

Homework 1

CS6550 - Computer Vision

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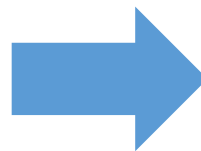
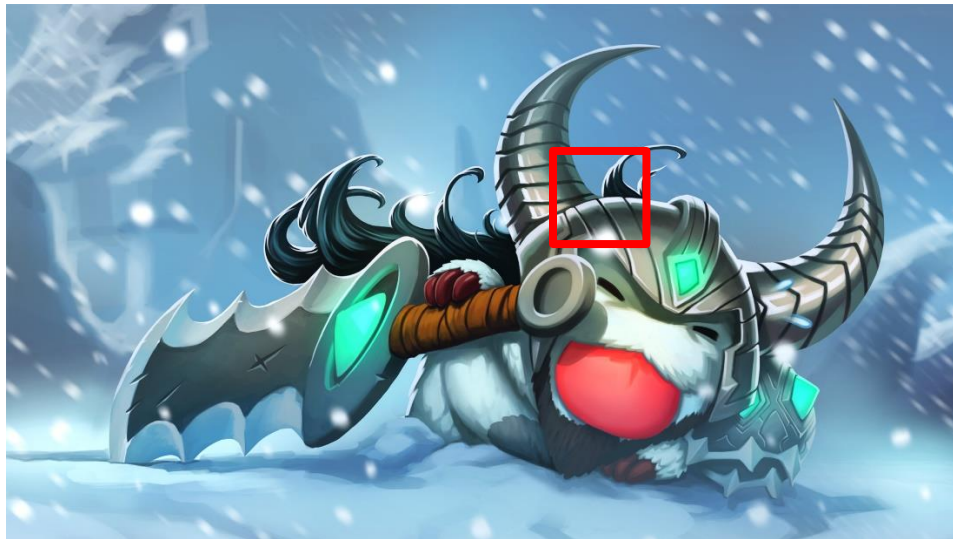
Part 1. Harris Corner Detection

Algorithm

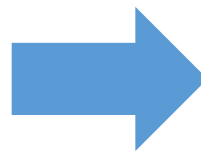
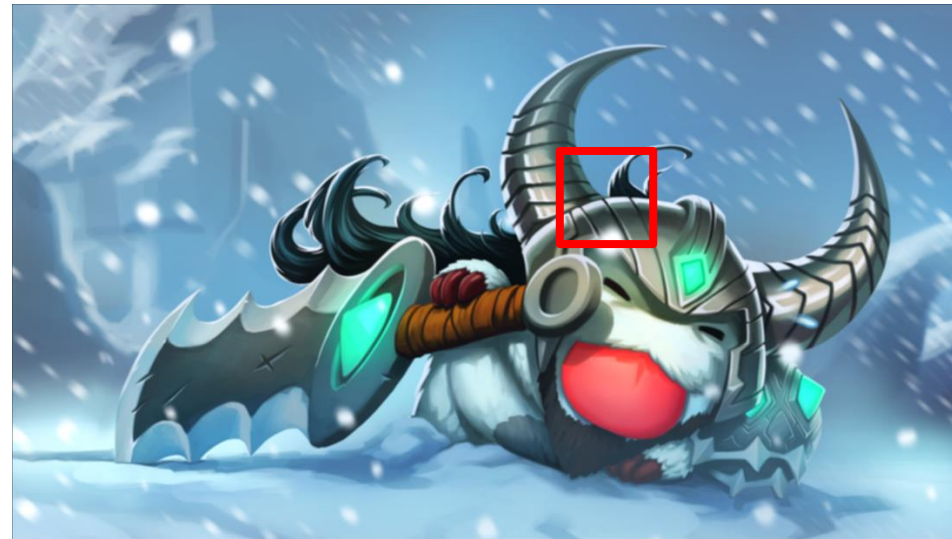
- Slides p.79

1. Filter the image with a Gaussian.
2. Estimate intensity gradient in two perpendicular directions for each pixel, $\frac{\partial f(x,y)}{\partial x}$, $\frac{\partial f(x,y)}{\partial y}$. This is performed by twice using a 1D convolution with the kernel approximating the derivative.
3. For each pixel and a given neighborhood window:
 - Calculate the local structure matrix A .
 - Evaluate the response function $R(A)$.
4. Choose the best candidates for corners by selecting a threshold on the response function $R(A)$ and perform non-maximal suppression.

