An Improved Detection Model for DCT Coefficient Quantization

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ABSTRACT

A detection model is developed to predict visibility thresholds for discrete cosine transform coefficient quantization error, based on the luminance and chrominance of the error. The model is an extension of a previously proposed luminance-based model, and is based on new experimental data. In addition to the luminance-only predictions of the previous model, the new model predicts the detectability of quantization error in color space directions in which chrominance error plays a major role. This more complete model allows DCT coefficient quantization matrices to be designed for display conditions other than those of the experimental measurements: other display luminances, other veiling luminances, other spatial frequencies (different pixel sizes, viewing distances, and aspect ratios), and other color directions.

1. INTRODUCTION

1.1 Discrete cosine transform-based image compression

The discrete cosine transform (DCT) has become a standard method of image compression. Typically the image is divided into 8\times8-pixel blocks, which are each transformed into 64 transform coefficients. The DCT transform coefficients $I_{m,n}$, of an $N\times N$ block of image pixels $i_{j,k}$, are given by

$$I_{m,n} = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} i_{j,k} c_{j,m} c_{k,n}, \quad m,n = 0,\ldots,N-1.$$  \hspace{1cm} (1a)

where

$$c_{j,m} = \alpha_m \cos\left(\frac{\pi m}{2N} (2j+1)\right), \quad \text{and} \quad \alpha_m = \begin{cases} \sqrt{1/N}, & m = 0 \\ \sqrt{2/N}, & m > 0 \end{cases}.$$  \hspace{1cm} (1b)

The block of image pixels is reconstructed by the inverse transform:

$$i_{j,k} = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} I_{m,n} c_{j,m} c_{k,n}, \quad j,k = 1,\ldots,N-1,$$  \hspace{1cm} (2)

which for this normalization is the same as the forward transform. Quantization of the DCT coefficients achieves image compression, but also causes distortion in the decompressed image. Specifically, quantization of coefficient $I_{m,n}$ induces an error image which is simply the associated basis function, with amplitude equal to the coefficient quantization error (neglecting the DCT normalization).

1.2 The Quantization Matrix

The JPEG compression standard\textsuperscript{1,2} requires that uniform quantizers be used for all the DCT coefficients. The quantizer step size used for each coefficient is determined by the user. A matrix is used to specify the quantization of the DCT coefficients, where the $m,n$th entry, $Q_{m,n}$, in the matrix gives the quantizer step size for coefficient $I_{m,n}$. Two example quantization matrices have been included in the JPEG standard. These
matrices are given in Tables K.1 and K.2 of reference(2) and in Table 5 of reference(4). One of these matrices is commonly used for graylevel images, and for the luminance component image of color images; the other matrix is used for chrominance images. These matrices were designed for a particular compression/viewing scenario, and it is not clear how they should be changed when used under different viewing conditions, or especially for compression in a different color space. In this paper we propose a quantization matrix design technique that can be applied under a wide variety of conditions: different display luminaances, veiling luminances, spatial frequencies, and color spaces.

2. DETECTION MODELS

2.1 Luminance-only Detection Model

Peterson, Peng, Morgan, and Pennebaker\textsuperscript{4} developed quantization matrices for compressing images in the RGB color space (a different matrix is used for each of the R, G, and B component images). The matrices were derived from measured detection thresholds for small patches of replicated DCT basis functions, produced on a monitor using an individual R, G, or B gun on a black background. With minor adjustments, the measured thresholds were converted to quantization matrices which performed well in informal tests.

Ahumada and Peterson\textsuperscript{5} proposed that the threshold measurements of Peterson \textit{et al.}\textsuperscript{4} could be predicted by a luminance-only detection model. The theoretical basis of their model is the assumption that the detectability of distortion in the decompressed RGB image can be predicted from the luminance contrast of the error image caused in a color component image by quantization of an individual DCT coefficient for a single block. That is, if the quantization error images associated with all the quantized DCT coefficients in all image blocks in all three color component images have amplitudes below their respective visibility thresholds, then no distortion will be visible in the decompressed image.

