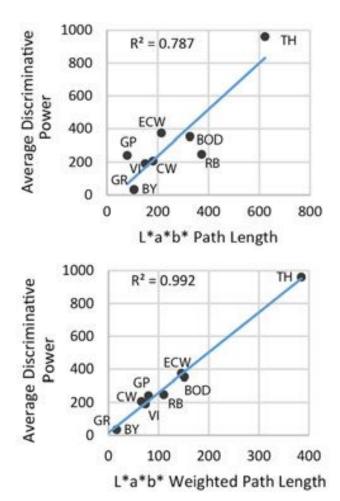
### Which colourmap should I use?

- Despite having the
  highest discriminative
  power, the thermal
  colourmap is confusing.
- Same also applies to divergent colourmaps, to some degree.



## **Reweighting CIELAB**

- Discriminative power should correspond to distance traversed by colourmap in uniform colour space
- Paper describes simplistic way to reweight CIELAB space to take into account the measured values in the paper
  - Equal weight is given to the
     10px, 15px, and 45px cases



# (Own) critique

- Instead of reweighting CIELAB in a way that is good for all datasets, maybe it would be better to collect data for many frequencies and reweight based on data that is currently being plotted
- Minimum discriminative power may be a better metric than mean discriminative power
- Outliers were manually removed
- Sample size a bit small: only 21 35 participants per colourmap

