Temporal filtering of luminance and chromaticity in macaque visual cortex
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SUMMARY

Contrast sensitivity peaks near 10 Hz for luminance modulations and at lower frequencies for modulations between equiluminant lights. This difference is rooted in retinal filtering, but additional filtering occurs in the cerebral cortex. To measure the cortical contributions to luminance and chromatic temporal contrast sensitivity, signals in the lateral geniculate nucleus (LGN) were compared to the behavioral contrast sensitivity of macaque monkeys. Long wavelength-sensitive (L) and medium wavelength-sensitive (M) cones were modulated in phase to produce a luminance modulation (L + M) or in counterphase to produce a chromatic modulation (L – M). The sensitivity of LGN neurons was well matched to behavioral sensitivity at low temporal frequencies but was approximately 7 times greater at high temporal frequencies. Similar results were obtained for L + M and L – M modulations. These results show that differences in the shapes of the luminance and chromatic temporal contrast sensitivity functions are due almost entirely to pre-cortical mechanisms.

INTRODUCTION

Signal processing in the visual system preserves some types of information while eliminating others. If perfect knowledge of neuronal activity at one stage of the visual system (e.g. visual cortex) allows for perfect reconstruction of activity at an earlier stage (e.g. the photoreceptors), then information is perfectly preserved between them. If, instead, multiple patterns of activity at an early stage produce indistinguishable patterns at a later stage, then information has been lost. The ability of an observer to detect a stimulus—to distinguish it from a blank—is a consequence of information that is retained throughout the entire visual system. A central goal of visual neuroscience is to understand which types of information are lost and at which stage.

A salient example of information loss in the visual system is the dependence of the visibility of a periodic stimulus on its temporal frequency. This relationship, the temporal contrast sensitivity function, plays important roles in industry (Legall, 1991) and medicine (Owsley, 2011; Tyler, 1981), but its biological basis is incompletely understood. This uncertainty is due in part to methodological differences between neurophysiological and behavioral studies. Temporal contrast sensitivity depends on many factors including observer species, background luminance, retinal eccentricity, stimulus size, and duration (Benardete and Kaplan, 1997b, 1999a; Lee et al., 1990; Lindbloom-Brown et al., 2014; Mergan, 1980; Pokorny et al., 2001; Snowden and Hess, 1992; Snowden et al., 1995; Solomon et al., 2002; Swanson et al., 1987; Van der Horst, 1969). In the current study, care was taken to match these factors between neurophysiological and behavioral measurements, providing a clearer picture of their relationship than has previously been available.

Chromatic modulations are easier to see than luminance modulations at low temporal frequencies, but at higher frequencies, the reverse is true (De Lange Dzn, 1958; Kelly and van Norren, 1977) (Figure 1A). The photoreceptors cannot be responsible for this difference because the same photoreceptor types that mediate luminance detection also mediate chromatic detection. Luminance stimuli modulate the long- (L) and medium wavelength-sensitive (M) cones in phase (L + M), whereas most chromatic stimuli modulate them in counterphase (L – M). Differences in the temporal filtering of these two stimulus classes must therefore be due to stages of the visual system where signals from the L and M cones are already mixed.

The purpose of this study was to quantify cortical and pre-cortical contributions to luminance and chromatic temporal contrast sensitivity. Cortical contributions were computed by comparing the behavioral
sensitivity of a macaque monkey to that of a computational observer of spikes in the lateral geniculate nucleus (LGN) (Figure 1B). Pre-cortical contributions were computed by comparing two computational observers: one of LGN spikes and one of simulated currents across the outer segments of modeled cone photoreceptors.

The main result of these comparisons was that information loss in the cortex was similar for L + M and L/C0M modulations, whereas information loss between the cones and LGN differed profoundly for L + M and L/C0M modulations. Differences in luminance and chromatic behavioral temporal contrast sensitivity are therefore due to processes occurring upstream of the LGN with minimal cortical involvement.

RESULTS

Two monkeys (M. mulatta) performed a 2-alternative, forced-choice contrast detection task that required them to report on which side of a computer screen a drifting Gabor stimulus appeared. Detection thresholds were measured as a function of stimulus location, temporal frequency, and the amplitude of L and M cone modulations. A model was developed that predicted detection thresholds as a function of all of these parameters jointly (Gelfand and Horwitz, 2018 and Figure 1A). Visual stimuli were constructed on the basis of this model and used to measure the signal-to-noise ratio (SNR) of LGN neuronal responses (Figure 1B). All stimuli were at the monkeys’ behavioral detection threshold, or equivalently, matched for SNR at the output of the visual system.

LGN single-unit responses

The spatial and spectral sensitivity of each recorded LGN neuron were characterized with a white noise stimulus (Horwitz, 2020). Spike-triggered averaging was used to locate the receptive field (RF) center and to identify the physiological type of each neuron. Fifteen neurons were classified as magnocellular (8 from monkey 1 and 7 from monkey 2) and 38 as parvocellular (19 from each monkey). Each recorded neuron was then stimulated with Gabor patterns centered on its RF that varied across trials in temporal frequency and L and M cone modulation phase (in-phase, L + M, or counterphase, L – M). The L and M cone contrasts were always equal, and their maximum was set by the limits of the display (0.19 for the L – M stimulus and 0.86 for the L + M stimulus). A blank stimulus was included to measure baseline firing statistics.

