$$\begin{split} \Delta E_{00}^* &= \sqrt{\left(\frac{\Delta L'}{k_L S_L}\right)^2 + \left(\frac{\Delta C'}{k_C S_C}\right)^2 + \left(\frac{\Delta H^*}{k_H S_H}\right)^2 + R_T \frac{\Delta C'}{k_C S_C} \frac{\Delta H'}{k_H S_H}} \\ & \text{Model the terminal below should be degree rather than the application of the  $k_L$  and  $k_L$  are the solution of the  $L^*$  and  $L^*$  are the solution of the solution of the  $L^*$  and  $L^*$  are the solution of the solu$$



# A Similarity Measure for Large Color Differences

Nathan Moroney, Ingeborg Tastl and Melanie Gottwals HP Labs

22<sup>nd</sup> IS&T Color and Imaging Conference, November 2014, Boston MA

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# **Outline**

### **Motivation**

• Large vs. Small Differences

### **Experiments**

- Hundreds of pairs of stimuli with  $\Delta E_{00}$  of 20
- Sorting of 9 pairs of differences from smallest to largest
- Web and laboratory

### Results

Observers rank pairs having equal  $\Delta E_{00}$  as being different

### Similarity metric

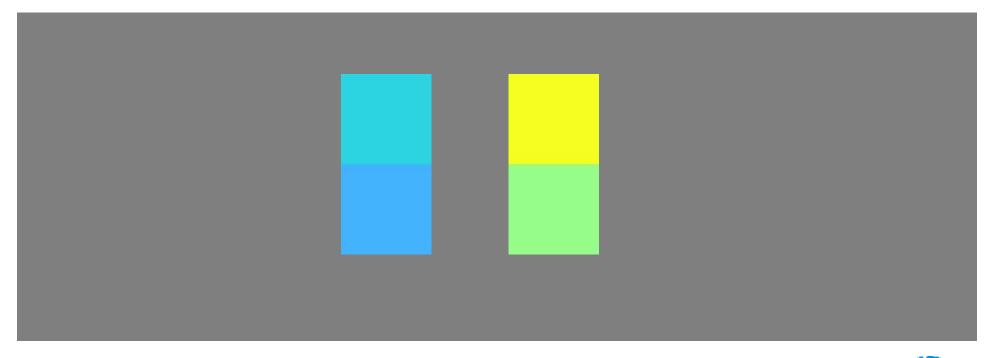
- Cosine similarity given categorical vectors
- Cosine similarity is small within a category & large across categories

### **Discussion and Speculation**



# **Motivation**

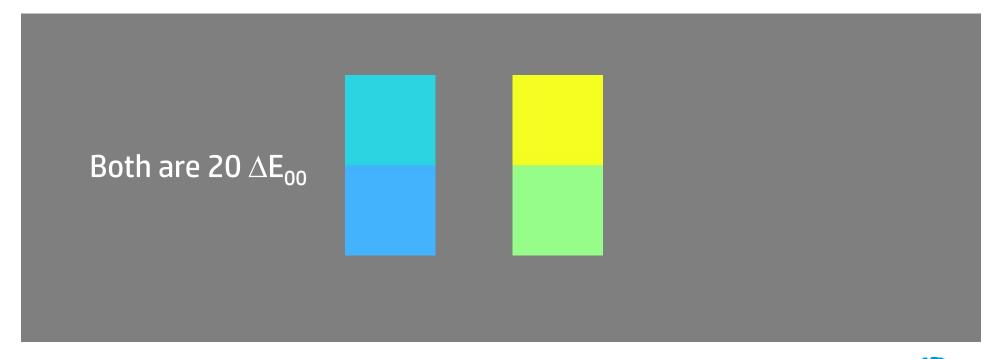
Which of these two pairs, has the larger  $\Delta E_{00}$ ?





