

Variations in normal color vision. IV. Binary hues and hue scaling

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We used hue cancellation and focal naming to compare individual differences in stimuli selected for unique hues (e.g., pure blue or green) and binary hues (e.g., blue-green). Standard models assume that binary hues depend on the component responses of red–green and blue–yellow processes. However, variance was comparable for unique and binary hues, and settings across categories showed little correlation. Thus, the choices for the binary mixtures are poorly predicted by the unique hue settings. Hue scaling was used to compare individual differences both within and between categories. Ratings for distant stimuli were again independent, while neighboring stimuli covaried and revealed clusters near the poles of the LvsM and SvsLM cardinal axes. While individual differences were large, mean focal choices for red, blue-green, yellow-green, and (to a lesser extent) purple fall near the cardinal axes, such that the cardinal axes roughly delineate the boundaries for blue vs. green and yellow vs. green categories. This suggests a weak tie between the cone-opponent axes and the structure of color appearance. © 2005 Optical Society of America

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1. INTRODUCTION

Conventional models of color appearance hold that the perception of color is organized according to a small number of privileged axes.^{1–5} In Hering's theory of color opponency, one of these axes represents variations in lightness or darkness while the other two encode the opposing dimensions of red vs. green and blue vs. yellow.⁶ By this account, the unique hues (colors that appear pure red, green, blue, or yellow) are special because they reflect the undiluted response of a single opponent process. All other hues are binary hues because they instead reflect mixtures of red or green with blue or yellow. For example, orange is composed of red and yellow, while purple is a blend of red and blue.^{7,8} Thus binary hues have a status subordinate to the unique hues because they have no representation in the model other than in terms of the contributions of the underlying unique hues. The stimuli corresponding to unique hues can be found by varying a spectral stimulus until it appears pure (e.g., to find the point at which a red stimulus appears untinged by blue or yellow).^{9–12} More generally, the red–green or blue–yellow responses to any stimulus can be measured by physically nulling the hue sensation (e.g., by adding a “green” light to the stimulus until any redness in the stimulus is canceled)¹³ or by scaling the component sensations (e.g., by judging the relative amounts of red and yellow that make up an orange stimulus).^{14–17}

Another approach to studying color appearance has

been to test for consensus in color naming across observers. Berlin and Kay¹⁸ found that languages have only a small number of basic color terms, in the sense that the terms are monolexic, used consistently by different speakers, and refer to color independent of particular objects. They also showed that the basic terms in different languages tend to be focused on very similar regions of color space, and that while languages vary in the number of basic terms, these follow a highly constrained order. For example, as refined in later work, a language with two terms is likely to have one encompassing white, red, yellow, and other “warm” colors with the other encompassing black, green, blue, and other “cool” colors. More recently, this broad pattern has been verified by analyses of color naming from the 110 unwritten languages sampled by the World Color Survey.^{19–22} The centroids of the stimuli labeled by basic color terms in these languages cluster strongly around similar points in color space, showing that respondents view the spectrum in very similar ways regardless of the varying number of categories into which their lexicons partition it. While counterexamples have been noted (e.g., Ref. 23), the similar clustering across languages suggests that the special and shared status of basic color terms may reflect special and shared properties of the human visual system or of the visual environment.

Like the unique hues, the evidence for basic color terms implies that some stimuli have a privileged status in color

appearance. Indeed, when given comparable stimulus sets, English-speaking observers select the same stimuli for unique hue settings as they do when choosing the best example or focal stimulus for red, green, blue, or yellow.^{24,25} However, basic color terms are not restricted to the set of primaries given by the three opponent axes. For example, English has 11 basic terms, which include the Hering primaries (white, black, red, green, blue, yellow, and a neutral gray) but also secondary colors (orange, purple, pink, and brown).¹⁸ Thus, by the criterion of consensus color naming, orange in English has a status similar to that of red or yellow and may have a status superior to that of a comparable mixture category such as yellow-green, for which there is not a basic term. Moreover, the stimuli labeled by different basic color terms do not support the independence of the luminance and chromatic dimensions assumed by many color-opponent models. For example, green and blue terms apply to stimuli over a wide range of lightness levels, while red is restricted to low values and yellow is used only for stimuli with a high lightness.^{18,21,26} Thus the specific structure of color appearance implied by the standard three-channel model of color opponency and by basic color terms differ, and this circumstance has led to suggestions that there may be an explicit neural process corresponding to each of the 11 basic categories.²⁶

In this study we examined the structure of color appearance by observing individual differences in color naming. Subjects with normal color vision have been previously shown to vary widely in the stimuli they select for the unique hues^{10,27–31} and in the focal stimuli they select for basic color terms.^{18,21,22,31} Thus a yellow that appears distinctly reddish to one observer might appear strongly greenish to another. In previous studies of these variations, we found that the stimuli observers choose for different unique hues are largely uncorrelated.¹⁰ For example, a subject whose unique yellow is more reddish than average is not more likely to choose a unique blue that is more reddish (or more greenish) than average. The independence of the unique hues is surprising given that many factors that affect visual sensitivity (such as differences in screening pigments or in the relative numbers of different cone types) should influence different hues in similar ways and thus predict strong correlations between them.¹⁰ However, a number of studies have shown that the unique hue loci are not in fact clearly tied to measures of visual sensitivity^{29,32–35} and may instead reflect learning or adaptation to specific properties of the color environment.^{34,36–39} By either account, our results suggest that the variations in the axes for the red–green and blue–yellow dimensions of color appearance—or between the two poles of the same opponent axis—are controlled by independent factors.

In the present study our aim was to look more closely at the patterns of variation in color naming by sampling color space more finely. In particular, we were interested in the range in color space over which hue choices are correlated and whether different patterns emerge for the unique hues and intermediate hues. For example, even if the selections for red and yellow are uncorrelated, to the extent that orange reflects the combined “responses” of red and yellow, the loci for orange might be expected to

covary with the loci of the underlying primaries. Alternatively, if focal orange is fine tuned by its own physiological or environmental constraints, then it might instead float freely between red and yellow. In turn, hues like orange and purple for which English has basic color terms might vary in different ways than blue-green or yellow-green, which may instead correspond more to the boundaries between categories. Comparing individual differences in the unique and binary hues might thus provide clues about the nature and number of the processes calibrating color appearance. A further goal of our study was to extend measures of individual differences in color appearance to include the dimensions of saturation and lightness and thus to characterize the patterns of variations more fully within the volume of color space. Our results show that the range of individual differences in color naming is similar for unique and binary hues and that there are again only weak correlations between the color categories from neighboring regions of color space. Thus, by these criteria, the unique hues do not emerge as special and do not alone fully anchor the structure of color appearance for an individual.

2. METHODS

Stimuli were presented on a Sony 20se monitor controlled by a Cambridge Research Systems VSG graphics card. The monitor was calibrated with a PR650 Spectracolorimeter, and gun luminances were linearized through look-up tables. The test colors were presented on a uniform 6 deg × 8 deg background provided by the monitor screen. The background had a mean luminance of 30 cd/m² and a mean chromaticity equivalent to Illuminant C (CIE 1931 $x=0.31$, $y=0.316$). (Note this differs from conventional studies of the unique hues, which have instead typically used narrowband stimuli presented on a dark background, but it has the advantage that we could explore the foci for moderately saturated lights under steady adaptation. To the extent that observers are adapted to the background, the results are unlikely to depend on the choice of the specific chromaticity chosen for the neutral background.^{16,40,41})

Color and luminance were specified in terms of a scaled version of Derrington, Krauskopf, and Lennie⁴² color space, in which the origin corresponded to the background color and contrast varied as a vector defined by the luminance, LvsM and SvsLM cardinal axes. Units in the space were related to the r , b chromaticity coordinates in MacLeod–Boynton⁴³ space and to Michelson luminance contrast (L_c) by

$$\text{LvsM contrast} = (r_{mb} - 0.6568) * 2754,$$

$$\text{SvsLM contrast} = (b_{mb} - 0.01825) * 4099,$$

$$\text{LUM} = 3 * L_c.$$

We used three sets of stimuli and procedures to measure individual differences in color judgments.