The Ahumada/Peterson luminance-only detection model approximates the log of the contrast sensitivity function (the dependence of the inverse threshold contrast on spatial frequency) by a parabola in log spatial frequency. The predicted log luminance threshold of the \(m, n\)th DCT basis function is

\[
\log T_{1, m, n} = \log \frac{s b_{1, l}}{r_{1, l} + (1-r_{1, l}) \cos^2 \theta_{m, n}} + k_{l, 1} (\log f_{m, n} - \log f_L)^2, \quad m, n = 0, \ldots, N-1. \tag{3}
\]

The minimum luminance threshold, \(s b_{1, l}\), occurs at spatial frequency \(f_{1, l}\), and \(k_{l, 1}\) determines the steepness of the parabola. The parameter 0.0 < \(s\) < 1.0 is to account for visual system summation of quantization errors over a spatial neighborhood. Such spatial summation causes a decrease in threshold. The spatial frequency, \(f_{m, n}\), associated with the \(m, n\)th basis function, is given by

\[
f_{m, n} = \frac{1}{2N} \sqrt{(m/W_x)^2 + (n/W_y)^2}, \tag{4}
\]

where \(W_x\) is the horizontal and \(W_y\) the vertical size of a pixel in degrees of visual angle. The model includes a factor \((r_{1, l} + (1-r_{1, l}) \cos^2 \theta_{m, n})\) which accounts for the imperfect summation of the two Fourier components present in basis functions having two cosine components \((m\) and \(n\neq 0)\), and also accounts for the reduced sensitivity due to the obliqueness of these Fourier components. The magnitude of the summation/obliqueness effect is determined by 0.0 < \(r_{1, l}\) < 1.0, and the angular parameter \(\theta_{m, n}\) is given by

\[
\theta_{m, n} = \arcsin \frac{2f_{m, n} f_{0, n}}{f_{m, n}^2}. \tag{5}
\]

Based on a fourth power summation rule for the two Fourier components\textsuperscript{5}, \(r_{1, l}\) is set to 0.6. The oblique effect can be included by decreasing the value of \(r_{1, l}\).

Ahumada and Peterson\textsuperscript{5} fit this model to the Peterson \textit{et al.}\textsuperscript{4} threshold data, and then used the grating detection data of Van Nes and Bouman\textsuperscript{5} to derive luminance dependencies for \(b_{1, l}, f_{1, l}\), and \(k_{l, 1}\), thus enabling the model to be used for a range of viewing conditions affecting luminance, contrast, and spatial frequency of the quantization errors. Since the single gun measurements of Peterson \textit{et al.}\textsuperscript{4} mainly varied the intensity of the
spatial modulation (chrominance remained relatively constant), the ability of the luminance-only model to predict visibility thresholds for modulations in combined luminance and chrominance directions was not adequately tested. Also, the replicated DCT basis functions used by Peterson et al. have Fourier transforms possibly more like those of grating stimuli than those of single basis functions. To address these issues, Peterson made new threshold measurements of single basis function, single monitor-gun test images superimposed on a white background (1931 CIE coordinates: \( X_0 = 37.27, Y_0 = 41.19, Z_0 = 29.65 \)). This configuration gives test stimuli having more significant chrominance modulation. Figure 1 shows the new measured thresholds for basis functions where \( m \) or \( n = 0 \).

A parabola representing a version of the luminance-only model is also shown in Figure 1. This model does a fair job of predicting the measured thresholds independent of color direction, except for the DC (\( m \) and \( n = 0 \)) thresholds, which are obviously different for the three color guns. We propose that the lower thresholds for the R and B gun DC basis functions are the result of chromatic detection mechanisms having greater sensitivity than the luminance mechanism. Thus, even for quantization in the RGB color space, a luminance-only model is not quite sufficient. Color mechanisms must be taken into account to determine appropriate quantization levels for the DC coefficients. More importantly, for images compressed using isoluminant color directions, a complete color space discrimination model for the DCT basis functions is clearly needed.

