

High-Dynamic-Range Imaging and Tonemapping

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Problems with setting the camera exposure level



- Under-exposed
 - Highlight details captured
 - Shadow details lost



- Over-exposed
 - Highlight details lost
 - Shadow details captured

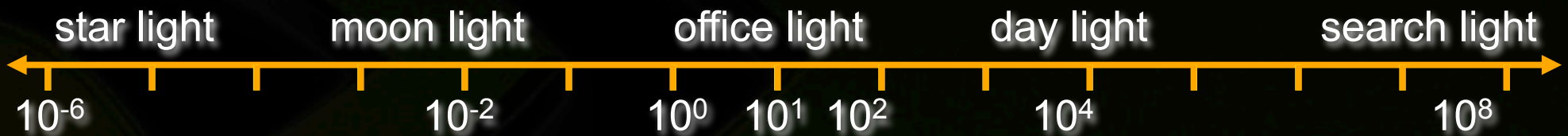
Dynamic range



Dynamic range



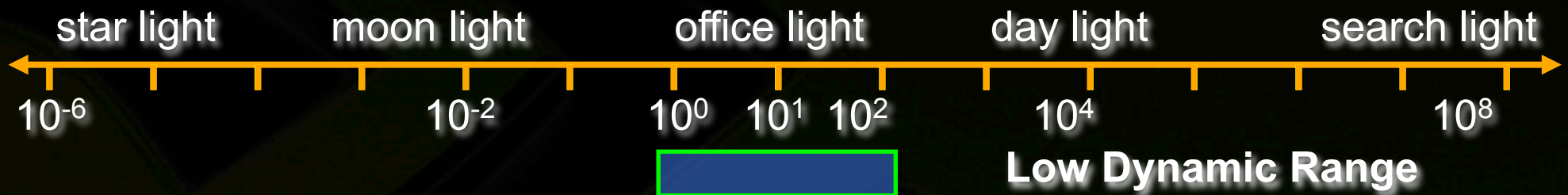
- Eye can adapt from $\sim 10^{-6}$ to 10^8 cd/m²



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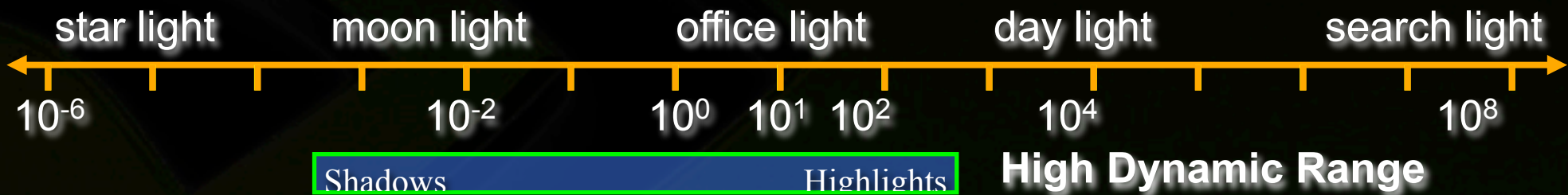
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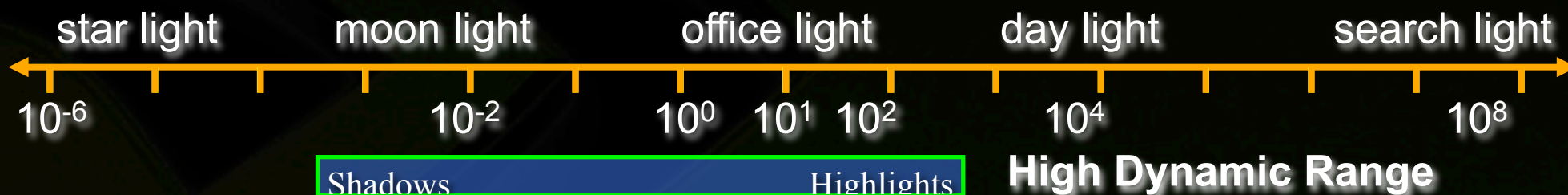
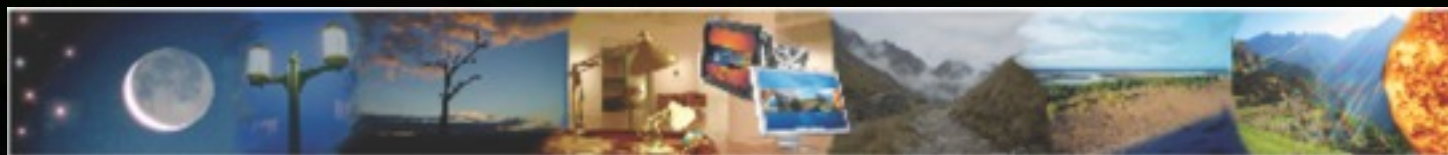


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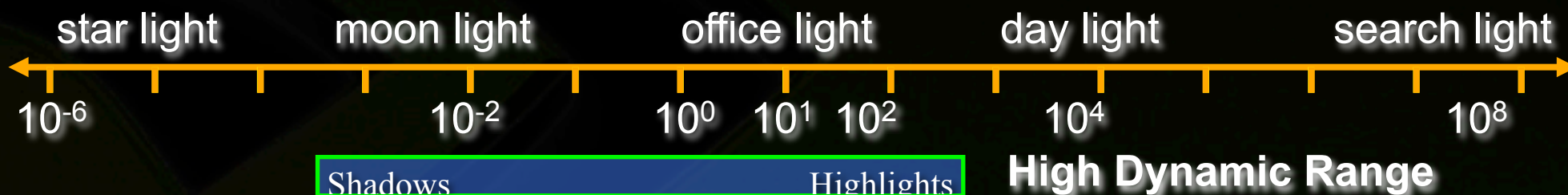
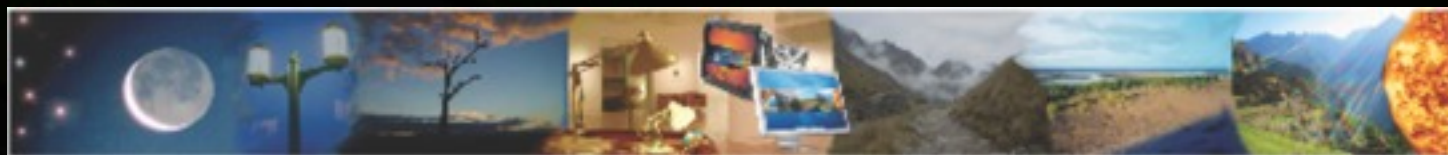
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- Sometimes 1 : 100,000 in a scene



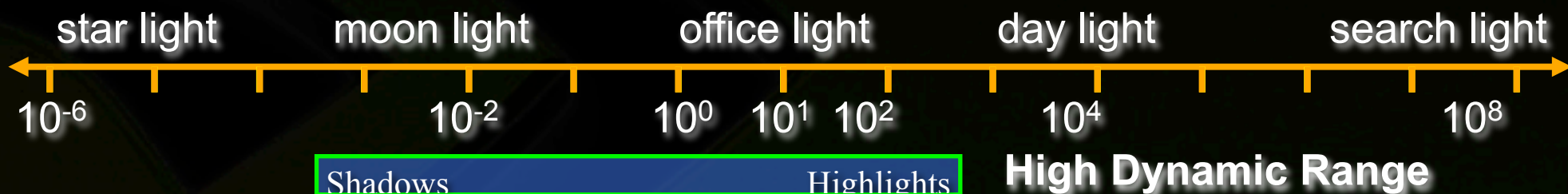
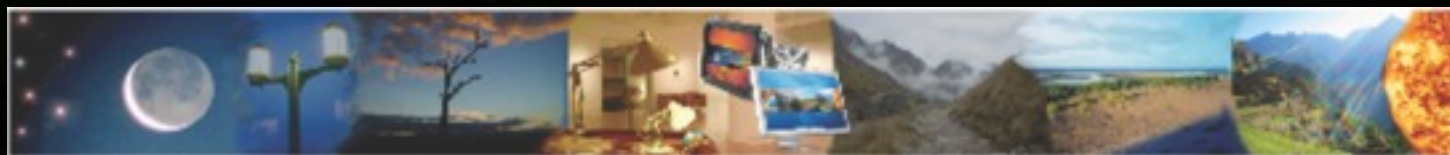
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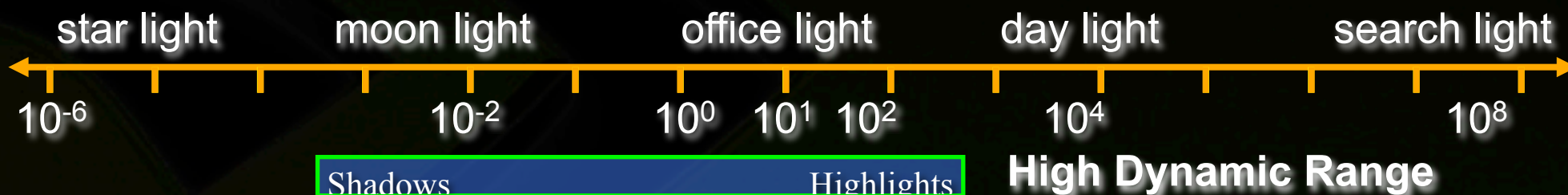
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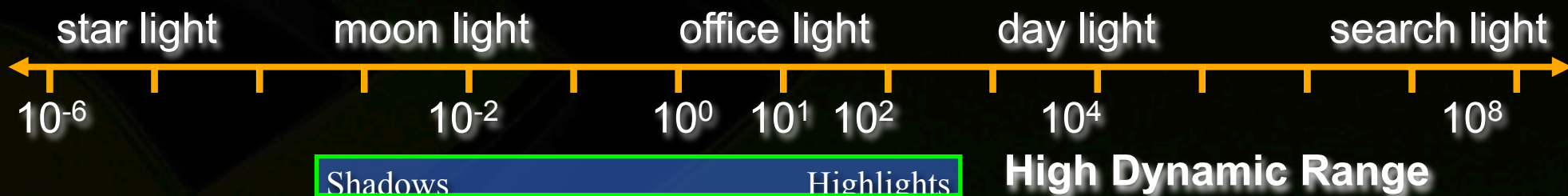
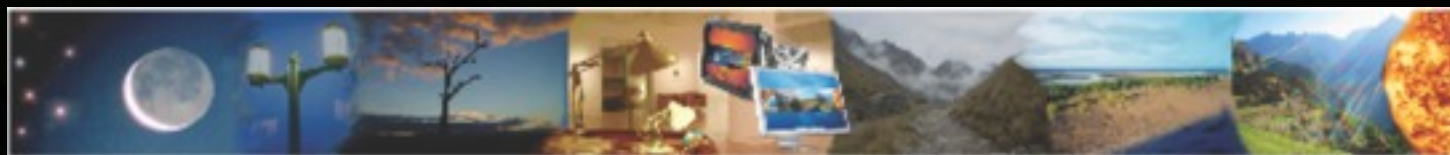
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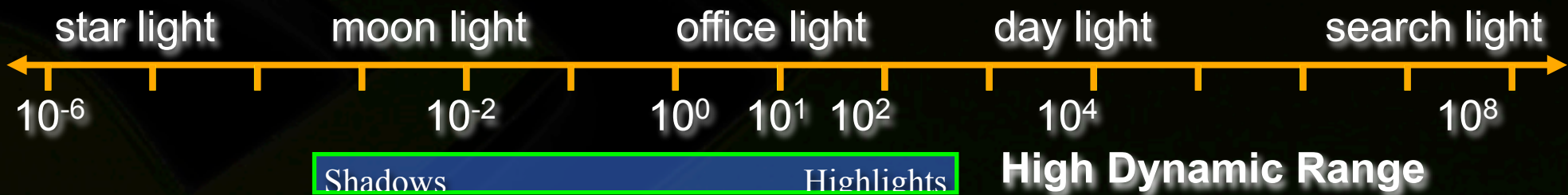
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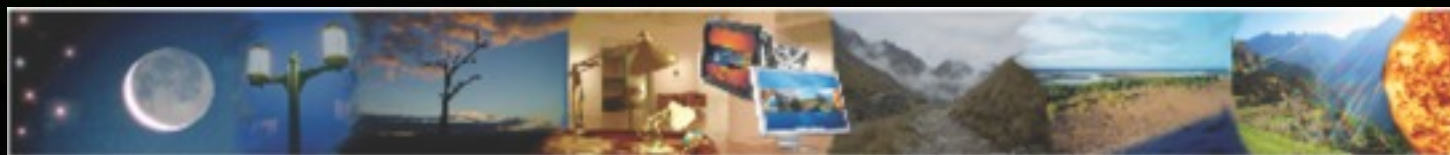


Shadows Highlights **High Dynamic Range**

- Without adaptation eye can handle about 1 : 10000
 - Scotopic Mesopic Photopic
 - Even 1 : 1000 easily enough for scenes with non-specular reflectance

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Shadows Highlights **High Dynamic Range**

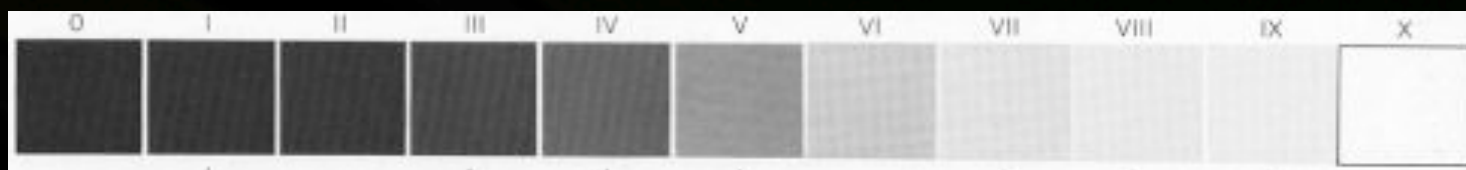
- Without adaptation eye can handle about 1 : 10000
 - Scotopic Mesopic Photopic
 - Even 1 : 1000 easily enough for scenes with non-specular reflectance
- Most displays can handle less than 1 : 100

Range of Typical Displays:

from ~ 1 to ~ 100 cd/m²

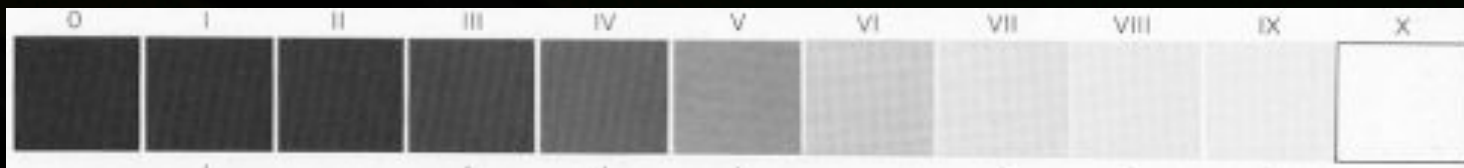


Dynamic Range = Contrast Ratio



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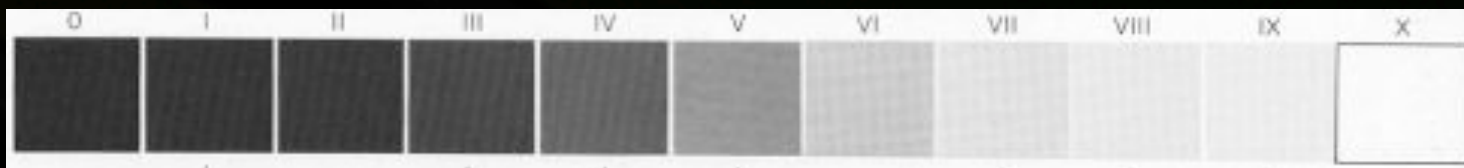
- Depends on the **overall brightness range** and the **smallest visible step**
 - the dynamic range of the image below is 10:1



- we could increase the dynamic range by either
 - using smaller steps
 - adding even brighter whites
 - adding even darker blacks

Dynamic Range = Contrast Ratio



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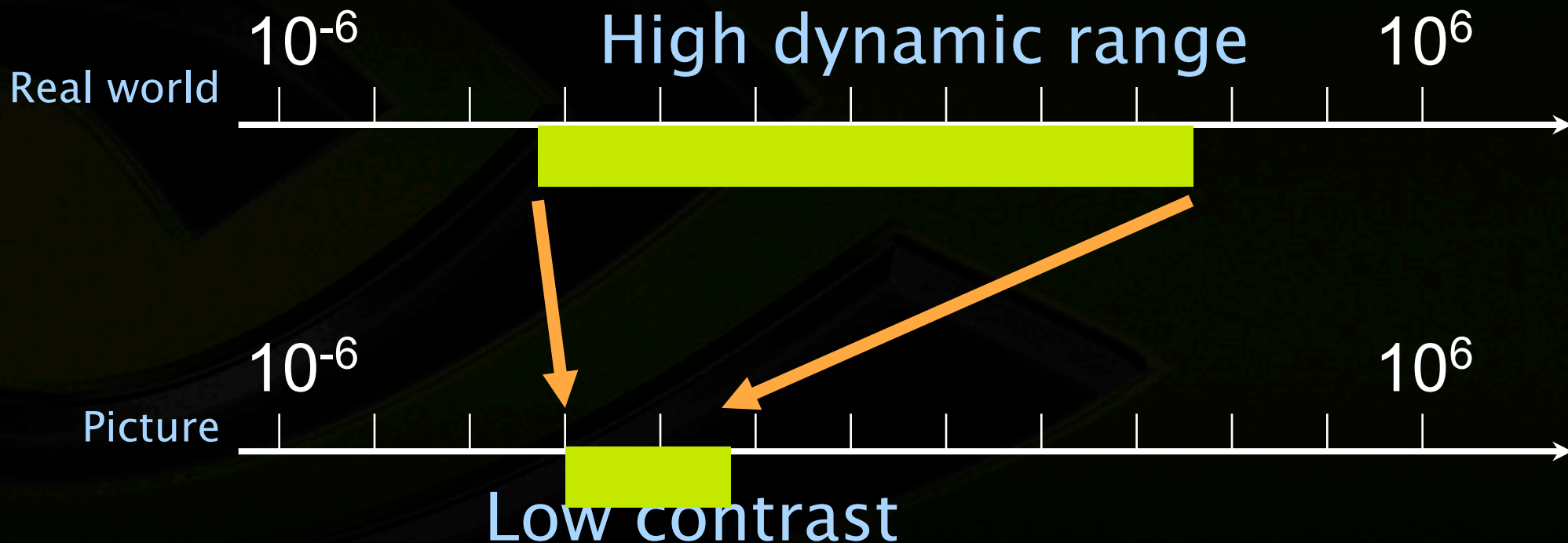


- we could increase the dynamic range by either
 - using smaller steps
 - adding even brighter whites
 - adding even darker blacks
- Originally audio term
 - ratio of the max intensity over the base noise level

Dynamic range of print



- Typically 1:20 to 1:50
 - Black  is ~50x darker than white 
 - Max 1:500



Why is it difficult ?



Why is it difficult ?



- Is it harder to obtain good blacks, or good whites?

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- Black is harder. It's hard to absorb all the light.
 - See the history of painting: good blacks appeared late

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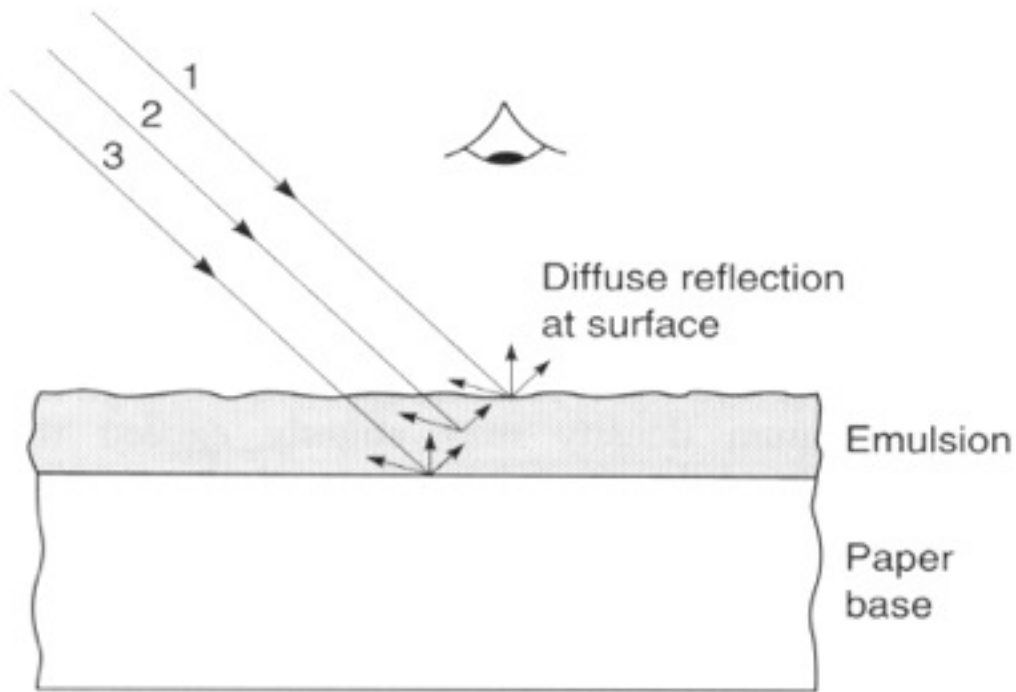


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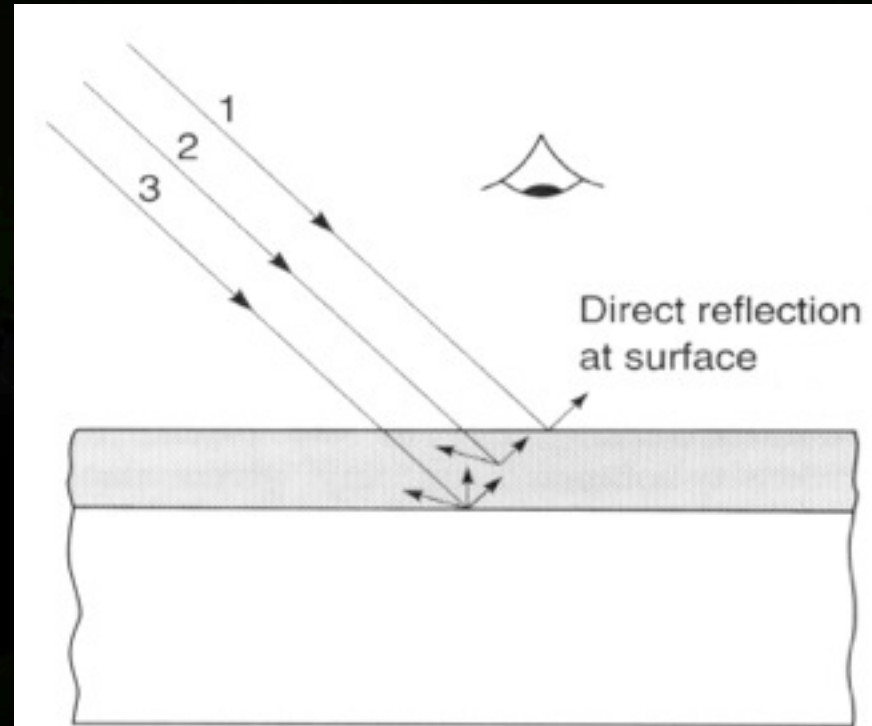
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- Is it harder to obtain good blacks, or good whites?
- Black is harder. It's hard to absorb all the light.
 - See the history of painting: good blacks appeared late
- We can achieve excellent white
 - Albedo $> 100\%$
 - How is this possible?
 - Use fluorescence
 - Most white materials (paper, paint, fabric) are fluorescent

Photo paper dynamic range



(a) Matt paper

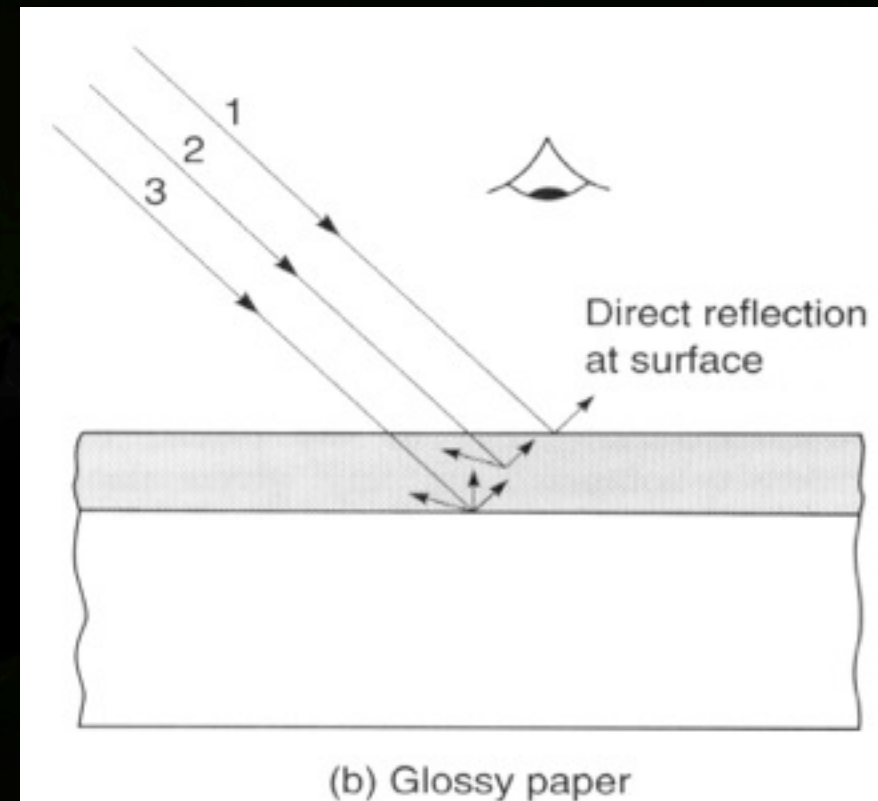
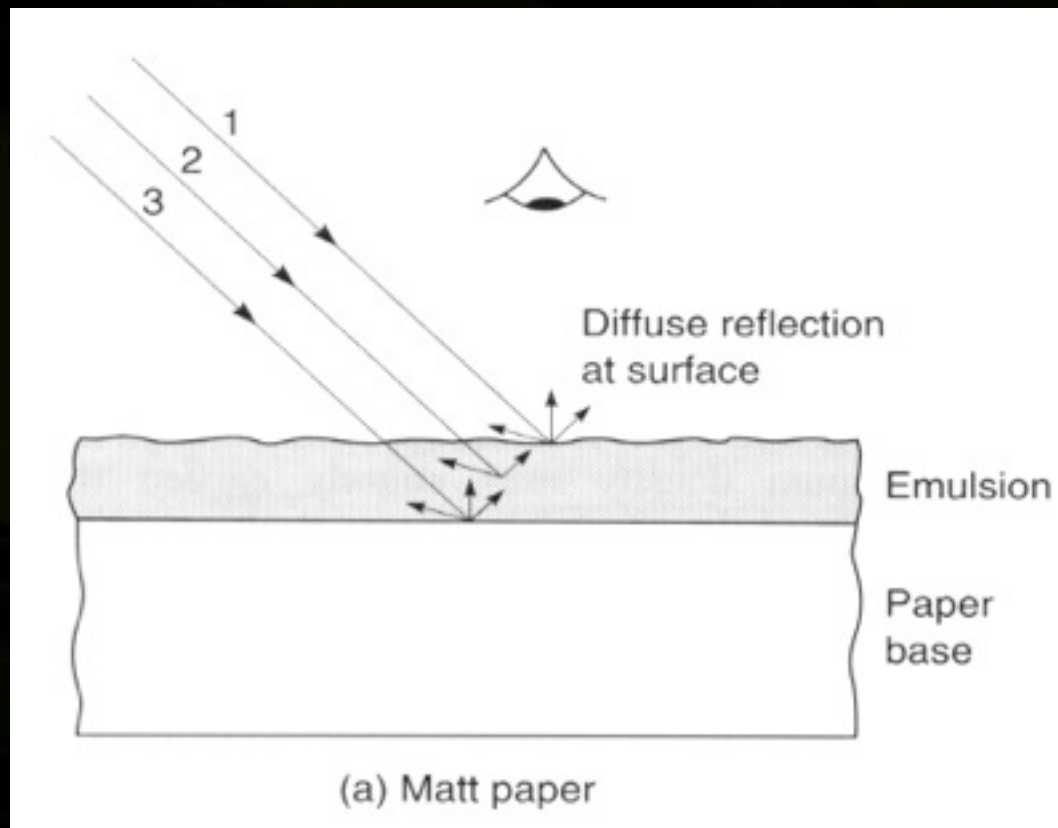


(b) Glossy paper

From The Manual of Photography, Jacobson et al.

Photo paper dynamic range

- Matte vs. glossy: which has the highest dynamic range?
 - Glossy because for some directions, it does not reflect light at all, while matte reflects equally in all directions

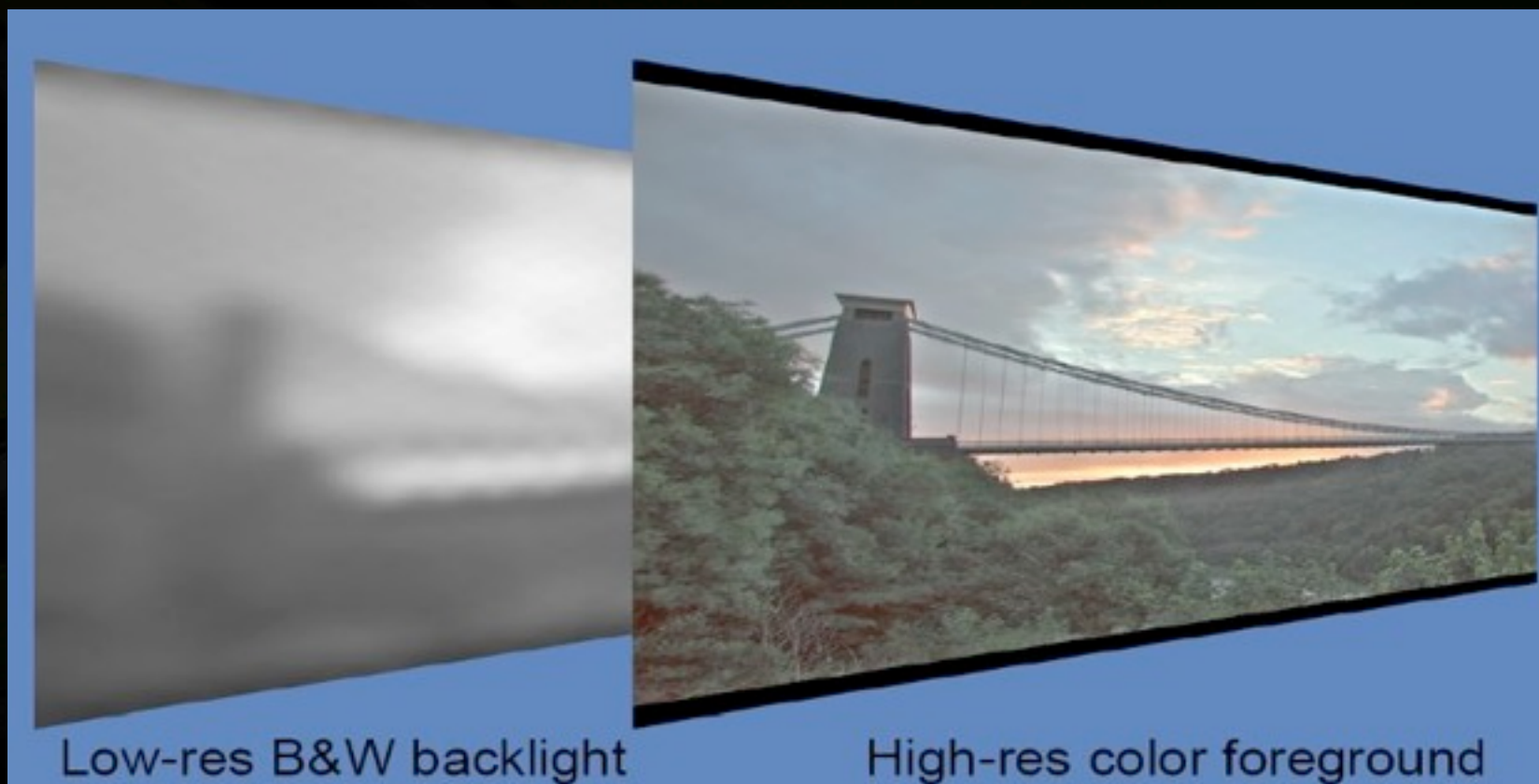


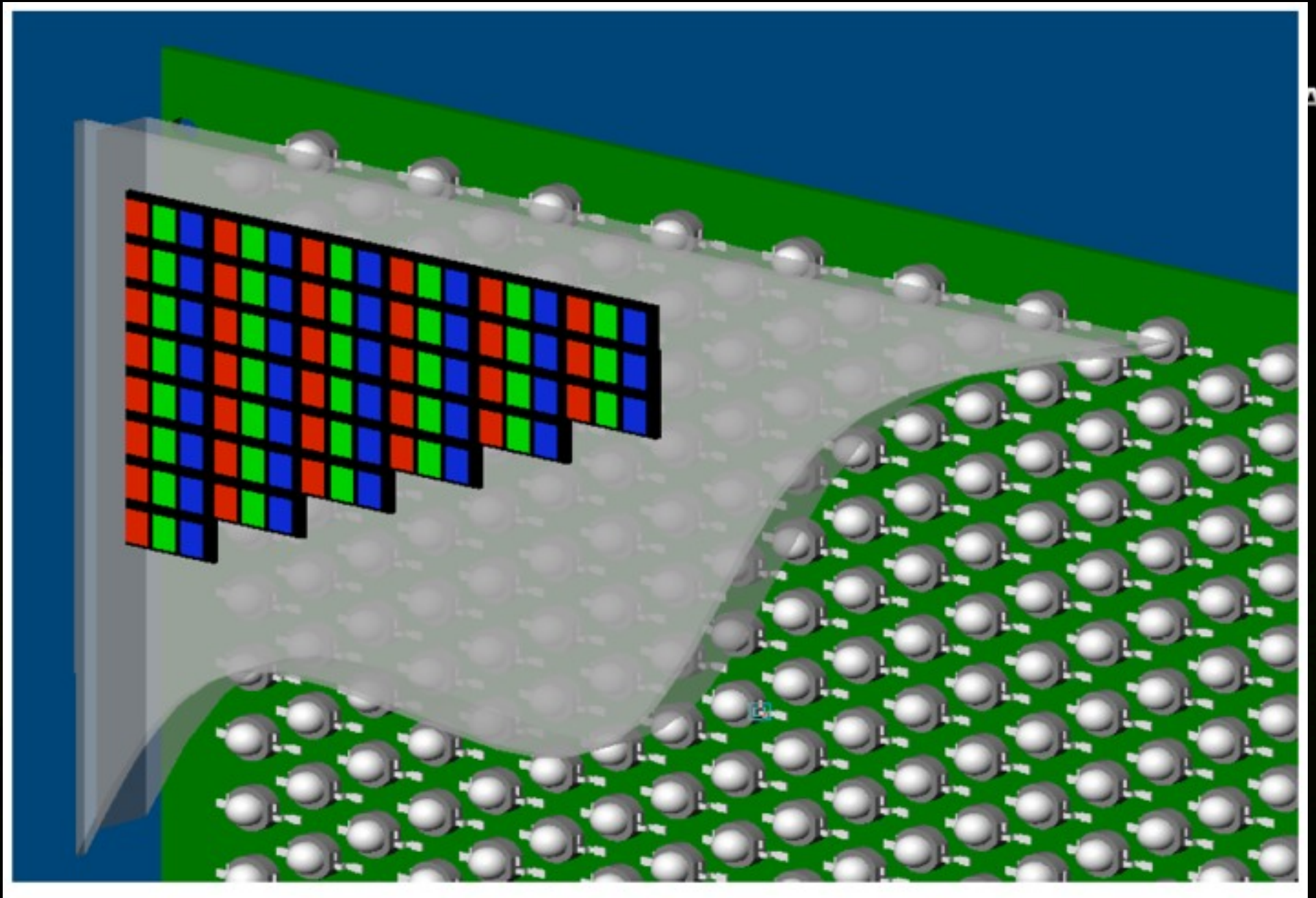
From The Manual of Photography, Jacobson et al.

Sunnybrook HDR display

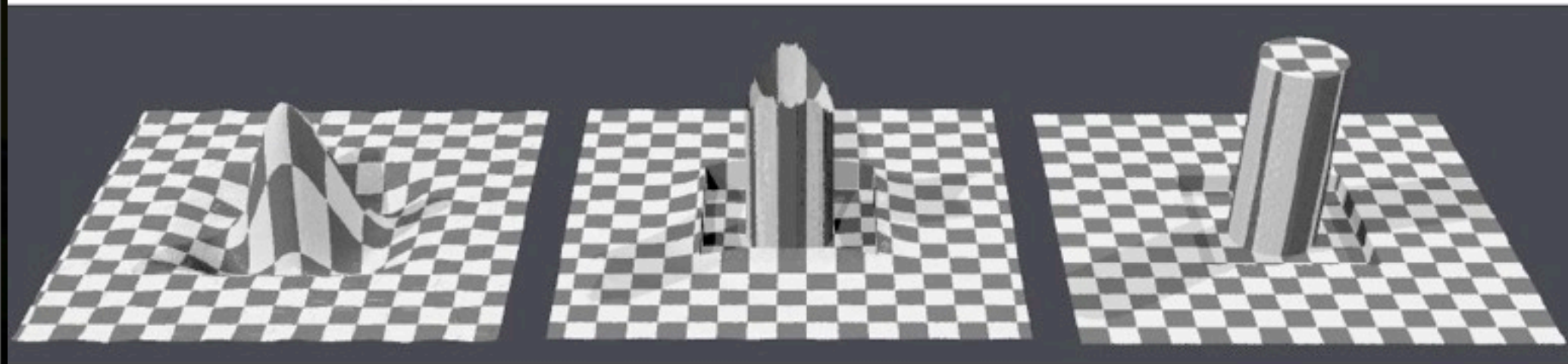
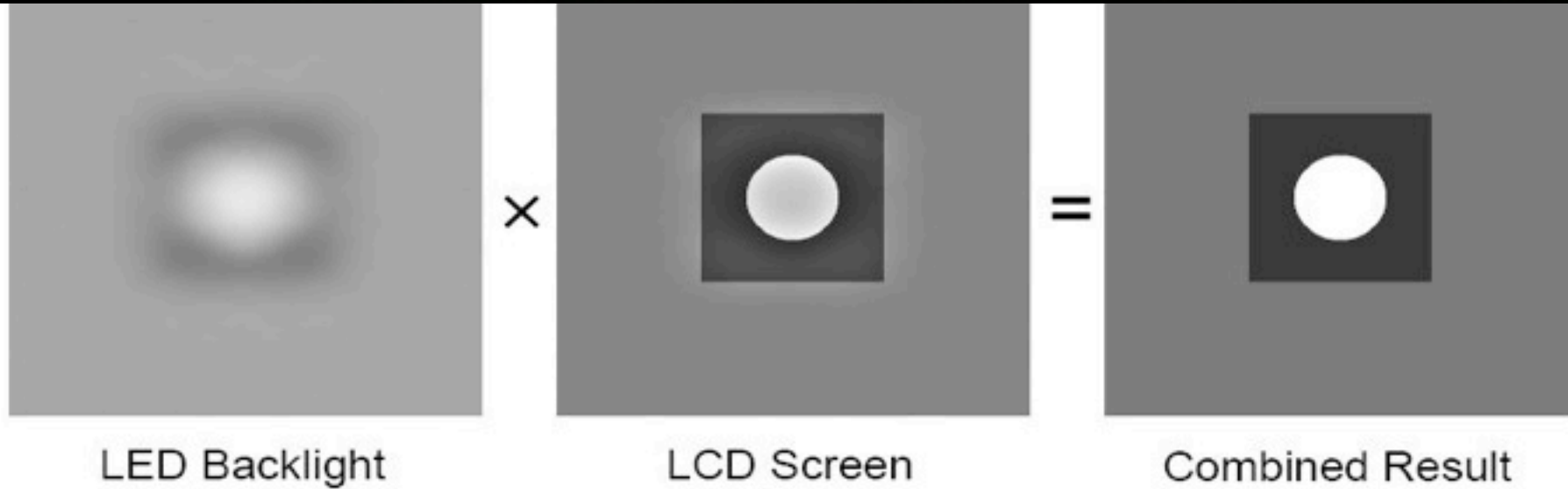


- Bright back light at low resolution
 - modulate with the front panel
 - over 1 : 10,000 doable





How it works



How humans deal with dynamic range



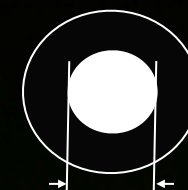
- We're sensitive to contrast (multiplicative)
 - A ratio of 1:2 is perceived as the same contrast as a ratio of 100 to 200
 - Use the log domain as much as possible
- Dynamic adaptation (very local in retina)
 - Pupil (not so important)
 - Neural & chemical
 - can adapt $\sim 10^{10}$
- Transmit the signal to brain
 - only $10^3 - 10^4$
 - spatial contrast-based processing already in the eye

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Dim Light



~ 6 mm

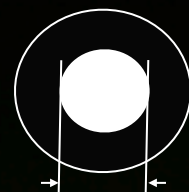
Pupil dilates
More light enters the eye

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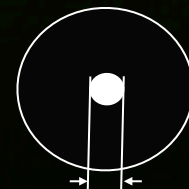


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More light enters the eye

Bright Light



~ 1 mm

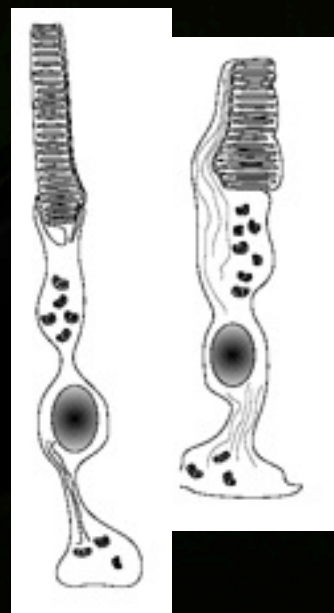
Pupil constricts

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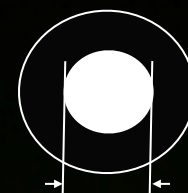
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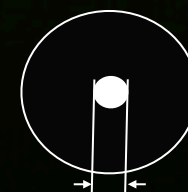
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More light enters the eye

Area ratio

$\sim 16 : 1$

Bright Light



~ 1 mm

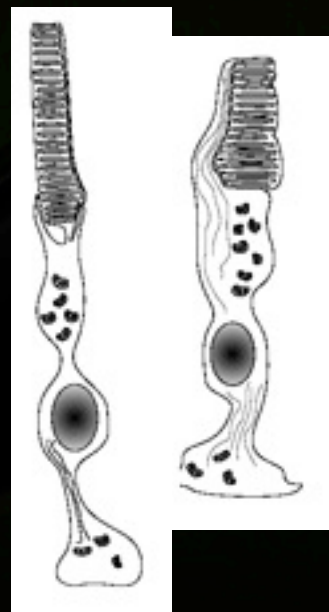
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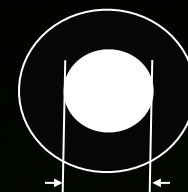
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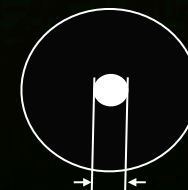
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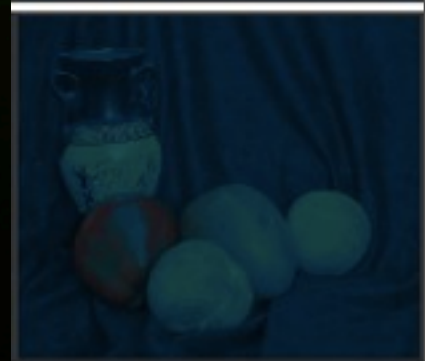
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Mesopic vision

- 3 cones + 1 rod map to 3 signals from eye to visual cortex



Mesopic vision

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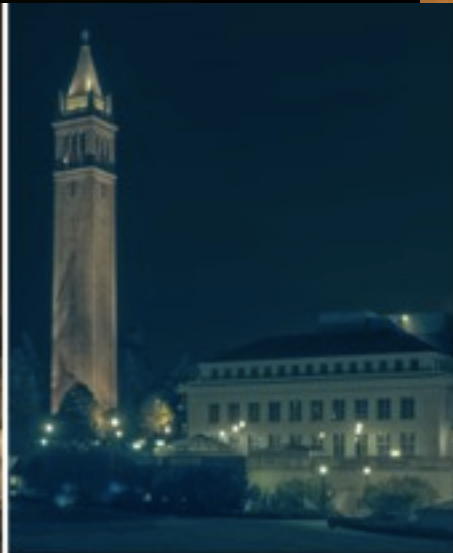
Images copyright Adam Kirk and James O'Brien.

Figure 6: Top: An HDR image of the Fremont Troll with no tone mapping. Bottom: The image has been tone mapped for low-light conditions.



