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#### **ICC-CMRF (2016-2017)**

#### An Efficient Uniform Color Space for High Dynamic Range and Wide Gamut Imagery

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## **Uniform Color Space**

- Predicts perceptual color difference
- > No inter-dependence b/w lightness, chroma, and hue





## **Criteria for a New UCS**



- **4** To uniformly encode high dynamic range signals e.g., 0-10,000 cd/m<sup>2</sup>
- **5** To uniformly encode wide gamut image signals e.g., Rec.2020





#### 1) CIELAB

#### **CIE/ISO** standard

#### **2) CAM16-UCS**

#### Accurately predicts small color difference data

#### 3) $IC_TC_P$

**Dolby's proposal for HDR and WCG** 

4)  $J_z a_z b_z$ 

**Current Proposal** 

• CIE, Colorimetry, 2004

• C. Li et al., Color Imaging Conf., 2016

• Dolby, 2016



### **Criteria and Visual Data**

No.	Criteria	Data sets	Purpose
1	Perceptual Color Difference	i) COMBVD ii) OSA	i) Training ii) Testing
2	Perceptual Uniformity	<ul> <li>i) COMBVD Ellipses</li> <li>ii) MacAdam Ellipses</li> <li>iii) Munsell Data</li> </ul>	i) Reference ii) Testing iii) Testing
3	Hue Linearity	<ul> <li>i) Hung &amp; Berns</li> <li>ii) Ebner &amp; Fairchild</li> <li>iii) Xiao <i>et al.</i></li> </ul>	<ul><li>i) Reference</li><li>ii) Testing</li><li>iii) Testing</li></ul>
4	Wide-range Lightness Prediction	i) RIT SL1 ii) RIT SL2	i) Testing ii) Training
5	Grey-scale convergence	Chroma-ratio metric (%)	Chroma-Ratio = $100 \frac{3C_w}{C_r + C_g + C_b}$



(3)

# Proposed J<sub>z</sub>a<sub>z</sub>b<sub>z</sub> space

$$\begin{bmatrix} X'_{D65} \\ Y'_{D65} \end{bmatrix} = \begin{bmatrix} bX_{D65} \\ gY_{D65} \end{bmatrix} - \begin{bmatrix} (b-1)Z_{D65} \\ (g-1)X_{D65} \end{bmatrix}$$
(1)

$$\begin{bmatrix} L \\ M \\ S \end{bmatrix} = \begin{bmatrix} 0.41478972 & 0.579999 & 0.0146480 \\ -0.2015100 & 1.120649 & 0.0531008 \\ -0.0166008 & 0.264800 & 0.6684799 \end{bmatrix} \begin{bmatrix} X_{D65} \\ Y_{D65} \\ Z_{D65} \end{bmatrix}$$
(2)

$$\left\{L, M', S', \right\} = \left(\frac{c_1 + c_2 \left(\frac{\{L, M, S\}}{10,000}\right)^n}{1 + c_3 \left(\frac{\{L, M, S\}}{10,000}\right)^n}\right)$$
 where  $n = 2610 / 2^{14}$   
 $p = 1.7 \times 2523 / 2^5$ 

$$\begin{bmatrix} I \\ a_z \\ b_z \end{bmatrix} = \begin{bmatrix} 0.5 & 0.5 & 0 \\ 3.524000 & -4.066708 & 0.542708 \\ 0.199076 & 1.096799 & -1.295875 \end{bmatrix} \begin{bmatrix} L \\ M' \\ S' \end{bmatrix}$$
(4)

$$J_{z} = \left(\frac{(1+d)I_{z}}{1+dI_{z}}\right) \qquad \text{where} \quad d = -0.56 \tag{5}$$



### **Results** (Uniformity)





#### **Results** (Uniformity in WCG)





#### **Results** (Quantitative)





#### **Results** (Hue Linearity)









### **Results** (Quantitative)





#### **Results** (Lightness Prediction)





#### **Results** (Lightness)





## Conclusions

- $J_z a_z b_z \rightarrow$
- Second best for small color difference
- Best for large color difference
- Best for wide gamut uniformity
- > Best for hue linearity
- Best for wide-range lightness prediction
- Plausible grey-scale convergence

J<sub>z</sub>a<sub>z</sub>b<sub>z</sub> gave overall best performance and should be confidently used



