Peripheral and Central Factors Limiting the Development of Contrast Sensitivity in Macaque Monkeys

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The aim of this study was to evaluate the contribution of peripheral and central factors to the development of visual sensitivity. We used the approach of (Pelli, 1981, 1990) to evaluate the hypothesis that intrinsic noise is high in infants compared with adults, and therefore sets an important limit on contrast sensitivity in infants. We measured contrast thresholds in the presence of various levels of dynamic spatiotemporal broadband noise in infant monkeys, and evaluated the developmental changes in contrast threshold and intrinsic noise. Our data show that intrinsic noise is high in infants and falls with contrast threshold during development. However, contrast thresholds in high-contrast noise also fall during development, although by a smaller amount. Therefore, while changes in intrinsic noise set an important limit on the development of contrast sensitivity across spatial frequencies, changes in non-additive sources of noise also contribute, particularly at high spatial frequencies. We interpret these results in terms of Pelli’s hypothesis about the sources of additive and non-additive noise affecting visual detection. In these terms, additive noise reflects peripheral factors and non-additive noise reflects central ones. Our results suggest that changes in peripheral sources of noise represent an important limit for the development of visual sensitivity. © 1997 Elsevier Science Ltd

Contrast sensitivity  Intrinsic noise  Visual development  Monkey

INTRODUCTION

Contrast sensitivity and spatial resolution are immature in newborn primates. These visual functions improve approximately 10-fold over the first year after birth in macaque monkeys (Boothe, Kiorpes, Williams, & Teller, 1988; Kiorpes, 1992). Recently, there has been increased interest in understanding the factors that limit visual sensitivity in newborn infants and determine the subsequent developmental time course. While there are changes in the optics of the eye that will lead to small improvements in performance, it is generally agreed that changes in the nervous system are of far greater importance. Very marked changes in photoreceptor outer segment morphology have been detailed by Yuodelis & Hendrickson (1986). Calculations based on these changes suggest that infant photoreceptors are substantially less efficient at capturing light than adult ones. Banks & Bennett (1988), Brown, Dobson, & Maier (1987), Brown (1993), and Wilson (1988) have recently evaluated these and other factors limiting infant visual development. These studies concluded that while one would expect an overall improvement in quantum efficiency during development in the infants, this expected improvement is not sufficient to fully account for the extent of the change in infant contrast sensitivity; changes in the signalling properties of the neurons carrying photoreceptor signals through the nervous system must also be involved. It is therefore desirable to try to learn the degree to which changes in peripheral and central mechanisms limit visual performance during development.

Psychophysical approaches based on noise masking measurements may help us understand the limitations on sensitivity by dividing them into peripheral (early visual pathways) and central factors. This idea was proposed by Barlow (1977) to characterize the efficiency of the human visual system, and was elaborated by (Pelli, 1981, 1990), Burgess, Wagner, Jennings, & Barlow (1981), and Kersten (1984), among others. The scheme proposed by Pelli (1990), under certain assumptions, attributes overall visual efficiency to the combined action of two stages, one presumed to correspond to the input stage of visual processing and the other to elements later in the visual
system. Pelli based his arguments on the well-known additive form of results from contrast detection experiments performed in masking noise, in which squared contrast thresholds rise in direct proportion to the squared contrast of added noise, plus a constant. From this constant, Pelli deduced the amount of additive intrinsic noise that affects contrast detection, expressed in units equivalent to the added noise. We have followed this scheme and similarly use “equivalent noise” to estimate the level of early “intrinsic” noise in the visual pathway. By measuring intrinsic noise and comparing its magnitude with visual sensitivity in young monkeys of different ages, we hoped to evaluate the hypothesis that infant contrast threshold is high as a result of high levels of intrinsic noise. The extent to which contrast thresholds and intrinsic noise fall together during development may establish the contribution of peripheral factors to sensitivity in young animals. In addition to intrinsic noise estimated in this way, there must also be non-additive “central” factors that affect thresholds; Pelli’s analysis derives these from thresholds measured at high noise levels. To the degree that these thresholds fall during development, central, non-additive factors may present the most important limit.