![Graph showing threshold ratio against frequency, cycles/deg](image)

**Figure 1:** Visibility threshold contrast ratio measurements from Peterson of single basis function, single monitor-gun test images superimposed on a white background, for basis functions where \( m \) or \( n = 0 \). Circles indicate R gun thresholds, diamonds indicate G gun thresholds, and squares indicate B gun thresholds. The points plotted at the far left of the graph are DC basis function (\( m \) and \( n = 0 \)) thresholds. The parabola-shaped curve represents a version of the luminance-only model of Equation (3).

### 2.2 The Luminance/Chrominance Detection Model

To account for the DC sensitivities in the data of Figure 1, we add two chromatic channels to the luminance-only model. A large number of different color spaces have been proposed as appropriate bases for chromatic discriminations. We have selected for our chrominance channels those favored by Boynton, a red-green opponent channel and a blue channel. The relation between these chromatic channels and the CIE 1931 XYZ color space is straightforward. The blue channel is just \( Z \), and the red-green opponent channel \( O \) is given by \( O = 0.47X - 0.37Y - 0.10Z \). This opponent channel is Boynton’s (Red-cone)–2(Green-cone) channel, with the Red and Green XYZ cone responses taken from MacLeod and Boynton. (We ignore the small correction developed by Vo & for going from the 1931 standard CIE values to the scientifically favored 1951 Judd CIE values used by MacLeod and Boynton.) Expressed in matrix form, the transformation from XYZ to
our YOZ opponent color space is

\[
\begin{bmatrix}
\text{YOZ} \\
\text{XYZ}
\end{bmatrix} = \begin{bmatrix}
0 & 0.47 & 0 \\
1 & -0.37 & 0 \\
0 & -0.10 & 1
\end{bmatrix}
\begin{bmatrix}
\text{XYZ} \\
\text{M}_{\text{YOZ}}
\end{bmatrix}
\]

(6)

We model the frequency response of the Y channel with the luminance-only model described above. To reflect this, we subsequently refer to threshold \( T_{\text{L},m,n} \) as \( T_{\text{Y},m,n} \). The parameters in the luminance channel model will subsequently be referred to with a similar change of subscript (\( L \rightarrow Y \)). To complete our luminance/chrominance model, we must also specify the shape of the frequency responses of the O and Z channels. Measurements of the spatial frequency responses of isoluminant chromatic modulations have typically found the chromatic sensitivity functions (the dependence of the inverse threshold contrasts on spatial frequency) to be low-pass in the frequency range of our basis functions and to be less sensitive at high spatial frequencies than the luminance channel\(^{11,12}\). We therefore model each of the O and Z log chromatic thresholds as a parabola, modified by setting it equal to its minimum value for all spatial frequencies to the left of the minimum. Since the data of Peterson\(^9\) are too sparse to estimate two separate chromatic channels in close proximity, we make the simplifying assumption, supported by the results of Mullen\(^11\), that both O and Z have the same shape spatial frequency response. The O and Z log chromatic thresholds for the \( m,n \)th DCT basis function can then be written:

\[
\log T_{O,m,n} = \begin{cases} 
\log \left( \frac{s b_O}{r_O + (1-r_O) \cos^2 \theta_{m,n}} \right), & \text{if } f_{m,n} \leq f_O \\
\log \left( \frac{s b_O}{r_O + (1-r_O) \cos^2 \theta_{m,n}} + k_O (\log f_{m,n} - \log f_O)^2 \right), & \text{if } f_{m,n} > f_O 
\end{cases}
\]

