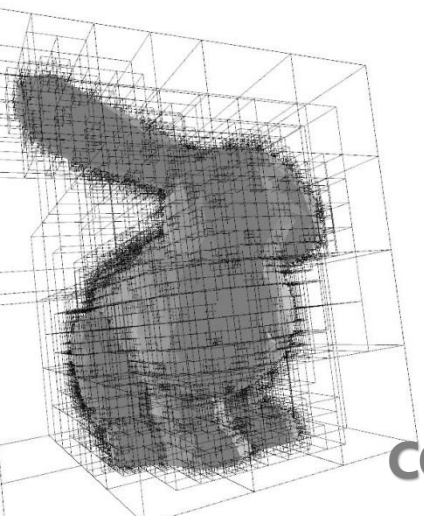


Image-Based Computer Graphics



Computational Photography

High Dynamic Range Imaging



Dynamic range problem

- How to record and display as human eyes?



With High Dynamic Range Imaging

- Normal vs HDR (after Tone Mapping)





Dynamic range problem

- The "dynamic range" of a scene is the contrast ratio between its brightest and darkest parts
- Typical images displayed on screen are 24-bits
 - 8-bits per color component (RGB)
 - 256 different intensity levels
- Real-world dynamic range is far greater than 256 intensity levels!
 - Eye can adapt from $\sim 10e-6$ to $10e6$ cd/m²
 - Often 1:100,000 in a scene

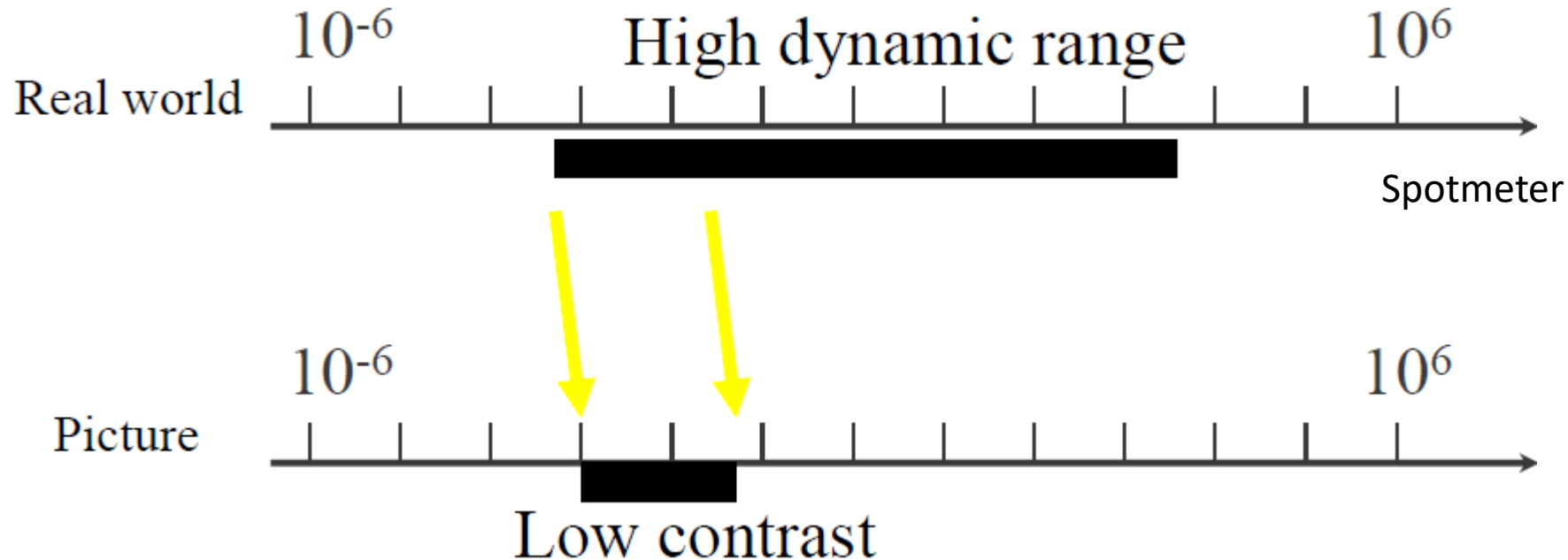


Dynamic range problem

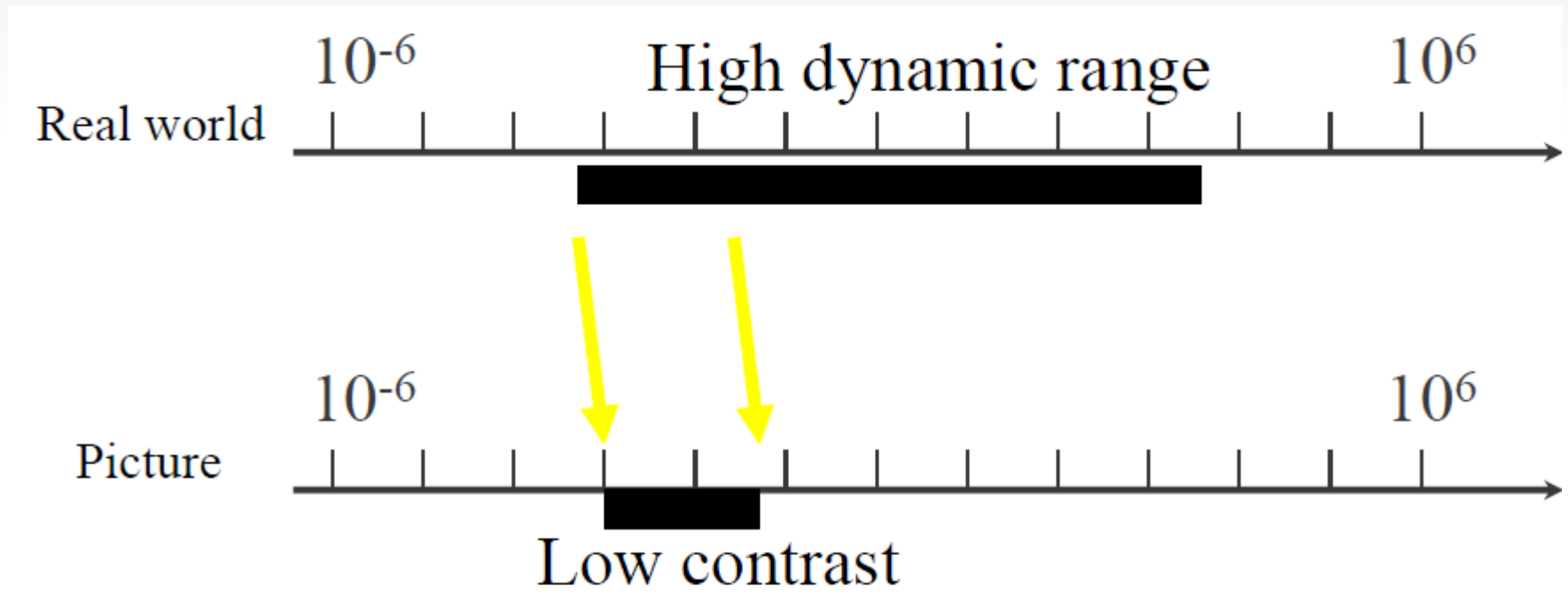
- Dynamic range of different media (approximate and debatable)
 - 10:1 photographic print (higher for glossy paper)
 - 20:1 artist's paints
 - 200:1 slide film
 - 500:1 negative film
 - 1000:1 LCD display
 - 2000:1 digital SLR (~11 bits)

Problem 1: How to record it?

- The range of illumination levels that we encounter is 10 to 12 orders of magnitudes
- Negatives/sensors can record 2 to 3 orders of magnitude