A representative magnocellular neuron responded to L + M modulations vigorously at high temporal frequencies and more weakly as temporal frequency was reduced (Figure 2A) (for similar data from a second magnocellular neuron, see Horwitz, 2020). This neuron also responded to L – M modulations but only at the highest frequencies tested and then only transiently (Figure 2B). An example parvocellular neuron responded more vigorously to L – M modulations than to L + M modulations (Figures 2C and 2D), although, as expected from their low contrast, none of the stimuli used in this study drove parvocellular neurons strongly (the example in Figures 2C and 2D is among the most responsive parvocellular neurons in the data set).
Figure 2. Responses of two LGN neurons to Gabor stimuli near behavioral detection threshold
(A) Raster plot of magnocellular responses to L + M modulations. Trials have been sorted by temporal frequency (left ordinate) which covaries with cone contrast (right ordinate, identical for L and M cones) to maintain a constant level of stimulus detectability. The temporal envelope of the Gabor stimulus is shown above the rasters.
(B) Identical to (A) but showing responses to L – M modulations.
(C and D) Identical to (A) and (B) but for a parvocellular neuron.
(E) Signal-to-noise ratio (d') calculated from responses in (A) (black) and from responses in (B) (magenta). Shaded bands represent ±1 standard error estimated by non-parametric bootstrap.
(F) Identical to (E) but for the parvocellular neuron.
See also Figures S1 and S2.
The SNR of each response was calculated by comparing it to baseline activity. This analysis assumes that the signal in the spike trains is at the fundamental temporal frequency of the stimulus (see supplemental information, transparent methods), but this assumption was not critical to the main results (see supplemental information, Figures S1 and S2). The example magnocellular neuron had greater SNR for L + M than for L – M modulations at all frequencies tested (Figure 2E). The example parvocellular neuron had greater SNR for L – M modulations than for L + M modulations above 6 Hz (Figure 2F).

The relationships among spiking responses, temporal frequency, L and M cone modulation, and cell type become clearer when data are averaged across neurons (Figure 3). As expected, magnocellular neurons were more sensitive to L + M modulations, and parvocellular neurons were more sensitive to L – M modulations (Wiesel and Hubel, 1966). The SNR of magnocellular and parvocellular responses increased smoothly from 1 to 20 Hz despite the fact that contrast changed with temporal frequency in different ways for L + M and L – M modulations over this range to render each stimulus near detection threshold.

The SNR of the average neuron (Figure 3) is lower than the SNR of neuronal populations. To estimate the SNR of neuronal populations, the SNR of individual LGN neurons was inflated by an estimate of how many LGN neurons were modulated by the stimulus, as described in the next section.

Population SNR analysis

Magnocellular neurons have greater contrast sensitivity than parvocellular neurons do at matched eccentricity, but they are less numerous, raising the possibility that, as populations, parvocellular neurons might have greater SNR (Croner and Kaplan, 1995). To estimate the SNR of neuronal populations, a model was constructed using parameters from the literature (Horwitz, 2020). The model provided a scale factor for each neuron that reflects how many times greater the SNR of a population of similarly sensitive neurons is. Scale factors were 2.1-fold greater (±0.4 SD) for parvocellular neurons than magnocellular neurons at matched eccentricity.

Parvocellular population SNR rose steeply with the temporal frequency of L – M modulations, and magnocellular population SNR rose similarly with the temporal frequency of L + M modulations (Figure 4A). Magnocellular and parvocellular populations were also weakly and similarly responsive to their non-preferred modulations, L – M and L + M, respectively. The similarity of these patterns is striking considering that these data were derived from recordings from two distinct populations of neurons responding to two sets of stimuli that varied in temporal frequency and L and M cone contrast in different ways.

To quantify how much temporal information was lost in the cortex, SNR in the LGN was compared to behavioral sensitivity. For this purpose, the monkeys’ percent correct at threshold, 82%, was converted to an SNR of 1.27 (Green and Swets, 1966, see methods). At high frequencies, the SNR of magnocellular and parvocellular neurons exceeded this level by approximately a factor of 7 in response to L + M and L – M modulations, respectively. At lower frequencies, SNR in the LGN was lower. In fact, at the lowest

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**Figure 3. Population signal-to-noise analysis**

(A and B) Signal-to-noise ratio (d’) averaged across magnocellular neurons (A) and parvocellular neurons (B). Points represent means across neurons, and shaded bands represent ±1 standard error of the mean.

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**Figure 4. Population SNR analysis**

(A) Magnocellular population SNR rose steeply with the temporal frequency of L – M modulations. (B) Parvocellular population SNR rose similarly with the temporal frequency of L + M modulations.
frequencies tested, parvocellular population SNR fell below behavioral SNR for both L + M and L – M modulations (Figure 4A and methods). Parvocellular neurons are the sole conduit by which low temporal frequency L – M modulations are transmitted from the eye to the cortex, so parvocellular SNR was underestimated.

The analysis in Figures 4A and 4B was based on spikes recorded between stimulus onset and disappearance, including the slow (166 ms) contrast ramps at the beginning and end of each stimulus presentation. No adjustment was made for response latency, which biased SNR downward. To examine the effects of spike counting window on SNR, the start and stop times for spike inclusion were varied independently over a 200-ms range (Figures 4C–4G). This analysis showed that delaying the spike counting window by ~120 ms relative to the stimulus presentation epoch (marked by an ‘X’) produced a greater d’ value than the window used in (A) and (B), which did not take response latency into account (lower left corner).