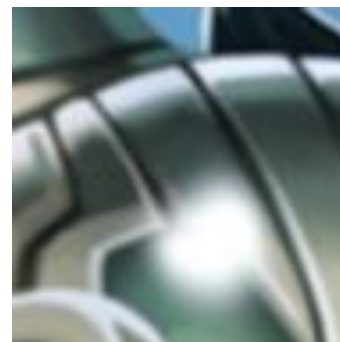
Gaussian Smooth



Gaussian



Gaussian



Compute Gradients

https://matplotlib.org/examples/color/colormaps_reference.html



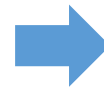
rgb2gray



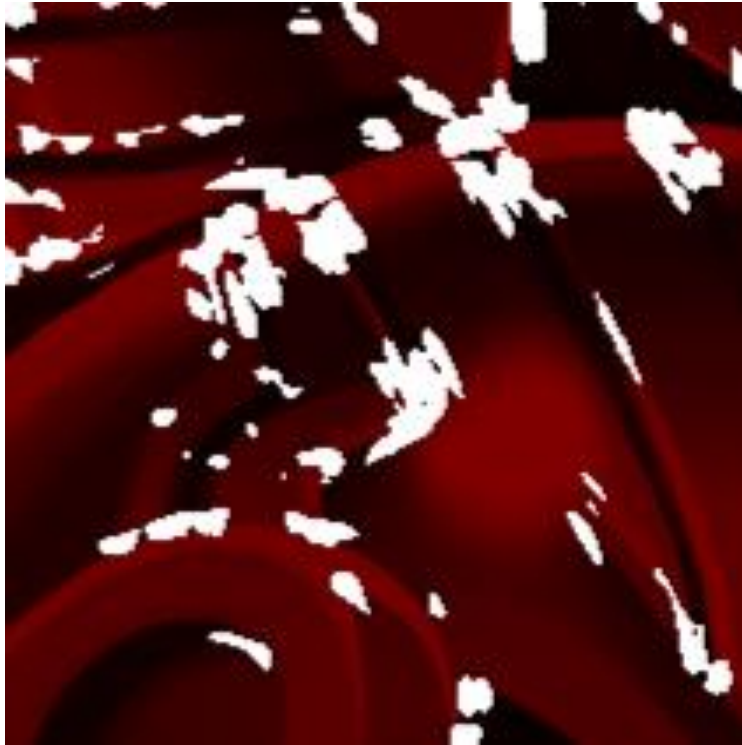
Direction



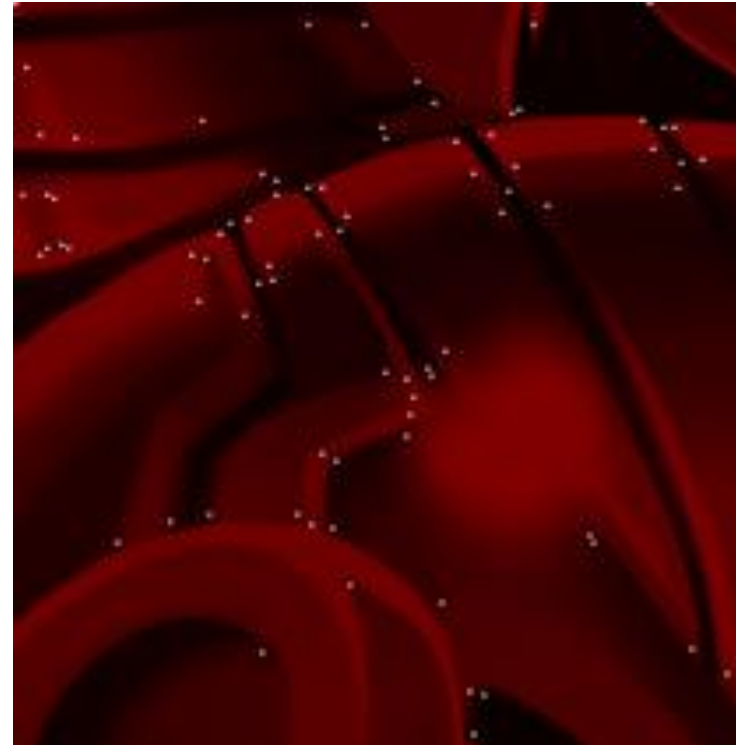
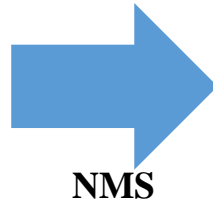
Magnitude



Non-maximal Suppression



Corner Response Function



Results



Origin



Rotate



Scale

Reference

- **Gaussian smooth**

<http://www.librow.com/articles/article-9>

- **Sobel edge detection**

<https://medium.com/datadriveninvestor/understanding-edge-detection-sobel-operator-2aada303b900>

- **Compute structure tensor**

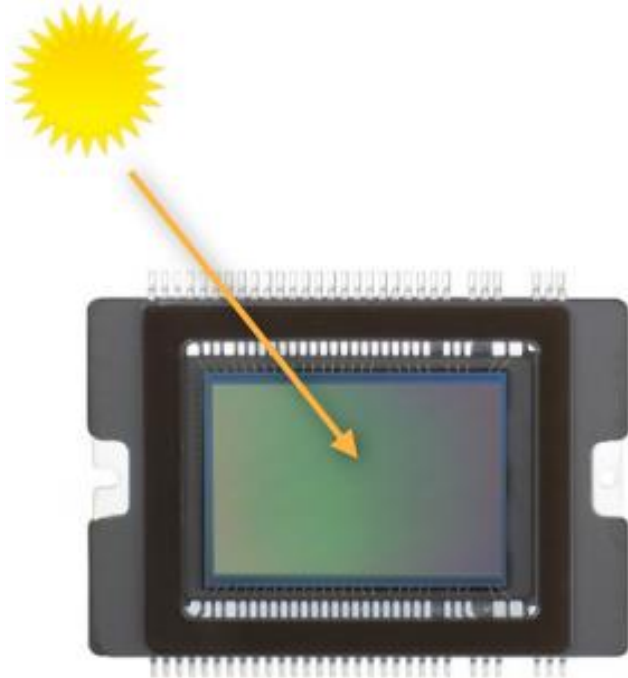
https://docs.opencv.org/3.4/dc/d0d/tutorial_py_features_harris.html

- **Non-maximal suppression**

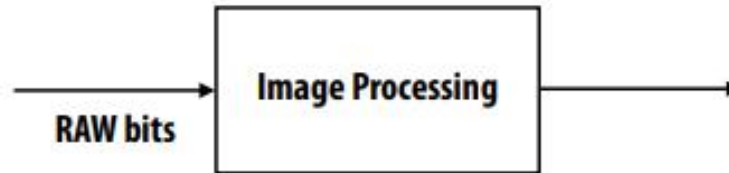
http://www.ipol.im/pub/art/2018/229/article_lr.pdf (p.313, Step 5.)

