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# Measurement of individual color space using a luminous vector field

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This study is intended to measure the geometry of the observer's color space when viewing a computer screen and to define individual variations from this data. A CIE photometric standard observer assumes that the eye's spectral efficiency function is constant, and photometry measurements correspond to vectors with fixed directions. By definition, the standard observer decomposes color space into planar surfaces of constant luminance. Using heterochromatic photometry with a minimum motion stimulus, we systematically measure the direction of luminous vectors for many observers and many color points. During the measurement process, the background and stimulus modulation averages are fixed to the given points in order to ensure that the observer is in a fixed adaptation mode. Our measurements result in a vector field or set of vectors  $(x, v)$ , where  $x$  is the point's color space position, and  $v$  is the observer's luminosity vector. In order to estimate surfaces from vector fields, two mathematical hypotheses were used: (1) that surfaces are quadratic or equivalently that the vector field model is affine, and (2) that the metric of surfaces is proportional to a visual origin. Across twenty-four observers, we found that vector fields are convergent and the corresponding surfaces are hyperbolic. The equation of the surface in the display's color space coordinate system, and in particular the axis of symmetry, varied systematically from individual to individual. A hyperbolic geometry is compatible with studies that emphasize a modification of the photometric vector with changing adaptations. © 2023 Optical Society of America

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## 1. INTRODUCTION

Human color perception is mediated by three types of retinal light-sensitive cells, the cone-photoreceptors called L, M or S. In order to generate the input required for luminance and motion perception, photoreceptor signals are linearly combined [1]. There is a functional separation between this input and the input for color perception at the level of retinal ganglion cells [2]. Due to the linearity of the luminance channel, it is anticipated that the human eye will exhibit a unique spectral efficiency function.

The spectral efficiency functions of human observers have been measured using heterochromatic flicker photometry [3] and correspond to a linear combination of L- and M- cone spectral sensitivity only (without S-cone input) at a fixed adaptation state [4]. There was agreement among several observers that led to the establishment of a standard by the commission internationale de l'éclairage (CIE), CIEV( $\lambda$ )-1924. These measures were confirmed by different kinds of experiments including minimal distinct

border, minimum motion, or test of the additivity principle known as Abney's law, for which the luminance of a compound light is equal to the sum of the luminances of its composing lights [5, 6]. However, several studies have suggested that the spectral efficiency function may be modulated by an observer's adaptation to the level of light or to a particular chromatic direction in the color space [4, 7].

In any given color space, such as a display's color space, the consistency of the efficiency function imposes a specific geometry on the standard observer. As will be demonstrated in section 2,  $V(\lambda)$  can be represented by a vector in the color space of a display. The vector defined by  $V(\lambda)$  is the same at any color point. As a result, all the vectors are parallel. The color points with equal luminance also lie on an iso-luminance plane orthogonal to the vectors. From a standard observer's perspective, photometric measurement by the visual system corresponds to the division of a color space into a pile of iso-luminance planes. However, because the spectral efficiency function may vary with

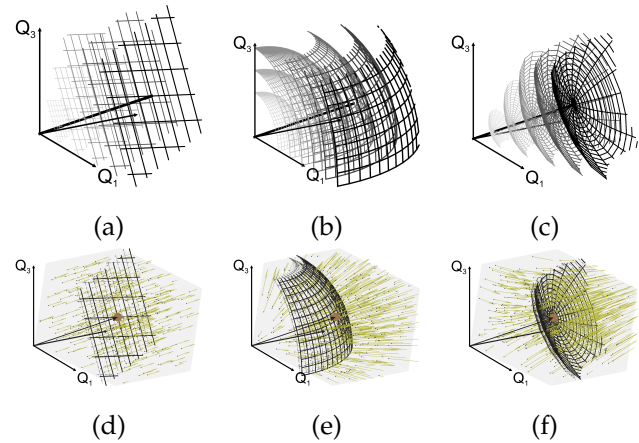
the observer's adaptation state, the iso-luminant surface could have a different geometry.

We define a luminous [8] vector field as a set of  $(x, v)$  pairs where  $x$  represents the coordinates of a color point  $p$  in color space and  $v$  represents the coordinates of the local luminous vector orthogonal to the iso-luminous surface at that point. We designed an experiment to estimate the direction and magnitude of the observer's luminous vector at several points in the display's color space. Our goal was to determine whether the direction of the observer's luminous vector corresponded to the direction given by  $V(\lambda)$ , forming a field of parallel vectors. A deviation from  $V(\lambda)$  would indicate broken linearity, as well as deformation of the iso-luminous surface by the visual system.

The iso-luminous surface of an observer can be reconstructed by predicting  $v$  from  $x$  using a vector field model. In general, it is difficult to reconstruct the surface from the vector field. To simplify our mathematical analysis, we used an affine model of the form  $v = 2Ax + a_0$  that allows only quadratic surfaces with constant curvature. Possible quadratic surfaces include planes (if degenerated), ellipses, and hyperboloids.

The constant  $a_0$  represents the origin of the observer's visual system in the affine model. The application of proportionality toward this visual origin enables a unique correspondence between the measured vector field and the observer's iso-luminous surface for all points in color space. When considering the proportionality between surfaces, the origin and a single surface define the color space metric as a whole. Whatever type of iso-luminous surface is used, be it planar Figure 1(a), elliptical Figure 1(b) or hyperbolic Figure 1(c), they decompose color space into slices of iso-luminous surfaces. The surface corresponds to a plane when the vector field is constant (Figure 1(d)), an ellipsoid if the vector field is divergent (Figure 1(e)), or a hyperboloid if the vector field is convergent (Figure 1(f)).