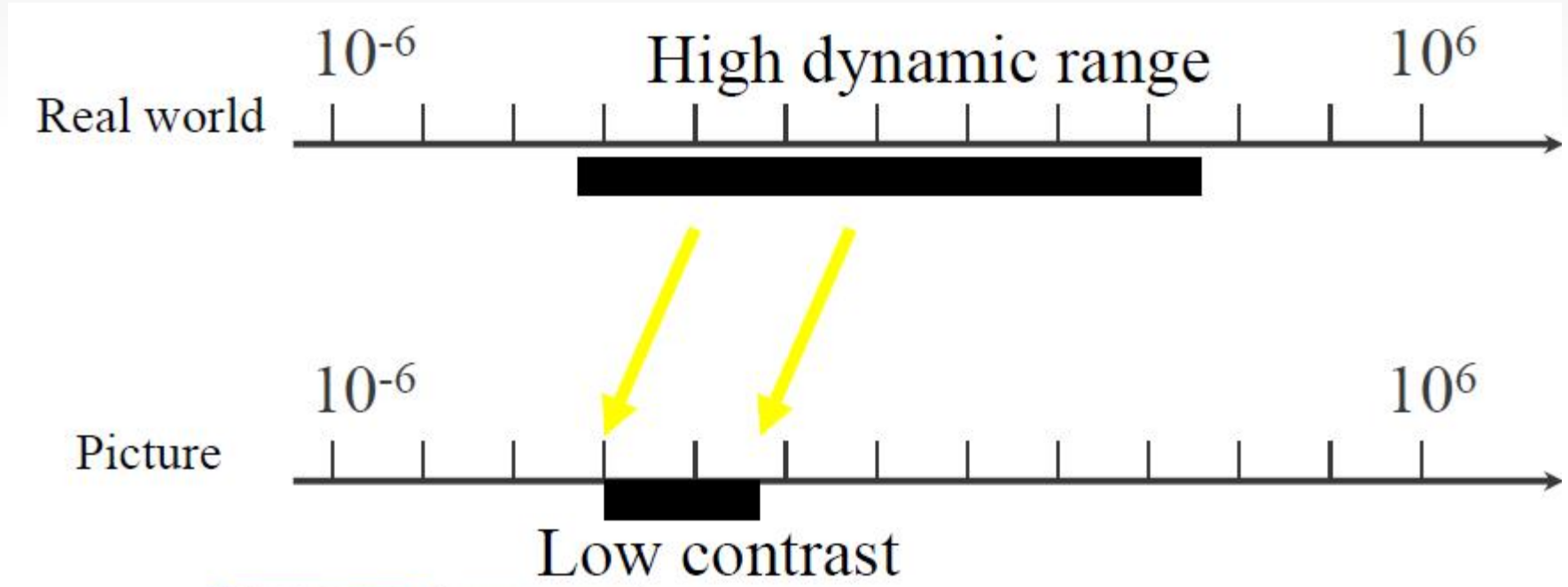
Multiple exposure photography



Shutter Speed: 10 sec

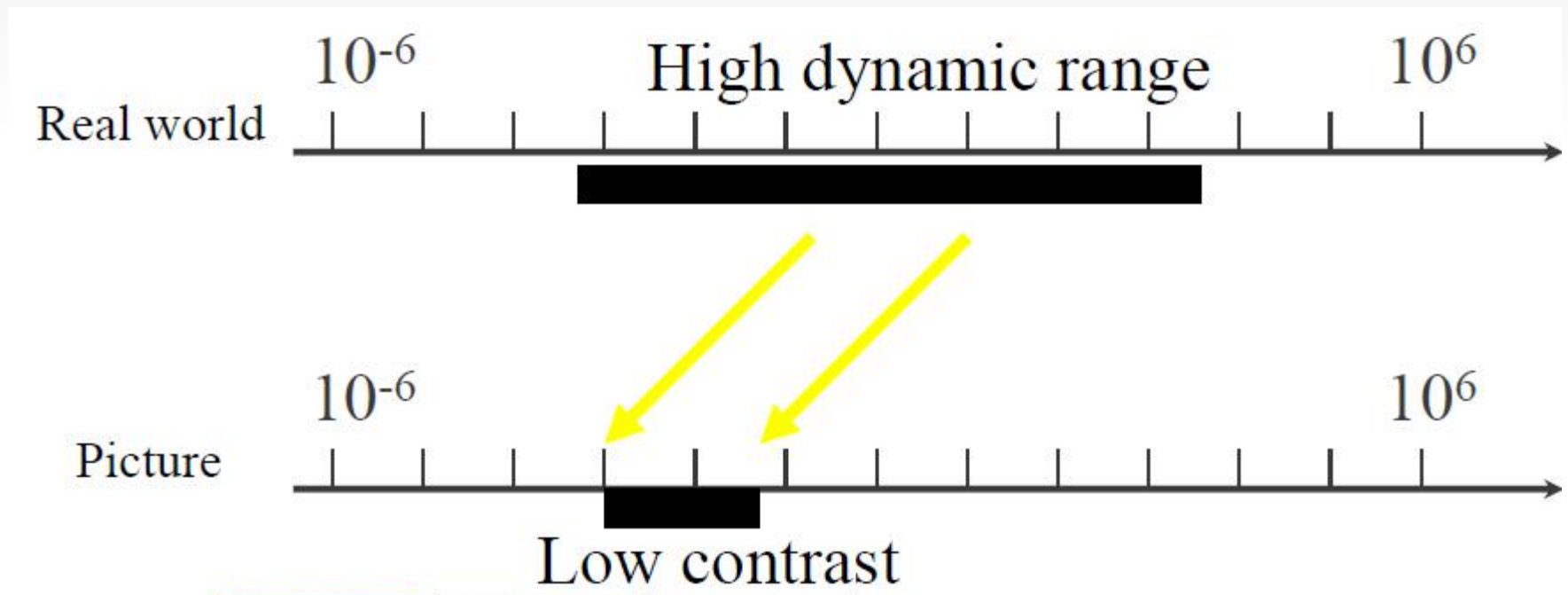


Multiple exposure photography



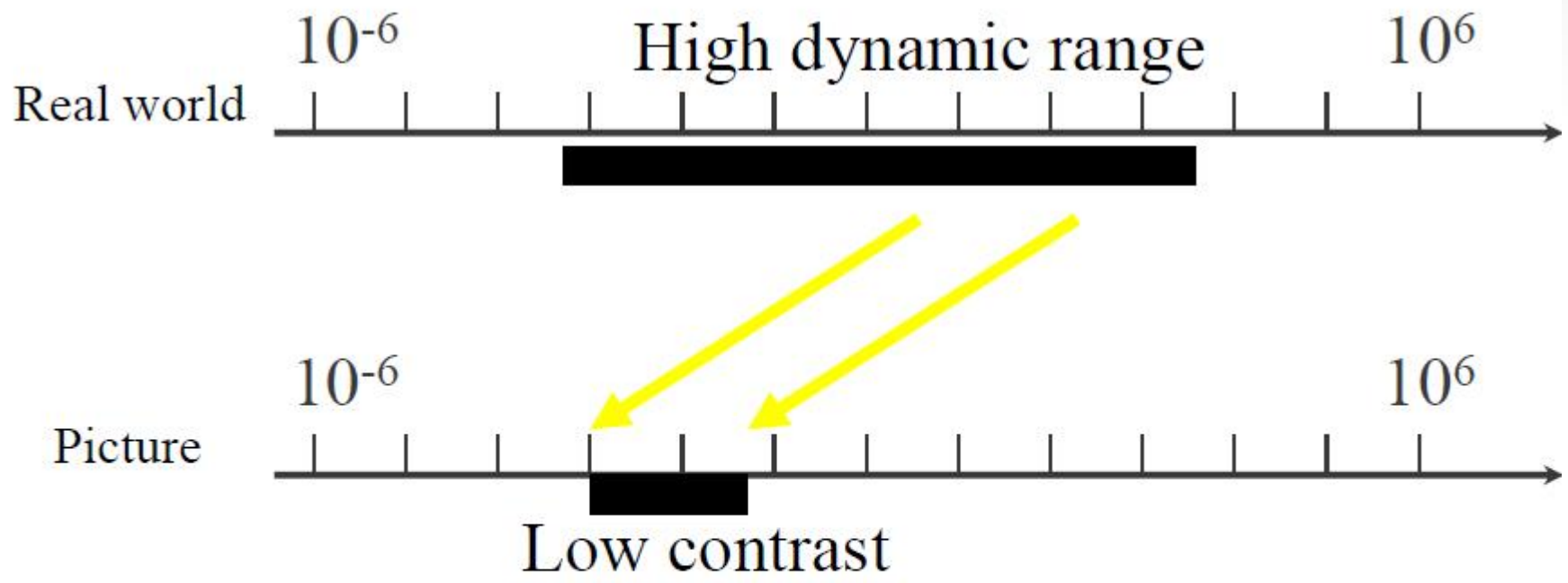
Shutter Speed: 1 sec

Multiple exposure photography



Shutter Speed: 0.1 sec

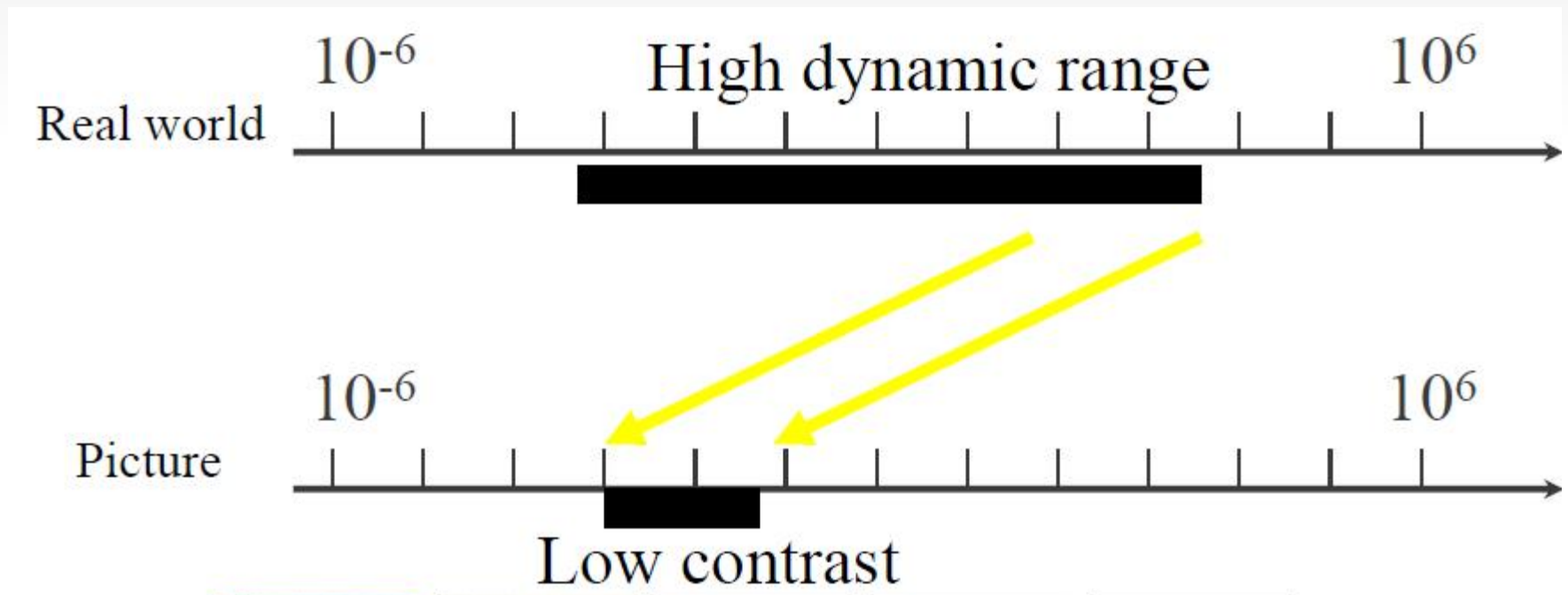
Multiple exposure photography



Shutter Speed: 0.01 sec



Multiple exposure photography



Shutter Speed: 0.001 sec



The ways to vary exposure

Shutter speed

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

Aperture

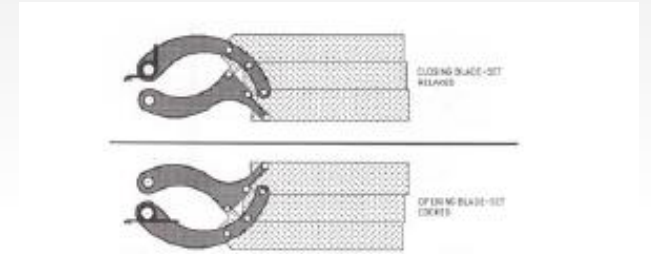
- Range: ~f/1.4 to f/22 (2.5 orders of magnitude)
- Cons: changes depth of field
- Useful when desperate