Part 2. Image Sensing Pipeline

Camera

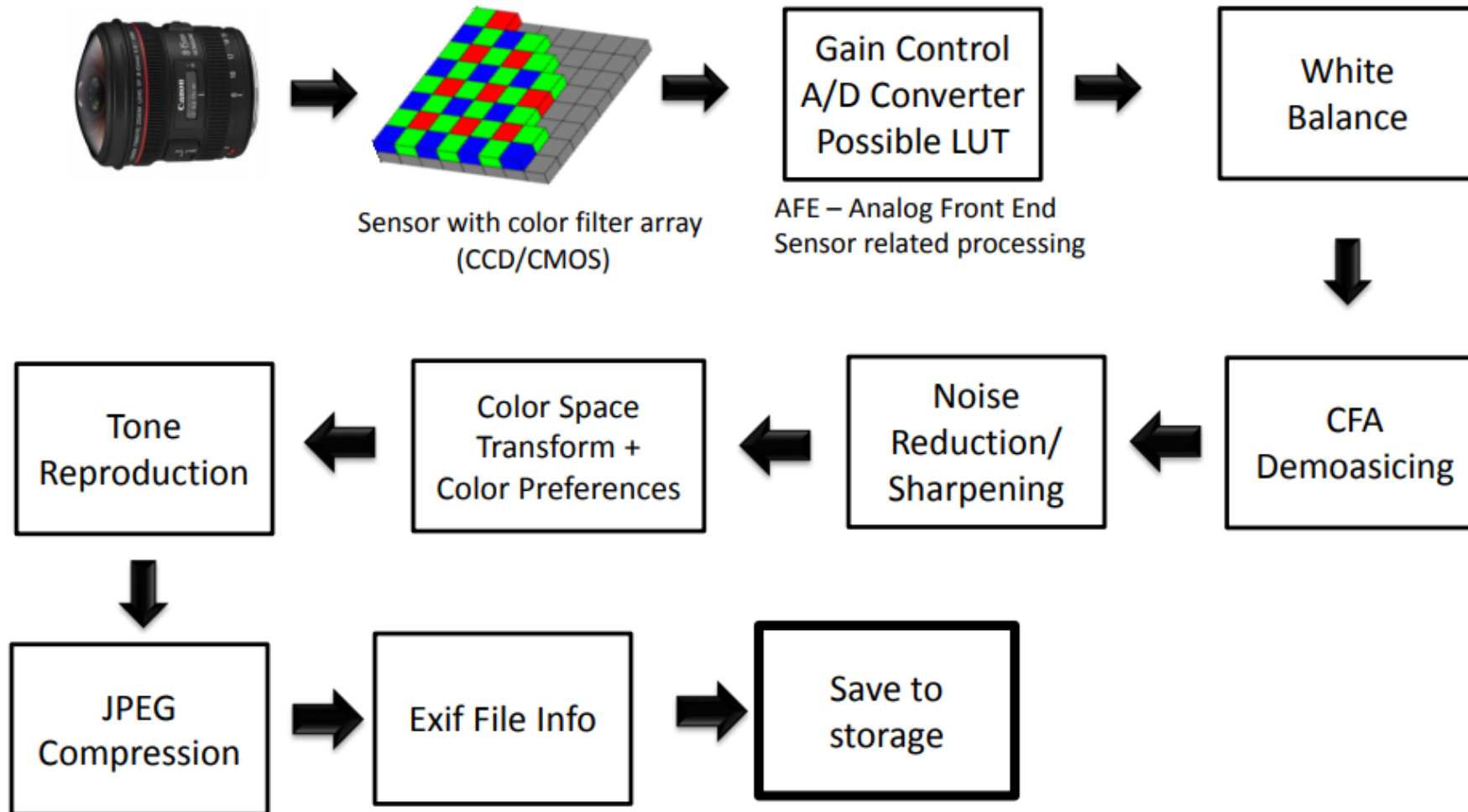


**Canon 14 MP CMOS Sensor
(14 bits per pixel)**

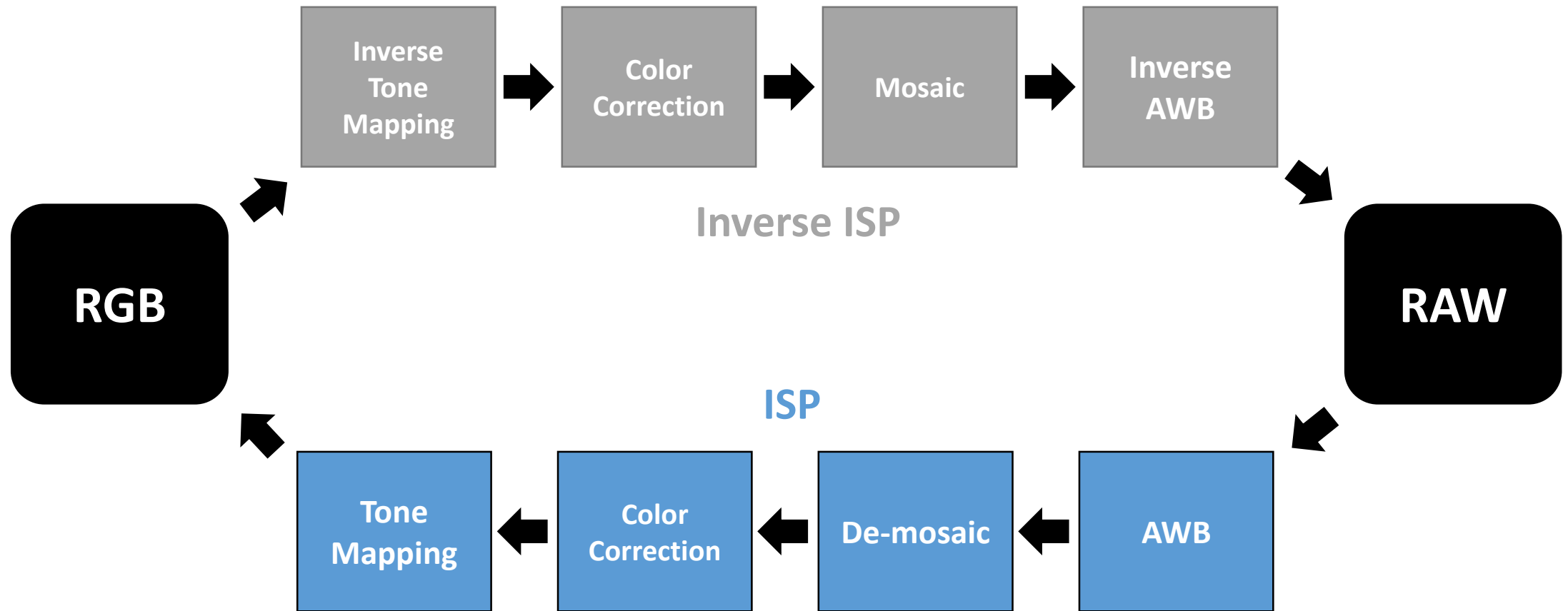


