## Tone mapping

Digital Visual Effects

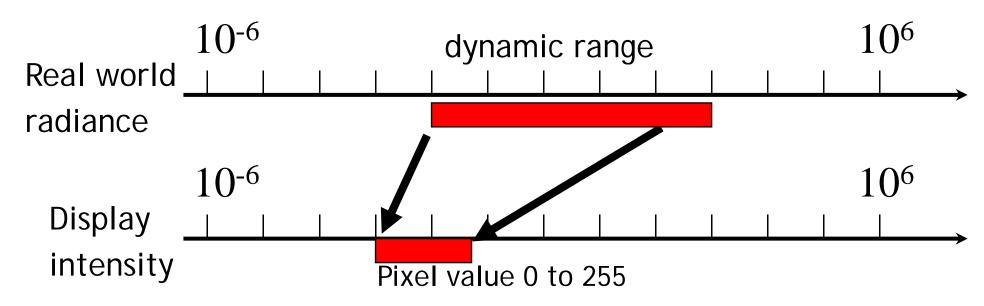
Yung-Yu Chuang

### Tone mapping

 How should we map scene luminances (up to 1:100,000) to display luminances (only around

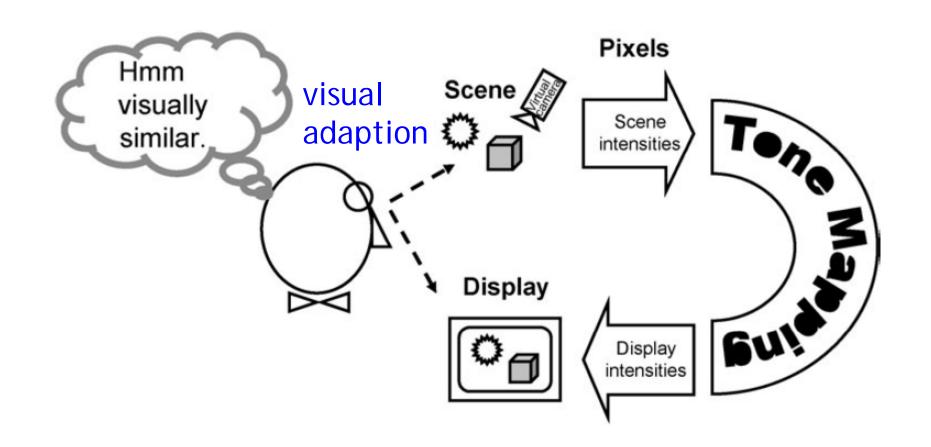
1:100) to produce a satisfactory image?

Linear scaling?, thresholding?



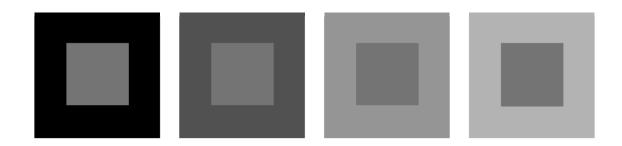
CRT has 300:1 dynamic range

## The ultimate goal is a visual match



We do not need to reproduce the true radiance as long as it gives us a visual match.

## Eye is not a photometer!



- Dynamic range along the visual pathway is only around 32:1.
- The key is adaptation

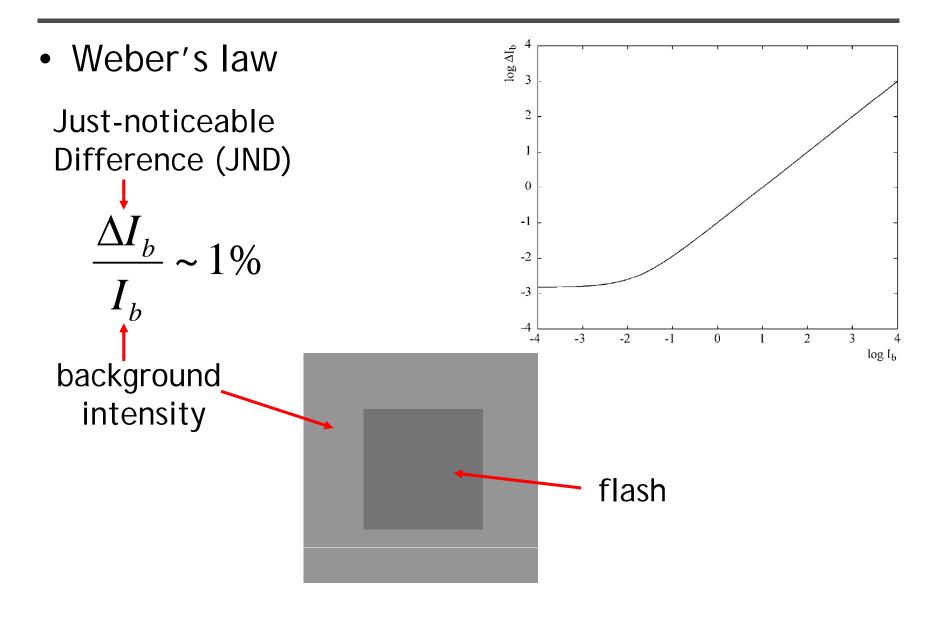
## Eye is not a photometer!





Are the headlights different in two images? Physically, they are the same, but perceptually different.

#### We are more sensitive to contrast



## How humans deal with dynamic range

- We're more sensitive to contrast (multiplicative)
  - A ratio of 1:2 is perceived as the same contrast as a ratio of 100 to 200
  - Makes sense because illumination has a multiplicative effect
  - Use the log domain as much as possible
- Dynamic adaptation (very local in retina)
  - Pupil (not so important)
  - Neural
  - Chemical
- Different sensitivity to spatial frequencies

#### **Preliminaries**

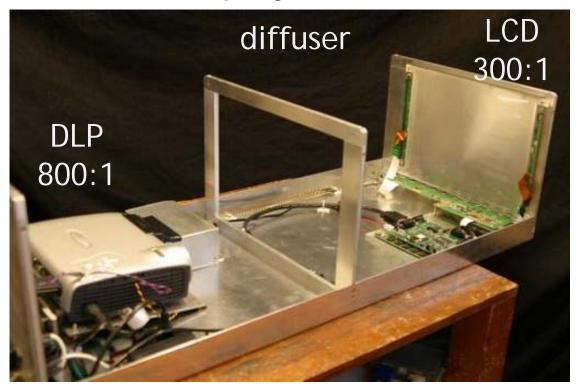
For color images

$$egin{bmatrix} R_d \ G_d \ B_d \end{bmatrix} = egin{bmatrix} L_d \, rac{R_w}{L_w} \ L_d \, rac{G_w}{L_w} \ L_d \, rac{B_w}{L_w} \ \end{pmatrix}$$

Log domain is usually preferred.

## **HDR Display**

 Once we have HDR images (either captured or synthesized), how can we display them on normal displays?



Theoretically, 240,000:1.

