A Binned-Profile Approach to the Color Blending Problem

in Transparent See-Through Displays

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Abstract

See-through displays allow users to view both digital content and physical objects at once. In such displays, light coming from background objects mixes with the light originated in the display, causing what is known as the color blending problem. Color blending is an important issue for the wider adoption of see-through displays and AR in general as it affect the legibility and color encodings of digital content, with a negative impact on the general usability of such displays. Color preservation aims at reducing the impact of color blending by finding an alternative color which, once blended with the background, would result in the original color. At the heart of color preservation is the capacity to predict how digital and background colors blend for a particular display.

In this paper we propose the binned-profile model for color prediction and preservation in see-through displays. The binned-profile model is based on the observations that each display renders colors differently and that background colors are changed by the display medium before blending. For a given display the model uses a colorimetric profile of how such display shows colors; with colors binned to a small set of “noticeably different” colors. We validate our model by measuring the accuracy of the predictions against other prediction models (direct model and chromatic adaptation transformations). Then, we introduce a color correction algorithm and measure the accuracy of the corrections. We investigated our approach with an extensive set of digital and background colors and different hardware configurations (projector-based and OLED displays). Finally, we elaborate on the usability and design implications of our approach for color preservation.

**Keywords**: Color Blending, Optical See-through Displays, Color Binning, Color Correction, Color Perception.

**Index Terms**: H.5 [Information Interfaces and Presentation]: H.5.1: Multimedia Information Systems — Artificial, Augmented, and Virtual Realities; H.5.2: User Interfaces — Ergonomics, Evaluation / Methodology, Screen Design, Style Guides

# Introduction

Optical see-through displays allow users to view both digital content and physical objects at once. They come in multiple form factors (e.g. head mounted displays or projection-based) and are used in augmented reality (AR) as a way to enhance the real world with digital information. Although other technologies can also be used for AR (e.g. video see-through displays), optical see-through displays have the advantage of letting users see the real world with their own eyes, without reducing its fidelity and preserving properties like lighting, texture, color, age and wear. Researchers investigate optical see-through displays for a wide range of applications including medical, maintenance, education and training (see [1][3][5] for a comprehensive list of applications); with a few consumer electronics have started to adopt them [6][Android glasses]. We can expect wider adoption of such technologies with the introduction of novel mobile AR platforms like Google Glass [], and the continuous development of transparent LCD (Samsung NL22B [[link](http://www.samsung.com/us/business/commercial-display-solutions/LH22NLBVLVC/ZA)], Eyevis [[link](http://www.eyevis.de/index.php?article_id=163&clang=1)], RichTech [[link](http://www.richtechsystem.com/html/transparent-video-showcase.html)]) and OLED displays (Futaba Corporation [[link](http://www.oled-info.com/futabas-oled-road-map-amoleds-2014-transparent-and-flexible-oleds-cars-2015)], Fujitsu [[link](http://www.fujitsu.com/be/Images/Workplace_of_the_Future.pdf)], Winstar [[link](http://www.winstar.com.tw/newspaper_ov.php?lang=en&ID=153)]).



Figure 1. Color blending including the screen distortions for background and digital colors.

An important aspect of optical see-through displays is that background light coming from real-world object mixes with the light emitted by the display, something that has been described as color blending [1]. Color blending is an important issue as it can affect the legibility and color-encoding of digital information, and compromise the general usability of such devices. Despite being a widely acknowledge problem for the adoption of optical see-through displays and general AR applications, little research exists on how to preserve digital colors exposed to color blending. To preserve a digital color a system should find an alternative digital color which, upon blending with such background, comes closest to the desired digital color. Existing solutions include blocking background light, and iterative correction and measuring of the digital content. PROBLEMS WITH THIS APPROACH.