In our experiment, we measured the luminous vector  $v$  at several color points of coordinate  $x$  in order to estimate the observer's luminous vector field (i.e. a set of  $(x, v)$  pairs). For the purpose of enabling orthogonality, we associated the display color space with a three-dimensional Euclidean space. There is a need in our study for having an orthogonal coordinate system for the display's color space. We will express observer's model in the display's coordinate system. If it is not orthogonal, the metric of the observer will be multiplied by the correlation metric of the display. In addition, we will use orthogonality between vectors or vectors to surfaces to design the minimum motion stimulus. This orthogonality will be simplified in Euclidean space. A variation of the minimum motion experiment [5] has been developed to estimate  $v$  at a fixed point  $p$  of coordinate  $x$ . In this experiment, we constrain the observer adaptation state to the given point  $p$  by covering the background of the display with the color  $p$ . The three-dimensional coordinates of  $v$  at a given point were then derived by using two sessions of minimum motion



**Fig. 1. Projective geometry.** The color space is decomposed by the visual system into a pile of iso-luminous surfaces proportional to each other towards an origin. It is possible to reconstruct the iso-luminous surface from a vector field by measuring a set of couples  $(x, v)$ , where  $v$  represents the photometric luminous direction for the observer at a particular point  $p$  of coordinate  $x$  under the assumption of quadratic surfaces and proportionality.

in two orthogonal planes. From the measured vector field we reconstruct the iso-luminous surfaces corresponding to the color space of the observer.

In the following section we will detail the geometry of the standard photometric observer and demonstrate that it engenders a decomposition of the color space into slices of iso-luminant planes because the spectral efficiency function  $V(\lambda)$  is considered fixed. Then we describe the whole procedure for conducting the experiment and analyzing the data.

## 2. GEOMETRY OF THE STANDARD PHOTOMETRIC OBSERVER

We call standard photometric observer, hypothetical observer whose spectral efficiency function would be the one defined by the standard CIE  $V(\lambda)$ -1924 and represented in the color space by a vector  $V_\lambda$ . Standard observer photometric measurements can be considered as the action of the vector  $V_\lambda$  on points  $p$  [9]. This action results in measuring the luminance value of the color point  $p$ . Luminance value  $L$  of a light  $p$  produced by the display is defined by the integral of the product between spectrum of light  $p(\lambda)$  and relative spectral efficiency function of the standard photometric observer  $V_\lambda$ , by the following formula [10]:

$$\text{Luminance}(p) = \int_{\lambda_m}^{\lambda_M} V(\lambda)p(\lambda)d\lambda = L \in \mathbb{R}^+, \quad (1)$$

where  $[\lambda_m, \lambda_M]$  indicates the wavelength interval for visible light.

Several authors have shown that the spectral function space of light corresponds to a real Hilbert space [11],

which we call the  $\mathcal{L}$  space. It is particularly true for the set of possible lights emitted by a computer screen. This is because they can be expressed as a linear combination of the primaries of the display phosphors. The integral of Equation 1 can be viewed as a scalar product in  $\mathcal{L}$ . Taking  $|f\rangle$  as the vector corresponding to the spectral functions  $f(\lambda)$  [12], the real Hilbert space of spectral functions of light  $\mathcal{L}$  can be defined with [13]:

$$\text{Linearity : } a|f\rangle + b|g\rangle = |af + bg\rangle \quad (2)$$

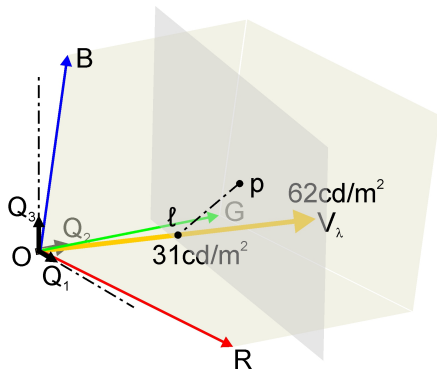
$$a, b \in \mathbb{R}, |f\rangle, |g\rangle, |af + bg\rangle \in \mathcal{L}$$

$$\text{Duality : } \langle f| = |f\rangle^t, |f\rangle \in \mathcal{L}, \langle f| \in \mathcal{L}^* \quad (3)$$

$$\text{Scal. Prod. : } \langle f|g\rangle = \int f(\lambda)g(\lambda)d\lambda \in \mathbb{R} \quad (4)$$

$$\text{Eucl. Norm. : } \||f\rangle\| = \sqrt{\langle f|f\rangle} \in \mathbb{R}^+ \quad (5)$$

Luminance value of a light  $p$  with spectrum  $p(\lambda)$  is given by the scalar product:  $\text{Luminance}(p) = \langle V|p\rangle$ , where  $\langle V|$  is the dual vector corresponding to luminous efficiency of the eye,  $V(\lambda)$ .



**Fig. 2. Standard photometric observer geometry.** Display color space is defined as a parallelogram  $p_0RGB$  ( $p_0$  is the black level, not shown here, see supplementary method S1-SA), mapped into a Euclidean space with basis  $OQ_1Q_2Q_3$ . Using this color space, spectral efficiency function of the eye  $V(\lambda)$  corresponds to a vector  $V_\lambda$ . A point  $p$ 's luminance is determined by the intersection of the plane orthogonal to  $V_\lambda$  passing through  $p$  with the line  $(OV_\lambda)$ . A luminance value of  $p$  is defined as the affix  $\ell$  on the graduated line  $(OV_\lambda)$ . In the display we use,  $\|V_\lambda\|^2 = 62 \text{ cd/m}^2$ . It is the plane orthogonal to  $V_\lambda$  that represents the iso-luminance plane. The luminance of all points located on this plane is the same as that of  $p$ .

