Model of parafoveal chromatic and luminance temporal contrast sensitivity of humans and monkeys

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Rhesus monkeys are a valuable model for studies of primate visual contrast sensitivity. Their visual systems are similar to that of humans, and they can be trained to perform detection tasks at threshold during neurophysiological recording. However, the stimulus dependence of rhesus monkey contrast sensitivity has not been well characterized. Temporal frequency, color, and retinal eccentricity affect the contrast sensitivity of humans in reasonably well-understood ways. To ask whether these factors affect monkey sensitivity similarly, we measured detection thresholds of two monkeys using a two-alternative, forced-choice task and compared them to thresholds of two human subjects who performed the same task. Stimuli were drifting Gabor patterns that varied in temporal frequency (1–60 Hz), L- and M-cone modulation ratio, and retinal eccentricity (2°–14° from the fovea). Thresholds were fit by a model that assumed a pair of linear detection mechanisms: a luminance contrast detector and a red-green contrast detector. Analysis of model fits indicated that the sensitivity of these mechanisms varied across the visual field, but their temporal and spectral tuning did not. Human and monkey temporal contrast sensitivity was similar across the conditions tested, but monkeys were twofold less sensitive to low-frequency, luminance modulations.

Introduction

A primary goal of neuroscience is to understand how sensory signals are converted into perceptual experiences. This broad phenomenon can be studied fruitfully through flicker sensitivity. Neurons in the early visual system respond to flicker above the critical flicker fusion frequency, implying a loss of high-frequency information between these neurons and those that mediate perception directly (Lee, Pokorny, Smith, Martin, & Valberg, 1990; Kremers, Lee, & Kaiser, 1992; Yeh, Lee, & Kremers, 1995; Engel, Zhang, & Wandell, 1997; Gur & Snodderly, 1997; Krolak-Salmon et al., 2003; Williams, Mechler, Gordon, Shapley, & Hawken, 2004; Vul & MacLeod, 2006; Jiang, Zhou, & He, 2007; Lee, Sun, & Zucchi, 2007; Falconbridge, Ware, & MacLeod, 2010). In addition, some neurons respond to imperceptible low-frequency modulations, demonstrating that information loss is not exclusive to high frequencies (Palmer, Cheng, & Seidemann, 2007; Hass & Horwitz, 2013). The loci and stimulus specificity of information loss in the visual system are largely unknown, and identifying them is an important step toward understanding visual awareness (Crick & Koch, 1998; Carmel, Lavie, & Rees, 2006).

With regard to temporal vision specifically, a significant obstacle to localizing information-processing bottlenecks is that existent neurophysiological and psychophysical measurements are difficult to compare. Several factors contribute. First, psychophysical measurements of temporal contrast sensitivity are made at low contrast, by definition, whereas most neurophysiological studies use high-contrast stimuli. Nonlinearities in neuronal contrast-response functions prevent accurate extrapolation of responses from high to low contrasts. Second, temporal contrast sensitivity varies across the visual field (Sharpe, 1974; Koenderink, Bouma, Bueno de Mesquita, & Slappendel, 1978a; Koenderink, Bouma, Bueno de Mesquita, & Slappendel, 1978b; Versu, Rovamo, Laurinen, & Nasanen, 1982; Wright & Johnston, 1983; Rovamo & Raninen, 1984; Tyler, 1985; Tyler, 1987; Pointer & Hess, 1989; Snowden & Hess, 1992) and with retinal illumination (De Lange Dzn, 1961; Kelly, 1972; Rovamo & Raninen, 1984; Snowden, Hess, & Waugh, 1995). Neurophysiological and psychophysical measurements

Citation: Gelfand, E. C., & Horwitz, G. D. (2018). Model of parafoveal chromatic and luminance temporal contrast sensitivity of humans and monkeys. Journal of Vision, 18(12):1, 1–17, https://doi.org/10.1167/18.12.1.
are rarely matched for these conditions. Finally, most neurophysiological measurements of flicker sensitivity have been made in animal models, and relatively little is known about the temporal contrast sensitivity of these animals (but see De Valois, Morgan, Polson, Mead, & Hull, 1974; Merigan, 1980).

To help bridge the gap between neurophysiological and psychophysical measurements of temporal contrast sensitivity, we made behavioral measurements in rhesus monkeys—the animal most frequently used to model human visual behavior. Specifically, we used a two-alternative, forced-choice (2AFC) task to measure contrast sensitivity of two rhesus monkeys as a function of three factors: temporal frequency, the relative modulation depth of the long wavelength-sensitive (L) cones and the medium wavelength-sensitive (M) cones (i.e., color direction in the LM plane of cone contrast space), and position in the visual field. We varied color because monkeys are highly sensitive to chromatic modulations under some conditions (Stoughton, Lafer-Sousa, Gagin, & Conway, 2012; Gagin et al., 2014; Lindbloom-Brown, Tait, & Horwitz, 2014). We also varied visual field location because chromatic sensitivity drops steeply with retinal eccentricity in humans (Anderson, Mullen, & Hess, 1991; Mullen, 1991; Stromeyer, Lee, & Eskew, 1992; Mullen & Kingdom, 2002), and most neurophysiological studies probe neurons with parafoveal receptive fields. For comparison, we also measured the temporal contrast sensitivity of two human observers under the same conditions as the monkeys.

