

**leti**  
cea tech

LP  
MC

UNIVERSITÉ  
**Grenoble**  
Alpes

# ALGEBRA FOR COLOR CORRECTION IN IMAGE SENSORS

GdR ISIS: géométrie et représentation de la couleur | Axel Clouet, David Alleysson, Jérôme Vaillant | 21/11/2018 | 1

- **PhD student at CEA-LETI (Grenoble) with LPNC:**
  - Supervisors:
    - Jérôme Vaillant (CEA)
    - David Alleysson (LPNC)
  - Title: color restitution in low light level image sensors
  - Main deals:
    - Poor signal to noise ratio
    - Illumination conditions

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Classical RGB



Poor contrast

Low light sensor



False colors

- **Introduction and context**
  - The goal of color sensor design
  - Classical color correction (state of the art)
- **Physics of light capture and sensor limitation**
  - Photon absorption, noise
- **Mathematical representation of light**
  - Discrete approximation
  - Sensor space, color space
  - Color correction interpretation
- **Challenge and prospects**

- **Color image sensor:**
  - Goal: acquire similar data as the human perception in given conditions:



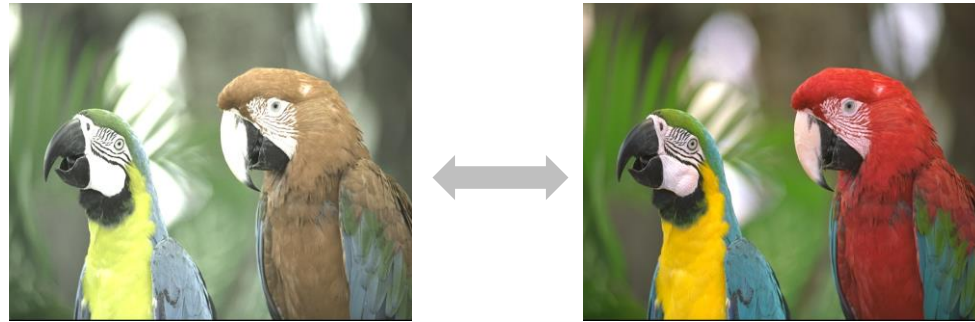
Color sensor



Human visual system

- Color image sensor:

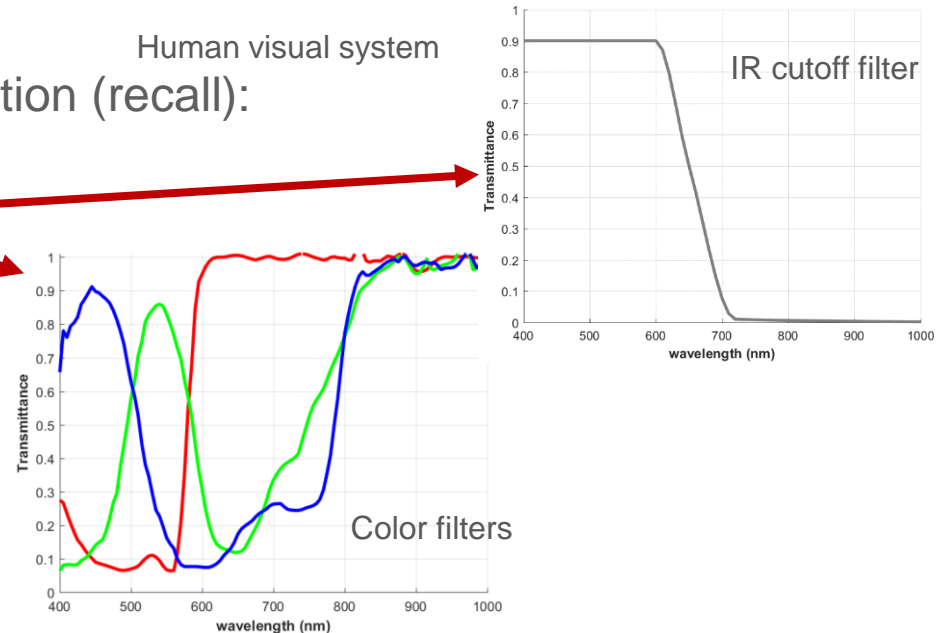
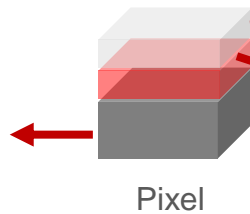
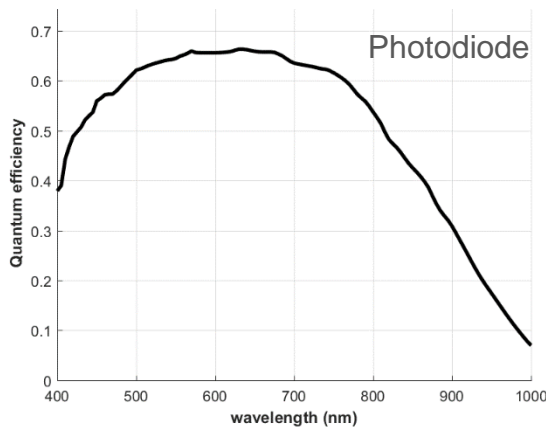
- Goal: acquire similar data as the human perception in given conditions:



Color sensor

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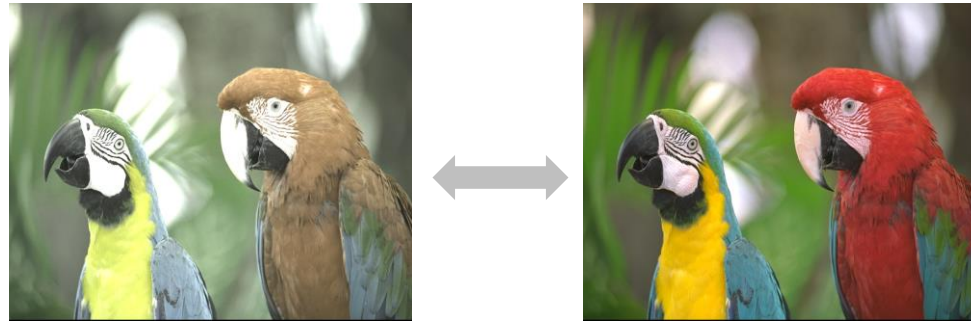
- Key components for color acquisition (recall):



# INTRODUCTION AND CONTEXT

- Color image sensor:

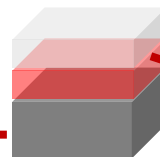
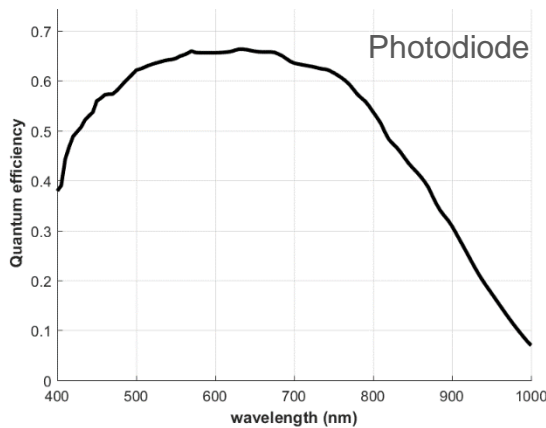
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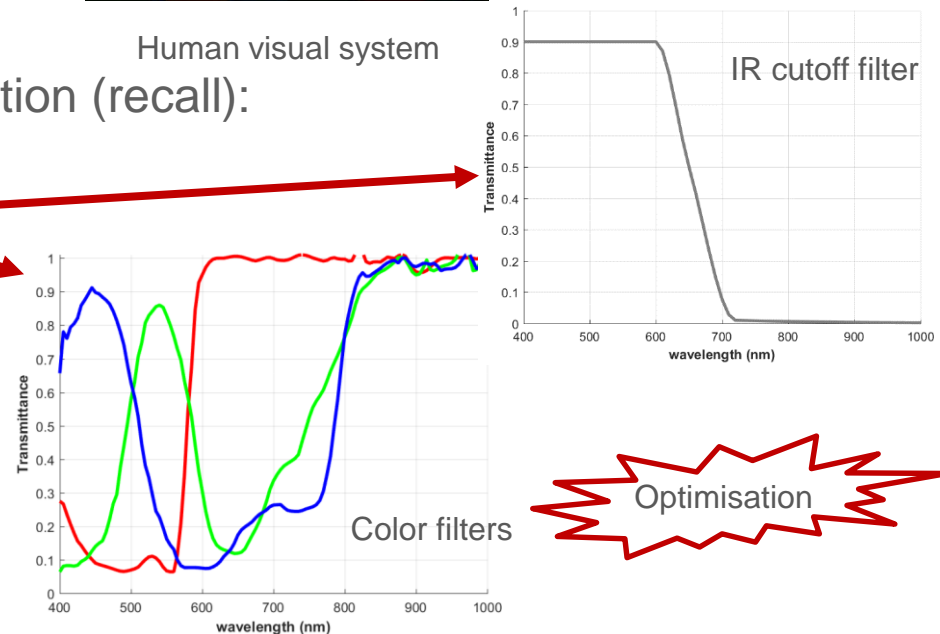
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Human visual system