Mesopic vision

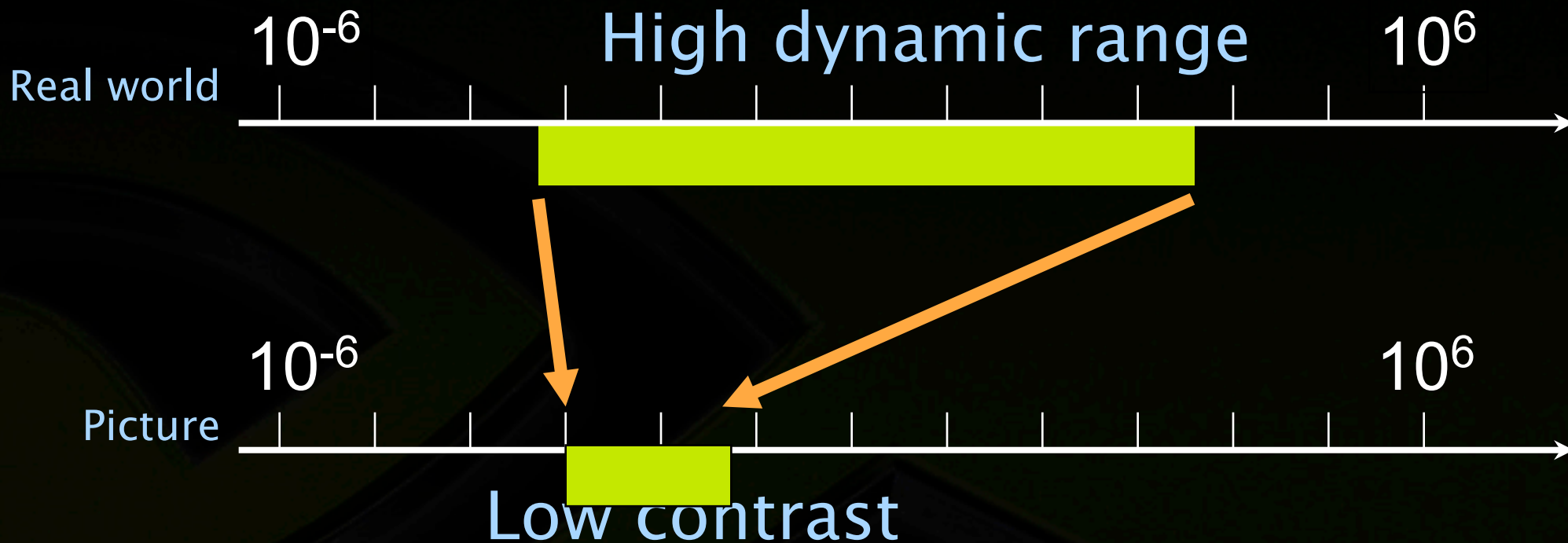
- 3 cones + 1 rod map to 3 signals from eye to visual cortex



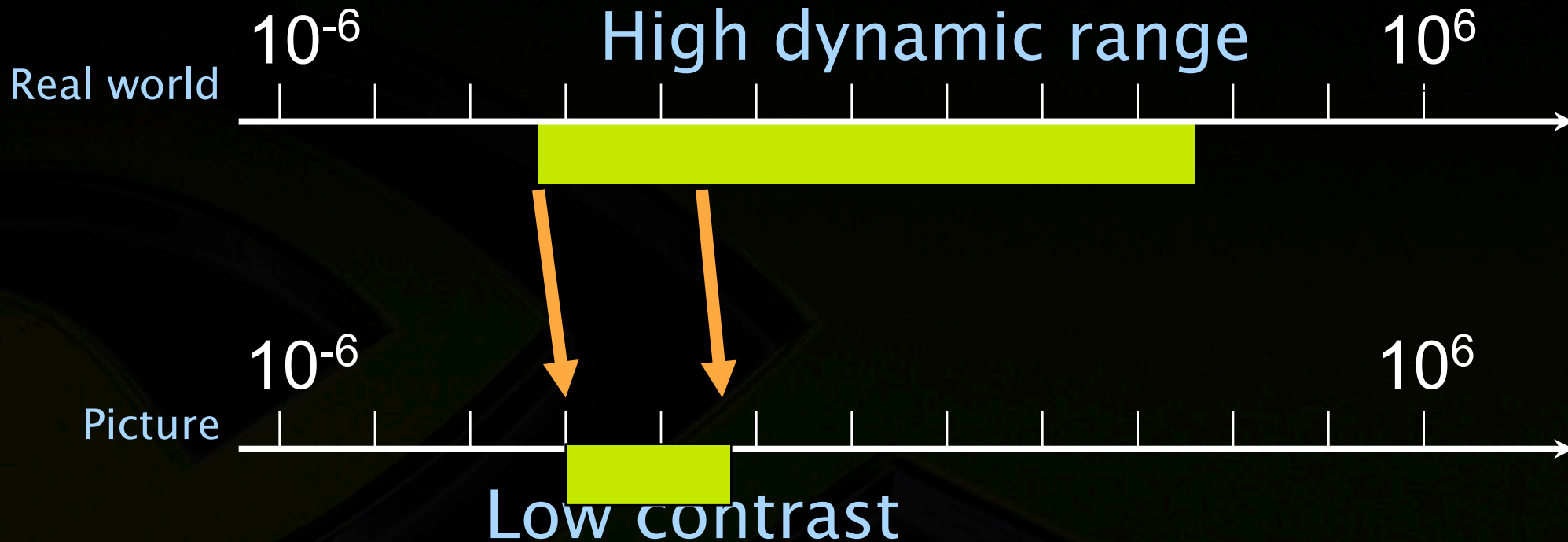
Images copyright Adam Kirk and James O'Brien.
Figure 4: Left: An HDR night scene with no tone mapping featuring UC Berkeley's Sather Tower. Right: The image has been tone mapped for low-light conditions. Artifacts on the clock face occur because the hands moved during image acquisition.

Images copyright Adam Kirk and James O'Brien.
Figure 6: Top: An HDR image of the Fremont Troll with no tone mapping. Bottom: The image has been tone mapped for low-light conditions.

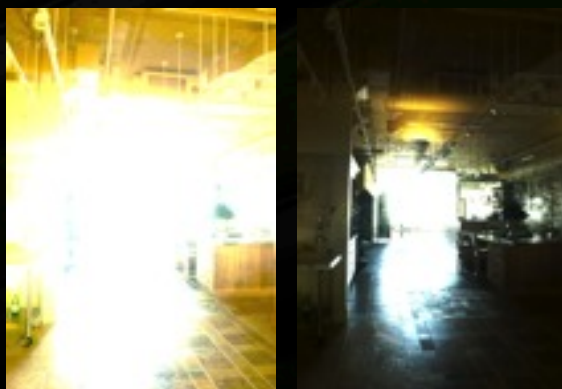
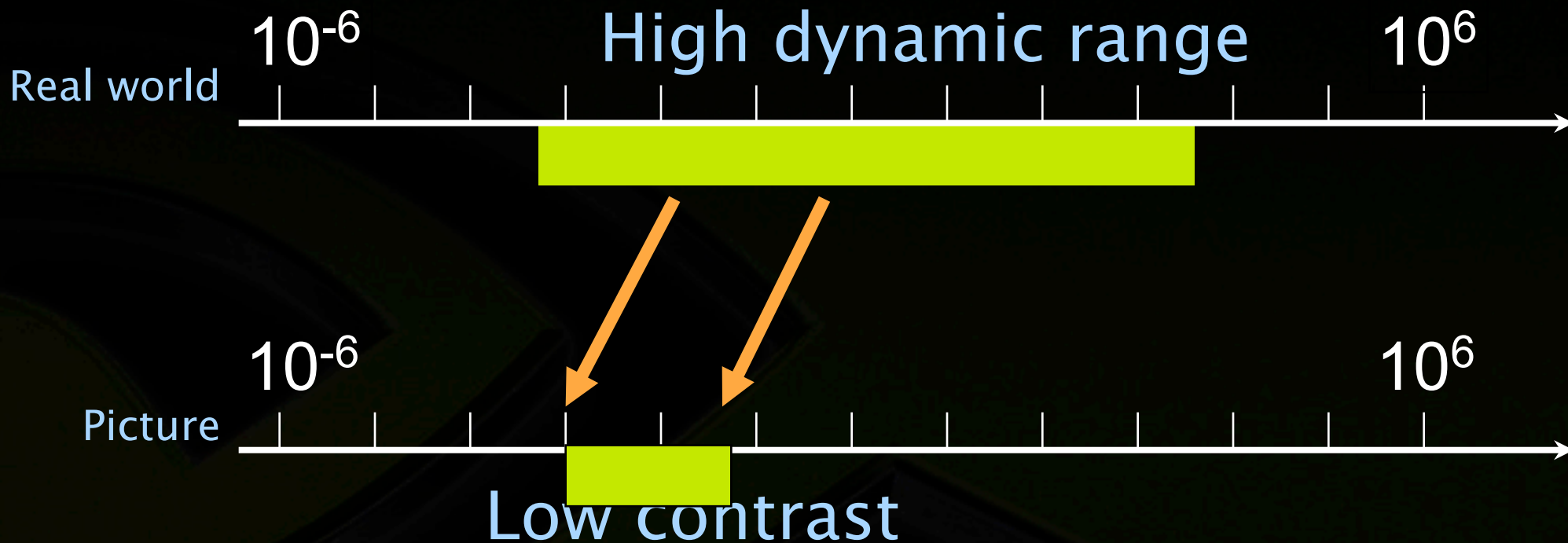
Multiple exposure photography



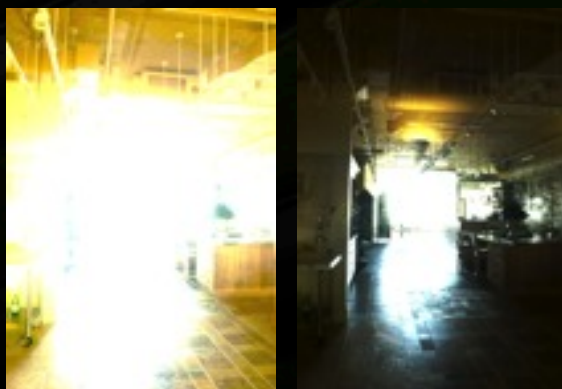
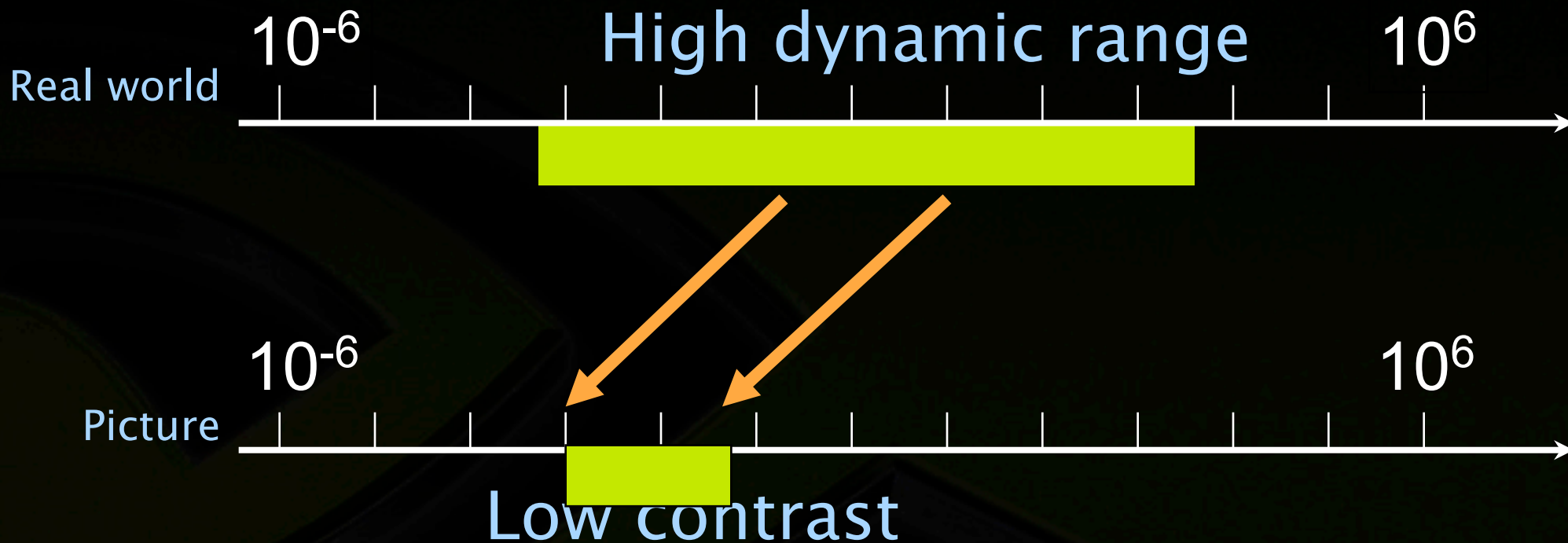
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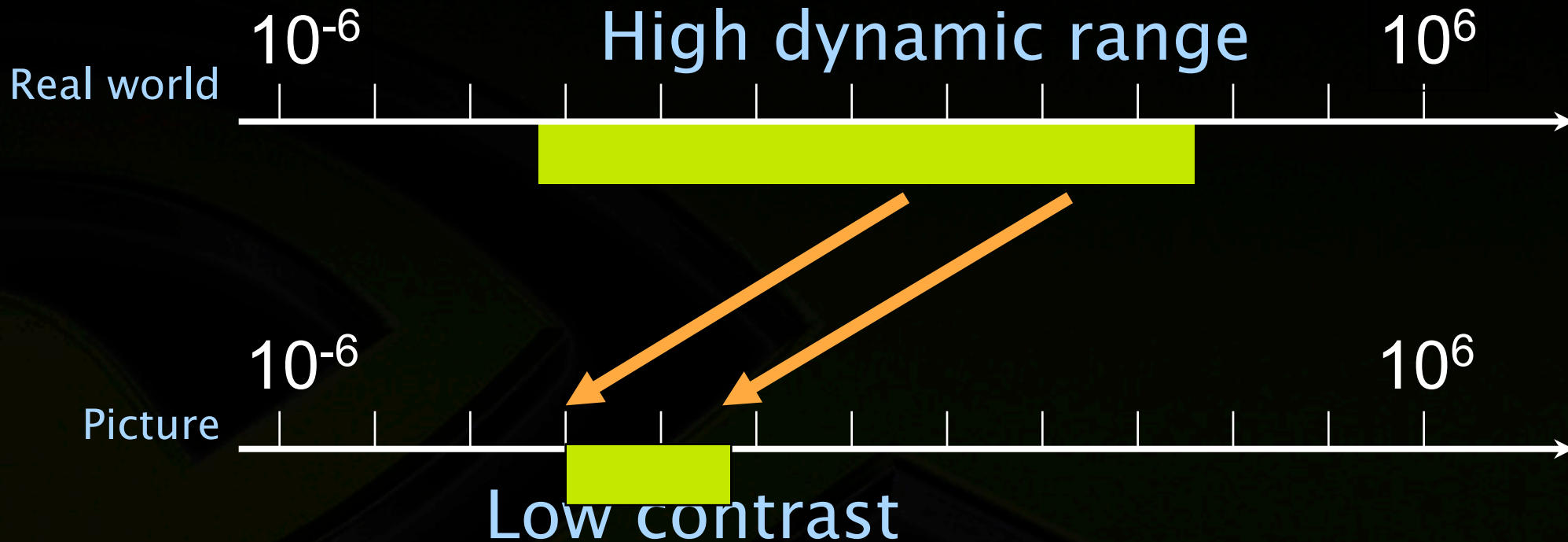
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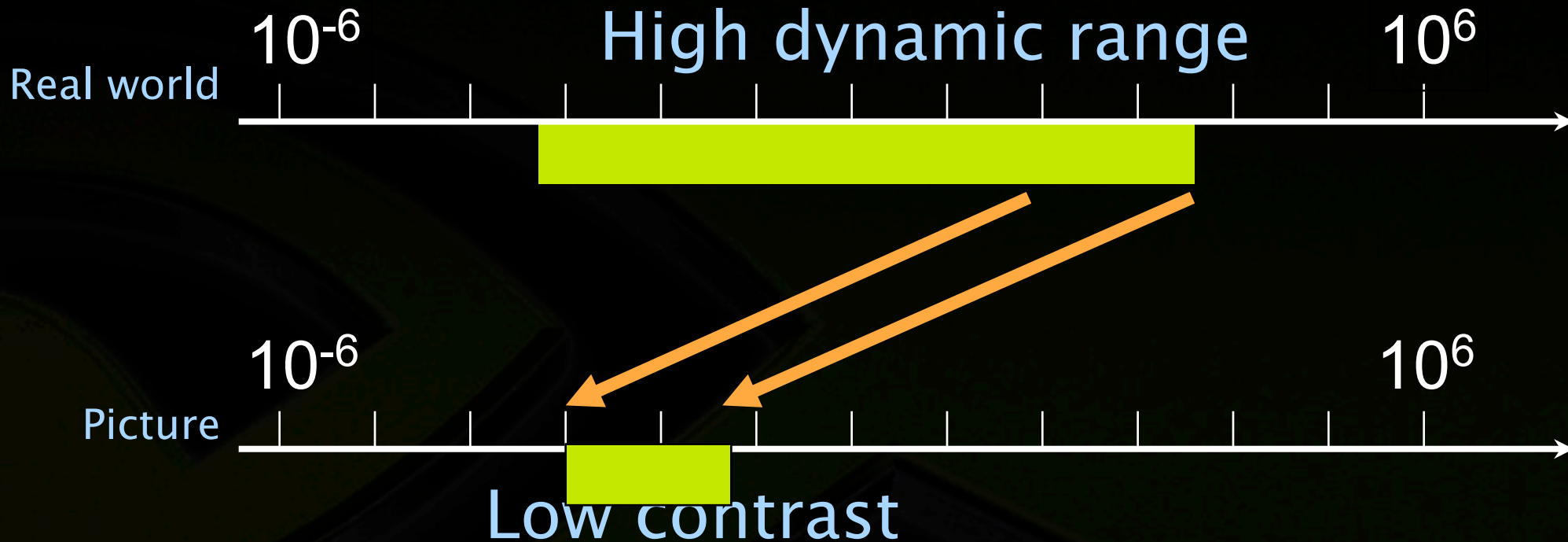
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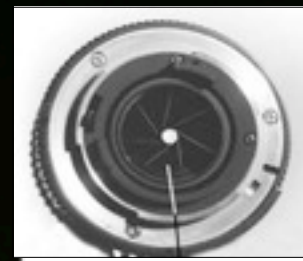


Multiple exposure photography



How can we vary exposure?

- Options:
 - Shutter speed
 - Aperture
 - ISO
 - Neutral density filter



Tradeoffs



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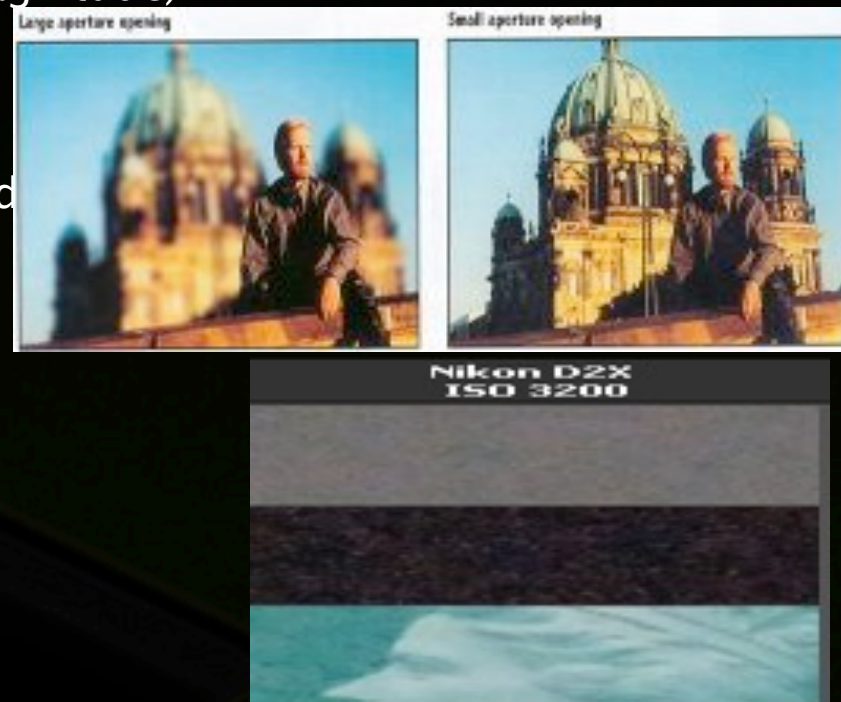
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 - Range: ~30 sec to 1/4000sec (6 orders of magnitude)
 - Pros: reliable, linear
 - Cons: sometimes noise for long exposure



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- Aperture
 - Range: ~f/1.4 to f/22 (2.5 orders of magnitude)
 - Cons: changes depth of field
 - Useful when desperate



Tradeoffs



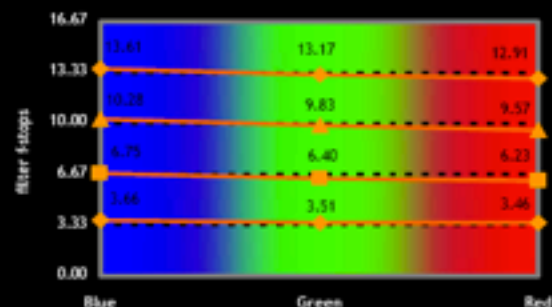
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 - Range: ~f/1.4 to f/22 (2.5 orders of magnitude)
 - Cons: changes depth of field
 - Useful when desperate
- ISO
 - Range: ~100 to 1600 (1.5 orders of magnitude)
 - Cons: noise
 - Useful when desperate
- Neutral density filter
 - Range: up to 4 densities (4 orders of magnitude) & can be stacked
 - Cons: not perfectly neutral (color shift), not very precise, need to touch camera (shake)
 - Pros: works with strobe/flash, good complement when desperate



Early HDR photos: Gustave Le Gray (~1850)



- Take two shots
 - one for the sky – direct light
 - one for the rest – reflected light
 - cut and paste the negatives, and develop



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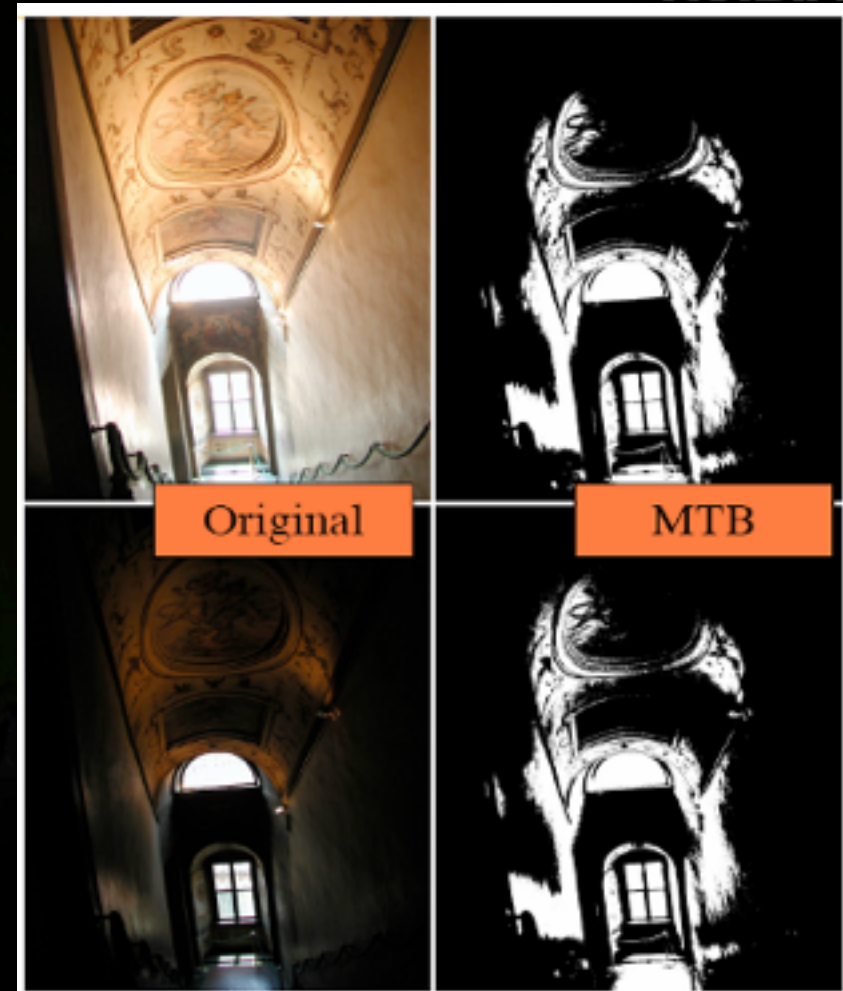


HP Robinson (1858) Fading Away, 5 negatives



Robust image registration of different exposures

- Exposure stack capture
 - camera may move
 - details look different
- Median-Threshold Bitmap (MTB)
 - use a black and white version of the image thresholded at the median
- Find the translation that minimizes difference
- Accelerate using pyramid





SIGGRAPH2005

Alignment Results



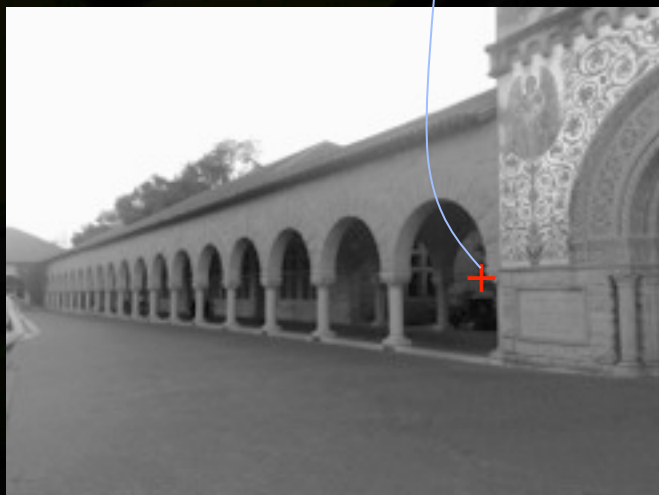
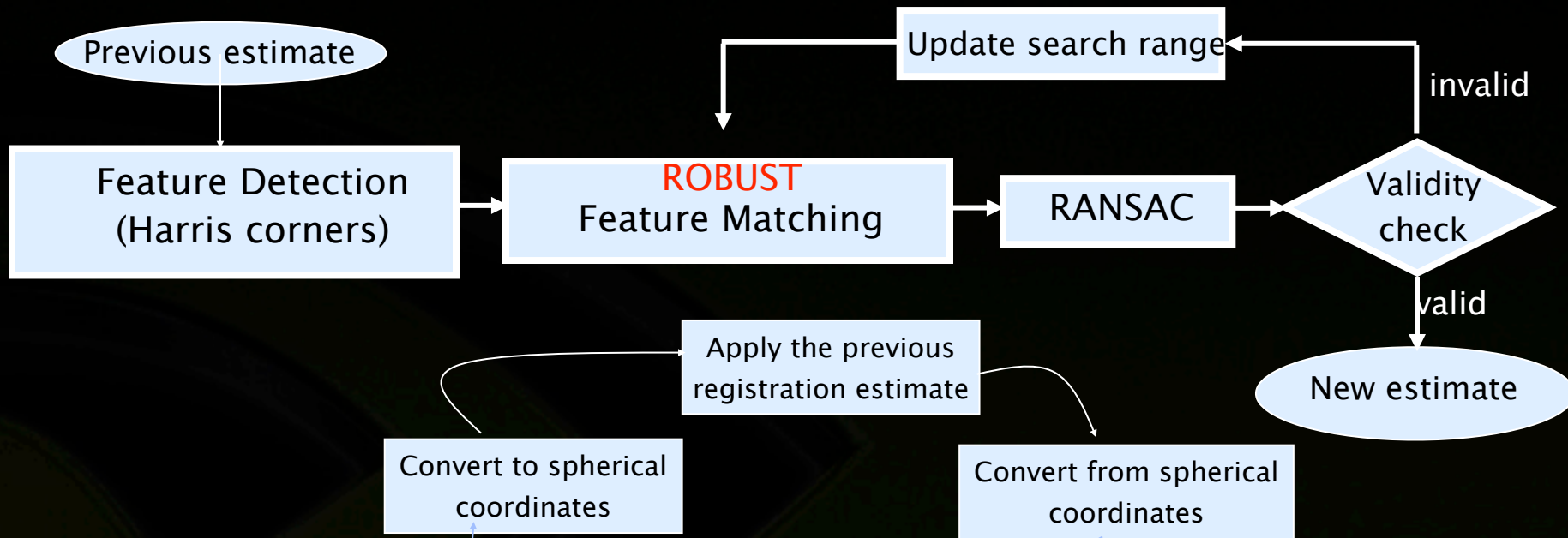
5 unaligned exposures

Close-up detail

MTB alignment

Time: About .2 second/exposure for 3 MPixel image

Feature-based registration



Best **normalized cross-correlation** match

Don't forget lens distortion!



- Modern lenses are pretty good
 - often the distortion is quite small
- Homography only works for planar scenes
 - with rotations around entrance pupil homography is OK
- But if you want to align and accumulate images
 - you should account for all sources of warpage



Monday, March 5, 12



Monday, March 5, 12



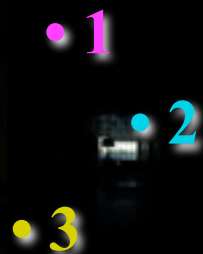
Monday, March 5, 12

Response curve calibration

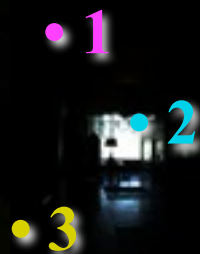


- Debevec and Malik (1997)
 - select a small number of pixels from the images
 - using all pixels would make the matrix too large
 - optimize the response curve with a smoothness constraint
- Robertson et al. (1999)
 - optimize over all pixels in all images
 - no need to solve a linear equation with a large matrix
 - iterate
 - calculate HDR image using the response curve
 - find a better response curve using HDR image
 - refinement (2003): better weights

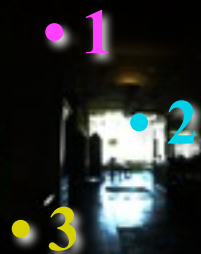
Calibration [Debevec & Malik 97]



$\Delta t =$
1/64 sec



$\Delta t =$
1/16 sec



$\Delta t =$
1/4 sec

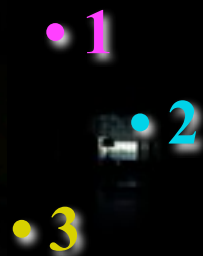


$\Delta t =$
1 sec

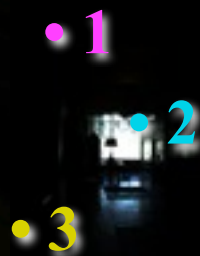


$\Delta t =$
4 sec

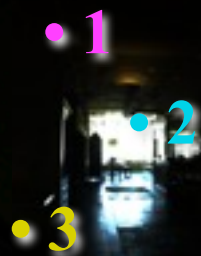
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1/16 sec



$\Delta t =$
1/4 sec



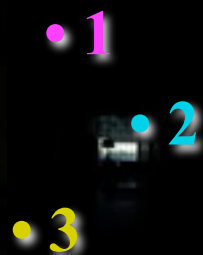
$\Delta t =$
1 sec



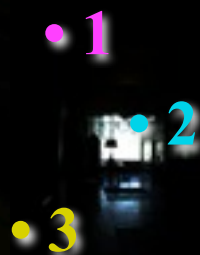
$\Delta t =$
4 sec

$$\text{Pixel Value } Z = f(\text{Exposure})$$

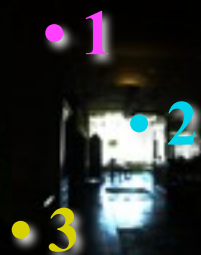
Calibration [Debevec & Malik 97]



$\Delta t =$
1/64 sec



$\Delta t =$
1/16 sec



$\Delta t =$
1/4 sec



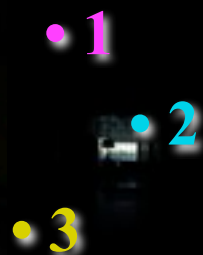
$\Delta t =$
1 sec



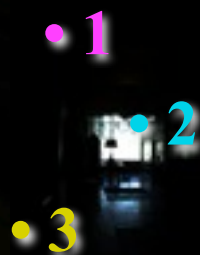
$\Delta t =$
4 sec

$$\text{Pixel Value } Z = f(\text{Exposure})$$
$$\text{Exposure} = \text{Radiance} * \Delta t$$

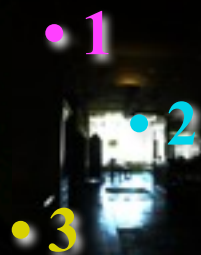
Calibration [Debevec & Malik 97]



$\Delta t =$
1/64 sec



$\Delta t =$
1/16 sec



$\Delta t =$
1/4 sec



$\Delta t =$
1 sec



$\Delta t =$
4 sec

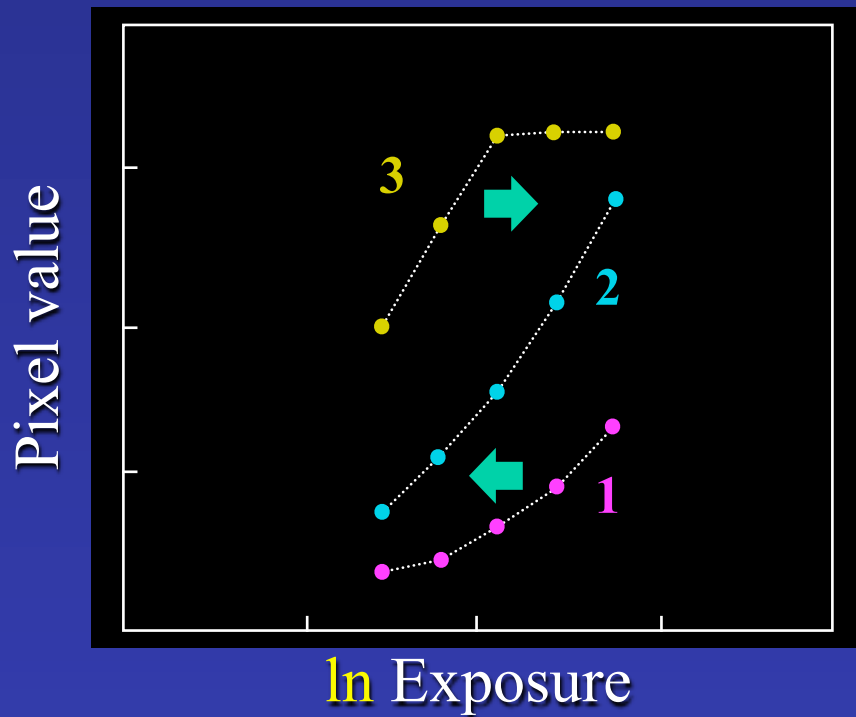
Pixel Value $Z = f(\text{Exposure})$

$\text{Exposure} = \text{Radiance} * \Delta t$

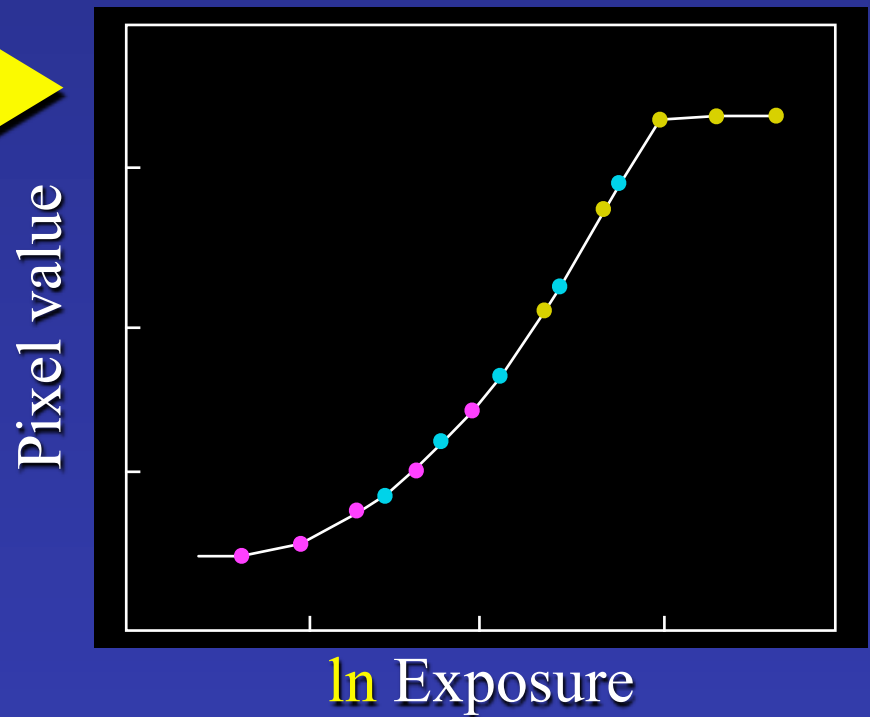
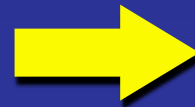
$\ln \text{Exposure} = \ln \text{Radiance} + \ln \Delta t$

Adjust exposure to find a smooth response curve

Assuming the same exposure for each pixel



After adjusting radiances to obtain a smooth response curve



The Math

- Let f be the response function: $Z_{ij} = f(E_i \Delta t_j)$
- Let g be the logarithm of the inverse response function: $g(Z_{ij}) = \ln f^{-1}(Z_{ij}) = \ln E_i + \ln \Delta t_j$
- Solve the overdetermined linear system:
 - unknown $E_i, g(\cdot)$

$$\sum_{i=1}^N \sum_{j=1}^P \left[\ln E_i + \ln \Delta t_j - g(Z_{ij}) \right] + \lambda \sum_{z=Z_{\min}}^{Z_{\max}} g''(z)^2$$

fitting term

smoothness term

Matlab code

```
%
% gsolve.m - Solve for imaging system response function
%
% Given a set of pixel values observed for several pixels
% in several images with different exposure times, this
% function returns the imaging system's response function
% g as well as the log film irradiance values for the
% observed pixels.
%
% Assumes:
%
% Zmin = 0
% Zmax = 255
%
% Arguments:
%
% Z(i,j) is the pixel values of pixel location number I
% in image j
% B(j) is the log delta t, or log shutter speed, for
% image j l is lamdba, the constant that
% determines the amount of smoothness
% w(z) is the weighting function value for pixel value z
%
% Returns:
%
% g(z) is the log exposure corresponding to pixel value z

function [g,lE] = gsolve(Z,B,l,w)
n = 256;
A = zeros(size(Z,1)*size(Z,2)+n+1,n+size(Z,1));
b = zeros(size(A,1),1);

%% Include the data-fitting equations
k = 1;
for i=1:size(Z,1)
    for j=1:size(Z,2)
        wij = w(Z(i,j)+1);
        A(k,Z(i,j)+1) = wij;
        A(k,n+i) = -wij;
        b(k,1) = wij * B(i,j);
        k = k+1;
    end
end

%% Fix the curve by setting its middle value to 0
A(k,129) = 1;
k = k+1;

%% Include the smoothness equations
for i=1:n-2
    A(k,i) = l*w(i+1);
    A(k,i+1) = -2*l*w(i+1);
    A(k,i+2) = l*w(i+1);
    k=k+1;
end

%% Solve the system using SVD
x = A\b;
g = x(1:n);
lE = x(n+1:size(x,1));
```

Results: Digital Camera

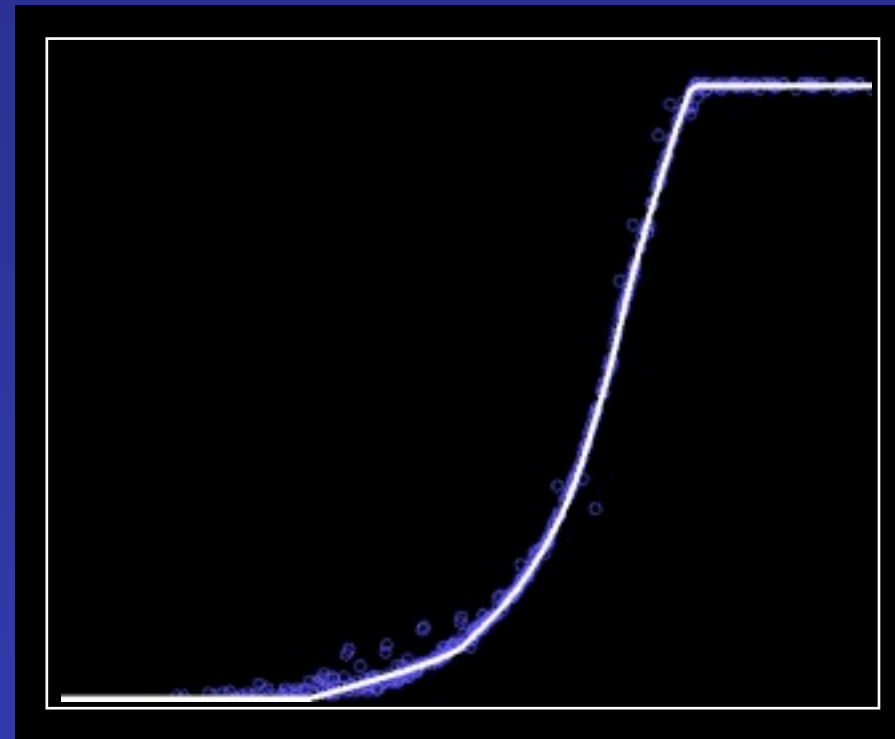
Kodak DCS460

1/30 to 30 sec

Recovered response curve



Pixel value



log Exposure

Reconstructed radiance map



HDR representations



HDR representations



- Portable Float Map, Floating point TIFF
 - RGB as three floats
 - $3 * 32b = 96b$

HDR representations

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 - RGB as three floats
 - $3 * 32b = 96b$
- Radiance
 - 8bit R, G, and B; 8bit shared exponent
 - $(R,G,B) / 255 * 2^{(E - 128)}$
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- OpenEXR
 - RGB each a half float, matches gfx HW
 - $3 * 16b = 48b$
- Many others that are better at compressing
 - but require conversions / LUT to operate on

Calibration [Robertson et al. 99]



- Input

- series of i images with pixel values $m = y_{ij}$ due to exposure $l_{ij} = \text{irradiance } x_j * \text{time } t_i$ $y_{ij} = f(l_{ij}) = f(x_j t_i)$
- weighting function $w_{ij} = w_{ij}(y_{ij})$
 - assume central values most sensitive

- Task:

- find pixel irradiance (luminance)
- recover response curve
- use EM (expectation maximization)

Calibration [Robertson et al. 99]

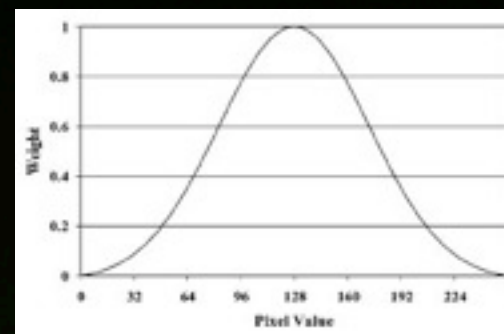


Input

- series of i images with pixel values $m = y_{ij}$ due to exposure $l_{ij} = \text{irradiance } x_j * \text{time } t_i$

$$y_{ij} = f(l_{ij}) = f(x_j t_i)$$

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$$x_j = f^{-1}(y_{ij}) / t_i$$

Calibration [Robertson et al. 99]

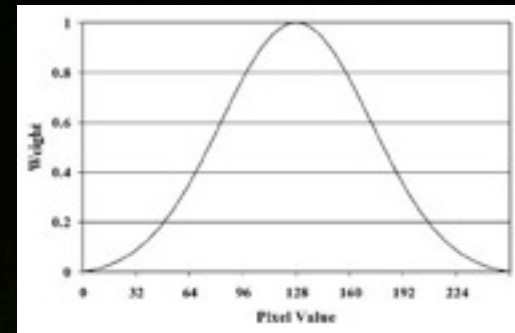


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$$y_{ij} = f(l_{ij}) = f(x_j t_i)$$

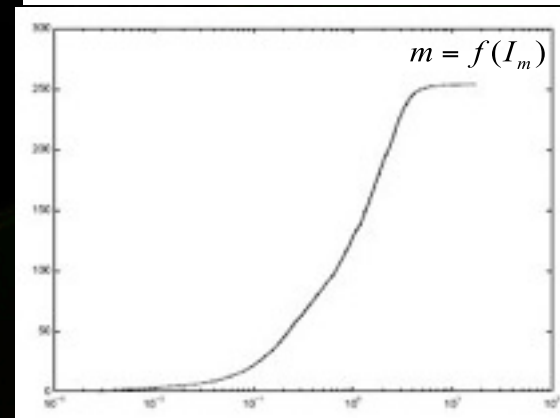
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Task:

- find pixel irradiance (luminance)
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$$x_j = f^{-1}(y_{ij}) / t_i$$



Calculate irradiances x_j given the images



- First iteration: unknown response curve
 - assume linear response with: $I_{128} = 1.0$
- Solve for irradiance at pixel j given
 - exposure times t_i
 - pixel values $m = I_{y_{ij}}$
 - weights w_m

$$\min_{x_j} \sum_{i,j} w_{ij} (I_{y_{ij}} - t_i x_j)^2 \Rightarrow x_j = \frac{\sum_i w_{ij} t_i I_{y_{ij}}}{\sum_i w_{ij} t_i^2}$$

Estimate response function for each m



$$I_m = \frac{1}{\text{Card}(y_{ij} = m)}$$

Estimate response function for each m



- For all pixel values $m = 0, \dots, 254$
 - ignore 255, that's saturated
 - collect every pixel with value m
 - calculate average I_m
 - note: weight is constant for a given m

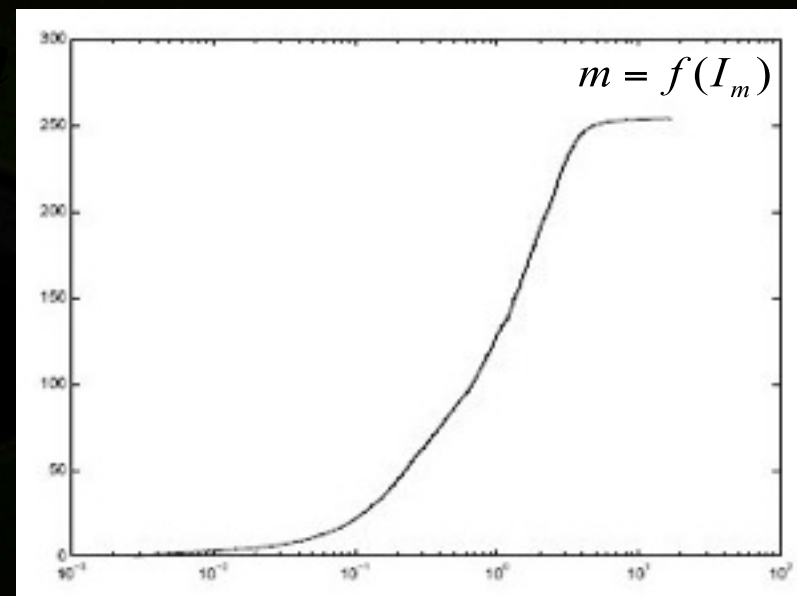
$$I_m = \frac{1}{\text{Card}(y_{ij} = m)}$$

Estimate response function for each m



- For all pixel values $m = 0, \dots, 254$
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$$I_m = \frac{1}{\text{Card}(y_{ij} = m)}$$



Estimate response function for each m

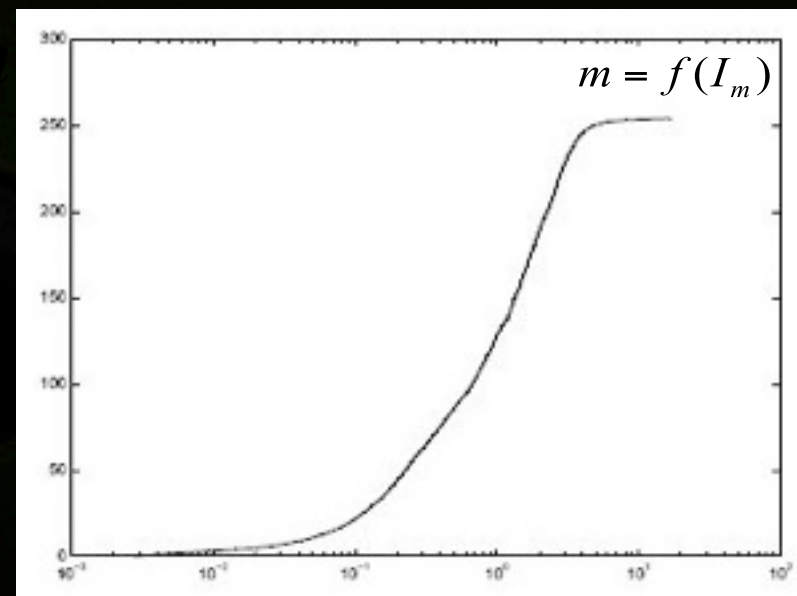


- For all pixel values $m = 0, \dots, 254$
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$$I_m = \frac{1}{\text{Card}(y_{ij} = m)}$$