We measured contrast thresholds in dynamic visual noise in infant monkeys, and evaluated the developmental changes in contrast threshold and intrinsic noise. Our data show that intrinsic noise is high in infants and falls during development, suggesting that peripheral factors are an important limit on the development of visual sensitivity. Contrast thresholds measured in high levels of noise fall as well, however, so we conclude that central changes must also be involved. Some of these results have been presented previously (Handel, Kiorpes, & Movshon, 1993; Kiorpes, Skoczenski, & Movshon, 1995).

**METHODS**

**Subjects**

Thirteen young *Macaca nemestrina* monkeys were subjects in this study. All animals were born at the Washington Regional Primate Research Center, and were hand-reared in the Visual Neuroscience Laboratory at New York University. All animal care conformed to guidelines approved by the New York University IACUC and the NIH Guide for Care and Use of Laboratory Animals. The animals ranged in age from 1 to 18 months at the time of testing. Their visual environment was a normal laboratory environment, which was enriched with a wide variety of appropriate visual and tactile stimuli. The animals were also given daily opportunities for interaction with other monkeys and humans.

**Stimuli**

Stimuli were presented on a Nanao T660i monitor with a mean luminance of 56 cd/m². The display subtended 36 deg at 60 cm, the viewing distance used for young infants. Typically, animals older than 1 year were tested at distances of 100–120 cm. Stimulus presentation was controlled by a computer via an ATVista graphics board (Truevision). We used patches of sinusoidal grating vignetted by a 2D spatial Gaussian, whose contrast was ramped on over 200 msec. Grating spatial frequency ranged from 0.5–4.2 cyc/deg. The standard deviation of the spatial Gaussian was 3 deg. Once the grating reached full contrast (after the 200 msec ramp which was accompanied by a tone), the animal was free to respond. Since the animals were freely viewing, the viewing duration was not controlled precisely. However, response latencies, when measured, were typically about 500 msec. We did not notice a tendency for younger animals to view the display for different durations than older animals did; in any case, variations in viewing duration around 500 msec have little effect on sensitivity and would not be likely to have a significant effect on the results. The grating patches were presented alone and in the presence of random spatiotemporal broadband noise. The noise was refreshed with a new random noise field at a rate of 53 Hz and was interleaved on alternate frames with the grating. The noise was continuously present and filled the entire display. Noise pixel size was 11.6 min (at 60 cm); there were 188 × 144 noise pixels across the screen. Noise Michelson contrast ranged from 0.01–0.50. In some early experiments we used noise in which the luminance distribution of pixel values was uniform, but for most experiments we used binary noise (i.e., noise whose pixels took on one of two values), because binary noise provides the greatest contrast energy at a given physical contrast. The details of the algorithm used for noise generation can be found in Gegenfurtner & Kiper (1992).

**Behavioral methods**

On each trial, a grating patch was presented on either the left or right side of the video display, 10 deg from center. The monkey’s task was a spatial two-alternative forced-choice; she indicated on which side of the display the grating patch had appeared. For very young infants (generally those younger than 12 weeks), we used a procedure that we call “reinforced preferential looking”. The procedure combines preferential looking and operant techniques. The animals were placed in a cage with a facemask mounted on one wall (Williams, Boothe, Kiorpes, & Teller, 1981). Their looking behavior was monitored via a video camera directed at the eyehole of the facemask. During training, the infant monkeys were rewarded for directing their gaze toward a 2 × 2 deg square of flickering, random noise located in the center of an otherwise blank display. A human observer watched the animal’s looking behavior on a video monitor connected to the camera. When the animal oriented toward the center of the display, a grating patch was presented to the left or right of center. The animal indicated the side on which the grating had appeared with a leftward or rightward directed eye movement. The human observer, who was blind as to the side of stimulus presentation, made a response on the computer keyboard.
to indicate the direction of the monkey’s eye movement. The monkey was rewarded with a squirt of infant formula or apple juice for the observer’s correct responses. Once the animal was trained to reliably detect and indicate the side of stimulus presentation, she was allowed to freely view the display. The flickering square was replaced with a small cross that could be used to re-orient the animal to center if necessary. We saw no sign that the animals used peripheral vision to detect the stimulus; the infants typically scanned the display looking for the stimulus, then maintained their gaze on one side or the other to indicate their response. Older animals were trained to pull one of a pair of grab bars located on the front of the cage to indicate the side of stimulus presentation. They freely viewed the display and were rewarded for correct bar pulls; errors were signalled by a tone. Further details of these procedures may be found in earlier reports (Kiorpes, Kiper, & Movshon. 1993; Kiorpes & Kiper, 1996).

