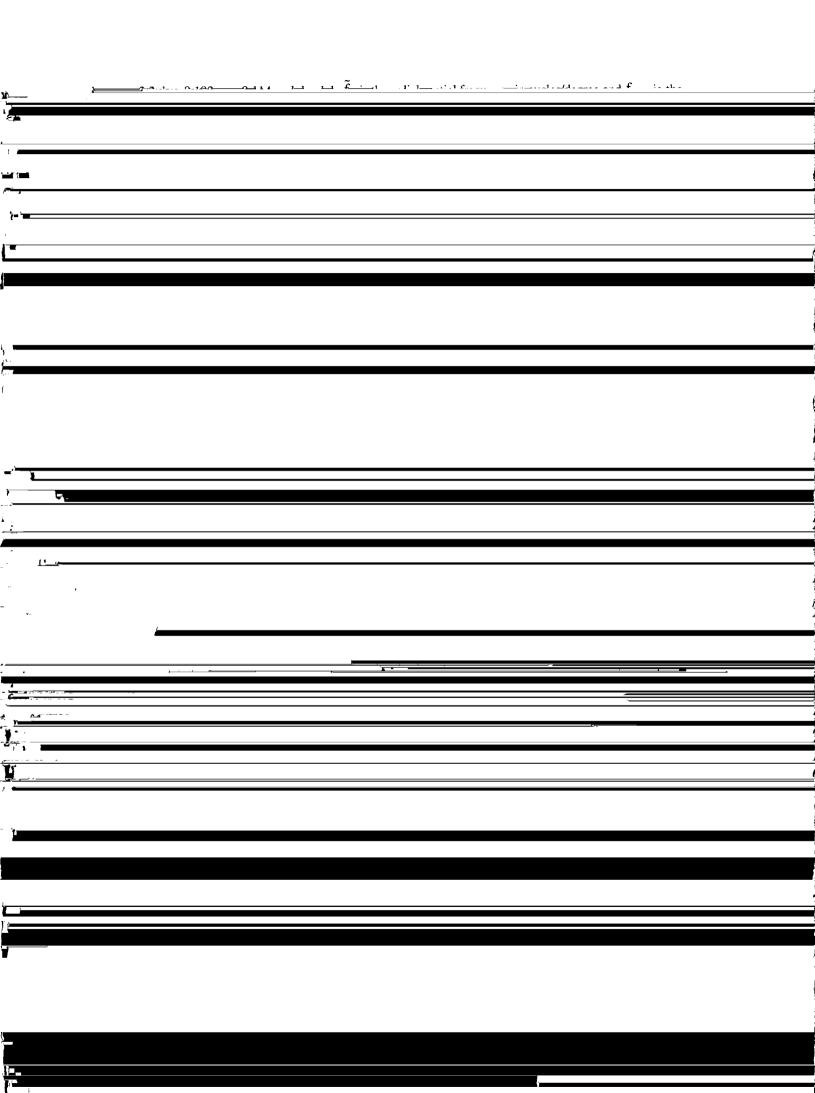
An Analysis of the Blue Noise Mask Based on a Human Visual Model Department of Electrical Engineering **ABSTRACT** The Blue Noise Mask (BNM) is a stochastic screen that produces visually pleasing blue noise. In its construction, a



$$\begin{split} E^{'2} &= \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \left[f(i,j) + h \left((i-i_0)_{mod \ N}, (j-j_0)_{mod \ N} \right) - g' \right]^2 \\ &= \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \left[f(i,j) - g' \right]^2 + \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} h^2 \left((i-i_0)_{mod \ N}, (j-j_0)_{mod \ N} \right) \\ &+ 2 \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} f(i,j) h \left((i-i_0)_{mod \ N}, (j-j_0)_{mod \ N} \right) \\ &- 2 \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} g' h \left((i-i_0)_{mod \ N}, (j-j_0)_{mod \ N} \right) \end{split}$$

$$i=0$$
 $j=0$ (8)

There are four terms on the right side of the above equation. Since g' is a constant and f(i,j) is known, the first term in equation (8) is fixed. The second term is a summation of the shifted filter squared over the support of the BNM. Due to the "wrap-around" property, this term is constant. For the same argument, the fourth term is also a constant. Let

$$\int_{1}^{\left(\frac{1}{2}-\frac{1}{2}\right)} \sum_{i=1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} \sum_{j=1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{N-1}{2}\right) di_{i} = \frac{1}{2} \int_{1}^{N-1} e(i + j) h_{i}\left(\frac{1}{2}-\frac{$$

where $G(i_0,j_0)$ is a function of i_0 and j_0 . It follows that minimizing E'^2 is reduced to minimizing $G(i_0,j_0)$.

Substituting (3) into (9) we have:

$$G(i_0, j_0) = \sum_{i=0}^{N-1} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} \sum_{m=0}^{N-1} h((i-m)_{\text{mod } N}, (j-n)_{\text{mod } N}) b(m, n) h((i-i_0)_{\text{mod } N}, (j-j_0)_{\text{mod } N})$$
(10)

$$R(m, n, i_0, j_0) = \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} h(k, l) h((k + (m - i_0))_{\text{mod } N}, (l + (n - j_0))_{\text{mod } N})$$
(15)

which is the circular autocorrelation (for simplicity, we will use "autocorrelation" for "circular autocorrelation" from now on) of h at $(m-i_0,n-j_0)$ and can be denoted as $R_h((m-i_0)_{mod\,N},(n-j_0)_{mod\,N})$. Substituting into (11) and making use of the symmetry of the autocorrelation function, we obtain:

$$G(i_0, j_0) = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} b(m, n) R_h((i_0 - m)_{\text{mod } N}, (j_0 - n)_{\text{mod } N})$$
(16)

In the frequency domain, R_h corresponds to $|H|^2$. It can be easily seen that $G(i_0, j_0)$ is the circular convolution of the binary pattern for level g with a new filter that is the autocorrelation function of h.

Examination of equation (11) tells us that to minimize $G(i_0, j_0)$, or the perceived error for level g' measured by the

