

# **Geolocation of Human Trafficking Images: A Perceptual Color-based Image Retrieval Approach**

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### **Personal Background**

- Police Officer since 2008
- Cybercrime Investigator at the LKA Baden-Württemberg since 2020
- Since 2021 mostly investigations in the field of CSA
- Actually working in an AI project



### **Research Background**

Why? To evaluate the effectiveness of colour-based descriptors for Content-Based Image Retrieval (CBIR) in human-trafficking investigations.

**How?** Perceptual approach using the Hotels-50k dataset

#### Hotels-50K:

**Sources: Travel Websites** from hotels worldwide









**1.027.871** Images **50.000** Hotels

#### Source: TraffickCam App

(images submitted by travelers to help combat trafficking), which are more visually similar to images from trafficking investigations.



**17.954** Images **5.000** Hotels



### **Colour Extraction**

#### **Criteria for Initial Colour Extractions**

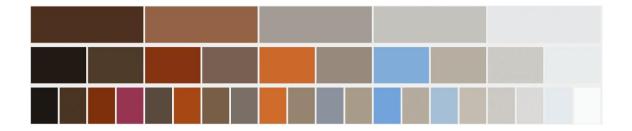
- Leveraged python colour palette extraction tool, Pylette<sup>1</sup>
- > Extracted **5, 10 and 20 dominant colours** per image
- Used the RGB colour model (Red, Green, Blue).
- Used Median-Cut to select the dominant colours
- > Sorted by **frequency.**



### **Colour Extraction**

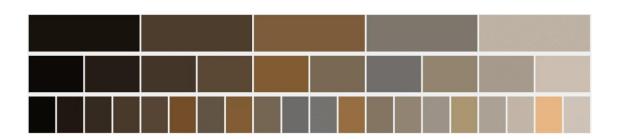
### 451, Hyatt Regency Waikiki Beach Resort & Spa





### 398, The Riverhouse Hotel & Convention Center



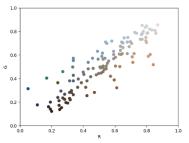


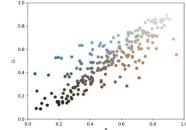


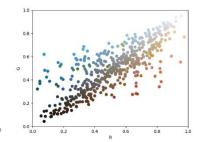
### Scatter Plots 2D & 3D

#### 451, Hyatt Regency Waikiki Beach Resort & Spa



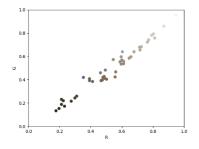


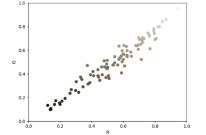


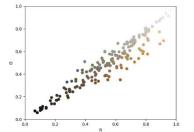


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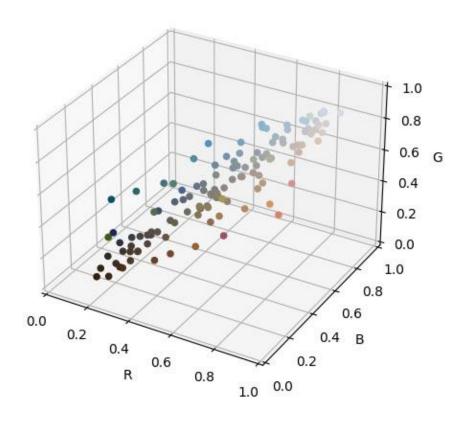






### **Curse of Dimensionality**





Increasing the number of extracted colours adds dimensional complexity

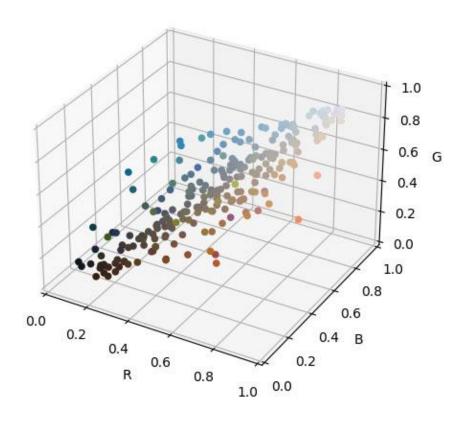
**Too few colours:** Risk of losing important visual information