A. Unique and Binary Hue Settings

In the first case, subjects made both unique hue settings (for red, green, blue, or yellow) and binary hue settings (for orange, purple, yellow-green, or blue-green). Stimuli were moderately saturated isoluminant pulses, presented at the full contrast for 1 s and ramped on and off with a Gaussian envelope (with a standard deviation of 250 ms). The stimuli all had the same maximum contrast of 80 in the space and thus varied only in hue angle within the LvsM and SvsLM plane, with isoluminance defined photometrically. The hues were presented in a central 2-deg field demarcated from the 6×8 deg background by a narrow black outline. Between stimuli the field remained at the same gray as the background.

For each setting, subjects first adapted to the gray background for 1 min. Hue loci were then estimated with a 2AFC staircase procedure. On each trial, the observer responded whether the target hue was biased toward one of the target's neighboring hues or the other. For example, for unique red, they responded whether the color appeared either too purple or too orange, while for purple they responded too blue or too red, etc. Successive hues were then varied using two randomly interleaved staircases, with the hue angle estimated from the mean of the final six of ten reversals from both staircases. During a 1-h session the eight hues were tested two times each in random order and were retested in a second session for each subject approximately one week later. Observers were 73 students at the University of Nevada, Reno (UNR). All subjects were screened for normal color vision by the Neitz Color Test⁴⁴ and the Ishihara pseudoisochromatic plates and were naïve with regard to the specific aims of the study.

B. Individual Differences in Hue, Lightness, and Contrast

In the second experiment, stimuli were varied not only in hue but also in lightness and saturation, in order to compare the variations for each color in terms of the three principal attributes of color appearance. Because this required varying the stimuli along three dimensions instead of one, we used a different procedure in which subjects were shown a palette of colors at a fixed contrast, and then selected the best example of a given color term from this palette. This procedure was thus more similar to the types of procedures used in cross-linguistic studies of color naming. In the present case, the palette was composed of a 9×9 array of stimuli that varied in hue across columns and in lightness across rows, with the lightness and hue steps equated within the scaled space defined above. Each circular patch subtended 1.15 deg with 1.3 deg between the patch centers. The background in this case subtended 15×20 deg. The term to be selected for was written in the upper left corner of the background. Subjects were first shown a broad range of colors spanning a hue angle of 112 deg centered at random points in color space around the nominal focal stimulus, and they selected the best patch for the term indicated by using a keypad to move a thin black ring over the array to highlight their choice. The next five trials then zoomed in at random points around the selected chip and showed a much finer color array spanning 45 deg in color angle that

was centered at random points around their color selection. During a given run all stimulus arrays had a fixed contrast, with the eight color terms and five repetitions presented in random order. Color terms again included the four unique hues and the four binary terms. Contrast across runs varied in random order in steps of 20 units, from 20 up to the maximum contrast available for a given region of the space. A separate new sample of 53 UNR students participated in this experiment. As before, these subjects were all screened for normal color vision and repeated the settings in two daily sessions.

C. Hue Scaling

To provide a still finer sampling of color space, in the final condition we used a hue scaling task to rate the color appearance of 24 isoluminant stimuli falling at intervals of 15 deg along a circle spanning the LvsM and SvsLM plane. These stimuli all had a fixed contrast of 80 and were again shown in a square 2-deg field, pulsed for 1 s as in the unique and binary hue settings described in the first condition above. The scaling procedure followed the procedure used by De Valois and colleagues.¹⁶ For each stimulus, subjects rated the hue by pressing separate buttons to indicate the relative amounts of red, green, blue, or yellow. For example, the response to a reddish orange might be three red presses and two yellow. Subjects were instructed to use at least five presses to score the color but were allowed to use more if they wanted to use finer scaling (e.g., seven red and one yellow for a red that appeared only slightly tinged with yellow). Each angle was presented five times in random order, and subjects repeated the settings on a second day. On a separate run during the session the hues were again shown, and subjects selected a color label for the hue by choosing from the four unique and four binary terms displayed at the bottom of the screen. A separate sample of 59 additional color-normal students took part in these settings.

3. RESULTS

A. Unique and Binary Hue Settings

Figure 1 plots the mean hue angles chosen by individual subjects for each of the color terms tested. The average angles across subjects and their range are given in Table 1. As in previous studies,²⁸ the range of variation in the hue settings is pronounced, to the extent that the range of focal choices for neighboring color terms often overlap. Thus some subjects chose as their best example of orange a stimulus that other subjects selected as the best example of red, while others selected for orange a stimulus that some individuals chose for yellow. In fact, there was only one narrow region of the color circle, between red and purple, that did not receive choices for any of the eight terms.

Surprisingly, the degree of consensus among observers did not clearly distinguish unique from binary hues, nor basic terms from nonbasic terms. For example, both blue and green spanned a relatively large range of hue angles, while the narrowest range was for blue-green. Thus there was much greater agreement between subjects about the border separating the blue and green categories than about the focal stimulus for either category. Of course,

this comparison depends on the choice of space. The cone-opponent space explicitly captures how the hues vary in terms of the dimensions underlying early postreceptoral color coding but makes little assumption about the salience of hue differences along different chromatic angles and thus may fail to reveal the perceptual magnitude of the spread for each hue. To explore this, Table 2 gives the mean and standard deviation of the hue angles converted

to the CIE u'/v' space, which is designed to roughly equate the perceptual distances between different regions of color space. Within this space the range for red and blue-green are greatly expanded, while yellow and purple are contracted. Yet it is still the case that as a group the unique hues do not differ from the binary hues in the degree of consensus.

Within the cone-opponent space of Fig. 1 the blue-green settings are not only narrow but are also notable for falling close to the $+M/-L$ pole of the $LvsM$ axis (especially since the empirically defined axis may be rotated slightly clockwise relative to the nominal axis that we used based on the standard observer).⁴⁵ Red is well known to be the only unique hue that lies near one of the cardinal axes.^{10,16,46,47} However, the fact that blue-green settings cluster tightly around the opposite pole of the $LvsM$ axis indicates that the blue and green categories (if not the unique points) may also be more closely associated with the cardinal axes than normally supposed. In particular, whether a stimulus appears more green or more blue (and whether a red appeared too blue or too yellow) depends roughly on whether it results in more or less S -cone excitation relative to the background. However, as with the unique red settings, this is only a very rough correspondence, for the range of individual differences in the color choices far exceeds the plausible range of variation in the stimulus angles isolating the $LvsM$ axes for different observers.^{45,48} It is also notable that the average settings for yellow-green fall close to one of the poles of the $SvsLM$ axis (and that purple skirts the opposite pole, though in this case the average differed more clearly from the S axis). Thus again, while focal yellow or green lies at intermediate angles in terms of the cardinal axes, the partition defining whether a color is too yellow or too green falls roughly at the $SvsLM$ axis and thus depends on whether the hue has a larger or smaller L/M ratio than the background. (Again this must at best be a very rough correspondence, yet the $SvsLM$ axis is more strongly affected than the $LvsM$ axis by factors such as variations in macular pigment density, and thus the range of potential variation in the $SvsLM$ cardinal axis is much larger.^{45,48})

Table 3 shows the correlations between the settings for the different color terms. Previously we found that there is little correlation either among unique hue choices¹⁰ or among different focal judgments for the unique hue terms.²² The present results confirm this and, moreover,

