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## 1 Presentation

### 1.1 Logistics:

Tuesday 8:15 - 9:30 Friday 9:45 - 11:00

### 1.2 Grading

- Homework 60%

- Final exam 20%
- Midterm exam 20%

### 1.3 Work

- Regular quizzes
- Homework/ project -> pre-requisite to take the final exam
  - won't accept any homework if it is late.

### 1.4 Health excuses

Not accepting health excuses.

## 2 Introduction to Computer Vision

The goal of computer vision to bridging the gap between pixels and meaning.

### 2.1 what is vision?

- humans: -> images -> sensing devices -> interpreting device -> interpretations
- computers: -> images -> cameras -> computers -> interpretation

### 2.2 what information to extract?

- Metric 3D information
- Semantics

### 2.3 what is color?

- the result of interaction between physical light in the environment and our visual system.
- A psychological property of our visual experiences when we look at objects and lights, not a physical property of those objects or lights.

## **2.4 two types of light sensitive receptors**

- cones: cone-shaped, less sensitive, operate in high light, color vision
- Rods: rod-shaped, highly sensitive, operate at night, gray-scale vision

## **2.5 trichromacy**

- three numbers seem to be sufficient for encoding color
- dates back to 18th century.
- don't have to be always RGB

## **2.6 RGB v. HSV**

- RGB: component
- HSV: hue, saturation, value

## **2.7 white balance:**

- it is the process of removing unrealistic color casts, so that objects which appear white in person are rendered white in you photo.
- when the white balance is not correct, the picture will have an unnatural color "cast".

### **2.7.1 Film cameras:**

different types of film or different filters for different illumination conditions

### **2.7.2 digital cameras**

- automatic white balance
- custom white balancing using a reference object

### **2.7.3 Von Kries adaptation**

- multiply each channel by a gain factor
- A more general approach would correspond to an arbitrary 3x3 matrix

#### 2.7.4 Best way: gray card

- take a picture of a neutral object (gray/white)
- deduct the weight of each channel, use the inverse of the weight to calibrate each channel.
- without gray card, we need to guess
  - using the average of the image, assume it is gray

#### 2.7.5 Brightest pixel assumption (non-saturated)

- highlights usually have the color of the light source
- Use weights inversely proportional to the values of the pixel

#### 2.7.6 Gamut mapping

### 3 Scale-invariant

## 4 Detectors

### 4.1 Harris detector

#### 4.1.1 mathematics

$$E(u, v) = \sum_{x, y} w(x, y) [I(x + u, y + v) - I(x, y)]$$

This equation is computational expensive, so we can use the first order Taylor

$$\begin{aligned} & \sum_{x, y} [I(x + u, y + v) - I(x, y)] \\ &= \sum_{x, y} [I(x, y) + uI_x + vI_y - I(x, y)] \\ &= u^2 I_x^2 + 2uv I_x I_y + v^2 I_y^2 \\ &= [uv] \left( \sum \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \right) [uv]' \end{aligned}$$

Thus, we can have a bilinear approximation

$$M = \left( \sum w(x, y) \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \right)$$

and

$$E(u, v) \approx [u, v]M[u, v]'$$

#### **4.1.2 cornerness**

We look at the two eigen value of the  $M$  matrix. With two big eigenvalues, it means it is a corner. If one is significantly bigger than the other, it is a edge. If both small, it is plain images. We write

$$R = \det(M) - k \cdot \text{tr}(M)^2$$

, where  $\det(M) = \lambda_1\lambda_2$ ,  $\text{tr}(M) = \lambda_1 + \lambda_2$

## **4.2**