• Normalize

- set $I_{128} = 1.0$



Estimate response function for each m



- For all pixel values $m = 0, \dots, 254$
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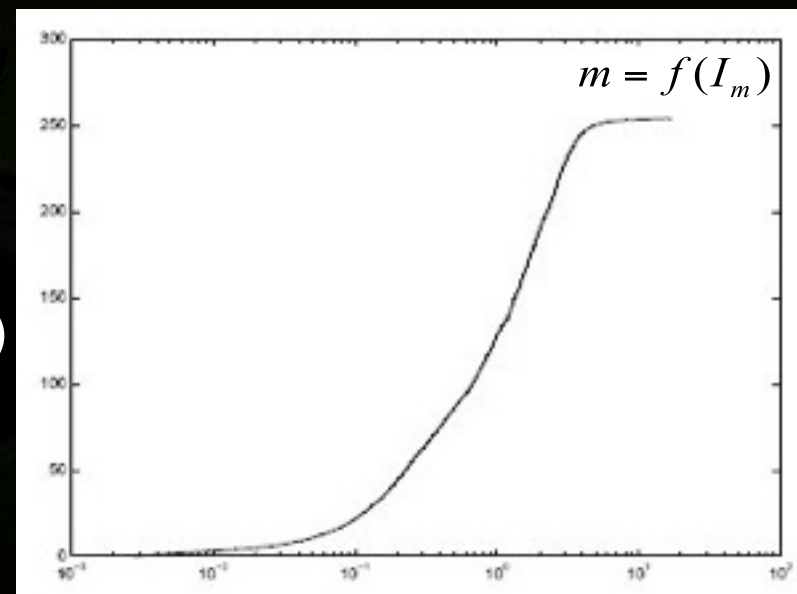
$$I_m = \frac{1}{\text{Card}(y_{ij} = m)}$$

● Normalize

- set $I_{128} = 1.0$

● Repeat (Expectation Maximization)

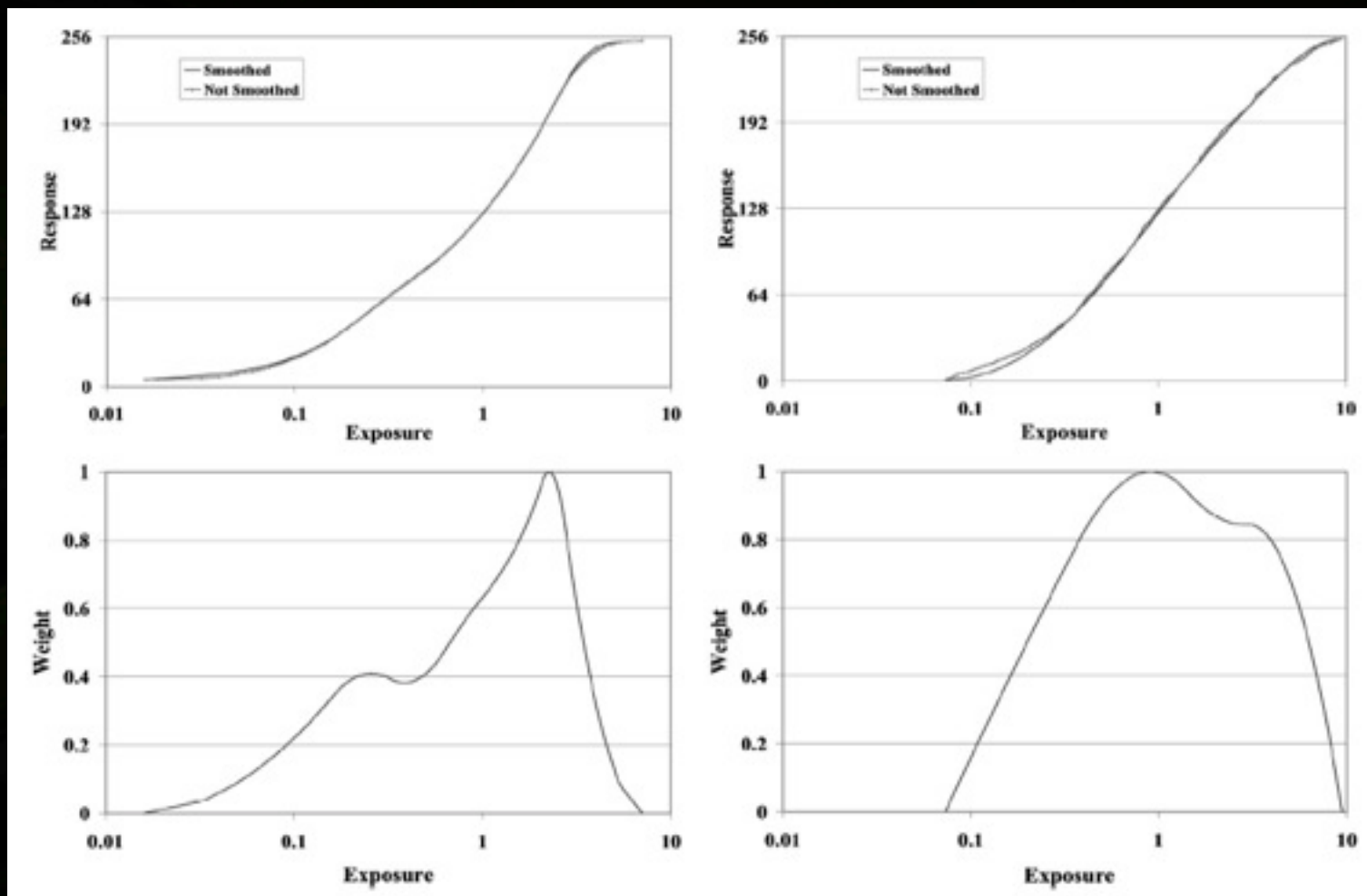
- estimate irradiances
- estimate response function



Calibration [Robertson et al. 03]

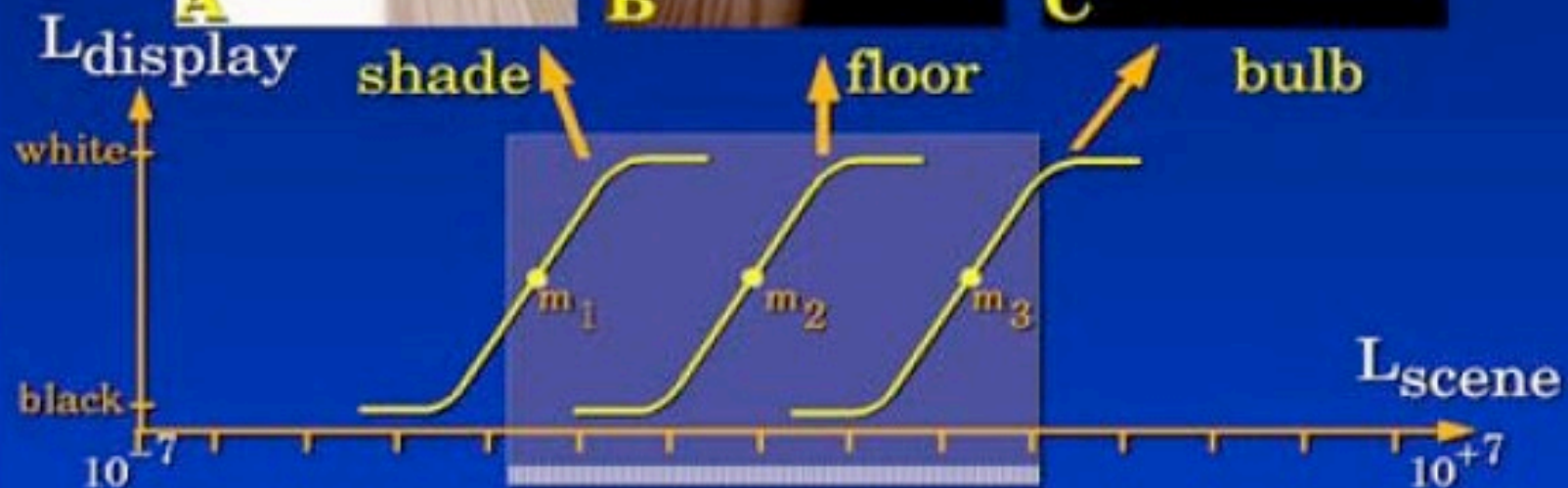


- Estimate also the weight
 - high derivative \rightarrow high weight



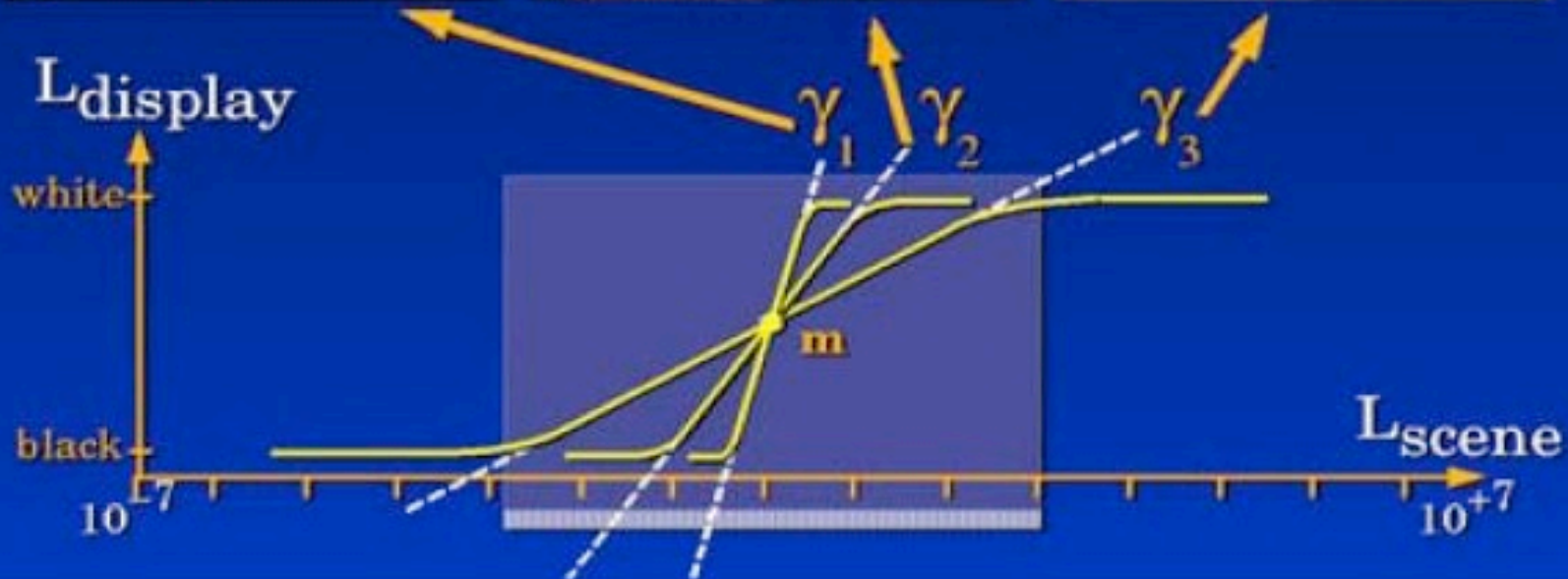
Tone mapping is not easy

Backgnd: Global Scale: m



Tone mapping is not easy

Backgnd: Global Contrast : γ



Tone mapping is not new



- Painters needed to deal with HDR forever
 - dynamic range of the world is much higher than that of paints
 - change the contrasts to give an effect
- Photographers have done it for a long time
 - dynamic range of the film is much higher than that of paper
 - developing prints required manual tone mapping



Early painters couldn't handle HDR

- Go for local contrast, sacrifice global contrast



Go for global contrast



- Local contrast suffers
 - a flat painting



NVIDIA Research

Simone Martini, c. 1328

Leonardo masters contrast: HDR painter



Leonardo invents Chiaroscuro



NVIDIA Research

Madonna by Giotto

Madonna by Leonardo

Leonardo invents Chiaroscuro



NVIDIA Research

Madonna by Giotto

Madonna by Leonardo

Caravaggio



Ansel Adams



Ansel Adams

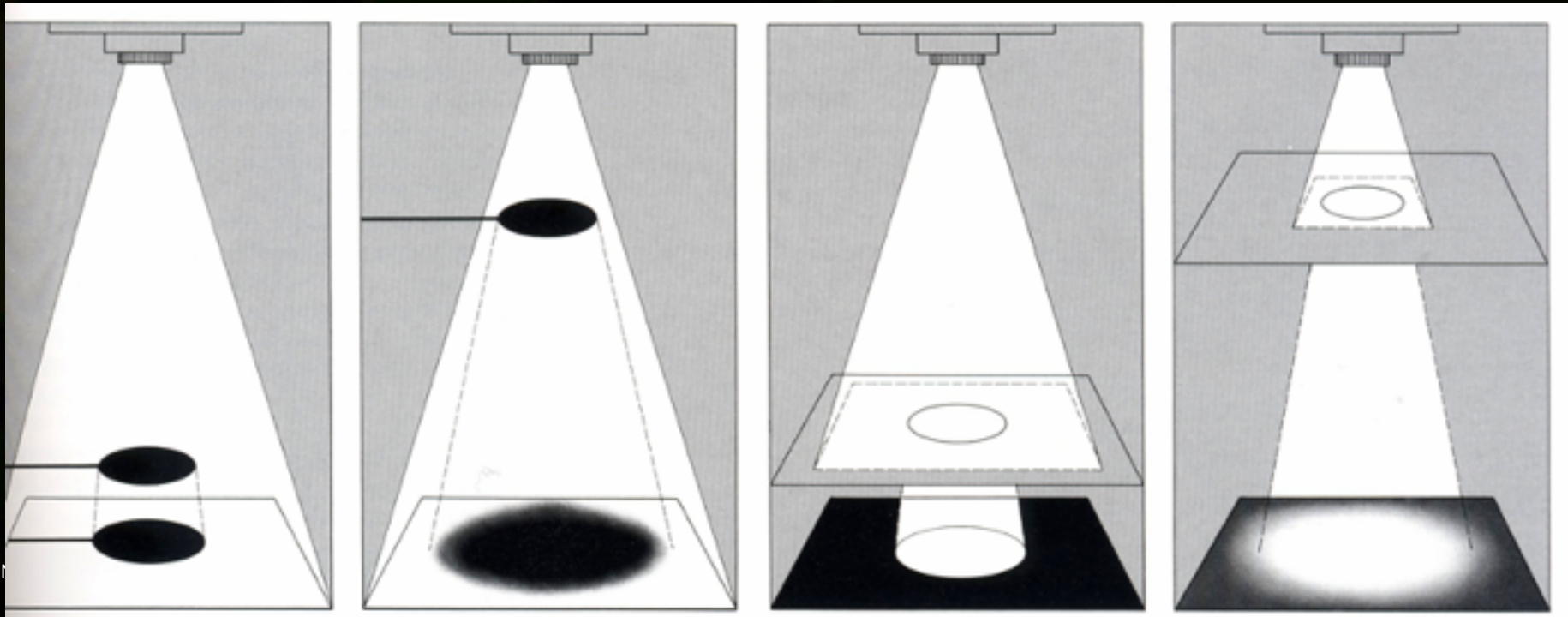


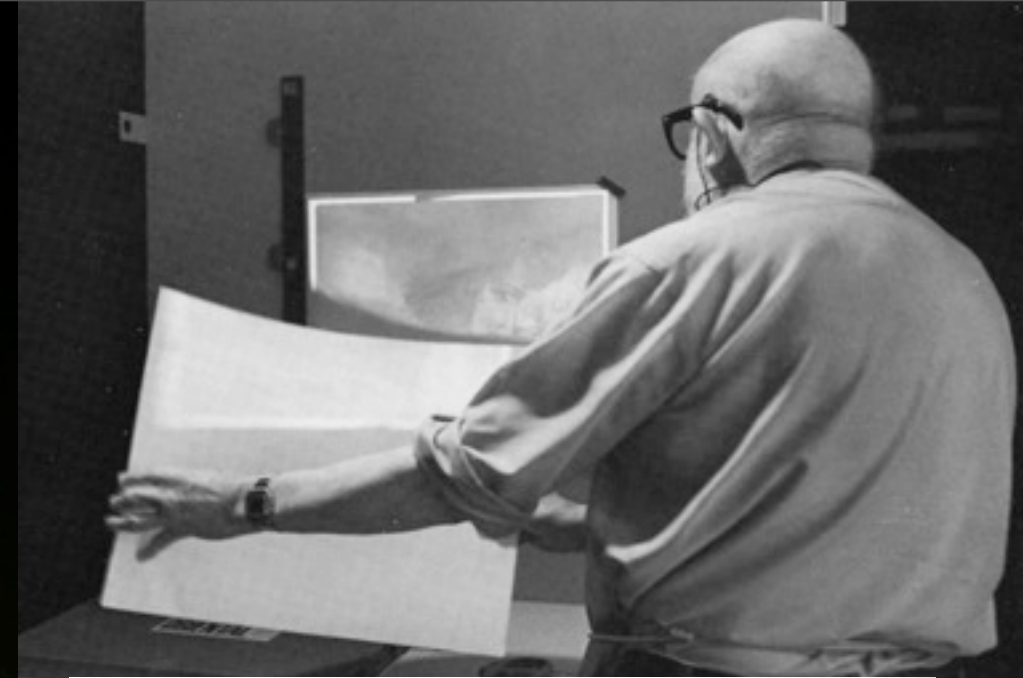
- Design and plan the photo while you are taking it
 - know the medium: both the film, development, and paper
 - standard film & development for the masses using Kodak Brownie
 - global tone map curve, OK on the average
 - virtuosos like Adams
 - capture full dynamic range on the film
 - add spatially varying contrast during development



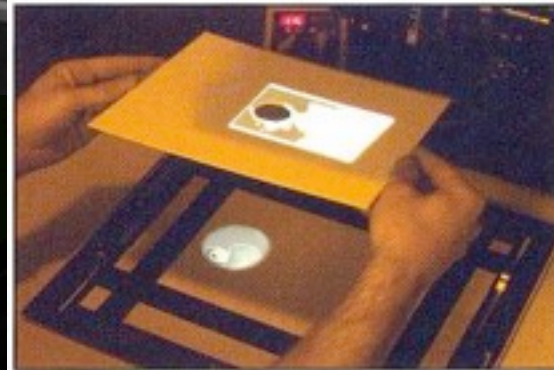
Dodging and burning

- Hide a part of the print during exposure
 - dodge → keep the bright color of the paper
- Let more light be exposed to a region
 - burn → creates a darker print
- Smooth circular motions & blurry mask avoid artifacts





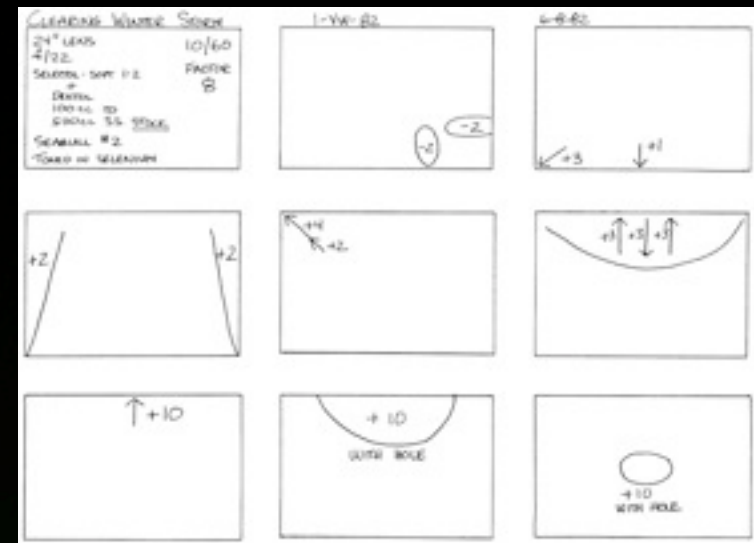
Dodging holds back light during the basic printing exposure to lighten an area.



Burning adds light after the basic exposure to darken an area.



Manual instructions –
repeat for each print



Straight print



After dodging & burning



Contrast reduction in the digital world

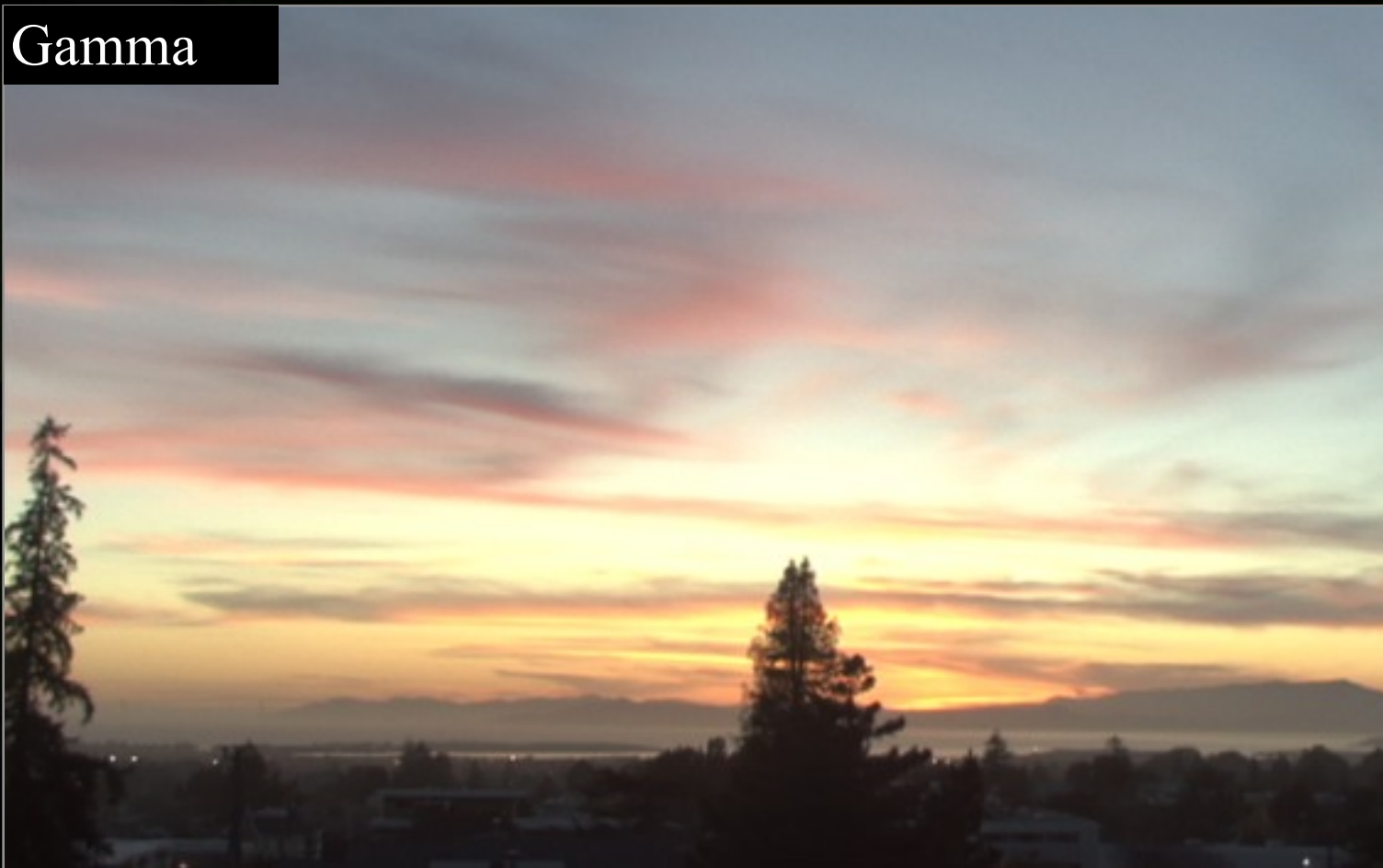
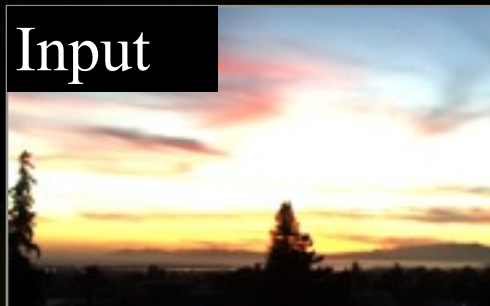


- Scene has 1:10,000 contrast, display has 1:100
- Simplest contrast reduction?



Naïve: Gamma compression

- $X \rightarrow X^\gamma$ (where $\gamma=0.5$ in our case)
- But... colors are washed-out

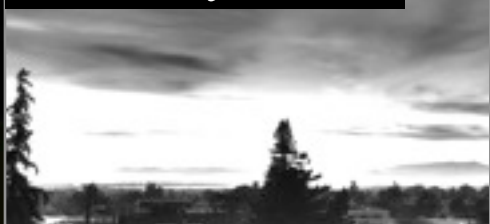


Gamma compression on intensity

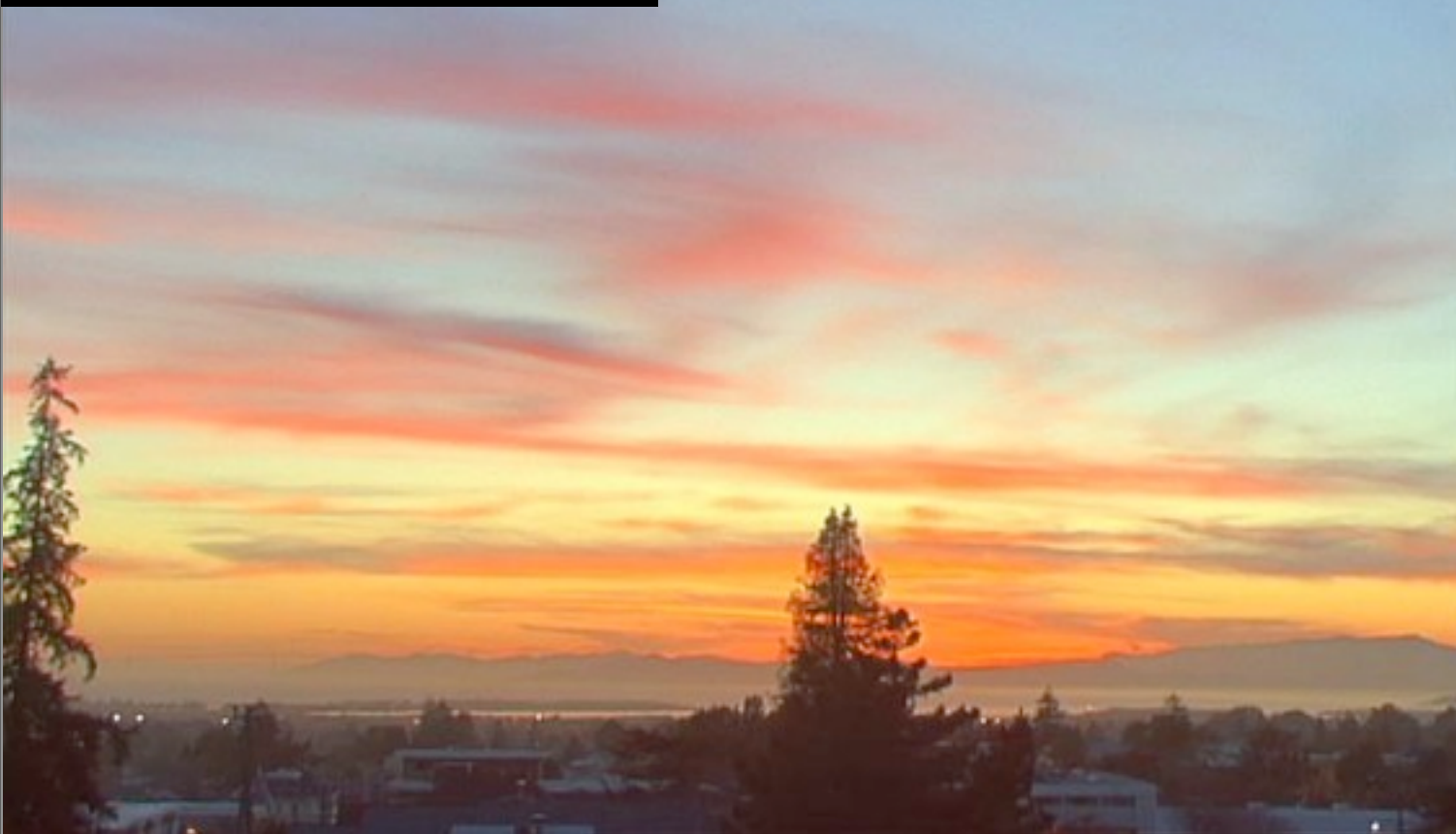


- Colors are OK,
but details (intensity high-frequency) are blurred

Intensity



Gamma on intensity



Color



Let highlights saturate

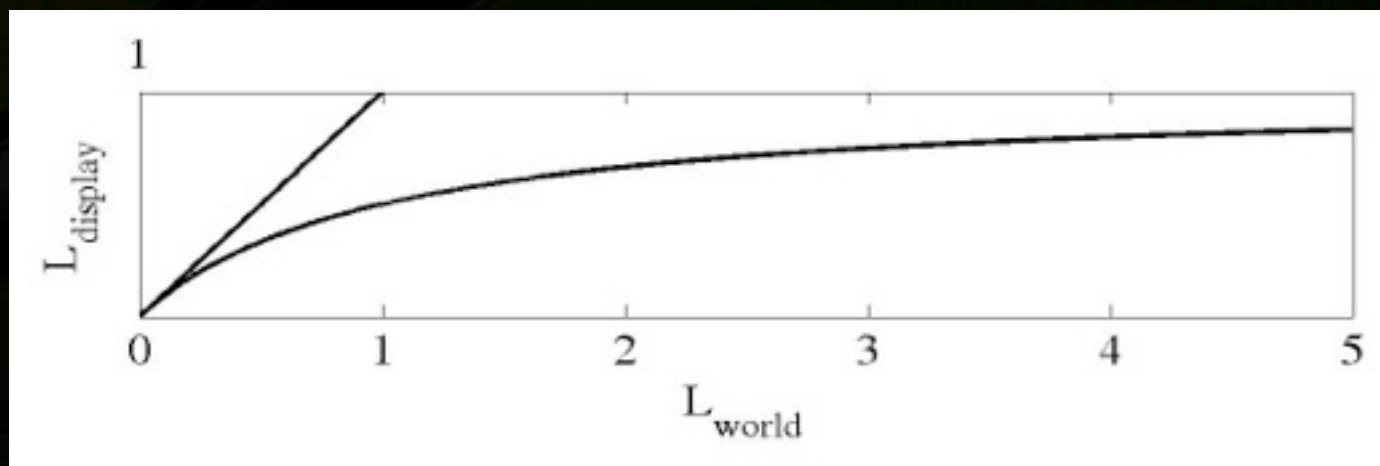
- Darkest 0.1% scaled to display device



Simple global operator (Reinhard et al.)



<http://www.cs.utah.edu/~reinhard/cdrom/tonemap.pdf>

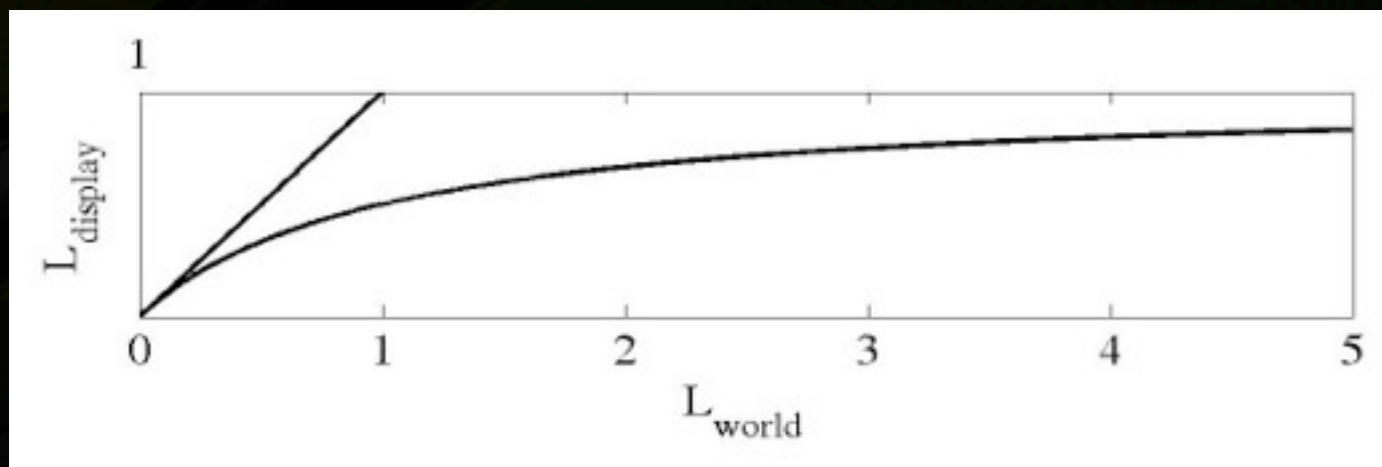


Simple global operator (Reinhard et al.)



- Compression curve needs to
 - bring everything within range
 - leave dark areas alone

<http://www.cs.utah.edu/~reinhard/cdrontonemap.pdf>

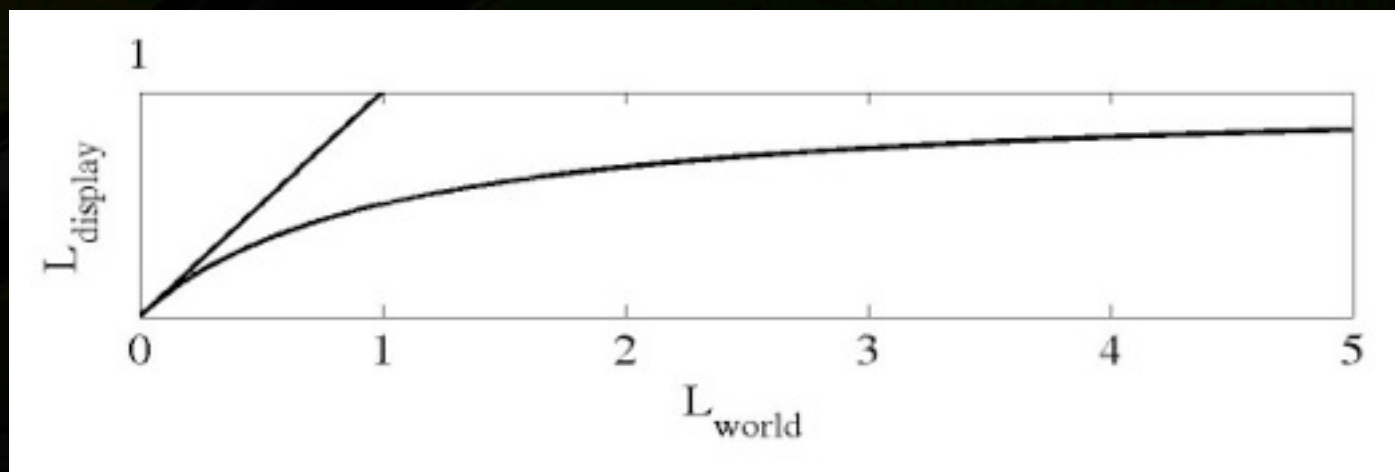


Simple global operator (Reinhard et al.)



- Compression curve needs to
 - bring everything within range
 - leave dark areas alone
- In other words
 - asymptote is 1
 - derivative at 0 is 1

<http://www.cs.utah.edu/~reinhard/cdrone/tonemap.pdf>



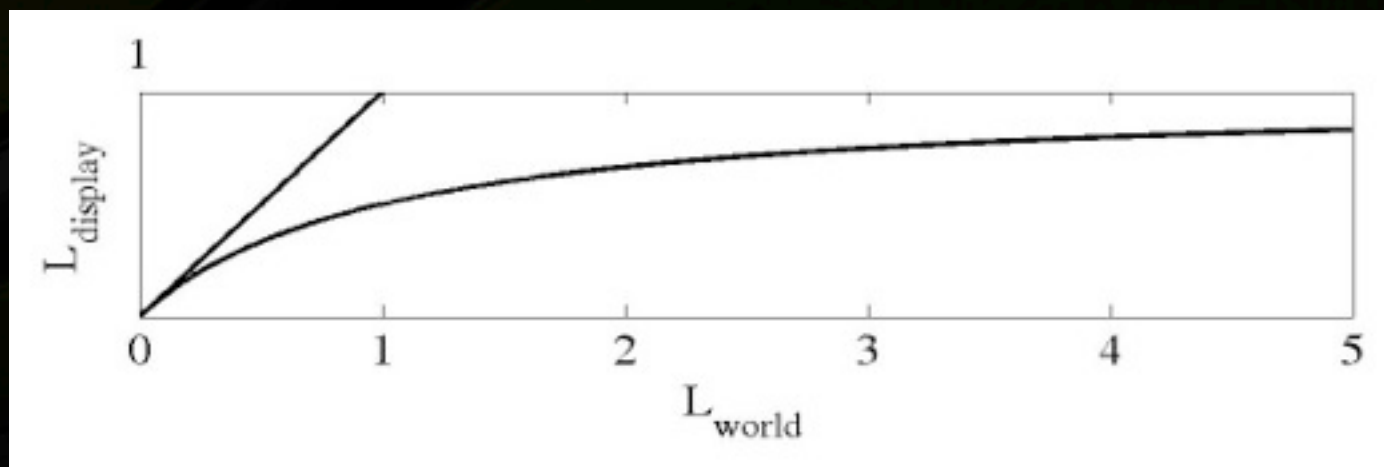
Simple global operator (Reinhard et al.)



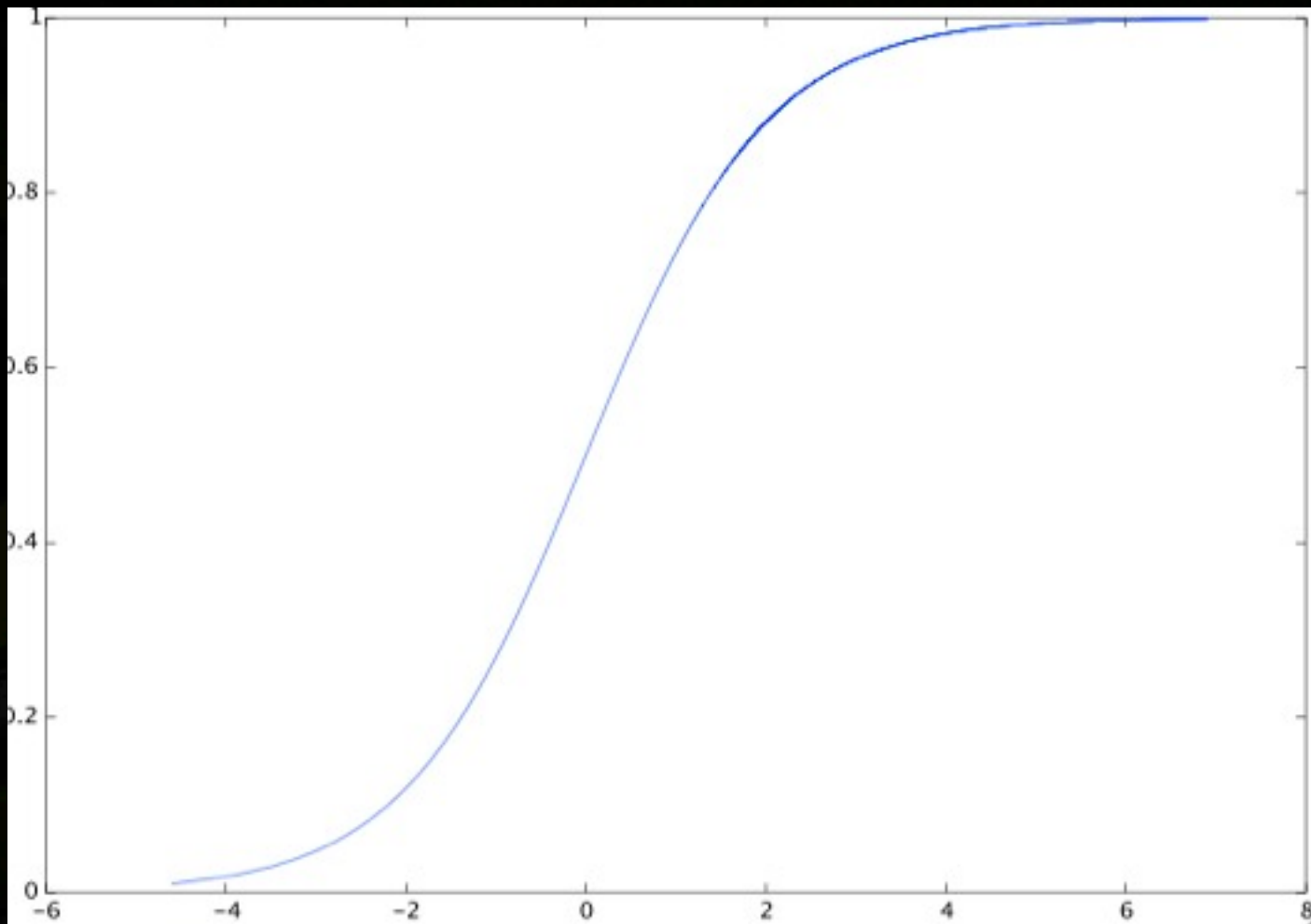
- Compression curve needs to
 - bring everything within range
 - leave dark areas alone
- In other words
 - asymptote is 1
 - derivative at 0 is 1

<http://www.cs.utah.edu/~reinhard/cdrone/tonemap.pdf>

$$L_{display} = \frac{L_{world}}{1 + L_{world}}$$



The same in log L closer to brightness perception





Reinhard operator

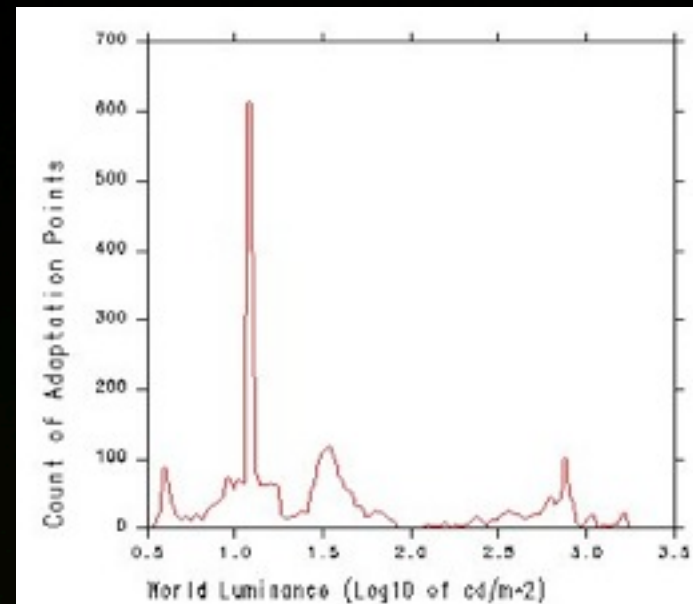


Darkest 0.1% scaled
to display device



Histogram adjustment [Ward et al. 1997]

- Histogram equalization
 - well-known method to increase contrast
 - luminance is not evenly spread, spread it
- Basic approach
 - lump pixels with 1deg area together
 - calculate histogram in $\log(\text{luminance})$ space
- Problem
 - doesn't just compress contrast, but also expands it
- Solution
 - put a ceiling to contrast by trimming large bins
 - not equalization, but adjustment



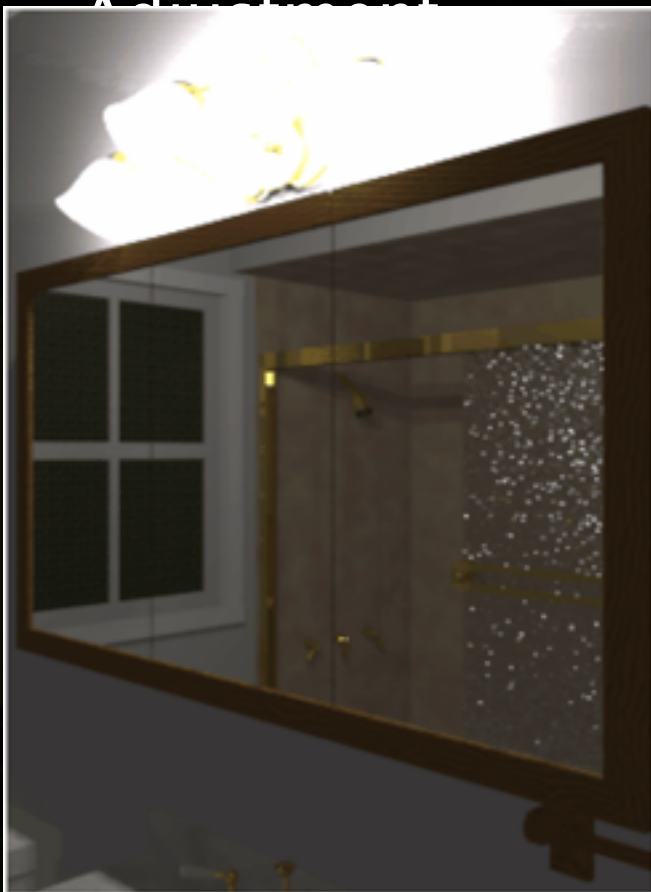
Equalization vs. adjustment



Linear

Equalization

Adjustment



Equalization vs. adjustment



Linear

Equalization

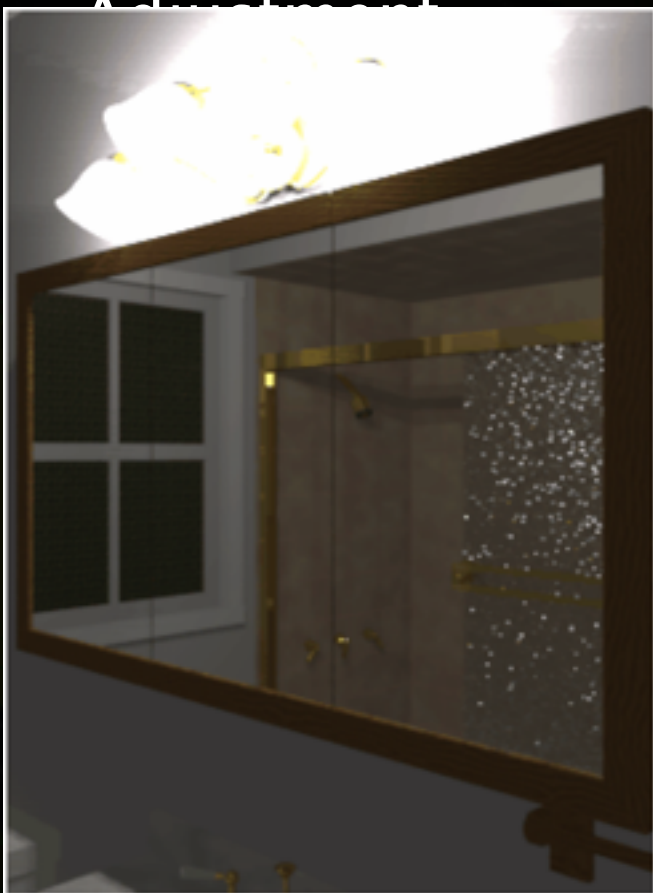


Equalization vs. adjustment



Linear

Equalization

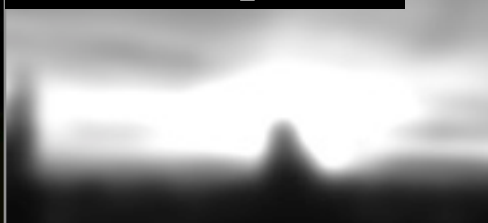


Oppenheim 1968, Chiu et al. 1993



- Reduce contrast of low-frequencies
- Keep high frequencies

Low-freq.