**Final Image Representation
(e.g., JPG file)**

Image Sensing Pipeline



In this homework



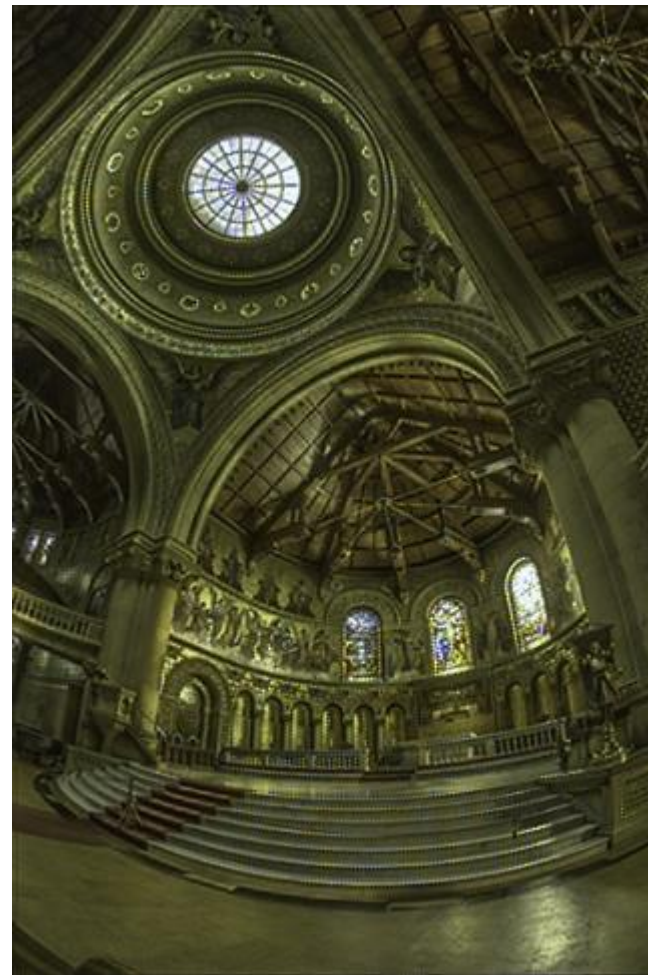
For Example (1/2)