Contrast threshold for each stimulus condition was established using the method of constant stimuli. Each threshold was based on at least 200 trials; we collected 50–100 trials at each of 3–5 contrast levels chosen to span the performance range from 50–100% correct. Contrast threshold for each spatial frequency was measured in the absence of noise and either in the presence of each of a series of noise masks, or in some cases in a mask of the highest noise contrast (0.50). Data collection was counterbalanced across noise contrast level. Threshold estimates and standard errors were calculated using Probit analysis (Finney, 1971) of the log-transformed data sets.

RESULTS

It is generally accepted that grating detection is mediated by mechanisms that are selective for both spatial and temporal frequency. For this reason, the contrast of the noise is not the appropriate measure of its strength; that is best given in units of spectral density, or contrast power per unit bandwidth, which is the expected contrast power of a sample of the noise in a unit interval of spatiotemporal frequency. We calculated the spectral density of the noise for each set of spatiotemporal characteristics used. Within any given set of conditions, there is, of course, a consistent relationship between noise spectral density and noise contrast. That relationship, for the binary noise conditions used throughout most of this study, is shown by comparing the upper and lower abscissas in Fig. 1. The upper abscissa shows the spectral density of the noise; the lower abscissa shows the corresponding noise contrast.

Figure 1 illustrates the variation in contrast threshold with the strength of the masking noise; the isolated points on the left side of the plot indicate the measured contrast threshold for each stimulus in the absence of noise (unmasked contrast threshold). As expected (Pelli, 1990), at low levels of added stimulus noise, thresholds were little affected and the function was flat. This is the range over which intrinsic noise exceeds stimulus noise. At high noise levels, thresholds rise in proportion to noise contrast and the function has a slope of 1. This behavior is well known and is described by the relationship:

\[ C = R_{sn} \sqrt{N^2 + N_{eq}^2} \]  

(1)

where \( C \) is contrast threshold, \( N \) is the contrast power of the masking noise, \( R_{sn} \) is a constant giving the observer’s overall signal-to-noise ratio, and \( N_{eq} \) is the “equivalent” intrinsic noise, referenced back to the visual scene in units equivalent to the masking noise power. When \( N = N_{eq} \), contrast threshold is elevated by \( \sqrt{2} \); this point is indicated on the abscissa with arrows for each curve. For the remainder of the paper, we will give noise strength in units related to the contrast of our “standard” noise (binary noise of our standard pixel size and frame rate), rather than the spectral density of the noise, since these units are more intuitive for most readers. For values given in contrast units we will use the terms \( N_C \) and \( N_{eq} \) in place of \( N \) and \( N_{eq} \). The constant \( R_{sn} \) gives the system’s overall signal to noise ratio at threshold, expressing the threshold contrast as a proportion of the summed intrinsic and masking noises. Intuitively, \( R_{sn} \) determines the vertical position of the masking function, while \( N_{eq} \) determines its horizontal position.