ISO

- Range: ~100 to 1600 (1.5 orders of magnitude)
- Cons: noise
- Useful when desperate

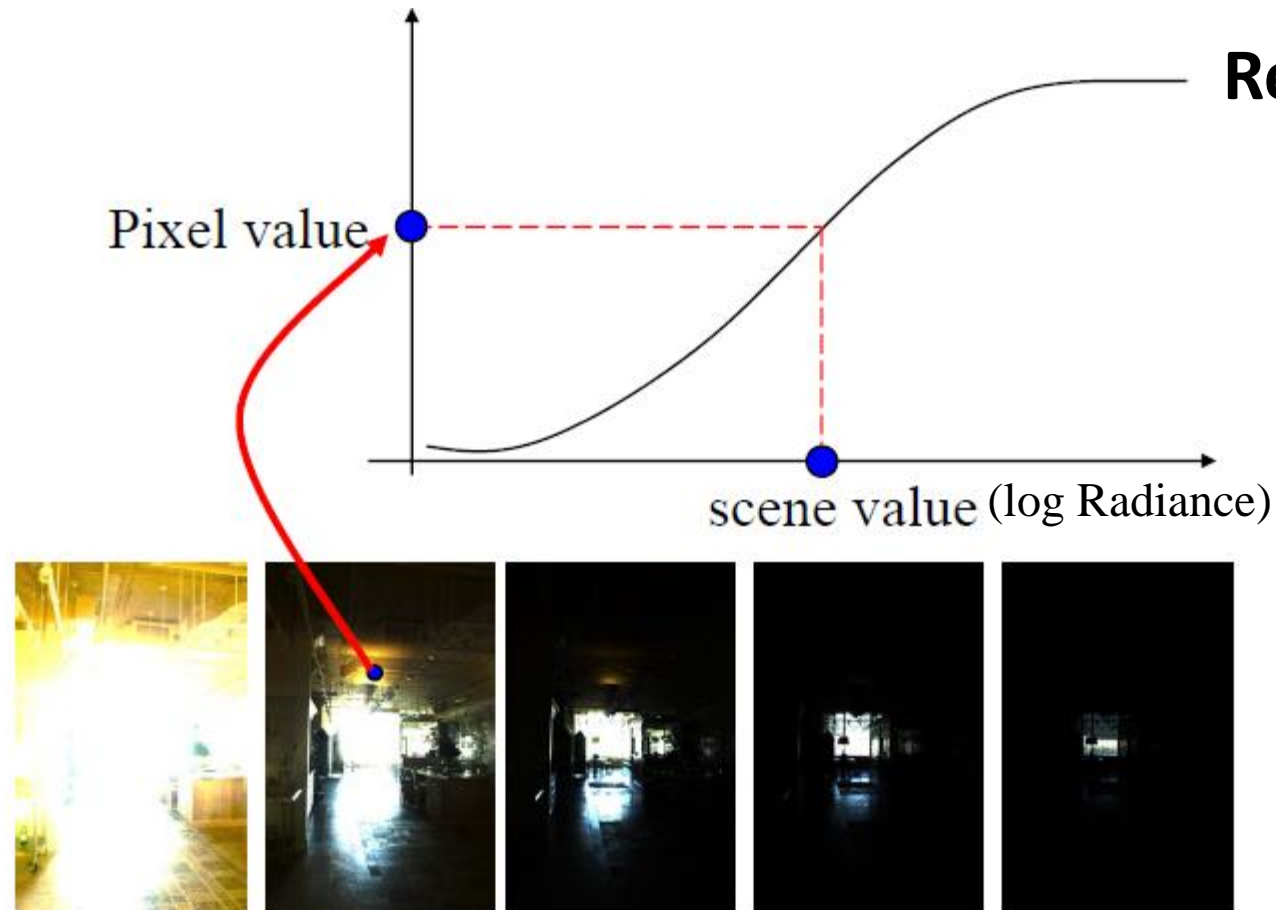
Neutral density filter

- 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



Response curve

- Given N photos at different exposure, recover a HDR color for each pixel, a response to its real **Radiance** (physical brightness)



Response curve

Compress the range by replacing radiance value (scene value) with its logarithm.

Then the dynamic range will be not too large to be displayed on a screen

Response curve

- Solutions

- See pixel values as response one scene point's radiance with different exposure parameters (radiance * time)

Image series



$\Delta t =$
10 sec



$\Delta t =$
1 sec



$\Delta t =$
1/10 sec



$\Delta t =$
1/100 sec

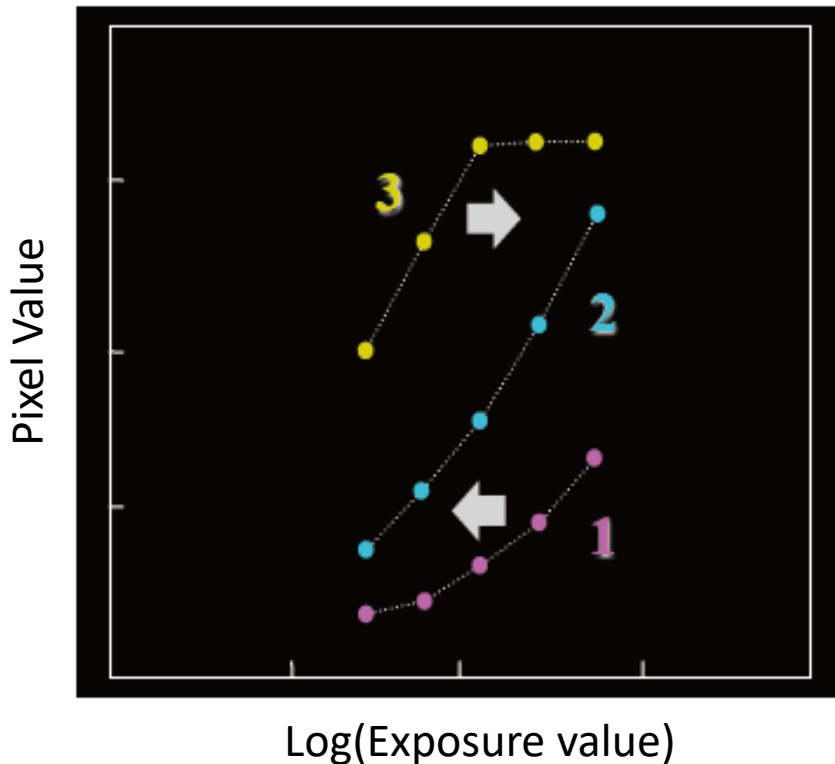


$\Delta t =$
1/1000 sec

*An alternative here : Vary scene luminance and see pixel values, assumes we control and know scene luminance

Response curve

- How to estimate a completed response curve?



For each pixel i and image j

$$\text{Pixel Value } Z_{ij} = f(\text{Exposure value})$$

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

$$\text{Log (Exposure value)} = \text{log(Radiance)} + \text{log } \Delta t$$

Response curve

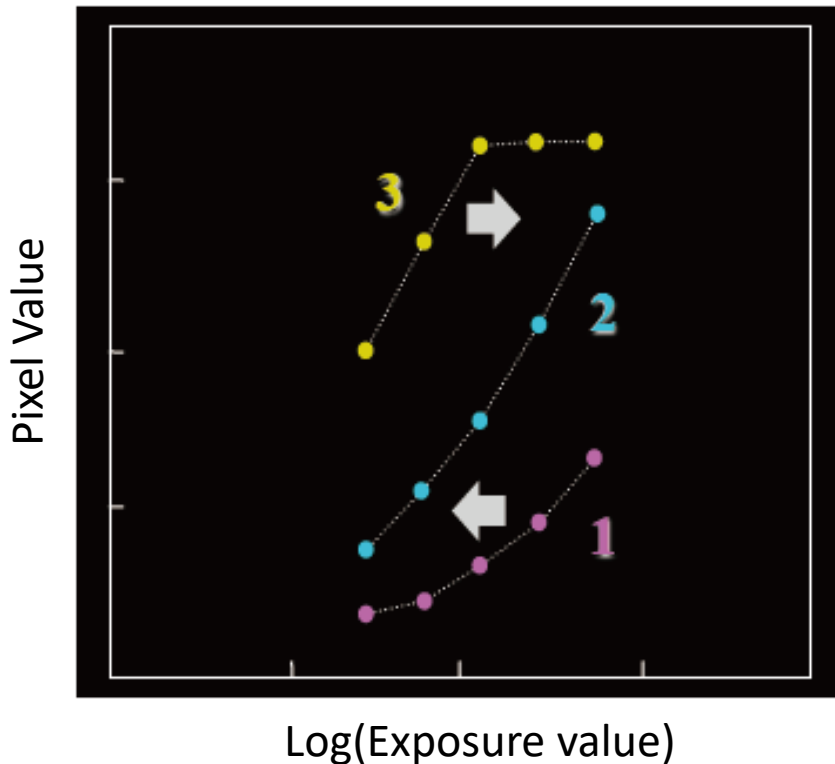
- How to estimate a completed response curve?