Geometry underlying standard photometric observer is based on the scalar product in  $\mathcal{L}$ . Call  $|P\rangle$  the matrix of size  $N \times 3$  containing the three display's primary emission spectra for  $N$  uniformly sampled wavelengths in the interval  $[\lambda_m, \lambda_M]$ . The orthogonal projection from spectral vector  $|x\rangle$  to point of coordinate  $x$  in three-dimensional display's color space is given by the linear map  $\alpha$  defined

by [14]:

$$\alpha: \mathcal{L} \rightarrow \mathbb{R}^3$$

$$|x\rangle \mapsto x = \alpha(|x\rangle) = \langle P|(P^tP)^{-\frac{1}{2}}|x\rangle = \langle Q|x\rangle \quad (6)$$

where  $Q$  is an orthogonal basis in which we may represent the primaries  $P$  by the vectors  $\alpha(|P\rangle)$ . Basis of the color space is given by  $\alpha(|Q\rangle) = \langle Q|Q\rangle = 1$ . This is the canonical basis of  $\mathbb{R}^3$  ( $1$  is the identity matrix of size  $3 \times 3$ ). Scalar product in  $\mathcal{L}$  corresponds to canonical Euclidean scalar product in  $\mathbb{R}^3$ :

$$\begin{aligned} x \cdot y &\stackrel{\text{def}}{=} x^t y = \alpha(|x\rangle)^t \alpha(|y\rangle) = (\langle Q|x\rangle)^t (\langle Q|y\rangle) \\ &= \langle x|Q\rangle \langle Q|y\rangle = \langle x|y\rangle \end{aligned} \quad (7)$$

Linear application  $\alpha$  defines then an orthogonal projection from  $\mathcal{L}$  to  $\mathbb{R}^3$ .

$V_\lambda$  is defined by  $V_\lambda = \alpha(|V\rangle)$  and is a fixed vector for the entire display's color space. An iso-luminant surface is a plane of level  $\kappa$  that is orthogonal to  $V_\lambda$ . An iso-luminant plane passing through  $p$  of coordinate  $x$  is defined by a set of points of coordinate  $y$  such that  $V_\lambda^t y = V_\lambda^t x$ . Coordinates of the point  $\ell$ , intersection between the plane orthogonal to  $V_\lambda$  passing through  $p$ , and the line  $(OV)$  is given by  $\ell = \kappa V_\lambda$  where  $\kappa = V_\lambda^t x / V_\lambda^t V_\lambda = L / \|V_\lambda\|^2$  is a factor of luminance compared to  $V_\lambda$  (For our display  $V_\lambda^t V_\lambda = \|V_\lambda\|^2 = 62 \text{ cd/m}^2$ ). Figure 2 illustrates colorimetric observer in the color space of our display.

Standard photometric observers do not adapt to the overall condition of the scene. Real observers do. Adaptation may result in a deformation of the vector field because the luminous vector will not remain constant. By relaxing the constraint of linearity for photometric measurements, one can use any surface shape for an iso-luminous surface. Restricting to proportional quadratic surfaces is a mathematical trick that allows the vector to be related to the point more easily. By doing so a projective formulation can be developed.

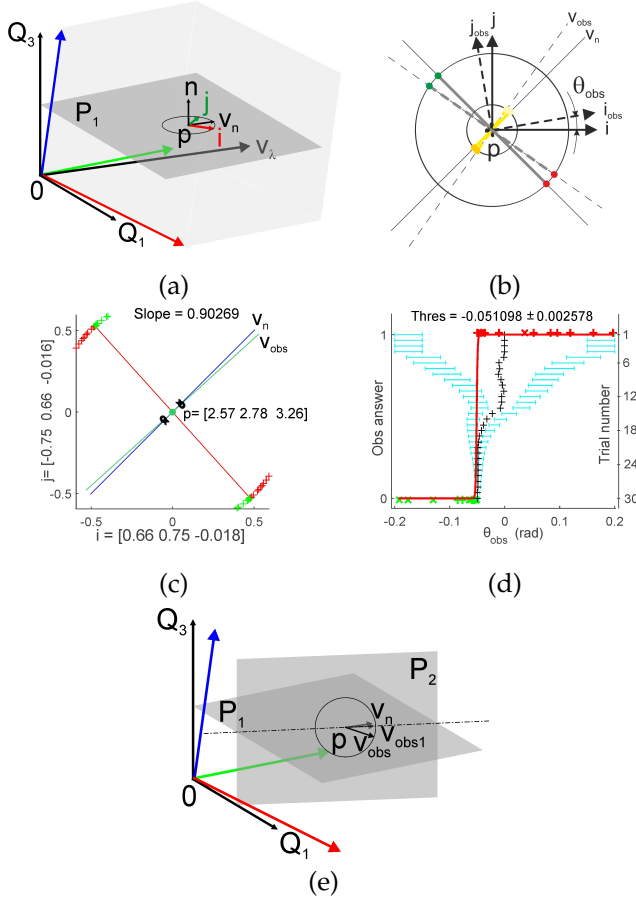
### 3. METHOD

The color space of the computer display was defined as Euclidean. Euclidean space is required for the definition of the observer's iso-luminous surface equation and for enabling orthogonality between vectors. For more information, please refer to supplementary method S1-SA. In this space, we are able to measure the luminous vector field using a minimum motion stimulus and compute the observer's iso-luminous surface based on the vector field.

#### A. Measuring luminous vector by heterochromatic photometry

We used a modified version of the minimum motion stimulus [5] in order to measure the luminous vector. Minimum motion stimulus consists of a spatial and color contrast modulated in time to produce motion. Motion perception occurs when the two colors that compose the stimulus are not of equal luminosity to the observer. Minimum

motion takes place in a plane containing heterochromatic modulation between the two colors and homochromatic modulation in an orthogonal direction. Minimum motion is intended to measure the difference in luminosity between two colors, but we have modified it to measure the iso-luminous line on a plane for a single color. The vector orthogonal to this line at the color point represents the projection of the luminous vector onto the plane.