As noted, throughout most of the study we used dynamic noise with a binary amplitude distribution. In some early cases, we used dynamic noise with a uniform amplitude distribution. Since the root mean square contrast of the amplitude distribution of the noise, presumably its masking strength, is given by the standard deviation of the amplitude distribution, and since uniform noise has lower standard deviation than binary noise, we compared equivalent noise contrast measured using

![Figure 1](attachment:image.png)

**FIGURE 1.** Representative masking functions for a 16-week-old monkey at two spatial frequencies, 1 and 4 c/deg (squares and triangles, respectively). The variation in contrast threshold is plotted as a function of masking noise contrast; noise spectral density is shown on the upper abscissa. The isolated data points are the measured unmasked thresholds for each spatial frequency. The open and filled arrows point to equivalent noise contrast (\( N_{eq} \)) for the 1 and 4 c/deg data sets, respectively.
uniform and binary noise distributions in three animals as a control. We calculated $N_{eq}$ for each data set using the spectral density of the noise distribution and then scaled the values to be equivalent in contrast for our standard noise. Our estimated $N_{eqC}$ was similar for the different noise conditions, so we included data collected using uniform noise in the population data. As stated above, in most cases we measured contrast threshold at each of a series of noise contrast levels, as well as in the absence of noise, as illustrated in Fig. 1. In some cases, we measured contrast threshold only in the absence of noise and in the presence of the highest noise contrast (0.50), to obtain a rapid estimate of equivalent noise contrast. Figure 2 illustrates the reliability of estimates of equivalent noise contrast based on only the end points of the series as opposed to the full series. These data are from nine animals tested at several ages and spatial frequencies. It is clear that the estimates of equivalent noise contrast are very similar, whether based on only the no-noise and highest noise points or based on the full series. Thus, we included estimates of equivalent noise contrast based only on two points in cases where there were no full series data at a given age; this was the case for most of the data presented for 2.1 c/deg.

To be sure that the theory and assumptions behind the noise masking paradigm were applicable to the monkey it was important to verify that the behavioral data conformed to the expected function. While this is apparent from individual data sets such as those shown in Fig. 1, it is important to establish whether the data as a whole showed any systematic variation from the expected function. Figure 3 shows that there was not. The data points are from multiple data sets collected from 10 monkeys at various ages. Each data set was fit by equation (1), and the data were then shifted horizontally and vertically on these log–log axes to make the fitted values of unmasked threshold and $N_{eqC}$ equal 1. The data are consistently and uniformly close to the standard curve, showing that equation (1) provided a good description of the data.

While all of our data fit the template well, the robustness of our estimate of intrinsic noise depended on the degree of masking observed. We numerically simulated the accuracy with which $N_{eq}$ was determined by the data. When $N_{eq}$ is less than 25–50% of the maximum available noise contrast, it is estimated with roughly $\sqrt{2}$ less reliability than the contrast threshold. When $N_{eq}$ approaches the maximum noise contrast, the reliability becomes appreciably worse. In cases where we did not have sufficient noise contrast to mask threshold, for example, some very young infants viewing high spatial frequency stimuli, we did not attempt to estimate $N_{eq}$. The results of these numerical simulations are reflected in our calculations of slopes involving $N_{eq}$ (see below).

Developmental data for two monkeys are shown in Fig. 4. Each panel is similar in format to Fig. 1. The open symbols in each case represent data at one young age and filled symbols represent data collected at one older age from the same animal under the same test conditions. Masking functions are shown for each animal at two spatial frequencies: 1 (upper panels) and 4 c/deg (lower panels). The arrows point to equivalent noise contrast ($N_{eqC}$) for each function. At the lower spatial frequency for these animals, 1 c/deg, there was little change in unmasked contrast threshold between the test ages (isolated points to the left of the curves) and there was little difference in equivalent noise contrast for the data sets (compare open and filled arrows). At the higher spatial frequency, there was a substantial decrease in unmasked contrast threshold between the younger and older test ages for both monkeys, and there was an obvious concurrent change in equivalent noise contrast. This result suggests that intrinsic noise, as estimated by equivalent noise contrast, changes during development as
does contrast threshold. However, it is important to note that the change in unmasked contrast threshold is not proportional to the change in equivalent noise contrast. If it were, each pair of curves would come together at high noise contrasts. Instead, the functions rise in parallel with one another but do not meet.

As many previous experiments have shown, contrast threshold decreases during development. The change in unmasked contrast threshold with age for the population of monkeys in this study is shown in Fig. 5 for all spatial frequencies tested. The data for each spatial frequency are plotted against a separate ordinate for clarity, with the lowest frequency at the top and the highest at the bottom. As expected, contrast threshold decreased linearly with log age. The overall extent of the improvement in threshold was about 1 log unit for the low spatial frequencies, but was greater for the highest frequency.