High-freq.



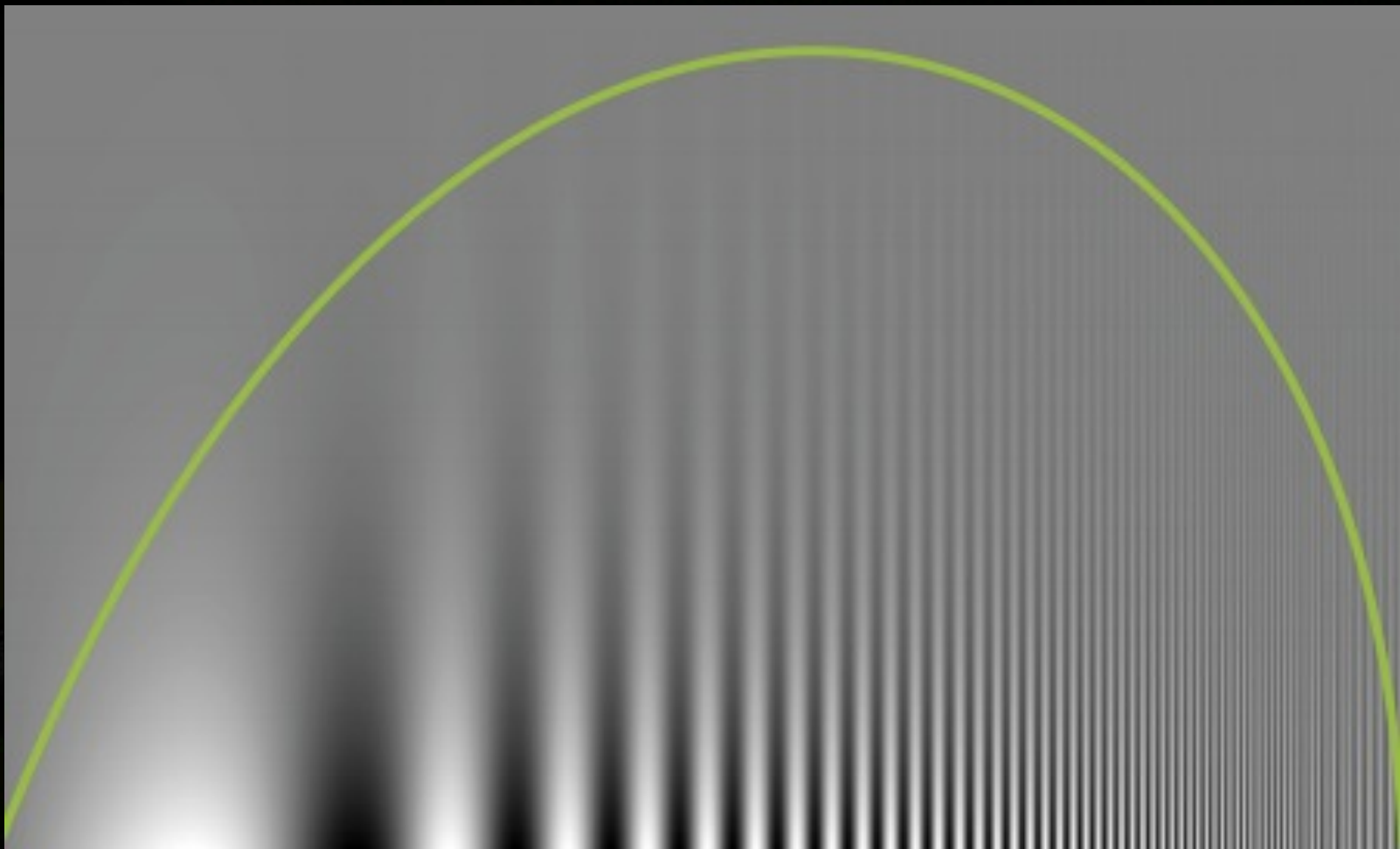
Color



Reduce low frequency



Contrast sensitivity function



Contrast sensitivity function



- Low sensitivity
 - to low frequencies
- Higher sensitivity
 - medium to high frequencies
- Most methods to deal with dynamic range
 - reduce the contrast of low frequencies
 - but keep the color

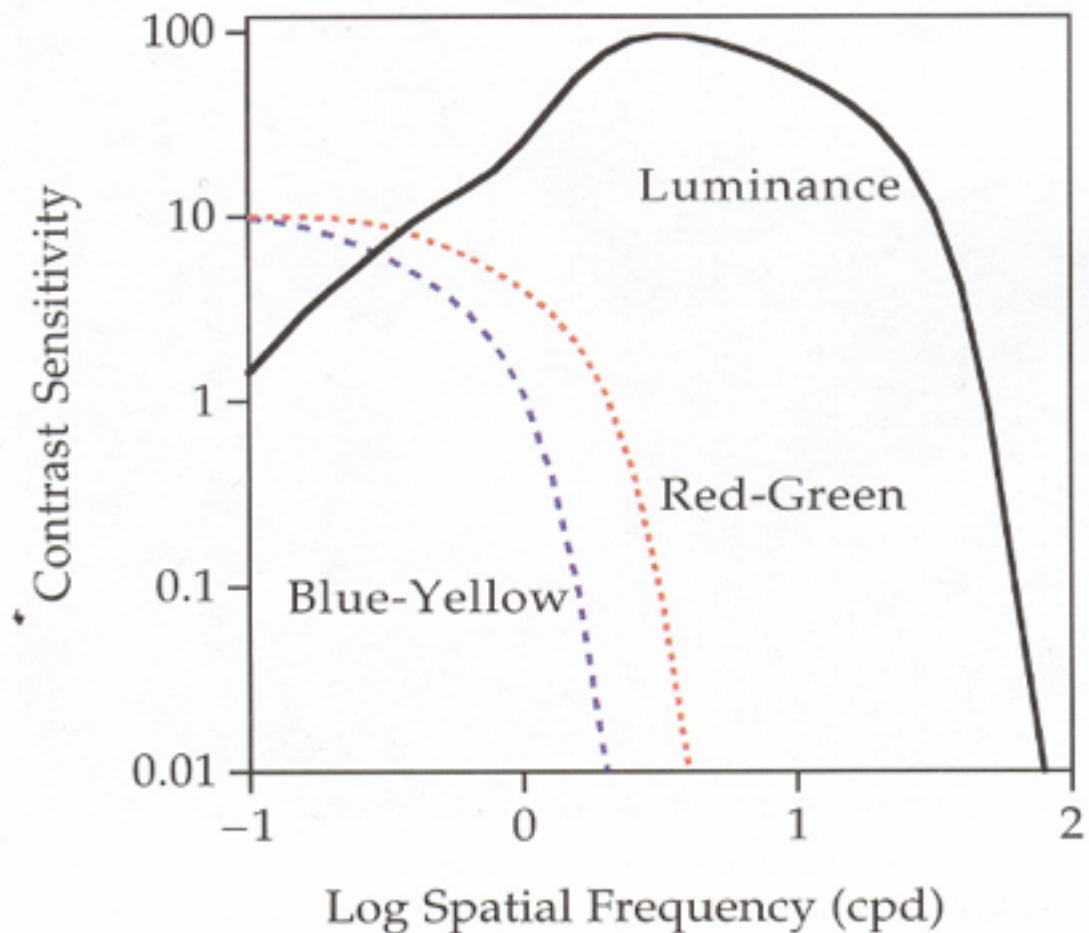


Figure 1-18. Spatial contrast sensitivity functions for luminance and chromatic contrast.

The halo nightmare

- For strong edges
- Because they contain high frequencies

Low-freq.



High-freq.



Color



Reduce low frequency



Durand & Dorsey 2002: Bilateral



Input HDR image



- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Durand & Dorsey 2002: Bilateral



Input HDR image



- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Intensity



Color



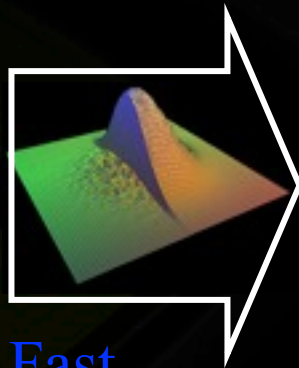
Durand & Dorsey 2002: Bilateral

Input HDR image



- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Intensity



Fast
Bilateral
Filter

Large scale



Detail

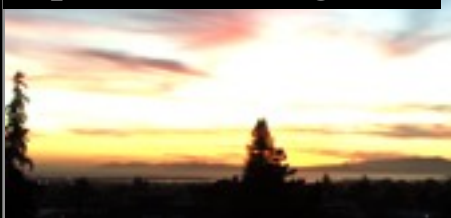


Color



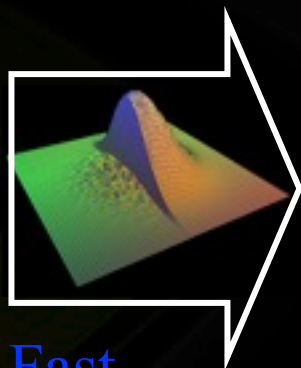
Durand & Dorsey 2002: Bilateral

Input HDR image



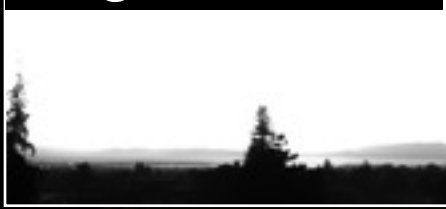
- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Intensity



Fast
Bilateral
Filter

Large scale



Detail



Reduce
contrast

Large scale



Color



Durand & Dorsey 2002: Bilateral

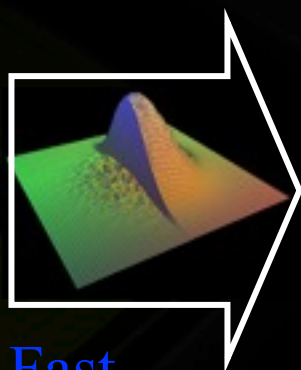


Input HDR image



- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Intensity



Fast
Bilateral
Filter

Large scale



Detail



Reduce
contrast

Large scale



Detail



Preserve!

Color



Color



Durand & Dorsey 2002: Bilateral



Input HDR image

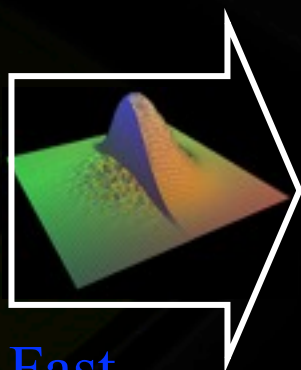


- Use non-linear filtering
 - to better separate details
 - without blurring across edges

Output



Intensity



Fast
Bilateral
Filter

Large scale



Reduce
contrast

Large scale



Detail



Preserve!

Detail



Color



Color





SIGGRAPH2005

Compressing and Combanding High Dynamic Range Images with Subband Architectures

Yuanzhen Li, Lavanya Sharan,

Edward Adelson

Massachusetts Institute of Technology

Halo Artifacts

Halos are widely believed to be inherent with multiscale methods.



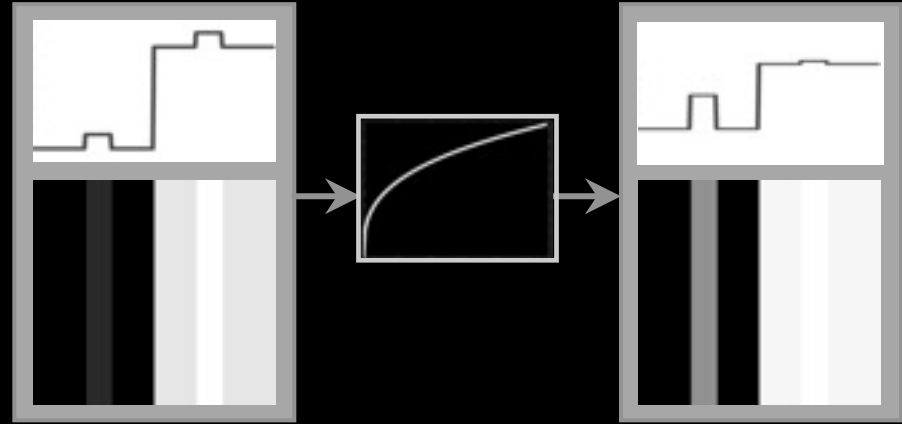
We fixed the halos

A subband/wavelet method that minimizes halos.



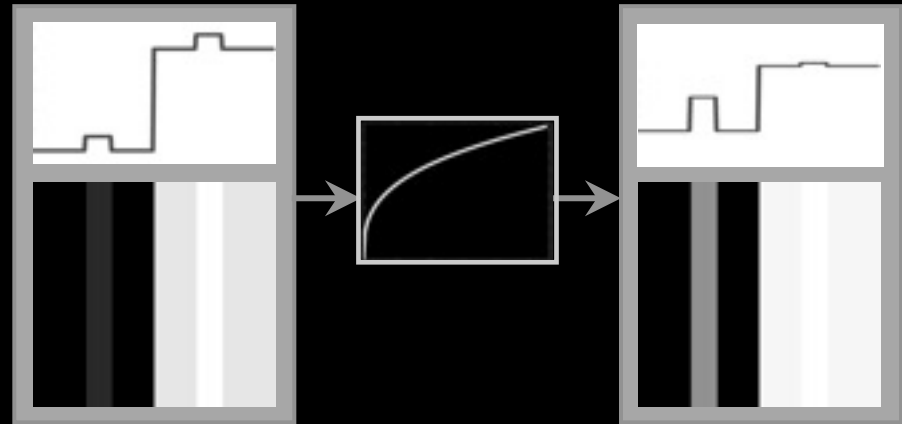
Range Compression

Method: Gamma or log on intensities
Problem: loss of detail

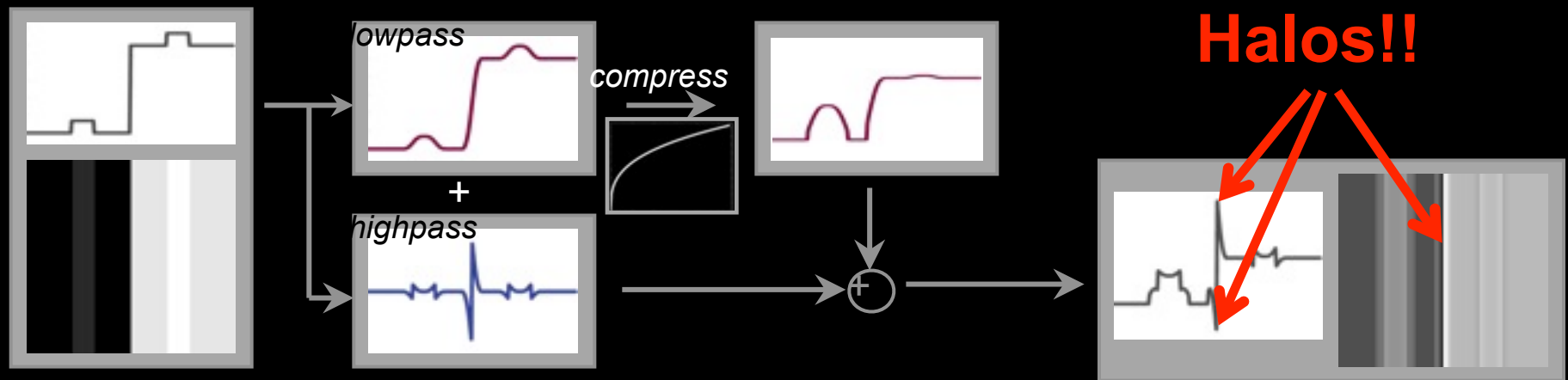


Range Compression

Method: Gamma or log on intensities
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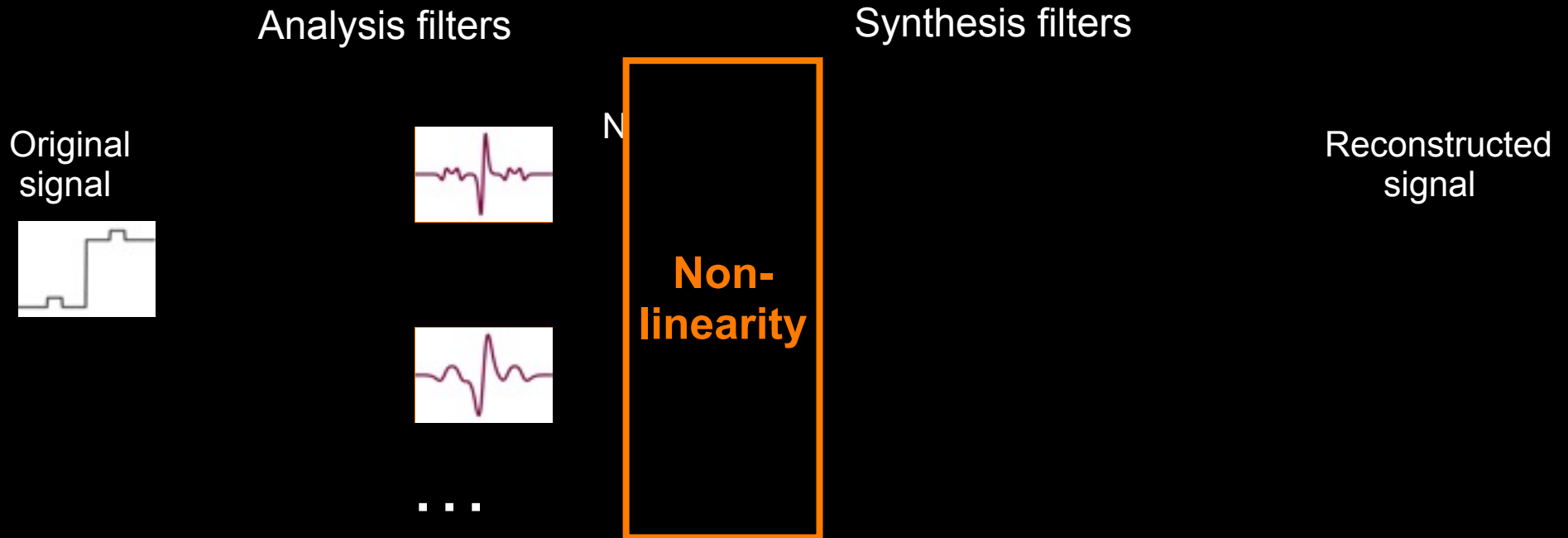
Solution: filtering
Problem: halos



How did we fix the halos?

- Analysis-synthesis subband architecture, e.g., wavelets
- Smooth gain control on subbands

Analysis-Synthesis Architecture

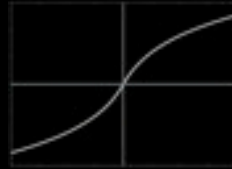


Point Nonlinearity on Subbands

Original subband



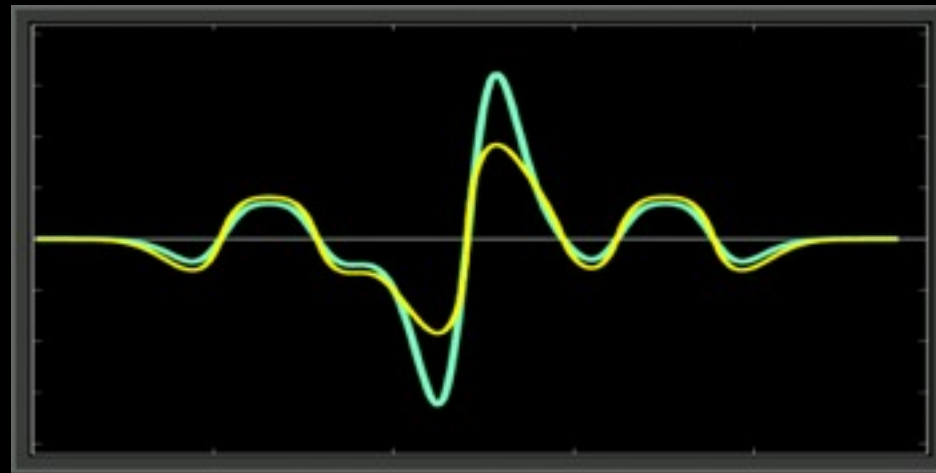
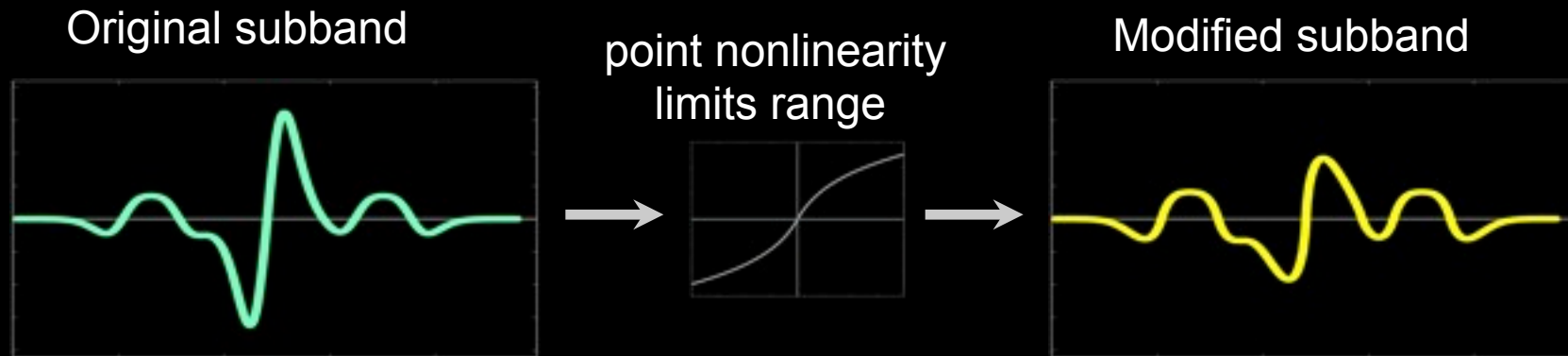
point nonlinearity
limits range



Modified subband

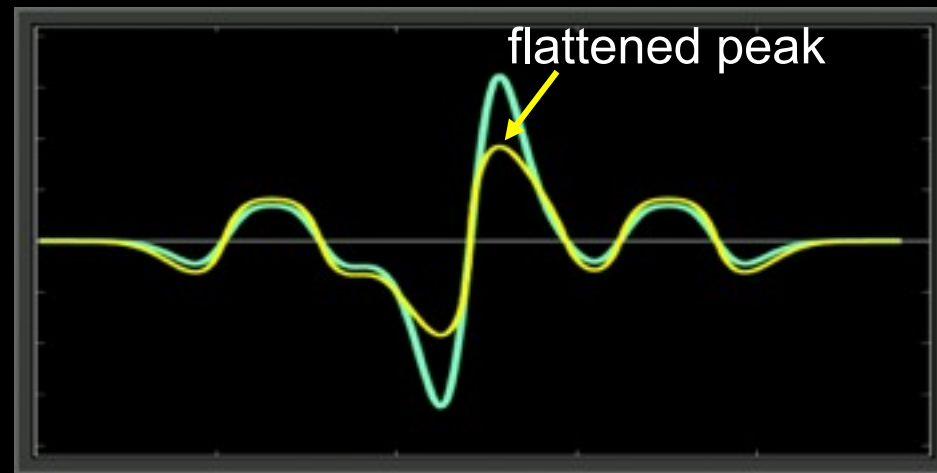
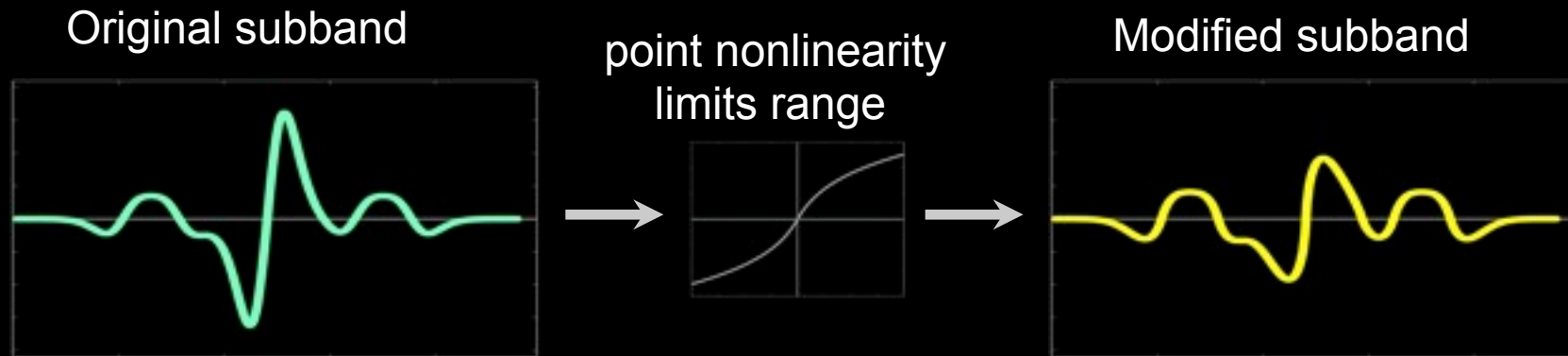


Point Nonlinearity on Subbands



Problem: Nonlinear distortion

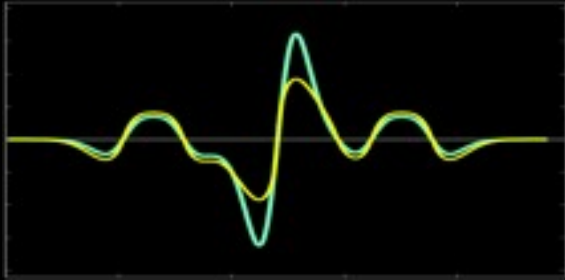
Point Nonlinearity on Subbands



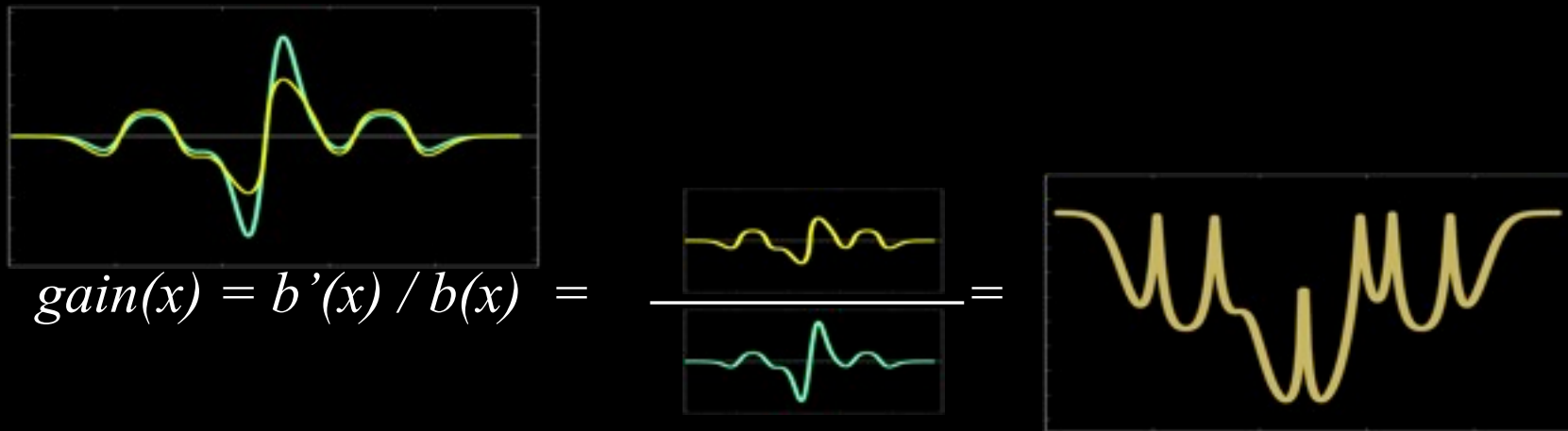
Problem: Nonlinear distortion

Smooth Gain Control

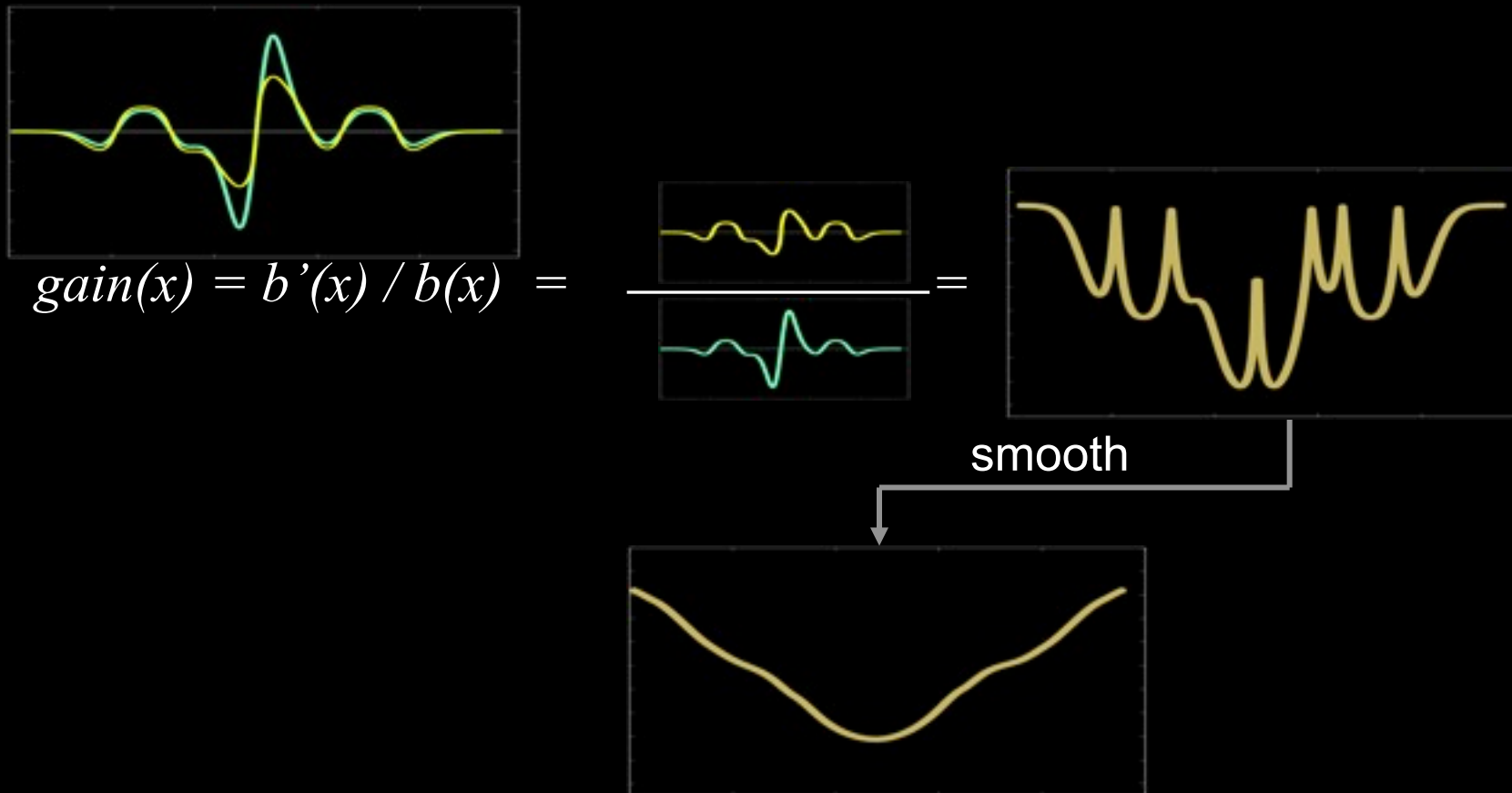
Smooth Gain Control



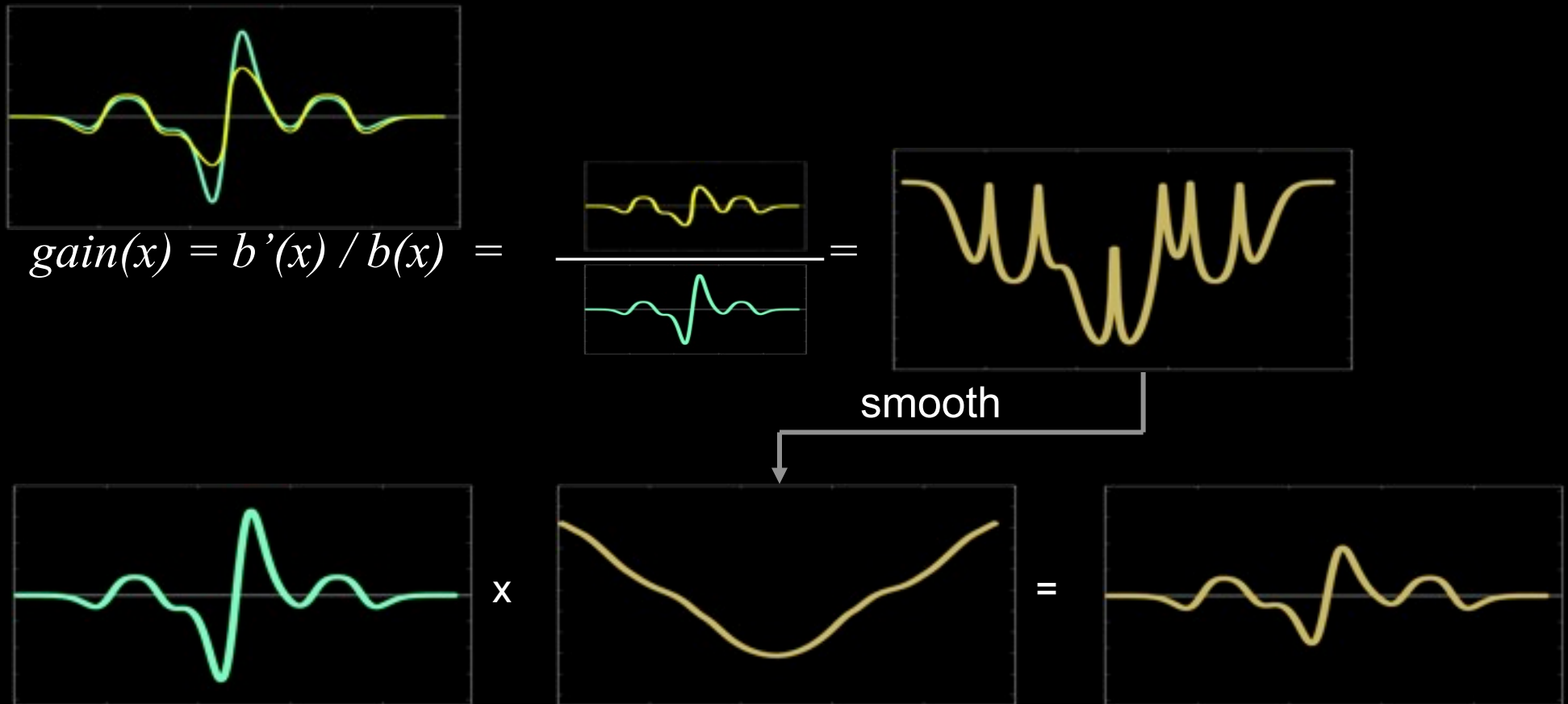
Smooth Gain Control



Smooth Gain Control

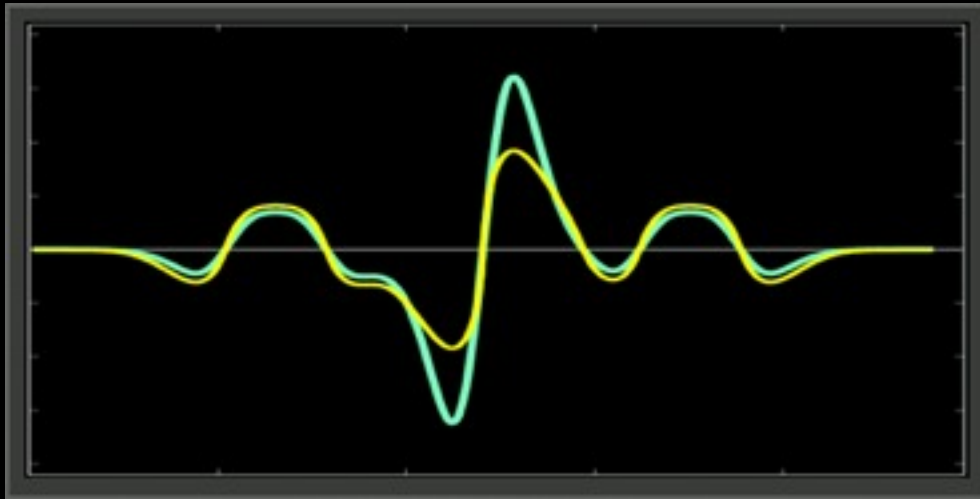


Smooth Gain Control



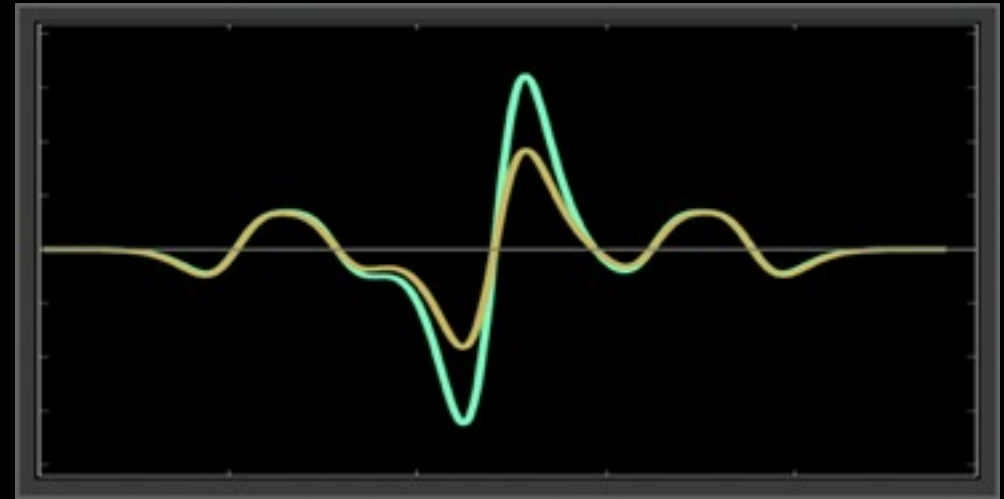
Smooth Gain Control Reduces Distortion

Point nonlinearity



Distorted

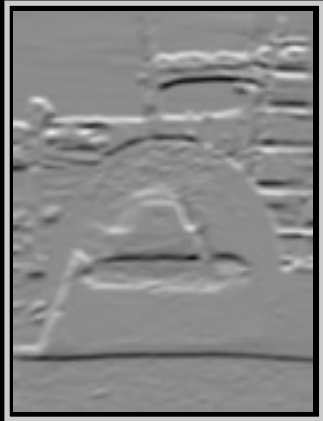
Smooth gain control



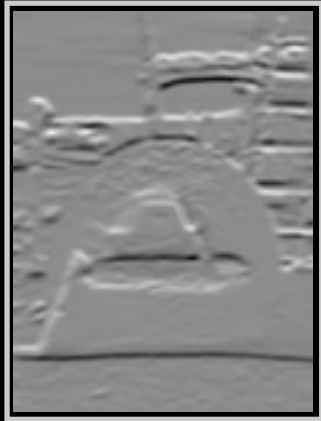
Distortion reduced

Smooth Gain Control on Subbands

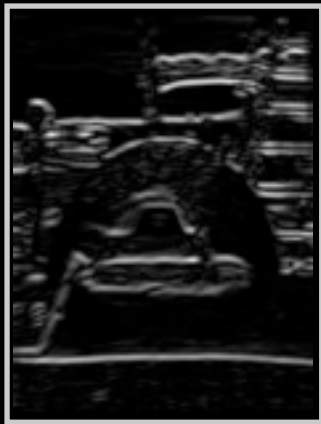
Smooth Gain Control on Subbands



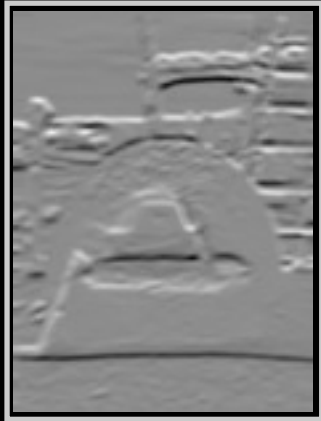
Smooth Gain Control on Subbands



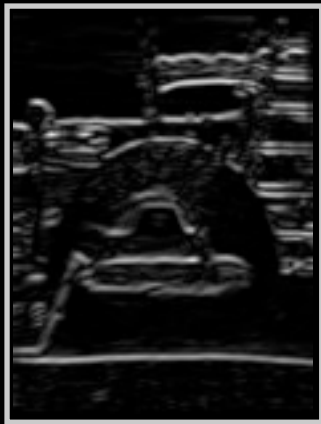
↓ rectify (by taking abs)



Smooth Gain Control on Subbands



↓ rectify (by taking abs)

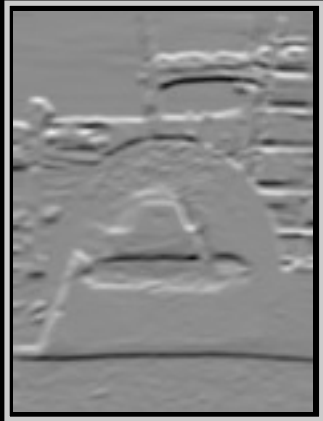


→ blur

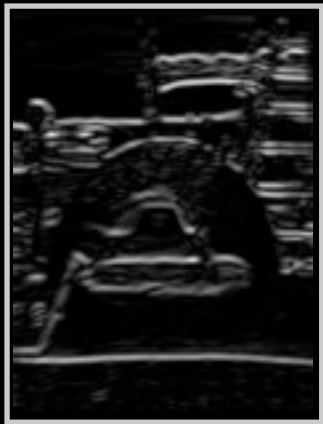


activity map

Smooth Gain Control on Subbands



↓ rectify (by taking abs)



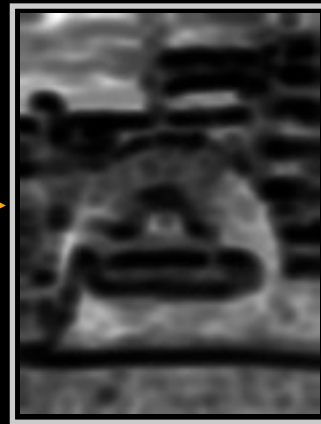
→ blur



activity map

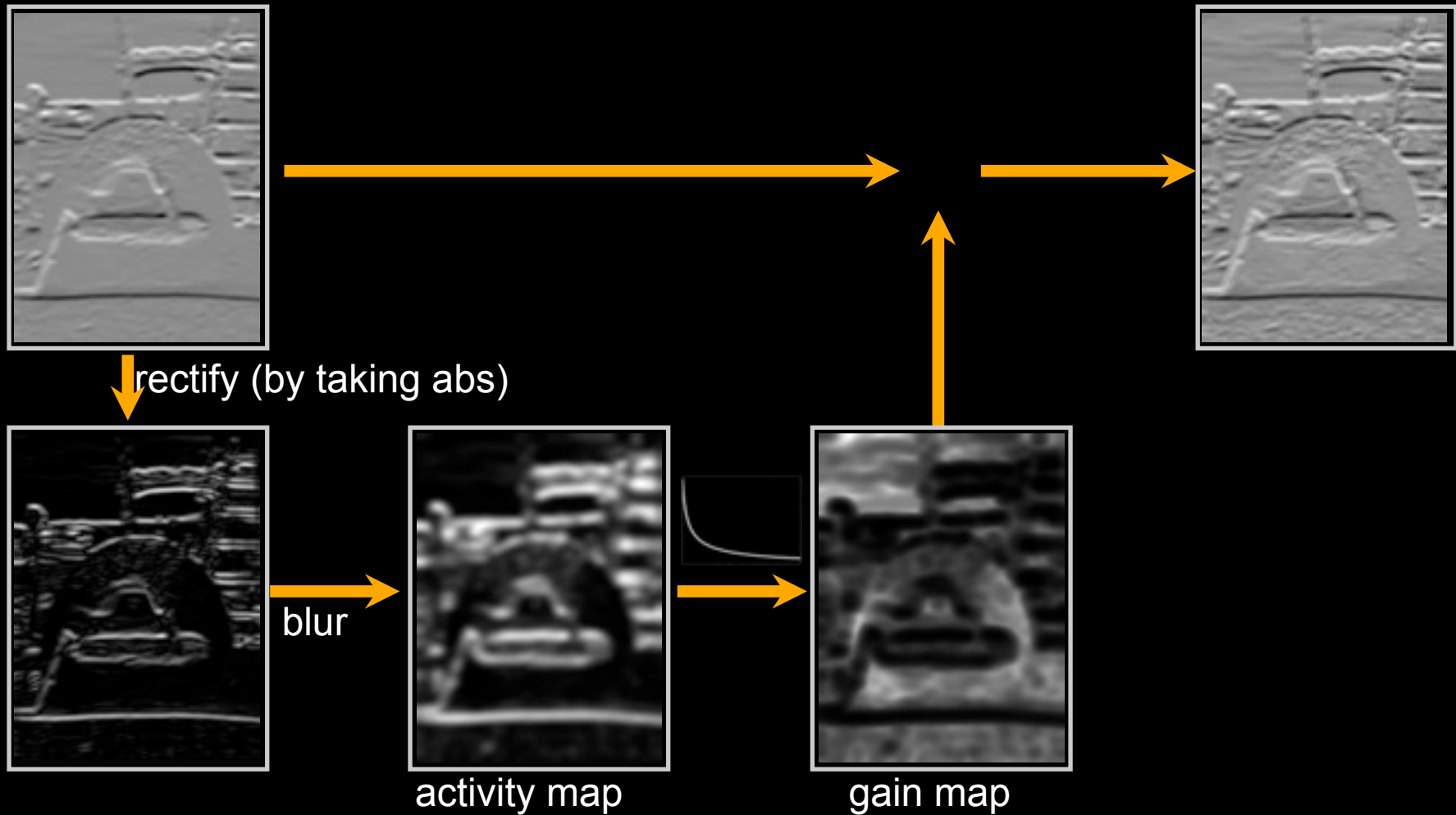


→



gain map

Smooth Gain Control on Subbands



Ours



Durand & Dorsey 2002



Reduced contrast

- Tonemapping reduces dynamic range
 - usually also contrast is reduced
 - for example the skyline below becomes less visible

reference HDR image (clipped)