(D–G) Identical to (C) but for 1.3, 1.8, 2.4, and 3.2 Hz modulations, respectively. (H and I) Identical to (A) and (B) but counting spikes from 120 ms after the stimulus appeared until 120 ms after the stimulus disappeared. See also Figure S3.

Across cell types and stimulus conditions, delaying the spike counting window by 120 ms affected SNR only subtly (compare Figures 4A to 4H and 4B to 4I). Over a broader range of spike counting windows, none was found that rendered parvocellular populations significantly more sensitive to low-frequency L – M modulations than the monkey (Figure S3). Over the same range of windows, magnocellular and parvocellular population SNRs were similar for L + M and L – M modulations, respectively (Figure S3).

These analyses show that low-frequency information was preserved with near-perfect fidelity downstream of the LGN, and the amount of information lost downstream of the LGN was nearly independent of whether L and M cone modulations were in phase or in counterphase. The difference between the luminance and chromatic temporal contrast sensitivity functions is therefore due primarily to information loss upstream of the LGN, which is quantified next.

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**Figure 4. Population signal-to-noise analysis**

(A and B) (A) Population signal-to-noise ratio (d’) as a function of temporal frequency for magnocellular neurons (triangles) and parvocellular neurons (circles) in response to L + M modulations (black) and L – M modulations (magenta). Bands represent ±1 standard error of the mean across neurons. Dashed line at 1.27 indicates the d’ inferred from behavioral sensitivity. Dashed rectangle represents the region magnified in (B).

(B) Population d’ for parvocellular neurons in response to 1 Hz, L – M modulations as a function of the start time (ordinate), and end time (abscissa) of the spike counting window. Contour is drawn at d’ = 1.27. A spike counting window delayed by 120 ms from the stimulus presentation epoch (marked by an ‘X’) produced a greater d’ value than the window used in (A) and (B), which did not take response latency into account (lower left corner).

(D–G) Identical to (C) but for 1.3, 1.8, 2.4, and 3.2 Hz modulations, respectively. (H and I) Identical to (A) and (B) but counting spikes from 120 ms after the stimulus appeared until 120 ms after the stimulus disappeared. See also Figure S3.
SNR loss upstream of the LGN

To measure how much information was lost between the cone photoreceptors and the LGN, cone photo-
current responses to the stimuli used in the LGN recordings were simulated using the model of Angueyra
and Rieke, (2013). SNR loss between the cones and the LGN exceeded SNR loss in the cortex and was
particularly severe at low temporal frequencies (Figure 5, diagonal cross hatches). Only 5% of the SNR avail-
able in cone outer segment currents in response to low-frequency L + M modulations reached the LGN
(Figures 5A and 5B). In response to L /C0 M modulations, information transmission efficiency was more
than doubled (Figures 5C and 5D). Above 5 Hz, the situation reversed; SNR loss for L /C0 M modulations ex-
ceeded SNR loss for L + M modulations. This analysis confirms differential retinal filtering of L + M and
L /C0 M modulations and shows that most of the information loss under the conditions tested occurred up-
stream of the LGN.

DISCUSSION

Much of the information in the light absorbed by photoreceptors fails to reach perception (Barlow, 1957;
Geisler, 1989, 2011). Identifying where and how this information is lost is a key step toward understanding
the biological basis of vision. The distinctive temporal properties of luminance and chromatic vision offer insight into this broader issue. The fact that information loss is temporal, not spatial, indicates a neural basis as opposed to an optical one. The fact that the same photoreceptor types mediate both aspects of vision indicates that the information loss is downstream of the photoreceptors. Previous studies have shown that low-frequency L + M modulations are selectively filtered in the retina and that high-frequency modulations are filtered in the cortex (Kaplan and Benardete, 2001; Kaplan et al., 1990). Contributions of the current study are the quantitative comparison of information loss upstream and downstream of the LGN and the demonstration that cortical filtering of L + M and L/C0M modulations is similar across temporal frequencies.

**Mechanisms of SNR loss in the retina and LGN**

The stimuli used in this study were approximately uniform within the RF of each LGN neuron studied. Consequently, center-surround antagonism reduced SNR in response to L + M modulations at low temporal frequencies (Figures 6A and 6B, top). At higher temporal frequencies, the delay of the surround became an appreciable fraction of the stimulus period, causing excitation from the center to move closer in time to the release of surround inhibition (Enroth-Cugell et al., 1983; Robson, 1966) (Figure 6B, bottom). This change in the relative timing of excitation and inhibition largely explains the weaker response of LGN neurons to low frequency L + M modulations than to higher frequency (5–10 Hz) L + M modulations (Benardete and Kaplan, 1997a, 1999a) (Figure 6C).

Most parvocellular neurons with parafoveal RFs receive input from a single cone type to the center of their RFs and a mixture of L- and M-cones to the surround. For these neurons, L – M modulations invert the influence of the surround relative to the center. A parvocellular L-ON neuron, for example, is excited by an increase in L cone contrast at the center and is disinhibited by a decrease in M cone contrast in the surround (Figure 6D). When close together in time, these influences combine to drive a strong response (Figure 6E, top). When the temporal frequency of the modulation is sufficiently high that excitation from the center coincides with inhibition from the surround, the response is reduced (Figure 6E, bottom & 6F). These center-surround interactions explain the low-pass temporal frequency tuning of parvocellular neurons to L – M modulations (Benardete and Kaplan, 1999b; Lankheet et al., 1998).