We performed a regression analysis to compare the change in threshold with age across spatial frequencies. This analysis showed that the slopes of the functions relating log contrast threshold and log age were close to −1 for 0.55 and 4.2 c/deg, but were somewhat lower for 1.0 and 2.1 c/deg. Table 1 presents regression slopes and correlation coefficients (in parentheses) for each comparison.

We next evaluated the change in equivalent noise contrast with age. Figure 6 shows the change in $N_{eqC}$ with age for the same population as Fig. 5. There was a more gradual change in equivalent noise contrast than in unmasked threshold with log age for all spatial frequencies. The regression analysis showed that the slopes of the functions relating log equivalent noise contrast and log age were lower overall compared with the change in threshold; they were close to −0.3 for all spatial frequencies except 0.55 c/deg (see Table 1).

These comparisons suggest, as did the individual data

<table>
<thead>
<tr>
<th>Spatial frequency (c/deg)</th>
<th>0.55</th>
<th>1.0</th>
<th>2.1</th>
<th>4.2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age/threshold</td>
<td>−0.91 (0.87)</td>
<td>−0.63 (0.81)</td>
<td>−0.65 (0.74)</td>
<td>−0.97 (0.87)</td>
</tr>
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<td>Age/$N_{eqC}$</td>
<td>−0.76 (0.75)</td>
<td>−0.28 (0.46)</td>
<td>−0.38 (0.53)</td>
<td>−0.39 (0.57)</td>
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<tr>
<td>Threshold/$N_{eqC}$</td>
<td>1.15 (0.77)</td>
<td>1.54 (0.71)</td>
<td>1.37 (0.81)</td>
<td>1.91 (0.79)</td>
</tr>
<tr>
<td>Age/$R_{in}$</td>
<td>−0.18 (0.27)*</td>
<td>−0.35 (0.64)</td>
<td>−0.28 (0.53)</td>
<td>−0.57 (0.81)</td>
</tr>
<tr>
<td>Threshold/$R_{in}$</td>
<td>3.66 (0.38)*</td>
<td>1.99 (0.62)</td>
<td>2.55 (0.59)</td>
<td>1.83 (0.80)</td>
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in Fig. 4, that both contrast threshold and intrinsic noise decrease during development, but not to the same extent. To evaluate the relationship between these measures, we plotted unmasked contrast threshold against equivalent noise in Fig. 7. The variation in contrast threshold with $N_{eqC}$ is shown for each spatial frequency, for all monkeys and ages, in the same format as Fig. 5. Clearly there was a strong correlation between unmasked contrast threshold and $N_{eqC}$ for all spatial frequencies. However, for all except for the lowest spatial frequency, the relationships appear to have a slope greater than 1 (dashed lines). A slope of 1 would be expected if changes in threshold were wholly due to changes in intrinsic noise. Table 1 includes slope estimates for these data sets, calculated taking into consideration estimation errors in both threshold and equivalent noise contrast (see Press, Teukolsky, Vetterling, & Flannery, 1992). Indeed, for all frequencies except 0.55 c/deg, these slopes are substantially greater than 1.

Thus, it seems that changes in intrinsic noise, as measured by equivalent noise contrast, account for some, but not all, of the variation in contrast threshold during development. The remaining variance may be captured by changes in threshold in the presence of high masking noise contrast, i.e., the signal to noise ratio measured by the constant $R_{sn}$ in equation (1). As noted earlier, the upper portion of the masking functions typically did not converge as the functions shifted with age (refer to Fig. 4). This suggests that there was also a change in contrast threshold in high noise with age. Figure 8 shows the
variation in signal-to-noise ratio ($R_{sn}$) as a function of age for each spatial frequency. At the low spatial frequencies these data appear quite flat, suggesting that there was only a slight variation in signal-to-noise ratio with age. But this was not the case at 4.2 c/deg. There was a substantial change in signal-to-noise ratio with age at this higher frequency; in fact, this was quantitatively more important than the change in equivalent noise contrast (see Table 1). To examine the relationship between signal-to-noise ratio and unmasked contrast threshold for each spatial frequency, we plot threshold as a function of $R_{sn}$ in Fig. 9. Because there was little variation in $R_{sn}$ for the lower spatial frequencies, the slopes for these data sets were quite steep (see Table 1) indicating little covariation. At 0.55 c/deg there was in fact no significant correlation between signal-to-noise ratio and either unmasked threshold or age. On the other hand, there was greater variation in $R_{sn}$ at 4.2 c/deg, and the relationship between signal to noise ratio and unmasked threshold is clear. Again, Table 1 includes slope estimates for these data sets, calculated taking into consideration estimation errors in both threshold and signal-to-noise.