**Fig. 3. Luminous vector estimation.** (a) Our first step in measuring the luminous vector at a point  $p$  is to define a plane  $P_1$  passing through  $p$ . In this plane, we construct a local coordinate system  $(p, i, j)$ . We rotate the local coordinate system around the normal vector  $n$  to the  $P_1$  plane to find the minimum motion angle for the observer that corresponds to the vector  $v_{obs1}$ . (b) The minimum motion stimulus is designed within the local coordinate system. (c) Illustration of the angular result for a typical session. (d) We use an adaptive procedure to estimate the angle for which motion is minimal to the observer (See supplementary method S1-SB). (e) We then define another plane  $P_2$  passing through  $p$ , orthogonal to  $P_1$  and collinear to the first estimate  $v_{obs1}$  in  $P_1$ , in which we estimate a second minimum motion angle using another local coordinate system. As a result, the vector  $v_{obs}$  in the second plane represents the luminous vector for the observer at point  $p$ .

Given a point  $p$ , we define a plane  $P_1$  passing through  $p$ ,

collinear with the vector  $Q_1 = (1, 0, 0)$  and  $v_n = V_\lambda / \|V_\lambda\|$  with  $V_\lambda = \alpha(|V|)$  corresponding to the standard observer's luminous efficiency vector  $V_\lambda$  normalized to 1 (Figure 3(a)). The vector  $v_n$  is an a priori for the observer's luminous vector. The angle difference between the observer's true luminous vector and  $v_n$  is measured using the minimum motion method. We defined an initial local coordinate system  $(p, i, j)$  where  $i$  and  $j$  are unitary orthogonal vectors ( $\|i\| = 1$ ,  $\|j\| = 1$  and  $i \cdot j = 0$ ) such that  $(i + j) / \sqrt{2}$  equals  $v_n$ . Homochromatic frames are defined as modulations centered on  $p$  and in the direction  $i + j$ , whereas heterochromatic frames are modulations in the direction  $i - j$  (Figure 3(b)). The heterochromatic and homochromatic modulations in plane  $P_1$  are respectively red-green and light-dark yellow. During the test we rotate the local coordinate system around the normal vector  $n$  to the  $P_1$  plane (Figure 3(c)) depending on the observer's response to find the minimum motion angle for the observer. The angle at which motion is minimum is estimated using an adaptive procedure (Figure 3(d)) (see Supplementary Method S1-SB). The vector corresponding to this angle is called  $v_{obs1}$ , which is the observer's luminous vector estimated at a point  $p$  in the plane  $P_1$ .

The minimum motion session is repeated for point  $p$  in plane  $P_2$ , orthogonal to  $P_1$  and collinear with  $v_{obs1}$  (Figure 3(e)). For heterochromatic and homochromatic frames, respectively, the modulations are blue-yellow and light-dark gray (the relative contribution of red and green to the gray depends on  $v_{obs1}$  estimation in  $P_1$ ). Because  $P_2$  is collinear to  $v_{obs1}$ , the result of the luminous vector estimation in the second plane directly defines the three-dimensional coordinates of the observer's luminous vector  $v_{obs}$  at point  $p$ .

## B. Stimulus

The minimum motion stimulus  $S(x, t)$  is defined by its spatial position on the screen  $x = [0, \dots, X - 1]$  with  $X = 192$  being the pixels extent, and its temporal variable  $t = [1, \dots, T]$  with  $T = 4$  being the number of frames composing the stimulus, according to the following equation:

$$\begin{aligned}
 S(x, t) = & p \\
 & + i \left( m \Pi(f_x x) \sin(2\pi f_t t) \right. \\
 & \quad \left. - d R \Pi(f_x x + 1/4) \cos(2\pi f_t t) \right) \\
 & + j \left( m \Pi(f_x x) \sin(2\pi f_t t) \right. \\
 & \quad \left. + d R \Pi(f_x x + 1/4) \cos(2\pi f_t t) \right), \quad (8)
 \end{aligned}$$

with  $f_t = 1/T$ ,  $f_x = Nb/X$  where  $Nb = 4$  is the number of bars of the same color in the stimulus. We used:

$$\Pi(x) = 2(\text{mod}(x, 1) < 1/2) - 1$$

to produce a spatial square wave of height bars.

In equation 8,  $(p, i, j)$  is the local coordinate system around  $p$ . The variable  $d \in [-1, 1]$  changes the direction

of motion to the left or right for all other parameters fixed. The value of  $d$  was randomly changed so that the observer wouldn't associate the variation in luminosity with the perceived direction of motion.

The variables  $m$  and  $R$  are the contrasts of the homochromatic and heterochromatic frames respectively. The two contrasts are modified with the point  $p$  to ensure that the stimulus is visible regardless of the luminance of  $p$ . These contrasts were set empirically. The contrasts for the reddest point  $p_1$  in the  $P_1$  plane is  $m_1 = 1/100$  and  $R_1 = 1/10$ . We set  $m_1 = 1/120$  and  $R_1 = 1/12$  in the  $P_2$  plane. For the other points, the contrasts are proportional to the luminance of the reddest point, for  $i = [2, \dots, 15]$ ,  $m_i = m_1 V_{\lambda}^t p_i / V_{\lambda}^t p_1$  and  $R_i = R_1 V_{\lambda}^t p_i / V_{\lambda}^t p_1$ . The correction is similar for plane  $P_1$  and  $P_2$ .

The stimulus is masked by a circular mask surrounded by a uniform background filled with the color of the point  $p$ . At a distance of one meter the visual angle of the stimulus is approximately 4.2 degrees. The head is left free but observers are placed against a table to fix the distance to the screen.