(7a)

and

\[
\log T_{Z,m,n} = \begin{cases} 
\log \left( \frac{s b_Z}{r_O + (1-r_O) \cos^2 \theta_{m,n}} \right), & \text{if } f_{m,n} \leq f_O \\
\log \left( \frac{s b_Z}{r_O + (1-r_O) \cos^2 \theta_{m,n}} + k_O (\log f_{m,n} - \log f_O)^2 \right), & \text{if } f_{m,n} > f_O 
\end{cases}
\]

(7b)

Note that Equations (7a) and (7b) are identical, except for the parameters \( b_O \) and \( b_Z \); \( T_{O,m,n} \) and \( T_{Z,m,n} \) share the parameters \( s, k_O, f_O, \) and \( r_O \). To obtain the overall model threshold \( T_{m,n} \) from the three channel thresholds, we use the "minimum of" combination rule:

\[
T_{m,n} = \min\{T_{Y,m,n}, T_{O,m,n}, T_{Z,m,n}\}
\]

(8)

In order to estimate the parameters in the model described above, we fit the model to the data of Peterson\(^9\) shown in Figure 1. Recall that the Peterson\(^9\) thresholds were measured for single basis functions. To reflect the absence of a spatial summation effect in this data, we fixed \( s = 1.0 \) during the fitting process. This fit resulted in the parameter values shown in Table 1 for \( k_Y, f_Y, k_Z, \) and \( f_{OZ} \). We chose \( r_{OZ} = 0.6 \), the same as \( r_Y \).

Boytont\(^10\) claims that at moderately high intensities, the Z channel's minimum threshold (\( s b_Z \) in our model) is approximately proportional to the background activity of the Z channel, and the minimum thresholds for the Y and O channels (\( s b_Y \) and \( s b_O \) in our model) are approximately proportional to the background Y. Based on the fit of our model to the Figure 1 data, we set the constants of proportionality to be: \( b_Y = 0.0219 Y_0 \), \( b_O = 0.0080 Y_0 \), and \( b_Z = 0.0647 Z_0 \), where \( Y_0 \) and \( Z_0 \) are the CIE values of average white. To determine a value for \( s \), we compared the thresholds measured in Peterson\(^9\) to those measured by Van Nes and Bouman\(^6\) for large test pattern sinusoidal gratings. The Peterson\(^9\) thresholds are consistently higher than the Van Nes and Bouman\(^6\) thresholds, a result attributable to spatial summation. Multiplication of the Peterson\(^9\) data by 0.25 brings them into approximate agreement with the Van Nes and Bouman\(^6\) data. We therefore chose \( s = 0.25 \). These results are summarized in Table 1.
Table 1. Parameter values estimated for the model of Equation (8).

<table>
<thead>
<tr>
<th>model channel</th>
<th>parameter values</th>
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<td></td>
<td>s</td>
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<tr>
<td>O</td>
<td>0.25</td>
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<tr>
<td>Z</td>
<td>0.25</td>
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</table>

As part of the model fitting, we also tried the Euclidean distance combination rule:

\[ T_{m,n}^{-2} = T_{Y,m,n}^{-2} + T_{O,m,n}^{-2} + T_{Z,m,n}^{-2}. \]  \( (9) \)

However, when the data of Figure 1 were fit using this rule, in order to prevent contributions from the chromatic channels at low spatial frequencies, \( f_{OL} \) was forced to be unrealistically low, and/or \( k_{OL} \) was forced to be unrealistically high. This led to our selection of the "minimum of" rule for \( T_{m,n} \).