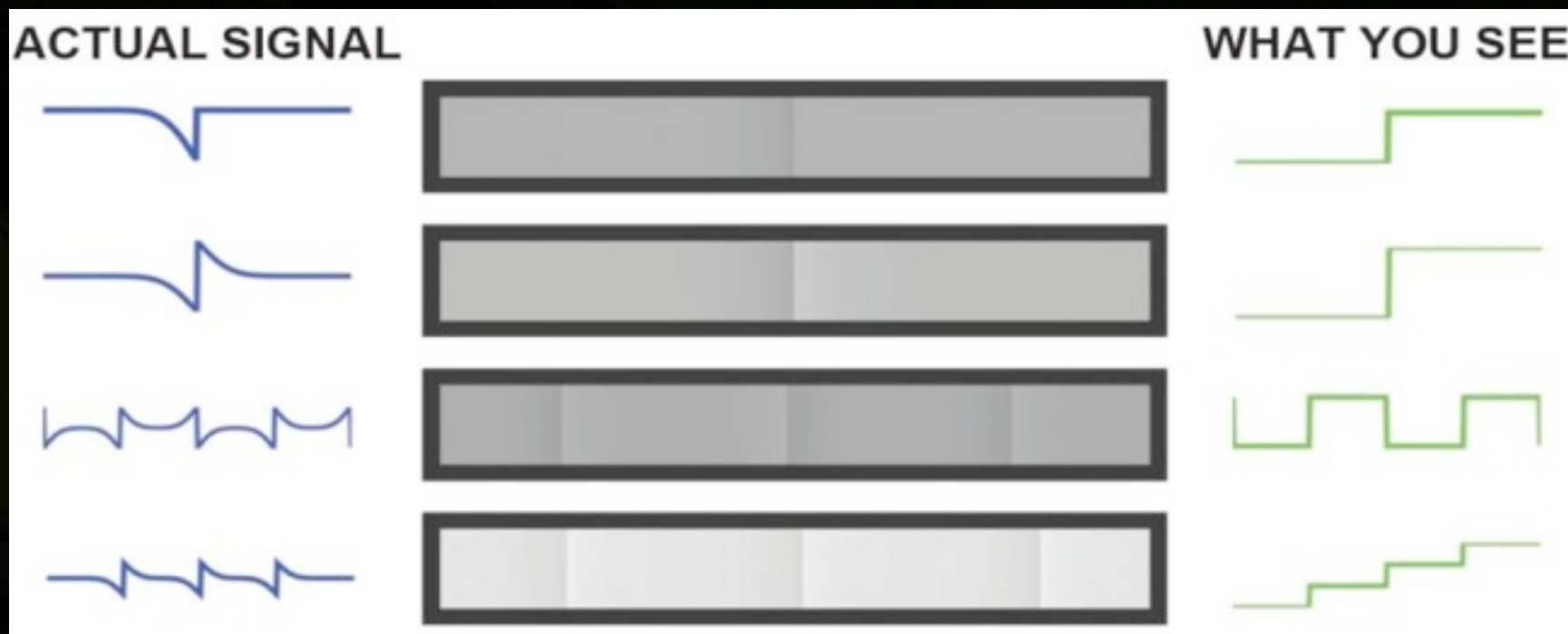
tone mapping



Countershading enhances contrast



- Make use of perceptual illusion
 - Craik - O'Brien - Cornsweet effect

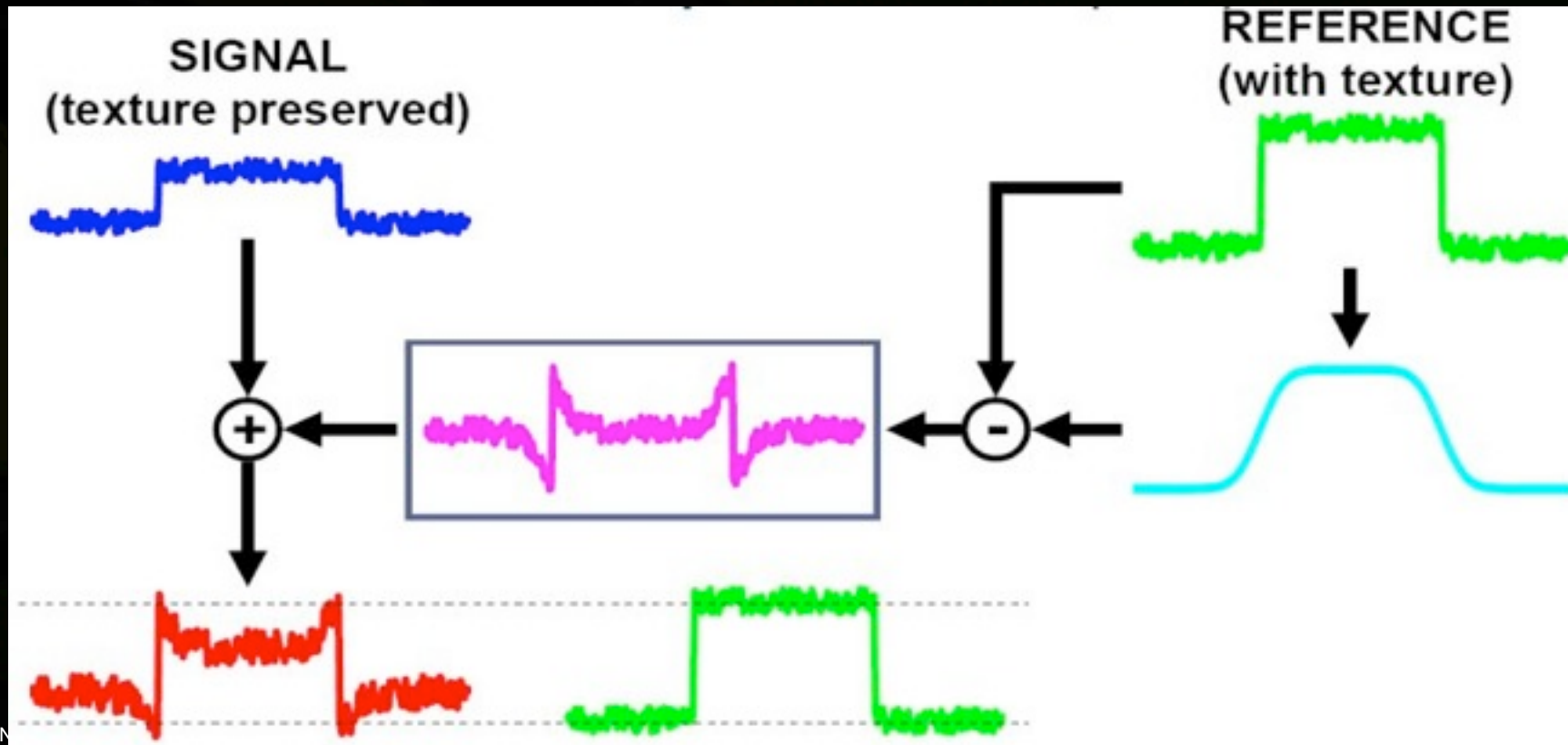


Unsharp masking

- Tonemapped HDR images lose contrast, add some back

Tonemapped LDR

Original HDR



Unsharp masking



- Too much contrast changes the visual appearance
- Global scale for contrast enhancements
 - features have different scales



Adaptive Countershading



- [Krawczyk et al. 07]
- Multiresolution local contrast metric adapts contrast to features
- Visual detection model avoids halos



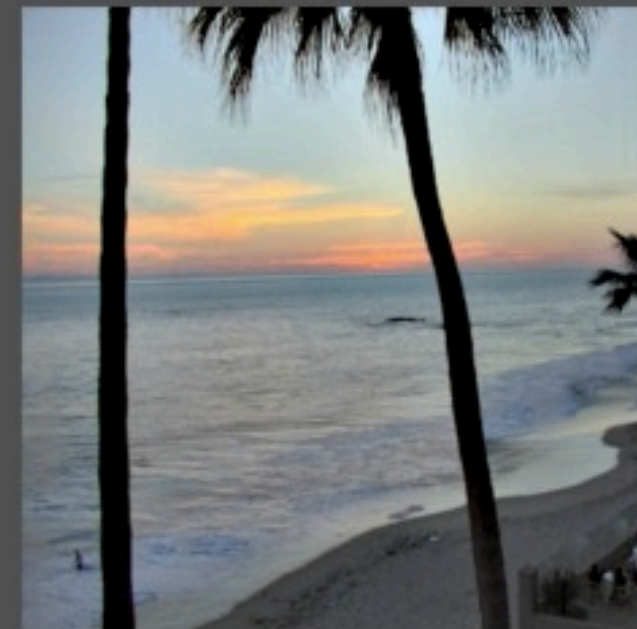
Unsharp Masking, Countershading and Halos: Enhancements or Artifacts?



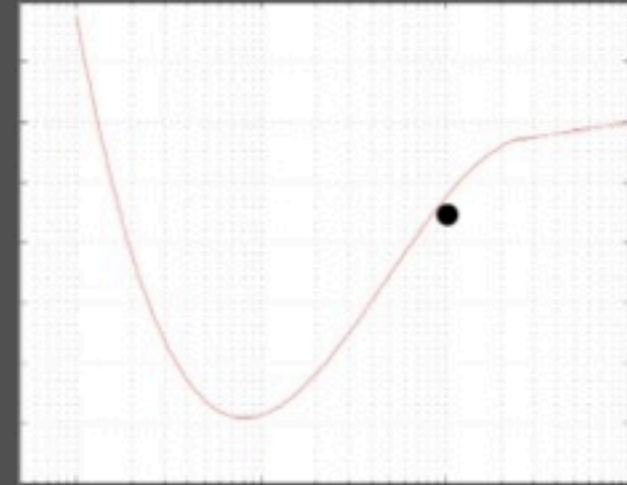
Sharpness

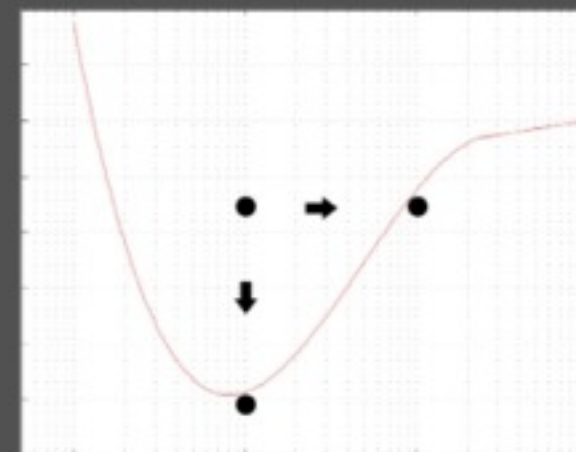
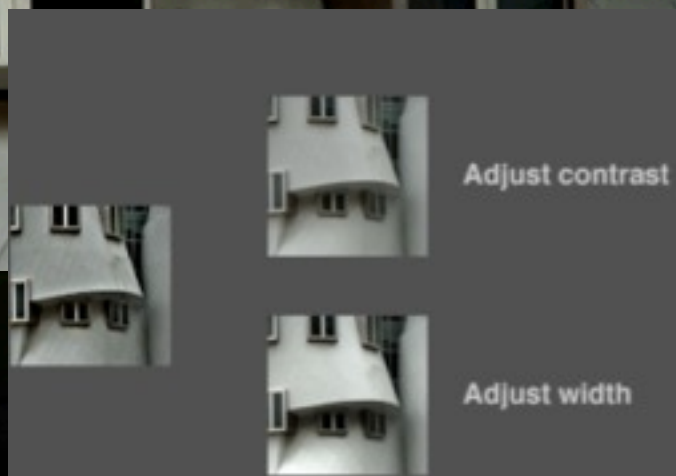
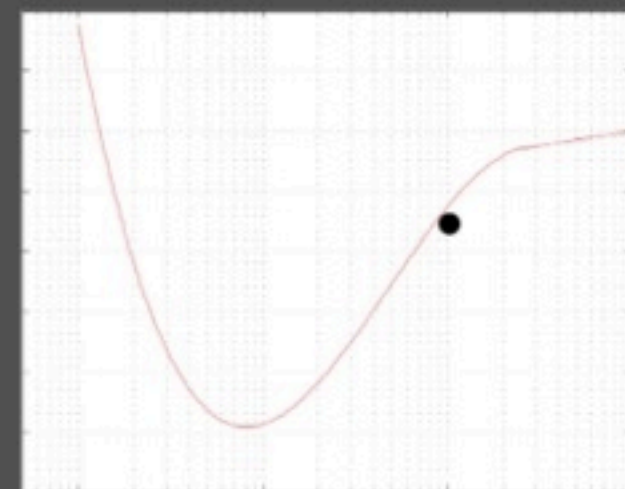


Haloes



Contrast





NVIDIA Research

0:30 – 3:00; 3:35 – 4:10



0:30 – 3:00; 3:35 – 4:10



0:30 – 3:00; 3:35 – 4:10



Exposure Fusion: Simplified HDR



Mertens, Kautz, van Reeth PG 2007

- Choose the best pixel from one of the images
 - Use heuristics for a smooth selection, such as
 - Exposure
 - Color saturation
 - Contrast



LDR
images



Weight
maps

The Laplacian pyramid



Gaussian Pyramid



The Laplacian pyramid



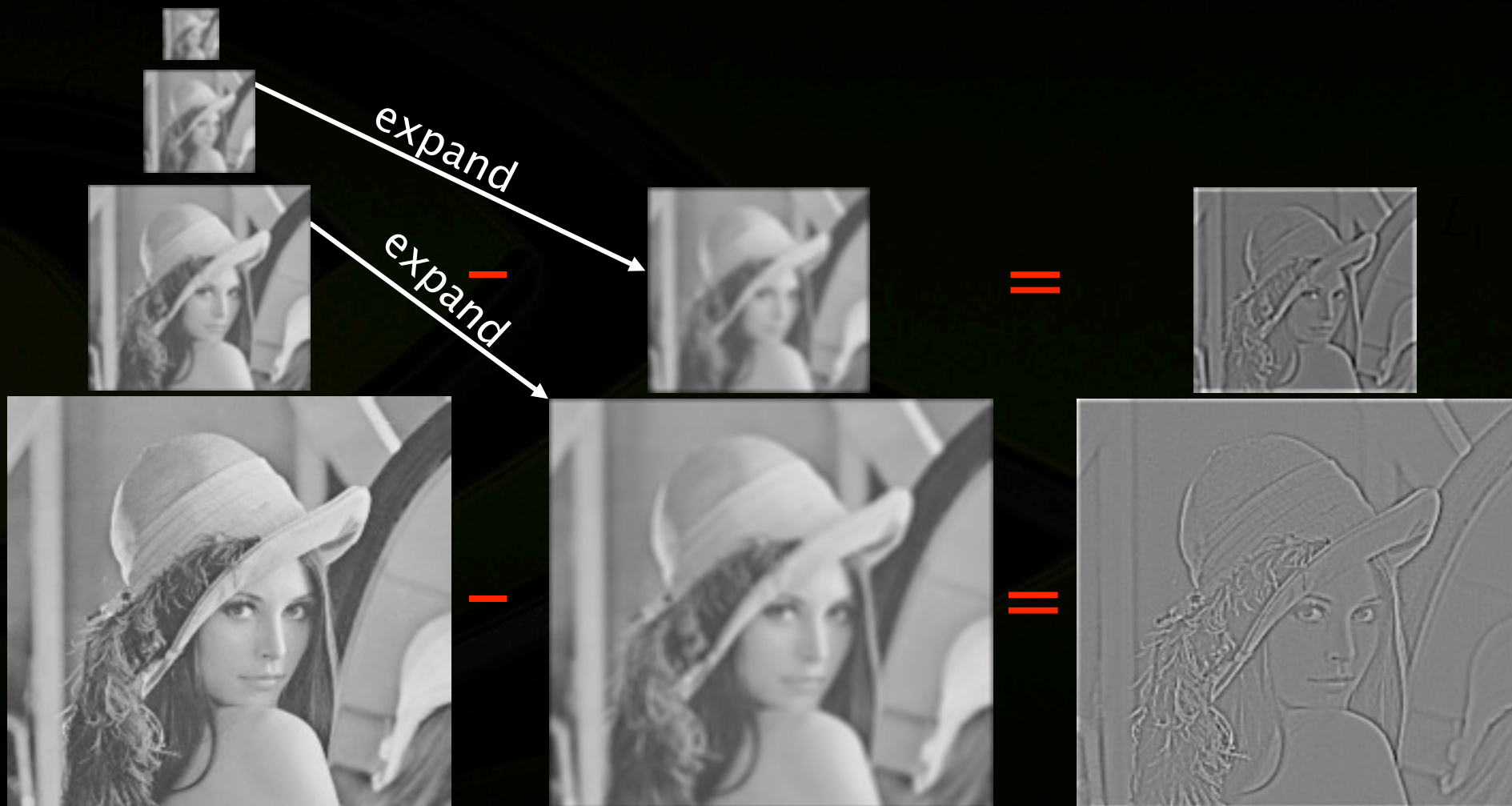
Gaussian Pyramid



The Laplacian pyramid



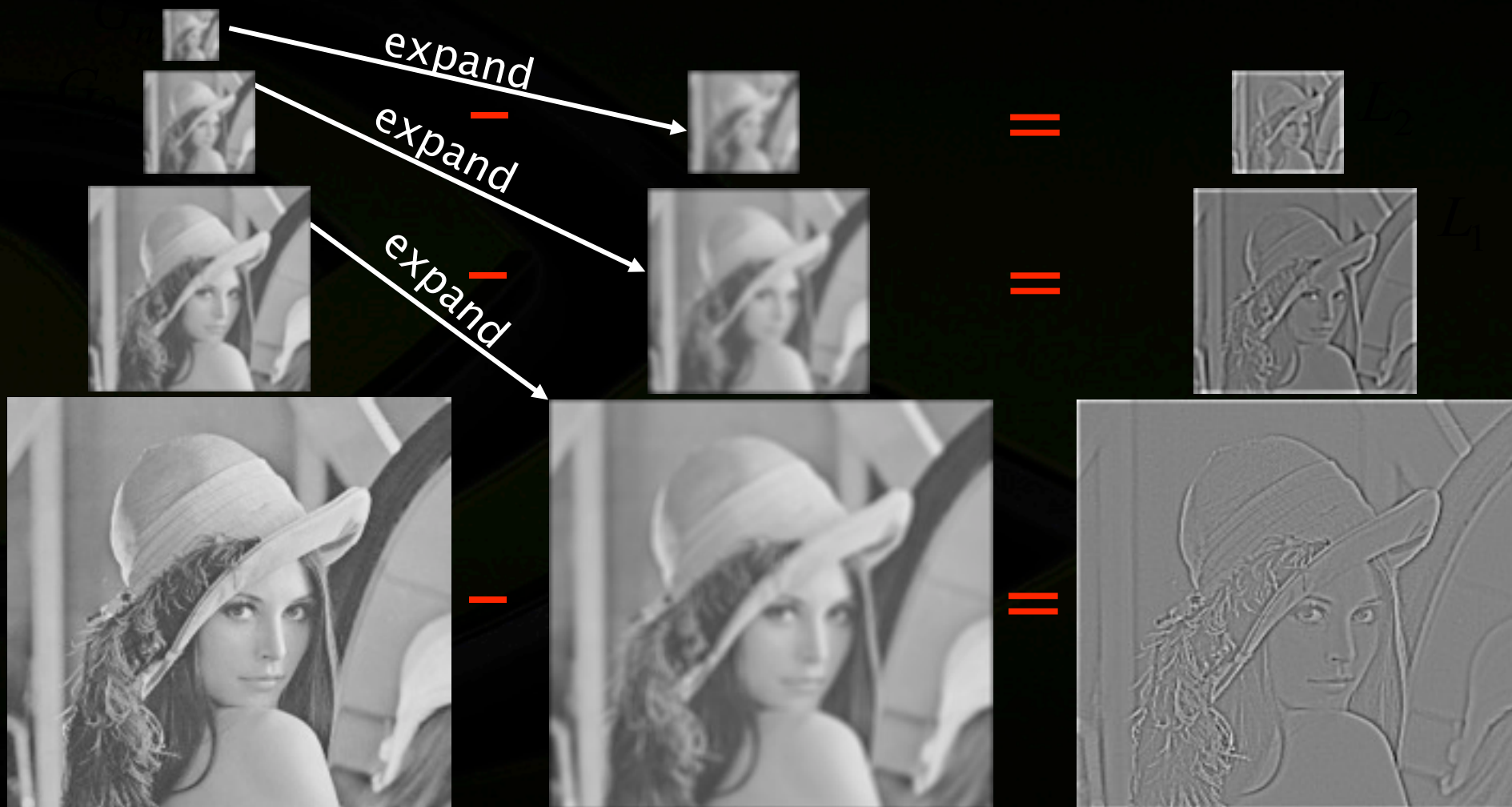
Gaussian Pyramid



The Laplacian pyramid



Gaussian Pyramid

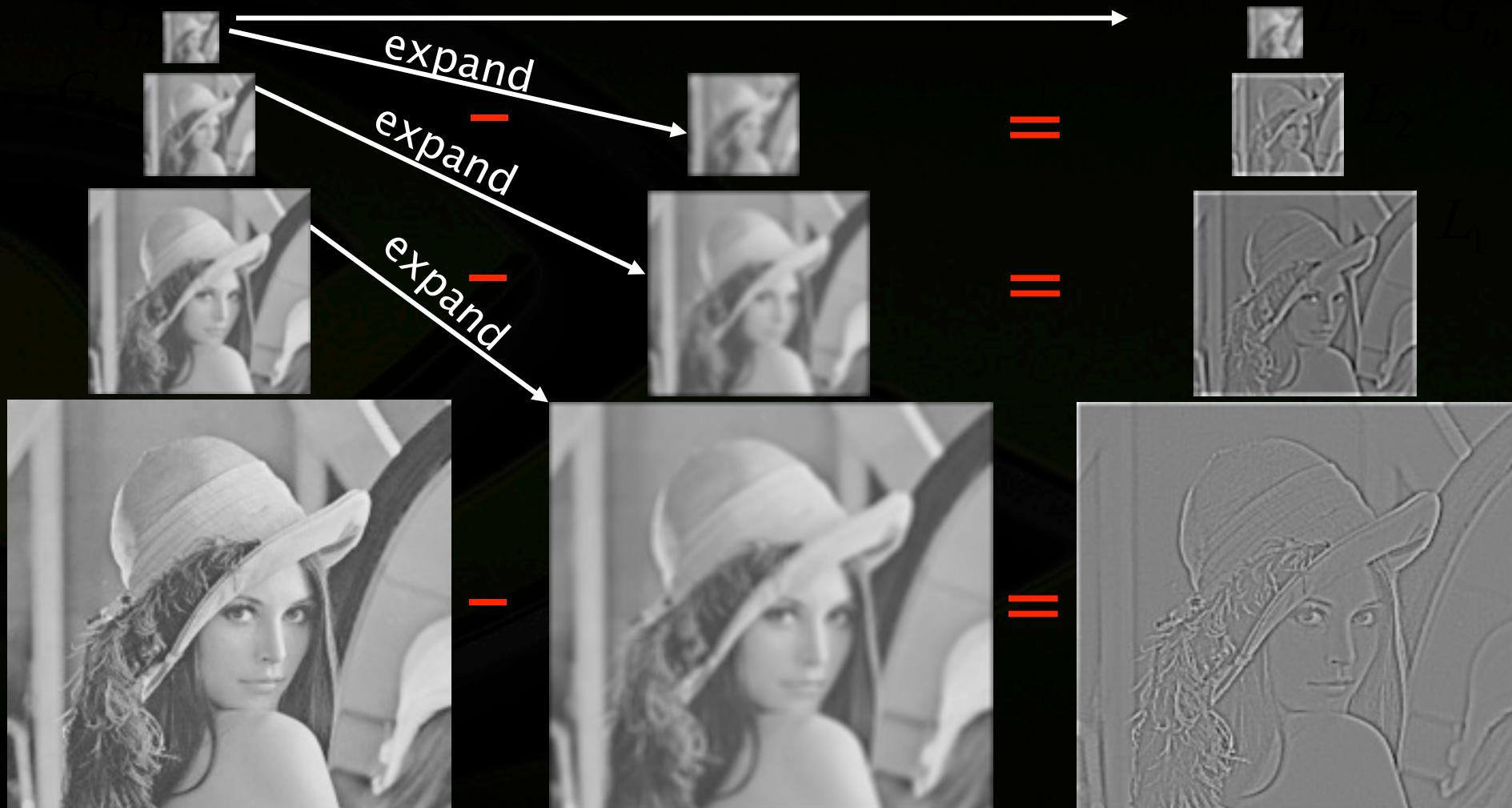


The Laplacian pyramid

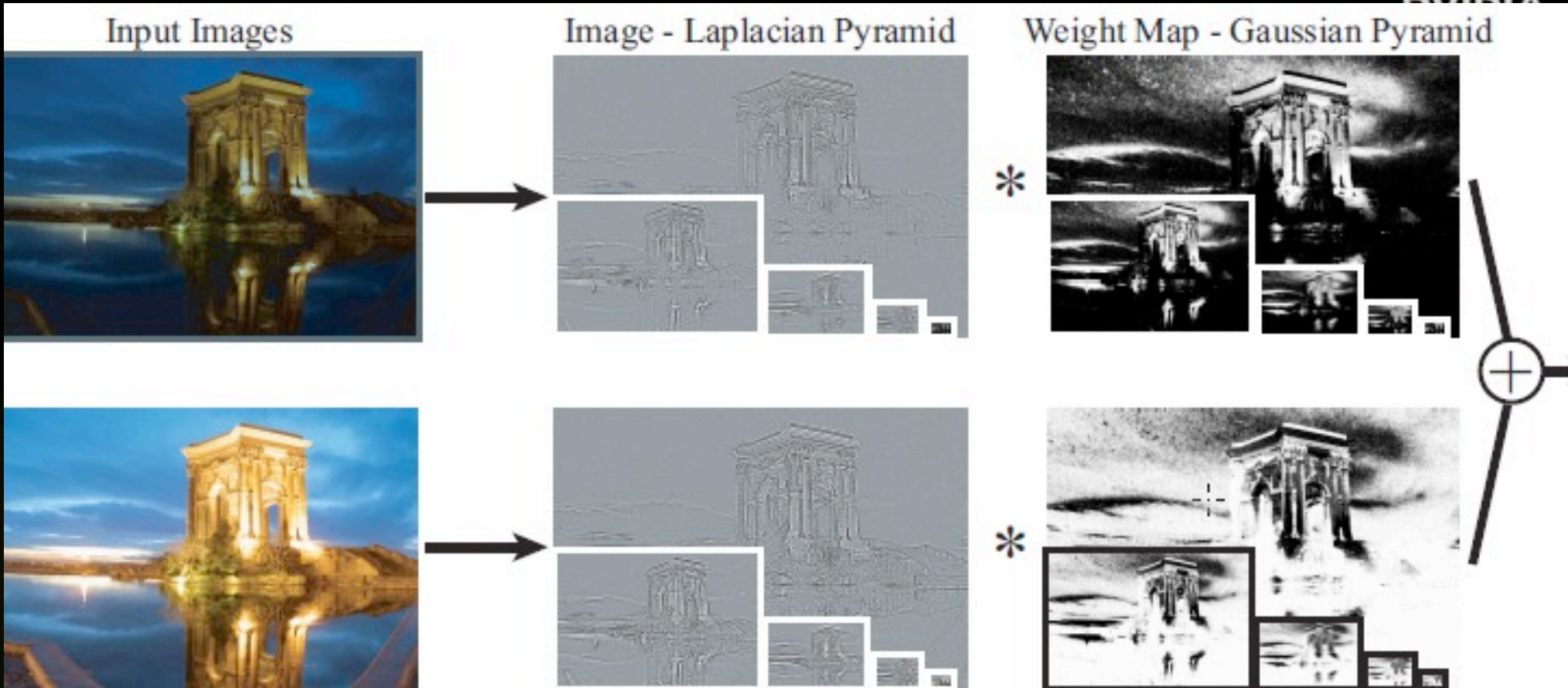


Gaussian Pyramid

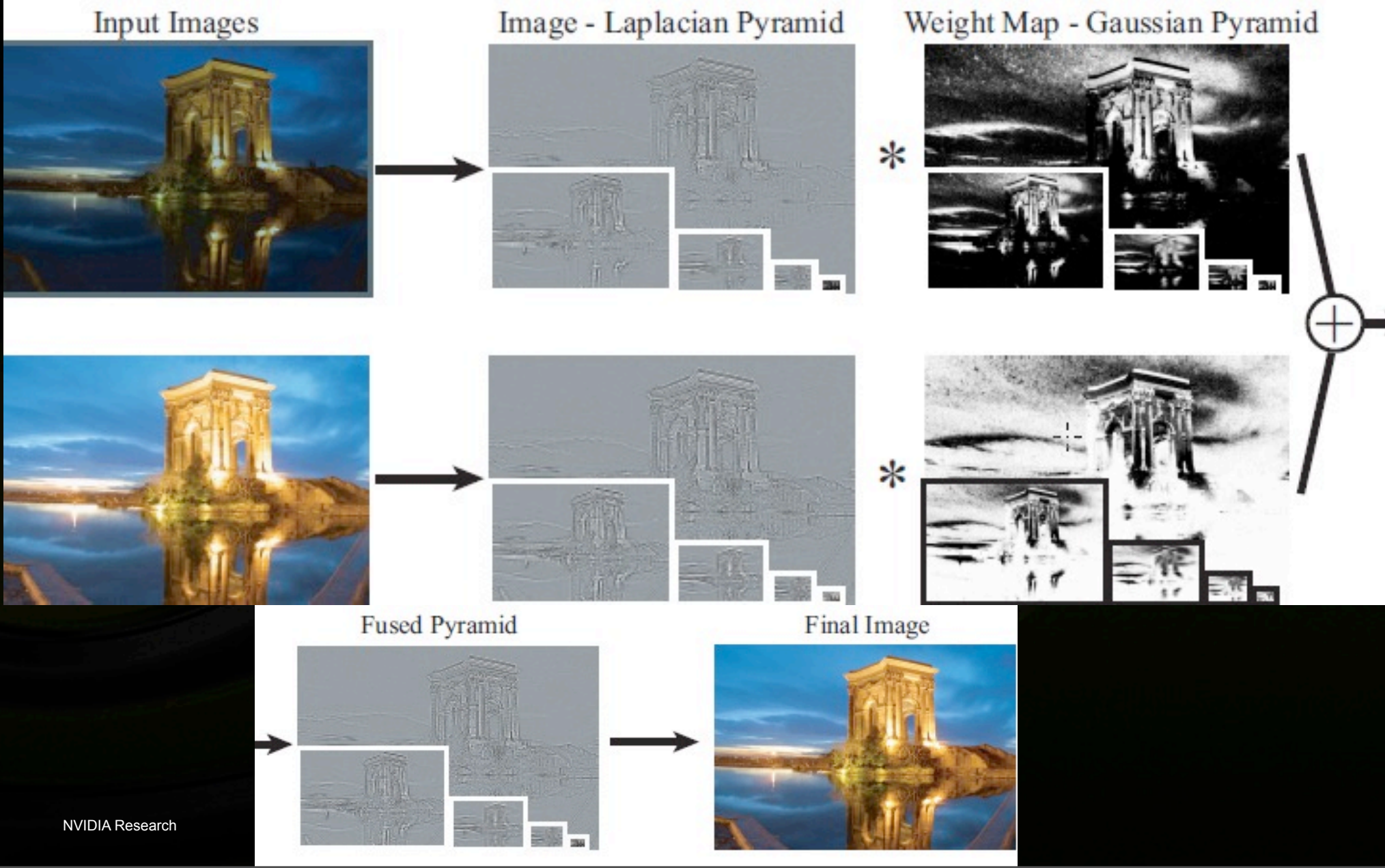
Laplacian Pyramid



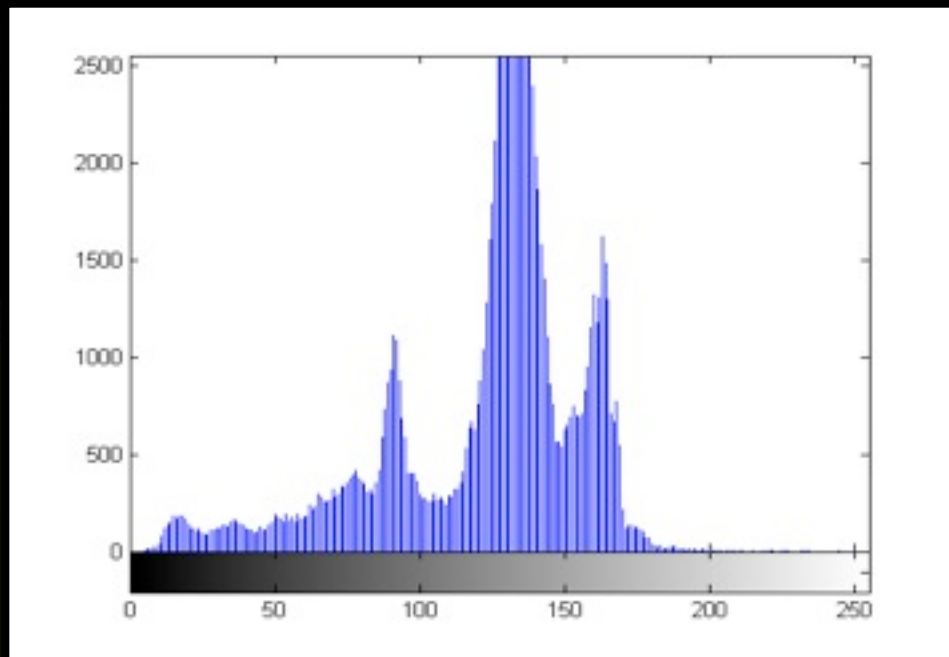
Multi-resolution fusion



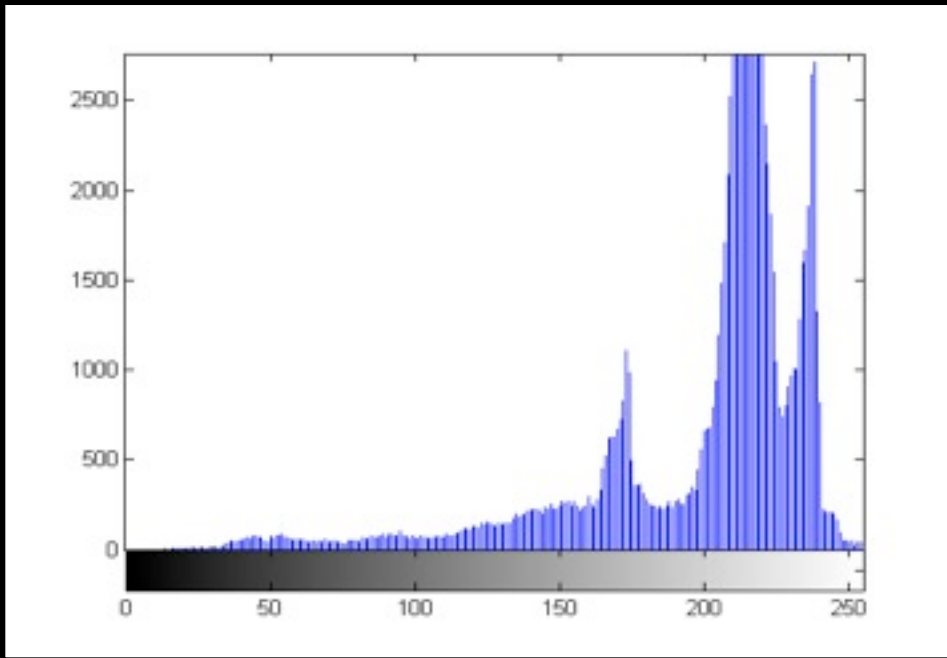
Multi-resolution fusion

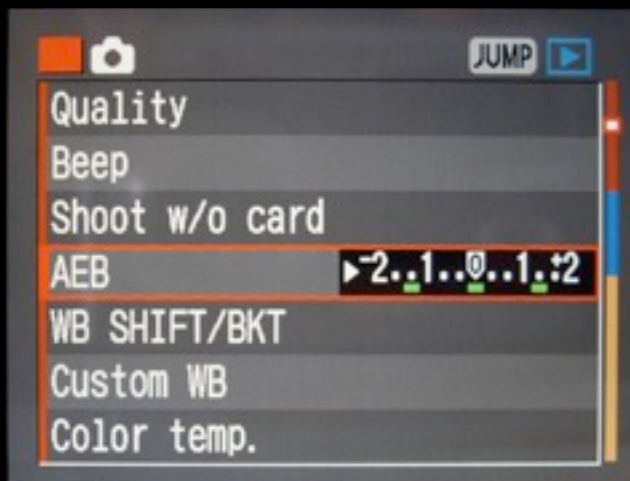


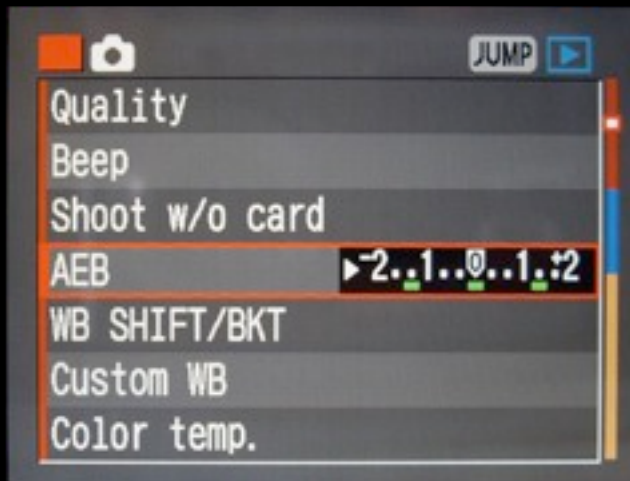
Metering



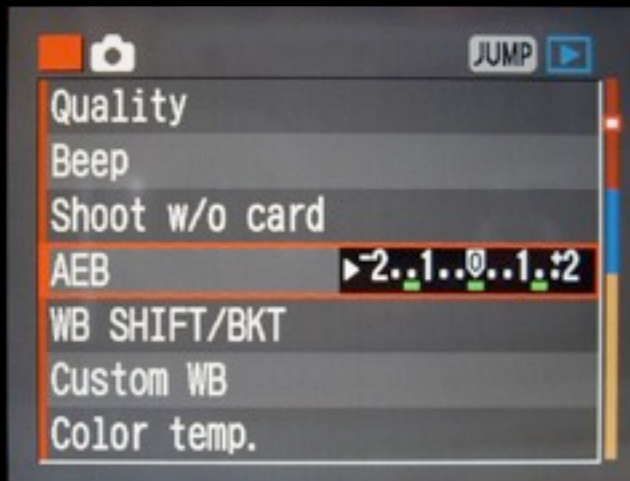
O. Gallo, M. Tico, N. Gelfand, K. Pulli
Metering for Exposure Stacks
Eurographics, Cagliari, Italy, May 13 - 18, 2012



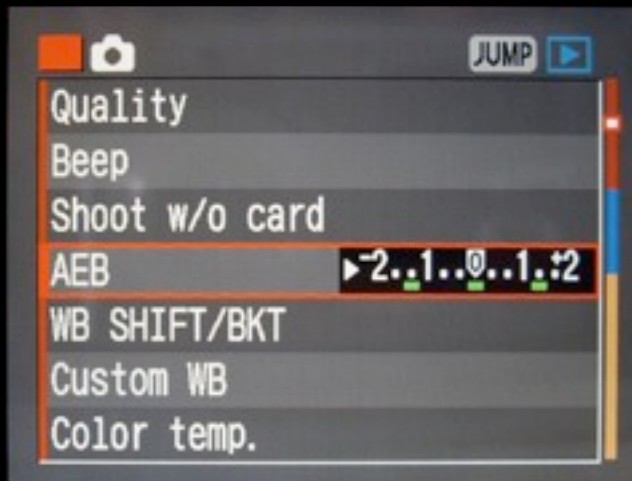




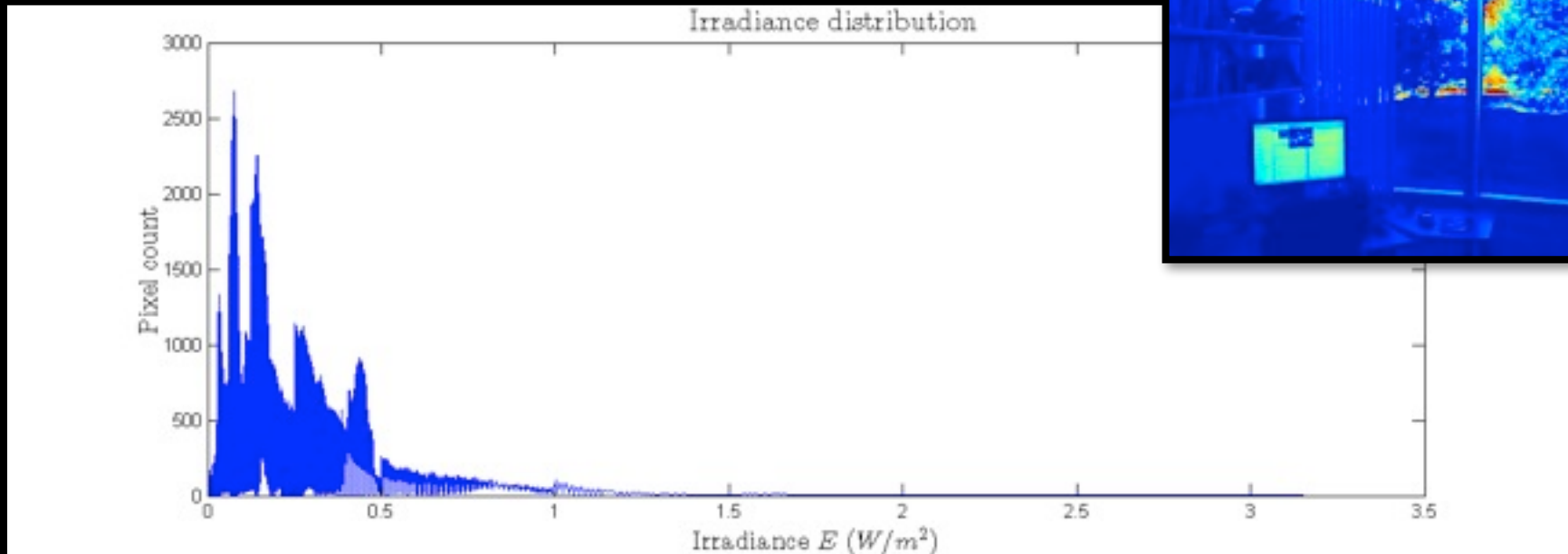
- “Minimal-Bracketing Sets for High-Dynamic-Range Image Capture”, *Barakat et al., Transaction on Image Processing 2008*
- “High Dynamic Range Imaging on Mobile Devices”, *Bilcu et al., ICECS 2008*
- “Optimal Scheduling of Capture Times in a Multiple Capture Imaging System”, *Chen and El Gamal, SPIE 2002*
- “Optimal HDR Reconstruction with Linear Digital Cameras”, *Granados et al., CVPR 2010*

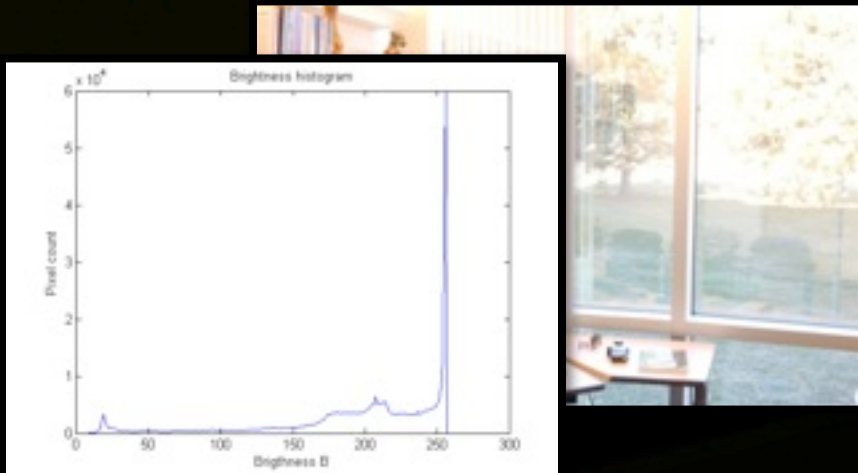
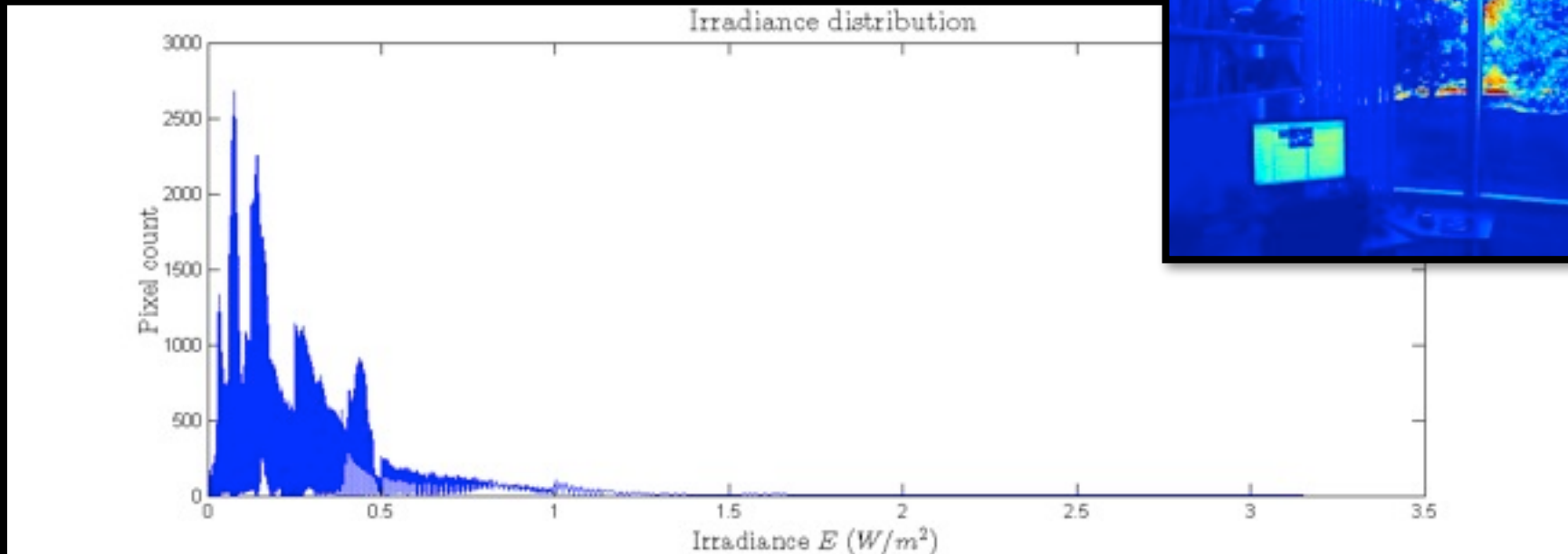