- Key components for color acquisition (recall):



Pixel



# INTRODUCTION AND CONTEXT

- **Color image sensor:**

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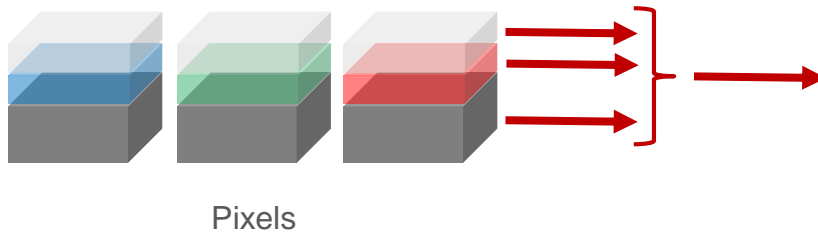


Color sensor

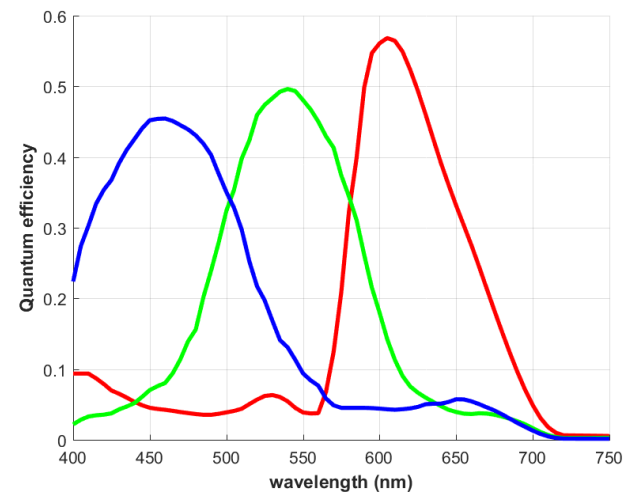


Human visual system

- Key components for color acquisition (recall):



Usual « spectral sensitivities »

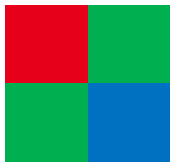




- **State of the art:**
  - Basic image processing:



Demosaicing

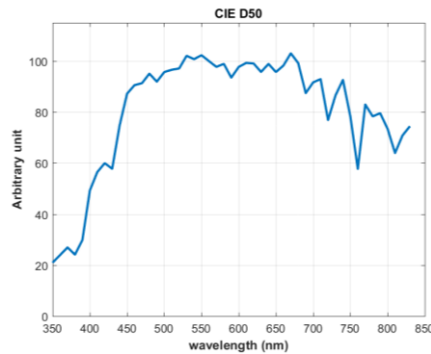
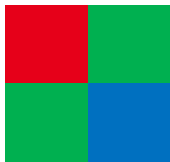


- State of the art:
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Demosaicing

White balance



Illuminant

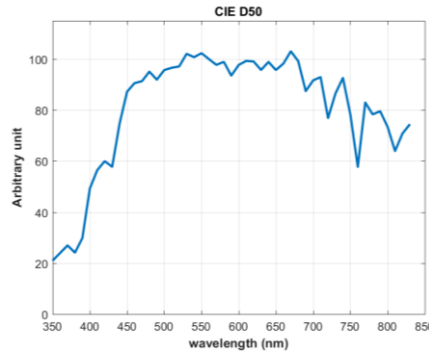
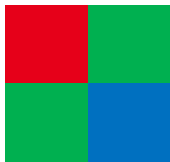
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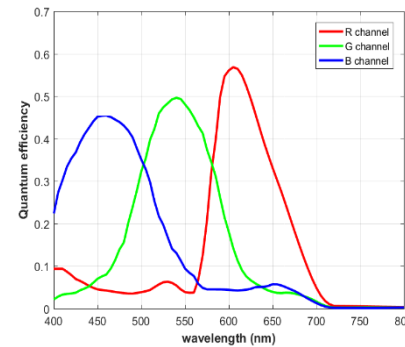
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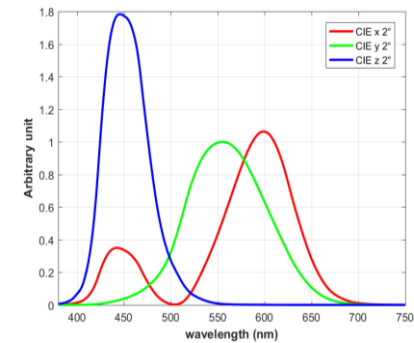
Color correction



Illuminant



QE

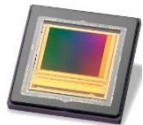


Color matching functions

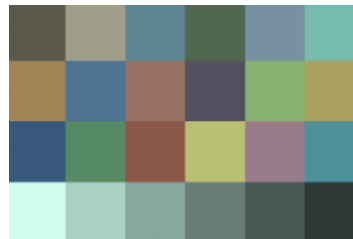
- State of the art:
  - Classical color correction:



$$T = \begin{bmatrix} R_{T1} & \dots & R_{T24} \\ G_{T1} & \dots & G_{T24} \\ B_{T1} & \dots & B_{T24} \\ 1 & \dots & 1 \end{bmatrix}$$



$$S = \begin{bmatrix} R_{S1} & \dots & R_{S24} \\ G_{S1} & \dots & G_{S24} \\ B_{S1} & \dots & B_{S24} \\ 1 & \dots & 1 \end{bmatrix}$$



$$\hat{M} = \underset{M}{\operatorname{argmin}} (\|T - M.S\|^2)$$

$$\hat{M} = T.S^T.(S.S^T)^{-1}$$

$$\hat{M} = \begin{bmatrix} K & [V] \\ 0 \dots 0 & 1 \end{bmatrix}$$

$$K = CCM.WB$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix}_{\text{offset}} = -K^T.(K.K^T)^{-1}.V$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix}_{\text{corr}} = \underbrace{\begin{pmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{pmatrix}}_{\text{CCM}} \cdot \underbrace{\begin{pmatrix} W_{11} & 0 & 0 \\ 0 & W_{22} & 0 \\ 0 & 0 & W_{33} \end{pmatrix}}_{\text{WB}} \left( \begin{bmatrix} R \\ G \\ B \end{bmatrix}_{\text{raw}} - \begin{bmatrix} R \\ G \\ B \end{bmatrix}_{\text{offset}} \right)$$

CCM

WB

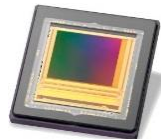
- **State of the art:**
  - Exemple of color correction (computed on hyperspectral image)
  - CIE-D65 illuminant
  - Teledyne Onyx sensor with infrared cutoff filter (slides before)



Color correction

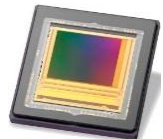


Corrected sRGB data



Raw data

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Raw data

Color correction



Corrected sRGB data

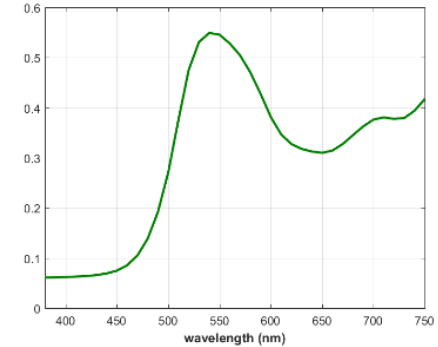
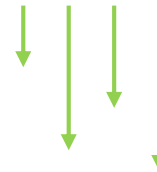
≈



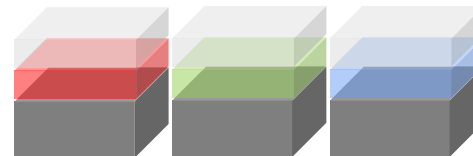
Reference sRGB data

- Photon absorption computation:

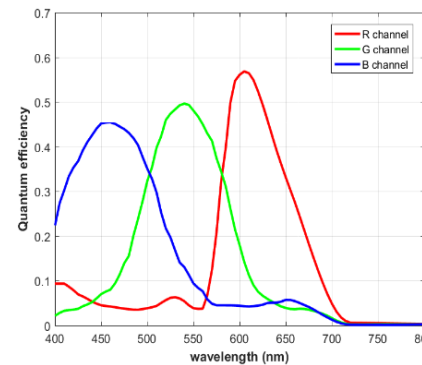
- Physical units:



