

Color Naming in Two Languages

Data Visualization Course Spring`16

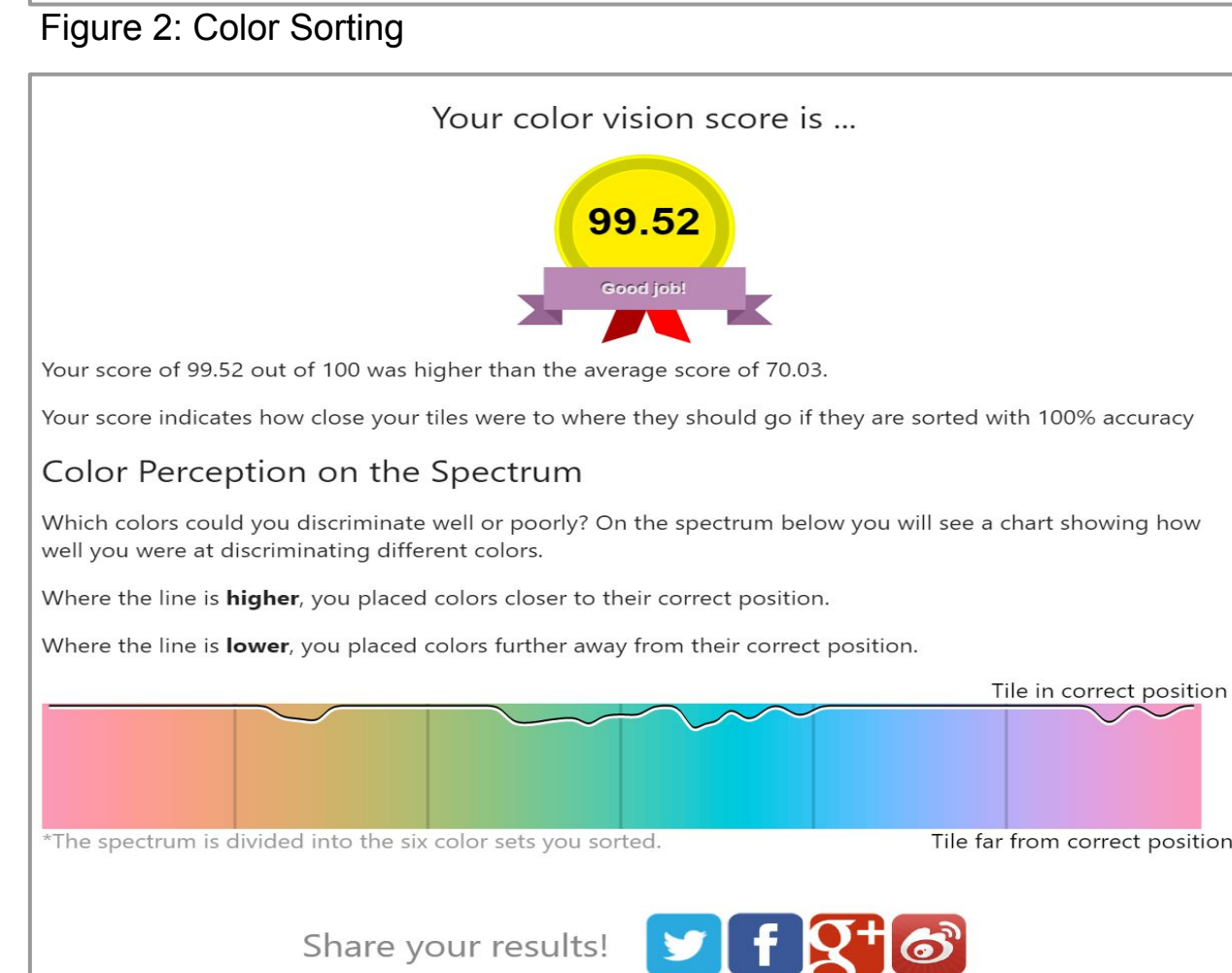
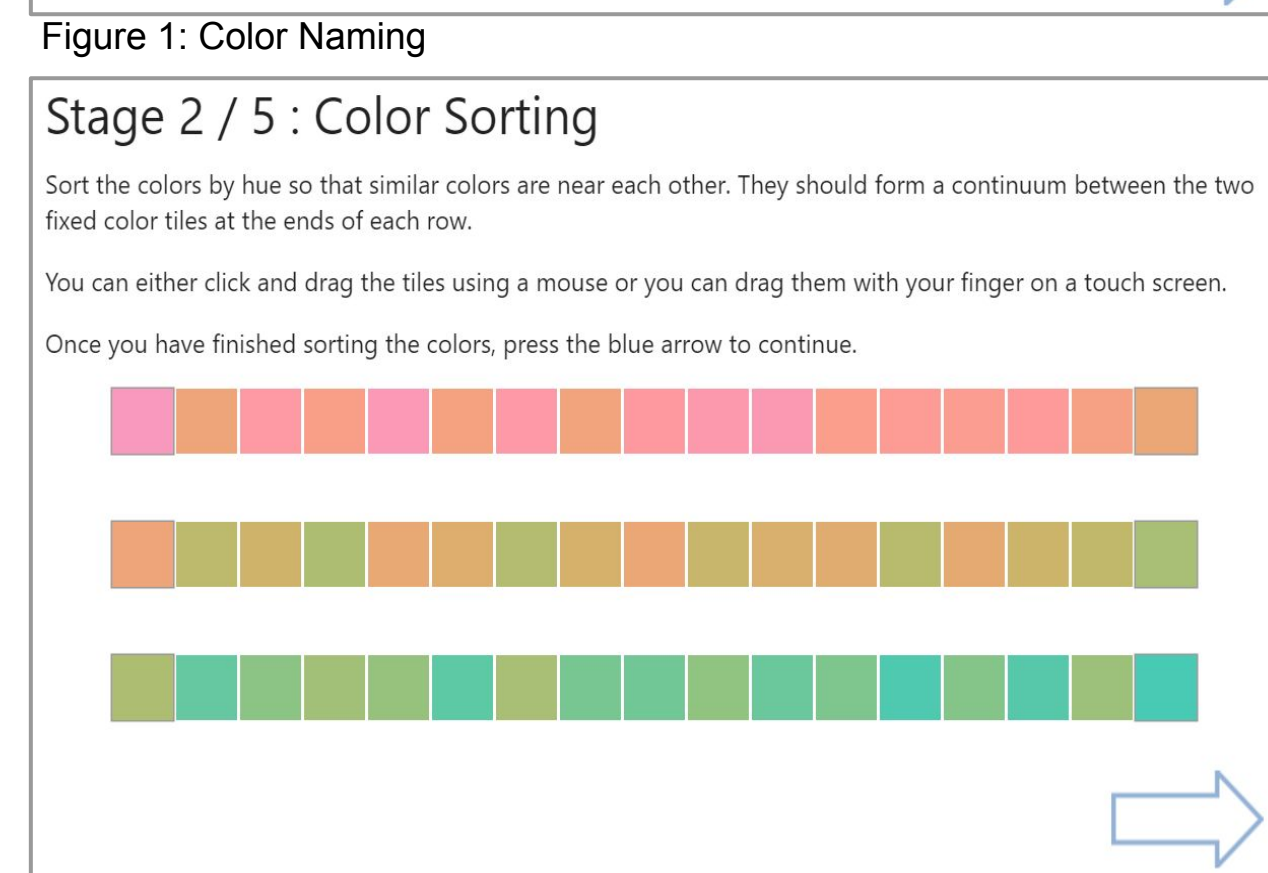