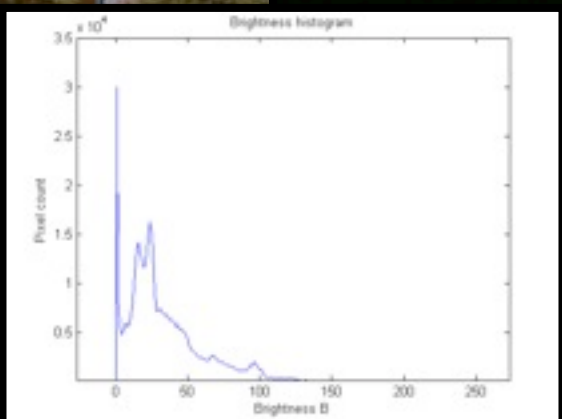
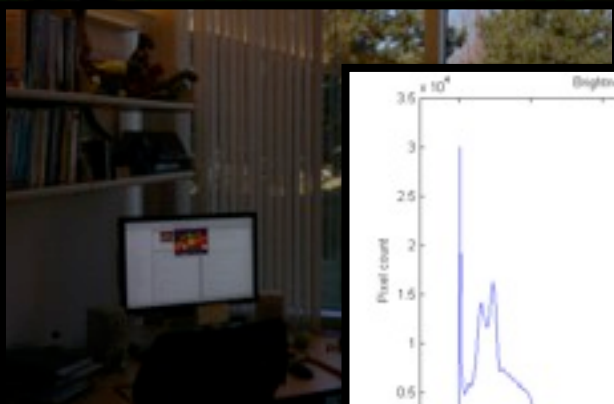
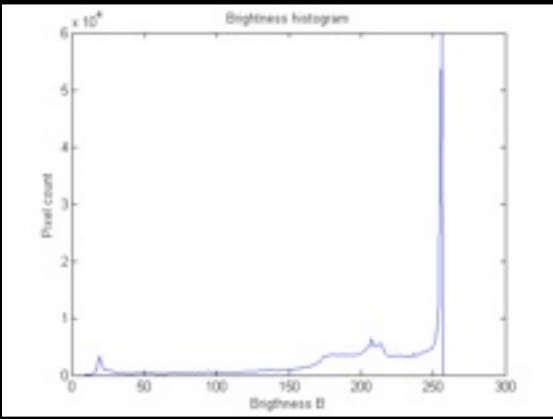
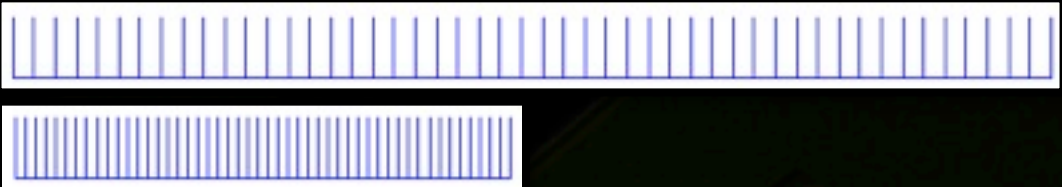
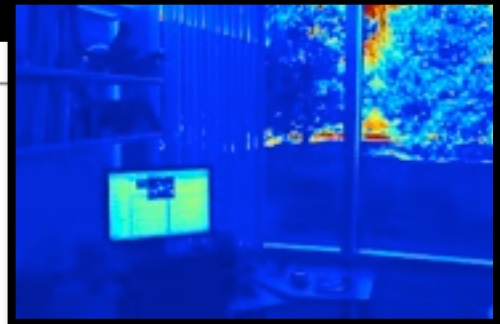
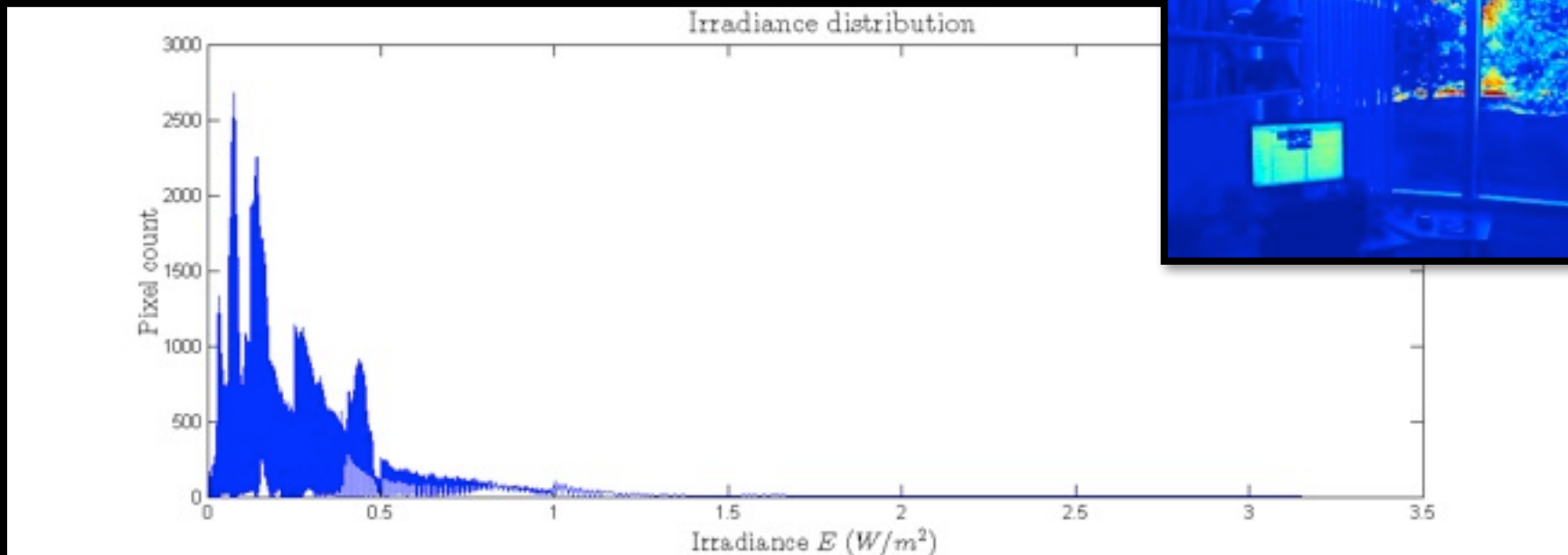
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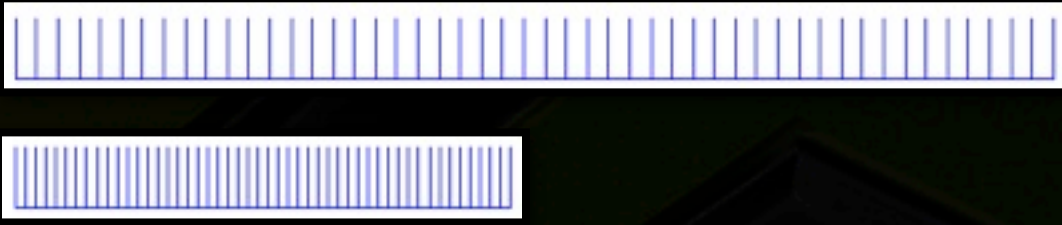
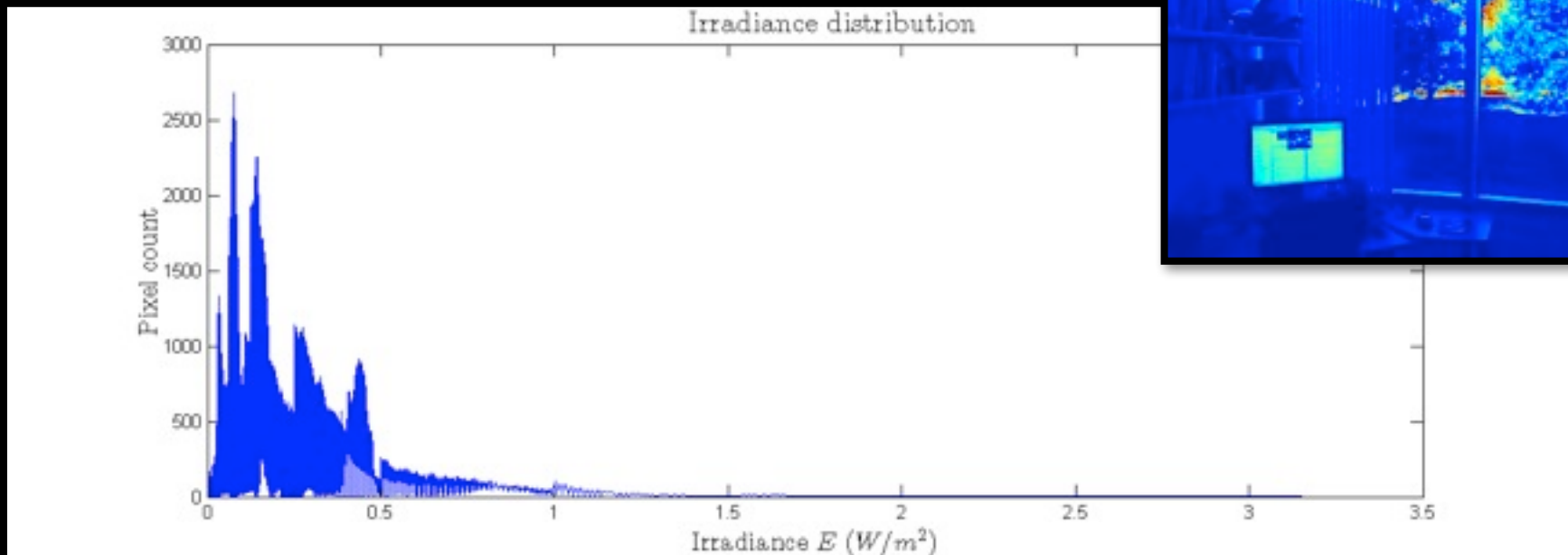


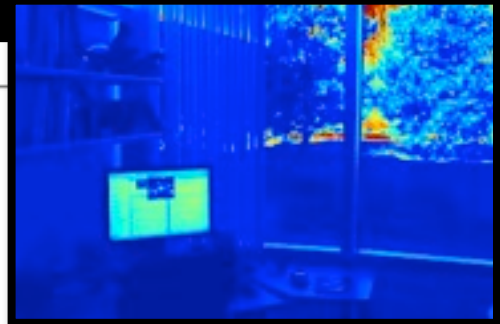
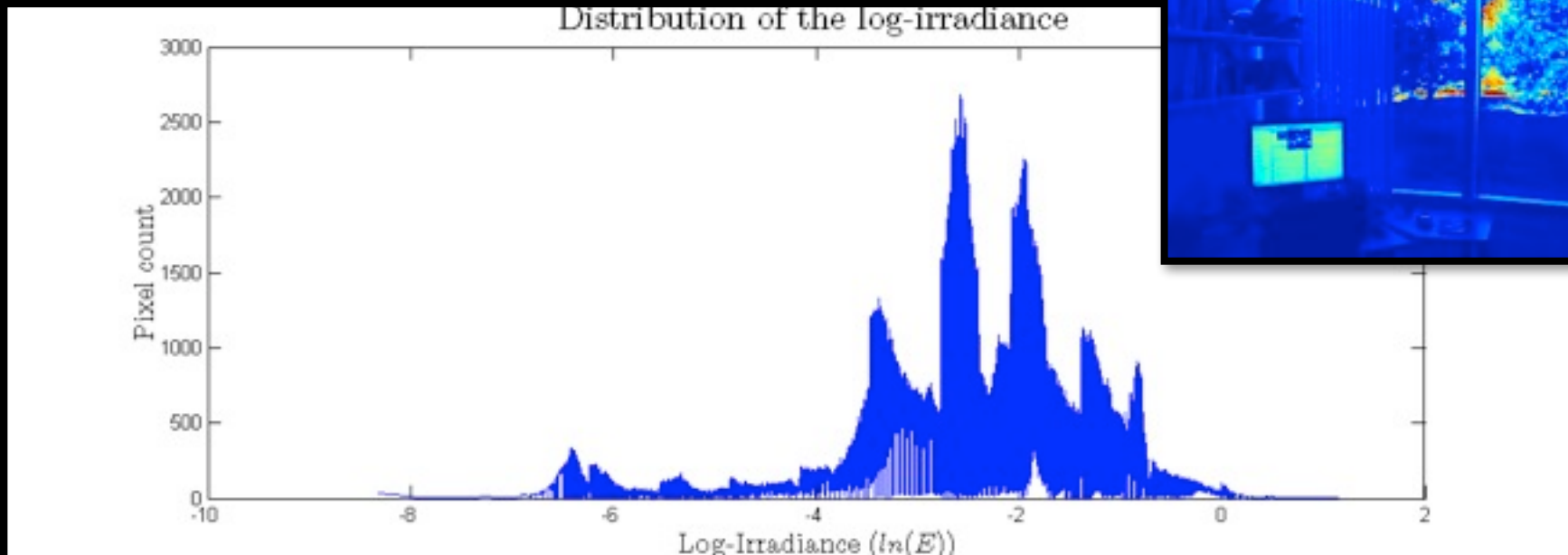
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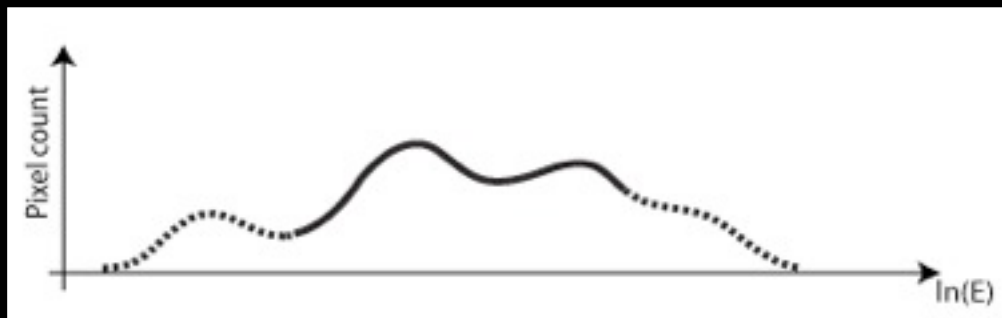


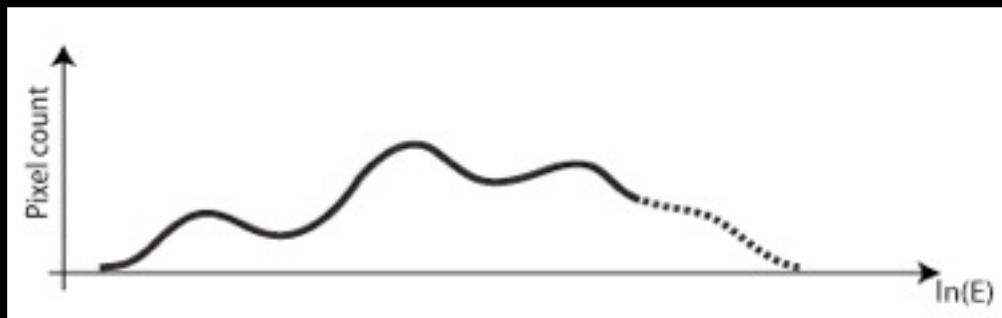


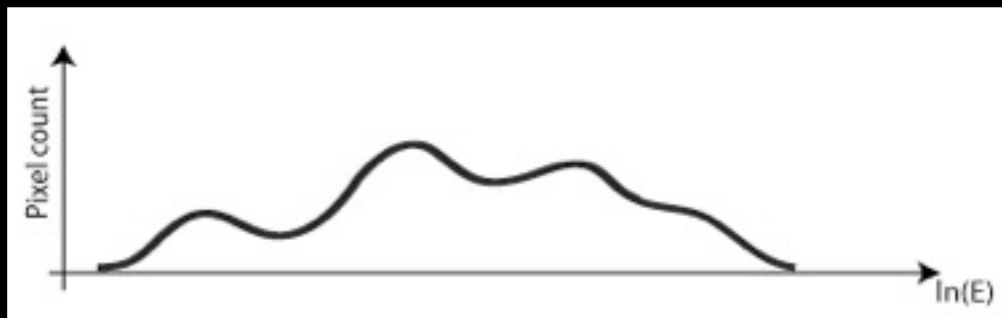


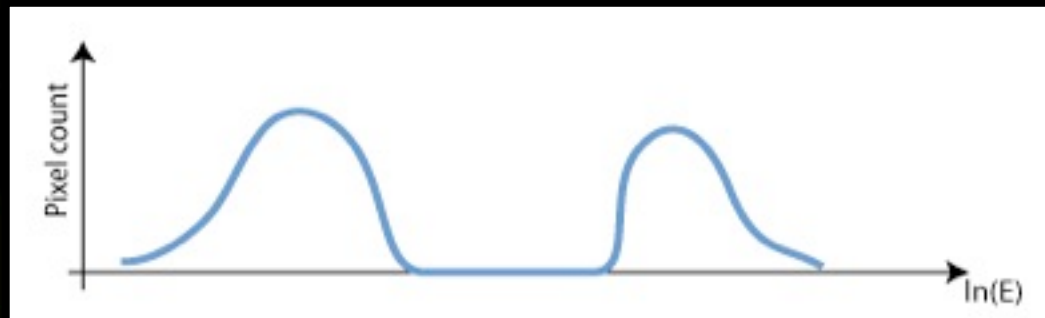


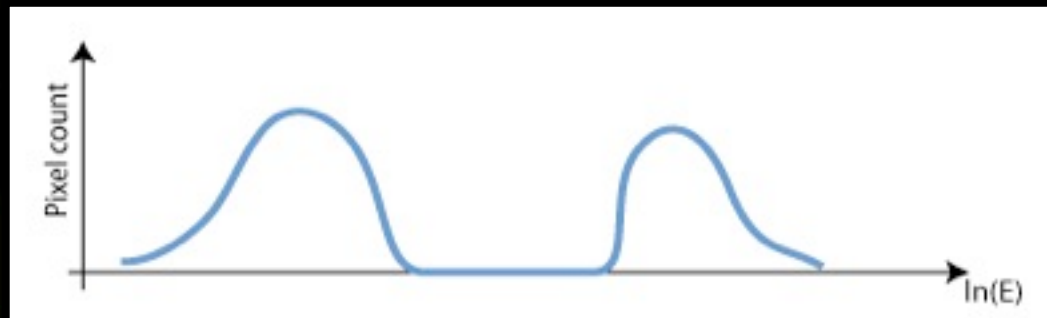








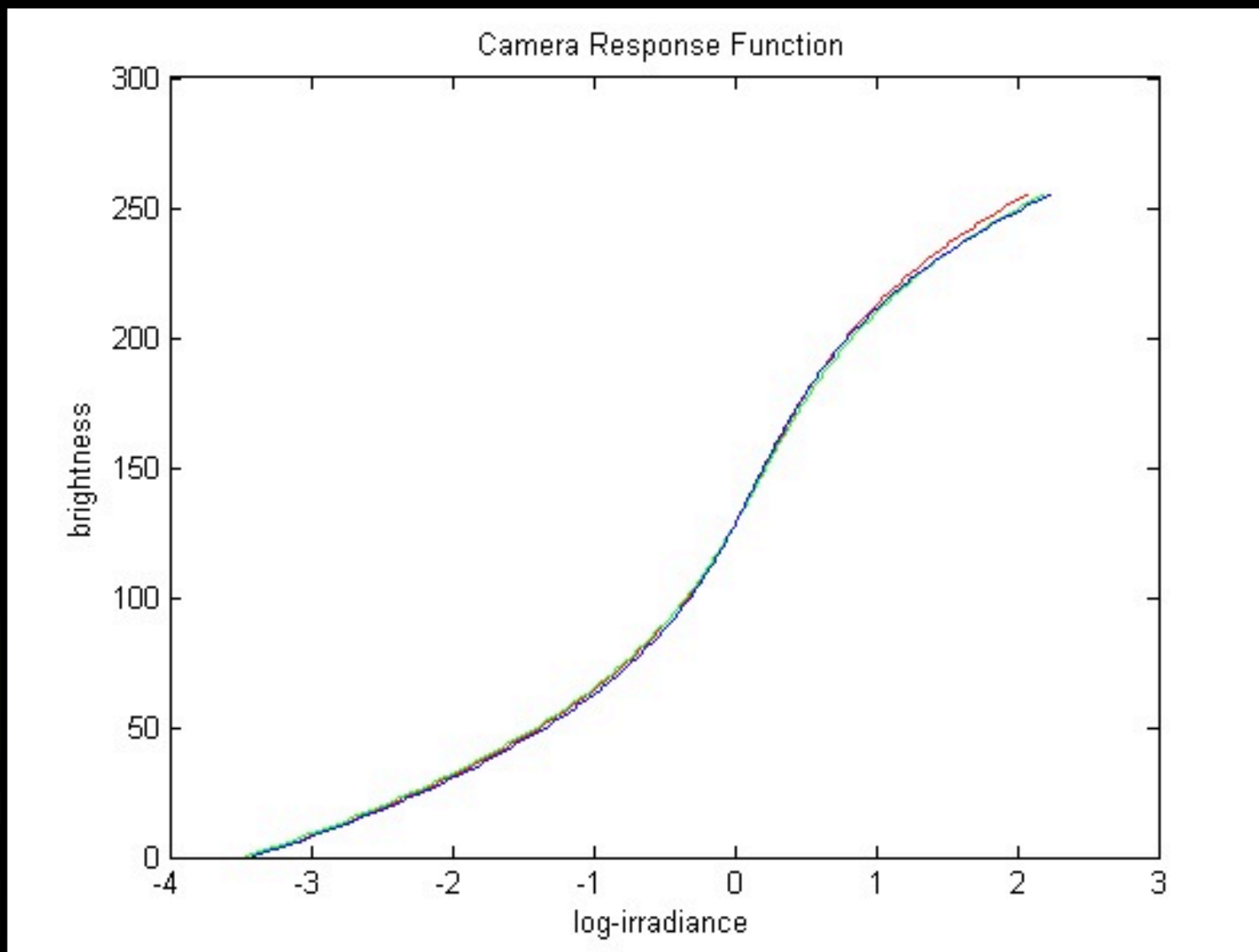


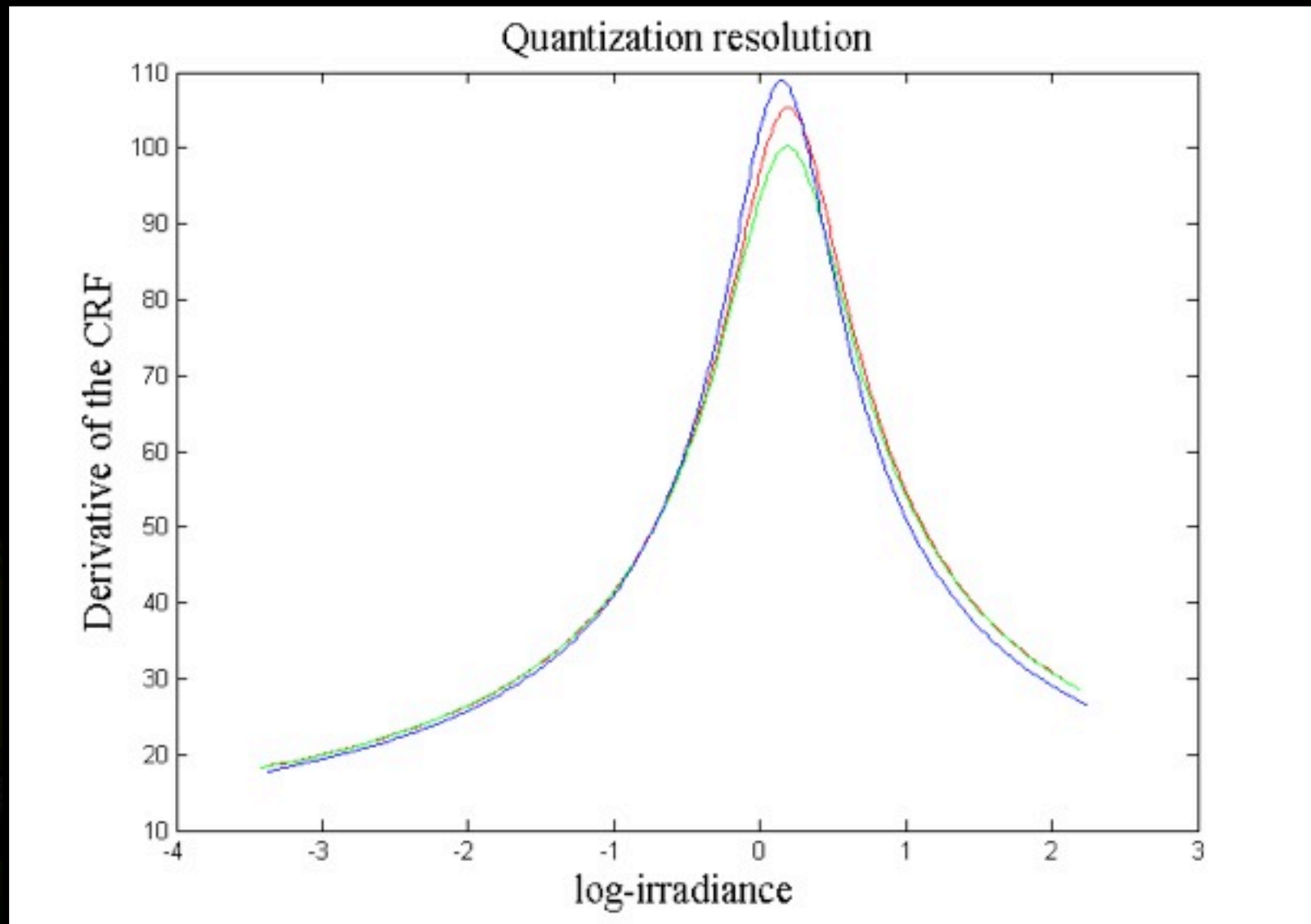


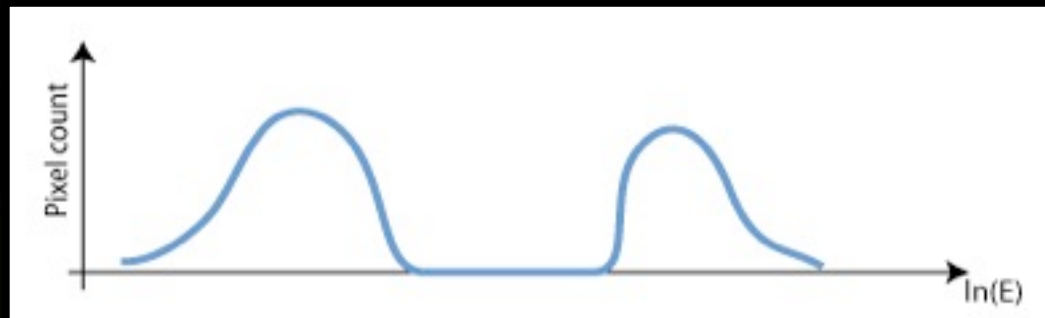
Δt_2

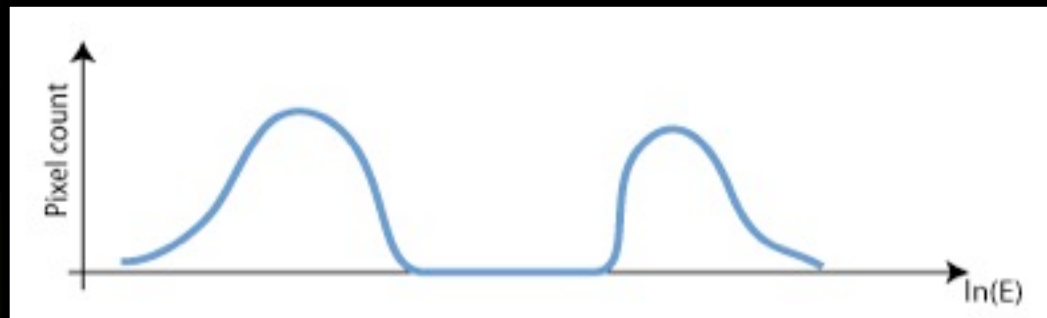


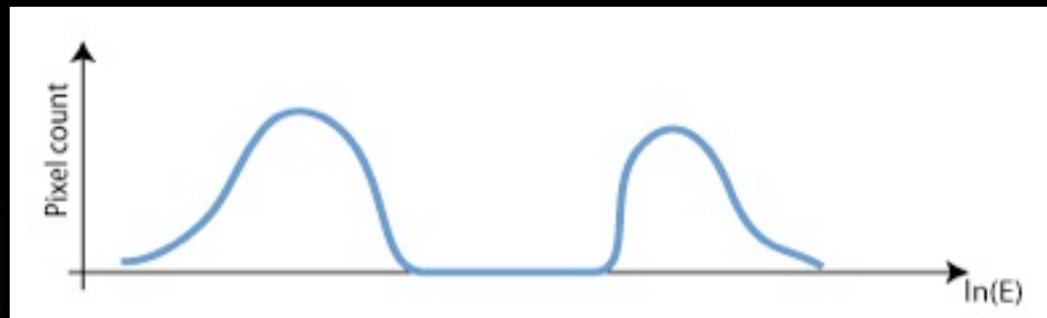
Δt_3









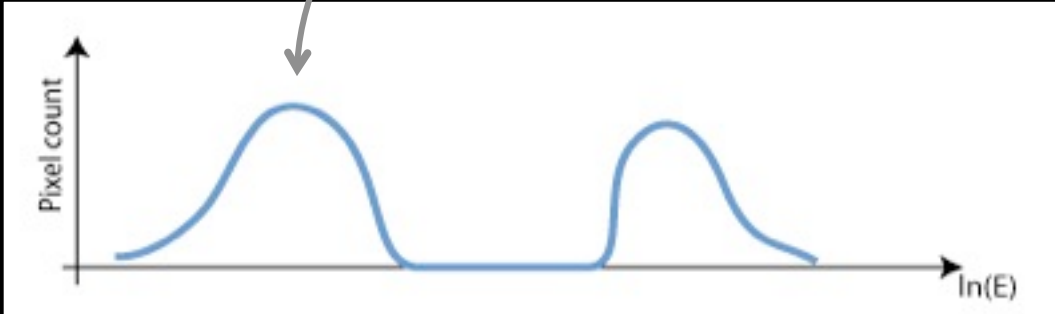


Δt_1

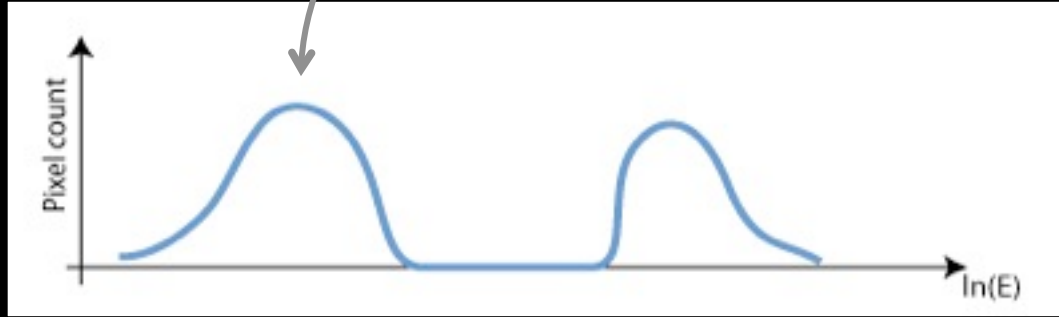


Δt_2

Histogram estimation

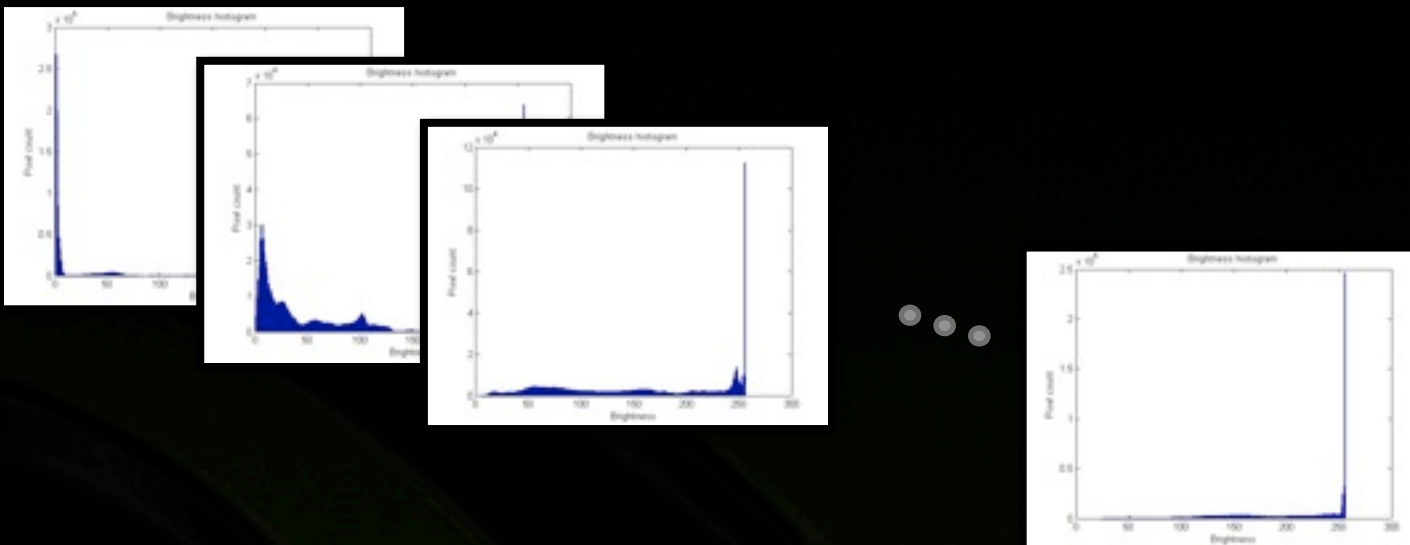


Histogram estimation

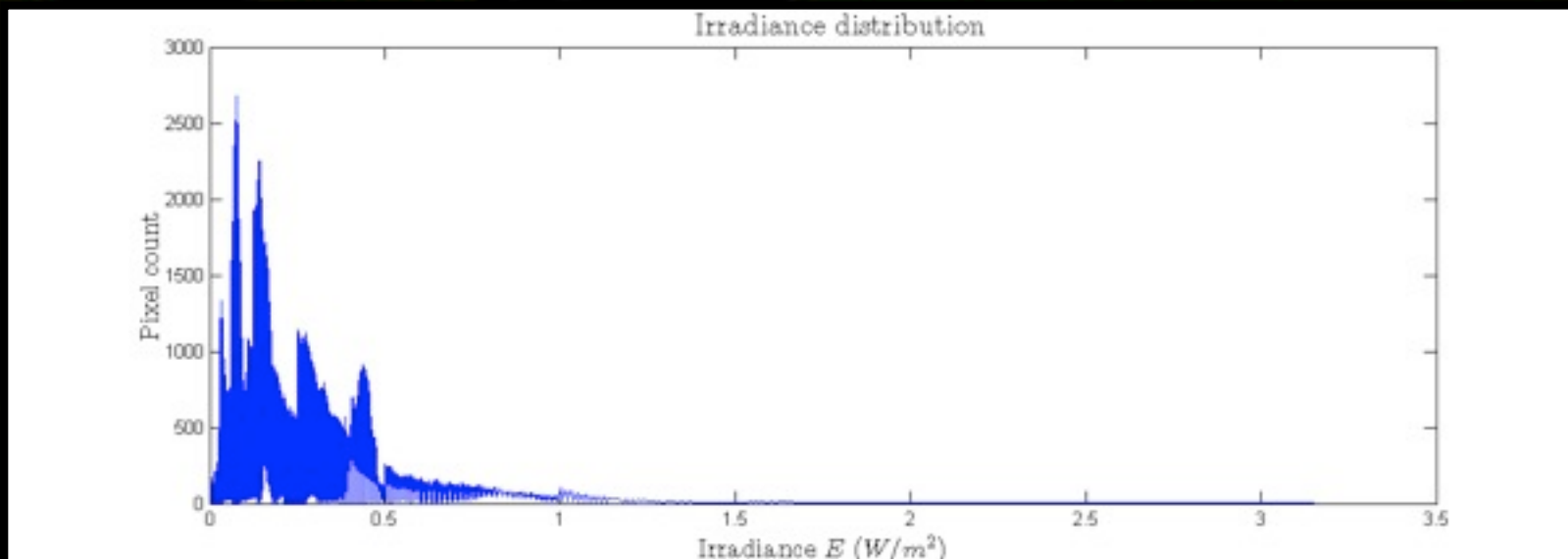
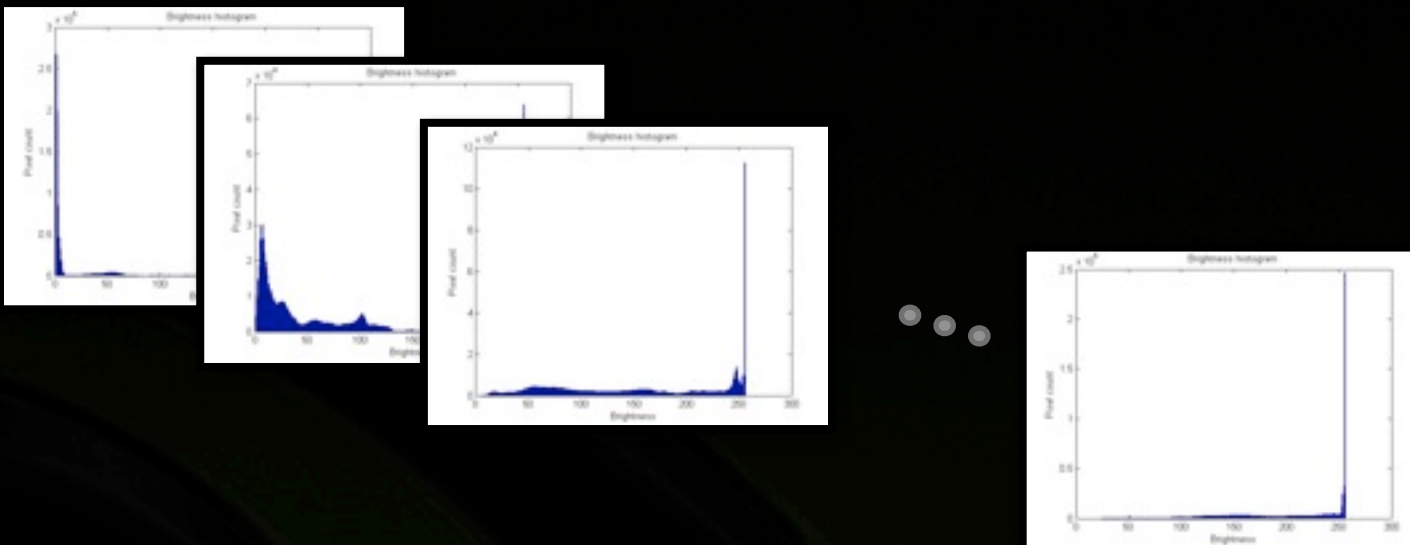


Exposure times

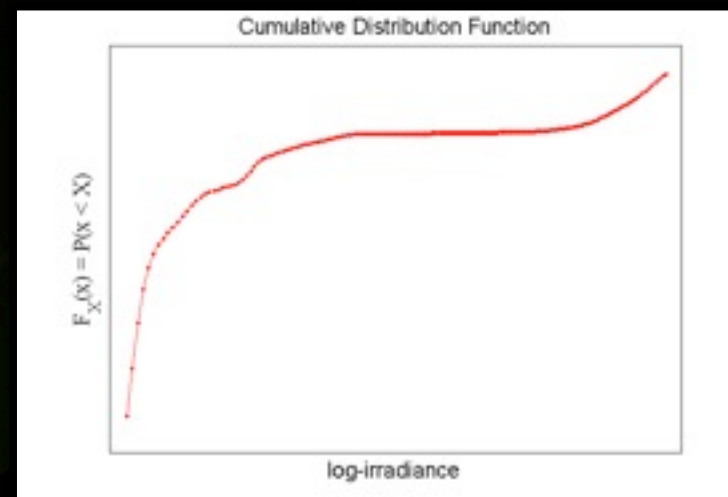
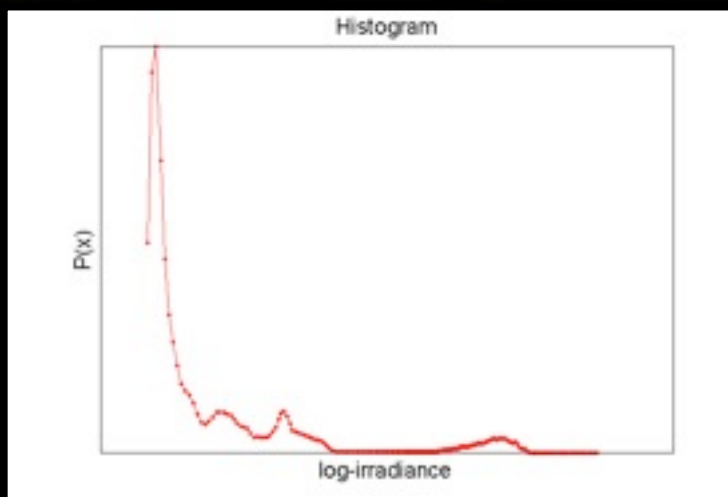
Histogram estimation



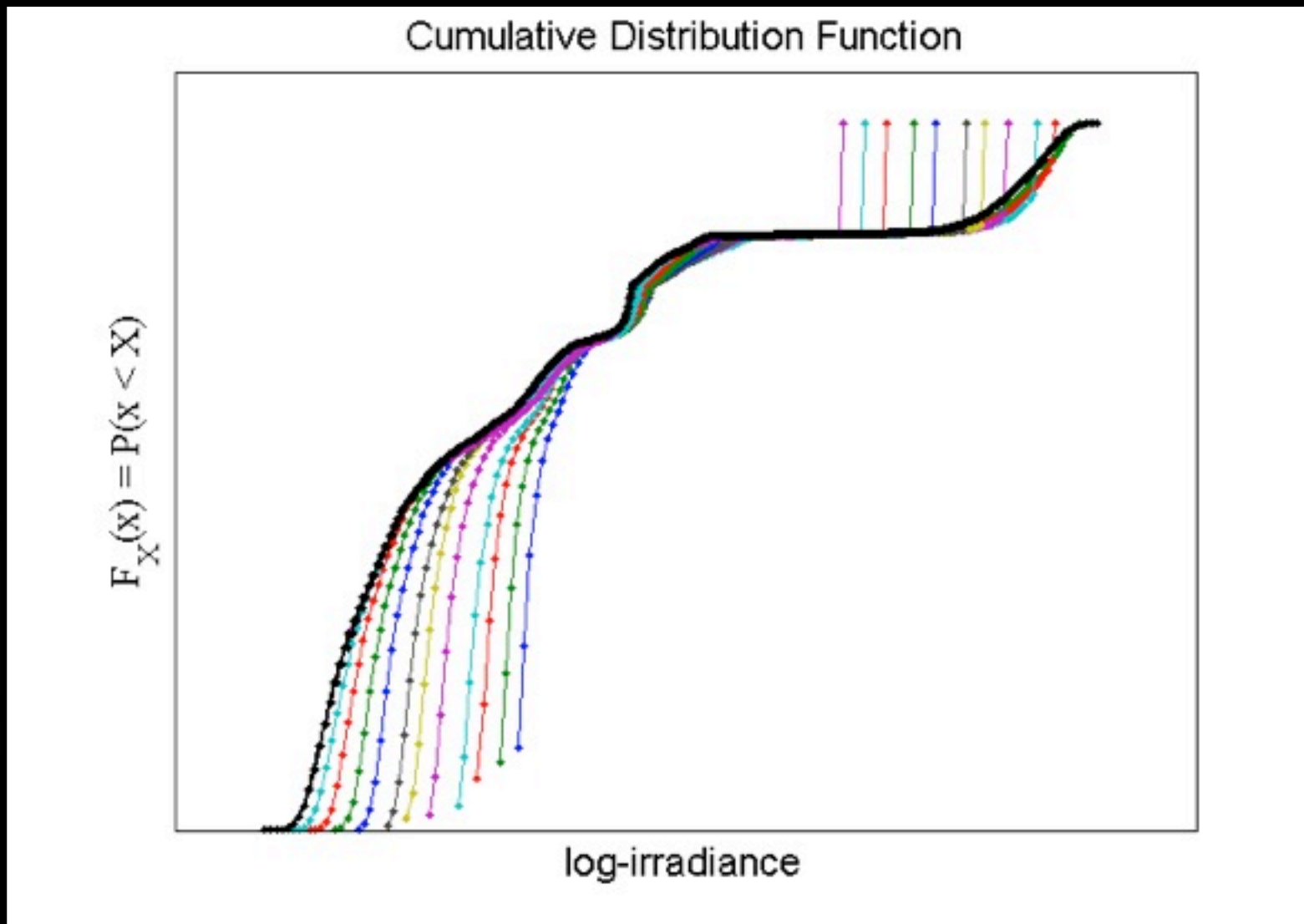
Histogram estimation

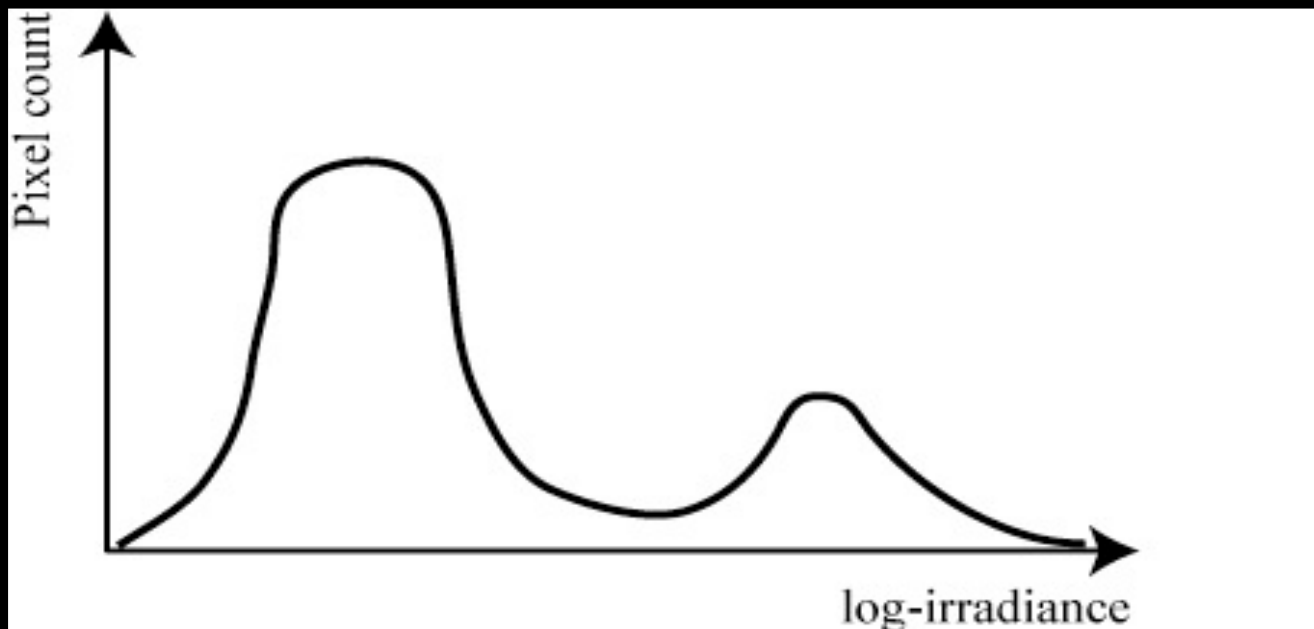


Cumulative Distribution Function (CDF)