The high-frequency roll-off of magnocellular and parvocellular responses is due largely to phototransduction, the dynamics of which depend on mean light intensity. Across a broad range of light levels, increasing
the mean intensity of a modulated light increases the speed of cone responses (Baudin et al., 2019) and retinal ganglion cell (RGC) responses (Purpura et al., 1990) and shifts the peak of the psychophysical temporal contrast sensitivity function to higher frequencies (De Lange Dzn, 1958). The ability to predict the shape of the high-frequency limb of the temporal contrast sensitivity function on the basis of the cone temporal impulse response across light levels suggests that that cortical filtering is independent of light level (Lee et al., 1990; Rider et al., 2019; Stockman et al., 2006).

Temporal filtering at the retinogeniculate synapse appears to be modest under most conditions (Alitto and Usrey, 2008; Benardete and Kaplan, 1997a, 1999b; Kaplan et al., 1987; Kaplan and Shapley, 1986). Many of the stimuli used in the current study had low contrast, making retinogeniculate transmission particularly efficient (Kaplan et al., 1987). High-frequency stimuli had higher contrasts, but the similarity in SNR of cone currents and LGN spiking responses at these frequencies suggests that information loss at the retinogeniculate synapse was minimal.

Mechanisms of SNR loss in the cortex

The SNR gap between the LGN and behavior is due, at least in part, to processes occurring in area V1 (Hawken et al., 1996). One mechanism that may contribute to high-frequency filtering in V1 is push-pull excitation-inhibition (Tolhurst and Dean, 1990). Simple cells in V1 receive spatially coincident excitation and inhibition that prevent high-contrast, non-preferred stimuli from driving a response (Troyer et al., 1998) and reduce sensitivity to high temporal frequency modulations (Krukowski and Miller, 2001; Krukowsk i et al., 2001). An intuition for the latter effect is that excitation and inhibition cancel when triggered simultaneously. The dominant inhibition required by the push-pull model ensures that cancellation is complete, and the slow kinetics of N-methyl-D-aspartate-sensitive channels in V1 neurons broaden the window of effective simultaneity (Eickhoff et al., 2007; Lester et al., 1990).

Most of the data supporting the push-pull model are from cat, but the same principles are likely at work in primates as well (Conway and Livingstone, 2006; Kremkow and Alonso, 2018). Monkeys have luminance-tuned simple cells, like cats do, but unlike cats, monkeys have a large population of cone-opponent V1 neurons. Some of these cone-opponent neurons combine visual signals antagonistically and roughly linearly across their RFs, consistent with the push-pull model (Conway and Livingstone, 2006; De and Horwitz, 2021). One possibility that is consistent with the results of this study is that push-pull excitation-inhibition reduces the SNR of high-frequency cone-opponent and non-opponent modulations similarly in V1.

Relationship to previous work

Two innovations set the current study apart from those previous. The first was holding fixed several factors between neurophysiological and behavioral measurements: the species and identities of the subjects, the intensity of the display background, the retinal eccentricity of the stimulus, and stimulus size. Two previous primate studies matched these parameters, but neither of them varied temporal frequency, and the one that varied color reported data from few neurons (Jiang et al., 2015a, 2015b; Sperling et al., 1978). A second innovation was the use of a cone current model to quantify information loss through the retina and retinogeniculate synapse (Angueyra and Rieke, 2013; Hass et al., 2015; Horwitz, 2020). Note that macaques are slightly more sensitive than humans to chromatic modulations under a range of conditions (Lindbloom-Brown et al., 2014; Stoughton et al., 2012) and are slightly less sensitive to low-frequency luminance modulations (Gelfand and Horwitz, 2018).

Results from this study are broadly consistent with those of Lee et al. (1990) who compared contrast detection thresholds of human observers to the responses of individual magnocellular-projecting (M) and parvocellular-projecting (P) RGCs. M RGCs responded strongly to luminance modulations and weakly to chromatic modulations. The reverse was true for P RGCs. Individual RGCs of both types were less sensitive than human observers at low frequencies and more sensitive at high frequencies. Results of the current study extend these observations by showing that the sensitivity of LGN populations and observers matches at low temporal frequencies, that the SNR of M and P populations is similar across temporal frequencies at contrast detection threshold, and that retinal circuitry is lossier than cortical circuitry except at high frequencies.

The idea that L – M and L + M temporal contrast sensitivity can be directly related to activity in the M and P pathways has been the subject of much debate. Single-unit recordings are ill-suited for settling this debate.
because, as shown in this study, many stimuli activate both pathways even at detection threshold. The only stimulus that achieved decisive pathway specificity in this study was the low temporal frequency, L – M stimulus, which modulated parvocellular neurons weakly but exclusively. Low-frequency L + M stimuli modulated magnocellular neurons more strongly than parvocellular neurons, but both populations carried measureable signal. At high frequencies, both magnocellular and parvocellular neurons responded briskly to L + M and L – M stimuli.

Spatial contrast sensitivity
Visual sensitivity under a range of conditions is bandpass for luminance contrast and low pass for chromatic contrast. Interestingly, this pattern is consistent whether modulations are temporal or spatial. A normative explanation is that L – M signals in natural scenes are small (Ruderman et al., 1998) but important (Carvalho et al., 2017; Rosenthal et al., 2018). Detecting these signals is facilitated by integration (low-pass filtering), a strategy that works over space or time due to the large, stationary nature of objects. L + M signals in natural scenes have greater amplitude, so they can be detected with less integration, permitting finer spatial and temporal resolution and the consequent benefits for visually guided action.