$$\text{Pixel Value } Z_{ij} = f(\text{Exposure value})$$

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

$$\text{Log (Exposure value)} = \text{log(Radiance)} + \text{log } \Delta t$$

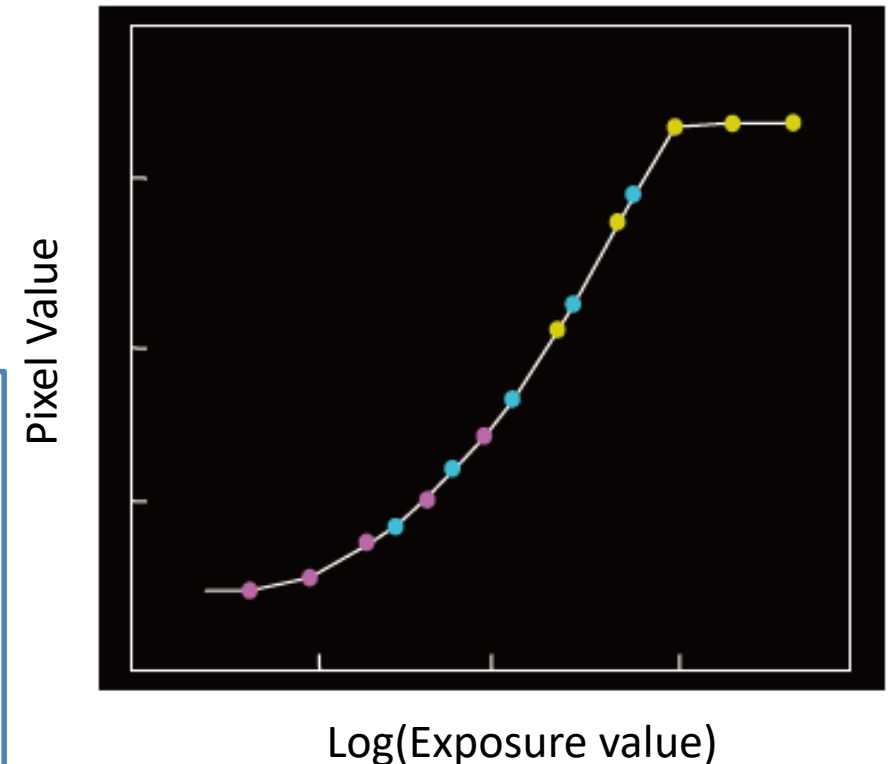
After adjusting radiances to obtain a smooth response curve



The position of each curve for each scene point is **NOT** decided yet. We are aiming at forming a target curve like the right image.

We are actually seeking for a real radiance value for every scene point. Eventually, a radiance value of a point is a translation for each segment of the curve.

Why?





Optimize the curve

Pixel Value $z_{ij} = f(\text{Exposure value}) \rightarrow \text{Exposure value} = f^{-1}(z)$, denote f^{-1} as g

Exposure value = Radiance * Δt

Log (Exposure value) = log(Radiance) + log Δt

- We need to take it as an optimization problem to find the “response” to the exposure value (equivalent to find the relationship between the pixel value and a real radiance).
- For each pixel site i in each image j , we want a function g of pixel value z :

$$\log(\text{Radiance}_i) + \log \Delta t_j = g_z$$

Known: $\Delta t_j, z_{ij}$

Unknown: g_z (z is the response pixel value for the logarithm exposure),
Radiance $_i$



Optimize the curve

$$\log(\text{Radiance}_i) + \log \Delta t_j = g_z$$

- We want the value to satisfy the above equation. But we only have observed relationship between **Radiance_i** and **g_z** (a difference by **Δt_j**).
- More importantly, for every observed exposure value **log(Radiance_i) + log Δt_j**, we actually have many pixels in j images with the same value, but may correspond to different pixel intensity level **z**.
- It is an over-determined system. We transfer it into an optimization problem, to minimize the difference between **log(Radiance_i) + log Δt_j** and **g_z**



Optimize the curve

$$\log(\text{Radiance}_i) + \log \Delta t_j = g_z$$

- If we have 0-M pixel intensity levels to represent the logarithm response curve, the actual unknown variables should be: g_0, g_1, \dots, g_M and the radiance values for sampled scene points
- For every observed point in image j at pixel i, we can get a difference term: $\log(\text{Radiance}_i) + \log \Delta t_j - g_z$, where z means the observed pixel intensity level

Optimize the curve

$$\log(\text{Radiance}_i) + \log \Delta t_j = g_z$$

- Minimize this energy to make the curve fit the observed points.

$$\sum_{i=1}^N \sum_{j=1}^P [\log(\text{Radiance}_i) + \log \Delta t_j - g_z]^2 \quad \text{fitting term}$$

- Add another smoothness term to get a smooth response curve

$$\sum_{i=1}^N \sum_{j=1}^P [\log(\text{Radiance}_i) + \log \Delta t_j - g_z]^2 + \lambda \sum_{z_{min}}^{z_{max}} (g''_z)^2 \quad \text{smoothness term}$$

$$g_{z-1} - 2g_z + g_{z+1}$$



Optimize the curve

$$\sum_{i=1}^N \sum_{j=1}^P [\log(\text{Radiance}_i) + \log \Delta t_j - g_z]^2 + \lambda \sum_{z_{min}}^{z_{max}} (g''_z)^2$$

Solve an over-determined linear system to get the minimum of this energy, because the minimum value is where the derivative is zero.

Consider a vector x formed by all the unknowns

By setting the derivatives to zero, we will get a form of:

$Ax = b$ (A is a matrix, b is a know vector), then solve the linear system.

The code in the original paper by Debevec

```
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);

k = 1;           %% Include the data-fitting equations
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

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

for i=1:n-2      %% Include the smoothness equations
    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

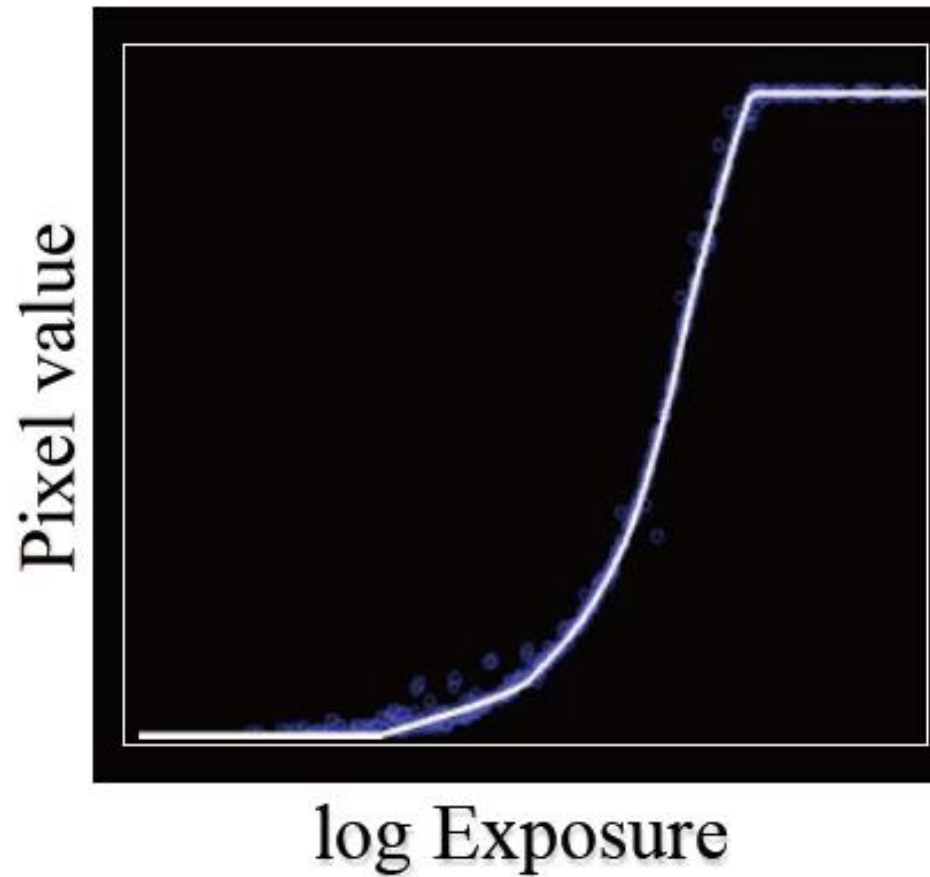
x = A\b;        %% Solve the system using SVD
```