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An effective approach to preserve digital color in see-through displays relies on its *prediction* accuracy, i.e. the capacity to estimate the blend resulting from a given background and digital colors and for a particular display. In this paper we argue that high prediction accuracy requires taking into account two distortions introduced by the display and shown in Figure 1: (1) the way a particular display renders colors, and (2) the effect of the display media on the background color. To address the first distortion we propose the binned profile prediction model: a model that divides the continuous universe of colors into discrete and finite bins and measures how the display actually renders each bin. To address the second distortion we measure the background color only after passing through the display. We compared our model with other approaches to the *direct model*, the *chromatic adaptation transformation* (*CAT) model*. The direct model ignores the effect of the display on the digital colors; the CAT model uses known transformation matrices to determine the way a display shows particular colors.

We used a colorimeter to measure the accuracy of the different prediction models on three transparent see-through displays. Results showed that prediction with the binned-profiled model and background display distortion outperforms other combinations. We compared the accuracy of color preservation using the binned-profile model against using the direct model, with and without the background display distortion. Results showed our model outperforms the direct model correcting a wider range of colors and with higher accuracy.



Figure 2. Examples of color blending

This paper contributes to the field of augmented reality in several ways: 1) we propose a novel approach to color prediction and preservation for optical-see through displays; 2) we validate our approach against other possible solutions; 3) we discuss the implications of color blending for situations where color preservation is not possible or contrast preservation is preferred; and finally 4) we discuss the challenges associated to incorporating our algorithm into everyday optical see-through display platforms.

# Background

Color blending is the phenomenon where background light coming from real-world object mixes with the light emitted by the display and changes it. Figure 2-top shows examples of color blending for a yellow text over three different background conditions: no background (black), red and blue. Figure 2-bottom shows the corresponding shift in color: the yellow text shifts toward orange when the background is red, while the green text shifts toward green when the background is blue. Field studies of AR applications with optical see-through displays reveal that the clarity and legibility of digital colors are affected by such changes in normal outdoors conditions; i.e. the colors in text and icons are altered (change in hue) or washed out (de-saturation) [Pingel and Clarke 2005]. Such changes affect the user interface and can render it useless: e.g. text might turn unreadable when washed out, or color encoded information such as red warning icons might lose their visual meaning.

Gabbard et al. studied such color changes in optical see-through displays [6] by building an experimental test-bed and examining foreground (27 colors on the edge of the RBG gamut) and background colors (6 common outdoor colors – foliage, brick, sidewalk, pavement, white and no background) of different lighting level and hues. His results showed how light background colors affect all other colors by pulling them towards white; while background colors of different hues pull all colors toward them. They defined the color blended and perceived by a user (CP) as a function of the light source (L1), the reflectance (RF) of background object (B), the light emitted by the display (L3), the interaction of both L1 and L3 in the display (ARD), and the human perception (HP). Equation 1 describes the interactions:

(1)

Our goal in this paper is to offer a solution to the color blending problem by means of colorimetric compensation: carefully selecting the color shown by the display so that the resulting blend comes close to the color originally intended. At the core of color compensation is the capacity to estimate how two colors blend; more specifically, how a color showed by the display blends with a particular background color. To do so we take equation 1 as our starting point and unravel the interaction of colors on the display (ARD) to account for two externally observable distortions. The first distortion is due by the fact that each display represents digital colors differently, and that it is such representation the one to consider when estimating color blending. Figure 3-left shows the color red (#FF0000) as displayed by different screens. Figure 1 illustrates this distortion as the difference in hues between the “digital color” (DC) and the “color shown” (CS): for a given digital color, different displays produce light of different hues. The second distortion is due to the display medium changing the background color before blending (BCD). Figure 3-right shows the foliage color as seen through different screens. Figure 1 illustrates this distortion as the difference in hues between the “bg color” and the “bg in display” color. In our formulation we simplify the light and reflectance of the background (the RF(L1,B) component of equation 1) into the single entity “background color” (BC). Moreover, we leave the influence of human perception of colors for our future work. Thus, we formulate color blending as follows:

(2)

Key to our understanding of color blending is the characteri-zation of the fdDC and fdBC functions. The fdDC function describes the way a particular display shows a given digital color. The fdBC function describes the way a background color is altered by the display medium.