# **Motivation**

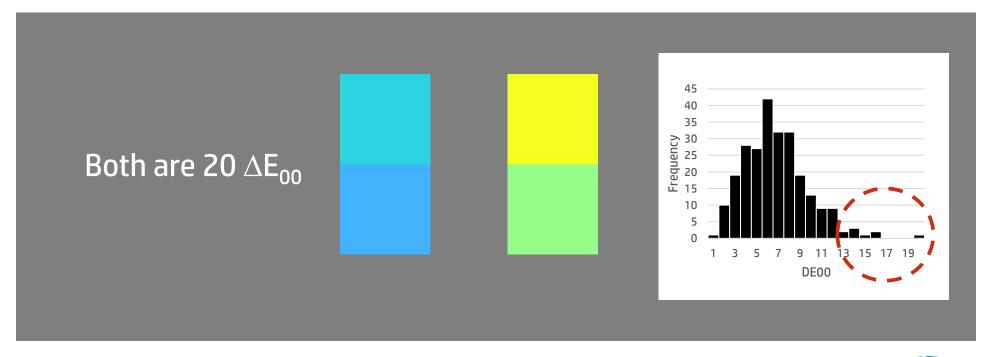
Using  $\Delta E_{00}$  they are equal





# **Motivation**

But isn't  $\Delta E_{00}$  only for small differences ? OK, so how to measure maximum error?





# Large vs. Small Color Differences

We'll fill this in by the end of the talk

	Small Differences	Large Differences
Application	Just Noticeable Difference	
<b>Central Question</b>	Do 2 colors match?	
Metrics*	$\Delta E*_{ab}$ , $\Delta E_{94}$ , $\Delta E_{00}$	
Underlying Metric	Euclidean distance, with weighting schemes	
Input	CIELAB coordinates, weights	
Output	"Geometric" distance, where a JND is approximately < 1	

<sup>\*</sup> Small color differences metrics not recommended for large differences or greater than ~5.



# **Experiments**

### **Sorting of Non-Repeating Random Vectors**

### Stimuli

- · Non-repeating random vectors
- Similar to a farthest-point sampling of vector endpoints
- All vectors within 0.0001 of 20  $\Delta E_{00}$ ,

### Task

- Sorting of multiple color differences
- Observer sorts 9 color differences from smallest to largest
- 21 blocks of 9 random color difference pairs

### Web-Based Experiment

285 participants; perform sorting of 1 block of differences; unknown displays

### Laboratory Experiment

• 12 participants; sort all 21 blocks of differences; sRGB mode HP DreamColor Z27x Display

Not regular or centroid plus offsets sampling

Not forced choice paired comparison of 2 pairs



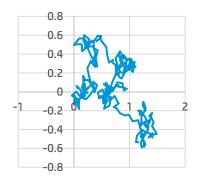
(Insert comment about Jim King here)

```
NonRepeatingRandomVectors(n, dist, t1, t2)
i = 0
while (i < n)
  test ave = F
  test ends = F
  start = RandomVector()
  end
        = start
  while ((Distance(start, end) - dist) < epsilon)</pre>
    end = RandomStep()
  ave = Average(start, end)
  for (i = 0 to number_vectors)
    if ((ave - ave[i]) < t1) test_ave = T</pre>
    d11 = start - start[i]
    d22 = end - end[i]
    d12 = strt - end[i]
    d21 = end - start[i]
    min diff = Min(d11, d22, d12, d21)
    if (min_diff < t2) test_ends = T</pre>
  if ((test_ave & test_ends) = T)
    AddVector(start, end)
    ++I
```

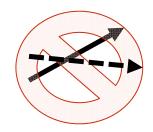


```
NonRepeatingRandomVectors(n, dist, t1, t2)
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    d21 = end - start[i]
    min diff = Min(d11, d22, d12, d21)
    if (min diff < t2) test ends = T
  if ((test_ave & test_ends) = T)
    AddVector(start, end)
    ++i
```