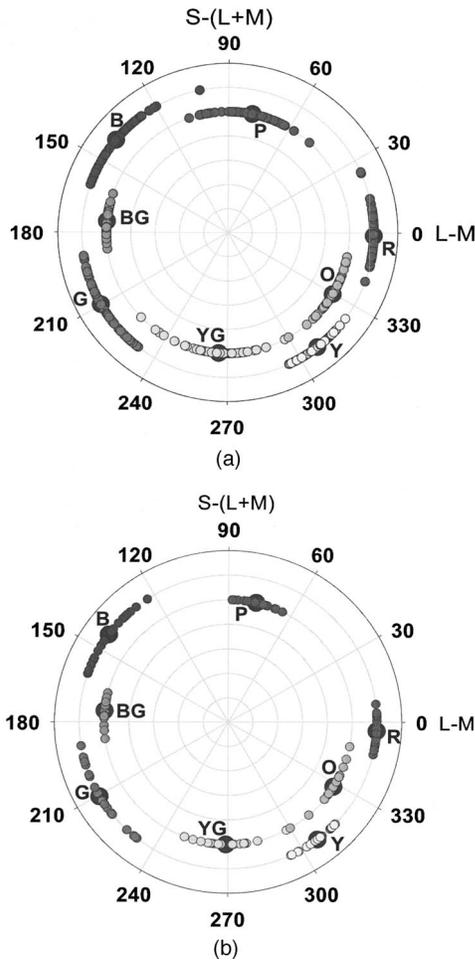


Fig. 1. Mean hue angles selected by individual observers for the eight different color terms. (a) all observers, (b) settings for the subset of observers who selected the hues most consistently.

Table 1. Mean Hue Angles for Unique and Binary Hues in the Scaled $LvsM$ and $SvsLM$ Space and the Range and Standard Deviation (SD) across Observers

	R	P	B	B-G	G	Y-G	Y	O
All subjects ^a								
Mean	-1.21	78.7	140.4	169.6	-150.3	-94.7	-52.0	-30.4
Range	44.5	60.7	59.1	26.6	43.9	73.5	28.8	48.3
SD	4.56	7.06	5.46	2.52	6.7	6.84	3.88	4.67
Most consistent subjects ^b								
Mean	-3.5	77.2	143.4	174.6	-149.6	-91.0	-53.1	-31.9
Range	20.1	24.0	37.5	21.8	42.9	48.4	21.1	48.3
SD	3.01	4.95	4.56	1.6	4.55	4.17	2.6	4.02

^aResults for all 73 subjects.

^bResults for the 21 subjects who set the hues most consistently.

Table 2. Mean and Standard Deviation (SD) of the Hue Angles within the u'v' Uniform Color Space

	R	P	B	B-G	G	Y-G	Y	O
All subjects								
Mean	-2.95	281.9	248.8	187.2	128.7	99.0	80.1	59.3
SD	18.2	4.17	10.0	12.8	11.4	5.43	4.38	10.5
Most consistent subjects								
Mean	-0.4	281.4	248.9	187.7	129.7	98.4	80.4	60.6
SD	19.5	4.0	10.4	12.6	11.8	3.98	4.28	11.0

Table 3. Correlations between Hue Angles Chosen for Different Color Terms^a

	R	P	B	B-G	G	Y-G	Y	O
All subjects								
R	0.57*	0.23*	-0.05	0.0	-0.29*	-0.06	-0.07	0.01
P		0.39*	0.31*	-0.04	-0.09	-0.07	-0.16	-0.10
B			0.67*	0.02	0.02	0.02	-0.29*	-0.17
B-G				0.60*	0.19	0.05	-0.14	-0.12
G					0.61*	0.12	-0.23*	-0.12
Y-G						0.57*	0.19	0.04
Y							0.31*	0.45*
O								0.66*
Most consistent subjects								
R	0.61*	0.27	0.06	0.13	0.19	0.09	-0.20	0.23
P		0.40*	0.12	0.22	0.16	0.12	-0.28	-0.13
B			0.83*	0.34	0.11	0.09	-0.32	-0.36
B-G				0.85*	0.25	0.13	-0.40	-0.06
G					0.84*	-0.18	-0.47*	-0.33
Y-G						0.83*	0.18	0.05
Y							0.77*	0.56*
O								0.82*

^aNote: Cells with asterisks along the diagonal show the correlation between repeated settings for the same hue across two sessions. * $p < 0.05$.

show that the settings for both unique and binary hues are also largely uncorrelated. The independence of the unique hues is surprising in two regards. First, many models of color appearance assume that the opponent hues (e.g., blue and yellow) are shaped by common factors (e.g., the equilibrium axis for the red–green dimension). Yet these factors do not appear to strongly constrain how individuals vary within each category. Second, as we noted at the outset, most conventional models of color appearance assume that the binary hues are represented only in terms of the underlying unique hues. Yet the individual choices for the binary categories cannot be predicted from the choices for either component unique hue.

The lack of consistent correlations between the different hues across subjects could occur if individual subjects were inconsistent in their hue settings. In fact, the diagonal cells in the matrix of Table 3 show the correlation between the settings for the same color across the two sessions, and these are low for some of the terms. (Note that this does not directly imply that subjects were unreliable in their settings but only that they were inconsistent relative to the range of variation across the group.) To test whether intraobserver variation was masking a dependence between the different hues, we reanalyzed the settings for the subset of observers who chose the focal stimuli with the highest reliability (as in our previous

study¹⁰). Subjects were chosen by excluding any observer whose range of four settings (two from each daily session) exceeded the mean range on any color by more than 1.5 standard deviations. This left a pool of 21 observers whose results are shown in Fig. 1(b) and in the lower halves of Tables 1 and 3. For this subset, the consistency of repeated settings was much higher, while the variance between observers was roughly halved. However, individual differences remained substantial. Moreover, the correlations among different hues remained weak. Thus the independence regarding the different hue settings is unlikely to be an artifact of noise in the observers' settings.

We also asked whether the weak dependence between unique and binary hues occurred because we correlated only pairs of colors. If blue and green vary independently, then if blue–green represented a “halfway point” between them, it might be tied more closely to the average of an observer's blue and green loci rather than to the setting for either color alone. We therefore compared the correlations between each hue and the mean of its two neighbors (Table 4). This comparison showed a consistent relationship with the bounding neighbors for yellow and for orange but still weak dependence on the bounding neighbors for the other colors and no clear tendency for unique and binary hues to behave differently. This again sug-

Table 4. Correlation between the Angles Chosen for Each Term and the Mean of the Angles for the Two Bounding Terms^a

Hue	All Subjects	Consistent Subjects
Primary		
Red vs. orange/purple	0.21	0.36
Green vs. blue-green/yellow green	0.18	-0.05
Blue vs. blue-green/purple	0.30*	0.27
Yellow vs. orange/yellow-green	0.39*	0.53*
Binary		
Purple vs. blue/red	0.40*	0.22
Blue-green vs. blue/green	0.15	0.38
Yellow-green vs. yellow/green	0.22	-0.11
Orange vs. yellow/red	0.31*	0.65*

^a * $p < 0.05$.

gests that there is little joint constraint on the individual foci for unique and binary hues.

B. Effects of Lightness and Contrast

In the next set of experiments we asked how the focal color settings depended on the lightness and contrast of the stimuli. As noted in Section 2, this required sampling a much wider range of colors, and we therefore changed the procedure so that subjects picked the focal stimuli from a palette. The selections for individual observers are shown in Fig. 2. For each of the eight colors, the top panel shows the chosen hue angles within the LvsM and SvsLM chromatic plane (similar to Fig. 1), and the bottom panel plots the elevation out of the isoluminant plane. Successive radii show the settings at contrasts ranging from 20 to 100 units. For orange, the monitor gamut limited the maximum contrast at higher lightnesses to 80, and for yellow, green, and yellow-green, the maximum was 60. Fewer settings are therefore shown for these colors (and are shown with a different scaling for the radii).