### 3. QUANTIZATION MATRIX DESIGN

Quantization errors in an arbitrary color space are interpreted in the following way. Suppose we wish to compress a color image whose pixels are computed as a linear combination of XYZ values,

\[ [\text{DEF}] = [\text{XYZ}]_{XYZ}^{\text{DEF}}. \]  \( (10) \)

That is, the DCT is to be performed on an image in color space DEF, and \( [\text{XYZ}]_{XYZ}^{\text{DEF}} \) is the transformation from XYZ to DEF. The image in DEF space can be thought of as being transformed to XYZ space, and then converted by the visual system to YOZ space for discrimination. We need to determine limits on the sizes of errors in each of the D, E, and F color space dimensions, in order for the resulting errors in the Y, O, and Z channels to all be below the thresholds established by our model. These DEF thresholds determine the quantization matrices. For example, a unit error in the amplitude of a DCT coefficient in dimension D induces errors whose amplitudes in the Y, O, and Z channels are given by the first row of \( [\text{DEF}]_{YOZ}^{\text{XYZ}} \):

\[ \text{DEF} \cdot [\text{M}_{\text{XYZ}}]_{YOZ} = [\text{DEF}]_{\text{DEF}} \cdot [\text{XYZ}]_{XYZ}^{\text{DEF}} \cdot [\text{M}_{\text{XYZ}}]_{YOZ}. \]  \( (11) \)

where \( [\text{DEF}]_{YOZ}^{\text{XYZ}} \) is the inverse of \( [\text{XYZ}]_{XYZ}^{\text{DEF}} \).

We now describe in detail the procedure to calculate \( Q_{D,m,n} \), \( Q_{E,m,n} \), and \( Q_{F,m,n} \), the quantization matrix entries for DCT coefficient \( I_{m,n} \) in the D, E, and F component images. First, using Equations (3) and (7), the display parameters \( W_x \) and \( W_y \), and the model parameters given in Table 1, the model channel thresholds, \( T_{Y,m,n} \), \( T_{O,m,n} \), and \( T_{Z,m,n} \), for the \( m,n \)th DCT basis function are calculated. Now let \( \gamma_{T_{D,m,n}} \), \( \alpha_{T_{D,m,n}} \), and \( \beta_{T_{D,m,n}} \) indicate the thresholds imposed on the quantization error in the D component by the model's thresholds for the Y, O, and Z channels, respectively. Each of the Y, O, and Z model channel thresholds are converted to a D threshold as follows:

\[ \gamma_{T_{D,m,n}} = \frac{T_{Y,m,n}}{|M_{1,1}|}, \quad \alpha_{T_{D,m,n}} = \frac{T_{O,m,n}}{|M_{1,2}|}, \quad \text{and} \quad \beta_{T_{D,m,n}} = \frac{T_{Z,m,n}}{|M_{1,3}|}. \]  \( (12a) \)

Similarly for E and F:

\[ \gamma_{T_{E,m,n}} = \frac{T_{Y,m,n}}{|M_{2,1}|}, \quad \alpha_{T_{E,m,n}} = \frac{T_{O,m,n}}{|M_{2,2}|}, \quad \text{and} \quad \beta_{T_{E,m,n}} = \frac{T_{Z,m,n}}{|M_{2,3}|}. \]  \( (12b) \)

\[ \gamma_{T_{F,m,n}} = \frac{T_{Y,m,n}}{|M_{3,1}|}, \quad \alpha_{T_{F,m,n}} = \frac{T_{O,m,n}}{|M_{3,2}|}, \quad \text{and} \quad \beta_{T_{F,m,n}} = \frac{T_{Z,m,n}}{|M_{3,3}|}. \]  \( (12c) \)

Then the combination rule is used to determine the D, E, and F thresholds. We use the "minimum of" rule:
\[ T_{D,m,n} = \min \{ Y T_{D,m,n}, Z T_{D,m,n} \} \]  
\[ T_{E,m,n} = \min \{ Y T_{E,m,n}, Z T_{E,m,n} \} \]  
\[ T_{F,m,n} = \min \{ Y T_{F,m,n}, Z T_{F,m,n} \} \]  

Finally, the D, E, and F quantization matrix entries are obtained by dividing the thresholds above by the DCT normalization constants \( (a_m) \) in Equation (1b):

\[ Q_{D,m,n} = \frac{T_{D,m,n}}{a_m}, \quad Q_{E,m,n} = \frac{T_{E,m,n}}{a_m}, \quad Q_{F,m,n} = \frac{T_{F,m,n}}{a_m}. \]  

The factor 2 results from the maximum quantization error being half the quantizer step size.