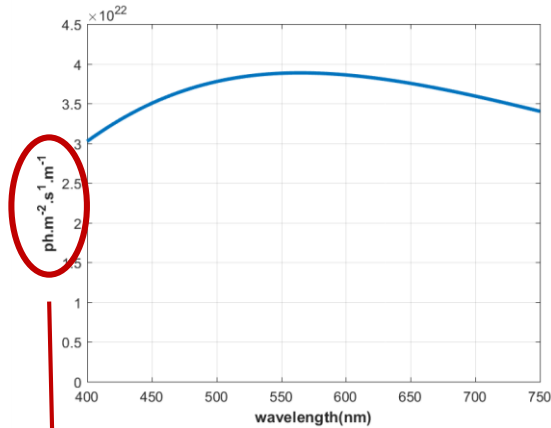
Reflectance rate  
 $R_p(\lambda)$



Pixels



Photon/electron  
conversion rate  
 $QE(\lambda)$



Black body 6500K

Physical absolute unit

- Photon absorption computation:

- Electron measure:

$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_0^{\infty} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

- Limitation of a sensor, the noise:

- Noise: uncertainty on the electron number measurement

- Main source of noise:

- Readout noise (standard deviation given by the manufacturer, some electrons)  
Gaussian
- Photonic shot noise ( $\sigma_{ph}^2 = signal$ )  
Poissonian



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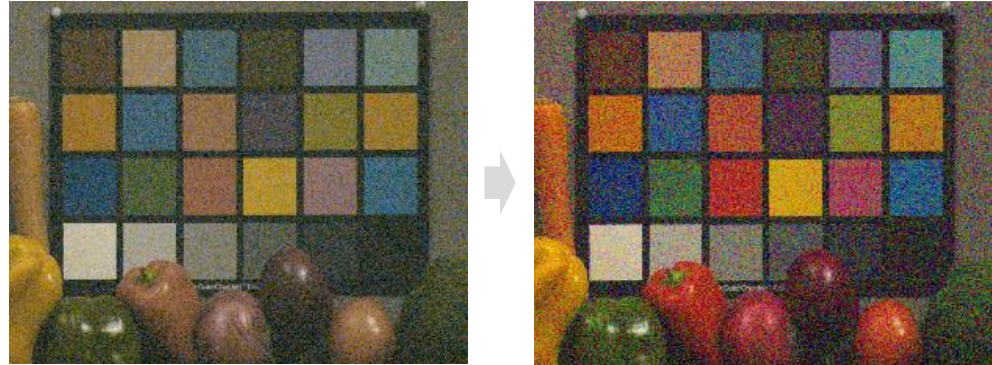
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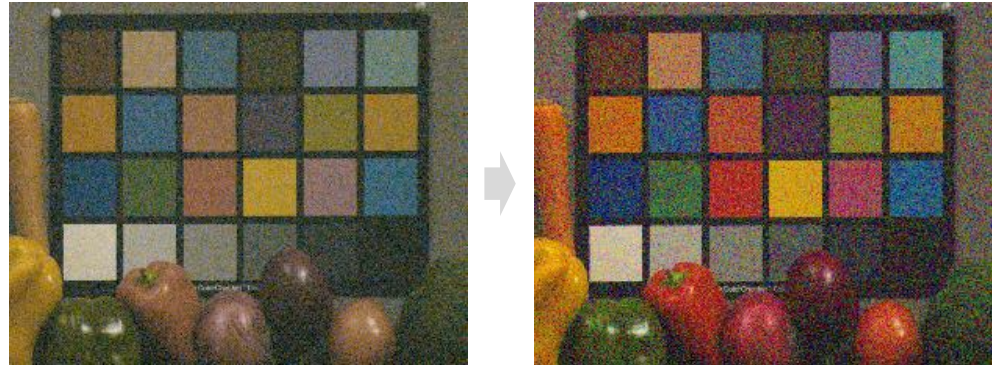
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$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_0^{\infty} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda \pm \Delta M_{e^-}$$

- **Color correction:**
  - Signal and noise amplification
  - Signal to noise ratio decrease

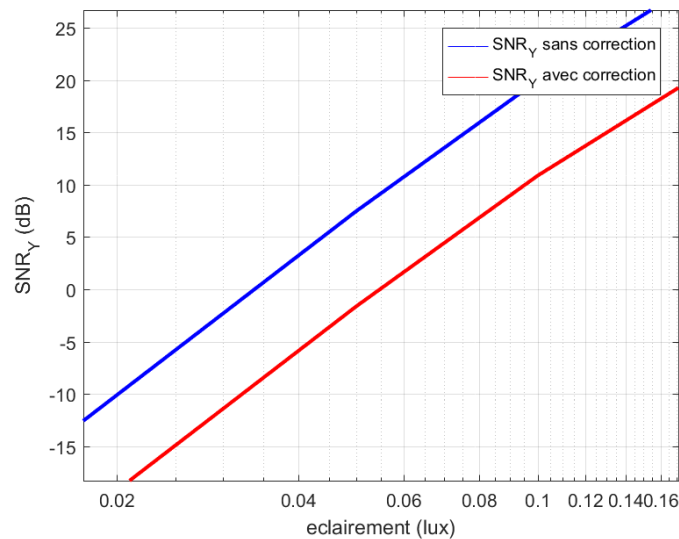


- **Color correction:**
  - Signal and noise amplification
  - Signal to noise ratio decrease



- **Quantitative exemple:**

Usual model: « without correlation between channels »

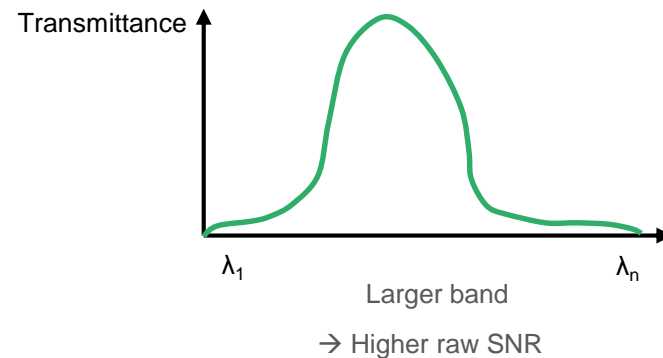
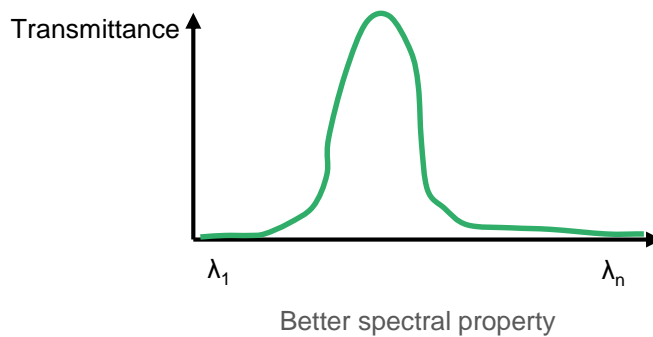


Grey patch 18%



## NOISE PROPAGATION


- Study of noise propagation → adapte spectral transmittances
  - Use large band filter for better raw SNR:



- Final SNR increase? **Not obvious**
  - Noise can be amplified according to spectral shape and number of spectral channels
  - → We need an algebraic representation for noise propagation through color correction

- From physics to algebraic representation:

$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_0^{\infty} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$


 Finite support

$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_{400nm}^{700nm} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

$$M_{e^-} \propto \int_{400nm}^{700nm} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

- From physics to algebraic representation:

$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_0^{\infty} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

↑ Finite support

$$M_{e^-} = \frac{N_{lux} \cdot T_{int} \cdot a_{pix}^2}{4 \cdot f_{\#}^2} \int_{400nm}^{700nm} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

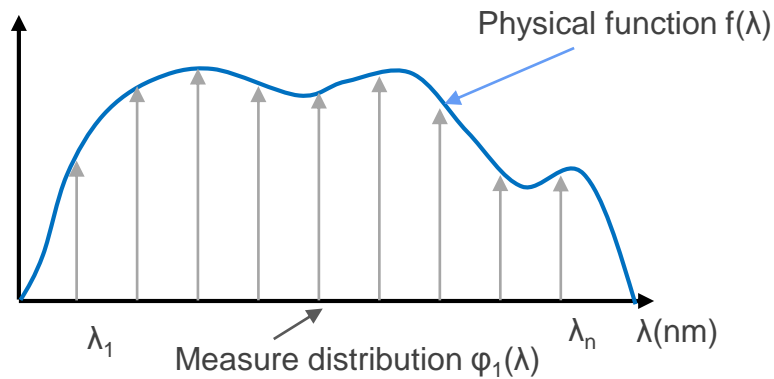
$$M_{e^-} \propto \int_{400nm}^{700nm} I_{lum}(\lambda) \cdot R_p(\lambda) \cdot QE(\lambda) \cdot d\lambda$$