To analyze the data, we built a model that described contrast sensitivity across the range of stimulus variations tested. The model was based on three established models, each of which described contrast sensitivity as a function of temporal frequency (Watson, 1986), color direction (Stromeyer, Cole, & Kronauer, 1985), and location in the visual field (Robson & Graham, 1981). These models had not been previously united, but we found that a simple combination predicted thresholds accurately without the need to assume complex interactions among the model parameters.

**Methods**

**Subjects**

Four subjects participated in this study: the authors (H1, a 23-year-old woman; H2, a 46-year-old man) and 2 nonhuman primates (M1 and M2, both male, *Macaca mulatta*). All procedures used with nonhuman primates were approved by the University of Washington Institutional Animal Care and Use Committee and adhered to the American Physiological Society’s Guiding Principles for the Care and Use of Vertebrate Animals in Research and Training. All procedures used with human subjects conformed to the Declaration of Helsinki and the policies of the University of Washington Human Subjects Division. Human subjects provided written, informed consent.

**Displays**

All subjects were tested in a room that was dark except for the light from a digital light-processing projector (ProPixx, VPixx Inc., Saint-Bruno, Canada) illuminating a rear projection screen (Da-lite Inc., Warsaw, IN) at 240 Hz. The screen subtended 46° × 26° of visual angle. The center of the screen was 61 cm in front of the subject and matched vertically and horizontally to the subject’s eye level. The chromaticity of the display background was (x = 0.3, y = 0.3), and the luminance was 90 cd/m².

**Psychophysical task**

Contrast detection thresholds were measured using a spatial 2AFC contrast detection task. Each trial began with the presentation of a 0.2° × 0.2° black fixation
point at the center of the screen (Figure 1). Five hundred milliseconds later, a Gabor stimulus appeared in the left or right hemifield. The fixation point disappeared 100 to 600 ms after the end of the stimulus presentation, and simultaneously, two targets appeared on the horizontal meridian. The subject was then required to indicate within 700 ms whether the stimulus had appeared on the left or right by selecting the corresponding target. Correct responses were accompanied by a tone and, for monkeys, a water reward.

Testing procedures

Monkey subjects were seated in a testing chair, with their heads stabilized by a head posting device. Eye position was tracked with a scleral search coil (Riverbend Instruments, Birmingham, AL). In 86% of the testing sessions, fixation was required to remain within a 1° × 1° window. In the remaining 14% of the testing sessions, the fixation window was enlarged to a maximum of 1.5° × 1.5°. Targets appeared 2° from the fixation point on the horizontal meridian.

Human subjects performed the same psychophysical task as the monkeys. In 42% (133 of 320) of the testing sessions, the subject’s reports were expressed via saccades to the same target locations as the monkeys’. In these sessions, head position was stabilized with a chin rest, eye position was tracked (EyeLink 1000 Plus, SR Research Ltd., Ottawa, Canada), and fixation was enforced. In the other 58% of sessions, subjects indicated their responses with a button box, and eye position was not tracked. The chin rest was used in most but not all of these sessions. Sixty percent of the button box sessions were conducted before the eye tracker sessions.

To examine the effect of response method on detection thresholds, we compared thresholds for 10 different combinations of color direction and temporal frequency, on the horizontal meridian, 5° from the fixation point. Threshold measurements were strongly correlated across response methods (r = 0.93 and 0.67 for H1 and H2, respectively) and did not differ significantly for either subject (paired t-tests: p = 0.86 and p = 0.11), indicating that the two response methods yielded similar threshold measurements.

Stimuli

The stimulus was an upward-drifting, horizontally oriented Gabor, with a spatial frequency of 1 cycle/° and a standard deviation of 0.15°. Stimulus contrast ramped up over 167 ms, remained constant for 334 ms, and then ramped down over 167 ms. The length of the stimulus duration mitigates the effect of the contrast envelope on the temporal frequency power spectrum.

Contrast detection thresholds were measured as a function of three variables: temporal frequency, color direction in the LM plane, and location in the visual field. Temporal frequency and color direction varied within blocks of trials and, on each trial, were selected from a set of two to four combinations that were chosen at the beginning of the block. Stimulus locations in the visual field were fixed within each block. Practice trials at the beginning of each block familiarized the subjects with the stimulus locations. Nevertheless, increases in spatial uncertainty with retinal eccentricity presumably manifest as increases in contrast detection thresholds (Pelli, 1985; Levi, Klein, & Yap, 1987).