RAW

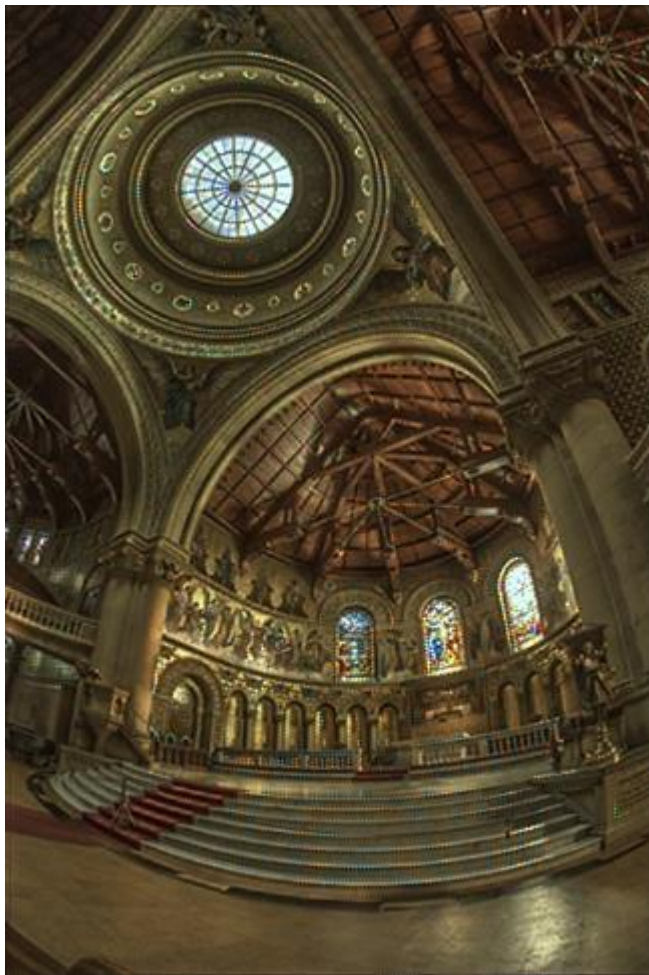


AWB

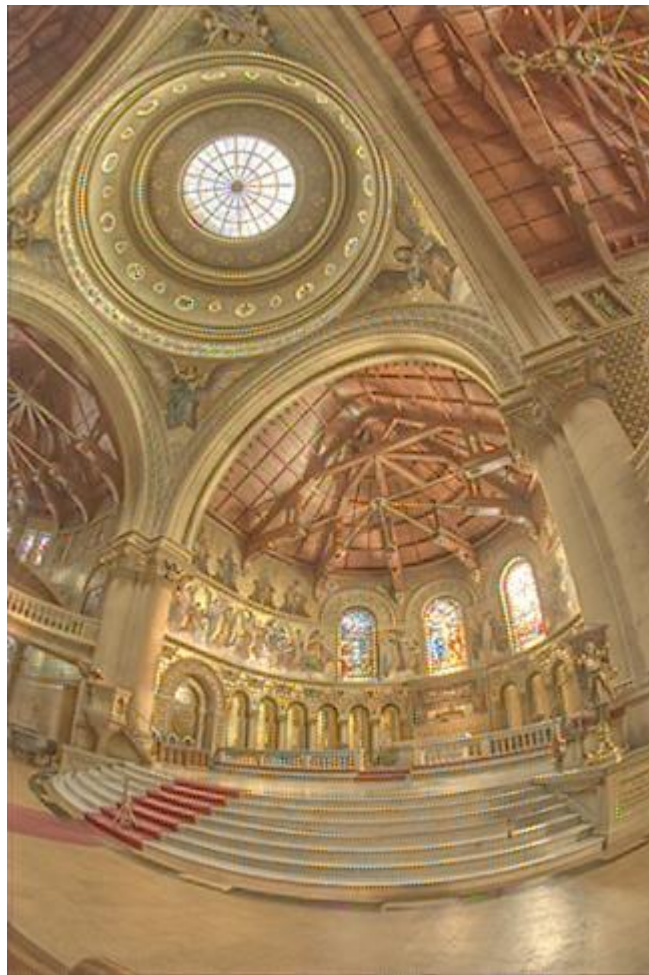


Demosaic

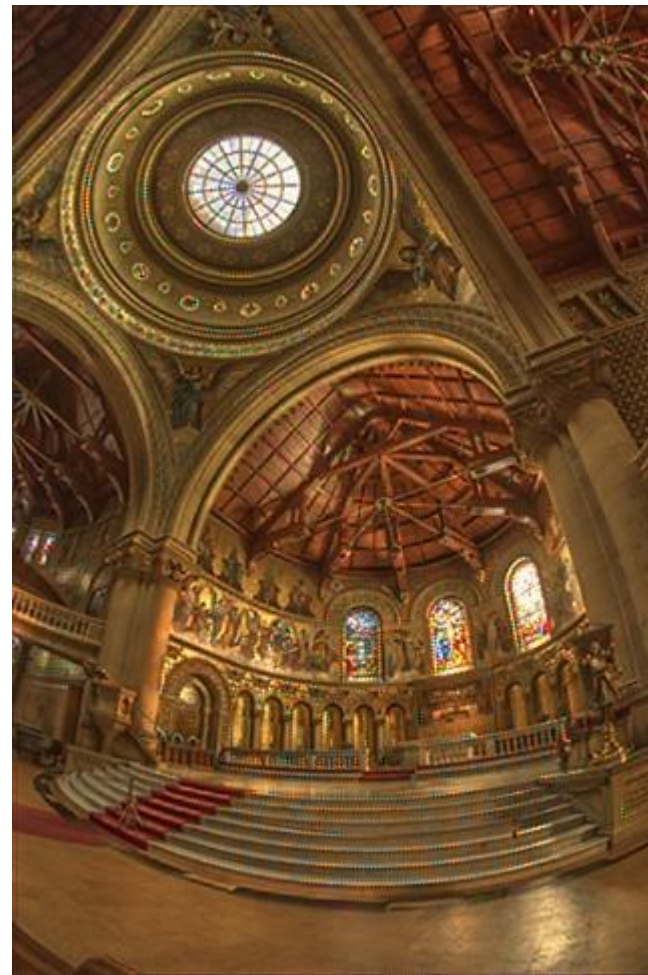
For Example (2/2)



Color Correction



CIE XYZ to RGB

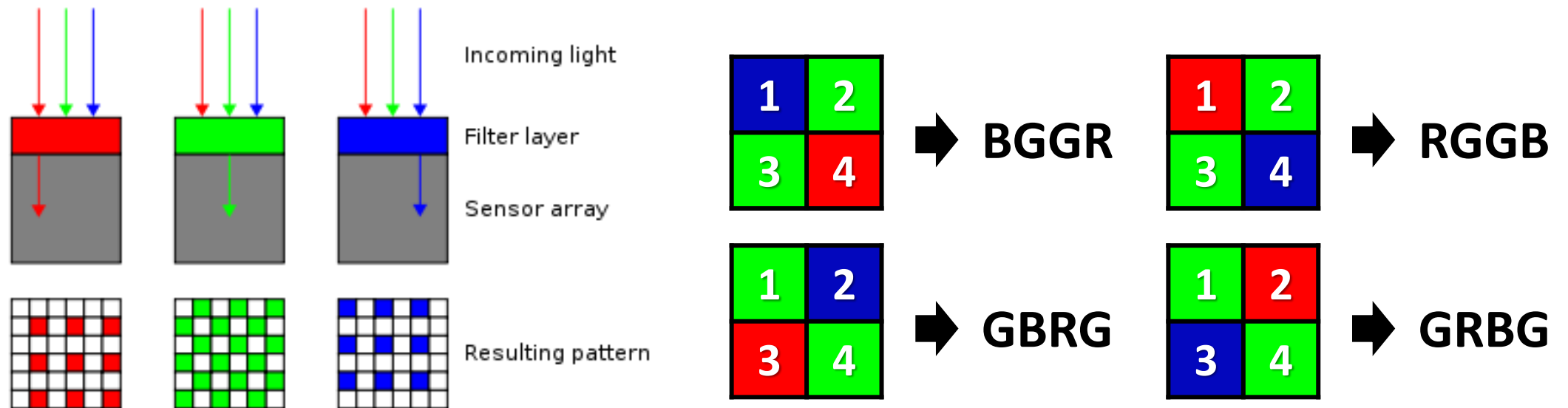


Tone Mapping

Raw Data (rawRGB)

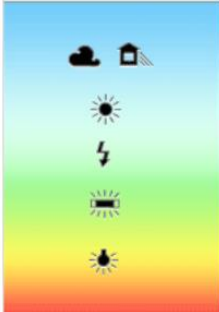
- **Bayer pattern (CFA, Color Filter Arrays)**