Histogram estimation





$$t = \operatorname{argmax}_t \sum_i P_{\tilde{E}}(\tilde{E}_i) \frac{[f'(\tilde{E}_i + \tilde{t})]^2}{[a'e^{\tilde{E}_i + \tilde{t}} + b'] [f'(\tilde{E}_i + \tilde{t})]^2 + \frac{\Delta^2}{12}}$$

HDR histogram

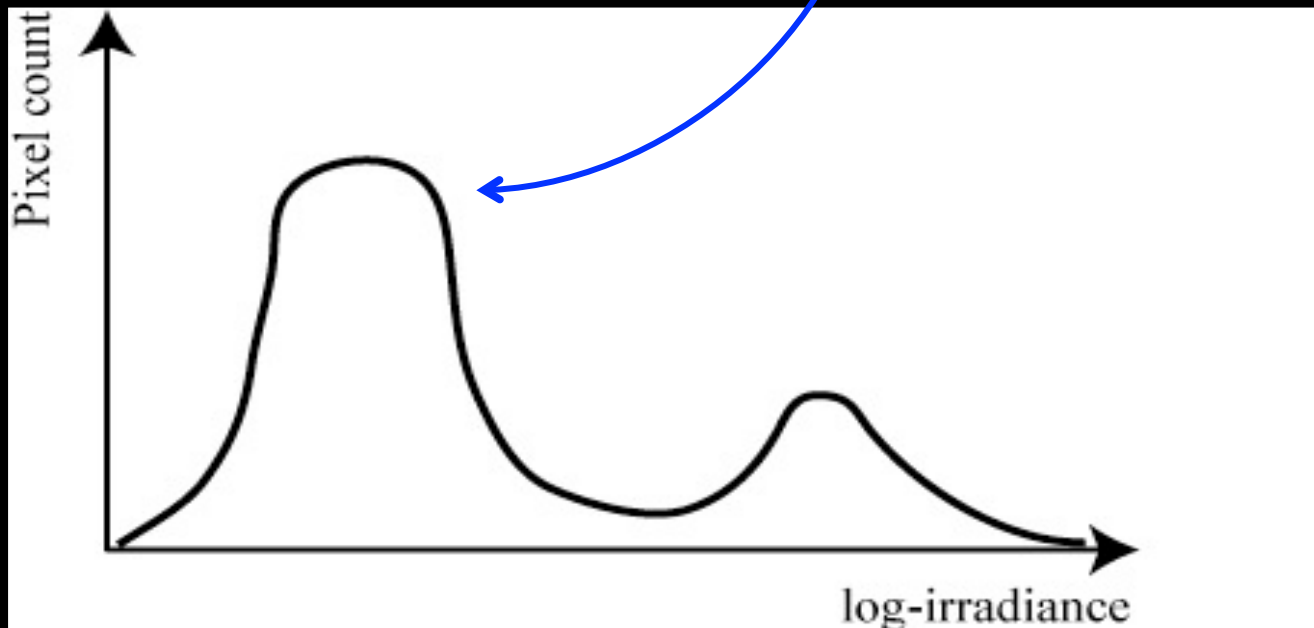
$$t = \operatorname{argmax}_t \sum_i P_{\tilde{E}}(\tilde{E}_i) \frac{[f'(\tilde{E}_i + \tilde{t})]^2}{[a'e^{\tilde{E}_i + \tilde{t}} + b'] [f'(\tilde{E}_i + \tilde{t})]^2 + \frac{\Delta^2}{12}}$$

Noise

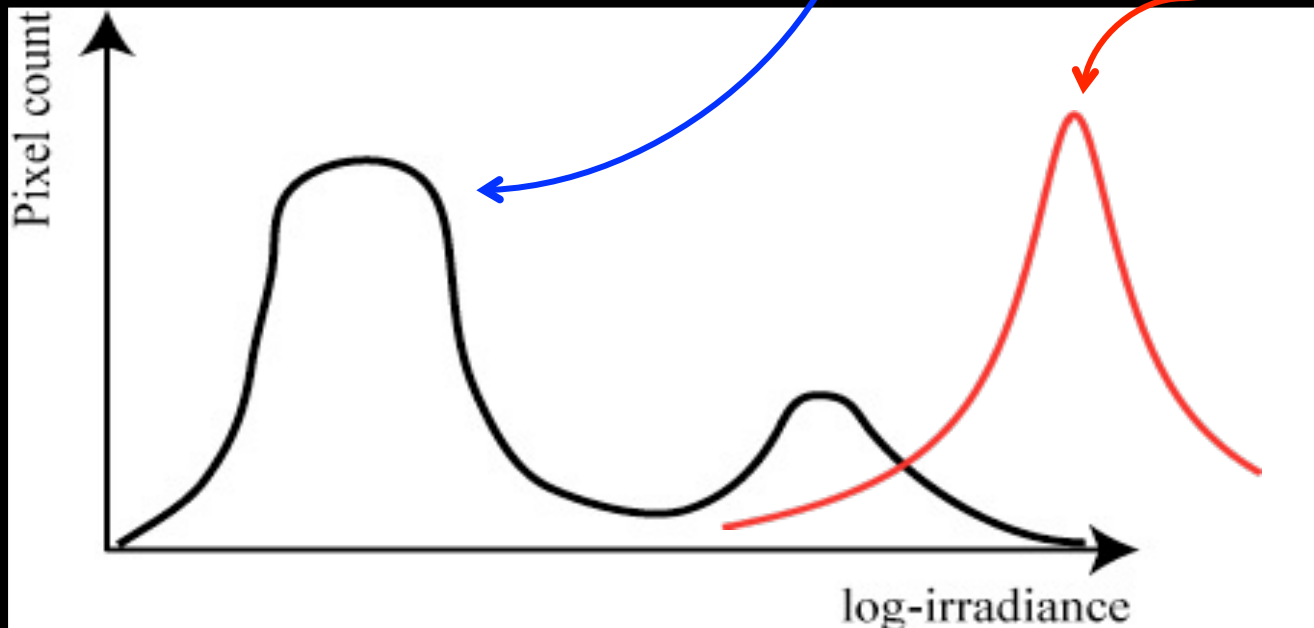
CRF

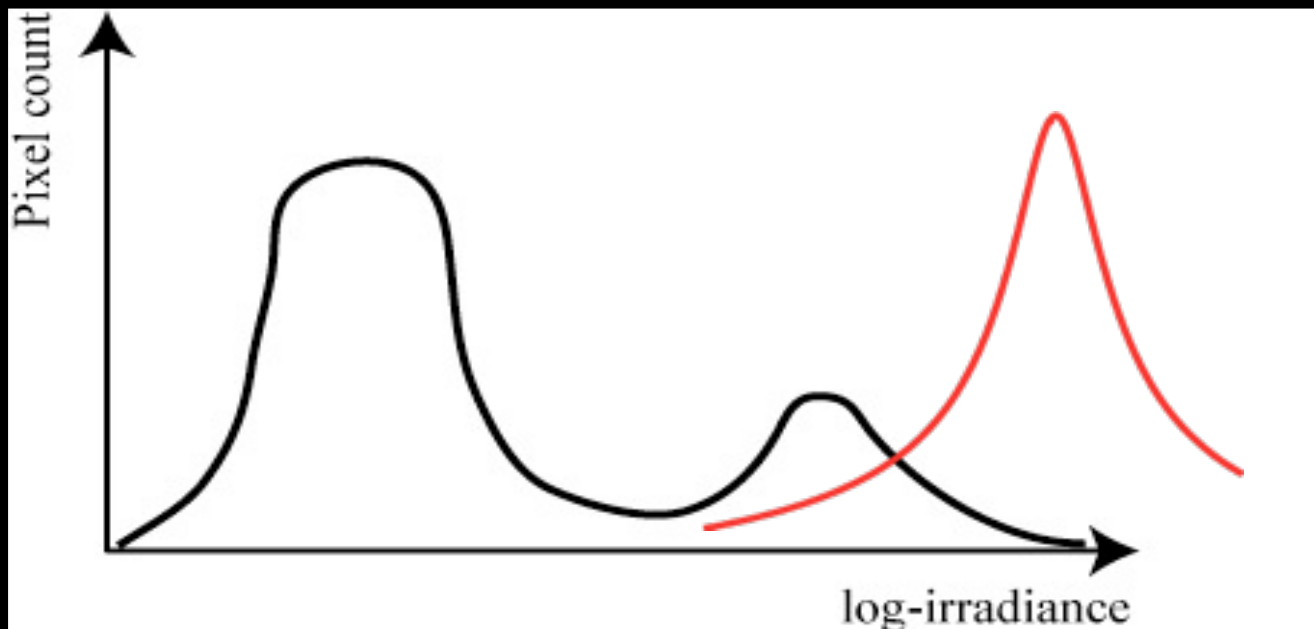
Quantization

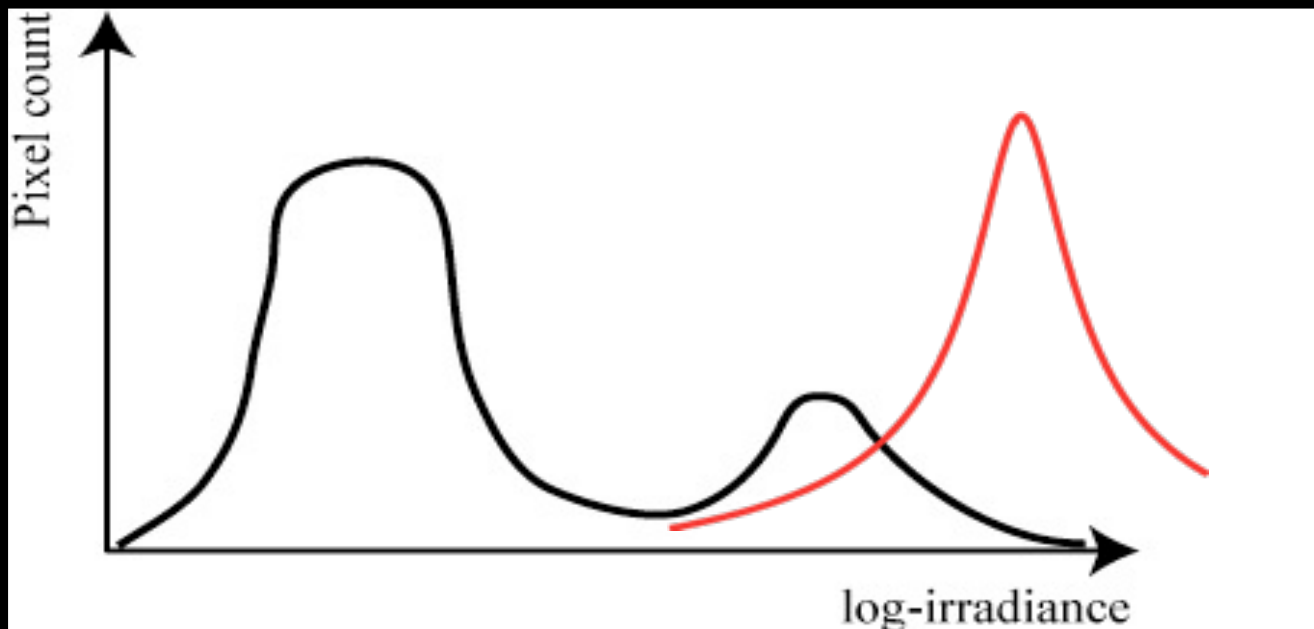
$$t = \operatorname{argmax}_t \sum_i P_{\tilde{E}}(\tilde{E}_i) \frac{[f'(\tilde{E}_i + \tilde{t})]^2}{[a'e^{\tilde{E}_i + \tilde{t}} + b'] [f'(\tilde{E}_i + \tilde{t})]^2 + \frac{\Delta^2}{12}}$$

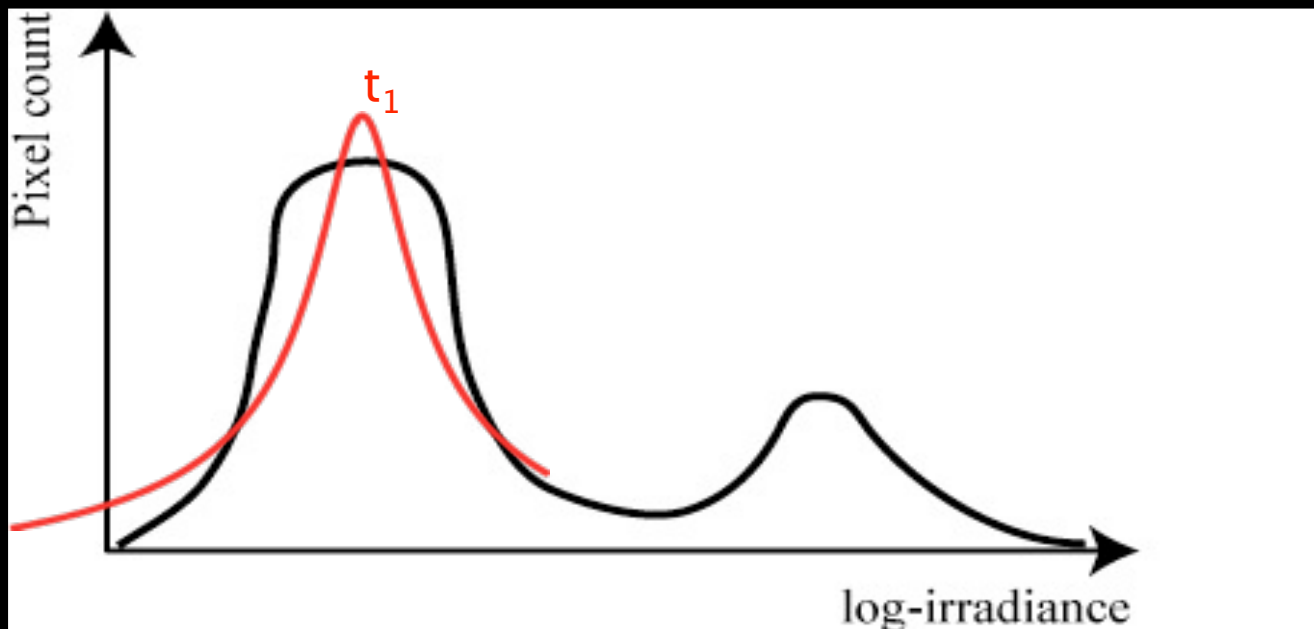


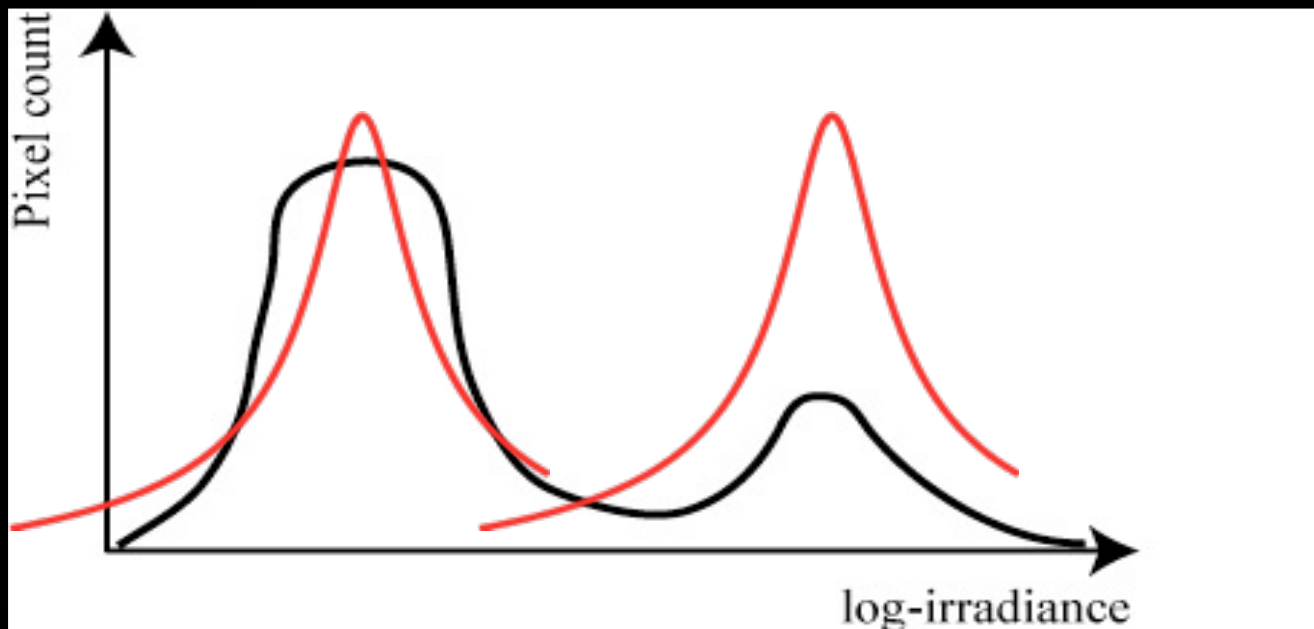
$$t = \operatorname{argmax}_t \sum_i P_{\tilde{E}}(\tilde{E}_i) \frac{[f'(\tilde{E}_i + \tilde{t})]^2}{[a'e^{\tilde{E}_i + \tilde{t}} + b'] [f'(\tilde{E}_i + \tilde{t})]^2 + \frac{\Delta^2}{12}}$$

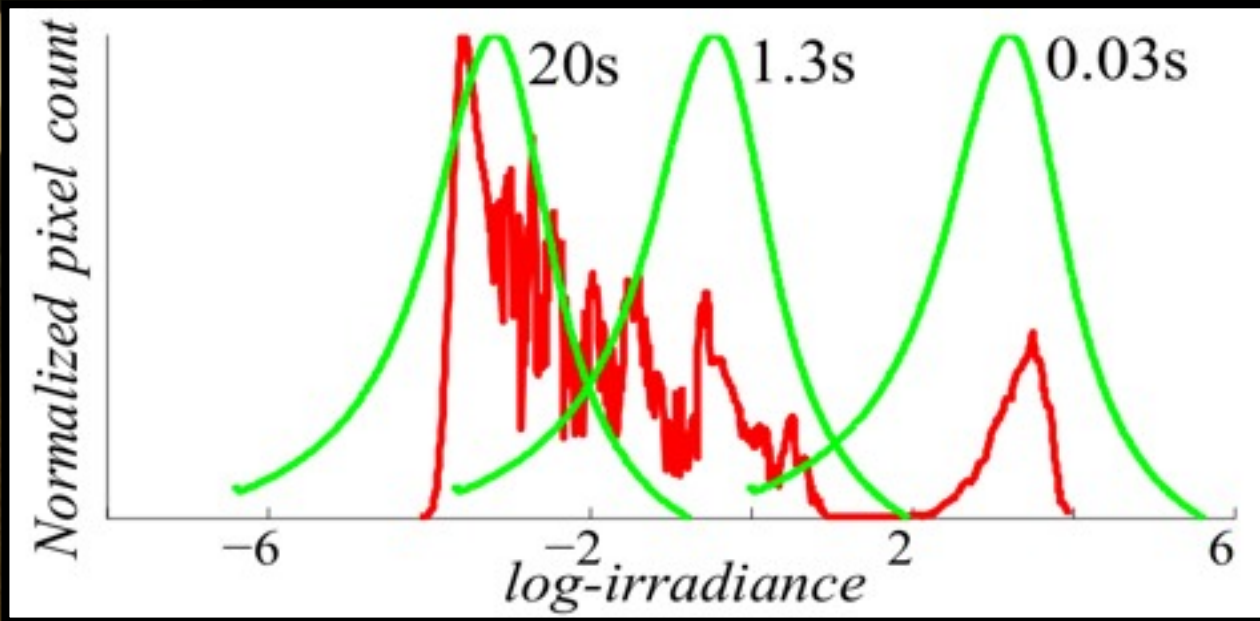
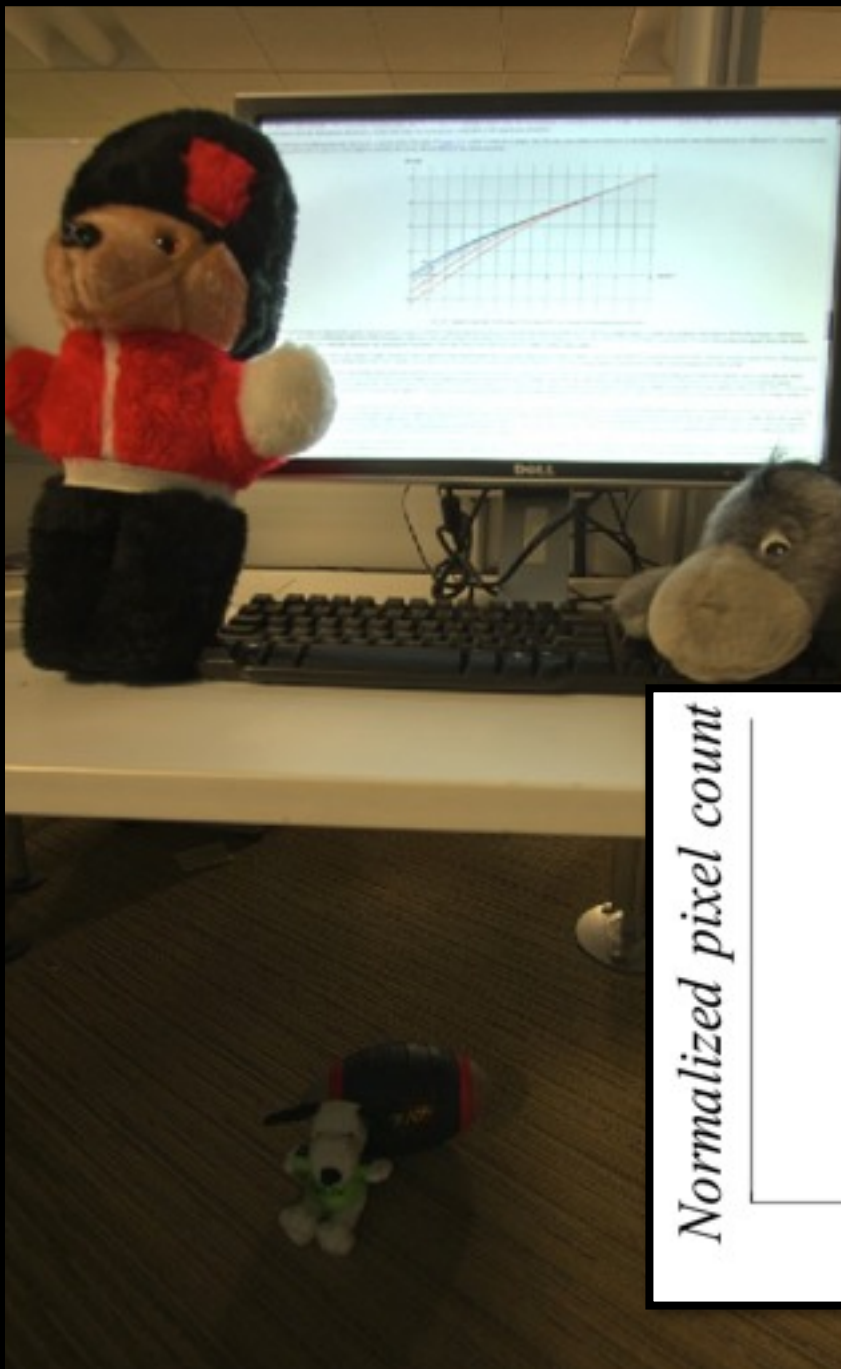


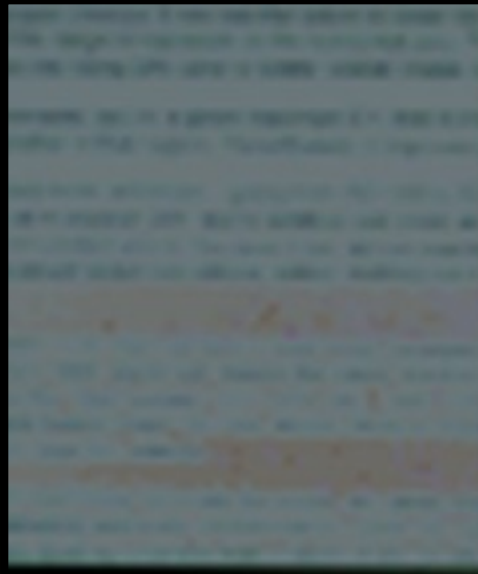
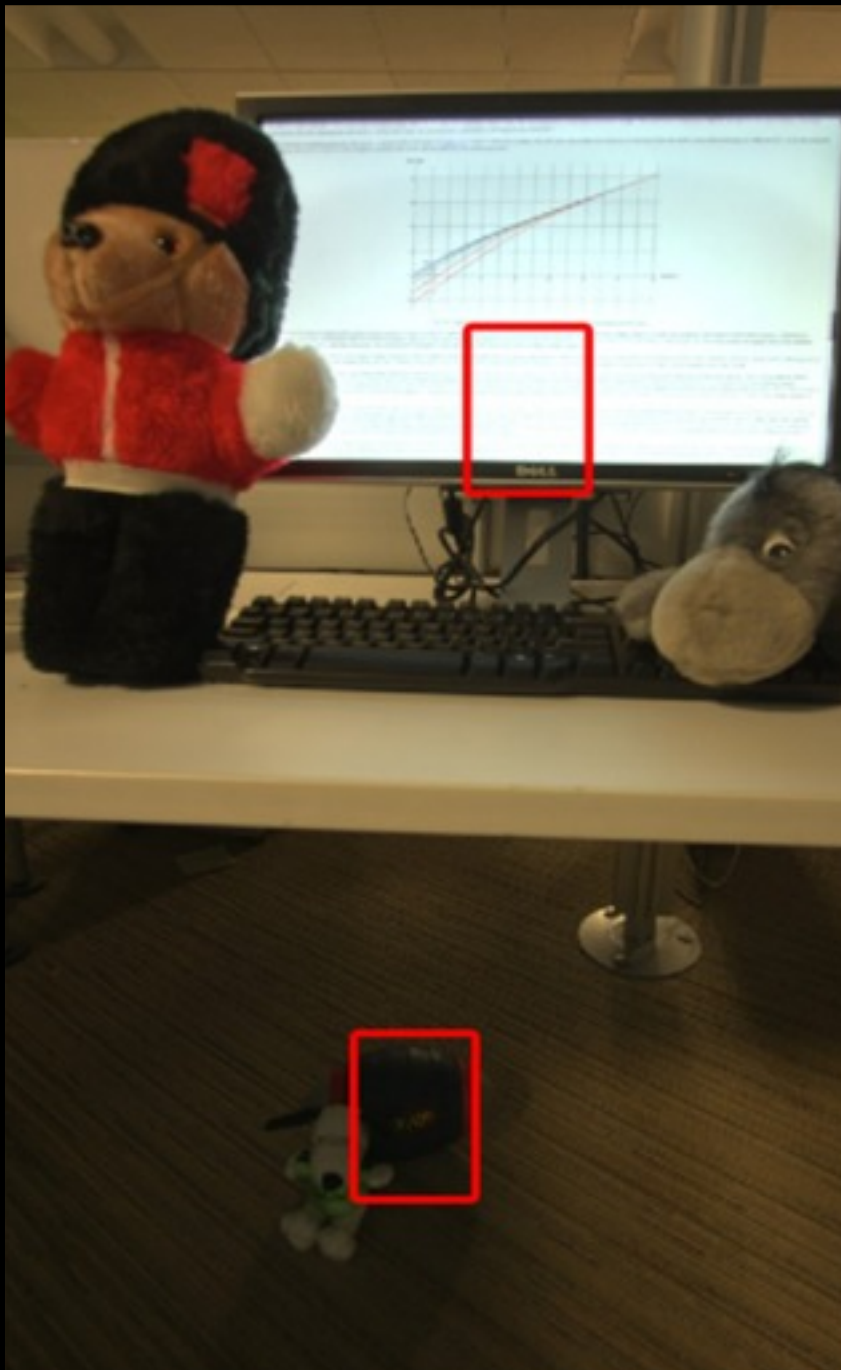




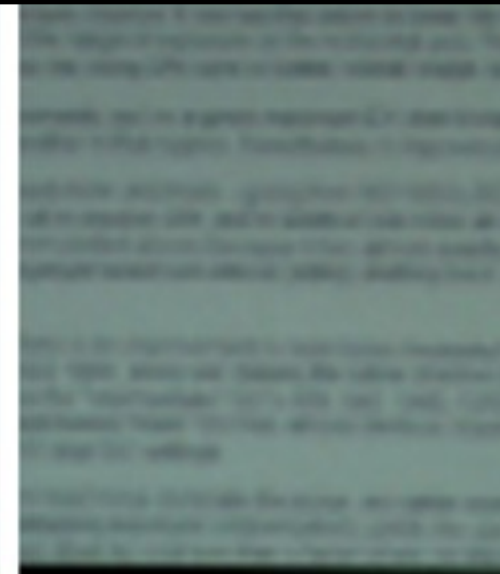






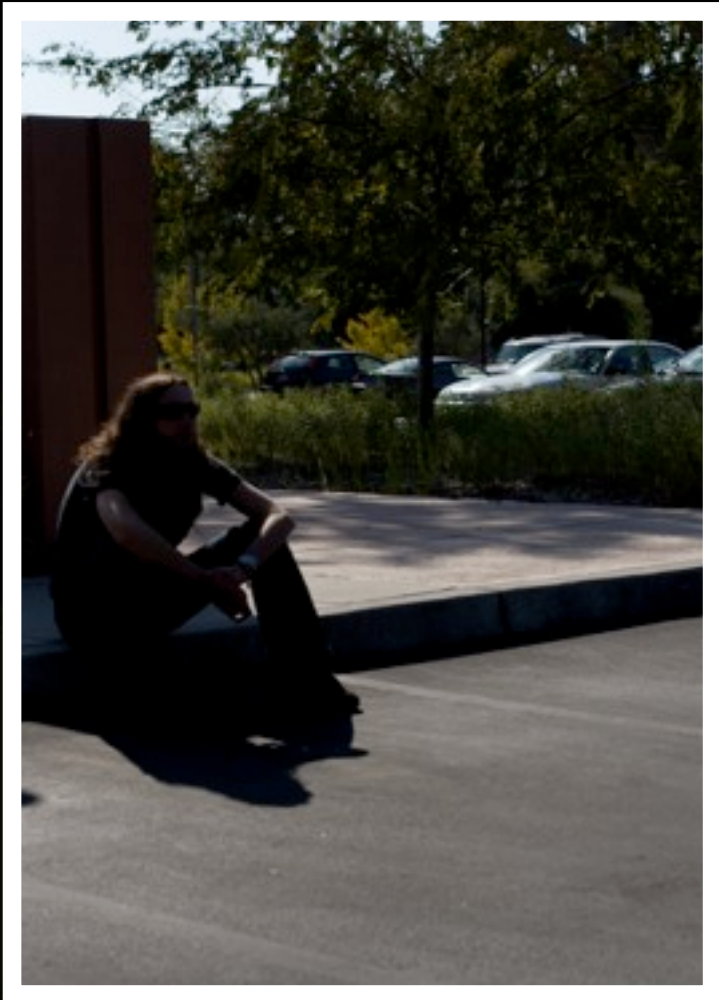


Range, 5 images
15dB

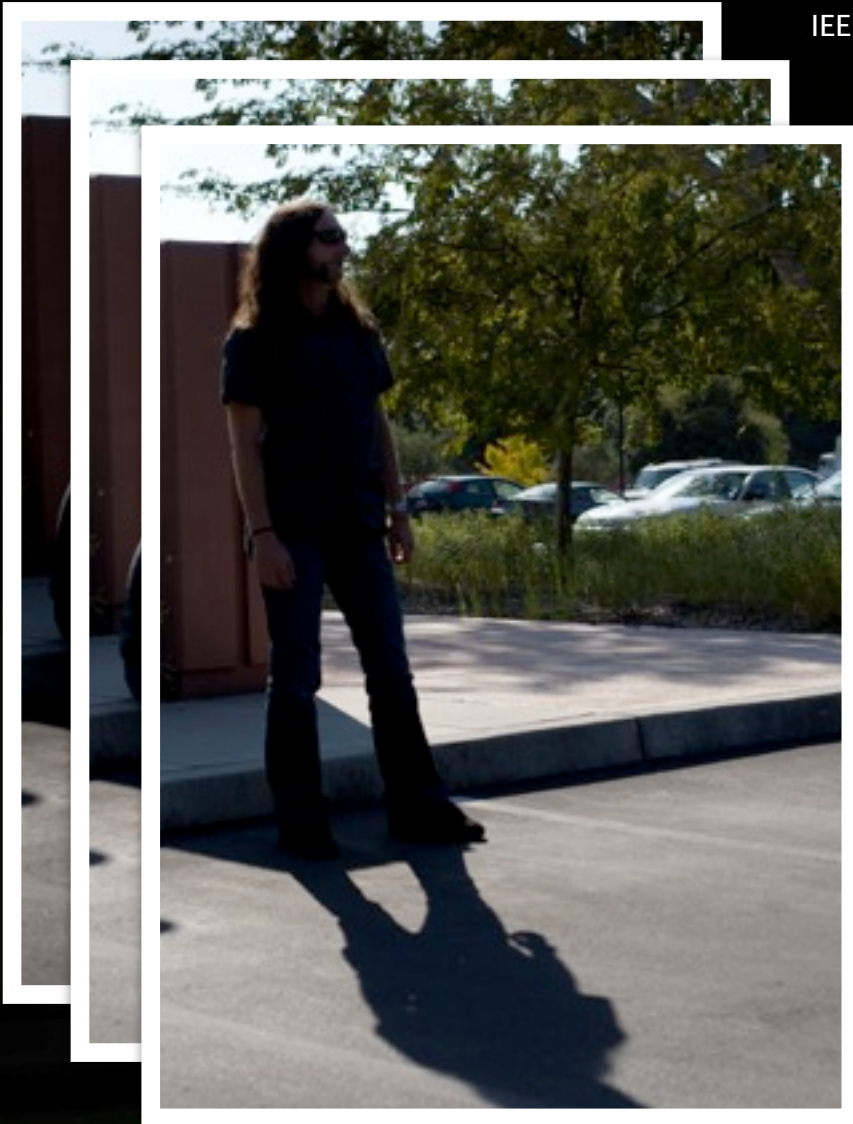


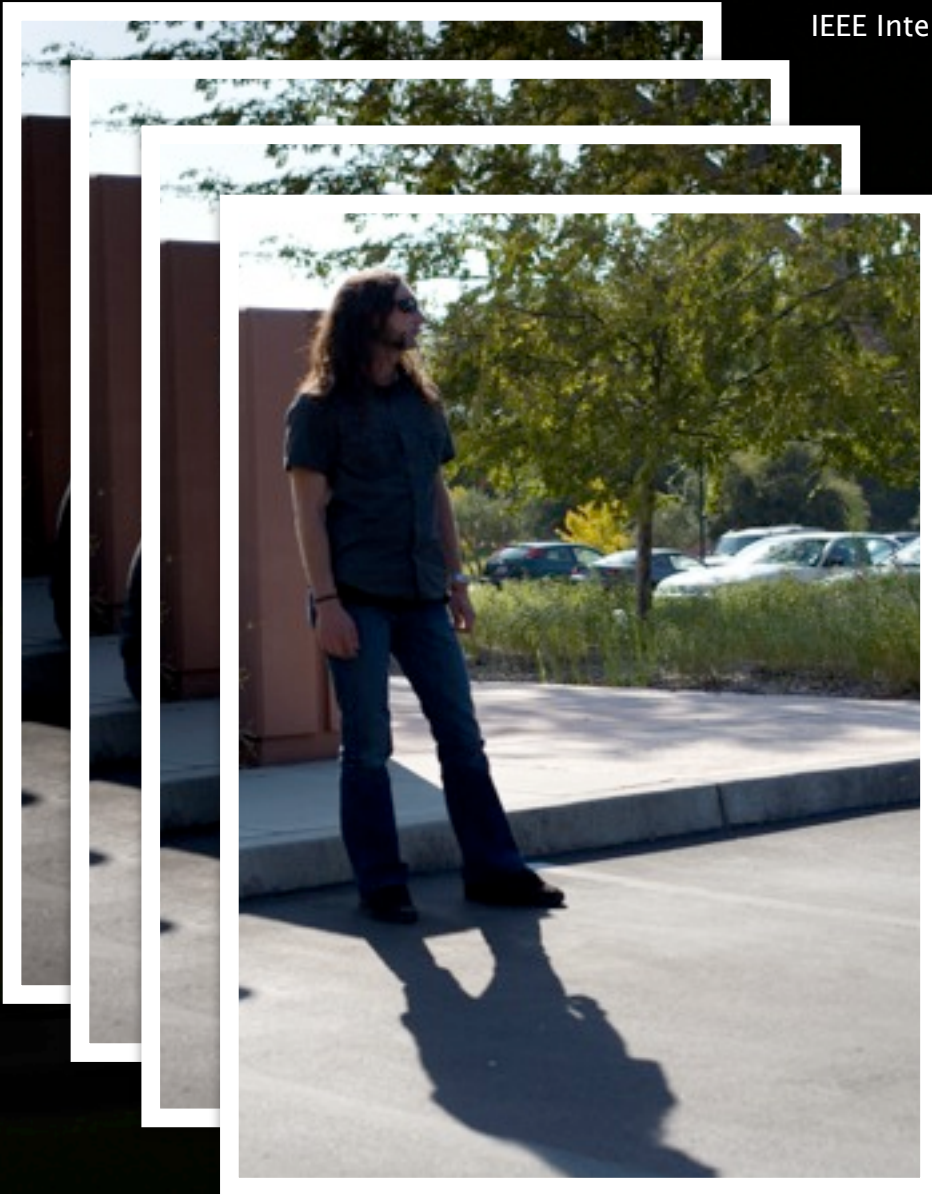
Ours, 3 images
25dB

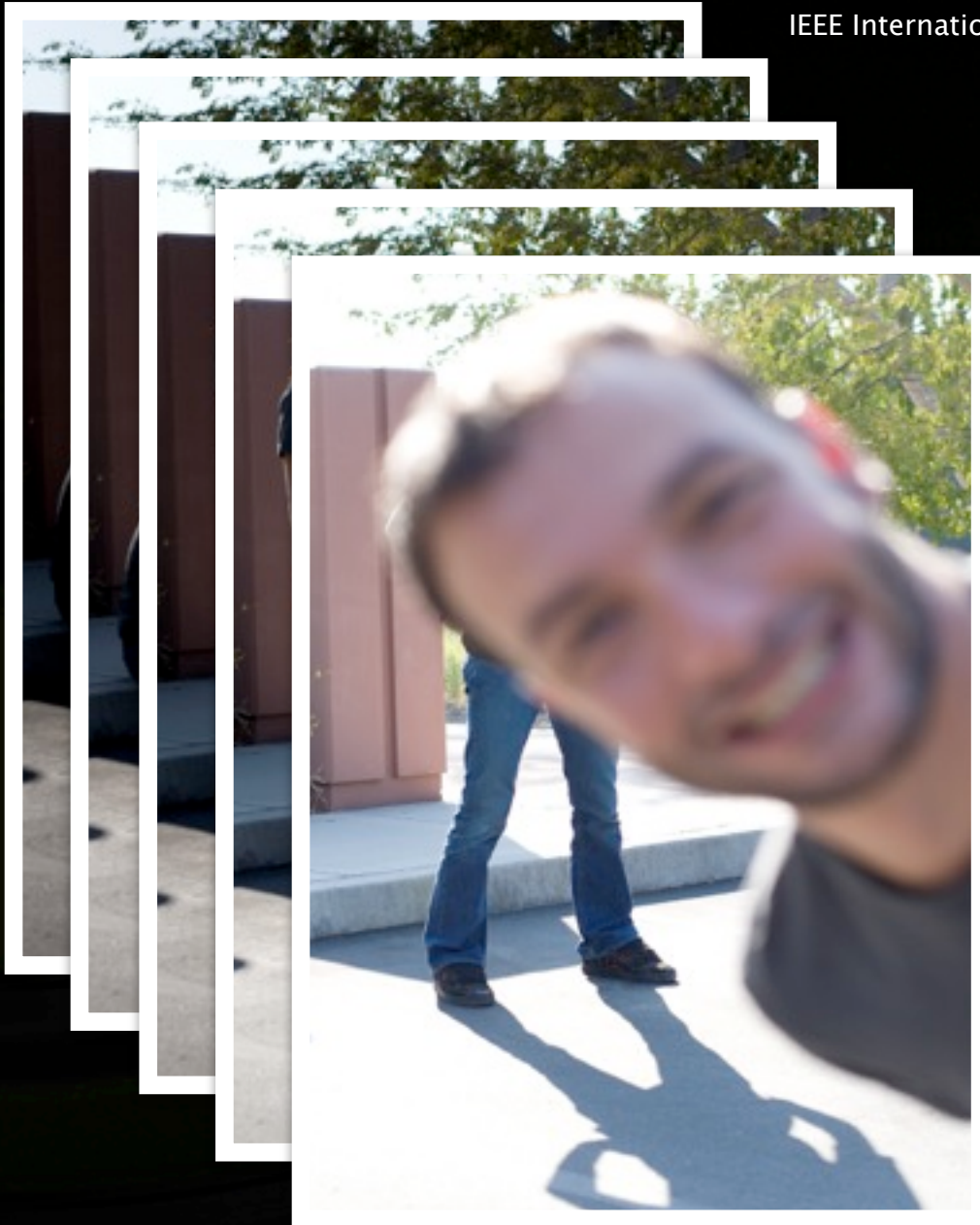


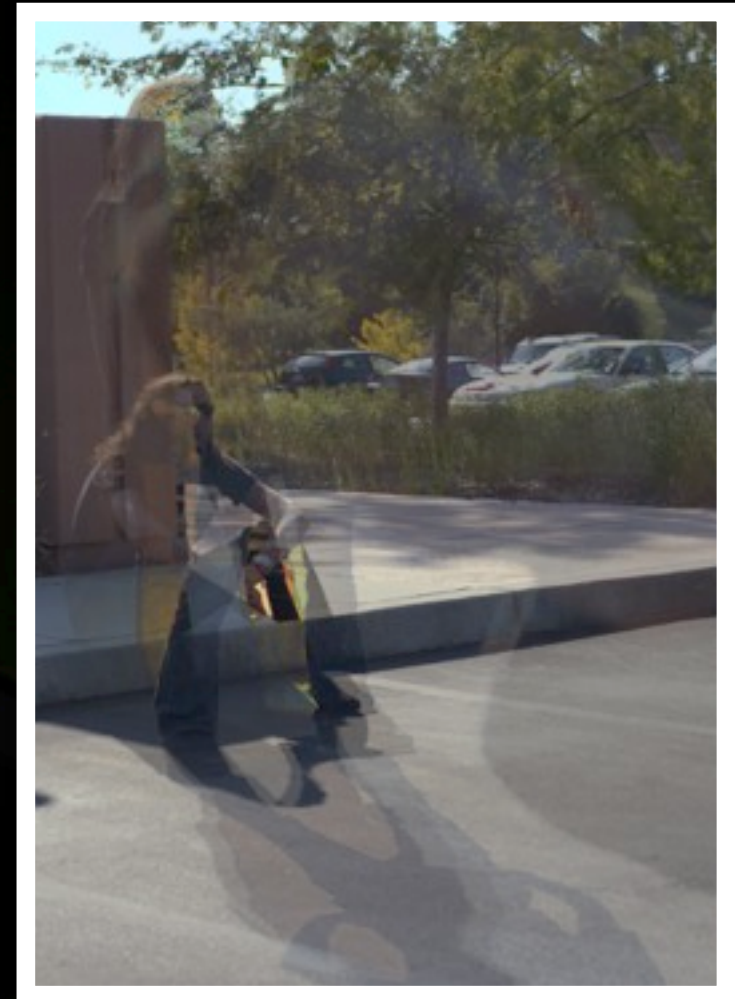
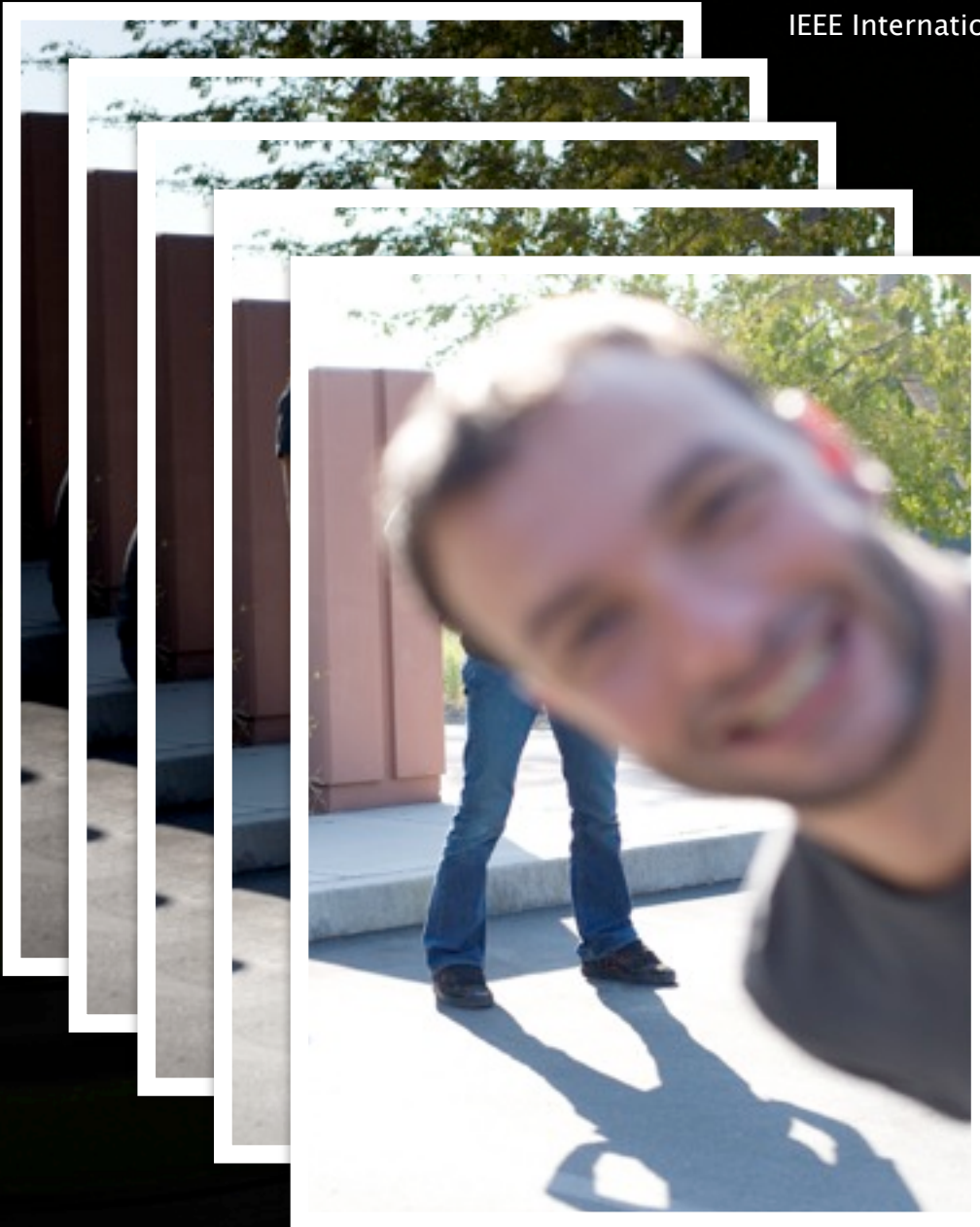