### Random walk

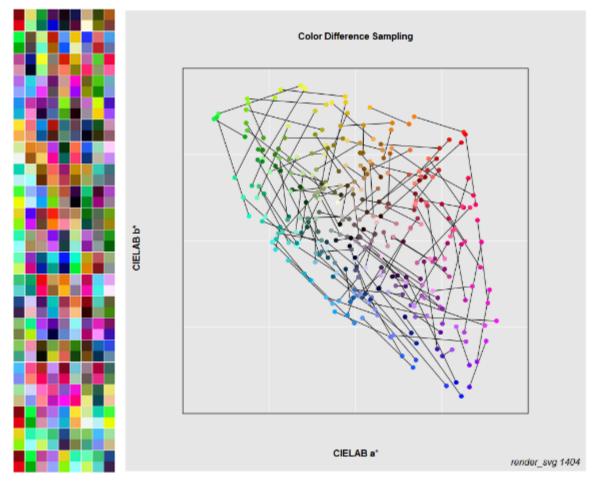


### Don't duplicate vectors



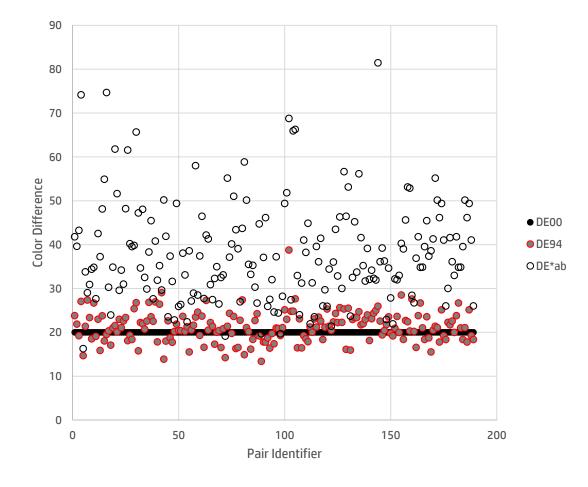


21 blocks rendered as RGB patches and as a CIELAB a\* versus b\* plot





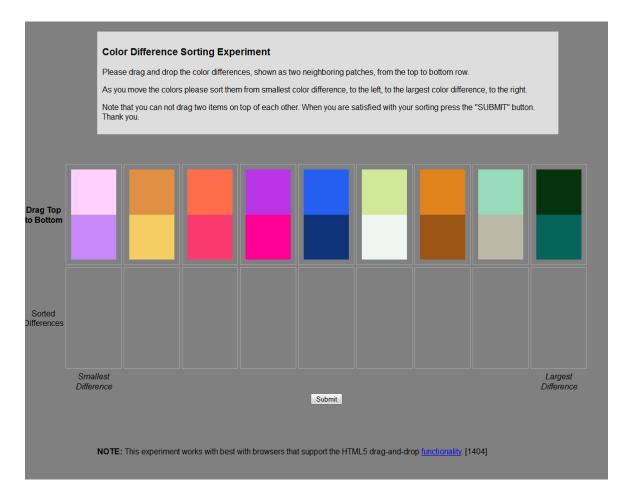
20  $\Delta E_{00}$  and the corresponding  $\Delta E_{94}$ and  $\Delta E^*_{ab}$  color differences





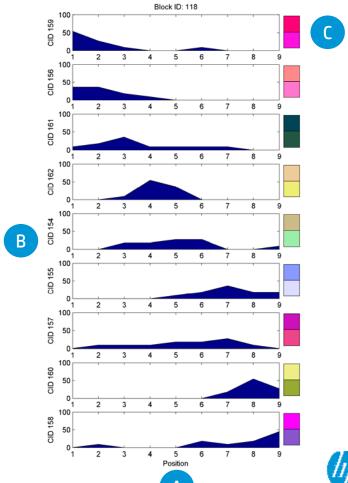
# Color **Difference Sorting**

HTML5 drag-and-drop interface with color patches rendered to 90x90 pixels in size