Some mechanisms underlying spatial and temporal visual filtering are shared. For example, low-frequency spatial and temporal modulations are filtered via center-surround RF antagonism (Robson, 1966), and high-frequency modulations are filtered via phototransduction (Cottaris et al., 2020). The spatial effects of phototransduction are linked to small eye movements produced during fixation. A small displacement of a high spatial-frequency grating can stimulate individual cone photoreceptors with contrast increments and decrements close together in time, causing cancellation.

Other mechanisms of spatial and temporal filtering differ, one of which is highlighted by the current results. The current study showed that the temporal filtering of luminance and chromatic modulations is similar in the cortex. In contrast, spatial filtering of luminance and chromatic modulations differs substantially. High spatial frequency luminance sensitivity is limited by midget ganglion cell density, implying near-perfect fidelity of cortical information transmission (Anderson et al., 1991; Banks et al. 1987, 1991; Dacey 1993). Chromatic spatial sensitivity, on the other hand, appears to be limited to a greater degree by cortical processes (Martin et al., 2001; Mullen and Kingdom, 2002; Mullen et al., 2005; Sekiguchi et al., 1993; Solomon et al., 2005) but see (Neitz et al., 2020).

Caveats
Several disparate data sets were converted to a common SNR metric to facilitate comparison across stages of the visual system. This conversion required mathematical models that could lead to erroneous conclusions if based on erroneous assumptions. The basis of each model, the approximations and assumptions made in their construction, and probable sources of error are discussed below.

The cone current model
The cone current model was based on patch clamp recordings from ex vivo macaque cones under light levels similar to those used in the current study (Angueyra and Rieke, 2013). The model approximates current noise as independent of the signal, which is reasonable at the moderate light levels used in this study (Figure 1 of Angueyra and Rieke, 2013). Cone signaling dynamics were approximated as independent of eccentricity, which is reasonable over the range investigated in this study (2–14°) (Sinha et al., 2017). Absolute detection thresholds predicted by this model are close to those measured behaviorally (Angueyra and Rieke, 2013; Koenig and Hofer, 2011).

Weaknesses of the model include the fact that it is based on a single, canonical temporal impulse response, noise spectrum, and cone distribution, all of which presumably vary across observers (cone distribution does; see Curcio et al., 1987). Indeed, results of this study provide indirect evidence for individual differences. LGN neurons in monkey 2 were more sensitive than those in monkey 1, relative to the cone model (Figure S). This was true for both magnocellular and parvocellular neurons, consistent with a systematic underestimate of cone sensitivity in monkey 2.

At the highest frequencies tested, magnocellular SNR slightly exceeded the SNR of simulated cone currents in monkey 2 (Figure 5B). This is unrealistic; SNR cannot increase between the cones and the LGN. The population model is not responsible for this discrepancy. The SNR of individual magnocellular neurons...
from monkey 2 exceeded the SNR of the simulated cones inside their RFs (Figure S4). The number of cones in monkey 2 or the high-frequency sensitivity of the simulated cones may have been underestimated. Cone current simulations were based on recordings made at 4,000–6,500 photoisomerizations per second, whereas cones in the monkeys’ eyes during the LGN recording experiments underwent approximately 7,400–8,800 photoisomerizations per second.

The LGN model

The LGN population model included correlations between neurons of a common type (magnocellular or parvocellular) but not between populations. Consequently, population SNR was computed for magnocellular and parvocellular populations separately. Computing SNR for both populations jointly requires additional assumptions that are ill-constrained by data.

The SNR of each LGN population is a lower bound on the SNR of both of them together. Note that this lower bound approached the theoretical upper bound imposed by the SNR of cone currents at high temporal frequencies (Figure 5). This leads to a prediction: the signals carried by magnocellular and parvocellular neurons with overlapping RFs are largely redundant in response to high temporal frequency modulations. This prediction is consistent with the idea that L – M signals carried by magnocellular neurons derive from the same circuits that mediate cone opponency in midget RGCs (Lee and Sun, 2009; Stockman et al., 2018). It is also consistent with the fact that the responses of midget and parasol RGCs with overlapping RFs share noise that is inherited from the photoreceptors (Ala-Laurila et al., 2011).

The behavioral model

The behavioral model was based on 13,760 detection trials from monkey 1 and 28,960 from monkey 2. Contrast sensitivity functions predicted from the model (Figure 1A) were similar to those from the literature and to those measured from human subjects performing the same task in the same testing apparatus (Gelfand and Horwitz, 2018; Merigan, 1980; Stavros and Kiorpes, 2008). The model accurately predicted contrast detection thresholds collected after the electrophysiological experiments (Horwitz, 2020). Error in estimated behavioral SNR was approximately 30% (see methods).

Conclusion

By comparing signal loss upstream and downstream of the LGN under identical conditions, this study showed that the differences between the luminance and chromatic temporal contrast sensitivity functions are due largely to processes upstream of the LGN.

Limitations of the study

Limitations of this study are discussed in detail in the section Caveats. The SNR of LGN spiking responses was estimated from single-unit responses and extrapolated to neuronal populations, a transformation that may be inaccurate. SNR estimates were obtained for magnocellular and parvocellular populations separately, and these estimates were not combined. Cone currents were simulated using a model based on cone light responses recorded ex vivo, a preparation which may have affected response properties.

Resource availability

Lead contact

Requests for further information, data, and analysis code should be directed to Greg Horwitz (ghorwitz@uw.edu).

Materials availability

This study did not generate new unique reagents.

Data and code availability

Data and code generated during this study are available at https://github.com/horwitzlab/LGN-temporal-contrast-sensitivity.