Colorimetric calculations were based on the Stockman, MacLeod, and Johnson (1993) 10° cone fundamentals. S-cones were not modulated, and all stimuli were presented at ≥2° from the fovea to avoid peak macular pigment density. L- and M-cone contrasts were defined as

\[ L\text{-cone contrast} = \frac{L_{\text{STIM}} - L_{\text{BACKGROUND}}}{L_{\text{BACKGROUND}}}, \]  

\[ M\text{-cone contrast} = \frac{M_{\text{STIM}} - M_{\text{BACKGROUND}}}{M_{\text{BACKGROUND}}}, \]

where \( L_{\text{STIM}} \) represents the L-cone excitation produced by the peak of the Gabor stimulus and \( L_{\text{BACKGROUND}} \) represents the L-cone excitation produced by the background. The quantities \( M_{\text{STIM}} \) and \( M_{\text{BACKGROUND}} \) are identical except for the M-cones.

Color direction was defined as

\[ \tan^{-1}\left( \frac{L\text{-cone contrast}}{M\text{-cone contrast}} \right), \]

and the modulation amplitude of the stimulus was defined as

\[ \sqrt{(L\text{-cone contrast})^2 + (M\text{-cone contrast})^2}. \]

The color direction and modulation amplitude of a Gabor pattern that modulates the L- and M-cones can be represented as the direction and length, respectively, of a vector in the LM plane of cone contrast space. Temporal frequency can be varied independently of L- and M-cone contrasts and is therefore represented as an orthogonal stimulus dimension. Thus, each Gabor stimulus is represented in a three-dimensional space (Figure 2).

Contrast detection thresholds for each color direction–temporal frequency combination were measured by the QUEST procedure (Watson & Pelli, 1983). The mode of the QUEST function after 40 trials was taken as an estimate of the threshold. The number of
threshold measurements from each subject is provided in Table 1.

Within each block of trials, the Gabor stimulus appeared at one of two locations that were mirror symmetric about the vertical meridian. We refer to these location pairs as the location (singular) of the stimulus, because knowing one member of the pair identifies the other. At each location tested, thresholds were first measured with four stimuli: 1 Hz L+M, 1 Hz L/C0M, 60 Hz L+M, and 60 Hz L/C0M.

Subsequent color direction–temporal frequency combinations were selected using an adaptive procedure based on Gaussian process regression (Rasmussen, 2004). Before each session, the subject’s thresholds were fitted with a nonparametric function that provided threshold predictions for every color direction–temporal frequency combination, along with error estimates associated with these predictions. Color direction–temporal frequency combinations were sampled where the estimated prediction error was greatest. The covariance of the Gaussian process was the product of a Matérn function of log temporal frequency and a periodic function of color direction (MacKay, 1998). Hyperparameters of the covariance function, which specify the variance and length scale of the fitted function, were refit after each block by maximum likelihood. Color direction–temporal frequency combinations for which the predicted threshold was outside of the gamut of the display were not tested.

Modeling contrast sensitivity: Effects of temporal frequency and color direction

Our model of temporal contrast sensitivity is based on one developed by Watson (1986). The Watson model assumes that detection is mediated by a single, linear bandpass filter that can be described as the difference of two low-pass filters, each with transfer function

\[ H(\omega) = \xi(H_1(\omega)) - \xi(H_2(\omega)), \]

where \( \tau \) is a time constant, \( \omega \) is temporal frequency in Hz, and \( n \) is the number of low-pass stages. The transfer function of the bandpass filter is the difference between the transfer functions of two low-pass filters:

\[ H_1(\omega) = \xi(i2\pi\omega + 1)^{-\tau}, \]

where \( H_1(\omega) \) and \( H_2(\omega) \) are the transfer functions of the low-pass filters defined by Equation 5, \( \xi \) is a gain parameter, and \( \xi \) controls the transience of the bandpass filter. When \( \xi = 0 \), the filter is low pass, and when \( \xi > 0 \), the filter is bandpass.

We extended this model to describe contrast detection thresholds across color directions in the LM plane. We assumed that detection is mediated by two linear mechanisms whose outputs are squared and summed. As a consequence, detection contours at any temporal frequency were constrained to be elliptical. We did not assume that the luminance and chromatic mechanisms were orthogonal. Therefore, the orientation of detection ellipses in the LM plane could, and in general did, change with temporal frequency.