Relative to the settings in the preceding cancellation task, mean hue angles in the present task tended to be biased away from the LvsM and toward the SvsLM axis, perhaps reflecting weaker sensitivity to SvsLM contrast in the palette stimulus. Note also that for these settings the stimuli varied along spheres of fixed radius within the cone-opponent space. Thus increasing or decreasing the lightness outside the isoluminant plane required a trade-off between luminance contrast and chromatic contrast and in this sense provided a measure of the relative importance of hue and lightness for the focal choices. That is, stimuli with a high lightness could be chosen only by sacrificing chromatic saturation. Nevertheless, for certain color terms the focal choices had a strong lightness component. For example, most subjects chose stimuli for red and purple that were darker than the background, while for yellow the focal choices had a higher lightness. This confirms previous studies in showing that lightness level is an important dimension of some focal colors and of yellow in particular.²⁶ Consistent with this, yellow also had the lowest variance in the lightness settings, while for all other colors the standard deviation of the lightness angles exceeded those for the corresponding hue angles. This

could indicate that lightness is less important to the judgment. However, an alternative is that subjects vary more in their preferred lightness values. In fact, we have recently found that the focal choices for different languages in the World Color Survey differ more in their lightness settings than in their hue settings (relative to the respective within-language variations),²² and thus it is likely that the variations in lightness levels do partly reflect actual variations in subject's preferences.

This is further suggested by the relationships among different lightness settings. Table 5 shows the correlation matrix among the eight color terms. Values below the diagonal give the correlations between the hue angles and are consistent with the preceding experiment in showing that the variations between hues are largely independent (though notably the strongest correlation is again between orange and yellow). The cells above the diagonal give the corresponding values for the lightness settings. The correlations are again weak overall, yet they are clearly stronger than for the hue settings, and in all cases the significant values are positive. This suggests that, unlike the hue settings, the lightness settings for individual observers revealed a general tendency to choose lighter or darker samples for their focal stimuli.

Unlike both hue and lightness, the correlations in the foci across different contrast levels were strong. Table 6 illustrates these for the red settings. (The pattern for the other colors was similar.) Because different contrasts and color terms were randomly intermixed during testing, such results suggest that in this task subjects were relatively consistent at selecting the same color-luminance angle in their settings, regardless of the contrast of the stimulus. In turn, this finding reinforces the conclusion that the choices for different terms are largely independent and that this independence reflects actual differences between observers rather than variance within the observers' settings. Moreover, it suggests that the differences between observers are largely captured by the color-luminance angles of their stimuli.

C. Hue Scaling

The preceding results showed that the variations in focal colors across neighboring color categories are largely independent. That is, the color a subject selects for red does not predict his or her selection for orange. In the final experiment we explored the pattern of variation not only across but also within color categories—for different shades of red or orange—by measuring individual differences in a hue scaling task. As noted in Section 2, in this case the stimuli were 24 hues spanning the LvsM and SvsLM plane at intervals of 15 deg. Subjects judged the hue by rating the relative amount of red, green, blue, or yellow. These ratings were then converted into a hue angle within a perceptual opponent space defined by the pure red–green (0–180 deg) and blue–yellow (90–270 deg) axes. For example, a stimulus that was rated three parts blue and two parts red would have an angle of $\tan^{-1}(3/2) = 56.3$ deg within the perceptual red vs. green and blue vs. yellow space. Figure 3(a) shows the relationship between the stimulus angle in the cone-opponent space and the average perceptual hue angle for the observers. On separate trials each observer also la-

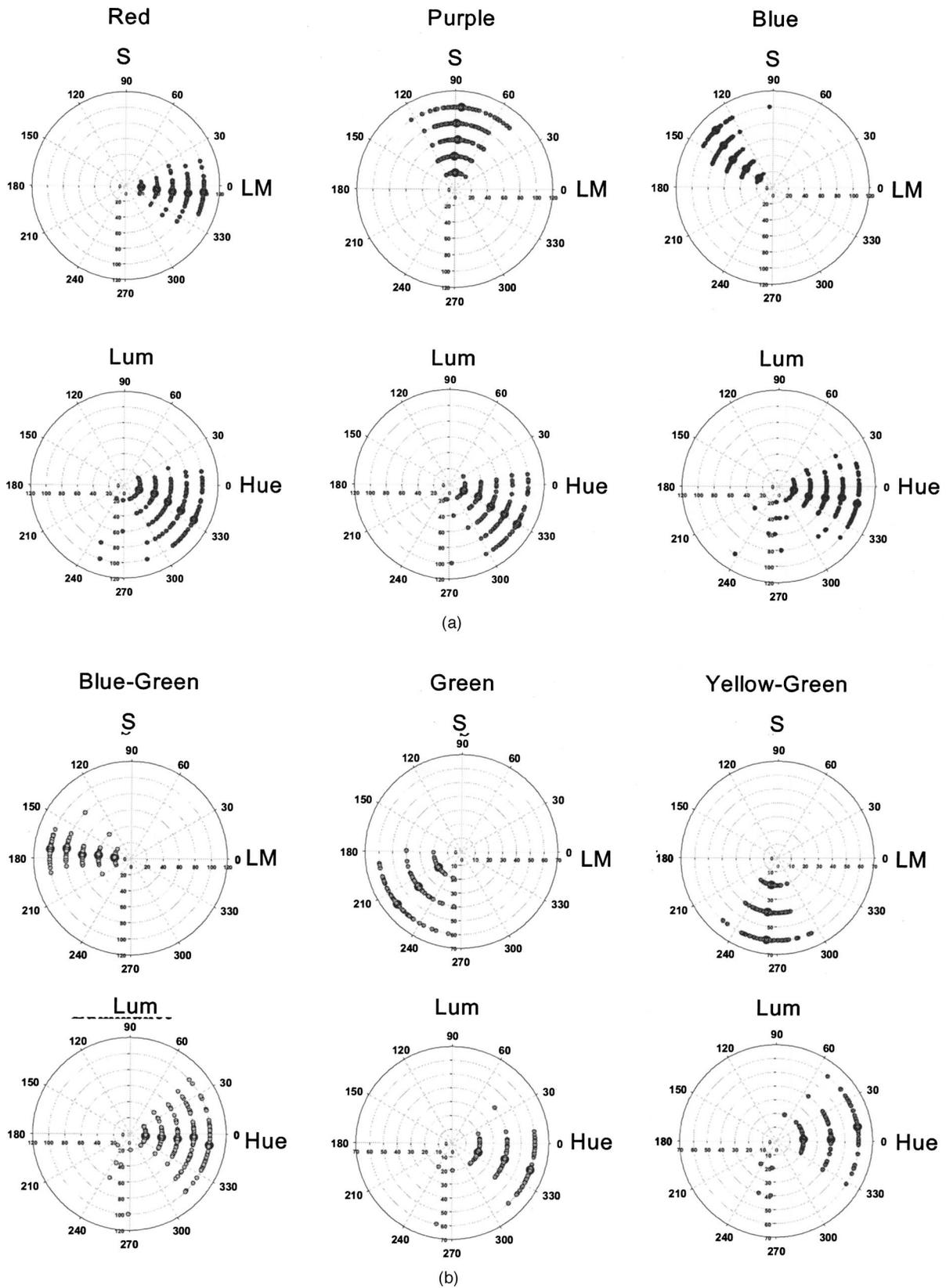
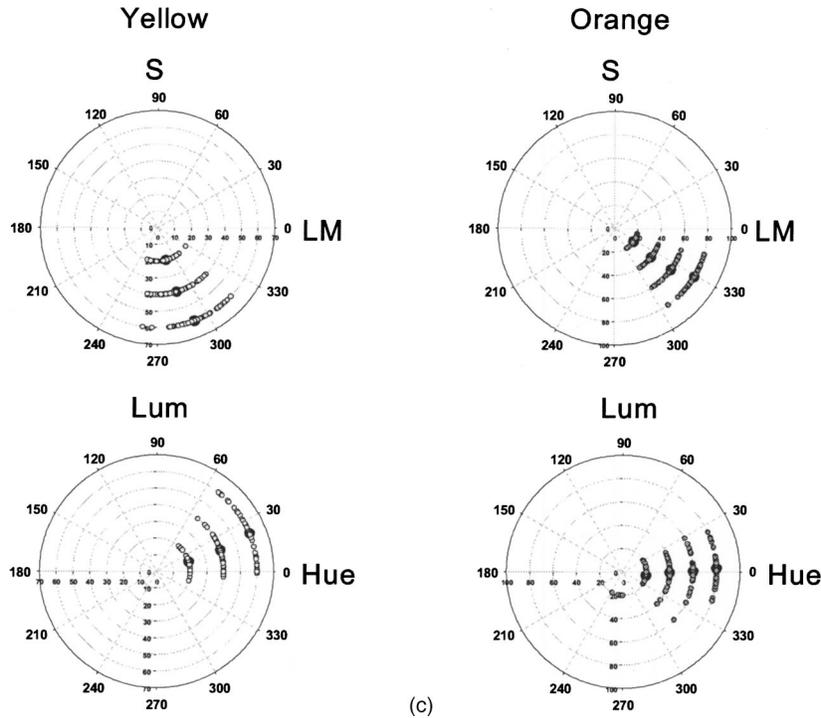