METHODS

All methods can be found in the accompanying transparent methods supplemental file.
SUPPLEMENTAL INFORMATION
Supplemental information can be found online at https://doi.org/10.1016/j.isci.2021.102536.

ACKNOWLEDGMENTS
I am deeply grateful to Fred Rieke and Juan Angueyra for valuable discussions, Charlie Hass for the MATLAB implementation of the cone population model, EJ Chichilnisky for feedback on the manuscript, Emily Gelfand and Lisa McConnell for excellent animal husbandry and training, and Zack Lindbloom-Brown for computer programming. This study was supported by NIH grants EY018849 and OD010425.

AUTHOR CONTRIBUTIONS
This is a single-author paper.

DECLARATION OF INTERESTS
The author declares no competing interests.

Received: October 19, 2020
Revised: March 2, 2021
Accepted: May 11, 2021
Published: June 25, 2021

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Supplemental information

Temporal filtering of luminance and chromaticity in macaque visual cortex

Gregory D. Horwitz
SUPPLEMENTAL INFORMATION

Transparent Methods

Experimental Model and Subject Details

Two macaque monkeys (M. mulatta, both male) were used in these experiments. Monkeys 1 and 2 were 13 and 7 years old, respectively, at the time of data collection. All procedures were approved by the University of Washington Institutional Animal Care and Use Committee. Monkeys 1 and 2 in the current study are monkeys 1 and 2 from Gelfand and Horwitz (2018) and from Horwitz (2020). The data analyzed in this report were collected contemporaneously with the data reported in Horwitz (2020) and partially overlap with this previous data set.

Method Details

Behavioral task

Behavioral detection thresholds were measured using a two-alternative, forced-choice contrast detection task (Gelfand & Horwitz 2018). Monkeys sat in a dark room, 61 cm away from a rear-projection screen that was illuminated by a digital light projector (Propixx, VPixx, Inc.) updating at 240 Hz. Stimuli were generated using routines from the Psychophysics Toolbox (Brainard 1997, Kleiner et al. 2007, Pelli 1997). The spectral power distribution of each display primary was characterized with a PR650 spectroradiometer (PhotoResearch, Inc.). The background was an equal-energy white metamer at 130 cd/m², producing approximate isomerizations per cone per second of L: 8900, M: 7400, and S: 2300.
The monkey initiated each trial by fixating a 0.2 x 0.2° spot at the center of the screen. An upward-drifting, horizontally oriented Gabor stimulus (1 cycle/° in a 0.15° standard deviation envelope) appeared either in the right or left hemifield. Stimulus contrast increased linearly over 166 ms, remained constant for 334 ms, and then ramped down over 166 ms. After a 100–600 ms delay, the fixation point vanished, and saccade targets appeared 2° to the right and left of fixation. The monkey received a liquid reward for making a saccade to the target in the hemifield in which the Gabor stimulus had appeared. Stimulus contrast, color direction in the LM plane, and temporal frequency varied across randomly interleaved trials, and stimulus location varied across days.

The duration of the stimuli used in this study exceeded psychophysical integration times, which are color- and temporal frequency-dependent (King-Smith & Carden 1976, Rovamo et al. 2003, Rovamo et al. 1996, Smith et al. 1984). A protracted stimulus was necessary to probe low temporal frequencies; low-frequency stimuli cannot be brief.

The data were fit with a model that predicts detection threshold as a function of temporal frequency, color direction in the LM plane, and location in the visual field. This model was used to adjust stimulus contrast to be near the monkeys' detection threshold (Gelfand & Horwitz 2018).

**Electrophysiology**

Responses of single LGN units were measured with extracellular tungsten microelectrodes (Fredrick Haer, Inc.) and recorded with a Multichannel Acquisition Processor system (Plexon, Inc.). Spike isolation was performed online with SortClient software and refined
offline with OfflineSorter software. Visual fixation was tracked with a scleral search coil (Riverbend Instruments, Inc.) and was required to remain in a 1 x 1° window. Liquid rewards were given for successful fixation. The visual display used in the electrophysiological experiments was identical to that used in the behavioral experiments.

*White-noise stimulation*

Each recorded LGN neuron was characterized with a white-noise stimulus that consisted of a 10 x 10 grid of 0.2° pixels. The light at each of these pixels was determined by independent, random draws from red, green, and blue Gaussian intensity distributions. The stimulus updated at 60 Hz (every four frames). Spike-triggered averaging was performed online to locate the receptive field of each recorded neuron and offline to classify it as magnocellular or parvocellular.

*Near-threshold Gabor stimulation of LGN neurons*

Following white noise characterization, each neuron was stimulated with a sequence of Gabor patterns centered on its receptive field. All stimuli were equated for detectability using the model of Gelfand and Horwitz (2018). Each neuron had a unique receptive field location and was therefore probed with a unique set of contrasts. Every stimulus modulated the L- and M-cones of the Stockman, MacLeod, and Johnson 10° standard observer with identical contrasts and did not modulate the S-cones. In randomly interleaved trials, L- and M-cone modulations were in phase, to create an L+M stimulus, or in counterphase, to create an L-M
stimulus.

Quantification and Statistical Analysis

LGN SNR calculation

Firing rate modulations of LGN neurons in response to the stimulus were nearly symmetric around the baseline rate. To quantify these responses, the modulation amplitude of LGN spike trains at the fundamental frequency of the stimulus was extracted from stimulus-present and -absent trials and compared. Both distributions of modulation amplitudes were standardized to make them approximately normal and to reduce their dependence on firing rate (Horwitz 2020). \( d' \) was defined as the difference between the means of these two distributions, divided by their pooled standard deviation. Neurometric sensitivity, the contrast at which \( d' = 1.27 \), was not measured.