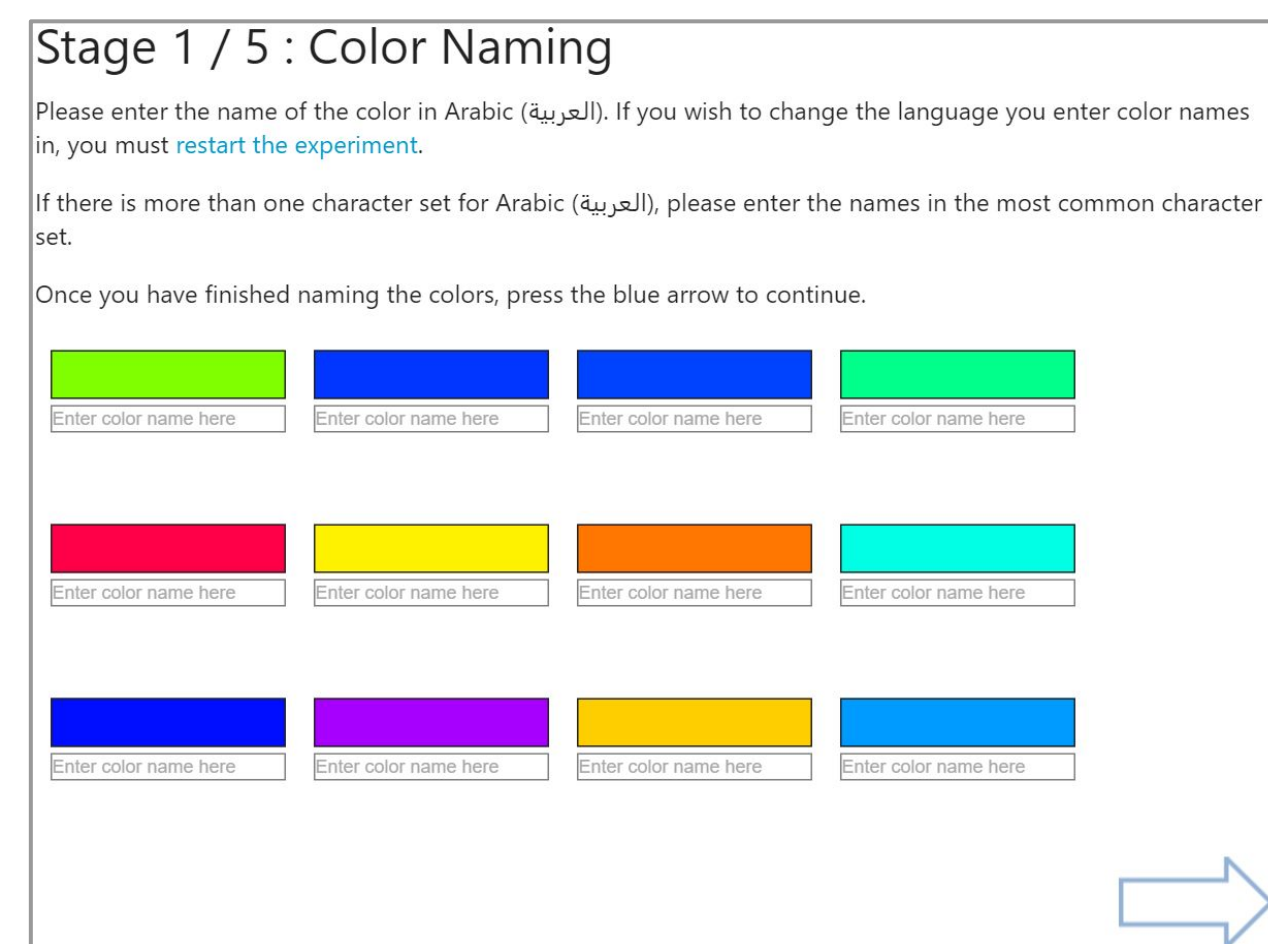
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Motive