- Green photo-sensors: luminance-sensitive elements
- Red and blue photo-sensors: chrominance-sensitive elements
- Twice as many green elements as red or blue to mimic the physiology of the human eye
- Four patterns: **BGGR**, **RGGB**, **GBRG**, **GRBG**



AWB (Auto White Balance)

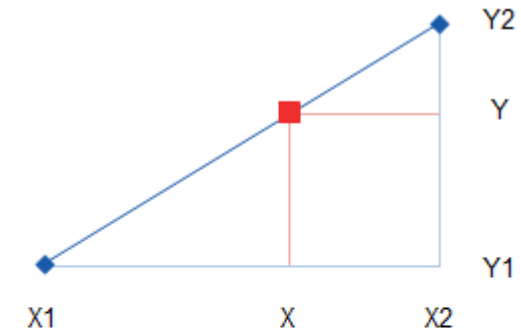
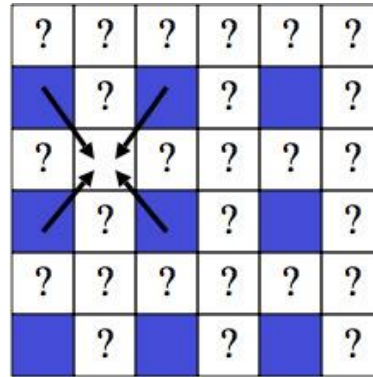
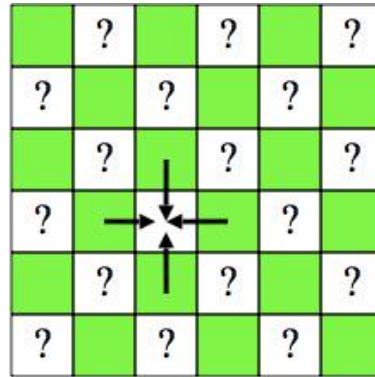
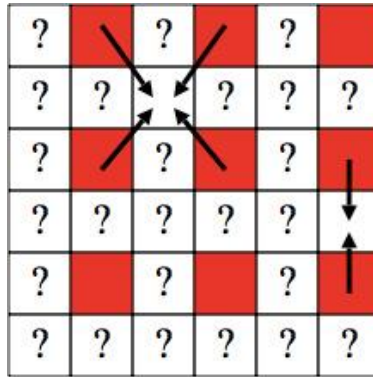
- **WB: mimics chromatic adaptation of the eye**
 - Can be manual settings
 - Stored in metadata
 - Otherwise, we can perform AWB

WB SETTINGS	COLOR TEMPERATURE	LIGHT SOURCES
	10000 - 15000 K	Clear Blue Sky
	6500 - 8000 K	Cloudy Sky / Shade
	6000 - 7000 K	Noon Sunlight
	5500 - 6500 K	Average Daylight
	5000 - 5500 K	Electronic Flash
	4000 - 5000 K	Fluorescent Light
	3000 - 4000 K	Early AM / Late PM
	2500 - 3000 K	Domestic Lightning
	1000 - 2000 K	Candle Flame

- **AWB: attempts to make what is assumed to be white map to “pure white”**
 - Two classical methods: gray world algorithm and white patch algorithm
 - In this homework, you only need to implement easiest AWB method
 - See more details in [white_balance.py](#)

De-mosaic (1/2)

- Easiest method – Linear Interpolation



$$\frac{(X - X1)}{(X2 - X1)} = \frac{(Y - Y1)}{(Y2 - Y1)}$$

$$Y = Y1 + (X - X1) \frac{(Y2 - Y1)}{(X2 - X1)}$$

De-mosaic (2/2)

- **Python Library - colour_demosaicing**

- *demosaicing_CFA_Bayer_bilinear*

- ***demosaicing_CFA_Bayer_Malvar2004***

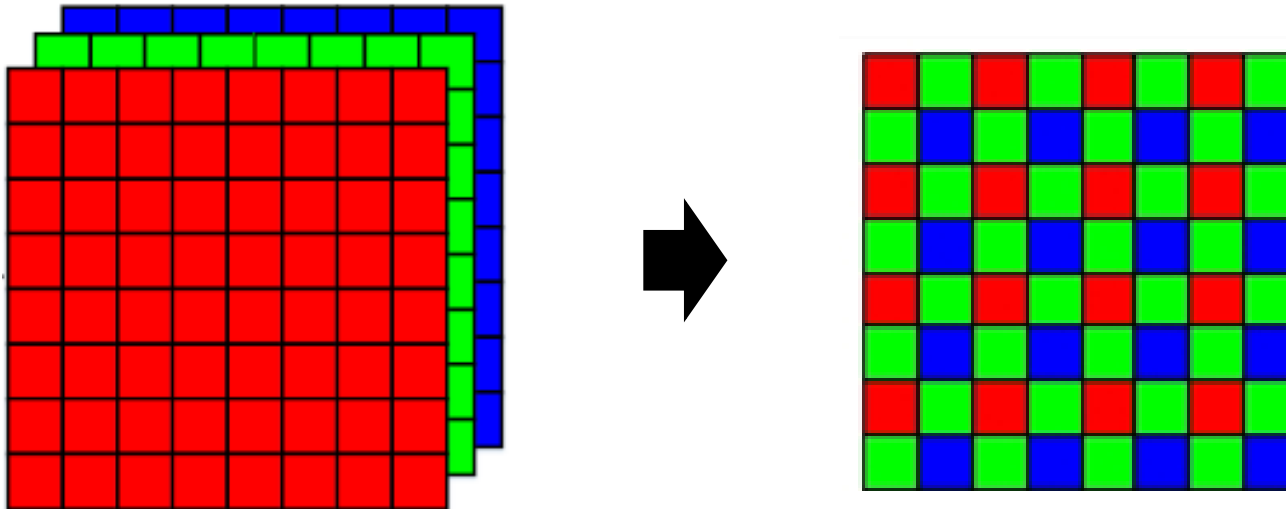
- *[MHCW04] Henrique S Malvar, Li-Wei He, Ross Cutler, and One Microsoft Way. High-Quality Linear Interpolation for Demosaicing of Bayer-Patterned Color Images. In International Conference of Acoustic, Speech and Signal Processing, 5–8. Institute of Electrical and Electronics Engineers, Inc., May 2004.*