To explore the nature of these functions we use three different optical see-through displays, a standard LCD display for background colors, and a colorimeter for color measurements (see section 4 for a detailed description of our experimental test-bed). To examine the colors in the background display, the digital colors on optical see-through displays and the resulting color blends we used the notations of the *Commision Internationale de l’Éclairage* (CIE) color model. We use the CIE 1931 XYZ color space for color measurement and addition required by equation 2. However the XYZ color space resembles the working of the human visual system which is more sensitive to colors in the blue or green hours. Therefore, we used the CIE 1976 Lab color space, a perceptually uniform color space, to calculate the perceptual difference between colors; e.g. the distance between a color and its shift when blended, or the distance between a blend prediction and the measured blend.



Figure 3. Left: The digital color #FF0000 and as displayed by different optical see-through displays. Right: The folieage color, and as it is seen through different optical see-through displays.

# Related Work

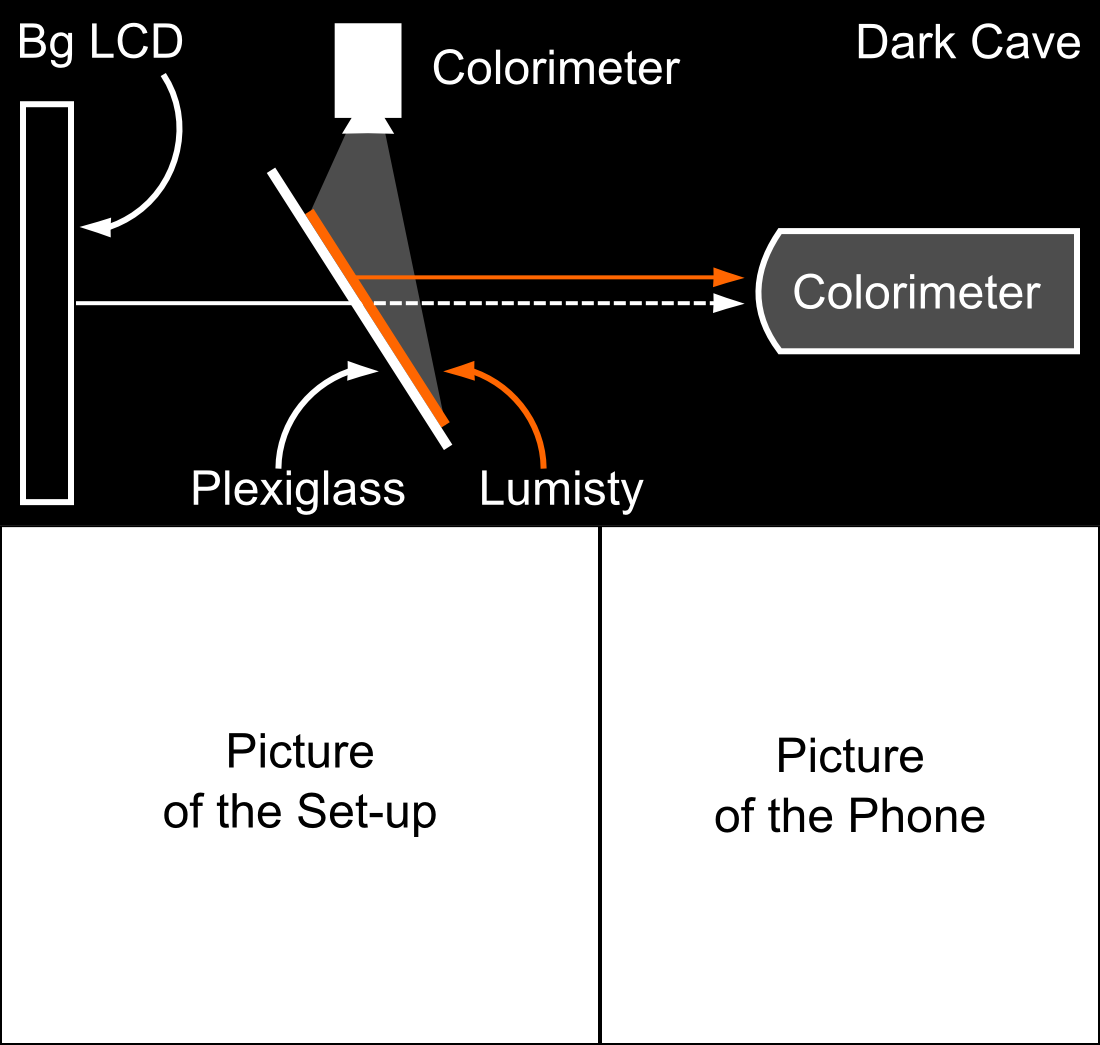


Figure 4. Experimental test-bed

Researchers have long discussed color blending as a significant perceptual challenge for the field of AR [13] especially in outdoor environments. Field studies of AR applications highlight that such inability so see the display clearly is worse with bright sunlight and with the sun lower in the sky [9]. In order to improve the display visibility users resort to strategies like looking for a dark spot (dark surface or shadow) or placing a hand in front of the display. Both strategies require users to switch context between their activity and the display and often missing important information. Strategies like these inspired researchers to investigate automatic ways to improve display clarity. A simple approach is to dynamically increase the intensity of the digital content (mentioned in [11]), however such solution is not always efficient [9]. Leykin and Tuceryan capture the field of view of the user and classify this image into zones where digital text would be readable or unreadable [14]. In a similar fashion, Tanaka et al. developed a layout system that relocates digital content to the darker areas of the display [17] taking into account restrictions like ordering of the components.

Color blending is also an important factor affecting the effective occlusion of physical objects by digital content; a feature particularly useful when the real environment is enhanced with 3D virtual objects that are intended to look real, such as in architectonical previewing. Without effective occlusion, the virtual object is perceived as translucent and unreal [4] and can confuse users [16]. Solving the occlusion problem keeps digital content from being affected by the physical objects in