### C. Procedure

We measure the luminous vectors around several points  $p$  of coordinate  $x$ . The points may have been picked at random in the Euclidean color space. In order to better understand the geometry of the visual system and compare it to the Euclidean geometry of the display's color space, we select points from spheres of constant radiometric norm (Figure 4). Spheres are defined by points  $p$  of coordinate  $x$  such that  $x^t x = r^2$ . There were two experiments conducted. In the long experiment, we measure fifteen points on five spheres of various radii ( $r = 1, 2, 3, 4, 5$ ), for a total of seventy five points and vectors. In the short experiment, we measure only six points on two spheres ( $r = 2, 4$ ) for a total of twelve points and vectors. The fifteen or six selected points lie on projective lines across radii (see dashed lines on Figure 4(a)(b)).

After choosing point  $p$ , we fill the entire screen with the color corresponding to  $p$  during the two minimum motion sessions in  $P_1$  and  $P_2$  without interruption. The minimum motion stimulus is a slight modulation around point  $p$  and has a spatial and temporal average equal to point  $p$ . As a result, the observer's visual system remains fixed on  $p$ 's color throughout the measurement.

An experiment presentation frame with white letters on a black background is presented before the first stimulus. Any button can then be pressed to move to the first point in the experiment. The duration of the stimulus is set at two seconds and the observer is given an additional one second to make a response before the next stimulus is presented. Observers are asked to respond left or right using the response box of the VSG2.5. When the observer responds the next trial begins. In case of no responses the trial is not used for the estimation of the minimum motion angle. A measure of the minimum motion angle in one plane includes thirty trials lasting approximately two

minutes. Two sessions (in  $P_1$  and then  $P_2$ ) are required to measure one vector. After two sessions, the screen returns to white writing on a black background, displaying a pause suggestion. The observer is invited to relax and close his eyes for a moment if desired. Any button can then be pressed to move to the next point in the experiment. For the short experiment, when all points on a sphere are measured, the experiment ends. The observer leaves the experiment room to ask the experimenter to restart the experiment on the next sphere. Measurements for the long experiment are divided into three parts.

After measuring the direction of the vector  $v_{obs}$  at all points  $p$ , we measured the vector length at all points. The length of the vectors is given as the minimum contrast  $m$  in the direction of the luminous vector, needed to detect the homochromatic frame from the background. The contrast on the heterochromatic frame  $R$  is set to zero. To do so, we used a contrast adjustment procedure that uses large and small, positive or negative, delta values.

### D. Apparatus

To avoid reflections, all experiments were conducted in an experimental box with opaque black tissue on the walls and table. In order to ensure the stability of light during calibration measurements or experiments, computers and monitors were turned on at least an hour before the measurement. In this period, we ran a warm-up procedure displaying random colors.

This display consists of a 19-inch Trinitron CRT (DELL D1626HT) connected with BNC cables and driven by a Cambridge Research VSG 2.5 board with Windows 7. Display resolution is 1024x768@100Hz, and palette animation is used to generate stimulus. The answer box is a Cambridge Research six-button infrared response box connected to the VSG board. The programs were developed using the VSG-MATLAB API. 8.107 and VSG Library and ToolBox ver. 1.217.

### E. Participants

Two authors conducted an experiment with seventy five vectors lasting five hours on three consecutive days. In contrast, twenty four naive observers conducted the experiment in less than one hour with twelve vectors. Two naive observers were excluded from the final sample. The first was due to a technical problem with the response box, and the second was due to the absence of data measurement for the vector length. Naive observers were recruited from psychology students at the university. They received one experimental point as a bonus for the final examination. Authors were forty nine years old at the time of the experiment. They had normal vision or corrected to normal. Naive observers are 14 males and 8 females, with an average age of 26 years and with normal or corrected to normal vision. None of the participants were dichromat based on their self-report. Experiment procedure had been approved by the ethical committee of the University of Grenoble Alpes under the number IRB 00010290-2017.

## F. Surfaces from vector field

After two sessions of minimum motion by points, giving the direction of the vector, and the adjustment procedure giving the magnitude of the vector, we have  $N$  pairs of points and vectors composing the vector field  $(x, v)$ . We write  $X$  and  $V$  as two matrices of size  $3 \times N$ , which contain respectively  $N$  points and vectors. The value of  $N = 75$  for the two authors and  $N = 12$  for twenty two observers.

A model of the visual system is a prediction of  $V$  from  $X$ . We choose an affine model of the kind  $v = 2Bx + a_0$  where the vector  $v$  is the  $i$ th member of the matrix  $V$  and the vector  $x$  is the  $i$ th member of the matrix  $X$  (See supplementary method S1-SC for more explanation). The matrix  $B$  is not symmetric and is replaced by  $A = (B + B^t)/2$  [17]. The matrix  $A$  and the vector  $a_0$  are the parameters of the observer's model.

For a quadratic surface of level  $k$  passing through a point  $p$  of coordinate  $x$ , we have  $x^t A x + x^t a_0 = k^2$  (See supplementary method S1-SD for more explanations). This equation is also written as:

$$(x - x_0)^t H (x - x_0) = 1 \quad (9)$$

with  $x_0 = (1/2)A^{-1}a_0$ ,  $H = U(\sqrt{|S|}/\rho)\mathcal{J}(\sqrt{|S|}/\rho)U^t$  with  $\rho^2 = k^2 + (1/4)a_0^t A^{-1}a_0$  and  $U$  and  $S$  are the eigenvectors and eigenvalues matrices of  $A$ ,  $A = USU^t$ . The diagonal matrix  $\mathcal{J}$  is the signature of  $A$  (containing in its diagonal the sign of the eigenvalues of  $A$ ). Depending on the sign of the diagonal matrix  $\mathcal{J}$ , the surfaces described by  $A$  are degenerate quadrics such as plane (if one eigenvalue is zero), ellipsoids (if the signs of the three eigenvalues are positive), or hyperboloids (if one or two eigenvalues are negative).