**5 Values** 



### **Curse of Dimensionality**





Increasing the number of extracted colours adds dimensional complexity

**Too few colours:** Risk of losing important visual information

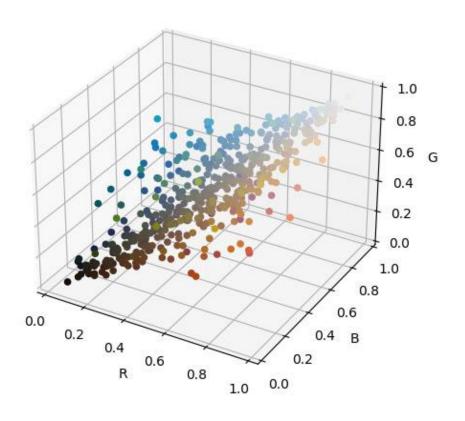
**Too many colours:** Risk of Over Fitting due to the curse of dimensionality

10 Values



### **Curse of Dimensionality**





Increasing the number of extracted colours adds dimensional complexity

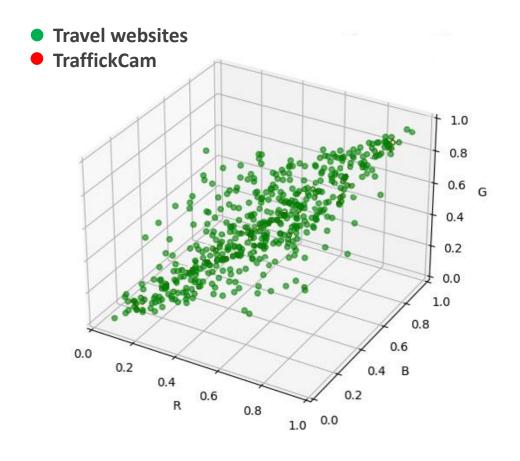
**Too few colours:** Risk of losing important visual information

**Too many colours:** Risk of Over Fitting due to the curse of dimensionality

Finding the right number of colour values is crucial to balance detail and computational complexity 20 Values



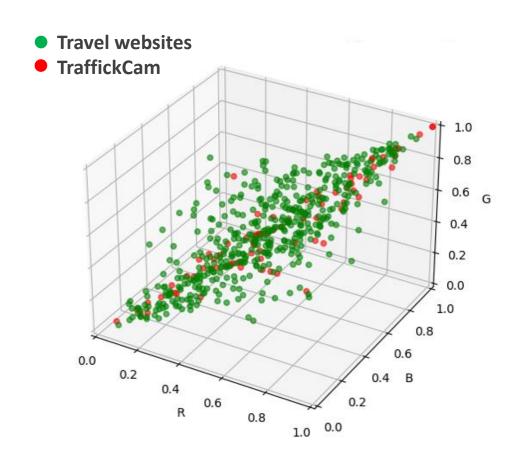
### **Comparing Trainsets and Teststes**







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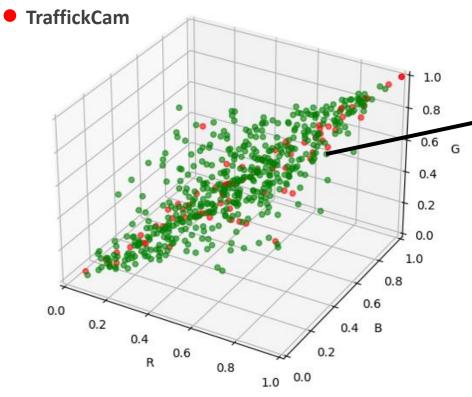


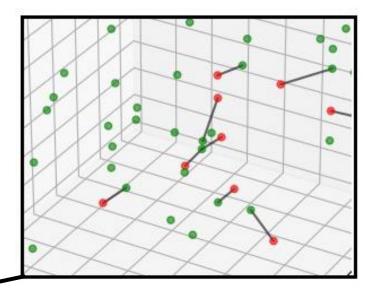




### **Euclidean Distance**







- Measures the straight-line distance between two points in RGB space.
- Calculated as the square root of the sum of squared differences between the corresponding color values (Red, Green, Blue).

$$d = \sqrt{(R_2 - R_1)^2 + (G_2 - G_1)^2 + (B_2 - B_1)^2}$$



## Dataset 1: "Unclean"

#### 3 x 100 different hotels

### **Travel Websites (Train)**













TraffickCam (Test)























### Dataset 1: "Unclean"

#### **Euclidean distances**

<b>Extracted Values</b>	Top-10	Top-20	Top-30	Top-40	Top-50	Top-60	Top-70	Top-80	Top-90	Top-100
1 Value	11,33	22,67	30,33	43,33	53,33	63,67	71,33	81,33	91,67	100,00
2 Values	10,33	22,67	32,67	44,33	52,33	62,00	74,00	84,00	91,67	100,00
5 Values	12,33	23,00	33,00	43,67	52,67	63,67	72,67	83,00	90,33	100,00
10 Values	12,33	22,00	33,00	42,67	51,67	62,67	71,33	82,33	90,00	100,00
20 Values	13,00	22,67	35,00	45,00	54,00	63,33	70,33	82,00	90,33	100,00

#### **Summary:**

- > Top-100 accuracy exceeded the 95% threshold, but Top-50 accuracy was only about 50%.
- The inclusion of out-of-scene images (e.g., tourist images with sea views) likely caused a high false positive rate.