Due to imperfect optical depth, 54,000:1 measured

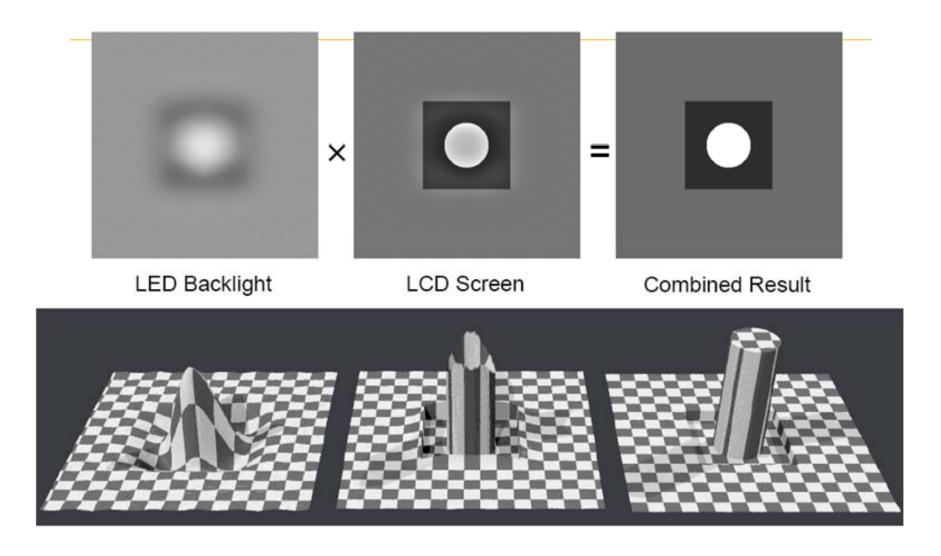
HDR display system, Sunnybrook Technology, SIGGRAPH2004

## Sunnybrook HDR display

- Use Bright Source + Two 8-bit Modulators
  - Transmission multiplies together
  - Over 10,000:1 dynamic range possible



## How it works



## Brightside HDR display



37" 200000:1

Acquired by Dolby

## Tone mapping operators

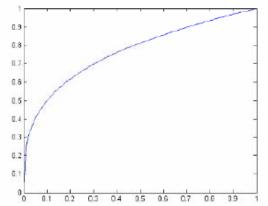
- Spatial (global/local)
- Frequency domain
- Gradient domain
- 3 papers from SIGGRAPH 2002
  - Photographic Tone Reproduction for Digital Images
  - Fast Bilateral Filtering for the Display of High-Dynamic-Range Images
  - Gradient Domain High Dynamic Range Compression

## Photographic Tone Reproduction for Digital Images

Erik Reinhard Mike Stark
Peter Shirley Jim Ferwerda
SIGGRAPH 2002

## Global v.s. local

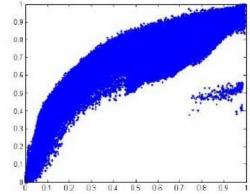






Example : Gamma Compression





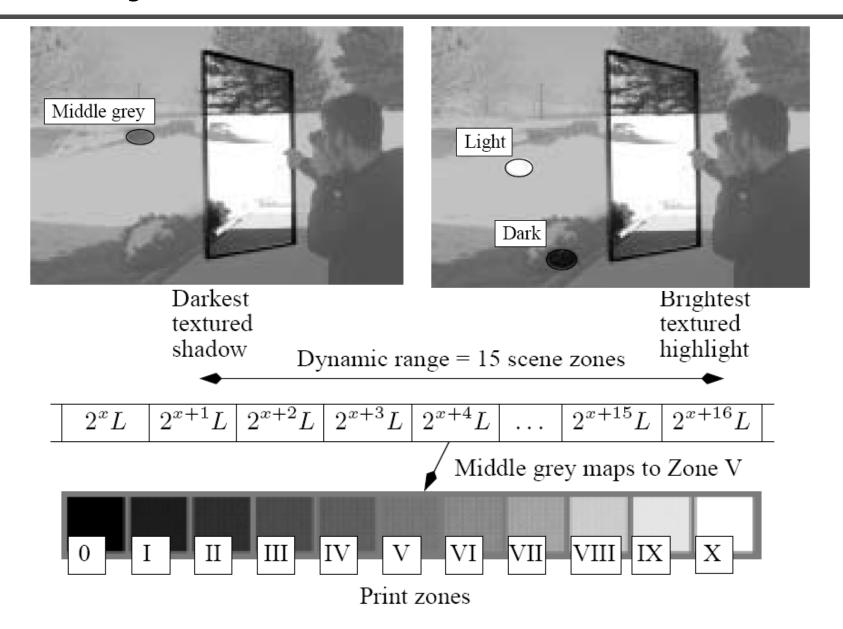


Example : Adaptive Histogram Equalization

## Photographic tone reproduction

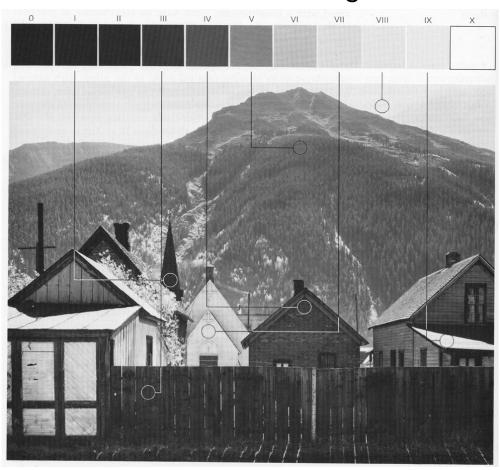
- Proposed by Reinhard et. al. in SIGGRAPH 2002
- Motivated by traditional practice, zone system by Ansel Adams and dodging and burning
- It contains both global and local operators

## Zone system



## The Zone system

- Formalism to talk about exposure, density
- Zone = intensity range, in powers of two
- In the scene, on the negative, on the print



Source: Ansel Adams

#### The Zones

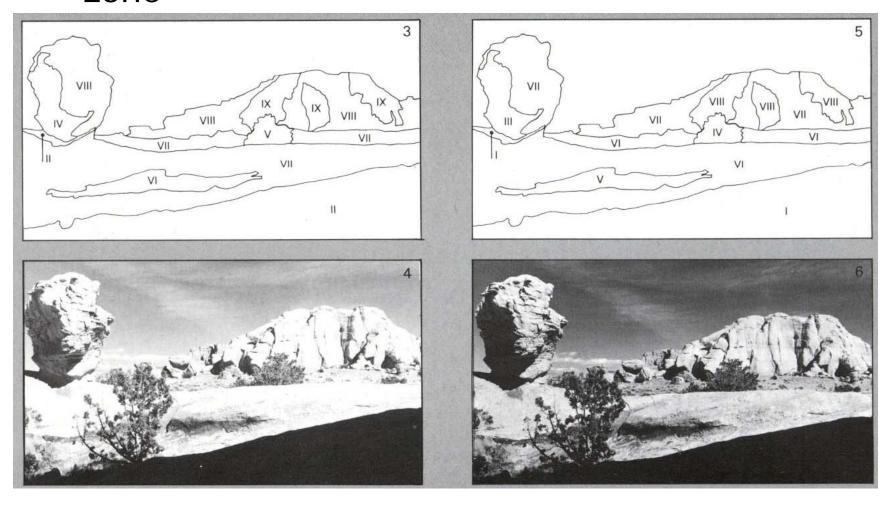
and buildings

#### V Middle grey: the pivot The Zones value; light foliage, dark skin O Solid black: the same as . the film rebate VI Caucasian skin, textured light grey; shadow on snow I Nearly black; just different from Zone 0 VII Light skin; bright areas with texture, such as snow in low sunlight **II** The first hint of texture **III** Textured shadow: the VIII Highest zone with any first recognizable shadow texture detail IV Average shadow value on Caucasian skin, foliage