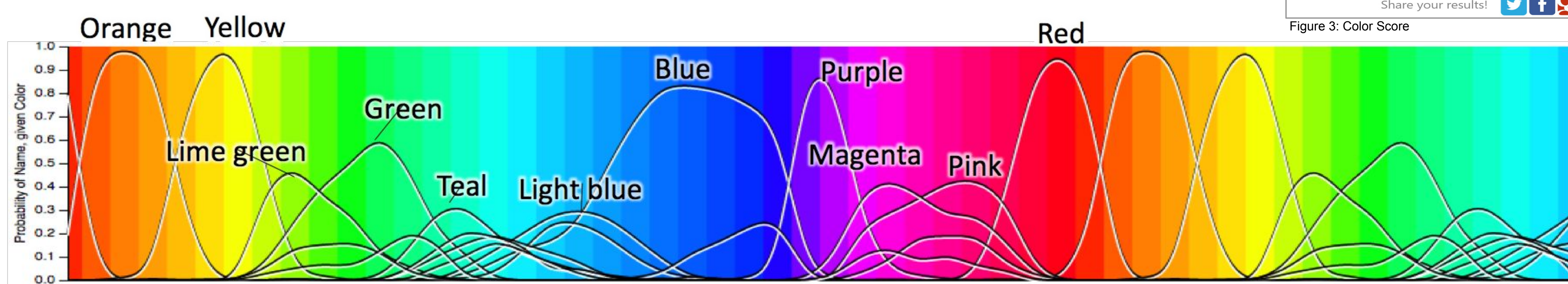
- ❖ Extend previous work on mapping different patterns of color naming across languages
 - Previous data sparse, some languages over-represented
 - Participants not all native speakers of language used
- ❖ New study for a more fine-grained exploration
 - Gathering denser data from few languages
 - Collecting more demographic information
- ❖ New data expands the potential to apply statistical tests
 - Where are the boundaries between color groups?
 - Are there statistically significant differences in color naming trends across languages?
- ❖ Prior research shows that language has dynamic effects on color perception
- ❖ Other work has explored the interaction between visual-perceptual differences related to language and color naming
- ❖ The goal of this study was to continue the exploration of this interaction while adding considerations such as situational lighting conditions and discipline-related color immersion