the background, thus solving the color blending problem. The main approach to solving occlusion has been to stop the light coming from the background by enhancing head-mounted displays with light blocking devices such as a transparent LCD [10][12][18] or spatial light modulators (SLM) [4]. In this approach a black/white depth mask of the scene is generated with the black pixels covering the area where digital content is not to mix with the background light. Therefore, digital colors projected on the black areas as seen in their original hue and lightness. Another solution is to control the illumination of the physical objects in a way that areas behind digital content remain in the dark. Noda et al. explored this approach by constraining digital objects into a dark room [15], while Bimber and Frölich implement it via occlusion shadows in a virtual showcase [2]. Finally, occlusion support has also been achieved by placing in spatial AR by placing the parts of the optical system behind the augmented object, such as Inami et al.’s usage of retro-reflective material as optical camouflage [8].

Our approach differs from the existing solutions as we aim not to change the location of user interface elements and not to add new hardware components to the see-through display; rather we seek to manipulate the color shown by the see-through display; an approach known as colorimetric compensation or color correction.

The field of projector-based spatial AR studied color correction as a way to enable projections on non-white or textured surfaces. Nayar et al. proposed a camera-based radiometric calibration model to compute the relation between the digital image and the projection on a textured surface [16]. Their approach requires a calibration phase where known images are projected on the projection surface and the resulting blended images are processed to obtain compensation matrixes. The calibration phase is repeated for each new projection surface or when lighting conditions change. Bimber et al.extended the range of projectable color by using a transparent film and multiple projectors taking into account the reflectance and absorption of the digital color by the projection surface [4]. Grossberg et al. extended the radiometric model to include ambient light [8]. While these works deals primarily in device dependent RGB space, higher correction accuracy is achieved by working on the device independent CIE XYZ color space [Ashdown, Menk]. Weiland et al. studied colorimetric compensation in see-through displays, and proposed a subtraction compensation model which is based on both color differences and the human eyes adaptive range. This model limits the amount of correction introduced by the compensation algorithm, as a way to guaranty that digital content is shown even when light backgrounds. Their results shows good compensation results although the approach is limited to rather static digital content and background settings.