- ***demosaicing_CFA_Bayer_Menon2007***

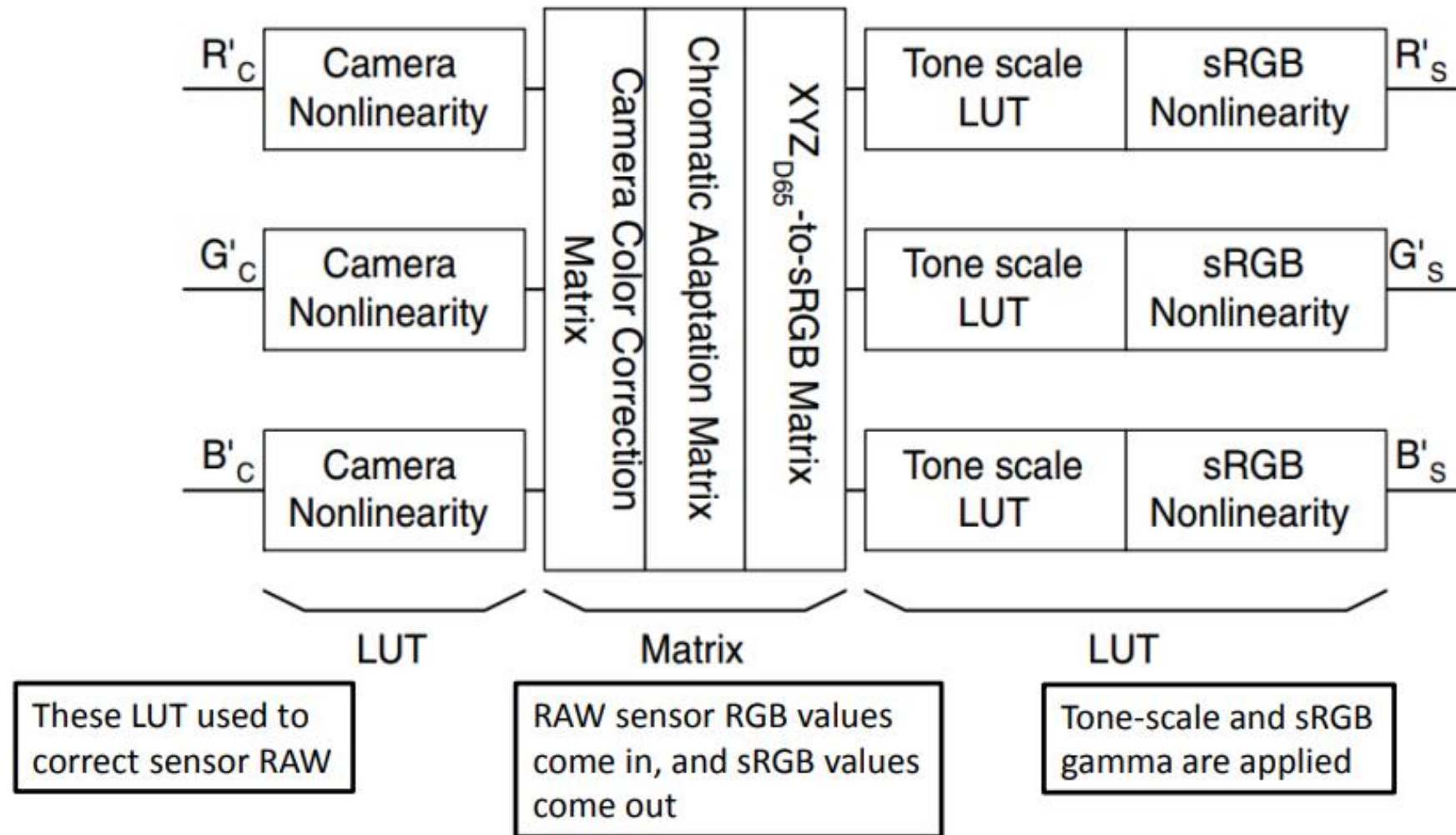
- *[MAC07] Daniele Menon, Stefano Andriani, and Giancarlo Calvagno. Demosaicing With Directional Filtering and a posteriori Decision. IEEE Transactions on Image Processing, 16(1):132–141, January 2007.*

Mosaic

- Discard the value of other 2 channels
 - $H*W*3 \rightarrow H*W$



Color Correction (1/3)



Color Correction (2/3)

- **CCM (Color Correction Matrix)**

- Transforms sensor native RGB values into CIE XYZ color space
- It is important that the white-balance has been performed correctly

Color images are stored in $m \times n \times 3$ arrays (m rows (height) \times n columns (width) \times 3 colors). For the sake of simplicity, we transform the color image to a $k \times 3$ array, where $k = m \times n$. The Original (uncorrected input) pixel data O can be represented as

$$O = \begin{bmatrix} O_{R1} & O_{G1} & O_{B1} \\ O_{R2} & O_{G2} & O_{B2} \\ \dots & \dots & \dots \\ O_{Rk} & O_{Gk} & O_{Bk} \end{bmatrix}$$

where the entries of row i , $[O_{Ri} \ O_{Gi} \ O_{Bi}]$, represent the normalized R, G, and B levels of pixel i . The transformed (corrected) array is called P , which is achieved by matrix multiplication with A , the **color correction matrix**. Imatest allows you to choose two different forms of A , either 3×3 or 4×3 .