One of the mechanisms (RG) was assumed to respond to the difference between L- and M-cone contrasts. The second mechanism (LUM) was assumed to respond to a weighted sum of L- and M-cone contrast. The sensitivity of each mechanism at frequency \( \omega \) was \( H_{RG}(\omega) \) and \( H_{LUM}(\omega) \), which are transfer functions that conform to the Watson (1986) model but have different parameter values. The

| Subject | Total No. of threshold measurements | Opponent ≤5 Hz | Nonopponent ≤5 Hz | Opponent >5 Hz | Nonopponent >5 Hz |
|---------|----------------------------------|----------------|-------------------|--------------|-------------------|
| M1      | 344                              | 56             | 86                | 69           | 133               |
| M2      | 724                              | 117            | 193               | 151          | 263               |
| H1      | 220                              | 35             | 60                | 39           | 86                |
| H2      | 272                              | 41             | 77                | 46           | 108               |

Table 1. Number of threshold measurements per subject. Notes: Color direction and temporal frequency conditions were distributed nearly continuously in the experiment but are binned coarsely in the table. Nonopponent and opponent stimuli are those in which L- and M-cone modulations had the same or opposite sign, respectively.
predicted contrast sensitivity across directions in the LM plane was therefore

\[
\text{Contrast sensitivity} = \sqrt{\left( H_{\text{RG}}(\omega) \cos(\frac{2\pi}{2}) L + \sin(\frac{2\pi}{2}) M \right)^2 + \left( H_{\text{LUM}}(\omega) \cos(\theta) L + \sin(\theta) M \right)^2}, \tag{7}
\]

where \( L \) and \( M \) are cone contrasts normalized so that \( L^2 + M^2 = 1 \), and \( \theta \) is a fitted parameter indicating the relative weighting of \( L \)- and \( M \)-cones to the LUM mechanism. The RG mechanism was assumed to weight \( L \)- and \( M \)-cone signals equally (Stromeyer et al., 1985; Gegenfurtner & Hawken, 1995; Stromeyer, Kronauer, Chaparro, & Eskew, 1995; Sankeralli & Mullen, 1996). Contrast threshold was defined as the reciprocal of contrast sensitivity.

**Modeling contrast sensitivity: Effects of stimulus position in the visual field**

The preceding model describes contrast sensitivity at individual locations in the visual field. To capture differences in contrast sensitivity across the visual field, we extended the model. Visual field locations were represented in polar coordinates, where \( r \) is the eccentricity of the stimulus in degrees of visual angle, and \( \phi \) is the position of the stimulus in the plane of the screen, relative to the horizontal meridian (Figure 1). These parameters can be written as

\[
r = \sqrt{h^2 + v^2} \tag{8}
\]

\[
\phi = \tan^{-1} \left( \frac{v}{h} \right), \tag{9}
\]

where \( h \) and \( v \) are the horizontal and vertical positions, respectively, of the stimulus in degrees of visual angle relative to the fixation point.

As described in the Results section, we tested several parametric forms of the relationship between \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \) (Equation 6) and \((r, \phi)\). The general form of the dependence was

\[
\log_{10}(\xi) = b_0 + b_1 r + b_2 r \cos(2\phi) + b_3 r \sin(2\phi), \tag{10}
\]

where \( \xi \) represents \( \xi_{\text{LUM}} \) or \( \xi_{\text{RG}} \), which govern the sensitivity of the LUM and RG mechanisms, respectively. Setting \( \phi = 0 \) shows that \( \xi \) changes with slope \((b_1 + b_2)\) along the horizontal meridian, and setting \( \phi = \pm \pi/2 \) shows that \( \xi \) changes with slope \((b_1 - b_2)\) along the vertical meridian. \( b_3 \) is a parameter that allows \( \xi \) to differ between the upper and lower visual fields. When \( b_3 \) is positive, \( \xi \) is greater in the upper hemifield, and when \( b_3 \) is negative, \( \xi \) is greater in the lower hemifield.

Note that \( b_3 \) does not affect \( \xi \) on the vertical meridian (where \( \phi = \pm \pi/2 \)), a region of visual space we were unable to test because of the logic of our left/right 2AFC task. All parameters were fit by minimizing the summed, absolute values of differences between the log-transformed measured and predicted detection thresholds.

**Results**

We measured contrast detection thresholds of two monkey and two human subjects as a function of three variables: temporal frequency, angle in the LM plane, and location in the visual field. Thresholds of subject M2 (Figure 3), measured at screen location \( r = 5, \phi = 0 \), capture many features of this broader data set.

Thresholds generally increased with temporal frequency, as shown by the flaring of the data points and the fitted surface along the temporal frequency axis (Figure 3A, B). To show the effects of color direction, the data have been plotted twice: once rotated so that the \( L+M \) axis is in the plane of the page (Figure 3A) and once rotated so that the \( L–M \) axis is in the plane of the page (Figure 3B).