Response Curve

Kodak DCS460
1/30 to 30 sec



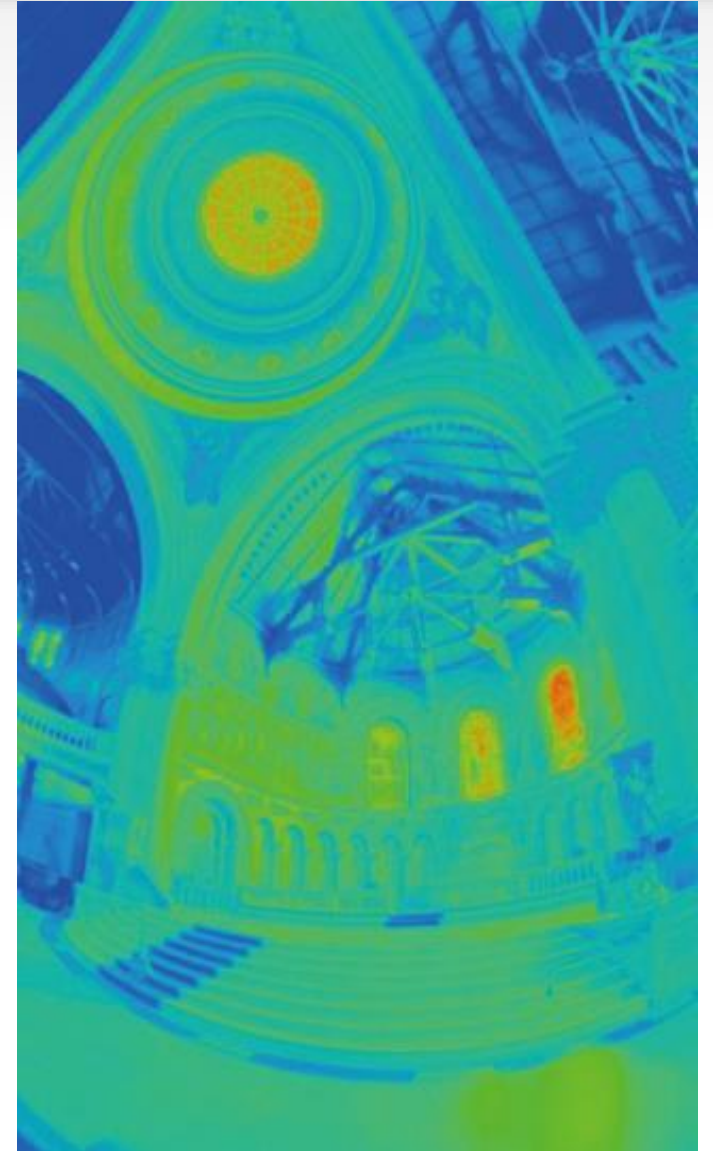
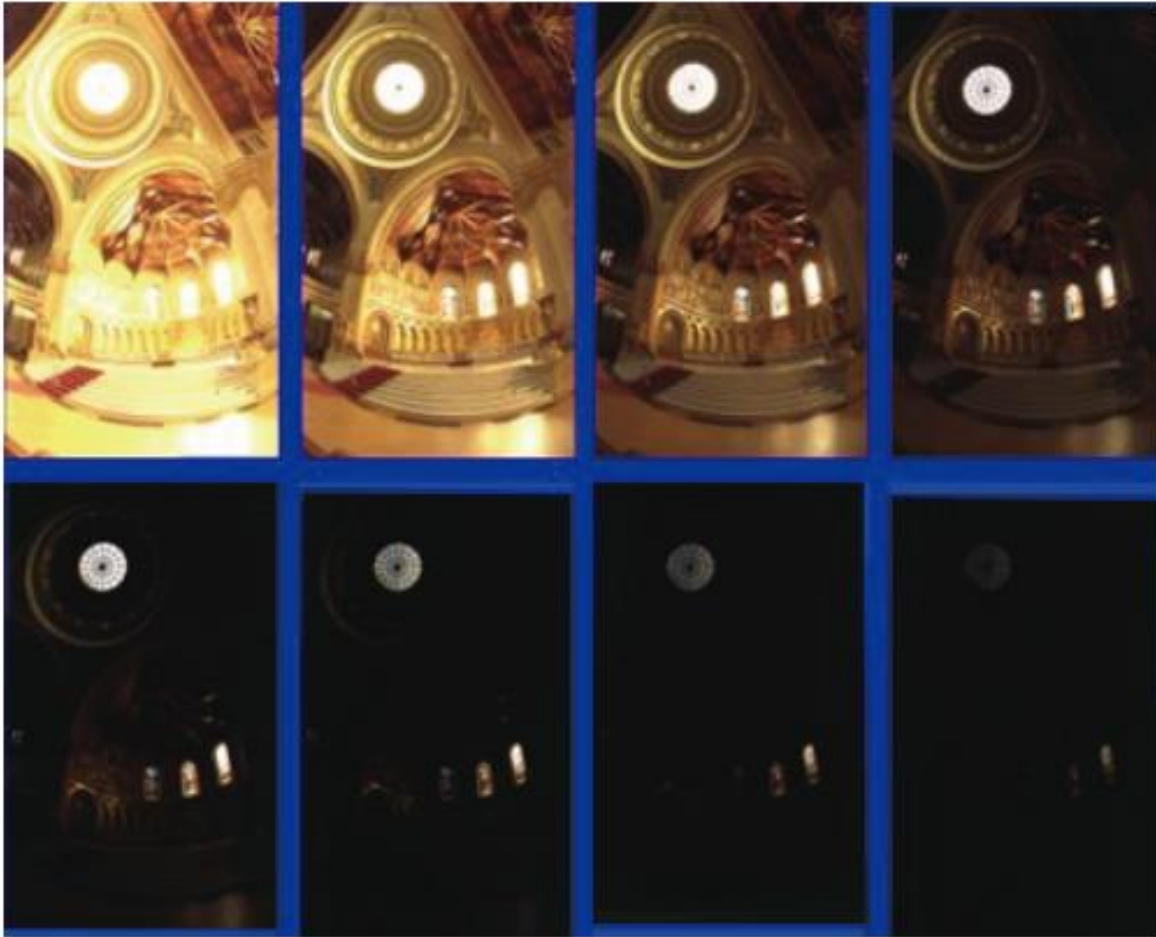
Recovered response curve



Reconstructed radiance map



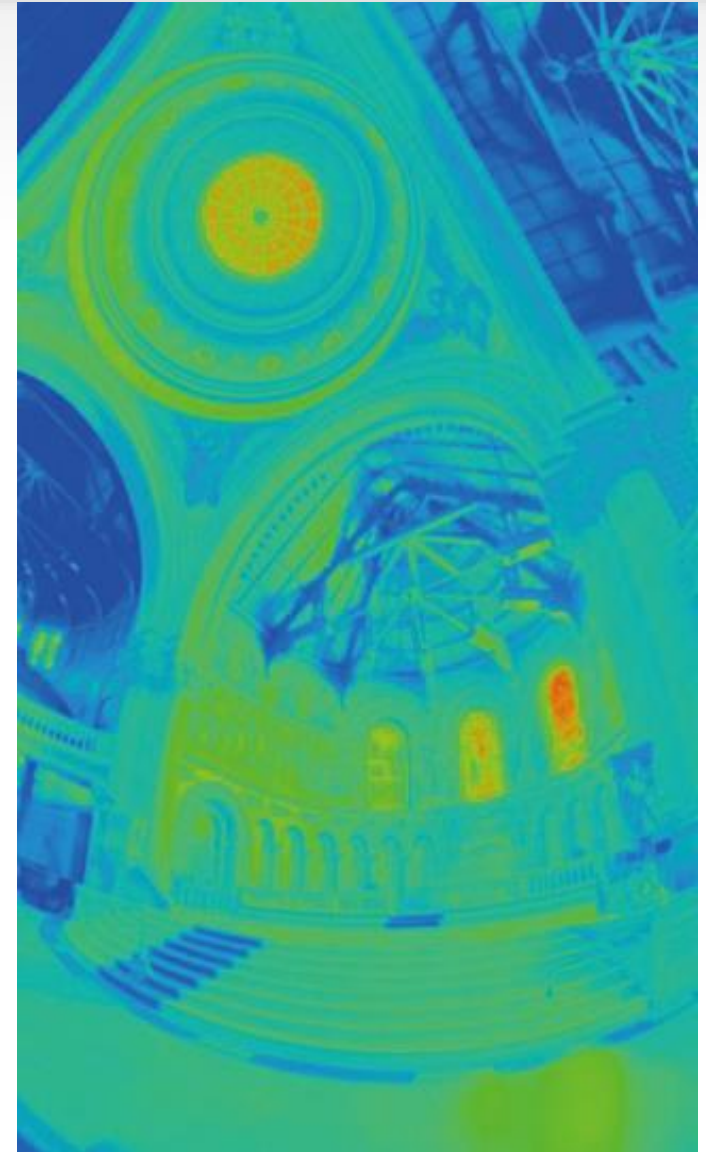
Reconstructed radiance map



Problem 2: How to display?



Linearly scaled
to display device



How to display?

- Input: high-dynamic-range image (Radiance map, floating point per pixel)
- Scene has *1:10,000* contrast, display has *1:100*, so simplest contrast reduction?



Naïve: Gamma compression

- $Z = r^\gamma$ In this case, $\gamma = 0.5$ can map the radiance into an acceptable range.
- Colors are washed-out.
 - Why ?
 - Larger RGB difference means larger color saturation

RGB: 80,93,120



Saturation: 51

41,93,159



Saturation: 150

It is applied independently on R, G & B. The linear relationship between 3 channels are not preserved by Gamma compression



Naïve: Gamma compression

- Keep the color, just compress the intensity.
- We can do it by Lab color space, or just extract the relative value of RGB channels.