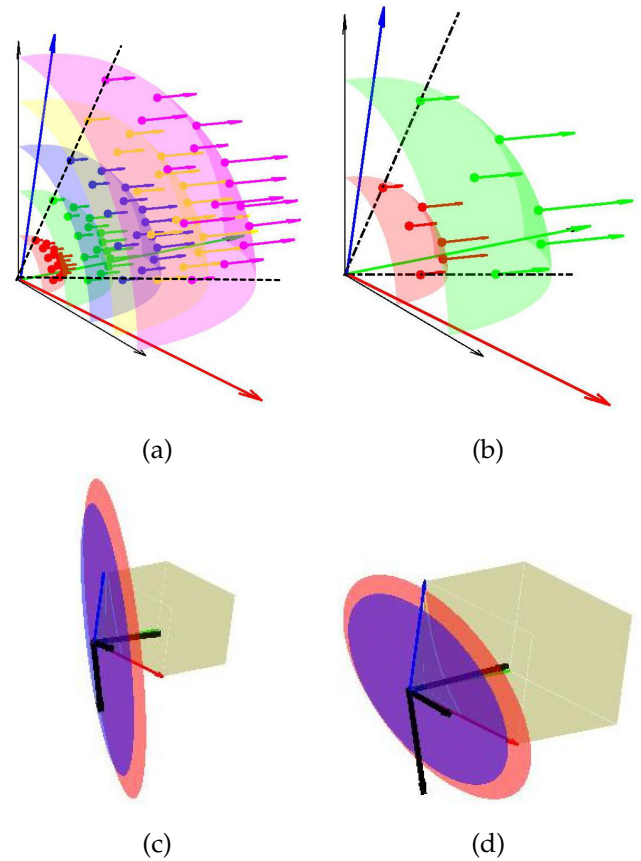
By posing  $z = (\sqrt{|S|}/\rho)U^t(x - x_0)$ , we have  $z^t \mathcal{J} z = 1$  which is the equation of the canonical surface of signature  $\mathcal{J}$ . Thus the estimated surface for each observer can be seen as a transformation  $T = (\sqrt{|S|}/\rho)U^t$  of the canonical surface placed at the origin  $x_0$ ,  $(x - x_0)^t T^t \mathcal{J} T (x - x_0) = 1$ . The observer model is in this case defined by an origin  $x_0$  and a coordinate transformation  $T$  of the canonical projective space in  $z$ .

The matrix  $T$  represents the observer's own color space since it contains the eigenvalues and eigenvectors of  $A$ . The columns of the observer's  $T$  matrix represent the color space axes for the observer. Using the variance of the coordinates of these axes, we can estimate the variability between observers.

## 4. RESULTS

Figure 4(a,b) shows the measured vectors for two observers. On the figure, the vector field is slightly convergent (not clearly visible since points are chosen on spheres). Thus, the iso-luminous surfaces of the observers are proportional hyperboloids. Due to proportionality, the pile of hyperboloid surfaces is included in a cone (the spectral cone for the observer). Cone and unitary hyperboloid estimates are shown in Figure 4(c,d). Table 1 shows the

three ordered eigenvalues of  $A$  [18] and the mean square error of the predicted  $\tilde{V}$  calculated using the affine model  $\tilde{V} = 2AX + x_0$  compared to  $V$ . (See supplementary material S1-SC and Dataset 1, Ref. [15], and Dataset 2, Ref. [16], for more information).



**Fig. 4. Results for two observers** (a)(b) Resulting vectors for two observers scaled for visibility. Two dashed lines illustrate projective lines used to place points on spheres of constant radiometric norm. See Dataset 1, Ref. [15], and Dataset 2, [16], for underlying values. (c) (d) Corresponding unitary hyperboloid (red) and cone (violet).

The signature of  $A$  has two negative eigenvalues for all observers we have tested so far, and for three observers, only one eigenvalue is negative, as shown in Table 1. To rule out the planar hypothesis we first checked that the mean eigenvalues of  $A$  from Table 1 were different from 0, on average, using t-test. The first ( $M=-0.156E-3$ ;  $SD=0.174E-3$ ),  $t(23) = -4.38$ ,  $p < 0.001$ , second ( $M=-0.056E-3$ ;  $SD=0.045E-3$ ),  $t(23) = -6.13$ ,  $p < 0.001$ , and third ( $M=4.6E-3$ ;  $SD=2.1E-3$ ),  $t(23) = 10.77$ ,  $p < 0.001$ , eigenvalues were all different from zero. Even though that could be implied from the sign of the t-test statistics, we also tested the sign of the eigenvalues using a sign rank test (using the exact solution for the function signrank in @MATLAB). The first and second eigenvalues in Table 1 were negatives and the third was positive (all  $p < 0.001$ ). Because the group study indicates a negative value for the second column,

we attribute the positive value for the three observers to experimental noise. The eigenvector associated with the