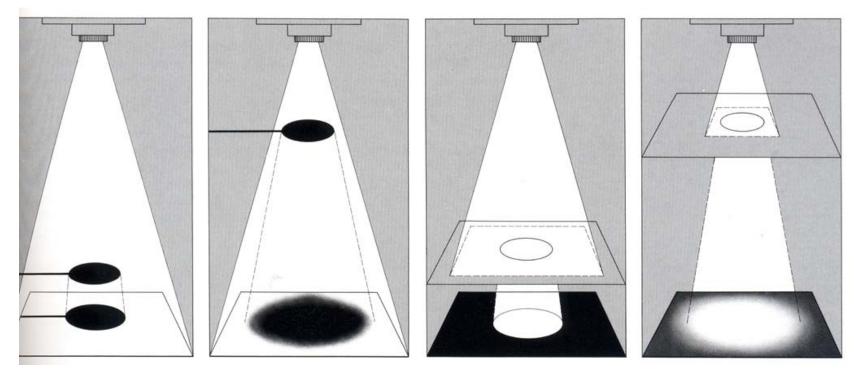
IX Pure untextured white

## The Zone system

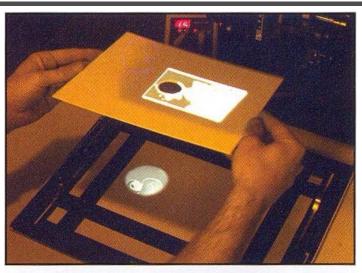
 You decide to put part of the system in a given zone

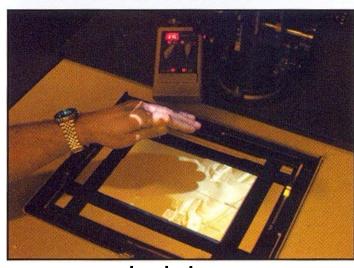


- During the print
- Hide part of the print during exposure
  - Makes it brighter









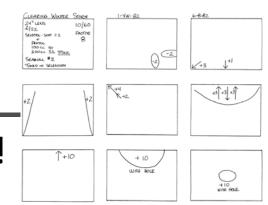


dodging

burning

From Photography by London et al.

• Must be done for every single print!







Straight print

After dodging and burning

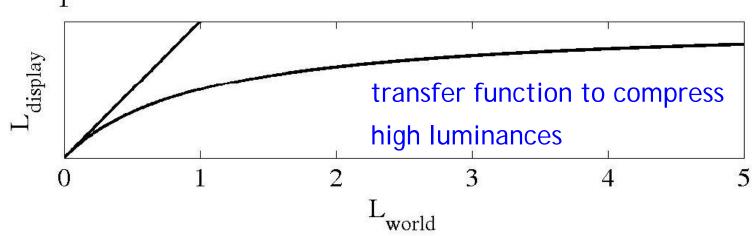
## Global operator

$$\overline{L}_{w} = \exp\left(\frac{1}{N}\sum_{x,y}\log(\delta + L_{w}(x,y))\right) \text{ key (how light or dark it is).}$$
Map to 18% of display range

Approximation of scene's for average-key scene

User-specified; high key or low key

$$L_{m}(x,y) = \frac{a}{\overline{L}_{w}} L_{w}(x,y) \qquad L_{d}(x,y) = \frac{L_{m}(x,y)}{1 + L_{m}(x,y)}$$



## Global operator

It seldom reaches 1 since the input image does not have infinitely large luminance values.

$$L_{d}(x,y) = \frac{L_{m}(x,y)\left(1 + \frac{L_{m}(x,y)}{L_{white}^{2}(x,y)}\right)}{1 + L_{m}(x,y)}$$

$$L_{white} = 0.5 \quad 1.0 \quad 1.5 \qquad 3 \qquad \infty$$

$$L_{d} \qquad \qquad L_{white} \text{ is the smallest luminance}$$

$$to be mapped to 1$$

$$0 \qquad 1 \qquad 2 \qquad 3 \qquad 4 \qquad 5$$

$$World luminance (L)$$



low key (0.18)

high key (0.5)

## Dodging and burning (local operators)

- Area receiving a different exposure is often bounded by sharp contrast
- Find largest surrounding area without any sharp contrast

$$L_s^{blur}(x,y) = L_m(x,y) \otimes G_s(x,y)$$

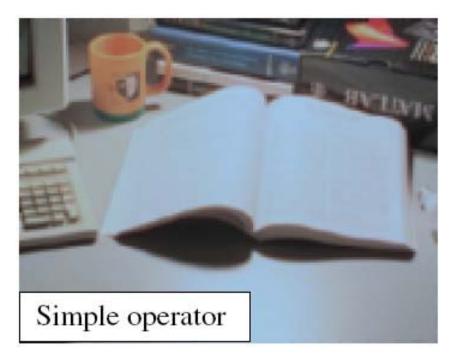
$$V_{s}(x, y) = \frac{L_{s}^{blur}(x, y) - L_{s+1}^{blur}(x, y)}{2^{\phi} a/s^{2} + L_{s}^{blur}}$$

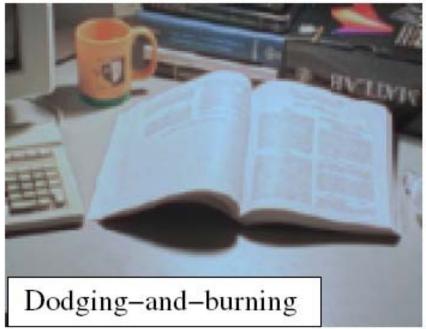
$$s_{\max}: |V_{s_{\max}}(x, y)| < \varepsilon$$

## Dodging and burning (local operators)

$$L_d(x, y) = \frac{L_m(x, y)}{1 + L_{s_{\text{max}}}^{blur}(x, y)}$$

- A darker pixel (smaller than the blurred average of its surrounding area) is divided by a larger number and become darker (dodging)
- A brighter pixel (larger than the blurred average of its surrounding area) is divided by a smaller number and become brighter (burning)
- Both increase the contrast





## Frequency domain

- First proposed by Oppenheim in 1968!
- Under simplified assumptions,

image

= illuminance \* reflectance

low-frequency high-frequency attenuate more attenuate less







## Oppenheim

- Taking the logarithm to form density image
- Perform FFT on the density image
- Apply frequency-dependent attenuation filter

$$s(f) = (1-c) + c\frac{kf}{1+kf}$$

- Perform inverse FFT
- Take exponential to form the final image

# Fast Bilateral Filtering for the Display of High-Dynamic-Range Images

Frédo Durand & Julie Dorsey

SIGGRAPH 2002

## A typical photo

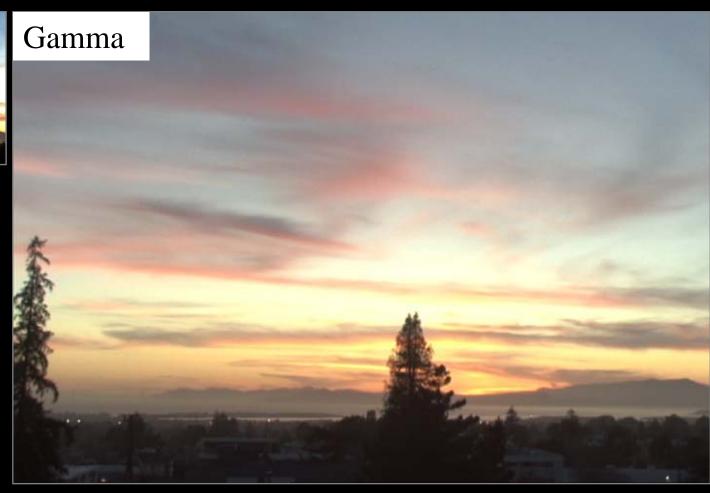
- Sun is overexposed
- Foreground is underexposed