Detection thresholds for low temporal frequency \( L+M \) modulations were greater than for low temporal frequency \( L–M \) modulations, as expected (Stromeyer et al., 1985). This feature of the data is manifest in the greater width of the fitted threshold surface in the \( L+M \) direction (Figure 3C) than in the \( L–M \) direction (Figure 3D). It can also be seen in slices through the detection threshold surface fit: detection ellipses (Figure 3E) and contrast sensitivity functions (Figure 3F). The bump in RG sensitivity at ~2 Hz (Figure 3F) was a consequence of noisy data fit with a flexible model. It was not present in data from M2 at other locations nor in equivalent data from human subject H1 (Figure 4).

**Modeling contrast sensitivity at individual visual field locations**

For each observer, we measured detection thresholds at 11 to 21 locations in the visual field and fit the data independently at each location. Each of these fits contains 13 parameters: six that control the contrast sensitivity of the LUM mechanism, six that control the contrast sensitivity of the RG mechanism, and one that controls the \( L:M \) ratio of the LUM mechanism (see Equation 7 in the Methods section). We iteratively refit data from each location using solutions from every other location as initial guesses to the solver (MATLAB, MathWorks, Natick, MA; fmincon) until none of
the fits improved. We confirmed that the final model described the data well in the sense that the distribution of the residuals was centered on zero, was narrow, and depended little on predicted threshold (Figure 5). A subtle decrease in the variance of the residuals with predicted threshold may be due to the exclusion from this analysis of thresholds beyond the display gamut, which occur preferentially under high predicted-threshold conditions.

Describing detection thresholds at each visual field location independently had two significant shortcomings. First, the model overfit the data; many parameters...
were used to fit few data points. Second, predictions were made only at locations in the visual field at which thresholds had been measured. In the next section, we describe an extension of the model with fewer parameters that generalizes to a continuum of visual field locations.

**Modeling contrast sensitivity across visual field locations**

To extend the model, we first looked for patterns in the fitted values of the 13 model parameters across locations in the visual field. For each subject, we plotted the best-fit value of each parameter as a function of location in the visual field and inspected the plots to identify trends. The parameters $\xi_{LUM}$ and $\xi_{RG}$, which specify the sensitivity of the LUM and RG mechanisms, respectively, stood out as strongly eccentricity dependent (Equation 6, data not shown). These two parameters were therefore allowed to change with visual field location in all model variants described below.

We considered the possibility that allowing $n_{LUM}$ (Equation 5), $n_{RG}$ (Equation 5), or $\theta$ (Equation 7) to vary across the visual field, in addition to $\xi_{LUM}$ and $\xi_{RG}$, would improve the model fit. $n_{LUM}$ and $n_{RG}$ affect the slope of the high-frequency roll-off of the LUM and RG mechanisms, respectively, and $\theta$ affects the L:M cone contrast sensitivity.
the LUM and RG mechanisms, but not consistent with the idea that the overall sensitivity of in all 12 cases: 3 models
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from each model.

out from each fit and used to calculate prediction errors
location. Individual threshold measurements were held
were constrained to have the same value at every
location. Individual threshold measurements were held
from each fit and used to calculate prediction errors
from each model.

Prediction errors were similar when computed from
models that allowed \( n_{\text{LUM}}, n_{\text{RG}}, \) or \( \theta \) to vary as from a
model that did not (one-sided Wilcoxon tests, \( p > 0.1 \)
in all 12 cases: 3 models \( \times \) 4 subjects). These results are
consistent with the idea that the overall sensitivity of
the LUM and RG mechanisms, but not \( n_{\text{LUM}}, n_{\text{RG}}, \) or
\( \theta \), varies across the region of the visual field that we
probed. We therefore focused exclusively on models
which only \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \) changed with visual field
location. In the next section, we discuss the parametric
form of this dependence.

Parametric description of \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \) across
visual space

Contrast sensitivity for all subjects dropped more
quickly along the vertical meridian than along the
horizontal meridian for both LUM (Figure 6A) and
RG (Figure 6B). We modeled this pattern in the data
with Equation 10 (see the Methods section) and
considered four variants of the model. Each model
variant applied different constraints to \( b_3 \), which
controls the asymmetry of detection thresholds above
and below the horizontal meridian. In the “symmet-
ric” variant, sensitivity was forced to be symmetric in
the upper and lower visual fields \( (b_3 = 0 \) for both
\( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \)). In the “yoked” variant, the upper
and lower visual field asymmetry was constrained to
be identical for both mechanisms (a single \( b_3 \)
parameter was shared by \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \)). In the
“luminance-only” variant, LUM sensitivity, but not
RG sensitivity, was allowed to differ between upper
and lower visual fields \( (b_3 = 0 \) for \( \xi_{\text{RG}} \)). In the
“unconstrained” variant, LUM and RG sensitivity
was allowed to differ independently and asymmetri-
cally in the upper and lower visual fields \( (b_3 \) was fit
separately for \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \)).