## Dataset 1: "Unclean"

#### 3 x 100 different hotels

### **Travel Websites (Train)**













TraffickCam (Test)



















"Clean" datasets = "Unclean" datasets – TraffickCam and unwanted scenarios

#### **Travel Websites (Train)**











#### TraffickCam (Test)















"Clean" datasets = "Unclean" datasets – TraffickCam and unwanted scenarios

#### **Travel Websites (Train)**













#### **Euclidean distances**

<b>Extracted Values</b>	Top-10	Top-20	Top-30	Top-40	Top-50	Top-60	Top-70	Top-80	Top-90	Top-100
1 Value	42,00	62,00	72,67	81,00	89,33	92,67	95,67	97,33	99,33	100,00
2 Values	48,33	66,67	75,33	83,00	88,00	92,33	96,00	98,00	99,33	100,00
5 Values	44,33	59,67	68,00	78,00	85,00	90,33	94,67	96,67	99,00	100,00
10 Values	42,33	60,00	70,00	79,33	86,67	90,67	93,33	96,33	98,67	100,00
20 Values	43,33	57,33	69,33	80,67	87,00	92,00	96,00	97,33	99,00	100,00

#### **Summary:**

➤ Significant improvement; **Top-70 accuracy** reached **95** % for both **2** and **20 descriptors.** 



"Clean" datasets = "Unclean" datasets – TraffickCam and unwanted scenarios

#### **Travel Websites (Train)**













## Dataset 3: "Three Beds"

"Three Beds" datasets = Optimised selection out of the "Clean" datasets

#### **Travel Websites (Train)**













## Dataset 3: "Three Beds"

"Three Beds" datasets = Optimised selection out of the "Clean" datasets

#### **Travel Websites (Train)**











## Dataset 3: "Three Beds"

#### **Euclidean distances**

<b>Extracted Values</b>	Top-10	Top-20	Top-30	Top-40	Top-50	Top-60	Top-70	Top-80	Top-90	Top-100
1 Value	51,00	68,33	80,33	88,00	92,33	95,33	96,67	99,00	99,67	100,00
2 Values	62,00	74,33	84,33	92,00	95,67	96,67	97,00	99,33	99,67	100,00
5 Values	55,33	66,33	77,00	83,00	89,67	93,33	96,00	98,33	99,33	100,00
10 Values	55,67	72,00	82,67	88,67	94,00	96,00	97,00	98,67	99,33	100,00
20 Values	53,33	70,33	80,33	86,67	90,67	94,33	96,00	98,67	99,67	100,00

#### **Summary:**

▶ 2 descriptors achieved the best results with a Top-50 accuracy over 95 %.



### **Key Insights**

**Colour palettes** can **greatly improve** the accuracy of hotel room identification.

For Euclidean Distance the 2-descriptor model was the most reliable, offering high accuracy with minimal complexity.

Cleaning the datasets (removing out-of-scene images) had a significant impact on accuracy.

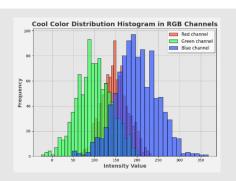


Too many descriptors introduced the risk of the curse of dimensionality.



### **Further Improvements**

Explore a histogram-based analysis to understand color distribution and introduce weighting within color spectra.



Further classification of training and test datasets based off perceptual context, such as bed, bathroom, & minibar.

Incorporating texture-based methods.

Exploring
alternative
color spaces,
such as
HSV/HLS.

White balance in the RGB space.



Test under realworld conditions by incorporating occlusion scenarios.



### **Advantages**

Independent of different angles or perspectives in room images.



Publicly available dataset.



Reproducible and traceable method, particularly in light of regulatory frameworks like the AI Act, where transparency and explainability of results are key requirements.

Methods are highly flexible and can be adapted to different datasets and applications by adjusting the number of descriptors and the distance metrics.

**Computationally simple**, making it easy to implement and scale across large volumes of images.



## Let's dive deeper!



Jessica Herrmann

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