## Gamma compression

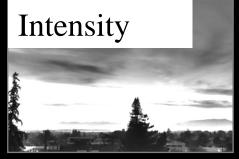
- $X \rightarrow X^{\gamma}$
- Colors are washed-out

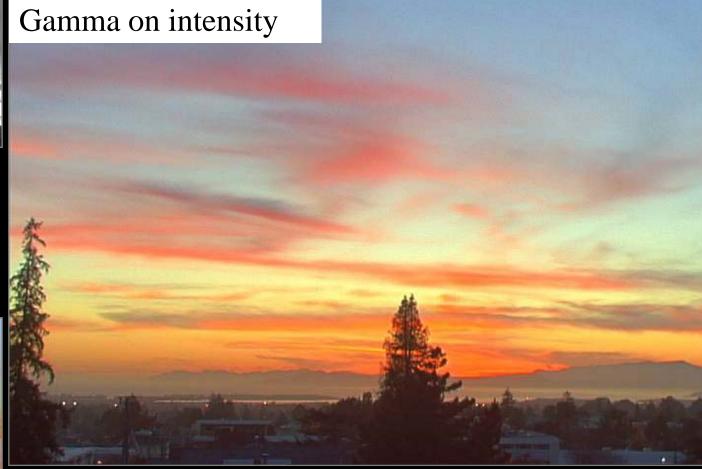




## Gamma compression on intensity

 Colors are OK, but details (intensity highfrequency) are blurred





Color

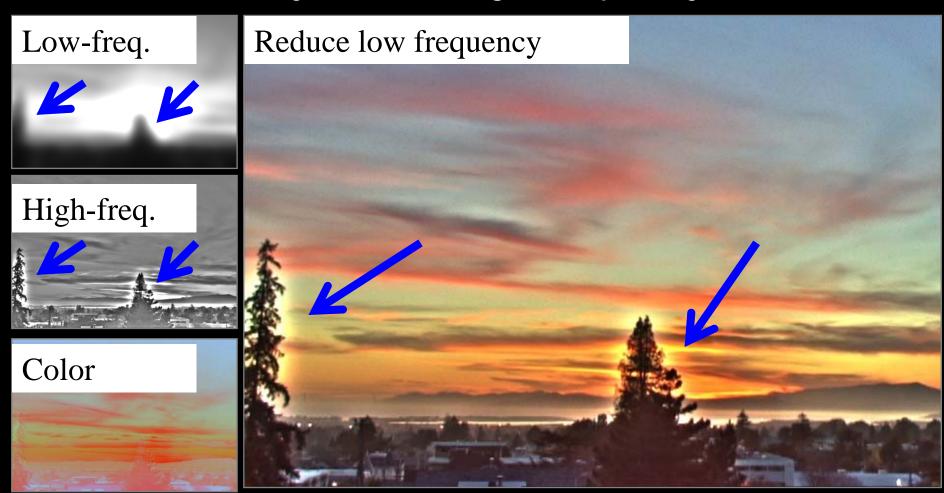
## Chiu et al. 1993

- Reduce contrast of low-frequencies
- Keep high frequencies



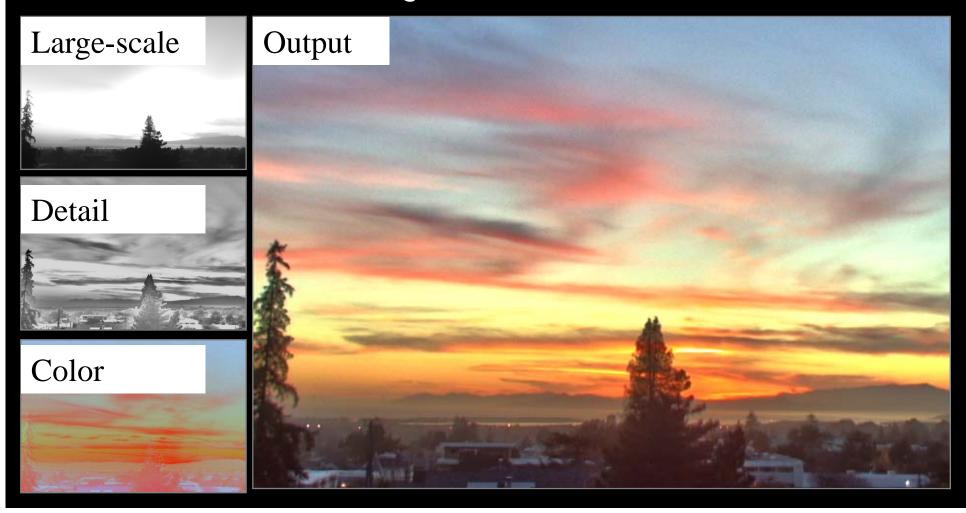
# The halo nightmare

- For strong edges
- Because they contain high frequency



## **Durand and Dorsey**

- Do not blur across edges
- Non-linear filtering



## Edge-preserving filtering

Blur, but not across edges



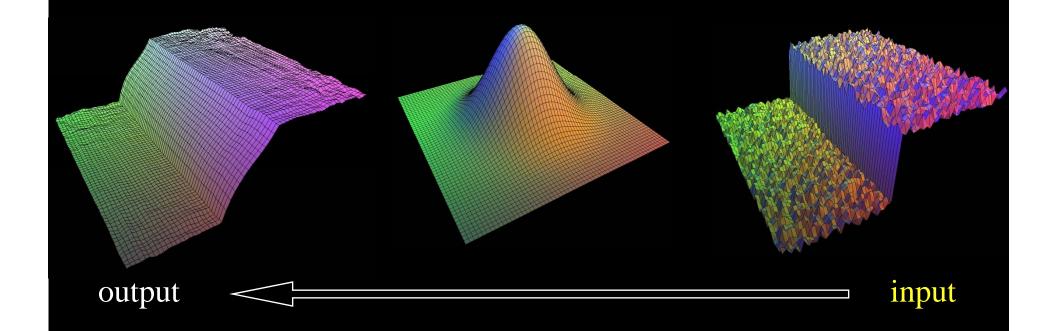




- Anisotropic diffusion [Perona & Malik 90]
  - Blurring as heat flow
  - LCIS [Tumblin & Turk]
- Bilateral filtering [Tomasi & Manduci, 98]