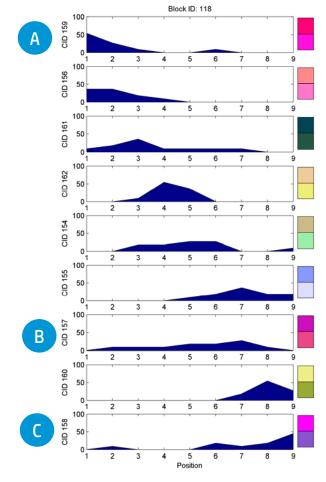




- Initial visualizations
  - Each set of 9 sub-plots is for 1 block
- A X-axis is sorted rank
  - Left: smallest difference
  - · Right: largest difference
- Y-axis is relative frequency
  - Larger: observers consistently used this rank
  - Zero: observers did not use this rank
- C Approximate color rendering
  - Shown to the right of each sub-plot
  - Approximate sorting top to bottom

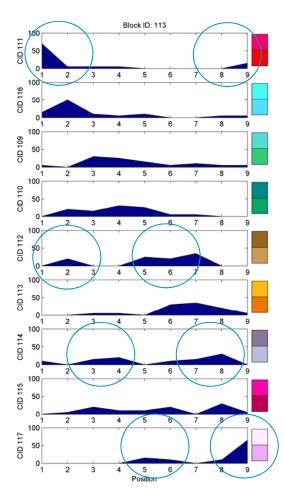


- A Smallest difference
  - Relatively narrower distribution
- **B** Intermediate differences
  - Relatively wider distributions
- Largest difference
  - Relatively narrower distribution
  - Qualitatively, at crossing of naming boundaries?
  - For ideal set of 9 equal differences
    - · Flat histograms for each pair
    - Not seen in the experimental data...





- Similar results seen across other blocks
- Similar results for web-based & laboratory
- **But also multi-modal distributions?** 
  - Complicates even basic analysis
  - Can occur for any ranking, shown circled to right
  - Almost as if there were multiple criterion for sorting....

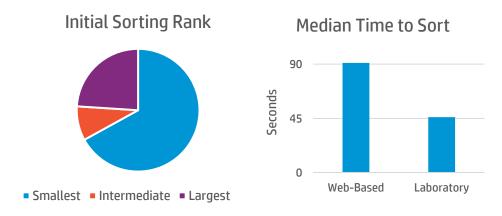


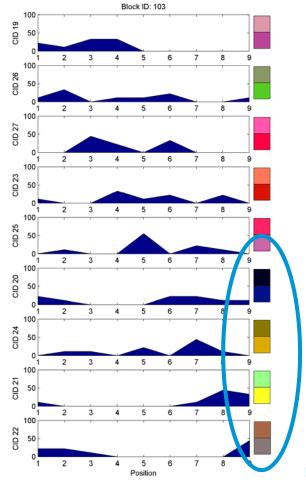


### Name boundary crossing

- Tends to result in a mode with a larger rank or sorted color difference
- Shown circled

### Other interesting results



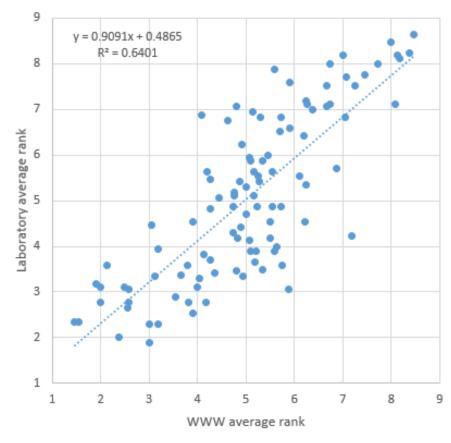




# Correlation of Web-Based and Laboratory Experiments

Limited to the approximately 2/3<sup>rds</sup> of pairs with uni-modal distributions\* for both experiments