3.1 Quantization in RGB space

For quantization in monitor-RGB space, we require the matrix to transform from RGB to XYZ space, \( \text{RGB} \rightarrow \text{XYZ} \). Assuming that \( R, G, \), and \( B \) take on values between 0 and 1, \( \text{RGB} \rightarrow \text{XYZ} \) is the monitor calibration matrix giving the XYZ values for unit changes in each of the RGB signals. For our monitor,

\[ \text{RGB} \rightarrow \text{XYZ} = \begin{bmatrix} 26.1 & 13.3 & 2.3 \\ 25.2 & 48.9 & 10.2 \\ 9.3 & 4.7 & 35.7 \end{bmatrix}. \]  

This matrix is post-multiplied by \( \text{XYZ} \rightarrow \text{Y}OZ \) to obtain \( \text{RGB} \rightarrow \text{Y}OZ \):

\[ \text{Y}OZ = [ \text{RGB} ] \text{RGB} \rightarrow \text{Y}OZ \rightarrow \text{RGB} = \begin{bmatrix} 13.3 & 7.1 & 2.3 \\ 48.9 & -7.3 & 10.2 \\ 4.7 & -0.9 & 35.7 \end{bmatrix}. \]  

The matrix \( \text{RGB} \rightarrow \text{Y}OZ \) gives the amplitude of the YOZ errors resulting from unit errors in RGB. These values indicate the sensitivity of the discrimination model YOZ channels to RGB errors. For example, a unit error in the R component leads to an error of 7.1 in the O channel of the model.

We can calculate the R, G, and B coordinate increments which induce a minimum threshold step in each of the Y, O, and Z channels. These are the the entries of \( \text{RGB} \rightarrow \text{Y}OZ \) divided into the appropriate minimum threshold: \( s_{bY}, s_{bO}, \) or \( s_{bZ} \), calculated using the expressions in Table 1 and the \( Y_0 \) and \( Z_0 \) values of our average white. For example, letting \( \text{RGB} \rightarrow \text{Y}OZ \)\( \rightarrow \text{RGB} \), signify the upper left corner entry in \( \text{RGB} \rightarrow \text{Y}OZ \), the increment in R which results in a minimum threshold change in Y is \( s_{bY} \) divided by \( \text{RGB} \rightarrow \text{Y}OZ \)\( \rightarrow \text{RGB} \). RGB minimum threshold increments calculated in this way are given in Table 2 for YOZ. Note that the minimum threshold for G is determined by the Y channel (0.0046 versus 0.0113 and 0.0469). That is, the Y channel imposes the strictest limit on G in order for a G change to not induce "too large" a change in YOZ-space. Similarly, the minimum threshold for R comes from the O channel (0.0116), and for B comes from the Z channel (0.0134). Following the procedure described above, using \( \text{RGB} \rightarrow \text{Y}OZ \), the model parameters in Table 1, and the \( Y_0 \) and \( Z_0 \) values for our monitor, we obtain the quantization matrices shown in Table 3 for our RGB color space.

Table 2. Minimum thresholds imposed on R, G, and B quantization errors by the Y, O, and Z model minimum thresholds.

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<th>( O )</th>
<th>( Z )</th>
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</tr>
<tr>
<td>B</td>
<td>0.0483</td>
<td>-0.0881</td>
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Table 3. RGB quantization matrices. The values in these matrices are obtained following the procedure described in Section 3. The \( Q_{0,0} \) value is located in the upper left corner of each quantization matrix. As specified in the JPEG standard, the values have been rounded to the nearest integer. JPEG also requires that values in the quantization matrix be \( \leq \) 255.