Study Design

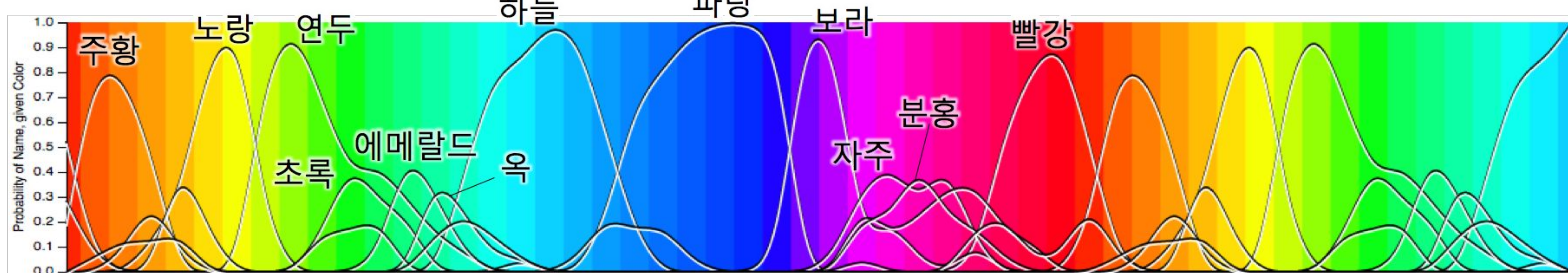
- ❖ We created an online “Color Perception” test and administered it on the LabInTheWild platform (labinthewild.org)
- ❖ Data Collected:
 - Demographic Information
 - including native language, education level, and situational lighting conditions
 - Color names of 36 colors [see Figure 1]
 - User viewed and named tiles (12 colors at a time)
 - Names were requested in users’ native language
 - Colors tiles chosen from max hue and value edges of RGB color cube
 - One color randomly chosen from 36 segments (bins)
 - Color Sorting [see Figure 2]
 - User sorted 6 lines of 15 colored tiles (3 at a time)
 - Colors chosen from a circle in Lab space centered on 0,0 in a,b and of uniform lightness.
- ❖ Motivation for users:
 - Users receive a color vision score and graphic to share on social media [see Figure 3]