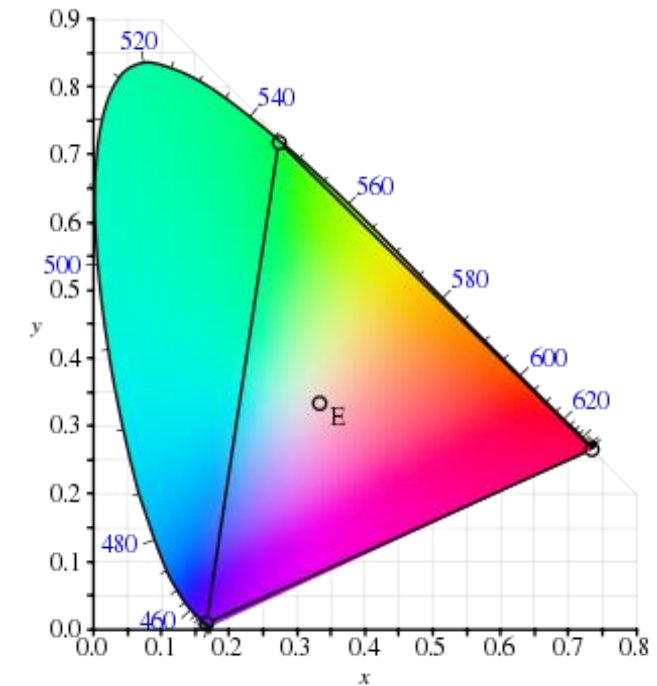
Each of the R, G, and B values of each output (corrected) pixel is a linear combination of the three input color channels of that pixel.

$$P = \begin{bmatrix} P_{R1} & P_{G1} & P_{B1} \\ P_{R2} & P_{G2} & P_{B2} \\ \dots & \dots & \dots \\ P_{Rk} & P_{Gk} & P_{Bk} \end{bmatrix} = \begin{bmatrix} O_{R1} & O_{G1} & O_{B1} \\ O_{R2} & O_{G2} & O_{B2} \\ \dots & \dots & \dots \\ O_{Rk} & O_{Gk} & O_{Bk} \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$$

Color Correction (3/3)

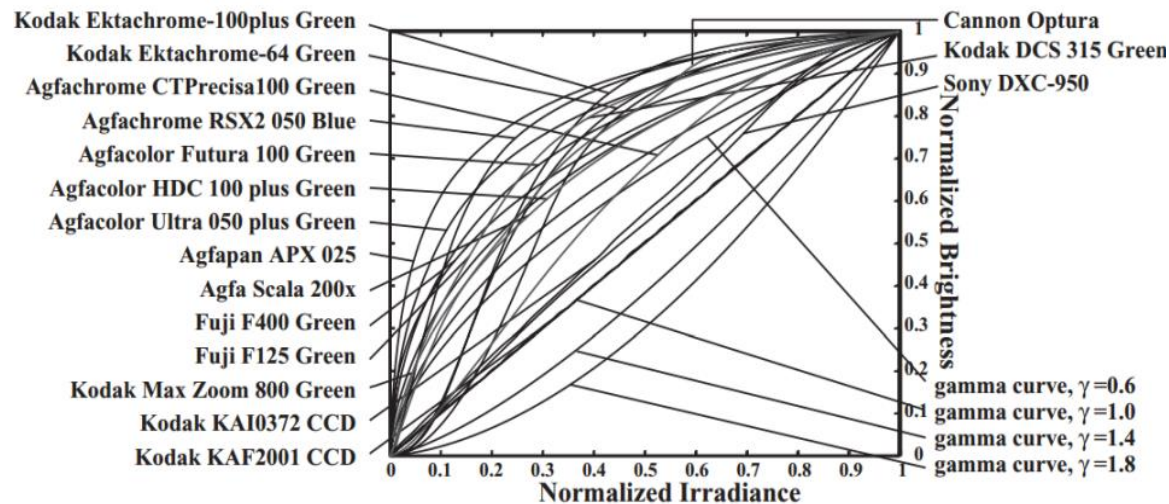
- **Transform Color Space**

- RGB \longleftrightarrow XYZ (CIE 1931)
- White point defined at $X = 1/3, Y = 1/3, Z = 1/3$
- $$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} 0.4887180 & 0.3106803 & 0.2006017 \\ 0.1762044 & 0.8129847 & 0.0108109 \\ 0.0000000 & 0.0102048 & 0.9897952 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$



Tone Mapping

- **Display cannot afford the brightness in real world**
 - use more sensor bits to store
- **High Dynamic Range (HDR) → Low Dynamic Range (LDR)**
 - compress sensor bits into 8-bit (RGB24)
 - makes images suitable to be viewed on a digital screen
- **Tone curve: Non-linear mapping of RGB tones**



Rules

- **Due: Wed, 10/16, 23:59**
- **You should use Python to complete this homework**
- **NO CHEAT !!**

For more details

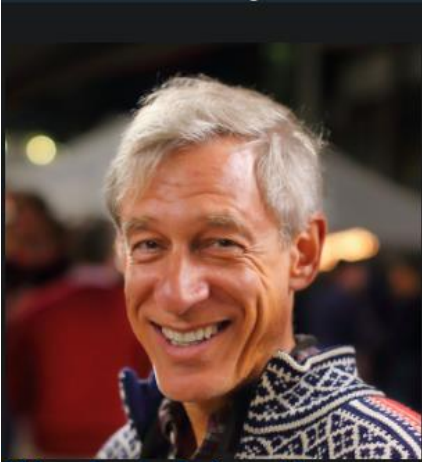
https://www.eecs.yorku.ca/~mbrown/CVPR2016_Brown.html

- M. S. Brown, “Understanding the In-Camera Image Processing Pipeline for Computer Vision”, *IEEE Computer Vision and Pattern Recognition - Tutorial*, June 26, 2016

<http://graphics.stanford.edu/courses/cs178/>

- CS 178 - Digital Photography (Spring 2014)
- Marc Levoy: **Google Pixel 3**

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