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Figure 2 plots all the measured R, G, and B gun, single basis function thresholds from Peterson\(^9\) (including those for the dual frequency \((m \text{ and } n \neq 0)\) basis functions), after correction by the summation/obliqueness factors of Equations (3) and (7). Figure 2 also shows the curves for the model threshold predictions \( T_{Y, m,n} \), \( T_{O, m,n} \), and \( T_{Z, m,n} \), using the parameters in Table 1, except with \( s = 1.0 \). This value for \( s \) was used to reflect the absence of a spatial summation effect in the single basis function data. In addition, the \( T_{O, m,n} \) and \( T_{Z, m,n} \) threshold prediction curves have been converted to luminance units, since all the threshold data plotted are in luminance units. This is accomplished by multiplying the \( T_{O, m,n} \) threshold predictions by \( 13.3 / 7.1 \), and the \( T_{Z, m,n} \) threshold predictions by \( 4.7 / 35.7 \). These factors are obtained from the RGBM<sub>Y</sub><sub>Z</sub> matrix. Figure 2 shows that for the B component, the DC and lowest spatial frequency thresholds are determined by the Z channel, and for the R component, the DC threshold is determined by the O channel. All the G thresholds are assumed to be determined by the Y channel. Note that the DC threshold for the Y channel (which we assume to be the DC threshold measured for G) is not predicted on a theoretical basis. The dot-dashed line in Figure 2 demonstrates that the measured DC threshold for G, and hence our DC threshold for Y, was found to be approximately equal to the minimum threshold of the Y channel.
3.2 Quantization in YC_rC_b space

The visual system is known to be much more sensitive to high spatial frequency information in luminance, compared to chrominance. In an attempt to put all luminance information in a single channel which can then be compressed with maximum fidelity, color images are often represented in the YC_rC_b color space for image compression. Pennebaker and Mitchell\(^2\) give the transformation from RGB to YC_rC_b as

\[
Y' = 0.3 \cdot R + 0.6 \cdot G + 0.1 \cdot B, \quad (17a)
\]

\[
C_r = \frac{(R - Y')}{1.6} + 0.5, \quad (17b)
\]

\[
C_b = \frac{(B - Y')}{2} + 0.5. \quad (17c)
\]

The additive constants keep the coefficients between 0 and 1, and can be ignored for our purposes. The matrix \(YC_rC_bM_{YOZ}\) is obtained from \(YC_rC_bM_{RGB}\) (the inverse of the matrix of coefficients above) as follows:

\[
YC_rC_bM_{YOZ} = YC_rC_bM_{RGB} \times RGBM_{XYZ} \times XYZM_{YOZ} = \begin{bmatrix} 1.0 & 1.0 & 1.0 \\ 1.6 & -0.8 & 0.0 \\ 0.0 & -0.3 & 2.0 \end{bmatrix} \quad (18)
\]

Figure 2: Visibility threshold contrast ratio measurements from Peterson\(^9\) of single basis function, single monitor-gun test images superimposed on a white background, for all basis functions (0 ≤ m, n < N).

Circles indicate R gun thresholds, diamond-shaped points indicate G gun thresholds, and squares indicate B gun thresholds. Large symbols indicate thresholds for basis functions for which m or n = 0; thresholds for basis functions for which m and n ≠ 0 are indicated with small symbols. The points plotted at the far left of the graph are DC basis function (m and n = 0) thresholds. All thresholds are plotted after correction by the summation/obliqueness factors of Equations (3 & 7), with \(r_Y = r_{OZ} = 0.6\). The solid, dashed, and dotted curves show the channel model threshold predictions for \(T_{Y,m,n}\) (Equation (3)), \(T_{O,m,n}\) (Equation (7a)), and \(T_{Z,m,n}\) (Equation (7b)), respectively, using the parameters and conversion factors as described in Section 3.1. The dot-dashed line shows that the measured G DC threshold, which is the DC threshold for the Y channel, is approximately equal to the minimum threshold of the Y channel.
The matrix \( Y_{\text{C}_b} \) gives the amplitude of the YOZ errors resulting from unit errors in \( Y_{\text{C}_y} \). These values indicate the sensitivity of the discrimination model YOZ channels to \( Y_{\text{C}_y} \) errors.