We compared these model variants using a leave-
one-out, cross-validated analysis of prediction error
similar to the analysis of \( n_{\text{LUM}}, n_{\text{RG}}, \) and \( \theta \) previously
described. We held out individual threshold measure-
ments, fit the four models (symmetric, yoked, lum-
nance-only, and unconstrained) to the remaining data,
recorded prediction errors between the model fits and
the held-out data point, and repeated this process for
each threshold measurement. The model with the
lowest prediction errors, for all subjects, was the yoked
variant (Figure 7).

The yoked model contained 18 parameters: 13 that
governed sensitivity as a function of temporal fre-
cquency, color direction, and visual field location.
and five that governed changes in two of the 13 parameters
\( (\xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \))
 across the visual field. Residuals from these model fits,
plotted as a function of predicted threshold, were
similar to those obtained when a separate 13-parameter
model was fitted to the data at each screen location
individually despite the 8- to 15-fold reduction in the
number of parameters (Figure 8, compare to Figure 5).

The median ratio between the measured and
predicted thresholds was 1.00, indicating that the
predictions were not systematically biased upward or
downward. The 10th and 90th percentiles of the ratios
were 0.77 and 1.40, respectively, indicating that 80%
of the measured thresholds were within a factor of \( \sim 0.7 \)
of the predictions. We conclude that the model fit most of
the data accurately.

Analysis of residuals

If the model were specified perfectly, we would
expect the residuals to be independent and identically
distributed across all combinations of temporal fre-
cquency, color direction, and visual field location.
Testing this hypothesis is difficult given the number of
independent variables, but to confirm the absence of
strong patterns in the residuals, we performed two
additional analyses. In each analysis, we pooled
residuals across two of the stimulus variables (e.g., \( r \)
and \( \phi \) location in the visual field) and examined them as

weighting to the LUM mechanism. We fit the data from
each subject using models in which \( \xi_{\text{LUM}} \) and \( \xi_{\text{RG}} \) and,
optionally, one of the set \( (n_{\text{LUM}}, n_{\text{RG}}, \) and \( \theta \),
were
allowed to vary across location. All other parameters
were constrained to have the same value at every
location. Individual threshold measurements were held
out from each fit and used to calculate prediction errors
from each model.

Prediction errors were similar when computed from
models that allowed \( n_{\text{LUM}}, n_{\text{RG}}, \) or \( \theta \) to vary as from a
model that did not (one-sided Wilcoxon tests, \( p > 0.1 \)
in all 12 cases: 3 models \( \times \) 4 subjects). These results are
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a function of the remaining two (e.g., color direction and temporal frequency).

First, we collapsed residuals across visual field locations and calculated the autocorrelation of median residuals as a function of color direction and temporal frequency (Figure 9, left side of each panel). This autocorrelation function was fairly flat for all subjects, consistent with independent residuals across temporal frequency and color direction.

Second, we plotted median residuals as a function of location (Figure 9, right side of each panel). Residuals for subjects M1 and H1 had little discernable structure. On the other hand, the model systematically overestimated subject M2’s sensitivity near the horizontal meridian along the line $h = 5^\circ$ and underestimated it further from the horizontal meridian (Figure 9B). This pattern is probably due to task training: visual field locations at which sensitivity was
overestimated were tested earlier than locations at which sensitivity was underestimated. For subject H2, the assumption that contrast sensitivity decays exponentially along the horizontal meridian is imperfect. For this subject, contrast sensitivity drops more gradually over the central 5° of the horizontal meridian than predicted from exponential decay (Figure 9D; Equation 10).

Human-monkey comparison

As expected from previous studies, human and monkey temporal contrast sensitivity was similar (De Valois et al., 1974; Merigan, 1980). Here, we extended these results to all directions in the LM plane and a variety of locations in the visual field from 2° to 14°. To test quantitatively for differences in temporal contrast sensitivity between humans and monkeys, we took the raw contrast sensitivity measurements for each subject, normalized them within each visual field location, and then pooled them across locations. Normalized luminance contrast sensitivity was greater for humans than monkeys from 1 to 1.5 Hz (two-way analysis of variance with subject as a random effect, $p = 0.056$). Model fits for each subject, evaluated at location $r = 5°$, $\phi = 0$, illustrate this difference (Figure 10).

Discussion

We measured the contrast detection thresholds of two humans and two monkeys as a function of three variables: temporal frequency, location in the visual field, and color direction in the LM plane. We built a model that successfully described thresholds for all observers over the range of stimulus variables tested (1–
60 Hz, 2°-14° of eccentricity, and all color directions within the LM plane).