**Table 1.** Analysis of  $A$  and  $MSE$

OBS	Eigenvalues of $A * 1E3$			mse	OBS	Eigenvalues of $A * 1E3$			mse
1	-0.134	-0.084	4.84	5.17e-06	13	-0.136	-0.098	5.98	3.79e-07
2	-0.101	-0.065	3.90	5.33e-07	14	-0.122	-0.052	2.87	1.40e-06
3	-0.060	-0.024	2.28	9.74e-08	15	-0.085	-0.060	2.58	1.46e-07
4	-0.095	-0.065	3.15	9.94e-08	16	-0.054	<b>0.007</b>	5.52	9.04e-07
5	-0.144	<b>0.025</b>	3.39	6.93e-07	17	-0.894	-0.120	7.50	1.45e-06
6	-0.414	-0.152	5.64	1.14e-06	18	-0.056	-0.025	2.50	1.57e-07
7	-0.088	-0.008	7.10	1.19e-06	19	-0.116	-0.082	4.55	3.79e-07
8	-0.068	-0.015	3.62	8.86e-07	20	-0.062	-0.028	3.24	5.89e-07
9	-0.105	-0.053	4.50	1.20e-06	21	-0.124	-0.066	2.97	2.50e-07
10	-0.189	-0.086	10.56	1.36e-06	22	-0.169	-0.101	3.30	4.55e-07
11	-0.188	-0.136	8.24	1.09e-06	23	-0.077	<b>0.001</b>	3.62	5.27e-06
12	-0.196	-0.038	6.41	1.22e-06	24	-0.068	-0.036	2.97	6.65e-08

third eigenvalue (third column of Table 1) is the axis of symmetry of the observer cone and unitary hyperboloid. Its direction is close to  $V_\lambda$  the spectral efficiency function of the eye. In the observer's color space, the direction of the luminous vector is the most precise direction since the third eigenvalue has a much higher absolute value than the two others. The ratio of eigenvalues indicates how precise an observer is for the other directions. Compared to the luminous direction, these ratios average 331 for the Blue/Yellow direction and 40.5 for the Red/Green direction.

The major differences between observers for the direction of the axis of symmetry are in the Blue-Yellow direction ( $Q_3$  direction) and not in the Red-Green direction ( $Q_1$  direction). The variance among observers of luminous axis coordinates along  $Q_3$  direction is  $2.7E-3$ , while it is  $1.6E-3$  in  $Q_1$  direction.

Table 2 presents, for observer number two, the Weber ratio as the ratio  $WR(r) = \|v_{obs}\|/r$  between the length of the luminous vector  $\|v_{obs}\|$  and the radius of the sphere  $r$  on which the corresponding point is taken. Along each projective line, Weber ratio is consistent across the five radii as shown by the relatively low standard deviation compared to the mean (Compare columns M and SD of Table 2). Projective lines can then be graduated by the Weber/Fechner psychophysical law. However as shown in Table 2, the Weber ratio changes with the color of the point (see ratio differences across the lines in Table 2). The observed variation corresponds to what was found in [19, Figure 6], that compared achromatic Weber fraction for different colors. Blue (line 15) has a smaller ratio than red (line 1) and green (line 5) has a particularly large ratio. A similar variation of Weber ratio across colors was also found for all observers (See supplementary material S2-SA). Our Weber ratio follows the same variation as [19] however the value of the ratio is five times higher in our study. It may be because we measured a contrast at constant average luminosity, while they measured a

luminosity contrast.

Dröslér [20] proposed that the constant parameters of psychophysical laws are projective invariant. Using the metric estimated for observer two, we checked if the observed change in the Weber constant from color to color could be adjusted on the basis of the observer's projective hyperbolic space. The factor for each projective line can be thought of as the hyperbolic norm of the point  $p(x)$  that we write  $|x| = \sqrt{(x-x_0)^t H(x-x_0)}$ , where  $x_0$  and  $H$  are the observer's color space model. We apply a correction for each line of the table by calculating  $WB_c(r) = (12.8E_r\{|x|\} + 0.28)^{-1}WB(r)$ . The two constants are calculated as the best polynomial fit between  $E_r\{|x|\}$  and  $E_r\{WB(r)\}$  and are observer's dependent ( $E_r\{\}$  is the expectation over  $r$ ). Corrected Weber ratio are further normalized by a constant for having the same average Weber ratio than non corrected Weber ratio. From  $WB_c(r)$  we compute a revised standard deviation  $SD_c(r)$  (Last line of Table 2). The value  $SD_c(r)$  is 3.2 times lower than non corrected SD. Also, the specific pattern of change in blue, green, and red was no longer observed across color points. This correction applies equally well to all observers, even naive observers for whom only two ratios per color point can be calculated (See supplementary material S2-SA). According to this post-hoc analysis, the psychophysical law constant for each projective line is a consequence of the hyperbolic norm of the observer's color space.

**Table 2.** Percent Weber ratio

$n^\circ$	$100\ v_{obs}\ /r$					M	SD
	1	2	3	4	5		
1	0.50	0.45	0.33	0.45	0.48	0.44	0.06
2	0.60	0.50	0.50	0.53	0.54	0.53	0.04
3	0.80	0.50	0.63	0.70	0.80	0.69	0.13
4	1.00	0.80	0.80	0.75	0.74	0.82	0.11
5	1.00	0.90	0.77	0.78	0.80	0.85	0.10
6	0.50	0.40	0.43	0.45	0.46	0.45	0.04
7	0.60	0.65	0.63	0.53	0.54	0.59	0.06
8	0.80	0.75	0.67	0.68	0.68	0.71	0.06
9	0.90	0.75	0.67	0.70	0.76	0.76	0.09
10	0.40	0.45	0.40	0.40	0.46	0.42	0.03
11	0.60	0.60	0.67	0.58	0.56	0.60	0.04
12	0.70	0.65	0.57	0.68	0.64	0.65	0.05
13	0.40	0.45	0.37	0.38	0.40	0.40	0.03
14	0.60	0.55	0.50	0.50	0.54	0.54	0.04
15	0.40	0.35	0.33	0.35	0.34	0.35	0.03
SD	0.21	0.16	0.15	0.14	0.15	0.16	
SD <sub>c</sub>	0.05	0.06	0.06	0.03	0.04	0.02	

## 5. CONCLUSION

We found that the color space of the observer has a hyperbolic geometry. Constant luminosity surfaces are hyperboloid surfaces, and color spaces are a pile of hyperboloid



sheets with varying radius  $k$  that correspond to different levels of luminosity. All of those hyperboloids are included in a cone defined for  $k \rightarrow 0$ . This is the spectral cone for the observer, or the locus of mostly saturated lights (i.e. monochromatic lights). The chromaticity diagram can be viewed as a hyperboloid surface rather than a planar surface in this geometry.