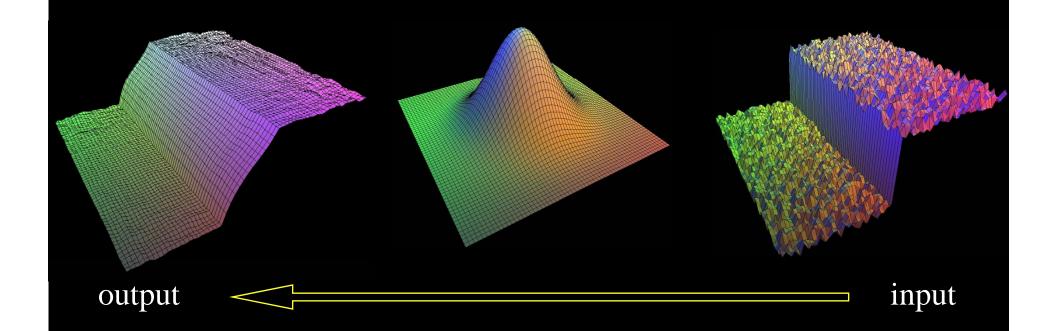
## Start with Gaussian filtering

• Here, input is a step function + noise



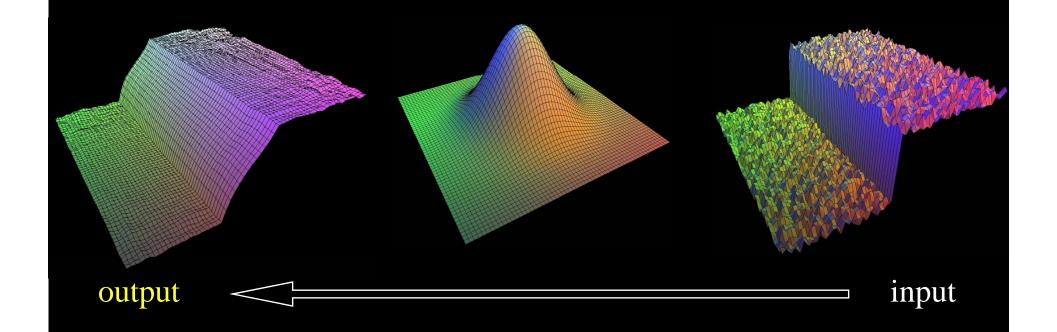
# Start with Gaussian filtering

• Spatial Gaussian f

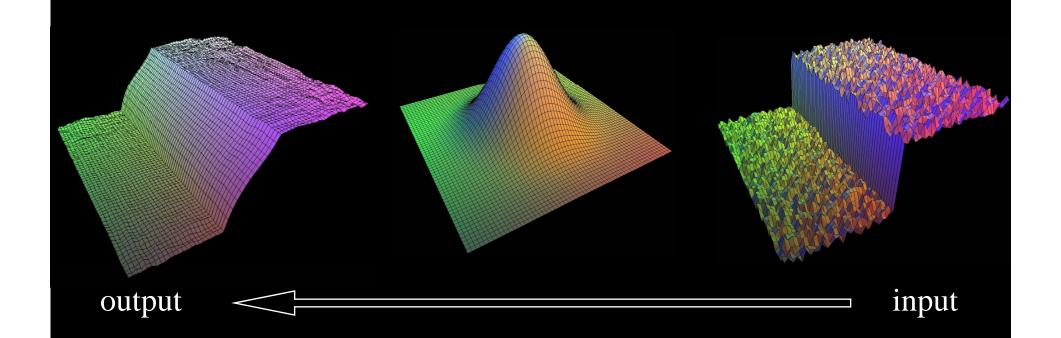


# Start with Gaussian filtering

Output is blurred

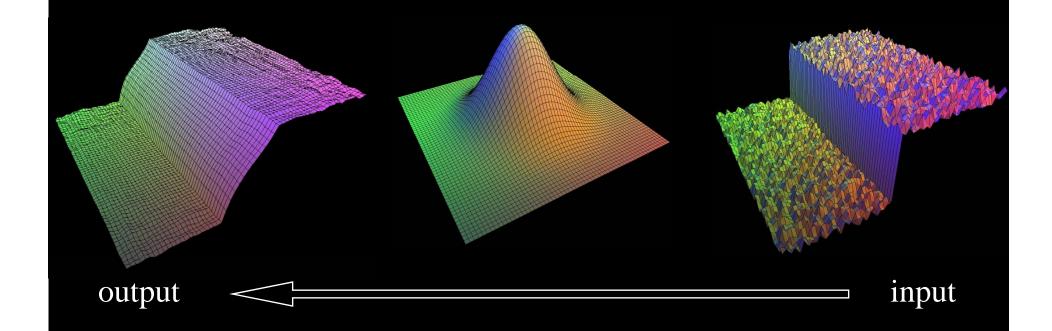


# Gaussian filter as weighted average



## The problem of edges

- Here, "pollutes" our estimate J(x)
- It is too different



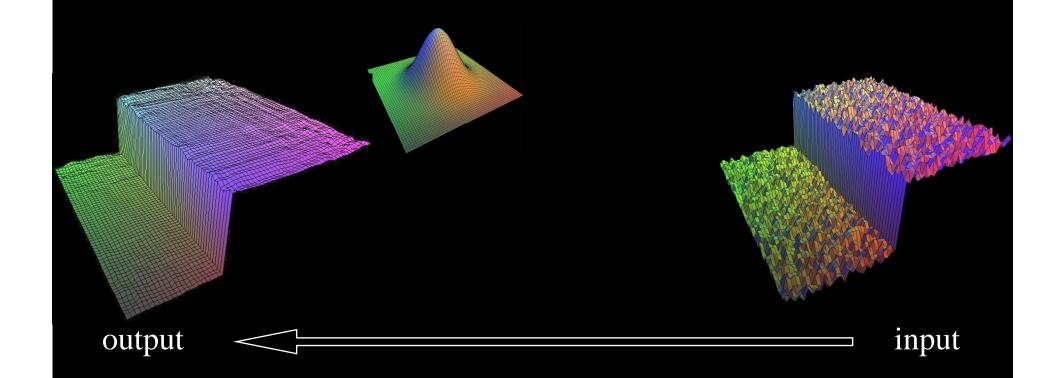
## Principle of Bilateral filtering

- [Tomasi and Manduchi 1998]
- Penalty g on the intensity difference



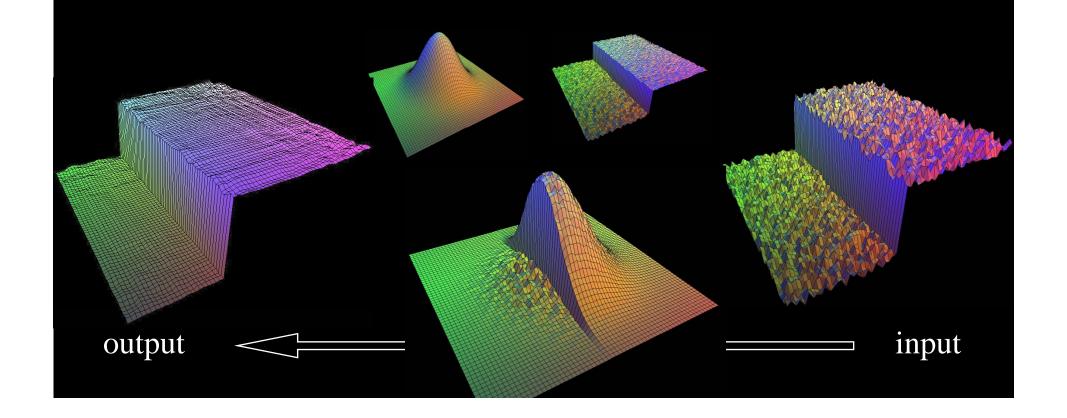
## Bilateral filtering

- [Tomasi and Manduchi 1998]
- Spatial Gaussian f



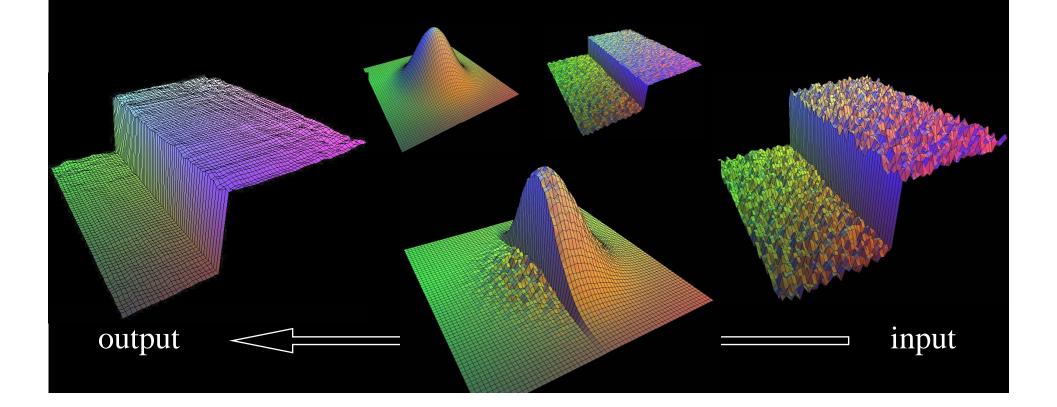