Fig. 2. Individual settings for the hue and lightness of each of the eight color terms. For each term, the top panel plots the selected hue angle projected onto the isoluminant plane (i.e., independent of the subject's lightness setting), while the bottom panel shows the elevation out of the isoluminant plane (i.e., independent of the subject's selected hue angle). Settings for stimuli of increasing contrast are plotted along circles of increasing radii. (Continues on next page.)



(c) Fig. 2. (Continued).

Table 5. Correlations between Hue Angles (below Diagonal) and Lightness Levels (above Diagonal) for Different Color Terms

	Lightness							
	R	P	B	B-G	G	Y-G	Y	O
R		0.02	0.11	0.12	0.21	-0.09	-0.06	0.09
P	0.15		0.30*	0.34*	0.17	0.08	-0.06	0.30*
B	-0.03	0.01		0.56*	0.29*	0.20	0.13	0.28*
B-G	0.11	-0.14	0.23		0.34*	0.10	0.10	0.36*
G	-0.12	-0.24	-0.02	0.14		0.16	0.09	0.49*
Y-G	0.07	-0.08	0.08	-0.02	0.26		0.21	0.29*
Y	-0.06	0.10	0.14	-0.16	-0.01	0.40*		0.13
O	-0.06	0.16	-0.03	-0.11	0.31*	0.10	0.48*	
	Hue							

beled the stimulus with one of the eight color terms. The distribution of these labels is shown in Fig. 4.

Not surprisingly, the ratings in the hue scaling task are qualitatively consistent with how stimuli were selected in the focal color task. It is again interesting to ask how these ratings are related to the cone-opponent axes used to define the stimuli. In Fig. 3(a) the arrows mark the intersection of each pole of the cardinal axes with the nominal perceptual axis (the four unique hues or the equal binary mixtures of these hues) that was nearest to the scaled hue. As before, the +L axis falls close to unique red, while the remaining cone-opponent axes lie close to the binary axes. (A similar pattern can be seen in the results of De Valois *et al.*⁴⁹) That is, the -L pole was, on average, rated as nearly an equal mixture of blue and green, while the -S pole was a balanced mixture of green and yellow. Thus, like the focal choices, these results point to a relationship between the structure of cone-opponent space

and the structure of color appearance (a relationship that is again very loose because of the large individual differences). Three of the cone-opponent directions therefore represent boundaries between the unique hue axes (e.g., whether a stimulus is more green or more blue), while the fourth is “unique” in that it is aligned with the red primary.

As with the focal color settings, subjects also varied widely in the hue scaling judgments. For the settings converted to angles in the RG–BY space, standard deviations for the individual stimuli ranged from 6 to 22 deg (mean = 14 deg). We asked whether the variation in the range for different stimuli might be predicted from the rate at which perceived color varies in different regions of color space. As Fig. 3(a) shows, perceived color as determined by the scaling task changes rapidly for stimuli moving from yellow to red, changing more slowly for transitions from red to purple. If individual differences in the ratings

Table 6. Correlations among the Hue Angles and Lightness Levels Chosen for Focal Red across Different Contrast Levels

Angle	Contrast 40	Contrast 60	Contrast 80	Contrast 100
Hue				
Contrast 20	0.62*	0.58*	0.88*	0.70*
Contrast 40		0.77*	0.59*	0.62*
Contrast 60			0.58*	0.71*
Contrast 80				0.73*
Lightness				
Contrast 20	0.57*	0.53*	0.64*	0.47*
Contrast 40		0.42*	0.49*	0.47*
Contrast 60			0.59*	0.42*
Contrast 80				0.60*

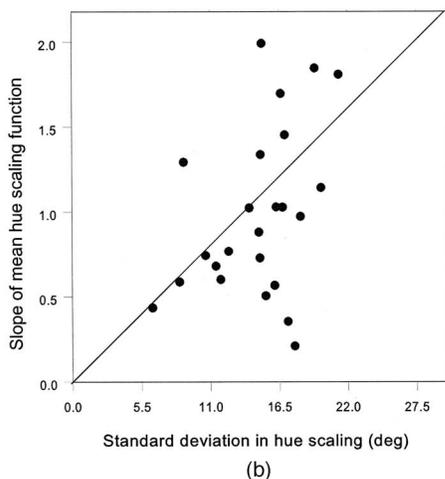
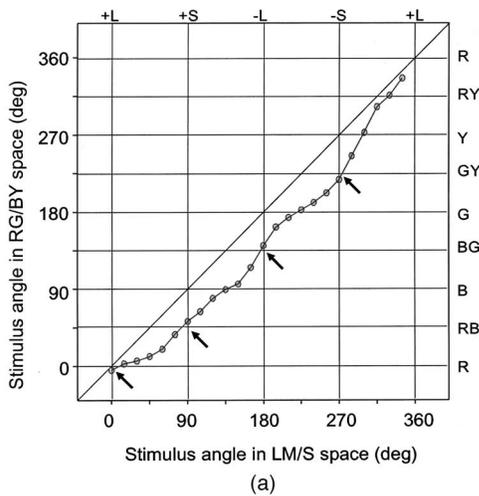


Fig. 3. (a) Average hue scaling function. Points plot the judged angle in a red–green versus blue–yellow perceptual color space as a function of the stimulus angle in the LvsM and SvsLM plane. Arrows point to the perceived hues of stimuli lying along the cardinal axes and the closest unique or binary color term. (b) Relationship between individual differences in the hue scaling and the local slope of the average hue scaling function.

reflected a fixed range of perceptual color difference, then this range should be related to the local slope of the hue

scaling function. These slopes were estimated from a polynomial fit to the mean hue scaling curve. Figure 3(b) compares the standard deviations in the ratings for each of the 24 stimuli (again with the ratings expressed as angles in the perceptual space) with the slope of the hue scaling function at each stimulus angle. There is little relationship between the two values, indicating that the variance in judgments probably does not depend on the salience of color differences in different regions of the space. This conclusion is also consistent with the analysis above showing that large differences in the ranges for different color terms remain when the stimuli are represented in a uniform color space such as $u'v'$ (Table 2).