According to literature, only the L- and M-cones input to the achromatic/motion channel we call  $V$ . Thus,  $V = aL + bM$  where  $a$  and  $b$  are the gains of the L and M channels, respectively. If adaptation modifies the spectral efficiency of the eye, it could be achieved by modifying those gains. If we further suppose that the color stimulus de-saturates the cone mechanism for which the stimulus is closest to (say red may de-saturate L-cones), then we can write those gains depending on the color position. If  $a = (M + S)$  and  $b = (L + S)$ , then the surface in LMS color space has the equation:  $V = 2LM + LS + MS$ . This is the equation of the hyperboloid of level  $V$  having L, M and S vectors as basis. Following this geometry, the luminosity of a color is given by a weighted linear combination of L- and M-cone mechanisms for which the weights are proportional to the sum of the two other cone responses.

Surprisingly enough, the quadratic metric we found is not inherited from the Hilbert scalar product (radiometric norm) but rather from a hyperbolic geometry that is likely imposed by visual processing. The transformation  $T$  and change of the origin  $x_0$  of the canonical unitary hyperboloid into the unitary hyperboloid of the observer represent the axes of the observer's color space. While LMS vectors lie on the envelope of the spectral cone, the axes estimated in  $T$  are based on a coordinate system having one component being the axis of symmetry of the observer's metric and the two others are orthogonal.

Weber's law and the hyperbolic metric are separate. The hyperbolic metric acts only as a factor in Weber's law from color to color. Colorfulness is the coordinate of a point on the unitary hyperboloid or equivalently a point on the Klein disk. It can be divided into saturation and hue. Luminosity is the affix to the projective line. Brightness is a monotonic function of luminosity and could be associated with the hyperbolic norm of the point  $p$ .

Metrics of color space are usually derived from color discrimination thresholds [21] showing a pseudo-Riemannian [20, 22] or non-Riemannian [23] structure. Our results support the pseudo-Riemannian hypothesis because iso-luminous surfaces are hyperbolic. In our experiment we fixed the adaptation state of the observer to the point of measurement. This ensured a single point of adaptation across the visual field of the observer. Our result is thus based on several adaptation states and this is probably the reason why we found a cone with a large aperture. According to our interpretation, under a single adaptation state, the surface of equal luminosity is already a hyperboloid, but with a smaller aperture than the one we measured. By aggregating several points of adaptation

we found a global cone with a larger aperture.

The luminous direction differs from the direction of the white on the screen. Indeed, we found that the luminous axis is close to  $V_\lambda$ , the direction of  $V(\lambda)$  in the color space of the screen. In contrast, the manufacturer of the screen defines the white with  $R = G = B$ , so the direction of the white is closer to  $w = (1, 1, 1)$ . Our measurements are not related to color appearance but to color physiology as we are taking heterochromatic photometry instead of whiteness measurements. From physiological to perceptual space, there is probably a projective transformation. We have shown that physiological color space has a hyperbolic metric. The perceptual space, which is probably an isomorphism transformed space from physiology [24], should also be hyperbolic.

There are individual variations between observers seeing the same points in the computer display color space. Their hyperbolic metrics differ, as evidenced by the difference between the eigenvectors matrix  $U$  and the eigenvalue matrix  $S$  of matrix  $A$ . Differences in the observer's color space may be compensated for in displaying images. Here, we suggest compensation can offer a way to reduce tiredness after spending long hours watching a display. Additionally, it can be used to enhance virtual reality immersion.

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**Disclosures.** The authors declare no conflicts of interest.

**Data availability.** All results data are available in Dataset 1, Ref. [15], and Dataset 2, Ref. [16].

**Supplementary Material.** See [Supplement](#) for supporting content.

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  12. If one works with  $N$  sampled spectral data with spectral step  $\Delta\lambda = 1$ , one can defined  $|f\rangle = [f(\lambda_m), \dots, f(\lambda_i), \dots, f(\lambda_M)]^t$  where  $\lambda_i = \lambda_m + (i - 1)(\lambda_M - \lambda_m)/N$  and  $i = 1..N$  are the sampling wave-lengths.
  13. Dirac's bra-ket notation (defined in Dirac, P. A. M. (1981). *The principles of quantum mechanics*. Oxford university press.) is not needed here because mathematics can be expressed with matrix-vector multiplication since the considered spectra are sampled. But we found bra-ket notation convenient for describing continuous spectrum (not shown here) and we had adopted it. Additionally, bar-ket notation distinguished between ket vector  $|f\rangle$  that described spectral stimuli and dual bra vector  $\langle g|$  that described visual color mechanisms. This decomposition between color stimuli and color mechanism is consistent with a representation of points for color stimuli and vector for color mechanisms [9].
  14. In general, the phosphor spectral matrix  $|P\rangle$  is not an orthogonal basis for light spectra that a computer display can emit. Correlation between phosphor spectra is a  $3 \times 3$  matrix given by  $\langle P|P\rangle = P^t P$ . It is usual in principal component analysis to replace correlated basis vector in  $|P\rangle$  by an orthogonal basis vector. The change of variable is  $|Q\rangle = (P^t P)^{-1/2}|P\rangle$  which guaranties that  $\langle Q|Q\rangle = \langle P|(P^t P)^{-1/2}(P^t P)^{-1/2}|P\rangle = 1$ .
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