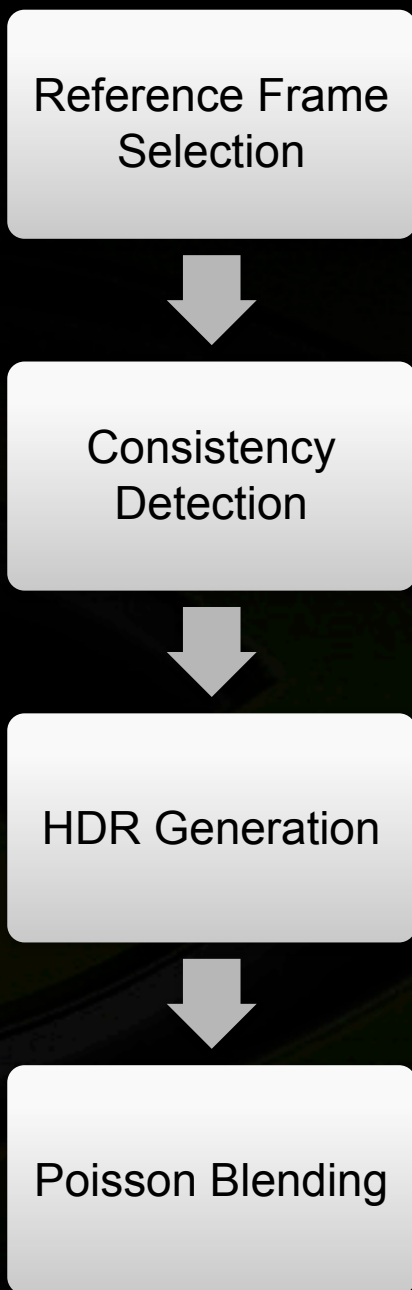






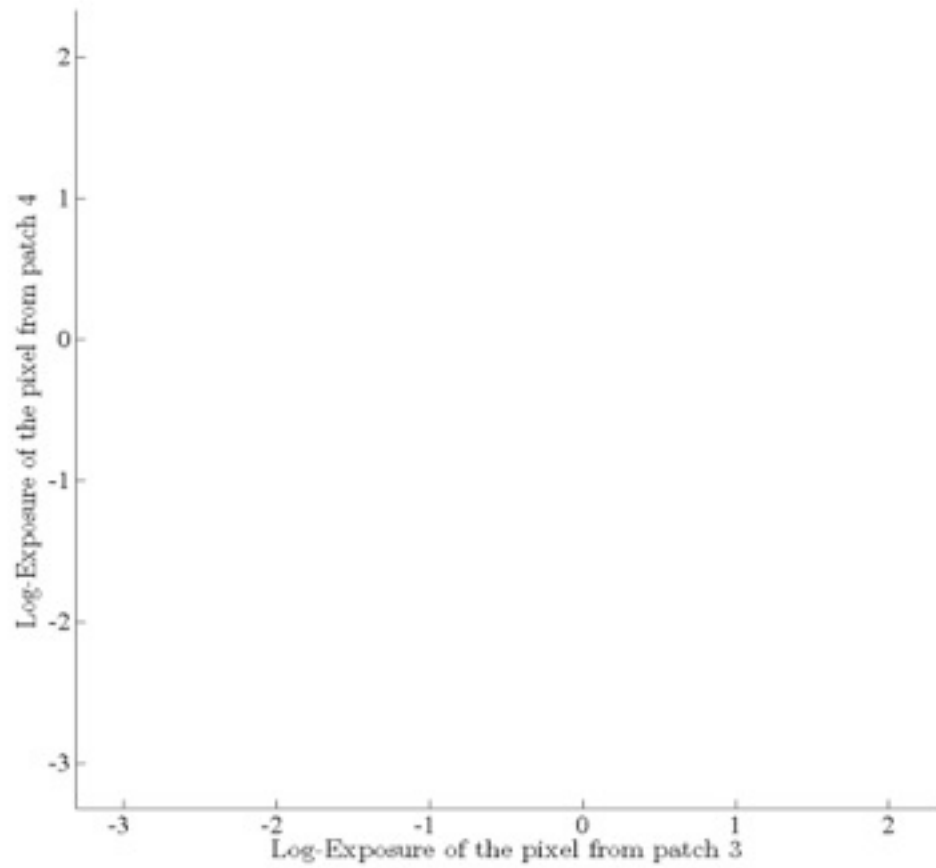
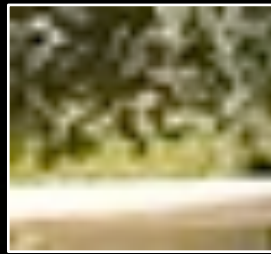


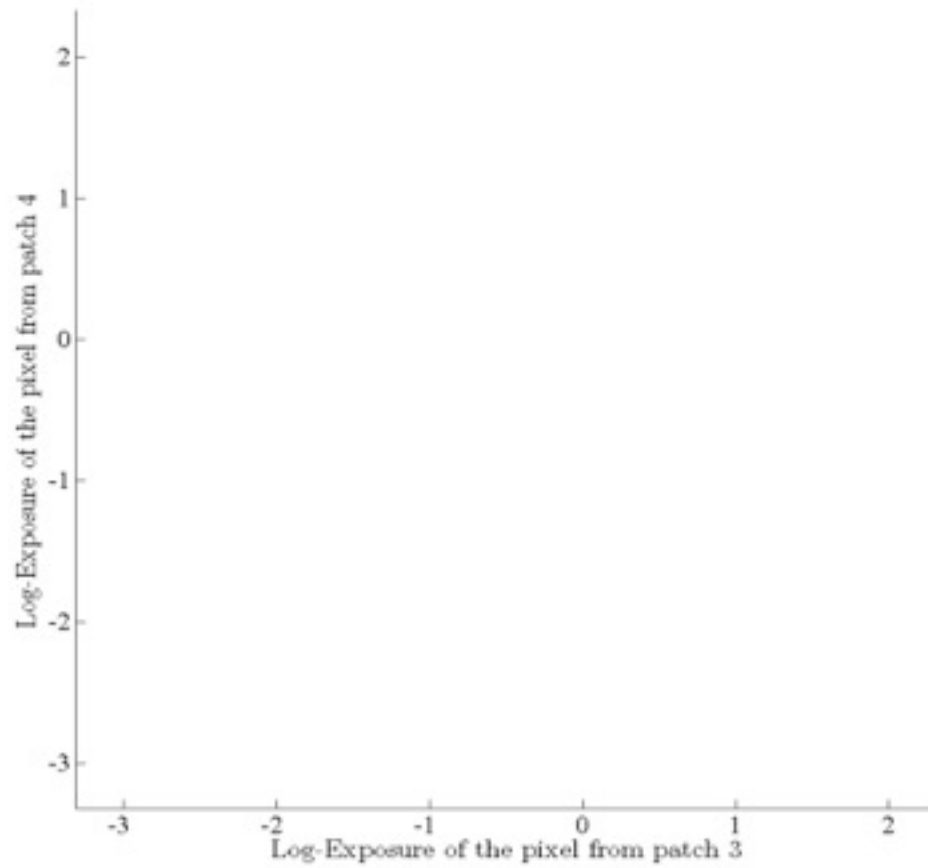
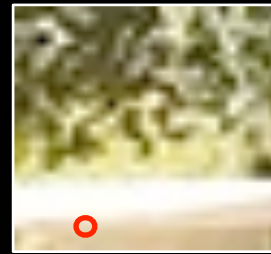


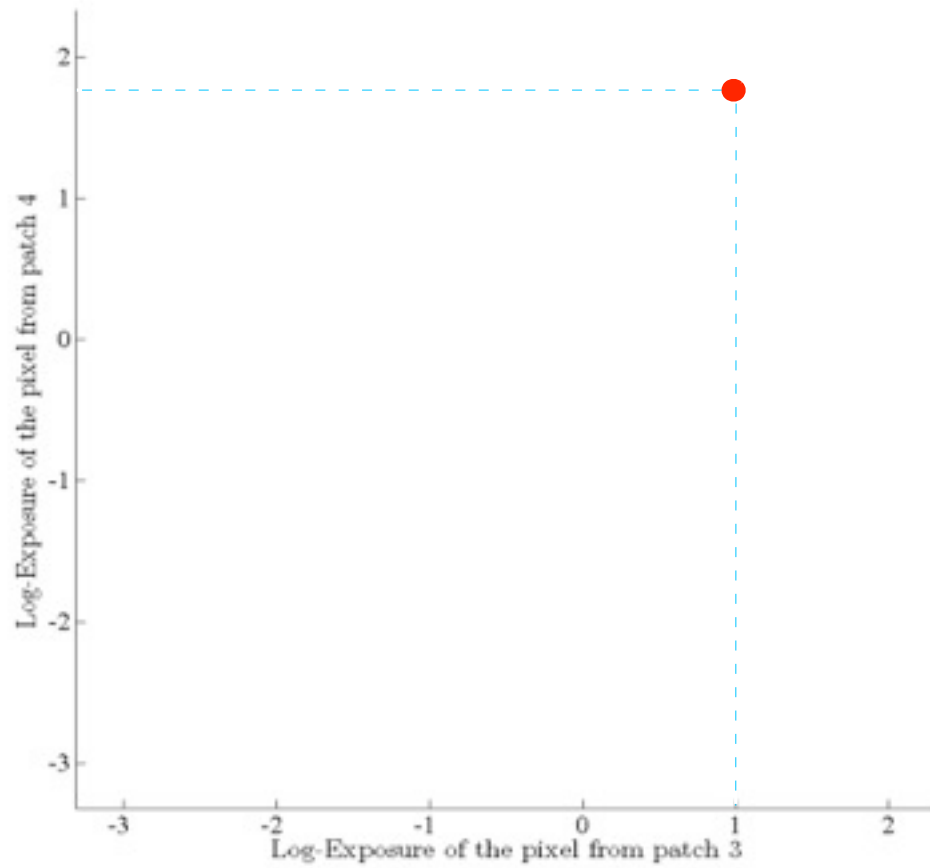
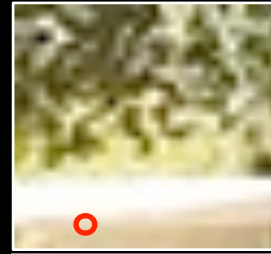
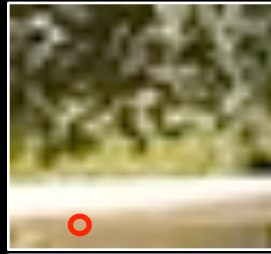


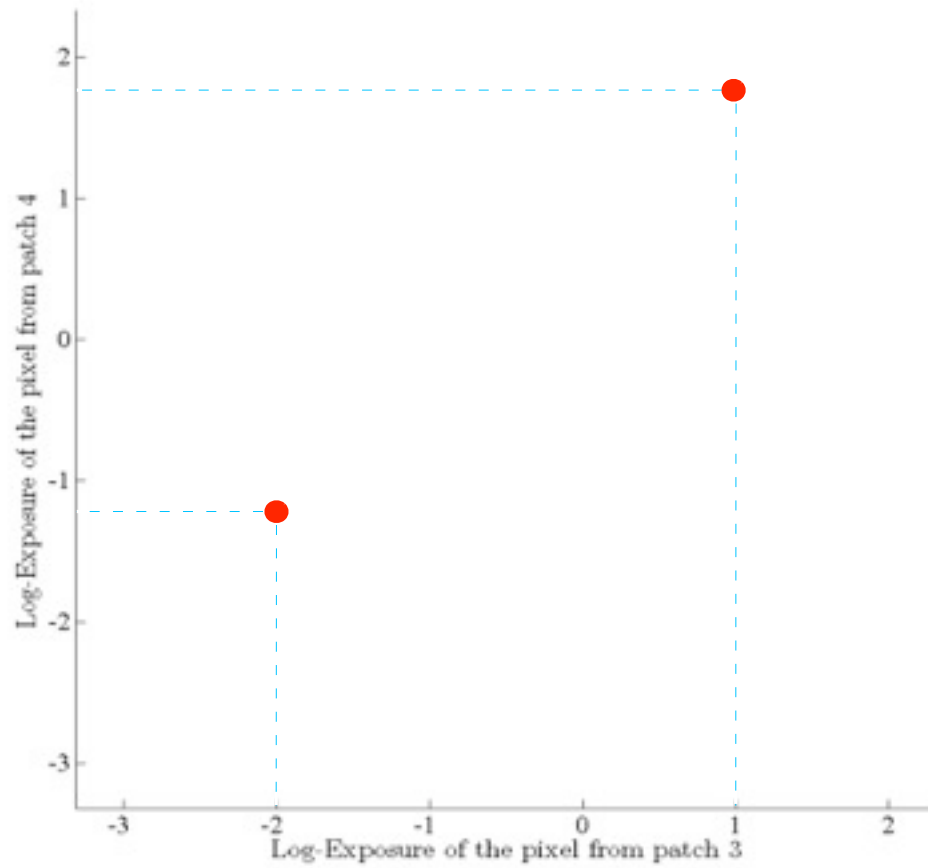
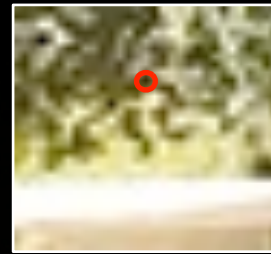
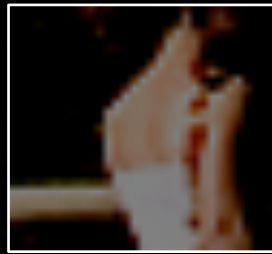


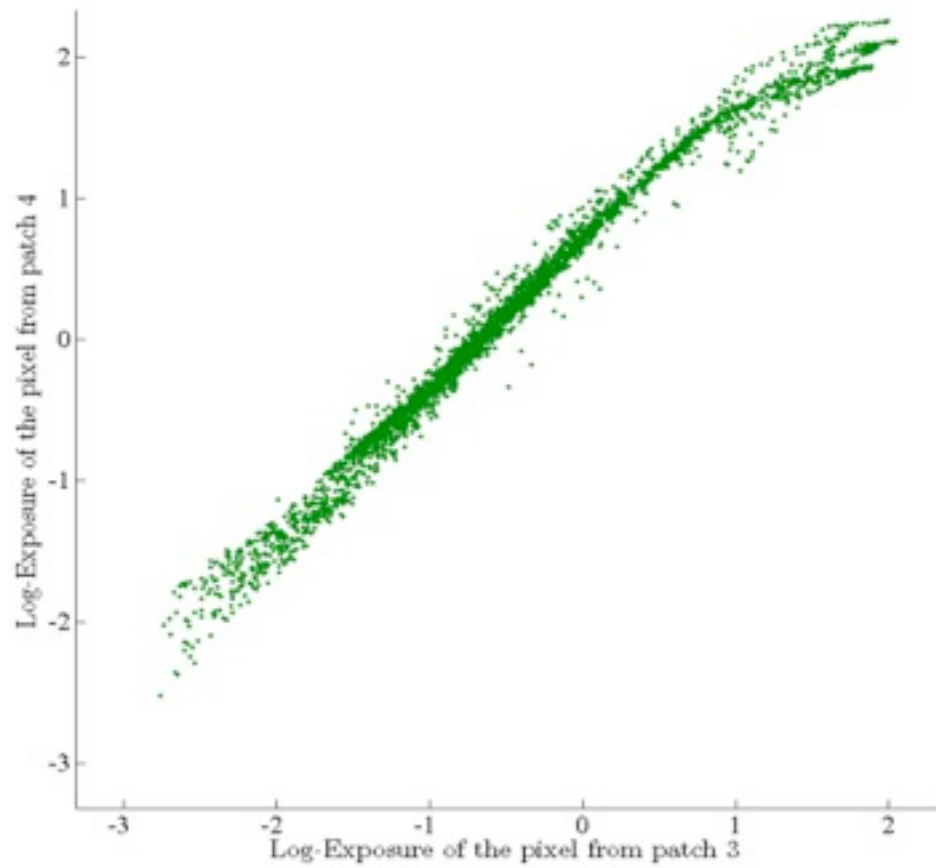
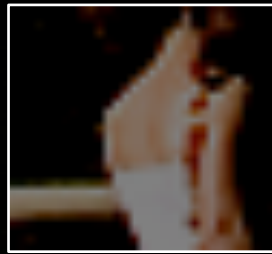


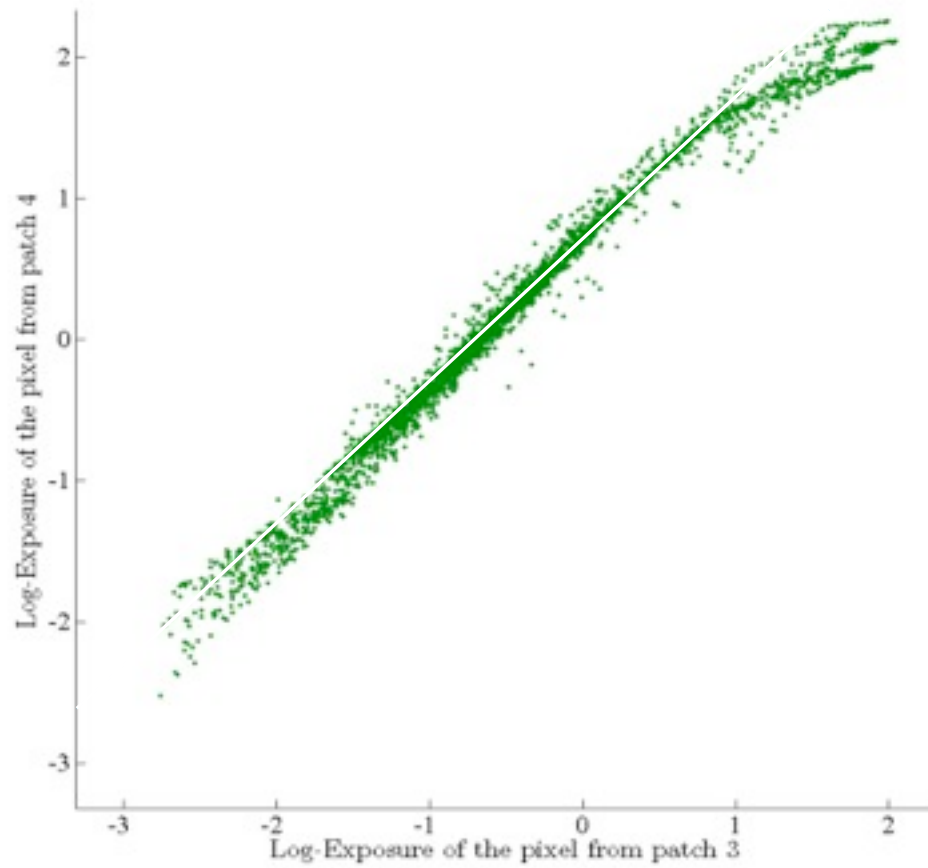


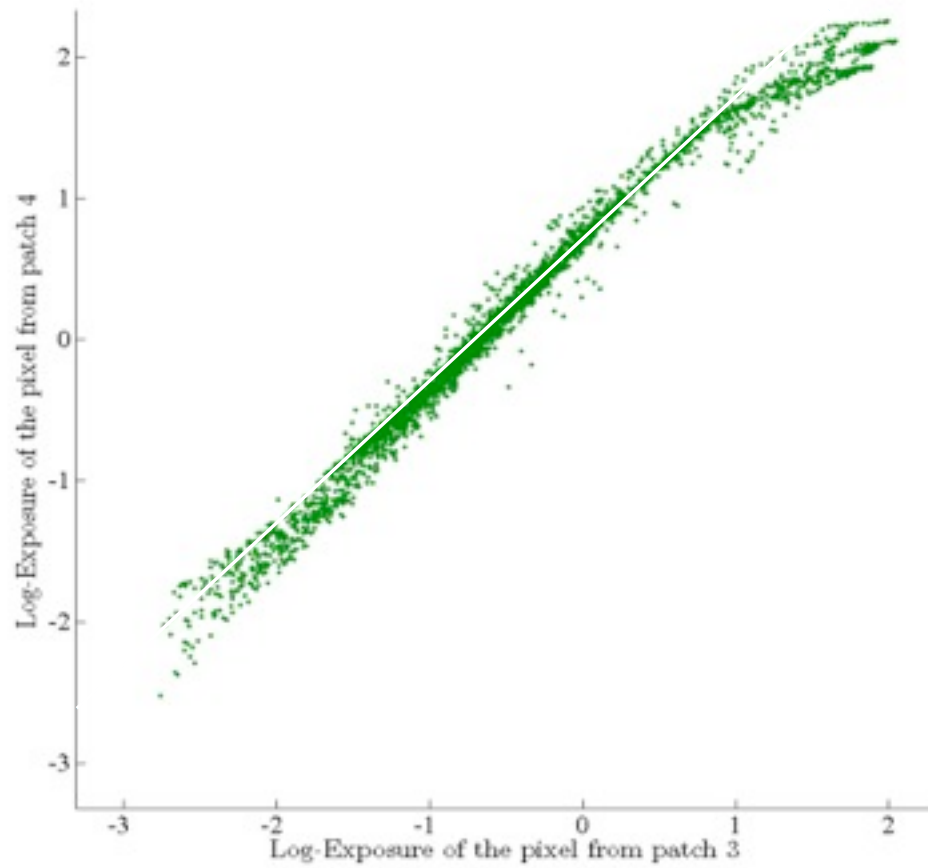
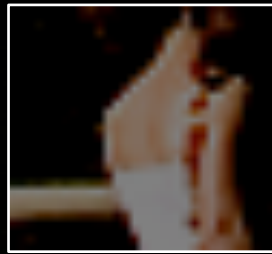




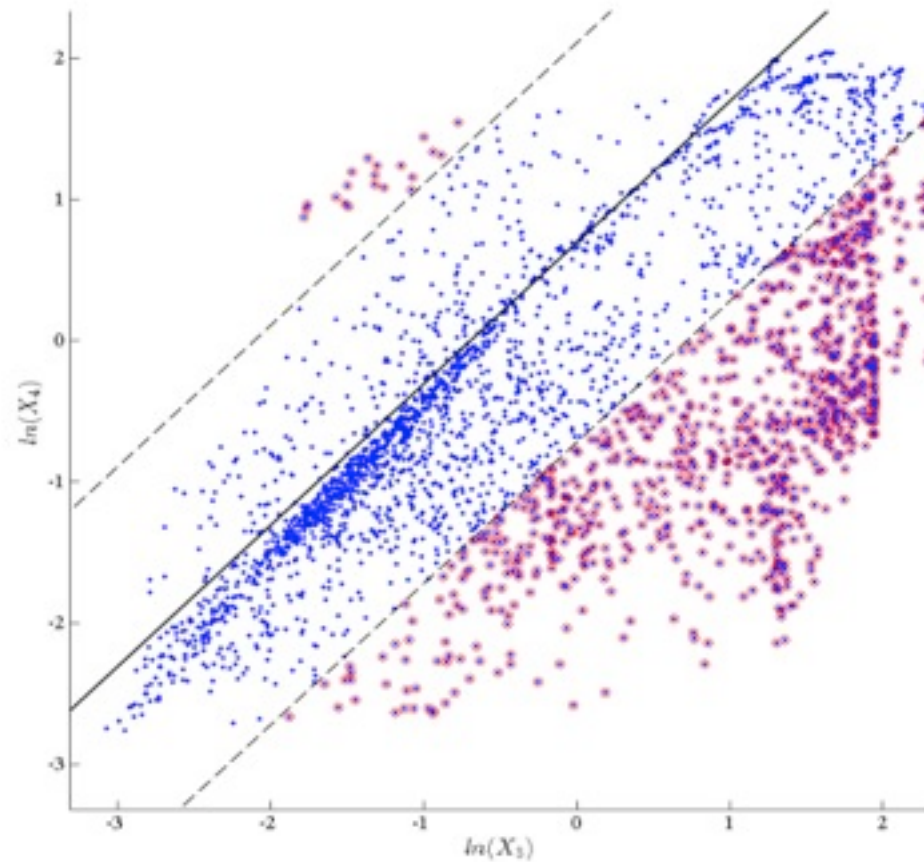


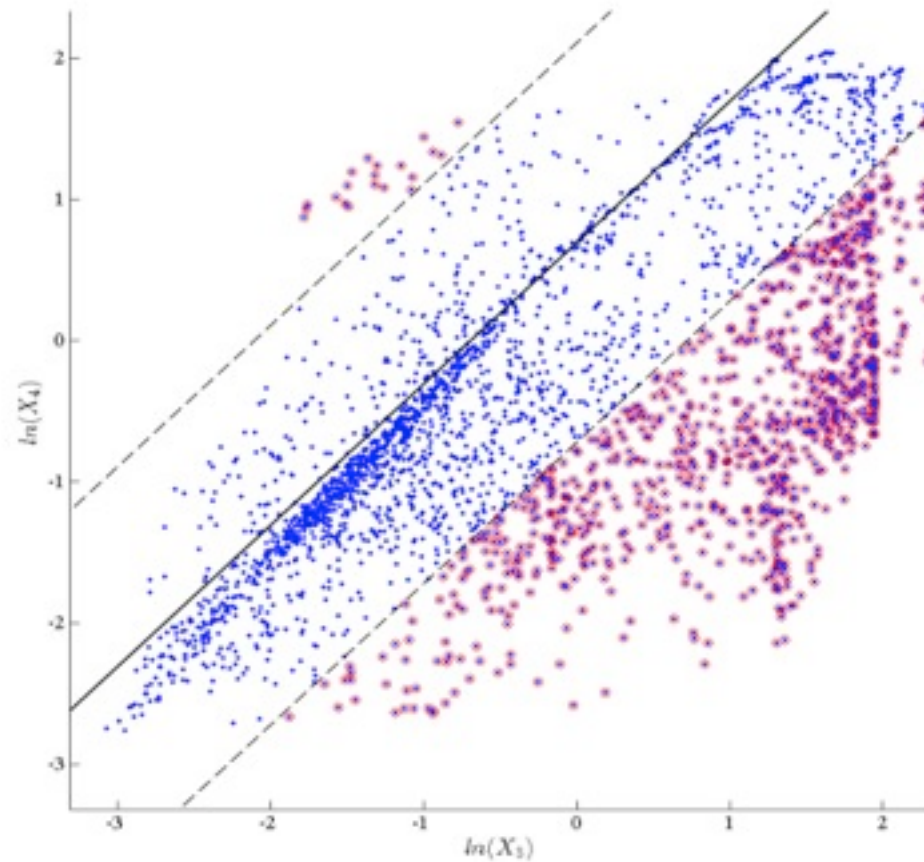


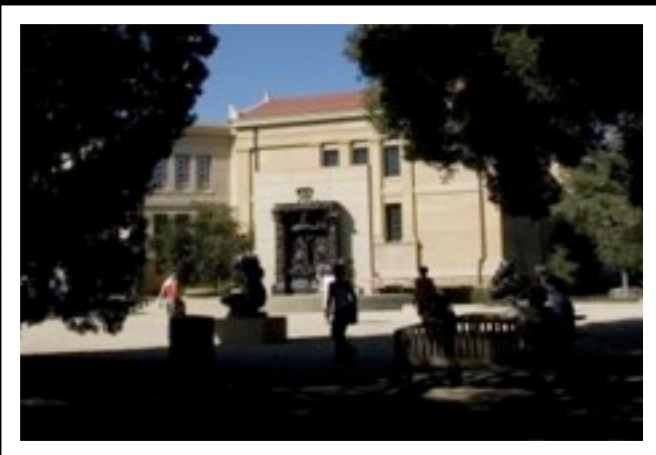


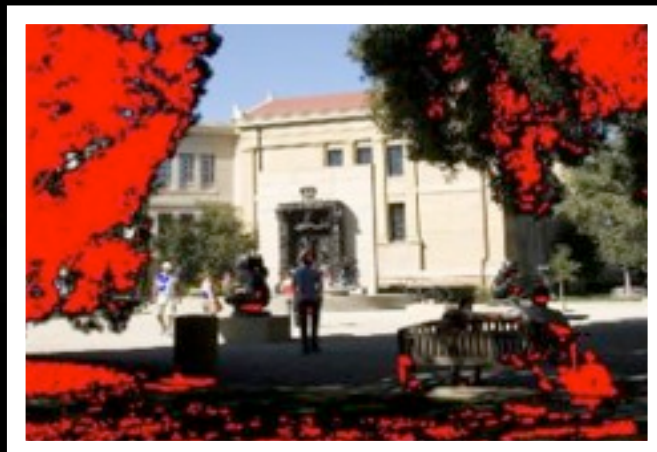


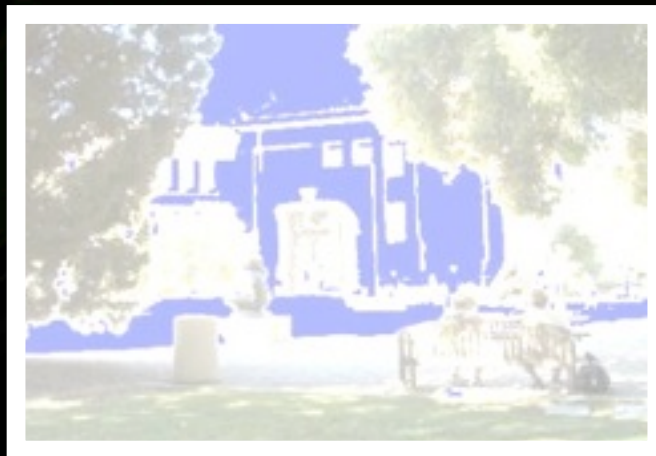
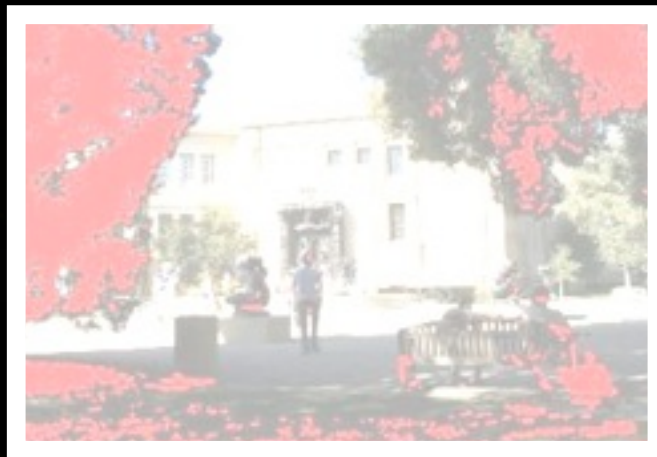
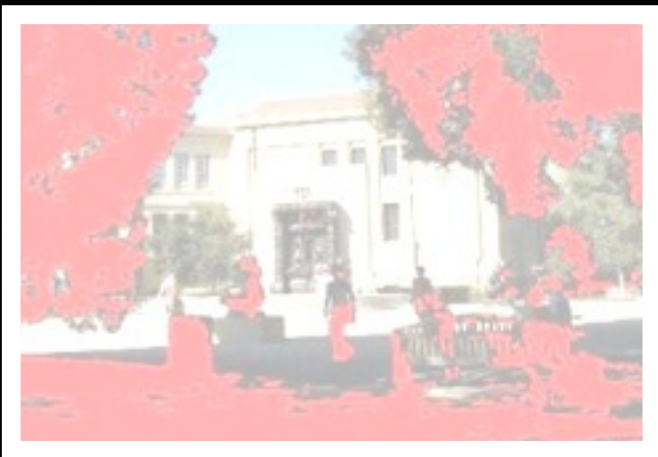


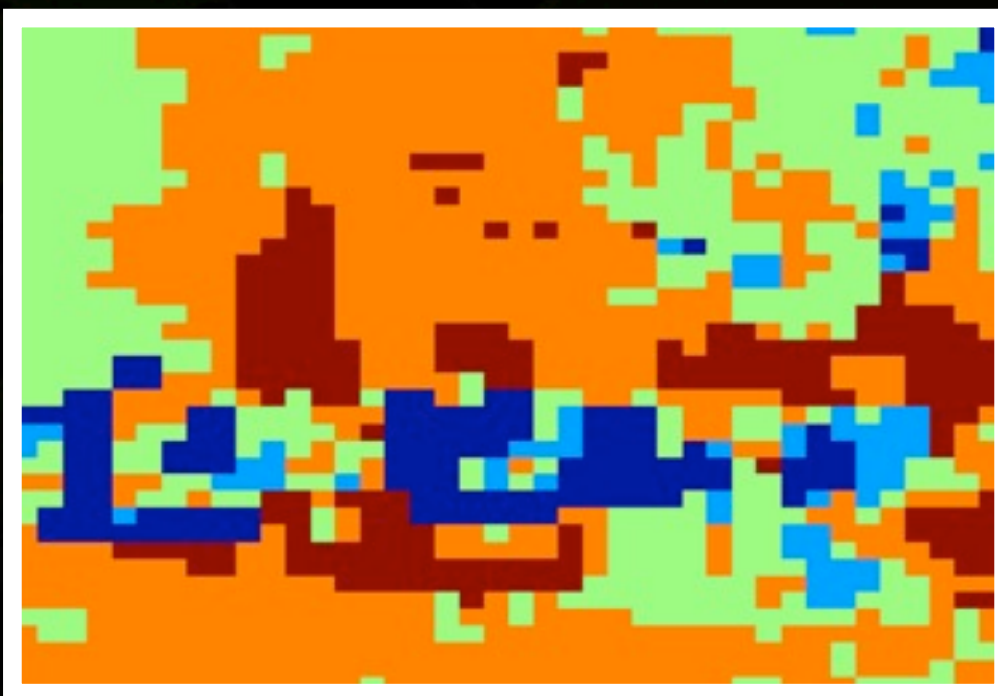












- 5 exposures
- 4 exposures
- 3 exposures
- 2 exposures
- 1 exposure



Monday, March 5, 12



NVIDIA Research

Monday, March 5, 12



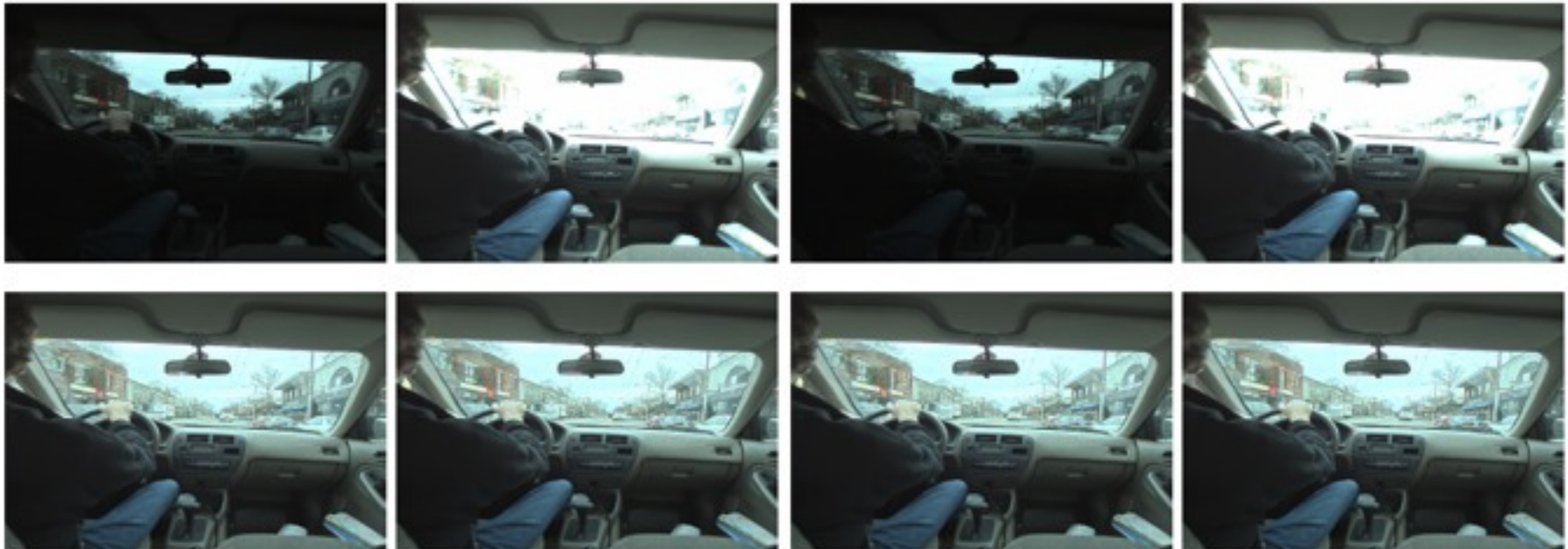
NVIDIA Research

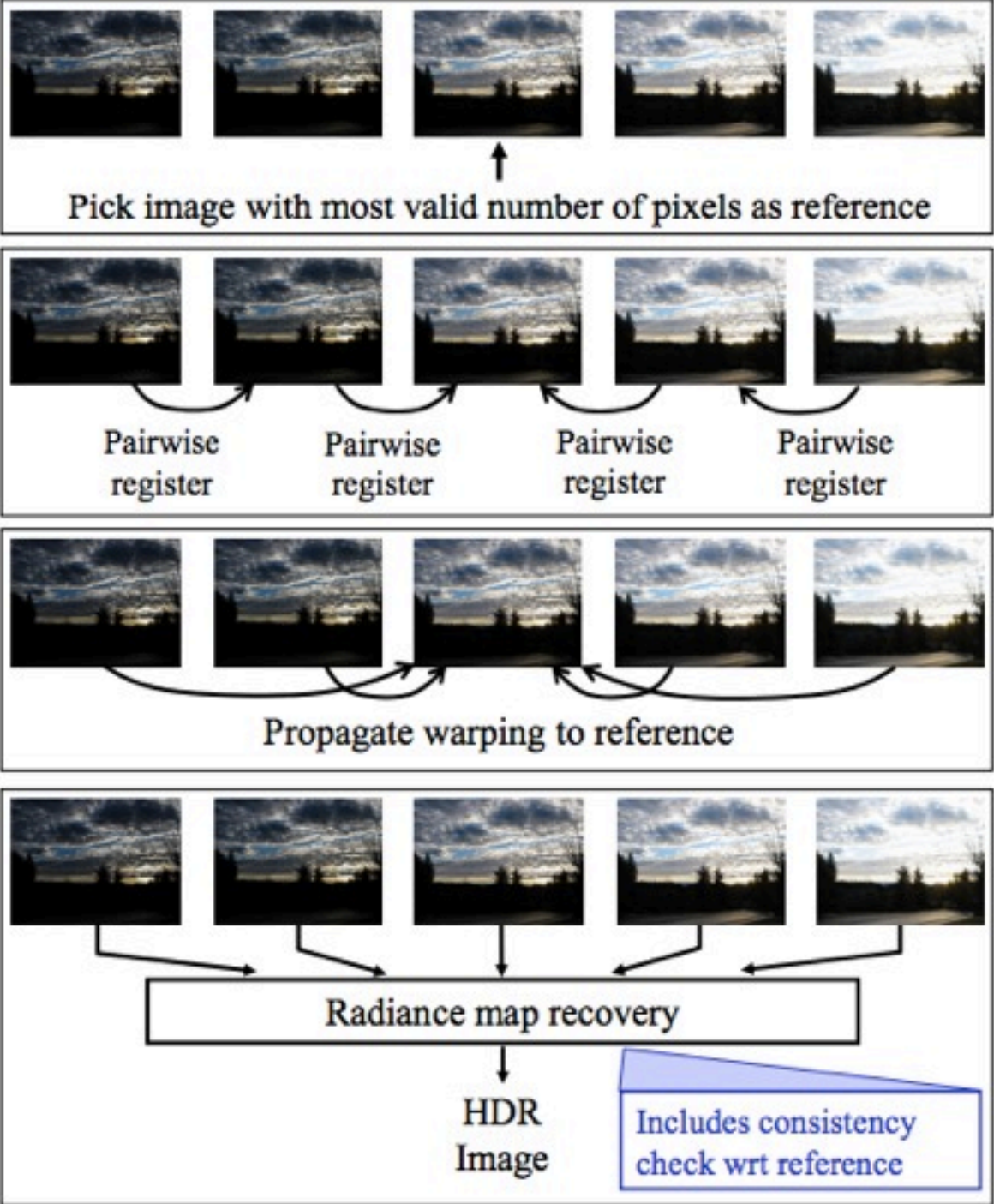
Monday, March 5, 12

HDR video



- Kang et al. 2003
 - automatic exposure control
 - register neighboring frames (motion compensation)
 - tonemapping





A Versatile HDR Video Production System

ACM SIGGRAPH 2011

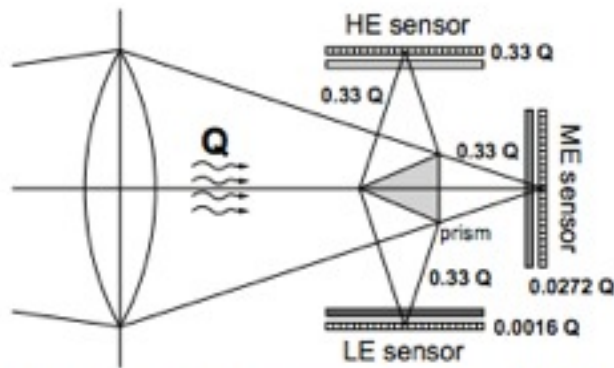


Figure 2: A traditional beamsplitting HDR optical system. Here a beamsplitting prism breaks up the light into three parts, one for each sensor fitted with different filters. Designs that use absorptive filters like this one make inefficient use of light.

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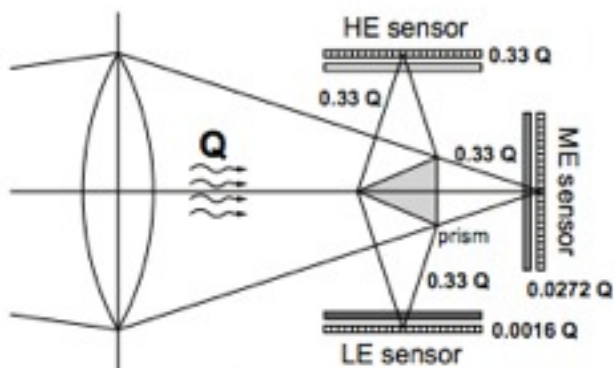


Figure 2: A traditional beamsplitting HDR optical system. Here a beamsplitting prism breaks up the light into three parts, one for each sensor fitted with different filters. Designs that use absorptive filters like this one make inefficient use of light.

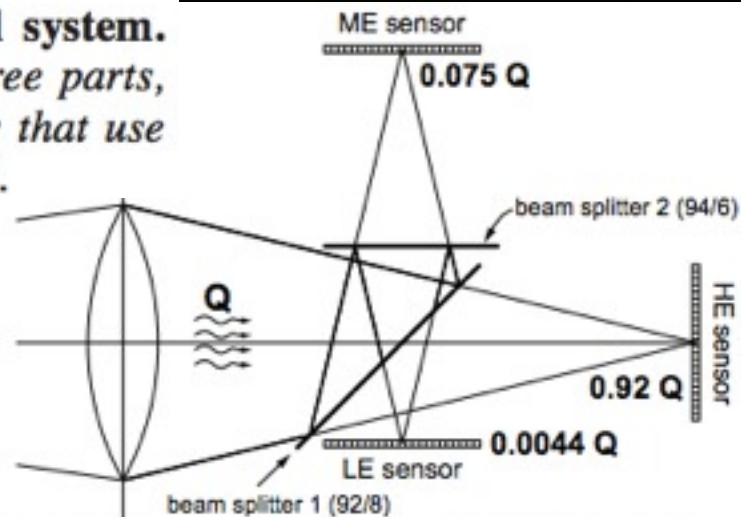


Figure 3: Illustration of our optical architecture. We also use beam splitters between the lens and sensors, but the key difference is that we re-use the optical path to improve our light efficiency. In the end, 99.96% of light entering the aperture arrives at the sensors. Light efficiency is important in all imaging applications.

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LDR image processing = asking for trouble



- Physically accurate image processing requires floats
 - 8bit or 16bit ints are not enough
 - inherent quantization between operations
 - e.g., applying gamma to brighten or darken maps levels that were separate to the same levels, can't separate any more
 - saturation
 - at the high end
 - can't deal with really bright pixels (direct light sources)
 - non-linearity
 - for better encoding, but not for physical processing

Image processing example: motion blur



- Processing LDR gamma-corrected images (sRGB) yields artifacts



blurred LDR



blurred HDR



blurred real photo