Because the details are also compressed, **details (intensity high-frequency) are blurred**

$R' = R / \text{intensity}$
 $G' = G / \text{intensity}$
 $B' = B / \text{intensity}$
important to use ratios
(makes it luminance invariant)

Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



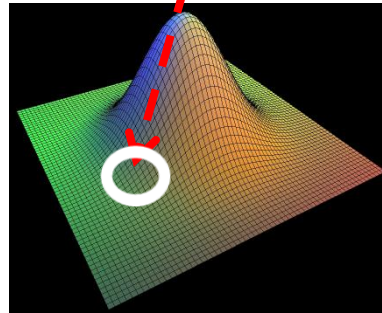
Recap Bilateral Filtering

- Weighted average of **SIMILAR** pixels

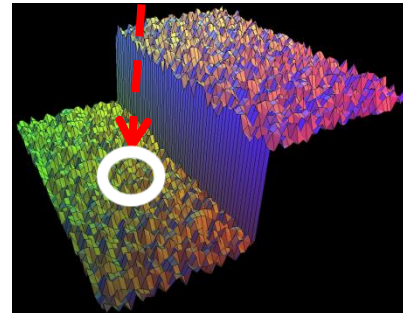
$$W_p = \sum_{q \in N} G_s(\|p - q\|) G_r(\|I_p - I_q\|)$$

$$B(p) = 1/W_p \sum_{q \in N} G_s(\|p - q\|) G_r(\|I_p - I_q\|) I_q$$

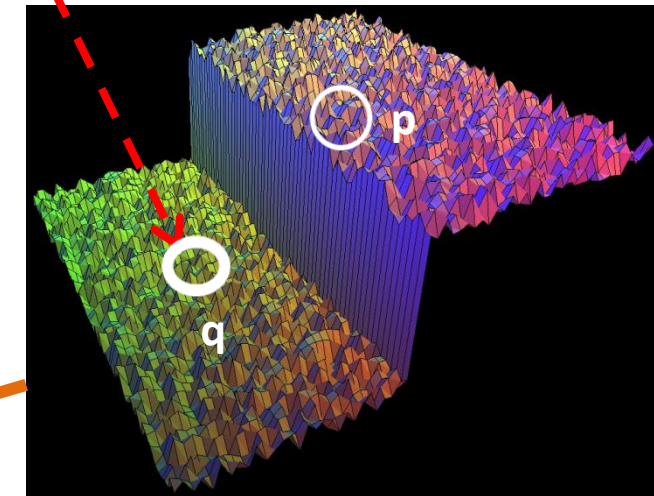
q			
		p	



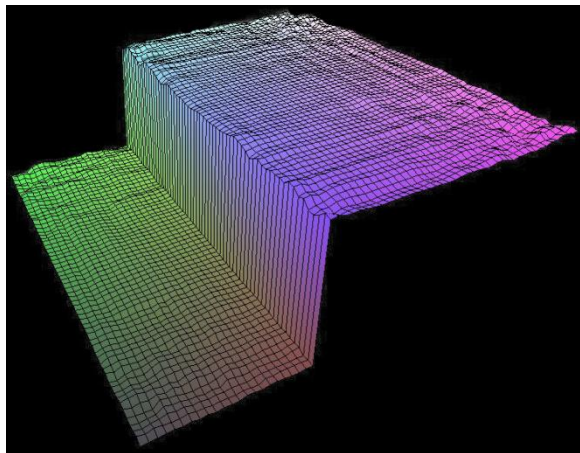
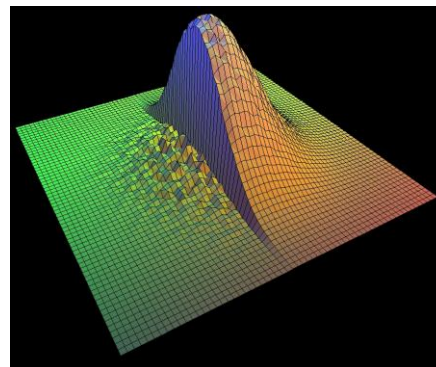
Spatial distance weight



Color difference weight



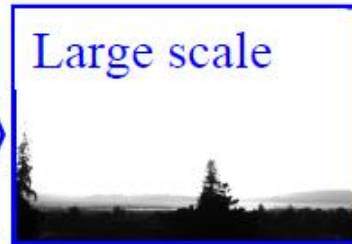
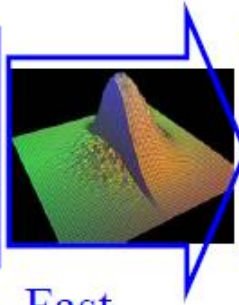
Input



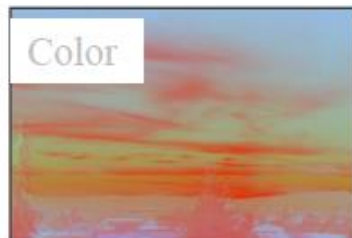
Output

Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



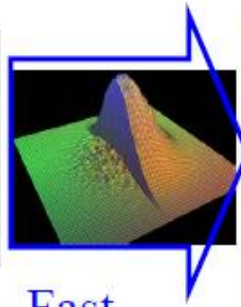
Fast
Bilateral
Filter



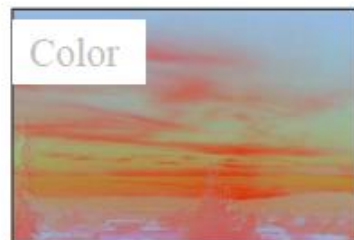
Spatial sigma: 2 to 5% image size
Range sigma: 0.4 (in log 10)

Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



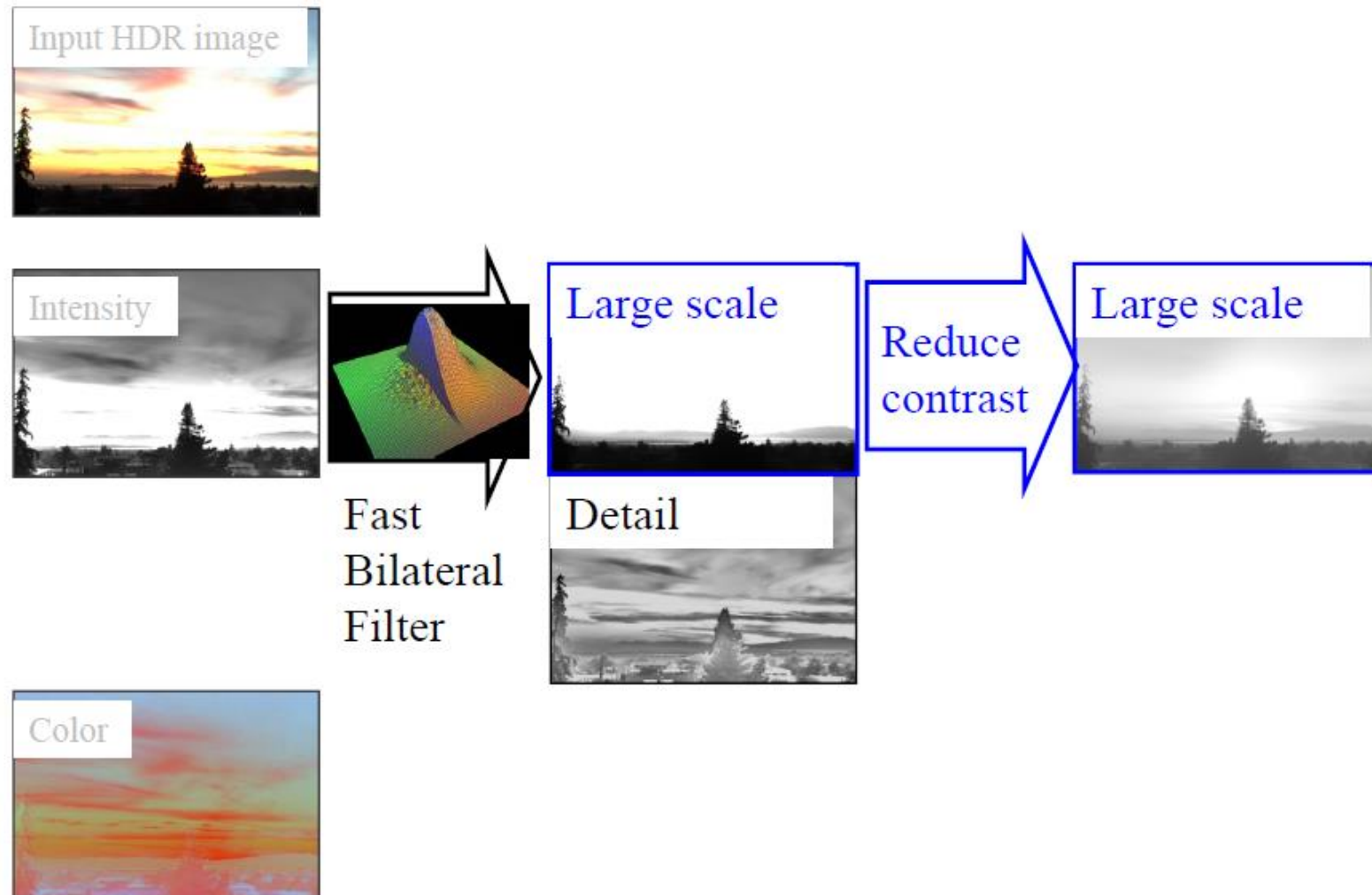
Fast
Bilateral
Filter



Detail = log intensity - large scale (residual)