English

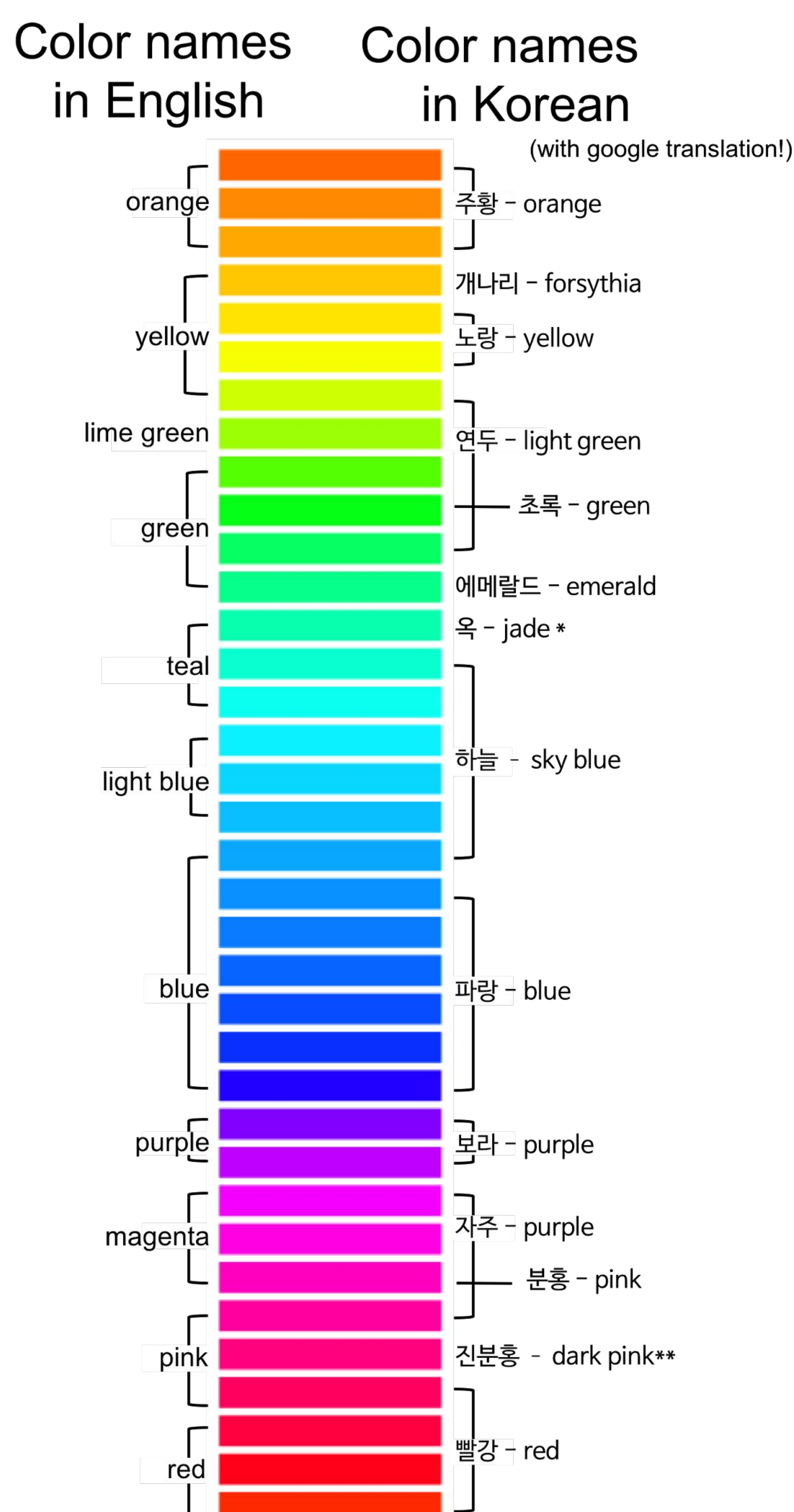


Korean



Analysis & Results

- ❖ Sparse preliminary data allowed for a limited quantitative exploration
- ❖ Data was collected from 347 participants
- ❖ More than a dozen languages were represented
 - 83.6% English, 8.9% Korean, 7.5% other, including Spanish, Chinese, Polish, and German
- ❖ There were enough Korean and English respondents to explore the pattern differences in color naming
- ❖ Colors across the Lab space were placed into 1 of 36 bins [as represented in the graphic to the right]
- ❖ Probabilities were calculated for the likelihood of a bin to be named one of 20 colors (ie “blue” or “red”)
- ❖ Probability distributions are represented in the line charts above
 - Greater values indicate a greater probability that a given bin will be given a particular color name
 - It is evident that there are some clear distinctions between English and Korean
 - Differences in the saliency of Blue-Green and Orange spectrums
- ❖ The ratio of distinct names out of total names given to a bin were calculated for Korean and a series of randomly sampled subsets of English (group N = 803)
- ❖ Paired-samples analyses revealed a significant difference between the English and Korean ratios for approximately 71% of pairwise comparisons
 - English tended to have larger ratios (more distinct words/total words applied to a bin)
- ❖ Further exploration of the more complete dataset will allow for increasingly robust conclusions
- ❖ Unsurprisingly, plots of the ratio distributions tend to mirror the general shape of the probability distributions above
 - Taken together, the two approaches begin to elucidate the nuances of color naming differences between the two focal languages



*'옥' is translated to 'jade' but '옥색', which means 'jade color', is translated to 'green'.
**'진분홍' is translated to 'Jinbunhong', which is actual pronunciation of '진분홍'. So we translated '진한 분홍'.

Limitations & Future Work

- ❖ Limited data space
 - To avoid to collect data sparsely, we limited the color space by sampling from only some edges of the rgb cube
 - We didn't get dense responses in languages except Korean and English
 - A clear next step is collecting more broad data
- ❖ Biased demographics
 - More than a half of participants have been or are in graduate school. (28.8% graduate school, 19.9% PhD, 2.1% Postdoc)
 - Instructions were in English, so our users presumably all could read English
- ❖ Other considerations
 - Spelling and structure differences may obscure or overemphasize color naming differences
 - Respondents reported eye-strain during the task
 - Feedback indicated we might need to give more explicit directions in future releases

Citations

- ❖ Chen, Y., Kim, Y., Thayer, K., and Wang, J. (2015). Color perception in different languages. Unpublished.
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- ❖ Farnsworth, D. (1957). The Farnsworth–Munsell 100 -hue test for the examination of color discrimination. Munsell Color Company.
- ❖ Reinecke, K. and Gajos, K. Z. (2015). Labinthewild: Conducting large-scale online experiments with uncompensated samples. In Proceedings of the 18th ACM Conference on Computer Supported Cooperative Work & Social Computing, CSCW '15, pages 1364–1378, New York, NY, USA, 2015. ACM.
- ❖ Reinecke, K., Flatla, D. R., & Brooks, C. (2016). Enabling designers to foresee which colors users cannot see. CHI' 16, May 07–12, 2016, San Jose, CA, USA. DOI: <http://dx.doi.org/10.1145/2858036.2858077>
- ❖ Winawer, J., et al. (2007). Russian blues reveal effects of language on color discrimination. In Proceedings of the National Academy of Sciences of the United States of America. National Academy of Sciences. <http://www.pnas.org/content/104/19/778>
- ❖ Wuergler, S. M., et al. (2012). Blue-green color categorization in Mandarin-English speakers. Journal of the Optical Society of America, 29 (2), 102–107. DOI: <http://dx.doi.org/10.1364/JOSAA.29>.