## Bilateral filtering

- [Tomasi and Manduchi 1998]
- Spatial Gaussian f
- Gaussian g on the intensity difference



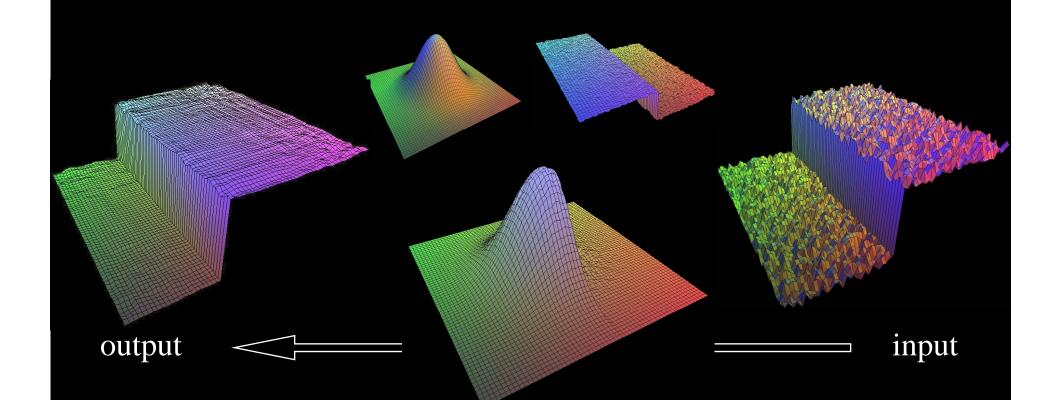
# Normalization factor

- [Tomasi and Manduchi 1998]
- k(x)=



## Bilateral filtering is non-linear

- [Tomasi and Manduchi 1998]
- The weights are different for each output pixel





Contrast too high!

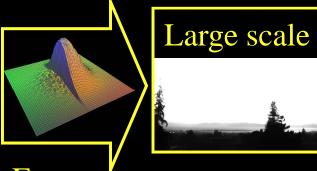










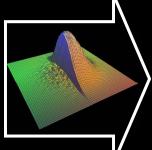


Fast
Bilateral
Filter







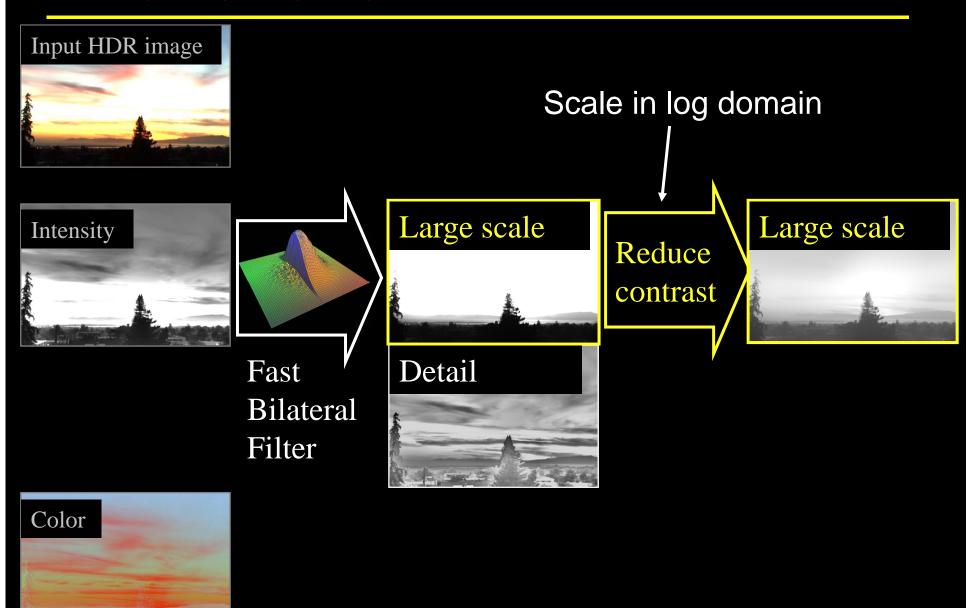


Fast
Bilateral
Filter



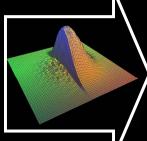












Fast
Bilateral
Filter







Reduce contrast

Preserve!

Large scale



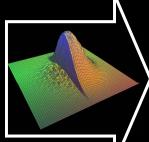


















Fast
Bilateral
Filter



Preserve!