In this paper we continue this line of work with see-through displays but aim at situations where the background is not static and illumination conditions continuously change. Our work uses background subtraction with special considerations for the actual nature of foreground and background colors. We proposed the binned-profile model, an approach which uses a measured display profile for foreground colors, and considers background colors as seen through the display. Further we perform our calculations using the device independent CIE XYZ and CIE LAB color spaces. Finally, we extend our study to both projector-based and transparent OLED displays.

# Experimental Test-Bed

We designed and built an experimental test-bed to generate background colors at different lighting conditions, show colors on multiple see-through displays, and measure the resulting color blending (Figure 4).

To generate different backgrounds we chose an XXXX LCD display calibrated at the standard white point of D65, a white that accurately reproduces the color spectrum as it exists outdoors. This approach to generating the background color is restricted by the color gamut of the LCD. Our test-bed design takes distance from previous systems [6] which prioritize the capacity to obtain background colors as seem in everyday outdoor settings; our design prioritizes the capacity to automatically produce a wide variety of colors. For our experiments we used background colors from the Macbeth color chart, as they mimic those colors of everyday natural objects like skin color, foliage and flowers. Figure 5 shows the difference between the theoretical background colors and the ones produced and captured in our test-bed.

Our test-bed works with three see-through displays: two projector-based and one transparent OLED. The projector-based displays use XXX mm thick transparent acrylic surface covered with a Lumisty MFY 2555 film ([http://www.lumistyfilm.com](http://www.lumistyfilm.com/)) and one of two projectors at 40 degrees. The first projector is an Epson 1705 at 2200 lumens, hereafter called the p2200 display. The second projector is an Epson VS35ow at 3700 lumens, hereafter called the p3700 display. For the transparent OLED display we used a Lenovo S800 phone [7] which has a 240x320 transparent OLED display at 167 ppi. The phone display is covered in acrylic and with a total XXX mm thickness. The test-bed has a holder for the displays at YYY cms in front of the background LCD.

To collect data we used a Konica Minolta CS-200 luminance and color meter at a 0.2 degrees angle (standard observer angle). The colorimeter sits at XXX cms from the see-through display holder. All displays and colorimeter are connected to the same controlling computer and are kept from any outside light by an enclosure (represented in Figure 4 as the dark cave).

# Color Prediction

In order to build an accurate color preservation system, it is necessary to have accurate estimations of how the resulting blend for a give pair of background and foreground colors. Providing such estimation requires unveiling the fdDC and fdBC distortion functions of equation 2.

In this paper we propose a model of the fdDC distortion function called the binned-profile model (BP). The BP model divides the RGB color space (over 16 million colors) into a smaller set of perceptually different bins. To create the bins we divided the CIE LAB color space into boxes of 5×5×5 – a method proposed by Heer and Stone [10] which guarantees all colors inside the box are within one noticeable difference; i.e. they are perceived as the same color by a human observer [17]. Figure XX shows the actual CIELAB color space and the binned result. Then, we measured how each bin is shown by each of our three display devices (see Figure YY) and gather the values into a look-up table. When predicting how a digital color blends with a particular background, the model determines the bin of the digital color and uses the display’s lookup table to know how such bin is actually shown (Color Shown in Figure 1). The system predicts how the two colors blend by adding this color to the background. Listing 1 describes this process in details.

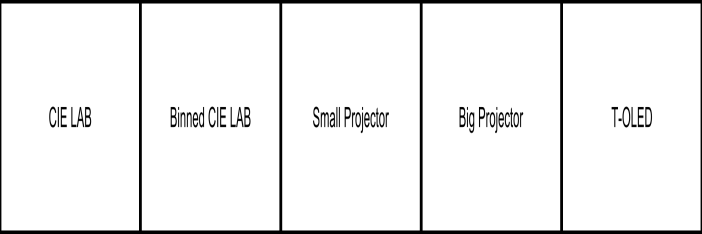


Figure 5. CIELAB color space and the binned reduction

BP\_prediction(display, foreground, background)

binned\_foreground = findBin(foreground)

display\_foreground = lookup(display , binned\_foreground)

prediction = addXYZ(display\_foreground, background)

return prediction

Listing 1. Binned-Profile prediction algorithm

We compare the prediction accuracy of our model against the direct model (DM) and three chromatic adaptation transformation models (CAT). Listing 2 presents the direct model, where the digital color is simply added to the background.