We obtained three main results. First, the model fitted contrast detection threshold data from both humans and monkeys with small adjustments to the parameters (Table 2). This confirms the similarity between monkey and human luminance contrast sensitivity and extends these results across color directions in the LM plane and visual field locations. Second, the model did not require complex interactions among parameters to fit the data adequately. This is not trivial: As the number of stimulus variables increases linearly, the number of variable combinations increases exponentially. Theoretically, for example, sensitivity to L-cone modulations in the upper visual field could have been poorly predicted by a model that assumes independent contributions of color direction and screen location to contrast sensitivity, but this was not the case. Third, we found, using the model as a guide, that monkeys were only half as sensitive to low-frequency luminance modulations as humans. A retrospective look at data from a previous study confirms this result, although this difference was not previously emphasized (Merigan, 1980, their figure 3).

Subjects M1, H1, and H2 were heavily trained on the task before data collection began (M1 is monkey A and H2 is human G from Lindbloom-Brown et al., 2014). Subject M2 was the least heavily trained subject but exhibited similar contrast sensitivity to the others, suggesting that all four subjects had attained near-asymptotic performance. Longer training periods would likely have been necessary had we used stimuli containing S-cone increments (Gagin et al., 2014).

Effects of eye size

Retinal illuminance depends on eye size and affects temporal contrast sensitivity (De Lange Dzn, 1958;
Kelly, 1961; Snowden et al., 1995). Monkey eyes are smaller than human eyes, so their retinal illuminance is relatively high. We considered the possibility that this size difference could account for the difference between humans and monkeys in low-frequency luminance contrast sensitivity but found it unlikely. When retinal illuminance is greater than 10 Td, human detection thresholds to low-frequency luminance modulations are largely independent of illuminance when they are measured in Weber contrast (Kelly, 1961). The background of our display (producing ~650 Td) was sufficiently intense that we would not expect low-frequency luminance contrast sensitivity to vary much, if at all, with the modest difference in retinal illuminance afforded by differences in eye size (Virsu & Lee, 1983; Smith, Lee, Pokorny, Martin, & Valberg, 1992).

Effects of stimulus size

Adjusting stimulus size to compensate for the cortical magnification factor, a procedure called M-scaling, approximately equates detection thresholds across retinal eccentricities (Rovamo, Virsu, & Nasanen, 1978; Strasburger, Rentschler, & Jüttner, 2011). M-scaling is sufficient to equate temporal contrast sensitivity across eccentricity under some conditions (Virsu et al., 1982) but not others (Rovamo & Raninen, 1984; Raninen & Rovamo, 1986). We did not M-scale our stimuli primarily because M-scaling that equates luminance contrast detection thresholds does not equate chromatic contrast detection thresholds (Noorlander, Koenderink, den Ouden, & Edens, 1983; Rovamo & Iivanainen, 1991; Vakrou, Whitaker, McGraw, & McKeefry, 2005; Masuda & Uchikawa, 2009). An important future direction is to extend the model to multiple stimulus sizes.

Assumptions of the model

In constructing the model, we relied heavily on results from previous studies. In this section, we present the assumptions of the model and direct the reader to the studies that supported these assumptions.

We chose a particular parametric form for the shape of the temporal contrast sensitivity function that is sufficiently flexible to fit a variety of data sets (Watson, 1986; Barten, 1993). We further assumed that detection contours in the LM plane are elliptical. This description, while demonstrably imperfect, is adequate under the stimulus conditions we used (Poirson, Wandell, Varner, & Brainard, 1990; Cole, Hine, & McIlhagga, 1994; Metha, Vingrys, & Badcock, 1994; Giuliani & Eskew, 1998). Detection thresholds of humans in the

Table 2. Parameter values from final fitted models.
LM plane are roughly elliptical across temporal frequencies (Noorlander, Heuts, & Koenderink, 1981) and retinal locations (Stromeyer et al., 1992), and we found that this is also true for monkeys.

We assumed that the orientations and sizes of detection ellipses were given by an energy calculation on the outputs of two linear detection mechanisms (Stockman & Brainard, 2010). Noise masking reveals more than two detection mechanisms in the LM plane (Hansen & Gegenfurtner, 2013; Shepard, Swanson, McCarthy, & Eskew, 2016), but two mechanisms dominate under the conditions of our experiment (Giulianini & Eskew, 1998; Stromeyer, Thabet, Chaparro, & Kronauer, 1999). We assumed that cone weights to the two postulated detection mechanisms do not change with temporal frequency. This approximation is imperfect but is reasonable when the L- and M-cones are in similar adaptation states (Stromeyer, Cole, & Kronauer, 1987; Gegenfurtner & Hawken, 1995; Stromeyer, Chaparro, Tolias, & Kronauer, 1997; Stockman & Plummer, 2005a; Stockman & Plummer, 2005b; Stockman, Jägle, Pirzer, & Sharpe, 2008). Under the conditions of our experiment, L- and M-cones absorbed ~8,900 and 7,400 photons/cone/s, respectively, and were therefore in an adaptation state similar to that produced by a moderate-intensity, 565-nm background. Under these conditions, flicker perception is dominated by a fast, cone-nonopponent pathway with little influence of the slow, cone-opponent pathway that might manifest as frequency-dependent cone weights to the LUM mechanism in our experiment (Stockman, Henning, Anwar, Starba, & Rider, 2018).