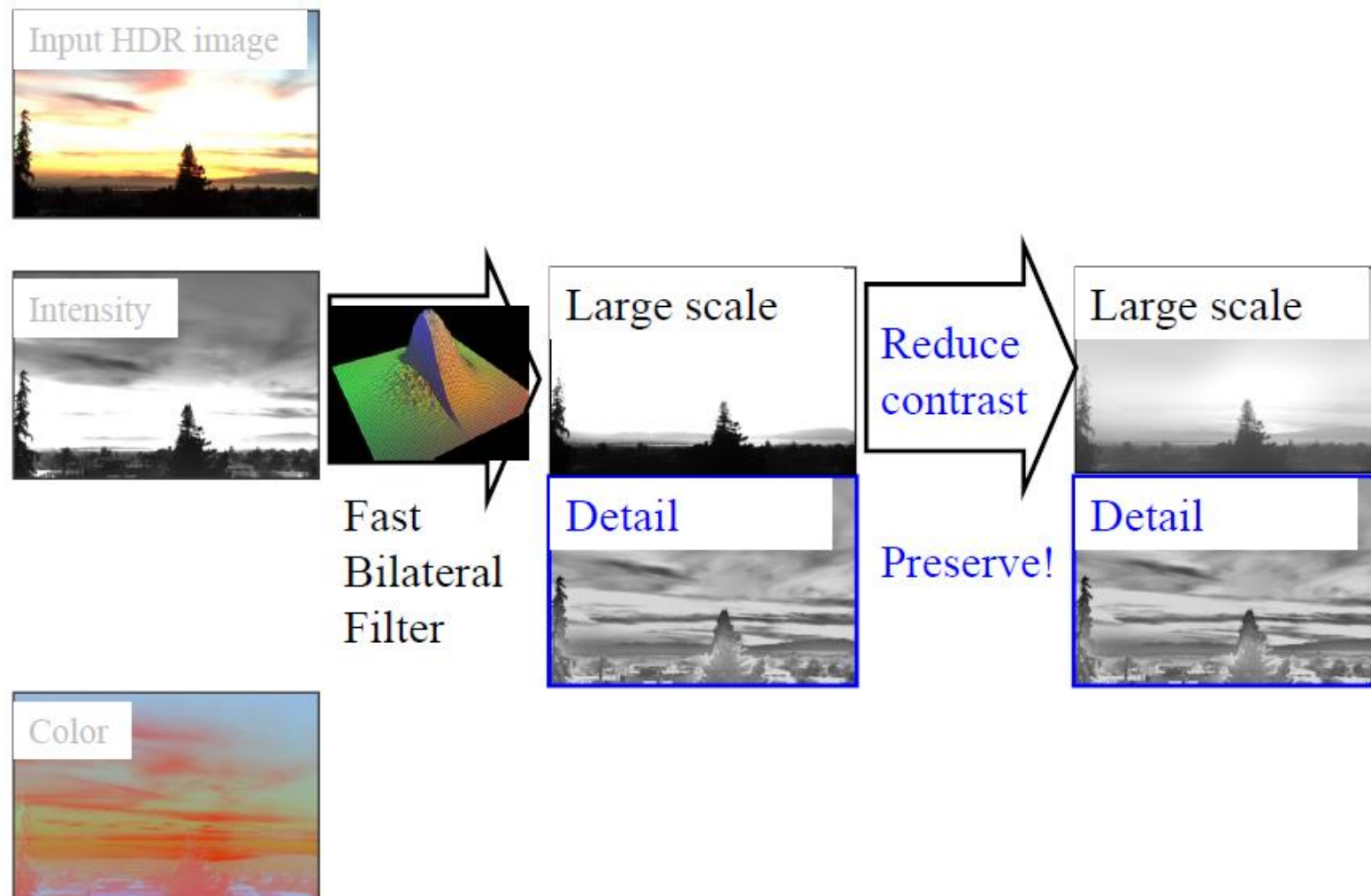
Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



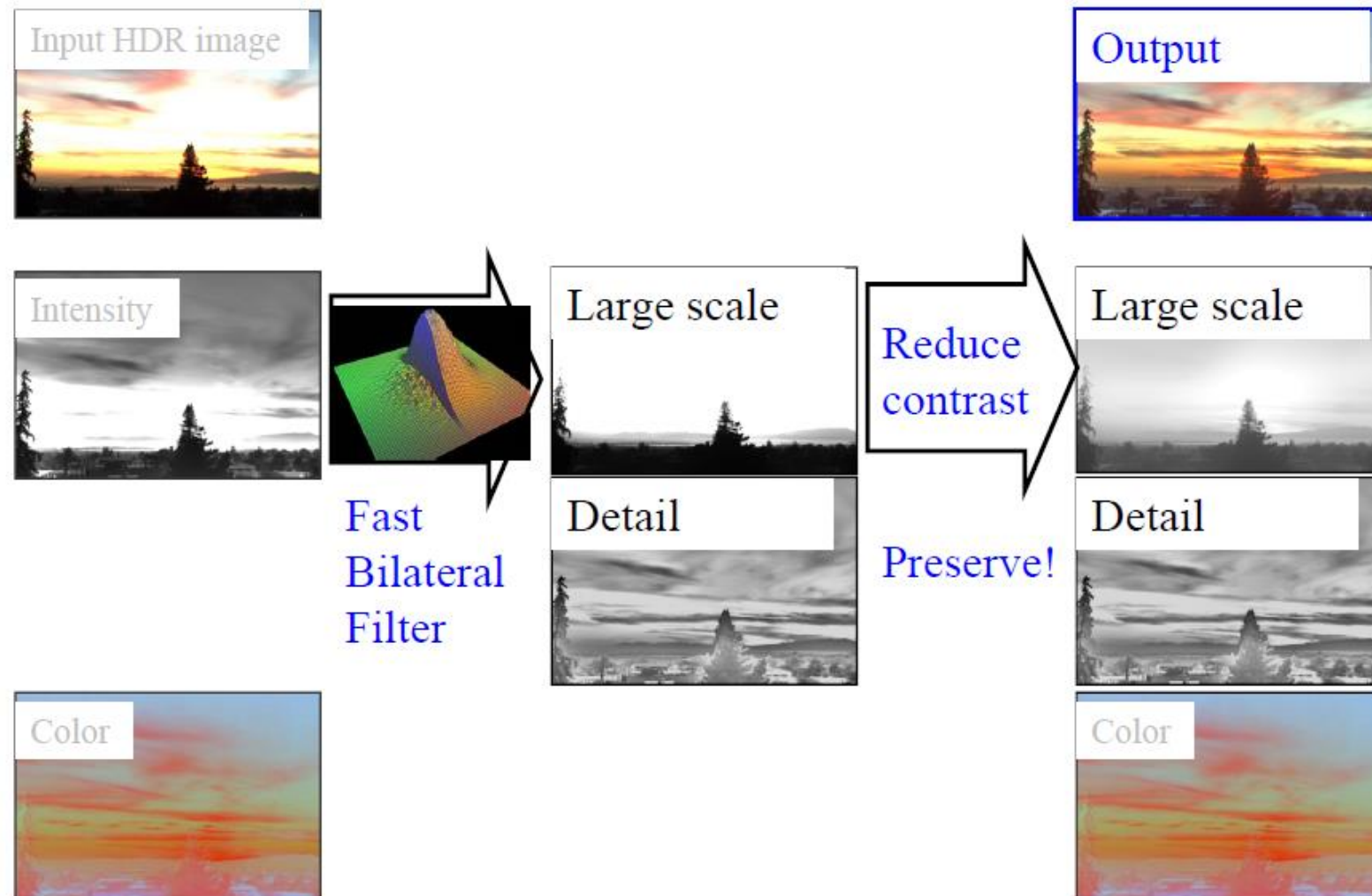
Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



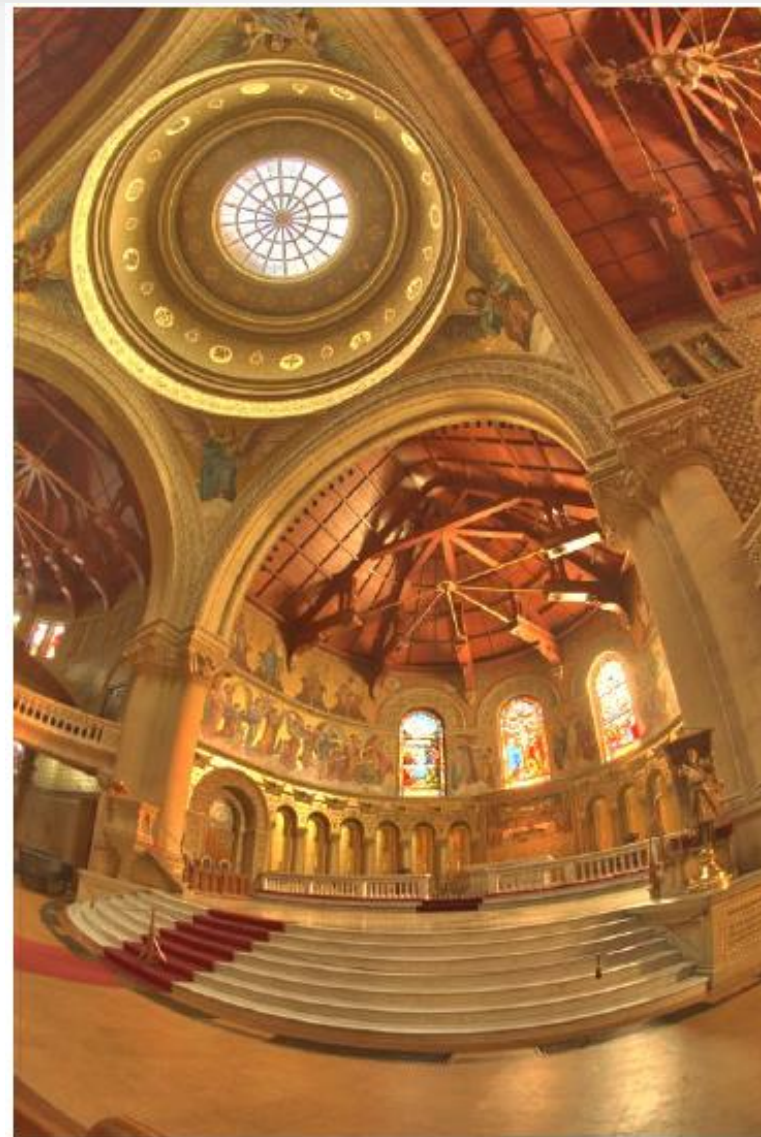
Tone Mapping by Bilateral Filtering

- Do not blur details by non-linear filtering (Bilateral filtering)