**DISCUSSION**

Our aim was to deduce from psychophysical measurements the extent to which peripheral and central factors limit the development of spatial contrast sensitivity. Our results show that intrinsic noise is high in young infants and falls with contrast threshold during development. This result is consistent with the conclusions of Brown (1994), who estimated intrinsic noise in human infants using a luminance discrimination paradigm. That intrinsic noise is higher in infants than adults suggests that there are important peripheral limits on contrast sensitivity in young infants. However, our data also show that thresholds at high noise levels fall during development, particularly at 4.2 c/deg, the highest spatial frequency we tested. This result is summarized in Fig. 10, and supports the notion that central factors must also contribute to contrast sensitivity development.

It is important to consider first what is measured by changes in $N_{eqC}$ and $R_{sn}$ ("intrinsic noise" and "signal to noise ratio"). Formally, equation (1) describes masking functions in terms of an additive and a multiplicative constant ($N_{eqC}$ and $R_{sn}$, respectively). We have followed Pelli's (1990) logic in attributing $N_{eqC}$ to peripheral factors and $R_{sn}$ to central ones, but it should be clear that this distinction may not be absolute. There is good reason to believe that noise in the peripheral elements of the visual pathway is approximately additive (e.g., Baylor, Lamb, & Yau, 1979; Croner, Purpura, & Kaplan, 1993; Troy & Lee, 1994), and it is also clear that non-additive factors contribute extensively to the variability measured in neurons in the central visual pathways (e.g., Tolhurst, Movshon, & Thompson, 1981, Tolhurst, Movshon, & Dean, 1983; Vogels, Spileers, & Orban, 1989; Britten, Shadlen, Newsome, & Movshon, 1993; Shadlen & Newsome, 1994). It is, however, going beyond the
available data to assign all additive factors to the periphery and all non-additive factors to central mechanisms, and to the degree that this assignment turns out to be incorrect our reasoning will need to be modified.

Our data do suggest that $N_{eqC}$ and $R_{sn}$ are to some degree independent and, therefore, that they measure different underlying processes. The best evidence for this is that the two parameters show quite different patterns of change with age at different spatial frequencies (Figs 6 and 8; Table 1), with $N_{eqC}$ showing the most relative variation at low spatial frequencies and $R_{sn}$ showing the most relative variation at the highest spatial frequency. At the lowest spatial frequency (0.55 c/deg), the changes in contrast threshold appear to be well-accounted for by changes in $N_{eqC}$, whereas at the highest spatial frequency, the changes in threshold are more strongly associated with changes in $R_{sn}$. If we consider the factors that relate most clearly to the changes in unmasked threshold, the simple correlations support the idea that $R_{sn}$ is relatively more important at 4.2 c/deg than at other frequencies (see Table 1), while $N_{eqC}$ has essentially uniform weight across spatial frequencies.

Our results at low spatial frequencies are consistent with those of Brown (1994), who measured the relative contribution of intrinsic noise to contrast discrimination thresholds in human infants and adults. Brown found a significant change in intrinsic noise between infants and adults using 0.55 c/deg grating stimuli, and a much smaller change in contrast-dependent, non-additive noise ($NC_1$ in Brown, 1994), which is comparable with our factor $R_{sn}$. It is not known whether infant humans would show a similar shift in the influence of additive and non-additive factors at higher spatial frequencies as we found in the monkeys.