In a fashion analogous to Table 2 for RGB, Table 4 shows the increments in \( Y', C_r, \) and \( C_b \) which induce minimum threshold changes in the \( Y, O, \) and \( Z \) channels. These are calculated in the same way as for RGB, but using \( Y_{\text{C}_y} \) instead of \( Y_{\text{C}_y} \). From Table 4, we see that the minimum thresholds for the chromatic channels \( C_r \) and \( C_b \) are determined by the model color discrimination channels \( O \) and \( Z \), respectively. In addition, the ratio of these thresholds to the \( Y \) channel thresholds are relatively low, so that the "minimum of" combination rule results in the \( O \) and \( Z \) channels determining \( T_{Y,m,n} \), \( T_{C_r,m,n} \), and \( T_{C_b,m,n} \) for a wider range of spatial frequencies than \( O \) and \( Z \) did for \( T_{R,m,n} \), \( T_{G,m,n} \), and \( T_{B,m,n} \). Table 5 shows the quantization matrices for \( Y_{\text{C}_y} \) color space.

Table 4. Minimum thresholds imposed on \( Y', C_r, \) and \( C_b \) quantization errors by the \( Y, O, \) and \( Z \) model minimum thresholds.

<table>
<thead>
<tr>
<th></th>
<th>( Y' )</th>
<th>( O )</th>
<th>( Z )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( Y' )</td>
<td>0.0034</td>
<td>-0.0733</td>
<td>0.0099</td>
</tr>
<tr>
<td>( C_r )</td>
<td>-0.0127</td>
<td>0.0048</td>
<td>-0.1066</td>
</tr>
<tr>
<td>( C_b )</td>
<td>-0.0323</td>
<td>0.1473</td>
<td>0.0071</td>
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Table 5. \( Y_{\text{C}_y} \) quantization matrices. The values are obtained as described in Section 3 and formatted as in Table 3.
4. CONCLUSIONS

We have presented a model for predicting visibility thresholds for DCT coefficient quantization error, from which quantization matrices for use in DCT-based compression can be designed. We estimated values for the parameters of our model based on experimentally measured visibility thresholds. The frequency parameters we estimated, \( f_Y \) and \( f_{OZ} \), agree fairly well with results others have reported for similar parameters. The values we have estimated for \( k_Y \) and \( k_{OZ} \) are similar to those estimated by others, however we have found these parameters to vary for the different experimentally measured thresholds. The value we have proposed for the obliqueness/summation parameters, \( r_Y \) and \( r_{OZ} \), only reflects summation and does not reflect an effect due to obliqueness. More data may be needed to more determine values for \( k_Y \), \( k_{OZ} \), \( r_Y \), and \( r_{OZ} \) more reliably; though those we propose here are reasonable and result in quantization matrices which perform well in preliminary tests. The value for \( s \) we have proposed is based on a limited amount of data. Further experiments are needed to determine the spatial extent over which summation occurs among DCT quantization errors, in order to estimate \( s \) more accurately.

The quantization matrices computed by the techniques described above take no account of image content. A promising extension of this model may be to optimize the quantization matrices for individual images or a class of images. That is, use an image-dependent approach to quantization matrix design. Watson\(^5\) has shown how this may be done for grayscale images, by taking into account local light adaptation, local contrast masking, and error pooling. Watson’s technique can be extended to the case of color images by adopting rules governing masking and adaptation within the O and Z channels.

5. ACKNOWLEDGMENTS

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