Population \( d' \) was defined as the \( d' \) for an individual neuron multiplied by a population scale factor (Horwitz 2020). The population scale factor depends on the number of neurons of a given type, magnocellular or parvocellular, that are modulated by the stimulus and is defined as:

\[
\text{Population scale factor} = \frac{4\bar{w}^T \bar{\mu}}{\sqrt{4.2\bar{w}^T \Sigma_{\text{pop}}^{-1} \bar{w}}}. \tag{1}
\]

where \( \bar{\mu} \) is a vector of \( n \) signals, with one element per neuron in the population, \( \Sigma_{\text{pop}} \) is an \( n \times n \) covariance matrix representing noise in the population, and \( \bar{w} = \Sigma_{\text{pop}}^{-1} \bar{\mu} \) is the vector of optimal weights for population read-out. The 4 in the numerator represents the increase in signal obtained by pooling over ON and OFF mosaics in the two eyes. The 4.2 in the denominator
represents the increase in noise incurred through this same pooling and includes a 0.2 that represents additional noise due to anticorrelation between ON and OFF mosaics within each eye (Ala-Laurila et al. 2011, Greschner et al. 2011, Mastronarde 1989).

To calculate $\mu$, RFs were modeled as 2-dimensional Gaussian functions truncated at 2 standard deviations. A hexagonal mosaic of RFs was constructed so that each RF touched its six neighbors at the 1 standard deviation boundary (Gauthier et al. 2009). The signal carried by the $i^{th}$ neuron, $\mu_i$, was defined as the integrated product of the stimulus envelope and the $i^{th}$ RF. The RF in the center of the hexagonal array was assumed to correspond to the neuron that was actually recorded. $\Sigma_{pop}$, the noise covariance matrix, was constructed by assuming that every neuron was equally noisy and that correlations between neurons were equal to their RF overlap (Ala-Laurila et al. 2011, Trong & Rieke 2008). RF sizes of magnocellular neurons were taken from Derrington & Lennie (1984). Parvocellular RF sizes were taken from Watson (2014) with a 20% reduction in diameter to convert from human to macaque (Dacey & Petersen 1992).

**Behavioral SNR calculation**

Detection threshold was defined as the contrast needed to support 82% correct choices on the contrast detection task (Gelfand & Horwitz 2018). Decisions in this task can be modeled as comparisons between draws from two independent, homoscedastic, Gaussian distributions. If the draw from the signal distribution exceeds the draw from the noise distribution, the trial is answered correctly. 82% correct is achieved when the means of the two distributions are 1.27 standard deviations apart.
Noisy estimates of detection thresholds, on which this model was based, produce noisy SNR estimates. The average cross-validated prediction error of contrast detection thresholds was 14%. A 14% change in contrast around threshold corresponds to a change from 82% correct to 79–84%, assuming a Weibull psychometric function with a slope of 3 (Wallis et al. 2013). This range corresponds to d' values from 0.89–1.7.

Cone current SNR calculation

The cone current model was developed by Angueyra and Rieke (2013). The implementation used in this study is identical to the one used in (Hass et al. 2015, Horwitz 2020) and is available on GitHub (https://github.com/horwitzlab/LGN-temporal-contrast-sensitivity). Each cone is modeled as a linear temporal filter, the output of which (the signal), is corrupted by additive Gaussian noise. Simulated cone currents were weighted over time and space using a filter identical to the signal. This resulted in two univariate, homoscedastic Gaussian distributions from which d' was calculated (difference in means divided by standard deviation).

Spike train distance-based SNR analysis

The analysis of SNR presented in the main text assumes that LGN signals are at the fundamental frequency of the stimulus (F1). To examine the validity of this assumption, spike train power spectra were computed by discrete Fourier transform (Supplemental Figure 1). Parvocellular neurons, as expected, responded dominantly with an F1-modulated response component, whether the stimulus modulated the L- and M-cones in-phase (Supplemental
Figure 1A) or in counterphase (Supplemental Figure 1B) (Kaplan et al. 1990). The dip in power at approximately 5 Hz is a consistent aspect of parvocellular spike trains even in the absence of contrast in the receptive field (Horwitz 2020).

Magnocellular neurons carry signatures of stimulus frequency in components of their response besides the F1. Frequency-doubled responses (F2) to high-frequency L+M stimuli were pronounced (Supplemental Figure 1C). The magnitude of the F2 component was tightly correlated with the magnitude of the F1 component across neurons within temporal frequency (mean $r = 0.88$, $p < 0.0001$, paired t-test) and across temporal frequencies within neuron (mean $r = 0.81$, $p < 0.0001$, paired t-test). The information carried by the F1 and F2 components is therefore largely redundant. A broadband increase in power at high temporal frequencies is an expected consequence of the rectangular spike counting window and the increase in spike modulation amplitude with temporal frequency (Supplemental Figure 1C & 1D)(Harris 1978).