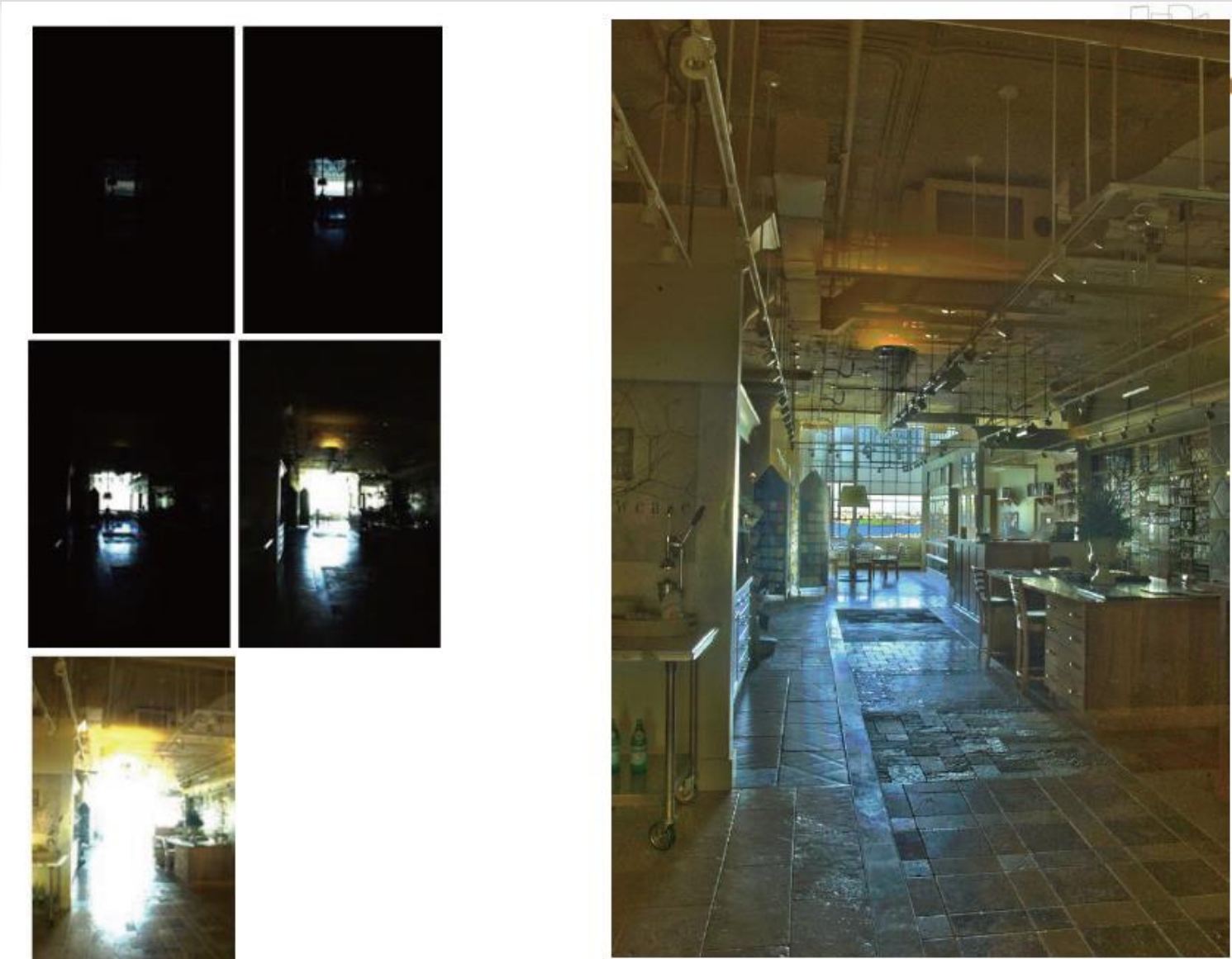
HDR Imaging results

- Tone mapped result of the four input images





HDR Imaging results





Acceleration

- Bilateral filtering is slow
- Fast algorithm: bilateral grid
 - <http://groups.csail.mit.edu/graphics/bilagrid/>
 - http://people.csail.mit.edu/sparis/publi/2009/ijcv/Paris_09_Fast_Approximation.pdf
 - <http://graphics.stanford.edu/papers/gkdtrees/>



HDR Storage

RGBE Format

- Denoted by “.pic” or “.hdr” extensions
(Developed by G. Ward for Radiance system)
- It is based on a 4 bytes representation
 - 8-bit mantissas (R, G, B)
 - Append to each pixel 8-bits for the common exponent
- Absolute accuracy of about 1%, covering a range of over 76 orders of magnitude
- Total 32-bit per pixel => 3 times smaller than RAW



HDR Storage

Other Formats

- LogLUV
- OpenEXR (.exr)
- Tiff (extension)
- Compressed formats

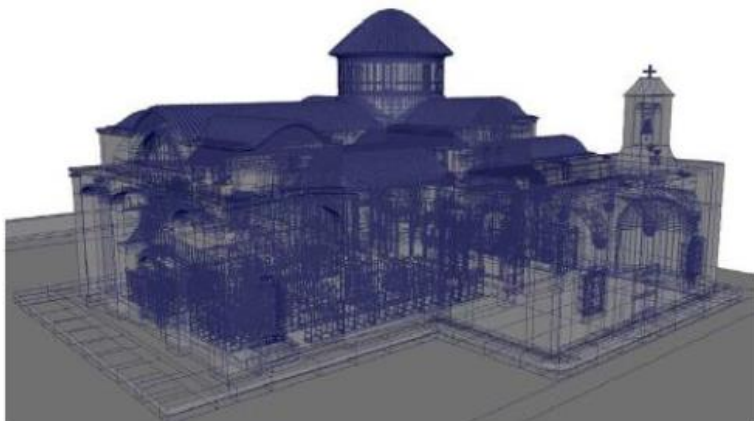


HDR Applications - IBL

Image Based Lighting (IBL)

- HDR images contain physical values of light
- Use HDRI to light virtual object
 - Commonly called augmented reality
- Used
 - Entertainment
 - Advertising
 - Design

IBL-Rendering with natural lighting

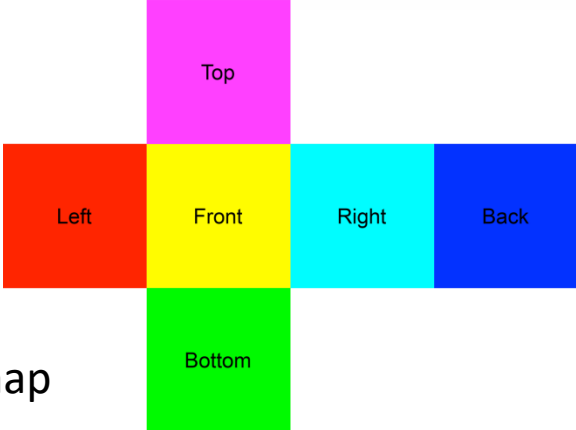




360 lighting map



equirectangular

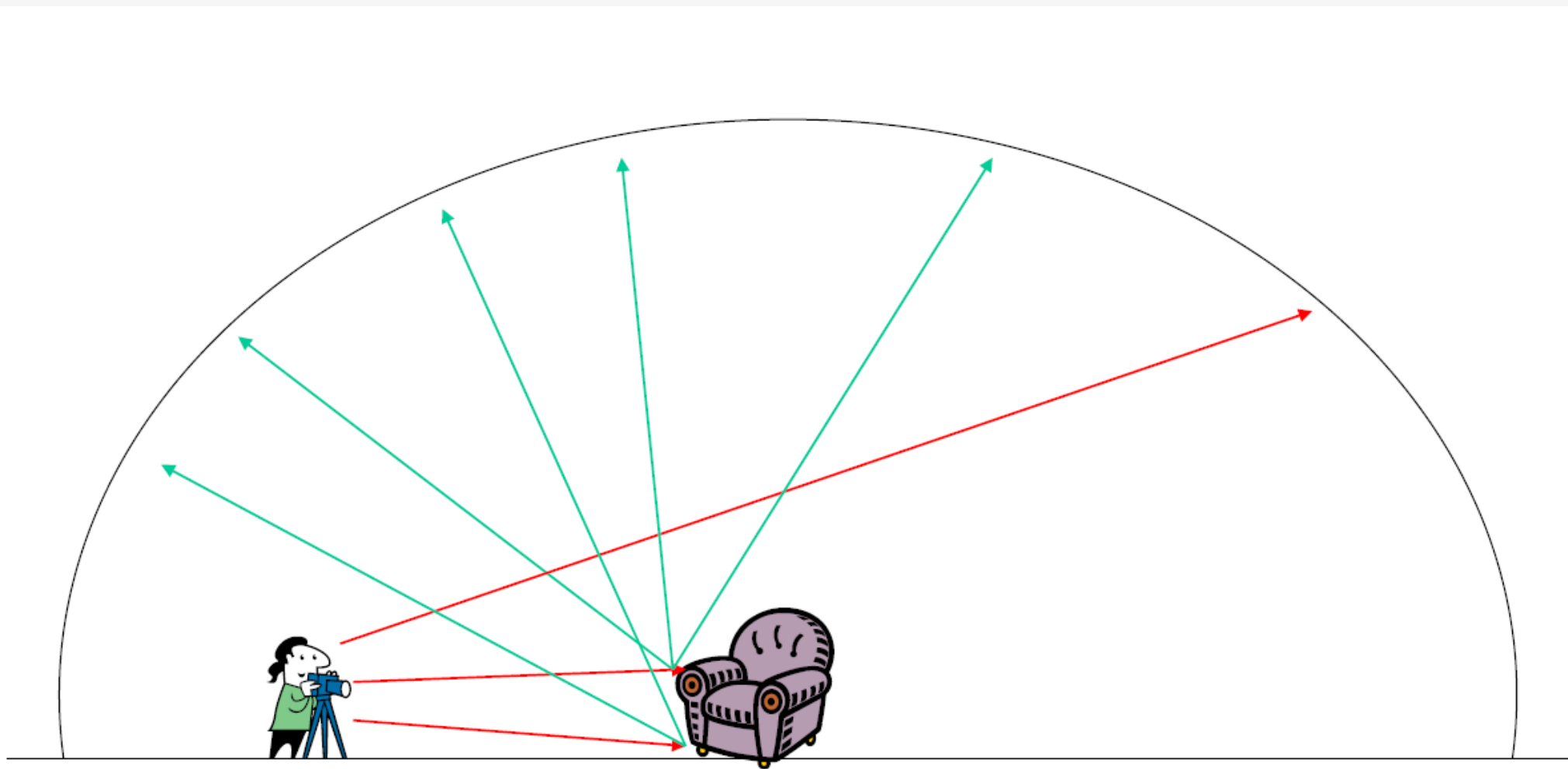


Cube map





IBL





Indoor Scene by IBL