The correlations between the ratings for the different chromatic angles are shown in Table 7. It is clear that there tend to be strong correlations between nearby hue angles yet only weak relationships between more distant angles. Thus the variations in each hue again depend on relatively local factors. This is further seen in Table 8, which reproduces the values for the eight angles closest to the foci for the eight color terms. These were determined from the modal values in the distributions of color labels in Fig. 4. Like the results for focal choices, few of the color terms are significantly correlated (though once again orange emerges as a possible exception). In the case of the hue scaling, this is all the more surprising, because subjects could rate the stimuli only in terms of the four unique hues, yet the variations in scaling binary hues like purple did not depend on how subjects differed in scaling red or blue.

One possible basis for this pattern of local correlations is that each hue angle covaries consistently only with its nearby neighbors. However, there is instead a discrete clustering of the correlations. To visualize this, we calculated for each stimulus the “center of mass” of its correlation coefficients, given by averaging the stimulus angles weighted by the coefficients. For these averages we used only coefficients that were significant and positive. The mean angles for each cluster are plotted as a function of the stimulus angle in Figs. 5(a) (for all subjects) and 5(b) (for the 30 most consistent subjects, whose repeated settings varied less than the median variance for all subjects). If the local correlations were centered at each stimulus angle, then these clusters would vary continuously and fall along the diagonal of the figure. Instead, there are clear steps, especially for the subset of consistent observers. One of these steps is centered on the +L pole of the LvsM axis and includes a wide span of stimulus angles ranging from -45 deg (orange) to 60 deg (red-dish purple). Over this span subjects differed consistently from each other in how they scaled the stimuli, while settings for neighboring stimuli just outside this cluster resulted in a new pattern of individual differences. Weaker clusters are also evident near 180 deg, the opposite pole of the LvsM axis, and at 270 deg, the +S pole of the SvsLM axis.

Recall again that in the hue scaling task the subjects were restricted to using four color terms (the perceived amounts of red, green, blue, and yellow). Thus it is possible that the clustering simply reflects how subjects weighted the independent primaries. However, predic-

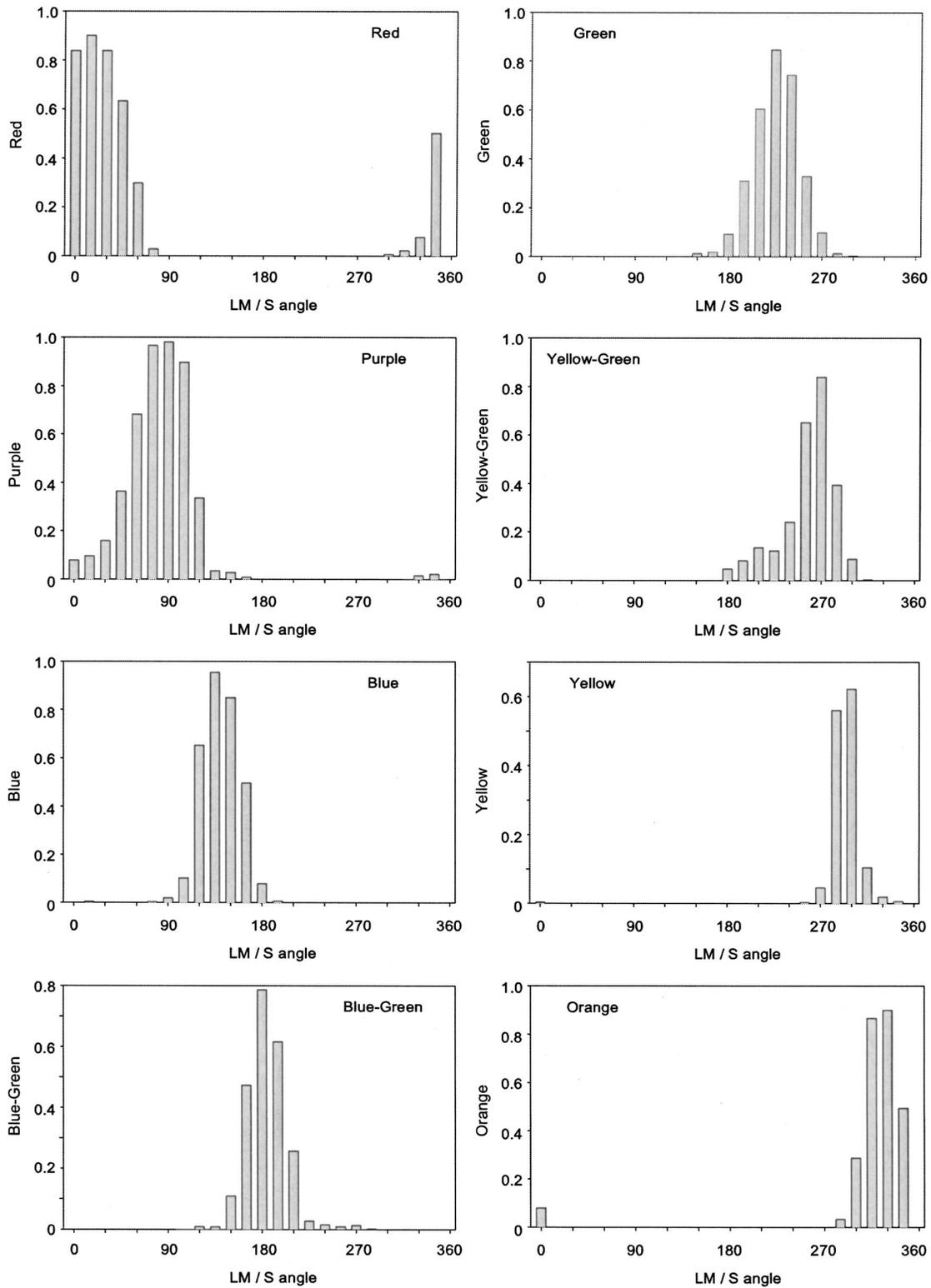


Fig. 4. Distribution of color labels for the 24 stimuli used in the hue scaling task. Each panel shows the number of times subjects chose a given color term as the label for the stimulus. The eight panels show results for the four unique hue terms or the four binary terms.

tions based on scaling or rotating the mean hue scaling response to red, green, blue, and yellow failed to fit the observed pattern of correlations or of factors derived from a factor analysis of the correlation matrix. Thus we are

uncertain of the basis for the clustering. Yet whatever its basis, the individual differences in hue scaling do not appear tied to differences in the relative strength or direction of mechanisms tuned to the unique hue directions.