It follows from the logic spelled out above that one interpretation of this result is that peripheral changes account for the majority of sensitivity development at low spatial frequencies, while central changes have greater weight at higher frequencies. Recall that during development contrast thresholds fall more at higher spatial frequencies than at lower ones (Fig. 5; Boothe et al., 1988). But inspection of Fig. 6 shows that $N_{eqC}$ changes similarly at low and high spatial frequencies. It thus seems reasonable to suggest that in the absence of changes in central factors, sensitivity would improve uniformly at all spatial frequencies, and that the enhanced development of sensitivity at high spatial frequencies is due to changes in some central process.

What might the peripheral factor(s) limiting sensitivity be? Several candidate peripheral mechanisms have been proposed and analyzed previously (Banks & Bennett, 1988; Jacobs & Blakemore, 1988; Brown et al., 1987; Brown, 1994; Allen, Bennett, & Banks, 1992; Wilson, 1993; Movshon & Kiorpes, 1993). Although such factors as growth of the eye and improvements in optical quality make modest contributions, the most important changes are likely to occur in the photoreceptors themselves, and in particular in the development of their outer segments into the efficient light-gathering devices seen in adults (Yuodelis & Hendrickson, 1986; Packer, Hendrickson, & Curcio, 1990). Such changes, like the changes in $N_{eqC}$, would have similar effects at all spatial frequencies within the limit of retinal sampling. Therefore, we tentatively suggest that the contribution of changes in $N_{eqC}$ to the development of sensitivity depends on this early peripheral mechanism. In pursuit of this idea, we are presently developing an ideal observer model for the infant monkey, with which we expect to be able to analyze quantitatively the effects of factors like photoreceptor density and efficiency on ideal contrast sensitivity, and relate the results to our behavioral data.

What central mechanism(s) may be important for contrast sensitivity development is far from clear. One possibility is that the important “central” factor is in fact to be found at the level of the lateral geniculate nucleus. We and others have suggested that neurons at the level of
the LGN set an important limit on the normal development of acuity (Movshon & Kiorpes, 1993; Blakemore & Vital-Durand, 1986). However, recent electrophysiological measurements in the LGN of infant macaque monkeys suggest that there are relatively modest postnatal changes in spatial and temporal resolution, and that infant LGN neurons show a rather adult-like contrast response (Hawken, Blakemore, & Morely, 1997; Movshon, Kiorpes, Hawken, Skoczynski, Cavanaugh, & Graham, 1997). These and earlier studies do consistently show an overall improvement in neuronal responsiveness postnatally, a change which would be expected to have a similar effect across spatial frequencies.

Another possibility is simply that the central factors represent the strength and reliability of central synaptic connections in the visual system. There is ample anatomical and physiological evidence that these connections undergo extensive postnatal development in the visual cortex (see Daw, 1995). The changes in \( R_{an} \) could simply reflect increases in synaptic gain or decreases in neuronal noise. But it is also possible that these effects are produced by changes in the spatial scale of the retinocortical projection produced by eye growth and photoreceptor migration (Packer et al., 1990), or by changes in the fineness of spatial selectivity in spatial-frequency selective elements in visual cortex (Blakemore, 1990; Chino, Smith, Hatta, & Han, 1997). As a consequence of the refinement of receptive field structure that may underlie the development of high spatial-frequency selectivity, the elements sensitive to a given spatial frequency, say 4 c/deg, in infants may not be the same as in adults. For example, assuming that, as photoreceptors migrate and the fovea develops its adult photoreceptor density, the central connections formed early in development are maintained, the best spatial frequency for a given foveal V1 neuron will increase. The relatively different pattern of change we see in \( R_{an} \) at the highest frequency may be reflecting such shifts. Ultimately, establishing which factors are the most important will require analysis of the physiological properties of neurons at several levels of the developing visual system, a project on which we have already embarked (Movshon, Hawken, Kiorpes, Skoczynski, Tang, & O’Keefe, 1994; Movshon et al., 1997; Skoczynski, O’Keefe, Kiorpes, Tang, Hawken, & Movshon, 1995).

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