To ask whether magnocellular spike trains carried stimulus-related signals that were missed by the analysis of F1 modulation, an auxiliary analysis was performed. Each spike train was represented as a point in a high-dimensional space, and the distance between each pair of spike trains was defined on the basis of how many spikes must be added, deleted, or moved to transform one to the other (Victor & Purpura 1997). For each stimulus condition, individual spike trains were extracted and the nine nearest neighbors identified. If five or more of these neighbors were responses to the stimulus, then the extracted spike train was classified as stimulus-present; otherwise, it was classified as stimulus-absent. These classifications were compared to ground truth to quantify correctly classified stimulus responses (hits) and incorrectly classified responses to the blank (false alarms). The signal-to-noise ratio ($d'$) was
calculated as $\Phi^{-1}(\text{hit rate}) - \Phi^{-1}(\text{false alarm rate})$, where $\Phi^{-1}$ is the inverse cumulative standard normal probability density function, hit rate is the number of hits divided by the number of stimulus-present trials, and false alarm rate is the number of false alarms divided by the number of stimulus-absent trials.

The results of this analysis agreed closely with the results of the F1-based SNR analysis under most of the conditions tested (Supplemental Figure 2). The one exception was that SNR in response to high-frequency, L-M modulations in monkey 1 was considerably higher under the spike train distance-based analysis than under the F1-based analysis. In this animal, high-frequency L-M modulations often produced transient, weakly entrained responses (e.g. Figure 2B). For the comparisons made in this report, however, the assumption that the signal is carried in the F1 response component is justified.

**Spike train distance-based SNR analysis methods**

The calculation of SNR based on spike train distances involves two free parameters. The first is the number of nearest neighbors used in the classification, which was set to 9. The second is the penalty associated with moving a spike 1 by second, which was set to 6 (the penalty of adding or subtracting a spike was 1). The values of these two parameters were found via a grid search that maximized decoding accuracy.

Two minor adjustments were made to $d'$ values calculated by this procedure to reduce variability and bias. The first adjustment avoids infinite values that would otherwise occur when a hit rate or a false alarm rate was 0 or 1. Zeros were replaced with $0.5/n$ and ones were replaced with $(n-0.5)/n$, where $n$ is the number trials, either stimulus-present or -absent.
(Stanislaw & Todorov 1999). The second correction compensates for a small downward bias produced by the fact that each trial was classified on the basis of other trials, which are likely to be of the opposite type. Consider an urn containing equal numbers of red and black balls: the nearest neighbors of a red ball, not including itself, are more likely to be black than red. To correct for this fact, \( \Phi^{-1} (E(\text{hit rate})) - \Phi^{-1} (E(\text{false alarm rate})) \) was subtracted from \( d' \), where \( E(\text{hit rate}) \) and \( E(\text{false alarm rate}) \) are the expected values of the hit rate and false alarm rate under the ball and urn model.

**KEY RESOURCE TABLE**

| REAGENT or RESOURCE | SOURCE | IDENTIFIER |
|---------------------|--------|------------|
| Rhesus monkeys (Macaca mulatta) | Washington National Primate Research Center | N/A |
| MATLAB | Mathworks | [https://www.mathworks.com/products/matlab.html; RRID: SCR_001622](https://www.mathworks.com/products/matlab.html; RRID: SCR_001622) |
| Plexon Sort Client | Plexon, Inc. | [http://www.plexon.com; RRID: SCR:003170](http://www.plexon.com; RRID: SCR:003170) |
| Plexon Offline Sorter | Plexon, Inc. | [http://www.plexon.com; RRID: SCR:003170](http://www.plexon.com; RRID: SCR:003170) |
Supplemental Figure legends

Figure S1. Spectral analysis of LGN spike trains, related to Figure 2. Power spectra are shown for parvocellular responses to L+M stimuli (A), parvocellular responses to L-M stimuli (B), magnocellular responses to L+M stimuli (C), and magnocellular responses to L-M stimuli (D). The fundamental frequency of each stimulus is shown along the abscissa (triangles). Frequency-doubled responses in (C) are indicated by black arrows.

Figure S2. Magnocellular signal-to-noise ratio analysis, related to Figure 2. Signal-to-noise ratio (d') was computed in response to L+M modulations (black) and L-M modulations (magenta) at the fundamental frequency of the stimulus (closed symbols) and from the performance of a k-nearest neighbors spike train classifier (open symbols). Symbols represent means across neurons, and shaded bands represent ± 1 standard error of the mean. Data are from monkey 1 (A) and monkey 2 (B).

Figure S3. Analysis of spike counting window on population signal-to-noise ratio, related to Figure 4. Population signal-to-noise ratio was calculated from parvocellular responses to L-M modulations (magenta) and magnocellular responses to L+M modulations (black). Spikes were counted for 0.05, 0.1, 0.2, 0.4, or 0.66 s (columns), starting 0, 0.05, 0.1, 0.15, 0.2, 0.25, or 0.3 s after stimulus onset (rows). Symbols represent means across neurons, and shaded bands
represent ±1 standard error of the mean. Horizontal lines indicate signal-to-noise ratios of 0 (gray) and 1.27 (dashed).

**Figure S4.** Signal-to-noise ratio of individual LGN neurons and the cones assumed to be inside their receptive fields, related to Figure 5. Plotting conventions are as in Figure 5, but population scaling has not been performed on the individual neuronal d' values, and only the cones inside the receptive field of each recorded LGN neuron were simulated.
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A

Monkey 1

B

Monkey 2

Frequency (Hz)

Signal-to-noise ratio (d')

F1-based (L+M)
Spike train dist. (L+M)
F1-based (L-M)
Spike train dist. (L-M)

Frequency (Hz)
Duration of spike counting window

Population signal-to-noise ratio (d')

Start time of spike counting window

Frequency (Hz)