Table 7. Correlations between the Rated Hues for Each of the 24 Stimuli in the Hue Scaling Task

	0	15	30	45	60	75	90	105	120	135	150	165	180	195	210	225	240	255	270	285	300	315	330	345
0	0.55*	0.82*	0.75*	0.57*	0.60*	0.17	-0.07	0.00	-0.14	-0.13	-0.15	-0.03	-0.22	-0.15	-0.25	0.17	0.07	0.06	-0.31*	-0.29*	0.17	0.52*	0.50*	0.70*
15		0.72*	0.86*	0.62*	0.66*	0.40*	0.01	-0.11	-0.14	0.02	-0.06	0.00	-0.21	-0.22	-0.30*	0.11	0.06	0.07	-0.22	-0.19	0.15	0.37*	0.30*	0.45*
30			0.65*	0.75*	0.75*	0.55*	0.06	-0.04	-0.14	0.10	0.09	0.06	-0.25	-0.21	-0.25	0.19	0.03	-0.01	-0.35*	-0.28*	0.11	0.43*	0.40*	0.45*
45				0.68*	0.80*	0.55*	0.12	0.10	-0.14	0.00	0.21	0.09	-0.22	-0.26	-0.22	0.23	0.04	0.02	-0.17	-0.20	0.25	0.36*	0.29*	0.37*
60					0.82*	0.67*	0.17	0.03	-0.24	-0.07	0.07	-0.03	-0.31*	-0.25	-0.23	0.14	0.05	0.03	-0.26	-0.30*	0.29*	0.45*	0.39*	0.49*
75						0.60*	0.30*	-0.01	-0.12	-0.03	0.08	0.07	-0.31*	-0.38*	-0.31*	-0.03	-0.15	-0.10	-0.26	-0.28*	0.11	0.21	0.12	0.05
90							0.27*	0.31*	0.21	0.05	0.25	0.09	-0.19	-0.15	0.01	-0.13	0.11	0.18	-0.02	-0.23	-0.04	0.02	-0.08	-0.03
105								0.41*	0.56*	0.03	0.22	0.07	0.09	-0.06	0.05	-0.01	0.11	0.13	0.04	0.04	0.07	0.13	0.15	0.14
120									0.51*	0.36*	0.34*	0.03	0.18	0.02	0.22	-0.11	-0.05	0.01	0.02	0.03	-0.13	-0.12	-0.02	0.02
135										0.25*	0.58*	0.14	0.08	0.04	0.19	0.06	0.18	0.02	-0.02	-0.16	-0.16	-0.07	-0.18	-0.11
150											0.40*	0.43*	0.24	0.11	0.24	0.20	0.23	0.02	-0.03	-0.20	-0.05	0.04	-0.24	-0.07
165												0.75*	0.62*	0.44*	0.39*	0.17	0.16	-0.09	-0.25	-0.26	0.01	0.07	-0.13	-0.01
180													0.62*	0.73*	0.47*	0.22	0.17	-0.07	0.03	0.12	0.04	-0.18	-0.28	-0.17
195														0.60*	0.62*	0.42*	0.24	-0.02	0.00	0.09	0.02	-0.15	-0.01	0.00
210															0.33*	0.51*	0.44*	0.18	0.02	-0.03	0.05	-0.15	0.07	0.05
225																0.30*	0.54*	0.29*	0.22	0.13	0.23	0.05	0.16	0.14
240																	0.47*	0.68*	0.33*	0.14	0.29*	0.07	0.02	0.13
255																		0.69*	0.57*	0.35*	0.13	0.01	0.07	0.03
270																			0.68*	0.66*	0.06	-0.42*	-0.37*	-0.39*
285																				0.71*	0.23	-0.25	-0.14	-0.22
300																					0.59*	0.36*	0.34*	0.38*
315																						0.69*	0.63*	0.62*
330																							0.49*	0.77*
345																								0.68*

Table 8. Correlations between Scaled Hues for Stimuli That Fell Closest to Each of the Four Unique Hues or Four Binary Colors

		R	P	B	B-G	G	Y-G	Y	O
		0	90	135	180	225	270	300	330
All subjects									
R	0	0.55*	-0.07	-0.13	-0.22	0.17	0.31*	0.17	0.50*
P	90		0.27*	0.05	-0.19	-0.13	-0.02	-0.04	-0.08
B	135			0.25*	0.08	0.06	0.02	-0.16	-0.18
B-G	180				0.62*	0.22	0.03	0.04	0.28*
G	225					0.30*	0.22	0.23	0.16
Y-G	270						0.68*	0.06	0.37*
Y	300							0.59*	0.34*
O	330								0.49*
Most consistent subjects									
R	0	0.89*	-0.04	-0.16	-0.42*	0.01	-0.45*	0.0	0.68*
P	90		0.62*	-0.06	-0.18	-0.13	-0.22	-0.07	-0.02
B	135			0.42*	0.06	0.26	0.11	-0.06	-0.05
B-G	180				0.80*	0.28	0.21	-0.13	0.49*
G	225					0.52*	0.31	0.20	0.10
Y-G	270						0.83*	0.14	-0.49*
Y	300							0.78*	0.30
O	330								0.81*

4. DISCUSSION

The processes underlying subjective color experience, and how they are derived from the opponent organization at early postreceptoral stages of the visual system, remain very poorly understood. The present results bear on two general questions about the structure of color appearance. First, they allowed us to examine whether judgments about color follow directly from how observers judge the red–green and blue–yellow dimensions that are assumed to underlie color appearance. Second, they provide a measure of the relationships between color appearance and the early cone-opponent axes.

In previous studies we found that the variations among individuals in the stimuli selected for unique hues are nearly independent, suggesting that each unique hue is controlled by independent factors.¹⁰ This is consistent with evidence showing little relationship between the unique hue settings and variations in visual sensitivity^{29,32–35} and with several studies indicating that the different poles of the color-opponent axes are mediated by separate processes.^{16,46,50–54} The present work tested whether the four independent processes coding the primary hues could account for how observers judged binary mixtures of the hues. Surprisingly, we again found that the stimuli selected for these binary hues were largely independent of the observers' unique hue settings and that consensus among observers was comparable for the unique hues and the binary hues. Thus by these specific criteria, there is little to distinguish between the unique and the binary hues and, in particular, little evidence that the binary settings reflect color judgments that are derived directly from underlying red–green and blue–yellow responses. Moreover, it suggests that the variations in color judgments depend on local factors, perhaps varying independently within each color category, rather

than global factors such as differences in sensitivity at peripheral stages of the visual system.¹⁰ For example, if there are indeed distinct neural processes tuned to each basic color term,²⁶ then our results suggest that the factors contributing to individual differences in these processes are largely category-specific. Similarly, if the terms instead reflect properties of the environment rather than the observer, such as the distribution of colors in the environment,⁵⁵ then the independence we found would suggest that the distributions for different clusters either vary or are learned in category-specific ways.

The one exception to this pattern was for orange, which correlated consistently with the yellow and, to a lesser degree, the red settings. Within our cone-opponent space, orange, yellow, and red lie close together, with an average separation of 50 deg in their hue angles. Thus the tendency for orange to covary with yellow may partly reflect the closer proximity to its primaries compared with other binary hues. Yet even for orange, the correlation with yellow or red was modest and not present for all conditions, and thus the orange settings were not strongly determined by the individual's red and yellow foci.