Discretisation:

$$\left\{ \begin{array}{l} M_{e^-} \propto \sum_{k=1}^n I_{lum}(\lambda_k) \cdot R_p(\lambda_k) \cdot QE(\lambda_k) \cdot \Delta\lambda \\ M_{e^-} \propto \mathbf{QE}^T \cdot \mathit{diag}(\mathbf{I}_{lum}) \cdot \mathbf{R}_p \end{array} \right.$$

Matrix form

- Discrete approximation of physics:
  - Measure or digital interpolation:

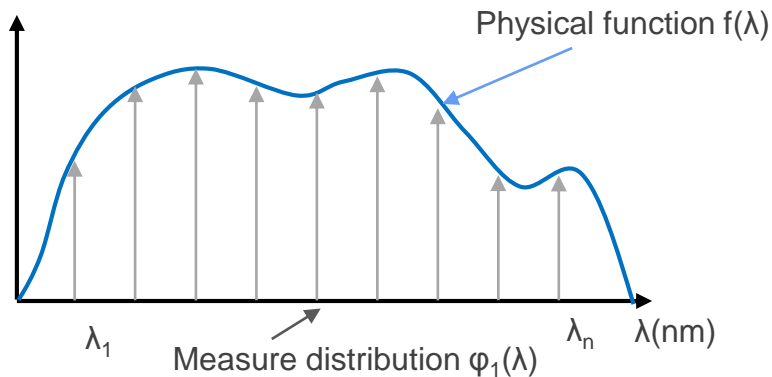


Vectorized function **f**

$f(\lambda_1)$	$= (f * \varphi_1)(\lambda)$
$f(\lambda_2)$	
...	
$f(\lambda_n)$	$= (f * \varphi_n)(\lambda)$

- **Discrete approximation of physics:**

- Measure or digital interpolation:



Vectorized function **f**

$f(\lambda_1)$	$= (f * \varphi_1)(\lambda)$
$f(\lambda_2)$	
...	
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- **Vector representation:**

- $n$ -space  $\rightarrow \mathbb{R}^n$  generated by the canonic base: a finite dimension Hilbert space

- $\vec{f} \in \mathbb{R}^n$

- $\vec{f} = \sum_{k=1}^n f_k \cdot \vec{e}_k$

- $[\vec{e}_1, \dots, \vec{e}_n] = [(1, 0, \dots, 0), \dots, (0, \dots, 0, 1)]$



- **Discrete approximation of physics:**

- Formalism recalling:

- Consider  $f(\lambda)$ , a spectral function
- $\vec{f}$  is the vectorized form of  $f$

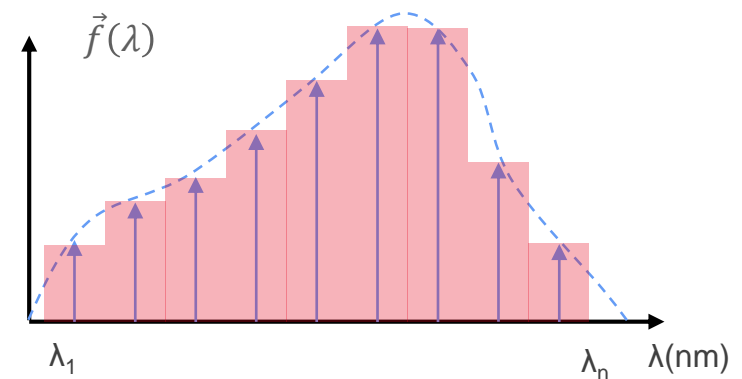
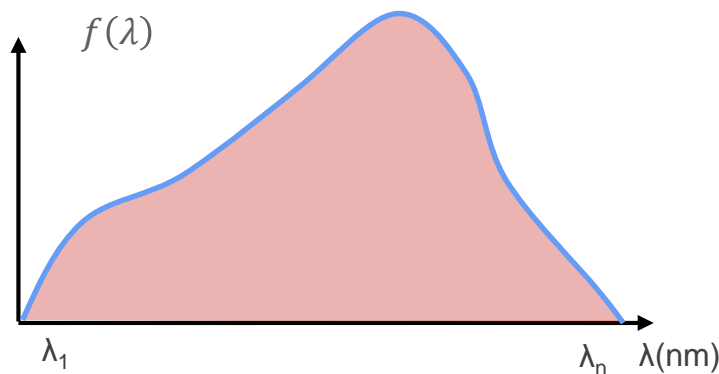
Integration in pre-Hilbert space:

$$\int_{\lambda_1}^{\lambda_n} f(\lambda) \cdot d\lambda$$

$\approx$

« integration » in  $\mathbb{R}^n$ :

$$\sum_{k=1}^n f(\lambda_k) \cdot \Delta\lambda = [\vec{f} \cdot \vec{1}] \cdot \Delta\lambda$$



- Discrete approximation of physics:

- Formalism recalling:

Vectorized function **f**

$f(\lambda_1)$
$f(\lambda_2)$
...
$f(\lambda_n)$

Vectorized function **g**

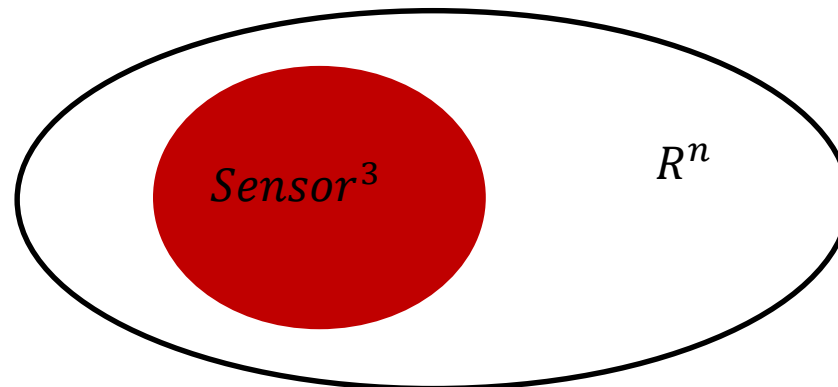
$g(\lambda_1)$
$g(\lambda_2)$
...
$g(\lambda_n)$

Matrix writing:

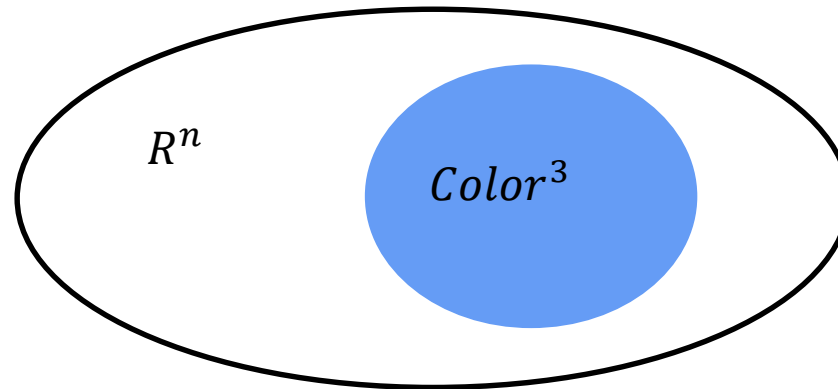
$$a = \vec{f} \cdot \vec{g} = \mathbf{f}^T \mathbf{g}$$

- For vectorial study no need to scale the scalar products (physical parameters of  $\Delta\lambda$ )

- **Sensor algebraic space:**
  - Consider an RGB image sensor having spectral channels such as  $(\vec{r}, \vec{g}, \vec{b})$  are  $R^n$  elements.
    - With :  $\vec{r} = \sum_{k=1}^n r_k \cdot \vec{e}_k$  ,  $\vec{g} = \sum_{k=1}^n g_k \cdot \vec{e}_k$  ,  $\vec{b} = \sum_{k=1}^n b_k \cdot \vec{e}_k$
  - This family is free and generates a 3 dimensions  $R^n$  subspace:



- **Color (or display) algebraic space:**
  - Consider a standard color space such as XYZ generated in  $R^n$  by  $(\vec{x}, \vec{y}, \vec{z})$ .
    - With :  $\vec{x} = \sum_{k=1}^n x_k \cdot \vec{e}_k$  ,  $\vec{y} = \sum_{k=1}^n y_k \cdot \vec{e}_k$  ,  $\vec{z} = \sum_{k=1}^n z_k \cdot \vec{e}_k$
  - This family is free and generates a 3 dimensions  $R^n$  subspace:

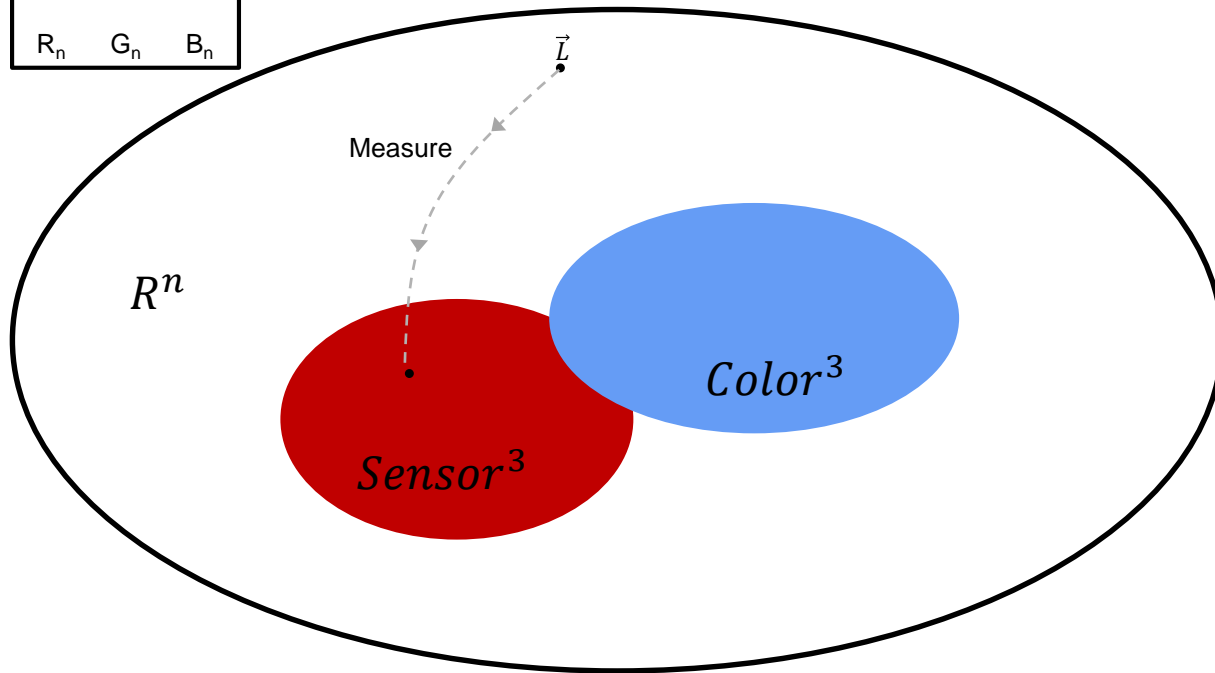
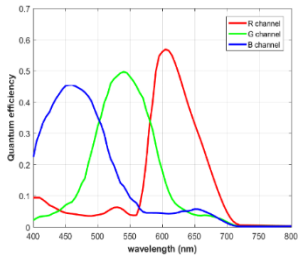


- **Schematic radiance projection:**
  - **L** radiance spectrum, **F** quantum efficiencies, **H** color matching functions

Measure operator:

$$F = \begin{matrix} R_1 & G_1 & B_1 \\ | & | & | \\ | & | & | \\ | & | & | \\ | & | & | \\ R_n & G_n & B_n \end{matrix}$$

$$\text{Measure} = F^T \cdot L$$



- Schematic radiance projection:
  - L radiance spectrum, F quantum efficiencies, H color matching functions

Measure operator:

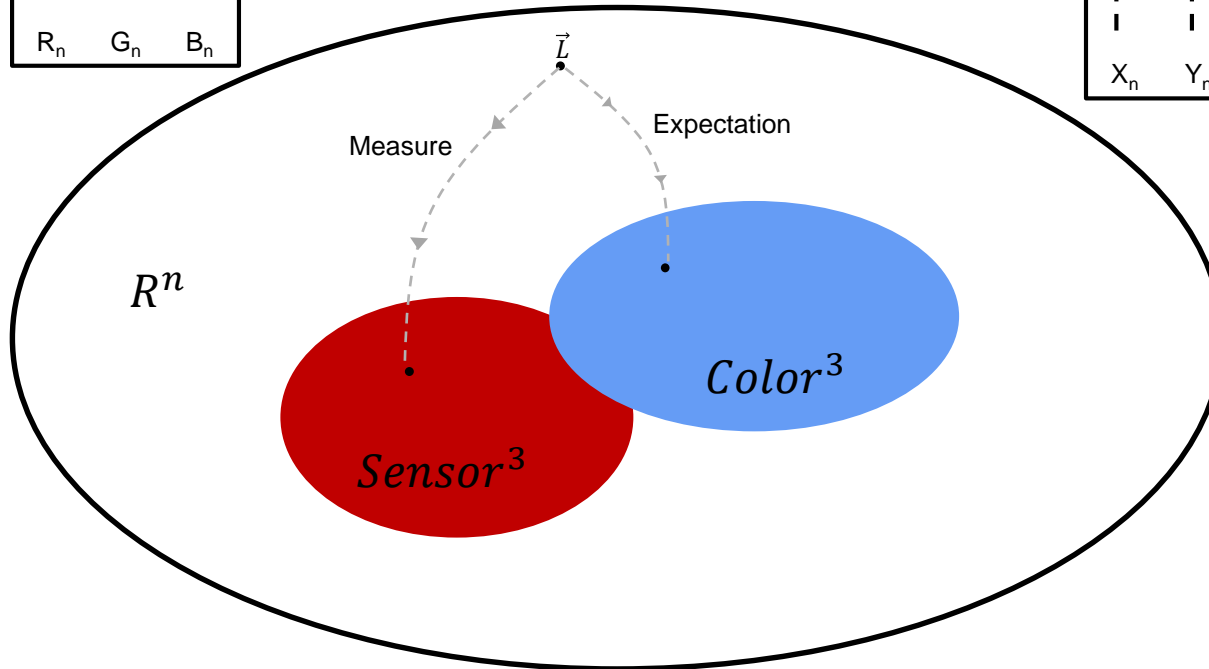
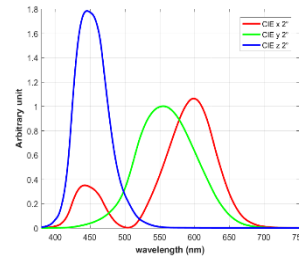
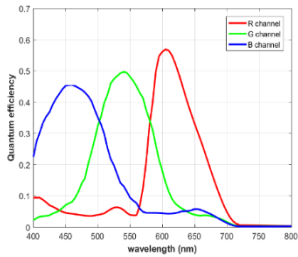
$$F = \begin{bmatrix} R_1 & G_1 & B_1 \\ \vdots & \vdots & \vdots \\ R_n & G_n & B_n \end{bmatrix}$$

$$\text{Measure} = F^T \cdot L$$

Expectation operator:

$$H = \begin{bmatrix} X_1 & Y_1 & Z_1 \\ \vdots & \vdots & \vdots \\ X_n & Y_n & Z_n \end{bmatrix}$$

$$\text{Expectation} = H^T \cdot L$$



- Schematic radiance projection:
  - L radiance spectrum, F quantum efficiencies, H color matching functions

Measure operator:

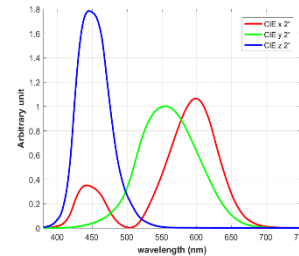
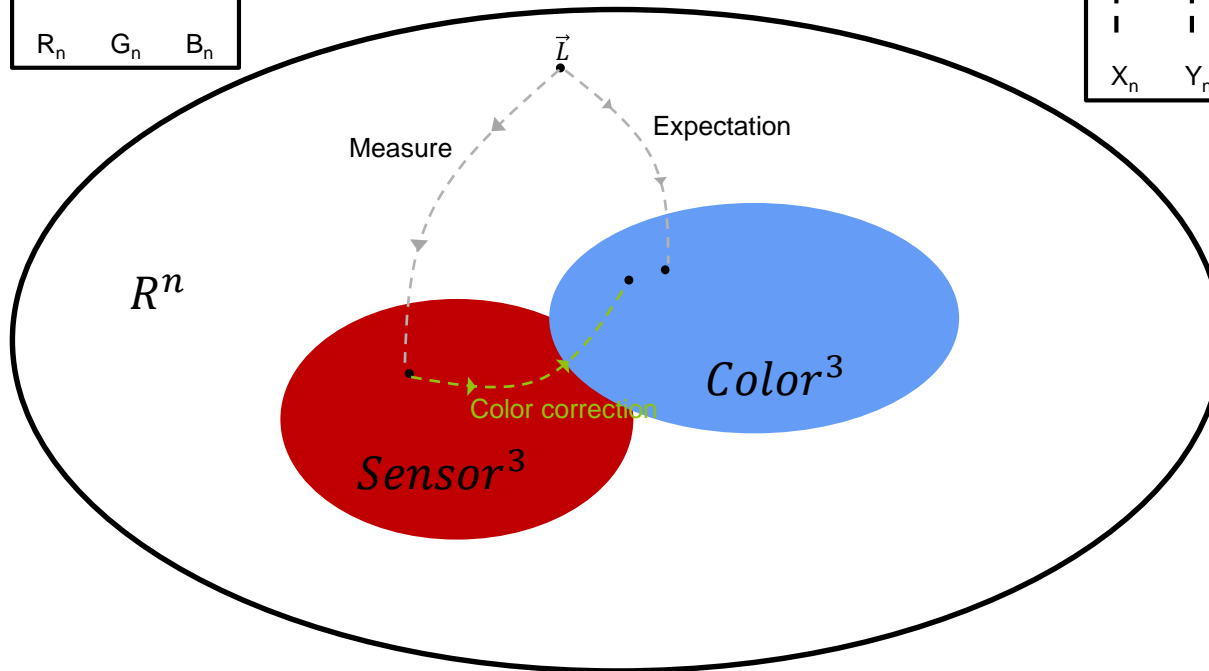
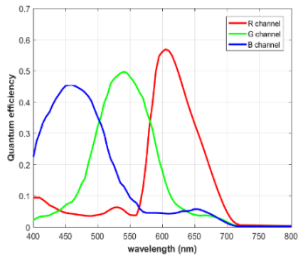
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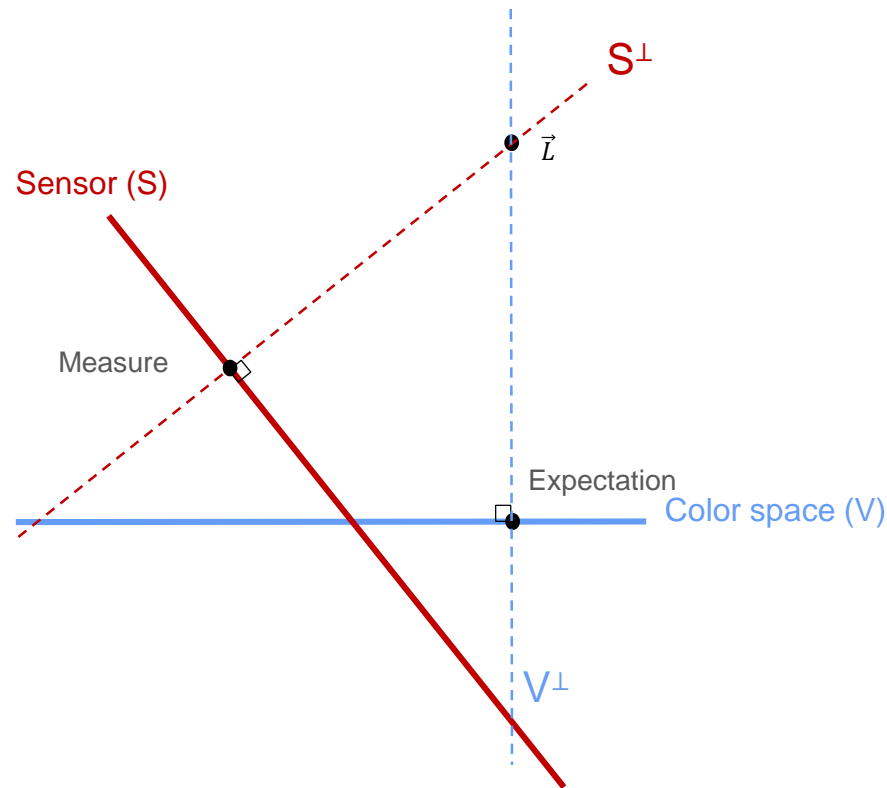
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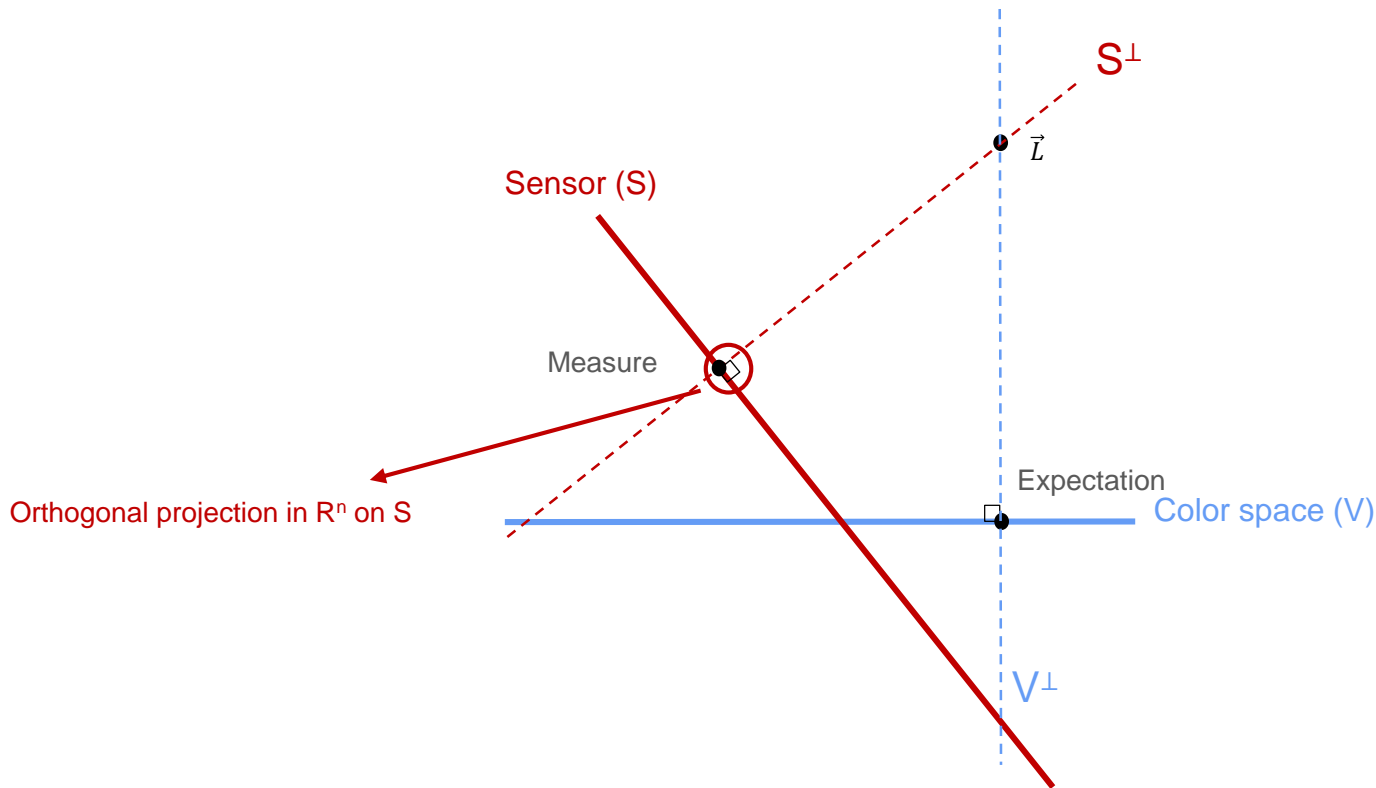


- Another schematic view of color correction:
  - Raw acquisition:

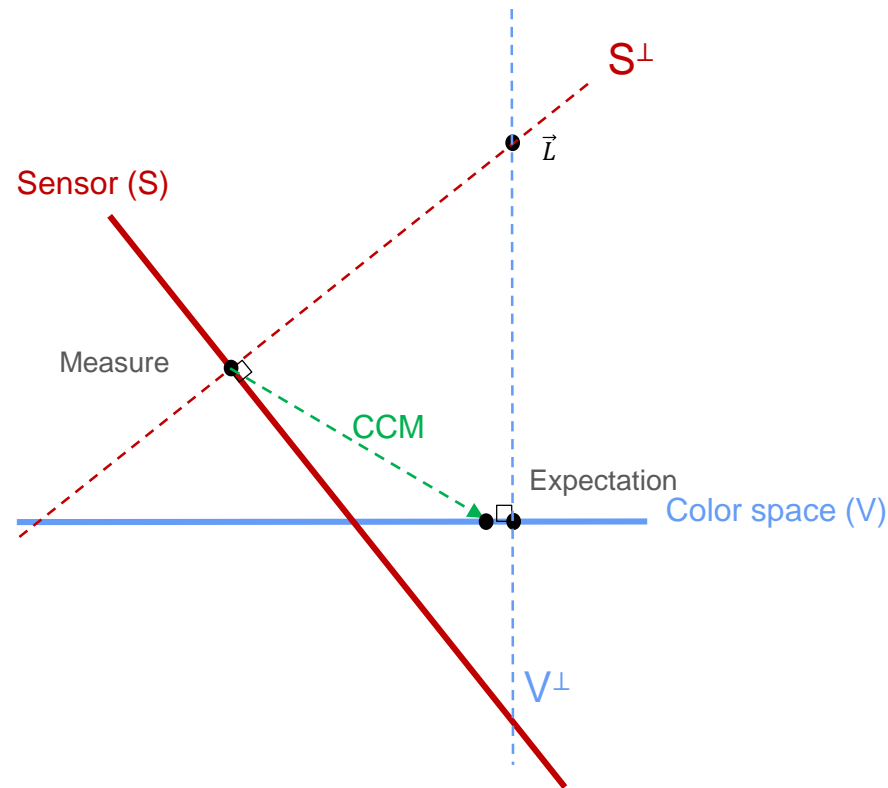




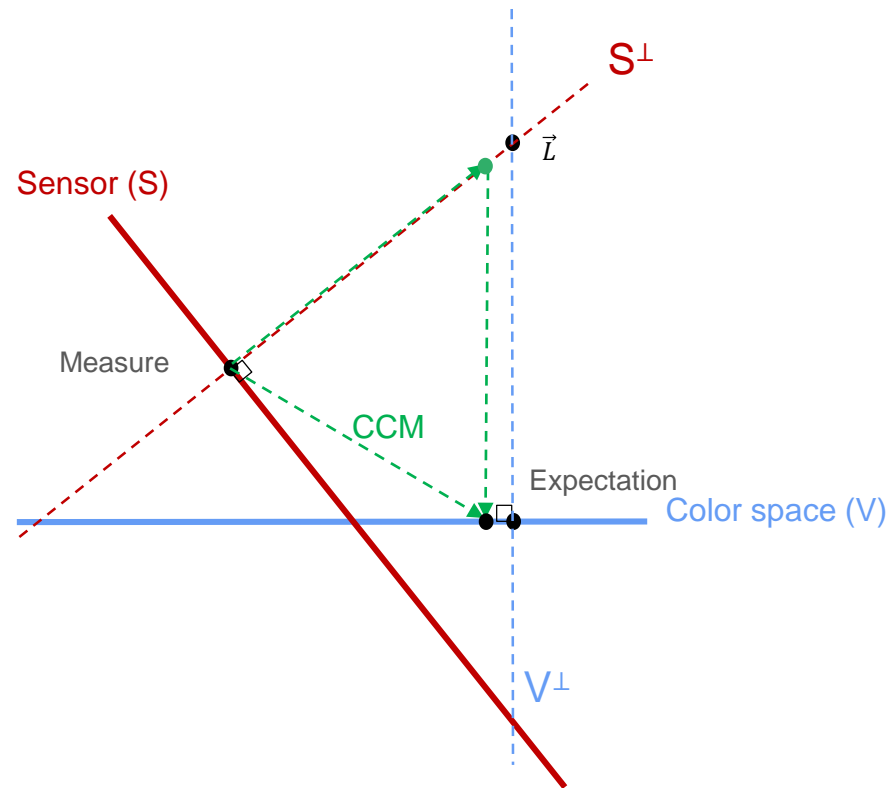
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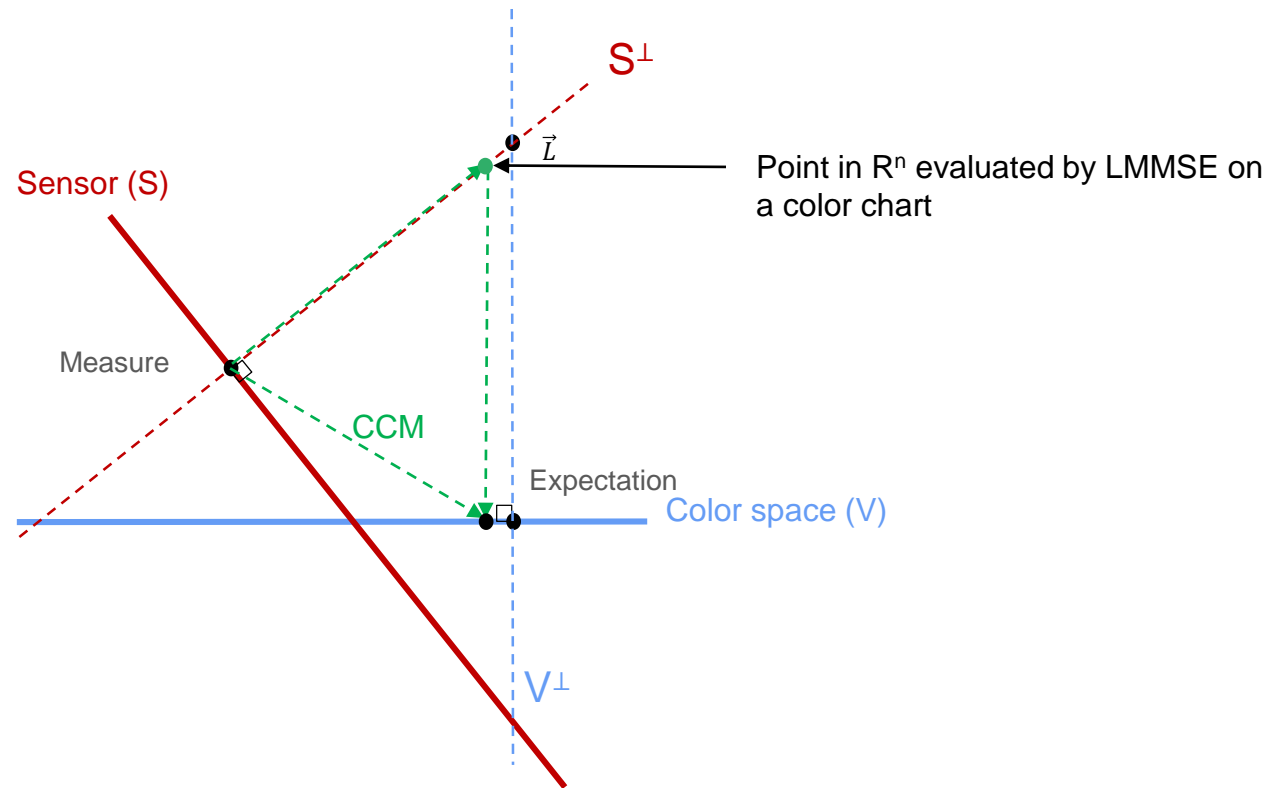
- Another schematic view of color correction:
  - Application of the CCM:



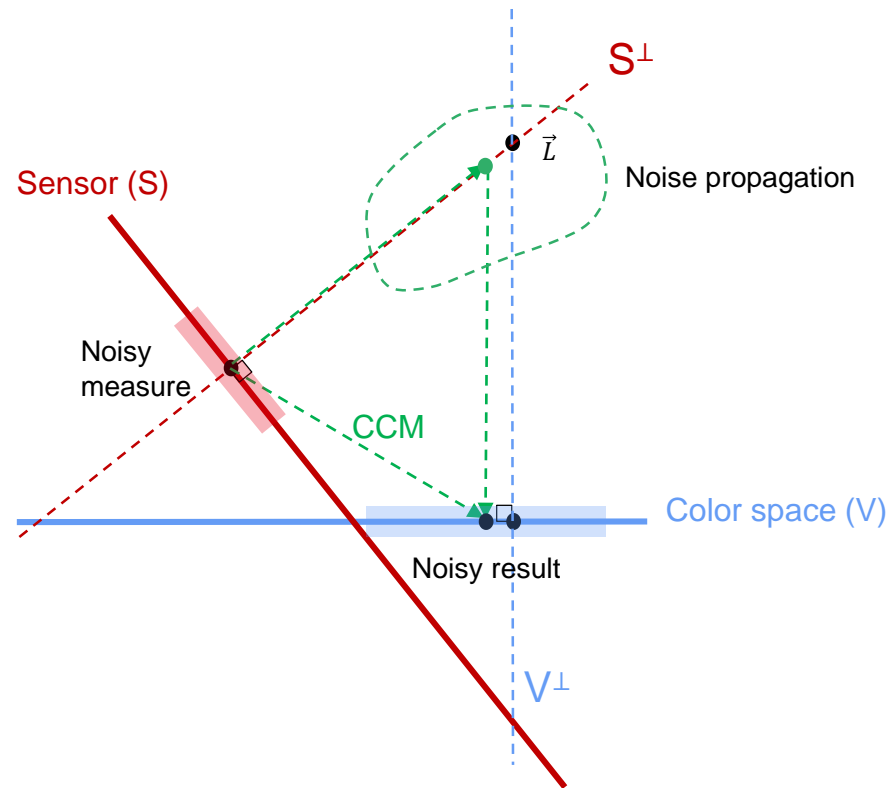
- Another schematic view of color correction:
  - Decomposition of the CCM:



- Another schematic view of color correction:
  - Decomposition of the CCM:



- Another schematic view of color correction:
  - Noise and CCM:

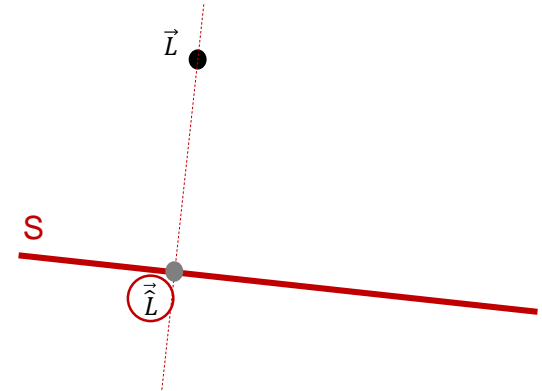


- **Color correction Kernel (not scaled):**

- 1) Radiance spectrum evaluation:

- Cohen operator (explicit form):

$$\hat{L} = F \cdot (F^T \cdot F)^{-1} \cdot F^T \cdot L$$



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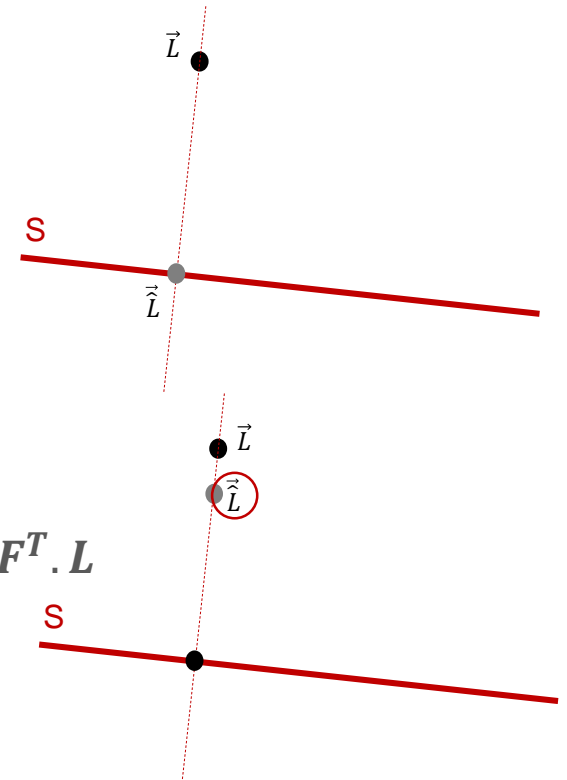
- Cohen operator (explicit form):

$$\hat{L} = F \cdot (F^T \cdot F)^{-1} \cdot F^T \cdot L$$

- Use of a dataset (state of the art method):

$$\hat{L} = z_{set} \cdot (F^T z_{set})^T \cdot (F^T z_{set} \cdot (F^T z_{set})^T)^{-1} \cdot F^T \cdot L$$

$n \times p$  matrix  
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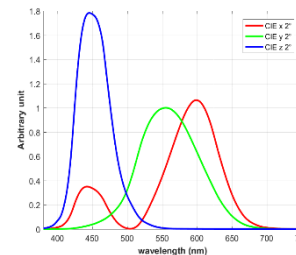
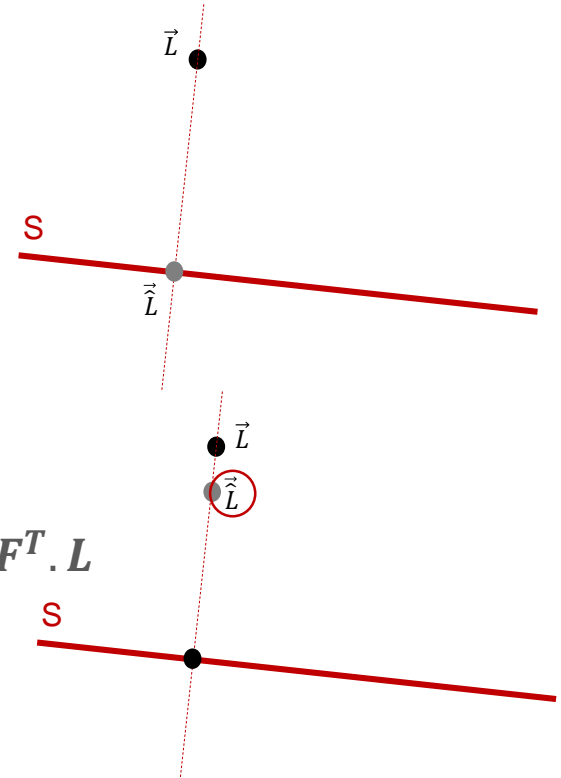
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- 2) Color space projection:

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix}_{corrected} = H^T \cdot \hat{L}$$

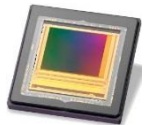




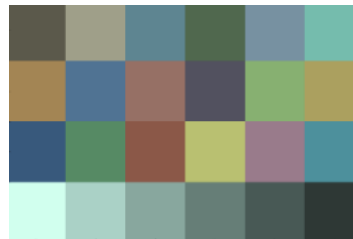
- State of the art (simplified):
  - Classical color correction:



$$T = \begin{bmatrix} R_{T1} & \dots & R_{T24} \\ G_{T1} & \dots & G_{T24} \\ B_{T1} & \dots & B_{T24} \end{bmatrix}$$



$$S = \begin{bmatrix} R_{S1} & \dots & R_{S24} \\ G_{S1} & \dots & G_{S24} \\ B_{S1} & \dots & B_{S24} \end{bmatrix}$$



$$\hat{M} = \underset{M}{\operatorname{argmin}} (\|T - M.S\|^2)$$

$$\hat{M} = T.S^T.(S.S^T)^{-1}$$

$$\hat{M} = CCM.WB$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix}_{corr} = \hat{M} \begin{bmatrix} R \\ G \\ B \end{bmatrix}_{raw}$$

- **Color correction matrix:**
  - Complete expression:

$$\hat{M} = C \cdot H^T \cdot z_{set} \cdot (F^T z_{set})^T \cdot (F^T z_{set} \cdot (F^T z_{set})^T)^{-1}$$

Scaling matrix
Color space projection
Spectral evaluation operator

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Scaling matrix    Color space projection    Spectral evaluation operator

- Analogy with the state of the art:

$$\hat{M} = C \cdot (H^T \cdot z_{set}) \cdot (F^T z_{set})^T \cdot (F^T z_{set} \cdot (F^T z_{set})^T)^{-1}$$

«  $\propto T$  »    «  $\propto S$  »

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- Scaling: in 8-bits acquisition  $\rightarrow$  display in sRGB standard

$$\begin{matrix} \left[ \begin{matrix} R \\ G \\ B \end{matrix} \right]_{sRGB} \\ \leftarrow [0, 255] \end{matrix} = \hat{M} \cdot \begin{matrix} \left[ \begin{matrix} R \\ G \\ B \end{matrix} \right]_{raw} \\ \leftarrow [0, 255] \end{matrix}$$

- **Scaling:**
  - Important for compact writing of CCM:  $3 \times 3$  fix matrix
  - Physical point of view: absolute scale  $\rightarrow$  classical geometry
  - Perception point of view: relative scale  $\rightarrow$  projective geometry?
    - To adapt formalism: homogenous coordinates
    - Homography as spectral color correction

- **Image sensor industry:**
  - Use of empirical color correction
  - Work in absolute physical units
- **Cognitive and mathematical color science:**
  - Mathematical approach
  - Relative scales
- **Strong link between domains useful to understand**

Thanks for attention!

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