DM\_prediction(foreground, background)

prediction = addXYZ(foreground, background)

return prediction

Listing 2. Direct model prediction algorithm

Chromatic adaptation transformation (CAT) is an established method to estimate the actual colors a display can reproduce based on the brightest white it can emit. In other words, CAT could potentially account for the fdDC distortion function of see-through displays. CAT is based on matrices and researchers have proposed CAT models which rely on different matrices. We chose three popular CAT models for our investigations on color blending: CAT1, CAT2 and CAT3. We selected those models due their populatiry in the literature and as a representative set of their kind. Listing 3 presents how we used CAT models for our blending predictions; we transformed the foreground color using the CAT matrix before adding it to the background.

CAT\_prediction(CATmatrix, foreground, background)

cat\_foreground = foreground × CATmatrix

prediction = addXYZ(cat\_foreground, background)

return prediction

Listing 3. CAT model prediction algorithm

Show the new conception of color blending as presented in figure 4.



Figure 6. Background color set in the experimental set-up.

## Data Collection

We used C# WPF framework to publish the digital colors in the see-through display and background color in LCD monitor. Cs-200 colorimeter was used to capture the color shown on the displays with and without the background. In without background condition the LCD monitor behind the screen was powered down to achieve a complete back background condition (No background). Our colorimeter provide us with XYZ value of the colors it measured, we took these values and converted it into a normalized LAB space for color distance calculations for each display. For both projection based see-through hardware we measured the XYZ white points of the lumisty surface at 5 different points , One near the each of the display’s four corners and one in the center to see the change in content due to projection angle. At each of the 5 points the white points for the both the projector remained the same. Then we chose to point the colorimeter at the center of the display for to facilitate the content display.

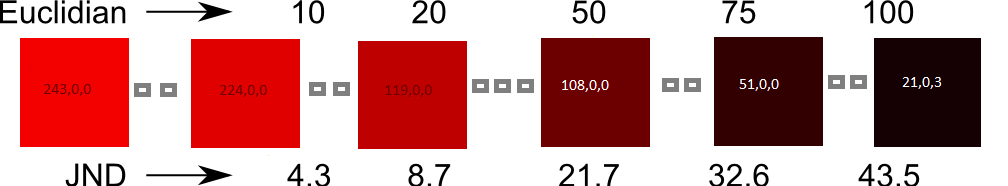


Figure 8. Examples of Euclidian distances and their corresponding just-noticeable difference.

We calculate seven different combinations of the white points;

1. One for each of the three displays without the background,

2. One for each of the three displays with white the background,

3. One for the background LCD’s white.



Figure 7. Single prediction result

We took the average out of 100 measures white point measurement for each condition. The background’s white point was calibrated to D65 and it was measured as (0.9504, 1, 1.0888).

For the Binned profile creation, each display was set in black background condition. 8376 perceptually different colors were displayed on the screen displayed. The displayed color where captured using the colorimeter and the captured xyz values where transferred into LAB space by using the reference white points given in the table 1.

|  |  |  |  |
| --- | --- | --- | --- |
| Display: | p2200 | p3700 | S800 |
| **Black Background**  X  Y  Z | 0.2655720  0.282182  0.481033 | 0.9504  1  1.0888 | 0.383264  0.395001  0.369982 |
| **White background**  X  Y  Z | 0.9504  0.990041  1.0888 | 0.9504  1  1.0888 | 0.724775  0.759896  0.727336 |

Table1: White points of all three displays in XYZ

For color prediction 827 colors which are equality spread across the perpetually different color space where chosen in random. These DC where shown under the influence of the 23 Macbeth chart background colors that can be reproduced in sRGB gamut. The color blends were measured using colorimeter place in front of the display.