### Color Difference Sorting: Uni-Modal Pairs





<sup>\*</sup> F. Schwaiger, H. Holzmann, and S. Vollmer, "bimodalitytest: Testing for bimodality in a normal mixture", R package version 1.0, (2013)

# **Similarity Metric**

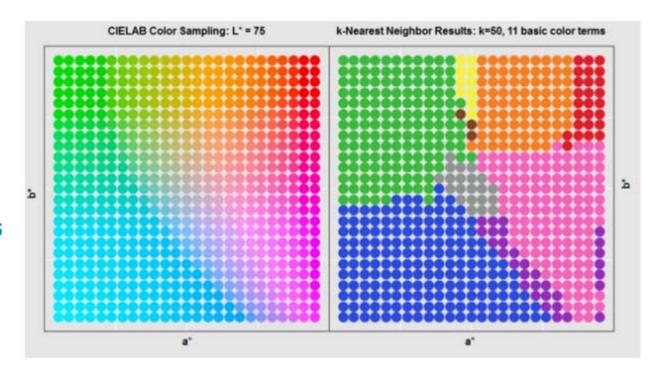
- Given these results can a similarity metric be designed that is:
  - Smaller within a color category
  - Larger across color categories
  - Not unlike results seen in Categorical Perception from vision science
- Not based on weighted Euclidean distances?
- Similar to other similarity metrics?
  - Value of 1 for identical
  - Value of 0 for dissimilar



# **Similarity Metric**

### **Use Categorical Vectors**

- K-nearest neighbors used to transform input CIELAB data, on left, to categorical counts, maximum shown color coded on right
- Start with basic 11 terms as the vocabulary
- Similar processing used for document classification





# **Similarity Metric**

### **Use Cosine Similarity of Categorical Vectors**

- K = 100
- Same example from the motivation slides earlier
- These two have  $\Delta E_{00}$  differences of 20...

$$\Delta S = \frac{\sum A \cdot B}{\sqrt{\sum A^2 \cdot \sum B^2}}$$

	Тор	Bottom	Тор	Bottom
sRGB	43, 212, 224	67, 179, 253	244, 254, 32	150, 253, 137
Gray	3	1	0	0
Blue	96	98	0	0
Green	0	1	0	83
Pink	0	0	0	0
Black	1	0	0	0
Green	0	0	0	0
Orange	0	0	0	1
Purple	0	0	0	0
Red	0	0	0	0
White	0	0	0	0
Yellow	0	0	100	16
Similarity	0.99		0.18	



# Large vs. Small Color Differences

We'll fill this in by the end of the talk

	Small Differences	Large Differences*
Application	Just Noticeable Difference	Consistently Describable Difference(s)
<b>Central Question</b>	Do 2 colors match?	When do 2 colors stop looking similar?
Metrics*	$\Delta E*_{ab}$ , $\Delta E_{94}$ , $\Delta E_{00}$	$\DeltaS$
Underlying Metric	Euclidean distance, with weighting schemes	Cosine similarity
Input	CIELAB coordinates, weights	Categorical or lexical vectors
Output	Geometric distance, where a JND is approximately < 1	Similarity measure, where 0 is completely dissimilar

<sup>\*</sup> Large color differences also relevant to image segmentation, analysis & retrieval.



# **Discussion & Speculation**

- Large color differences do not necessarily have a single sorting
  - Can be consistently sorted by multiple criterion
  - More sophisticated analysis needed to detect systematic trends in multi-modal sorting
- Small difference metrics probably not a good idea for maximum errors
  - At a minimum probably want to visualize
- Cosine similarity of categorical vectors is a promising metric
  - Initial stages of optimization but already useful in ways that differ from the  $\Delta E$ 's
  - Training data and algorithms are key aspects of the metric
- Experiment and data are public and ongoing
  - Same data could also be used to investigate more uniform color spaces
  - Would like to expert-source additional analysis & related experiments

