ABSTRACT
In Content-Based Image Retrieval (CBIR) two query-methods exist: query-by-example and query-by-memory. The user either selects an example image or selects image features retrieved from memory (such as color, texture, spatial attributes, and shape) to define his query. Hitherto, research on CBIR interfaces was absent. Hence, a usability evaluation of existing (CBIR) interfaces was done. Additionally, a study concerning human color cognition was conducted. Based on the resulting findings, a prototype color selection interface was developed. In a theoretical experiment, using Fitts' law, was proven that the prototype interface is highly efficient for CBIR purposes. This all resulted in a first set of design guidelines for CBIR color selection interfaces.

Author keywords
Content-Based Image Retrieval, CBIR, query-by-memory, color memory, color categories, color selection, Fitts' law, user-interface.

ACM Classification Keywords

INTRODUCTION
The content of the World Wide Web consists to a great extent of images [8]. Given the exploding market on digital photo and video camera's, the fast growing amount of image content further increases the need for image retrieval systems. Unfortunately, most image retrieval systems are text-based and do not provide the means for searching on image content.

Content-Based Image Retrieval (CBIR) methods, however, are capable of searching in image collections, even when (adequate) textual annotations are missing. Instead of text they use either an example image as query (query-by-example) or a set of image features (query-by-memory). In the latter case, a complex interface has to be designed for facilitating the feature selection process, based on the users memory. This paper focuses on the design of a color-selection interface for query-by-memory in CBIR systems.

LITERATURE OVERVIEW
Hitherto, all research toward CBIR systems focused on the feature matching process underlying image retrieval. For query-by-example purposes, well performing feature extraction and pattern recognition technology is paramount. For query-by-memory retrieval, these techniques are equally important, but furthermore, a well-designed user interface is required. Nevertheless, no extensive review on user-interfaces of CBIR systems is present today. Existing CBIR reviews, such as that of Venters and Cooper [10] emphasize the various image retrieval techniques, but not their interfaces. Others such as Steiner [9] only briefly discuss the usability of 36 freely available web based color-selectors, in a non-CBIR setting.

Everly and Mason [6] did conduct a usability study specific on color-selection. They assessed speed, accuracy, and ease of use of four color-selection user-interfaces: Apple's Crayon (see Figure 1), HSV, RGB, and CMYK color-selectors. On all three criteria the Crayon color-selector, which uses a discrete presentation of 60 colors, outperformed the other three color-selectors.
This can be explained by the origin of the HSV, RGB, and CMYK color-selectors: graphic design. Color-selectors in graphic design, available years before the first CBIR system was born, have other demands than color-selectors for CBIR. Subtle level crossings for example, do not have to be made in color-selection for CBIR, but are very custom in graphics design.

**EVALUATION OF 10 COLOR SELECTION INTERFACES**

In order to complete and sustain the latter study for CBIR user interfaces [6], 10 MSc.-students in Human-Computer Interaction were asked to evaluate the interfaces of 10 CBIR systems. They were all naive users in the field of CBIR. Concerning color selection, their findings can be summarized as a triplet of issues:

1. Most color selection interfaces require that the user is familiar with the presented color space. Imagine using three sliderbars, representing the Red, Green, and Blue axes of the RGB-model (see Java Sun RGB-sliderbars), for defining a color such as pink,
2. The students experienced the color selection interfaces as non-intuitive and often too "complex" (e.g., The interfaces provide multiple ways to define a color, which is confusing for the users.), and
3. The amount of variations of colors presented is eye-appealing, but judged as being not necessary. Hence, in CBIR the need for differentiation between more-or-less similar colors was absent. The presentation of hundreds of colors was considered as overwhelming and with that as being inefficient. Nevertheless, current CBIR systems that allow the user to specify colors (see Table 1 and Figure 2), still exhibit such an interface.

The next two sections will underline that for basic color selection by memory, CBIR interface design can be optimized using (only) a limited set of colors.

**THE 11 COLOR CATEGORIES**

When defining a color by memory humans have to rely on color memory. It is well known that humans have a relatively poor color memory over the long term [4]. They tend to remember colors as members of categories. Most people distinguish 11-color categories: red, green, blue, yellow, brown, purple, pink, orange, black, white, and gray [1,4].

There is a range of explanations for the existence of color categories; one of the strongest is the Sapir-Whorf view [11]. According to this view, linguistic categorization can influence non-linguistic perception and cognition. So, if colors appear more frequently than other colors, they are recognized more rapid than these.

Van den Broek et al. [2] confirmed the assumption that people quantize colors in 11 color categories. Next, they conducted two experiments that proved the difference between color categorization by the cognitive processes color discrimination and color memory. A color look-up table was derived from the experimental results. It was successfully used to design improved color matching techniques, yielding retrieval results that better match the query a user has in mind [3].