#### **Demonstration (Image Segmentation)**

Segmentation (7 regions) using K-means clustering algorithm

Original



CAM16-UCS





IC<sub>T</sub>C<sub>P</sub>













#### Deliverables

**1. Journal Paper:** *Optics Express* (*IF=3.3*)

https://www.osapublishing.org/oe/abstract.cfm?uri=oe-25-13-15131

2. MATLAB Code for J<sub>z</sub>a<sub>z</sub>b<sub>z</sub>







## **Model Development**

> Initially, we start using a structure similar to  $IC_TC_P$ 

(1)

(2)

 $(\mathbf{3})$ 

$$\begin{bmatrix} L \\ M \\ S \end{bmatrix} = \begin{bmatrix} \alpha_{1,1} & \alpha_{1,2} & 1 - \alpha_{1,1} - \alpha_{1,2} \\ \alpha_{2,1} & \alpha_{2,2} & 1 - \alpha_{2,1} - \alpha_{2,2} \\ \alpha_{3,1} & \alpha_{3,2} & 1 - \alpha_{3,1} - \alpha_{3,2} \end{bmatrix} \begin{bmatrix} X_{\text{D65}} \\ Y_{\text{D65}} \\ Z_{\text{D65}} \end{bmatrix}$$

$$\left\{L', M', S', \right\} = \left(\frac{c_1 + c_2\left(\frac{\left\{L, M, S\right\}}{10,000}\right)^n}{1 + c_3\left(\frac{\left\{L, M, S\right\}}{10,000}\right)^n}\right)^p$$

$$\begin{bmatrix} I \\ a_{z} \\ b_{z} \end{bmatrix} = \begin{bmatrix} \omega_{1,1} & \omega_{1,2} & 1 - \omega_{1,1} - \omega_{1,2} \\ \omega_{2,1} & \omega_{2,2} & 1 - \omega_{2,1} - \omega_{2,2} \\ \omega_{3,1} & \omega_{3,2} & 1 - \omega_{3,1} - \omega_{3,2} \end{bmatrix} \begin{bmatrix} L \\ M \\ S \end{bmatrix}$$





## **Model Development**

> Then we introduced a linear given below to correct hue linearity

$$\begin{bmatrix} X_{D65} \\ Y_{D65} \end{bmatrix} = \begin{bmatrix} bX_{D65} \\ gY_{D65} \end{bmatrix} - \begin{bmatrix} (b-1)Z_{D65} \\ (g-1)X_{D65} \end{bmatrix}$$
$$\begin{bmatrix} L \\ M \\ S \end{bmatrix} = \begin{bmatrix} \alpha_{1,1} & \alpha_{1,2} & 1-\alpha_{1,1}-\alpha_{1,2} \\ \alpha_{2,1} & \alpha_{2,2} & 1-\alpha_{2,1}-\alpha_{2,2} \\ \alpha_{3,1} & \alpha_{3,2} & 1-\alpha_{3,1}-\alpha_{3,2} \end{bmatrix} \begin{bmatrix} X_{D65} \\ Y_{D65} \\ Z_{D65} \end{bmatrix}$$
$$\{L, M, S, S, \} = \begin{bmatrix} c_1 + c_2 \left( \frac{\{L, M, S\}}{10,000} \right)^n \\ 1 + c_3 \left( \frac{\{L, M, S\}}{10,000} \right)^n \\ 1 + c_3 \left( \frac{\{L, M, S\}}{10,000} \right)^n \end{bmatrix}$$
$$\begin{bmatrix} I_z \\ a_z \\ b_z \end{bmatrix} = \begin{bmatrix} \omega_{1,1} & \omega_{1,2} & 1-\omega_{1,1}-\omega_{1,2} \\ \omega_{2,1} & \omega_{2,2} & 1-\omega_{2,1}-\omega_{2,2} \\ \omega_{3,1} & \omega_{3,2} & 1-\omega_{3,1}-\omega_{3,2} \end{bmatrix} \begin{bmatrix} L \\ M' \\ S' \end{bmatrix}$$

(1) 0.15 0.1 (2) yellowness-blueness 0.05 0 -0.05 -0.1 (3) -0.15 -0.2 -0.2 -0.1 0.1 0 redness-greenness (4)



### **Model Development**

Another simple equation to tune perceptual lightness



$$J_{z} = \left(\frac{\left(1+d\right)I_{z}}{1+dI_{z}}\right)$$
(5)