To calculate the accuracy of the color values predicted by various models, we measure the distance between the colors shown on the display and the colors predicted by the models for each display. We also measure the change in color on blending to quantify the actual change in the color under the influence of the background. Normalized LAB space is used for distance calculations with the use of white points shown in table 1 respectively. In case of prediction‘s accuracy calculation the white points of white background condition was used and for color distance calculation between the blended color and the color to show the white points of no background conditions was used for each display respectively.

Describe how we computed the blending prediction for each model.

## Results

Show one sample 3D image for a given display with the direct model and the binned profile to demonstrate the differences in the blue dots.

Show the BIG bar charts image (5 models X 3 displays X 27 backgrounds)

# Color Preservation

Color correction is based on basic rules of color physics where the two color values can be added to from the third. As shown in many previous works in projection based spatial AR[16][4][8] the backgrounds effect on the digital content can be neutralized by counter balancing the digital color with a negative equivalent of the background. As shown in [Ashdown, Menk] with these operations did in device independent color space archives higher correction accuracy. Based on the results from our prediction models results the binned profile model which considers the effect of the screen on the background was found to be most accurate. We decided to do the color correct based on this highly precise model for correction.

The Binned profile based correction is built on the capacity of the system to predict how the color changes accurately. The correction algorithm makes use of the linear addition present in color. To preserve we try to counter balance the background’s color in the digital color. Such that when the corrected color is rendered on the screen, it turns into the correct foreground color on blending.

Driect\_preservation(foreground, background)

ColorToShow=SubtractXYZ(foreground, background)

//Check ColorToShow with respect to sRGB the gamut

If ColorToShow inside gamut -> true

return ColorToShow

else

ColorToShow=NearestLABInRGBSpace

return ColorToShow

Listing 3. Direct Model’s correction algorithm

Although the previous work have used direct model as the solution to correct color [Weiland] or have suggested as a possible approach to correct color[gabbard]. As the results from our study on model precisions accuracy shows how the screen changes the background and how the display shows the colors have a very high influence on the color blending. Based on our results we build bin profile model based algorithm to correct colors taking these distortions into account.

Correction algorithm is given in listing 4 works as follows, any given digital color is first mapper with its equivalent displayed color using bin profile, and then linear color subtraction is applied on XYZ values of the displayed color to find the counter balanced color. This counter balanced color is again matched with bin to find the color to be displayed on the display. As we are working with only 8000 odd colors , to make sure to choose the right corrected color , we make use of the bin predication which was proven to be accurate to choose a color. Such that the color chosen will produce best possible result on blending with the back ground.

Bin\_Profile\_preservation (foreground, background)

**Step1:** *Bin Look up*

binned\_foreground = findBin (foreground)

CounterBalancedC=SubtractXYZ (binned\_foreground, background)

**Step2:** *Check Counter Balanced Color with respect to Bin Profile*

If CounterBalancedC inside gamut -> true

Return CounterBalancedC

Else

CounterBalancedC=Nearest LAB in BinProfile

**Step3:** *Find best color to show from the bin look ups*

CorrectionAccuracy = maximum double value

CEquivalent = findBin (CounterBalancedC)

BinIndex=0;

Loop bin profile

{

BinIndex++;

**Step3.1:** *Use prediction for accuracy*

ColorToShow=BP\_prediction (display, CEquivalent, background)

PredictedDifference=ColorDifferenceLAB(CToShow, binned\_foreground)

CorrectionAccuracy> PredictedDifference

{

CorrectionAccuracy= PredictedDifference

CEquivalent = Next nearest Color in Bin (CounterBalancedC)

Continue loop;

}

ColorToShow= findBin(BinIndex)

Return ColorToShow;

}

**Step4:** Render the ***ColorToShow***

Listing 4. BPTD correction algorithm

# Discussion

Colors that can be corrected regardless of the background

Camera-based color correction

# Conclusions

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Figure 9. Prediction results