For purple, which like orange corresponds to a basic color term, the foci appeared much less tied to the foci for the blue and the red component colors. Thus in terms of the individual variation, purple appeared to behave like a distinct category. It is possible that this is because purple is far removed from other focal colors in the space. (Similarly, in Munsell space the hue circle is divided into roughly five equal arcs, with the four unique hues and purple as principal hues.⁵⁶) The independence of the purple category means that while it may be possible to perceptually decompose a purple into red and blue,⁷ the "red" that is contributing to purple may not be the same "red" that is mediating judgments within the red color

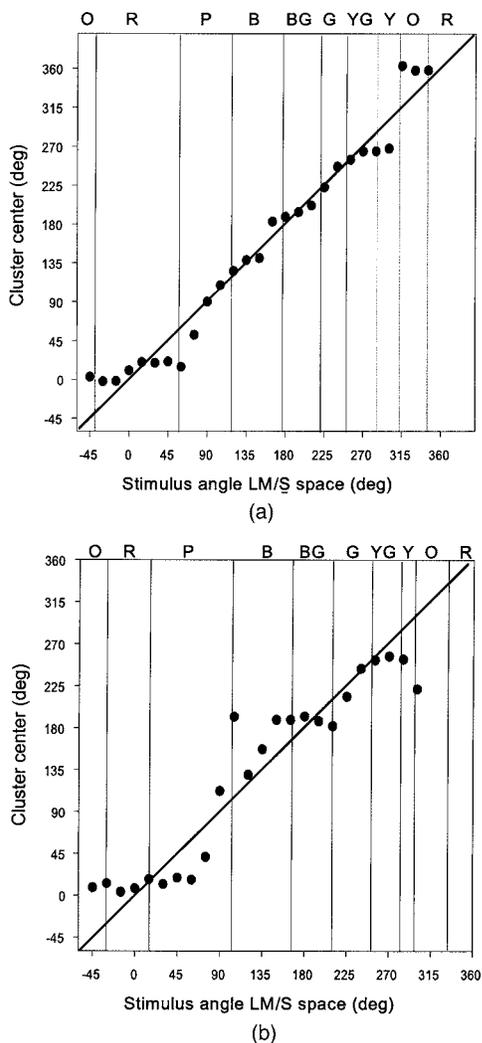


Fig. 5. Clustering in the correlations between the scaled hues for different stimuli. Clusters for each stimulus angle were calculated by averaging the stimulus angles weighted by the correlation coefficients (excluding nonsignificant or negative coefficients). (a) All subjects, (b) most consistent subjects.

category, since the purple and red foci vary in independent ways. This point is illustrated most clearly for the hue scaling results, where settings for purple (e.g., stimuli at 75 or 90 deg) varied independently of the settings for neighboring bluish-reds (e.g., 45 or 60 deg).

It may seem paradoxical that binary hues could be independent of the red–green and blue–yellow primaries, since the former are defined in terms of the latter. In particular, the settings for blue–green and yellow–green seem very likely *a priori* to reflect judgments about the boundaries between the primary color categories. In this case the independence we found may mean only that the category boundaries do not vary systematically with the category foci. In this regard, purple may be like the yellow–green and blue–green terms because it represents the border between red and blue. Our analysis tested only for the relationships between focal choices and thus does not exclude a relationship between the unique and binary hues based on other factors, such as the underlying spectral sensitivities of the opponent color dimensions, or the

possibility that the “rules” for defining the color boundaries vary independently of the rules for the category foci. If the shapes of the perceptual spectral sensitivities can vary in ways that are not tied to the focal choices, then it is not necessary that mixture hues covary with the foci. The extent to which these spectral sensitivities are nonlinear, and the conditions under which these nonlinearities are manifest, remain unclear.^{51,52,57–60} However, individual differences in color appearance based on different linear cone combinations predict stronger dependence for the binary hues than we observed, for in that case the spectral sensitivities are completely determined by and covary with the focal direction.

The factors shaping the location of different color categories remain unknown. It is well established that the red–green and blue–yellow dimensions of color appearance are not the dimensions along which chromatic information is encoded at early postreceptoral levels. For example, cells in the lateral geniculate do not show the return of a “red” response at short wavelengths that is predicted by the “red–green” perceptual channel.⁶¹ Instead, retinal and geniculate cells are tuned to stimulus variations along the LvsM and SvsLM cardinal axes,⁴² and psychophysical measures of sensitivity and adaptation similarly point to an organization in terms of these axes.⁴ Zone models of color coding have illustrated the types of transformations that could convert from early postreceptoral to the perceptual axes.^{62–64} Yet whether such transformations occur or are even necessary in principle and whether cortical color coding might instead involve very different representations (for example in terms of multiple chromatic channels) is still debated, because neural mechanisms that correspond clearly to the red–green and blue–yellow axes have yet to be identified.^{49,65,66} Moreover, evidence for these transformations would still leave unanswered the question of why the unique hues are oriented along particular axes. One answer to this question has been that the unique hues are special because they reflect special properties of the environment. For example, previous authors have pointed out that the blue–yellow axis falls close to the daylight locus and have suggested that unique yellow might reflect a normalization of the L- and M-cone responses to the average color in the observer’s environment.^{34,36,37,67} Similarly, Yendrikhovskij⁵⁵ has recently argued that the foci and relative salience of basic color terms could be predicted from how the color characteristics of natural images are clustered in the volume of color space. By these accounts, then, the location of the unique hues is shaped by salient properties of the environment.

Of the four unique hues, only red falls close to one of the cardinal axes,^{10,46,47} and in color naming red emerges as an earlier and more robust dimension than other hues.^{18–20} What salient property of the world might it signal? Recently, Webster and Kay²² suggested that both the special prominence of red as a color term and the fact that it lies near the +L axis might be related to the fact that the LvsM axis is believed to have evolved for discriminating edible fruits and leaves from the background foliage.^{68–70} Thus red might be special because it behaves in some sense like a trigger feature for ripeness, the spectral stimulus that drove the evolution of primate trichro-

macy. This account suggests a close functional connection between the cardinal axes and unique red and thus between the structure of cone-opponent and perceptual color space. On the other hand, it is clear that this connection must be a loose one, for individuals vary widely in unique red and far more than they vary in the stimulus direction isolating the LvsM axis.^{10,45,48}

The remaining unique hues are oriented along directions intermediate to the cardinal axes and thus cannot be tied in a similar way to isolated activity in one of these axes. This has provided compelling evidence for the dissociation between the dimensions of color appearance and precortical color organization and has tended to imply that the cardinal axes may impose little constraint on color categories. However, we found that the foci for binary hues do tend to fall near the cardinal axes. That is, focal settings for blue-green clustered nearly as close to one pole of the LvsM axis as the unique red settings did to the other, and similarly, the hue scaling functions showed that the $-L$ axis is very close to the stimulus that appears to be an equal mixture of blue and green. While this leaves open the question of what determines the best examples of blue and green, it raises the possibility that the boundary that partitions these categories may in part be related to properties of the representation of color at early postreceptoral levels. In the same way, the yellow-green boundary that divides hues into more yellow or more green fell close to the $-S$ axis, while purples fell near the $+S$ axis (though in the case of purple the average deviates significantly from this axis, and any connection between the S axis and the red-blue boundary is thus more tenuous). These effects were also mirrored in how subjects varied in their hue scaling. The differences among individuals appeared to reflect local variations that were roughly centered around the cardinal axes and thus again around differences in red on the one hand and in the binary hues of blue-green and yellow-green (and perhaps purple) on the other. (This asymmetry suggests another possible basis for the special prominence of red, since it is the only unique hue aligned with, and possibly more directly following from, a cardinal axis.)

A connection between the unique hue boundaries and the cardinal axes is consistent with a model of color coding proposed by De Valois and De Valois.⁶³ They suggested that the cardinal axes are recombined in the cortex to form channels tuned to the unique hues. Specifically, in their model, inputs from the SvsLM dimension are used to modulate the signals from the LvsM dimension, rotating the hue mechanisms either clockwise or counterclockwise off the LvsM axis depending on the sign of the S input. This model fails to account for the average locus of unique red we found in the present study and in previous studies,^{10,31,46} since again this locus remains very close to the LvsM axis (though for their small samples a bias toward $+S$ was found^{16,49}), but is qualitatively consistent with the loci for blue, green, and yellow. Moreover, a notable feature of their proposal is that the cardinal axes provide the reference frame for repartitioning color space and thus could account for our findings that the boundaries separating some of the unique hue categories tend to lie along the cardinal axes. As before, it is important to emphasize that any consideration of the average stimulus

angles for a color term must be tempered by the enormous variations across observers, so that any connection between an individual's color judgments and the cardinal axes must be a weak one. Yet in any case, such results point to possible ties between the dimensions defining color appearance and early color coding.

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