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1. They are robust to variability between people (i.e., all people are different and so is their color perception.) [1,4],
2. They are robust to variability within people (i.e., and, for example, also changing perceptual abilities.) [1,4],
3. The small number of categories makes their use computationally very efficient [3], and
4. No color space is needed for displaying the color selection interface [6].

Below we will discuss another advantage of using this limited number of discrete color categories from a human motor point of view.

A THEORETICAL EXPERIMENT: FITTS' LAW APPLIED ON COLOR-SELECTION USER INTERFACES

Introduction

An interface has a certain complexity from the perspective of human perceptual and motor skills. In 1954 Fitts [7] defined the Index of task Difficulty (ID), which best fitted his empirical results on human perceptual and motor skills.

The ID is embedded in Fitts' law [7], which is expressed in:

\[ T = a + b \cdot ID, \]

where \( T \) is the selection time (i.e., the time needed to select the target), \( a \) and \( b \) are empirically determined constants that depend on the task to be fulfilled (i.e., the setting in which the target has to be selected), on motor, and on perceptual characteristics of the user. Fitts' ID provides an indication for the motor workload of the color selection interface: the lower the ID, the lower the motor workload. An important goal as Dix et al. [5] already stated: "Speed and accuracy of movement are important considerations in the design of interactive systems."

Procedure

The procedure discussed here can be applied on each selection system. First, for each cell (or color) \( \varphi \) of the color selector under investigation, the Euclidean distance (\( D_\varphi \)) was determined. This was defined as the distance from the center of the color-selection user-interface to the cell \( \varphi \):

\[ D_\varphi = \sqrt{x_\varphi^2 + y_\varphi^2}, \]

where \( x_\varphi \) and \( y_\varphi \) are respectively the horizontal distance and the vertical distance between the center of the color-selection user-interface and cell \( \varphi \).

Second, the minimum of width and height (\( W_\varphi \)) of the cells was determined. Next, the ID for each individual cell \( \varphi \), was determined:

\[ ID_\varphi = \log_2 \left( \frac{D_\varphi}{W_\varphi} \right) + 1. \]

Third, the \( ID_\varphi \)'s of all cells \( \varphi \) of the color-selector were summed and divided by the number of cells (or colors) (\#\( \varphi \)) present in the color-selection interface:

\[ ID = \frac{\sum_\varphi ID_\varphi}{\#\varphi}. \]

This resulted in the ID of the color-selection interface as a whole (see Table 1).

Results

The procedure discussed above was applied on nine color-selection user-interfaces. Four of them were part of an application (marked \( \alpha \); see also Figures 1, 2, and 3), one of them was a stand-alone color chooser (marked \( \beta \)), and four of them were color-selectors of CBIR systems (marked \( \epsilon \)). One of the latter was a prototype color-selector (see Figure 4). The results of the analysis were placed in Table 1. This table provides the name of the application, the average ID, the amount of colors (#colors) present, and the difference in ID (\( \Delta \)), relative to the prototype color-selector.

<table>
<thead>
<tr>
<th>Name application</th>
<th>ID</th>
<th>#colors</th>
<th>( \Delta )</th>
</tr>
</thead>
<tbody>
<tr>
<td>MS PowerPoint standard( ^a )</td>
<td>2.32</td>
<td>144</td>
<td>1.97</td>
</tr>
<tr>
<td>MS PowerPoint custom( ^a )</td>
<td>3.27</td>
<td>810</td>
<td>2.79</td>
</tr>
<tr>
<td>MS Paint and EditPlus( ^a )</td>
<td>2.37</td>
<td>48</td>
<td>2.03</td>
</tr>
<tr>
<td>Apple's Crayon picker( ^a )</td>
<td>2.08</td>
<td>60</td>
<td>1.77</td>
</tr>
<tr>
<td>Java Sun Swatches( ^b )</td>
<td>3.32</td>
<td>310</td>
<td>2.83</td>
</tr>
<tr>
<td>QBIC( ^c )</td>
<td>4.76</td>
<td>16384</td>
<td>4.05</td>
</tr>
<tr>
<td>SIMPLIcity( ^c )</td>
<td>4.76</td>
<td>32768</td>
<td>4.05</td>
</tr>
<tr>
<td>VisualSEEK( ^c )</td>
<td>1.88</td>
<td>56</td>
<td>1.60</td>
</tr>
<tr>
<td>Prototype color-selector( ^c )</td>
<td>1.17</td>
<td>12</td>
<td></td>
</tr>
</tbody>
</table>

Table 1. The color-selectors studied, with their Index of task Difficulty (ID), their amount of colors (#colors), and the difference with the prototype color-selector (\( \Delta \)).

![Figure 3. The MS PowerPoint standard color selector.](image-url)
This paper discusses a prominent component of CBIR user-interfaces: the query-by-memory color selector. A first set of design guidelines is provided for query-by-memory CBIR color-selection interfaces. Based on these, a prototype interface was developed that is more efficient and has a high usability compared to other systems.

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