We also assumed that cone weights to each mechanism do not vary with retinal eccentricity. This assumption is supported by the near-constant L:M cone ratio to the RG mechanism across the visual field (Newton & Eskew, 2003; Sakurai & Mullen, 2006; Hansen, Pracejus, & Gegenfurtner, 2009) and to the LUM mechanism over the region of visual space we probed (Anderson et al., 1991; Knau, 2000). Further support for this assumption comes from our observation that allowing \( n \), the L:M ratio of the LUM mechanism in the model, to vary across the visual field did not improve prediction accuracy.

We assumed that log-transformed contrast sensitivity declines linearly with eccentricity with a slope that depends on the angle in the plane of the display screen (Robson & Graham, 1981). Our results confirmed the observation that the slope of this relationship is steeper near the vertical meridian than near the horizontal meridian (Pointer & Hess, 1989; Pointer & Hess, 1990; Abrams, Nizam, & Carrasco, 2012). Our results also confirm that low-frequency chromatic sensitivity is greater than low-frequency luminance sensitivity at the fovea (Chaparro, Stromeyer, Huang, Kronauer, & Eskew, 1993), and this relationship can reverse in the periphery due to the steeper decline in chromatic sensitivity with retinal eccentricity (Mullen, 1991; Mullen & Kingdom, 1996; Mullen & Kingdom, 2002; Mullen, Sakurai, & Chu, 2005). We found that chromatic and luminance contrast sensitivity was similarly asymmetric between upper and lower visual fields.

We assumed that the shape of the temporal contrast sensitivity function of the luminance and chromatic detection mechanisms does not change with eccentricity over the region of visual space that we probed. The assumption, which is supported by previous results (Wright & Johnston, 1983; Snowden & Hess, 1992), was built into the model by allowing only \( \xi_L \) and \( \xi_R \) to change across the visual field. We tested this assumption by asking whether allowing \( n_L \) or \( n_R \) to vary across the visual field improved threshold predictions, and we found that it did not.

**Future directions**

The contrast detection literature is vast, and extracting core principles from it and synthesizing them into a concise, accessible format is useful. For example, using the model, we can communicate large data sets with few numbers and interpolate contrast sensitivity for conditions that we did not test. The model can be used to identify stimuli for which detection is maximally or minimally constrained by signals in the early visual system (Geisler, 1989; Angueyra & Rieke, 2013; Brainard et al., 2015; Hass, Angueyra, Lindsblom-Brown, Rieke, & Horwitz, 2015) and to identify stimuli that are differentially visible between subjects. Our model spans only a few stimulus dimensions but could in principle be merged with models that predict contrast sensitivity on the basis of stimulus parameters that we did not vary (e.g., background illumination, spatial frequency, stimulus size, and S-cone modulation). Our code and data are available on GitHub (http://www.github/horwitzlab).

Our model helps to bridge the gap between neurophysiological and psychophysical studies of temporal contrast sensitivity. Measurements of neuronal responses at psychophysical detection threshold are difficult to obtain in part because detection thresholds depend on stimulus parameters in complex ways. A classic approach to this problem is to identify a suprathreshold stimulus that excites an isolated neuron strongly and then titrate a stimulus parameter (e.g., contrast) to measure psychophysical and neuronal detection thresholds simultaneously. This approach can be inefficient; psychophysical trials are longer than fixation trials, and estimating a distribution of noisy neuronal responses requires many repeated trials. Moreover, the assumption that suprathreshold stimu-
lus preferences are predictive of neuronal sensitivity at the behavioral detection threshold may be inaccurate. The model we present helps meet these challenges. Using the model, a battery of stimuli can be synthesized that are matched for detectability but differ in other respects (e.g., temporal frequency and color). These stimuli can be presented at the receptive fields of recorded neurons during detection task performance or passive fixation. This approach may be useful for revealing the neuronal basis of contrast sensitivity. For example, magnocellular, parvocellular, and koniocellular neurons in the lateral geniculate nucleus all respond to L+M modulations, and what contributions each makes to contrast sensitivity is poorly understood. Stimulation of neurons of each type with threshold-contrast L+M modulations and comparing their relative sensitivity will provide an upper bound on each population’s contribution.

Keywords: monkey, temporal modulation, contrast sensitivity, detection

Acknowledgments

The authors thank Zack Lindbloom-Brown for computer programming and Beth Buffalo for generous assistance with human eye movement measurements. They also thank Abhishek De, Yasmine El-Shamayleh, and Patrick Weller.

Commercial relationships: none.

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