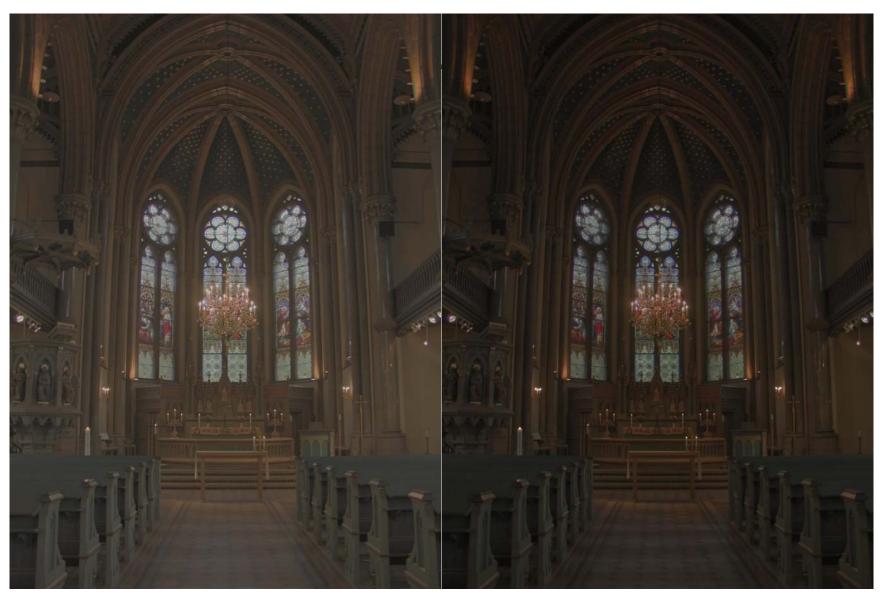
Detail





#### Bilateral filter is slow!

- Compared to Gaussian filtering, it is much slower because the kernel is not fixed.
- Durand and Dorsey proposed an approximate approach to speed up
- Paris and Durand proposed an even-faster approach in ECCV 2006. We will cover this one when talking about computational photogrphy.



Oppenheim

bilateral

# Gradient Domain High Dynamic Range Compression

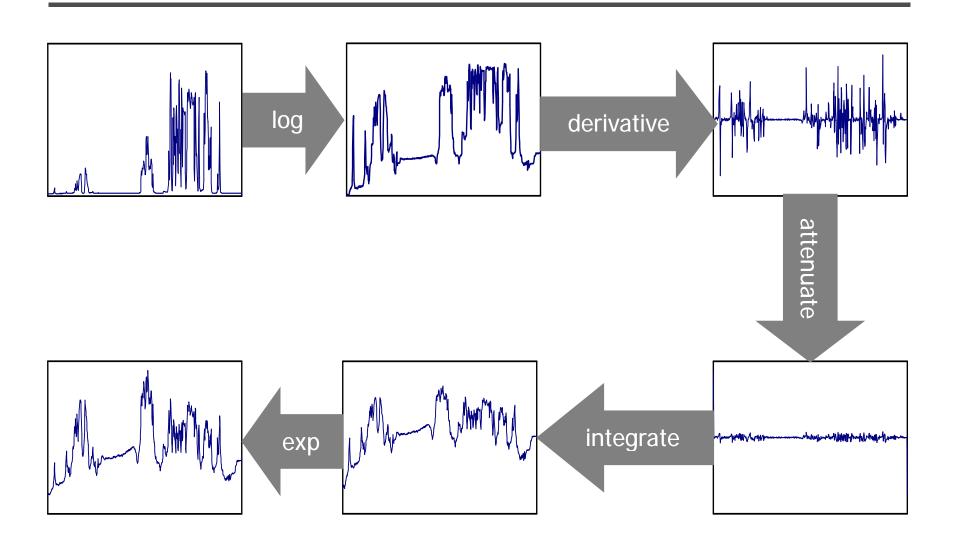
Raanan Fattal Dani Lischinski Michael Werman

SIGGRAPH 2002

## Log domain

- Logorithm is a crude approximation to the perceived brightness
- Gradients in log domain correspond to ratios (local contrast) in the luminance domain

#### The method in 1D



#### The method in 2D

- Given: a log-luminance image H(x,y)
- Compute an *attenuation map*  $\Phi(\!|\!| \nabla H \!|\!|)$
- Compute an attenuated gradient field G:

$$G(x, y) = \nabla H(x, y) \cdot \Phi(\|\nabla H\|)$$

Problem: G may not be integrable!

#### Solution

- Look for image I with gradient closest to G in the least squares sense.
- $\emph{\textbf{I}}$  minimizes the integral:  $\iint F(\nabla I,G) dxdy$

$$F(\nabla I, G) = \|\nabla I - G\|^2 = \left(\frac{\partial I}{\partial x} - G_x\right)^2 + \left(\frac{\partial I}{\partial y} - G_y\right)^2$$

$$\frac{\partial^2 I}{\partial x^2} + \frac{\partial^2 I}{\partial y^2} = \frac{\partial G_x}{\partial x} + \frac{\partial G_y}{\partial y}$$
 Poisson equation

Solve 
$$\frac{\partial^{2} I}{\partial x^{2}} + \frac{\partial^{2} I}{\partial y^{2}} = \frac{\partial G_{x}}{\partial x} + \frac{\partial G_{y}}{\partial y}$$

$$\int_{G_{x}} (x, y) - G_{x}(x - 1, y) + G_{y}(x, y) - G_{y}(x, y - 1)$$

$$I(x + 1, y) + I(x - 1, y) + I(x, y + 1) + I(x, y - 1) - 4I(x, y)$$

$$\begin{bmatrix} ... & 1... & 1.41... & 1... \\ I & = 1... & 1... \\ I & = 1... & 1... & 1... \\ I & = 1... & 1.$$

## Solving Poisson equation

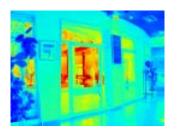
- No analytical solution
- Multigrid method
- Conjugate gradient method

#### **Attenuation**

- Any dramatic change in luminance results in large luminance gradient at some scale
- Edges exist in multiple scales. Thus, we have to detect and attenuate them at multiple scales
- Construct a Gaussian pyramid  $H_i$

# Attenuation $\varphi_k(x, y) = \left(\frac{\|\nabla H_k(x, y)\|}{\alpha}\right)^{\beta - 1} \beta \sim 0.8$ $\alpha = 0.1 \overline{\nabla H}$















gradient magnitude





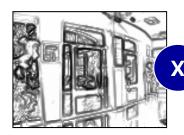


attenuation map

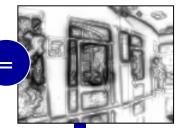
# Multiscale gradient attenuation



interpolate







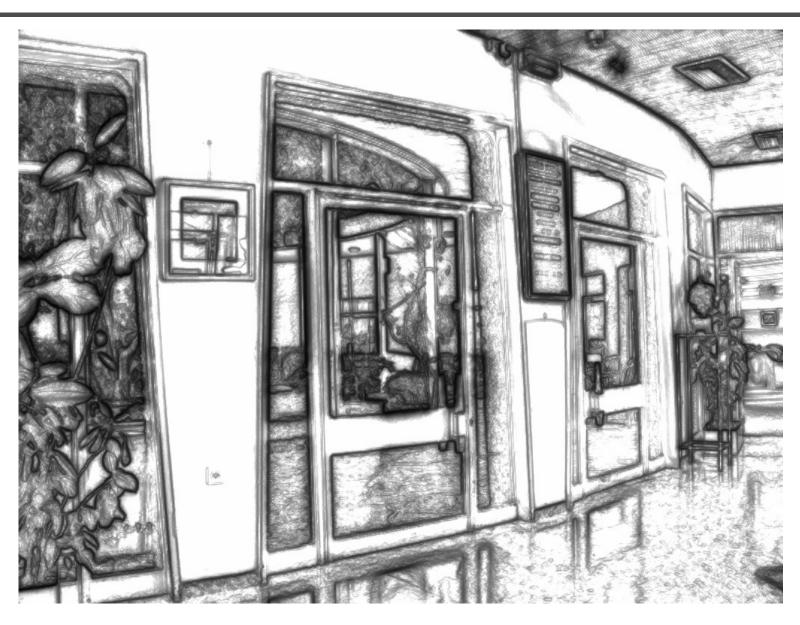
interpolate







# Final gradient attenuation map

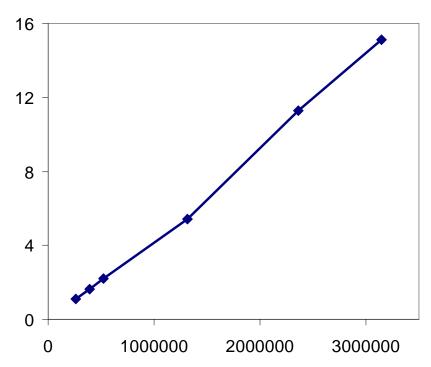


#### Performance

• Measured on 1.8 GHz Pentium 4:

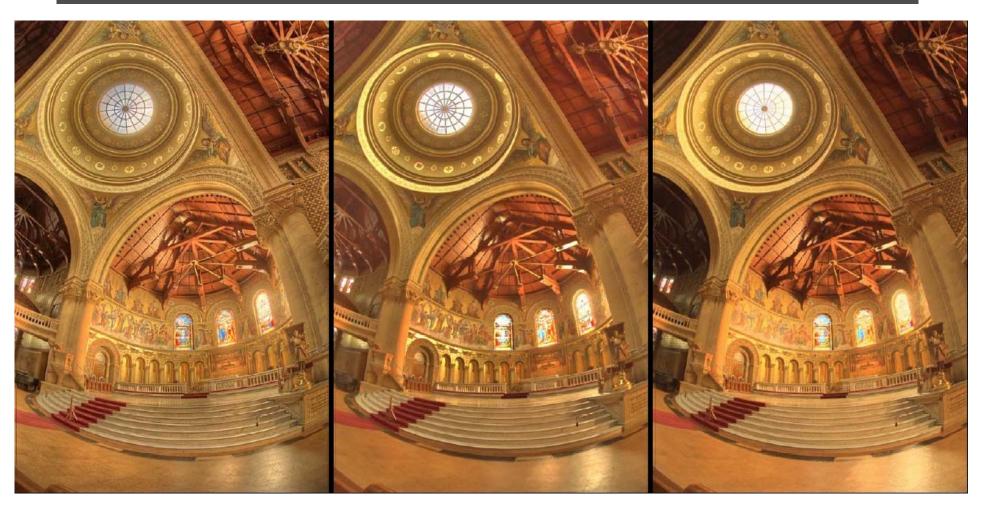
- 512 x 384: 1.1 sec

- 1024 x 768: 4.5 sec



 Can be accelerated using processor-optimized libraries.

## Informal comparison



Gradient domain [Fattal et al.]

Bilateral [Durand et al.]

Photographic [Reinhard et al.]

## Informal comparison



Gradient domain [Fattal et al.]

Bilateral [Durand et al.]

Photographic [Reinhard et al.]

## Informal comparison



Gradient domain [Fattal et al.]

Bilateral [Durand et al.]

Photographic [Reinhard et al.]

# Evaluation of Tone Mapping Operators using a High Dynamic Range Display

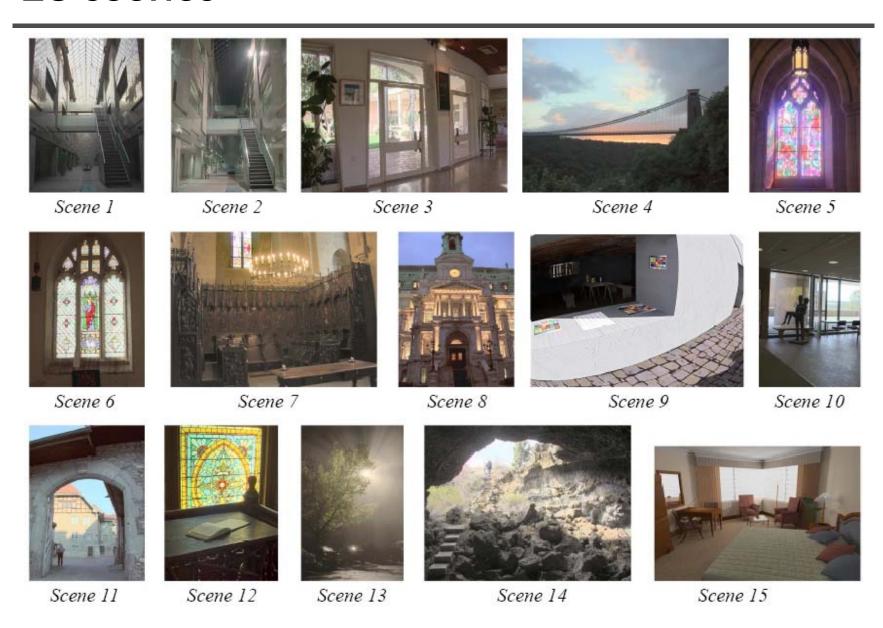
Patrick Ledda Alan Chalmers Tom Troscinko Helge Seetzen

SIGGRAPH 2005

## Six operators

- H: histogram adjustment
- B: bilateral filter
- P: photographic reproduction
- I: iCAM
- L: logarithm mapping
- A: local eye adaption

#### 23 scenes



## **Experiment setting**



#### Preference matrix

- Ranking is easier than rating.
- 15 pairs for each person to compare. A total of 345 pairs per subject.

	$tmo_1$	$tmo_2$	tmo <sub>3</sub>	$tmo_4$	tmo <sub>5</sub>	tmo <sub>6</sub>	Score
$tmo_1$	_	1	0	0	1	1	3
$tmo_2$	0	-	0	1	1	0	2
tmo <sub>3</sub>	1	1	-	1	1	1	5
tmo <sub>4</sub>	1	0	0	-	0	0	1
tmo <sub>5</sub>	0	0	0	1	-	1	2
tmo <sub>6</sub>	0	1	0	1	0	-	2

preference matrix (tmo2->tmo4, tom2 is better than tmo4)

#### Statistical measurements

- Statistical measurements are used to evaluate:
  - Agreement: whether most agree on the ranking between two tone mapping operators.
  - Consistency: no cycle in ranking. If all are confused in ranking some pairs, it means they are hard to compare. If someone is inconsistent alone, his ranking could be droped.

# Overall similarity

• Scene 8



	P	H	В	L	I	A	Total
P	_	24	46	42	10	32	154
H	24	-	44	32	8	12	120
В	2	4	-	8	2	4	20
L	6	16	40	-	4	12	78
I	38	40	46	44	-	38	206
A	16	36	44	36	10	-	142

## Summary

3712

823

```
      Overall Similarity: Color

      P
      H
      A
      L
      B

      3402
      2994
      2852
      1902
      1696

      Bright Detail

      A
      P
      H
      B
      L

      688
      569
      549
      474
      347
```

#### Dark Detail

P	A	I	L	H	B
815	793	583	491	485	283

## Not settled yet!

- Some other experiment said bilateral are better than others.
- For your reference, photographic reproduction performs well in both reports.
- There are parameters to tune and the space could be huge.

#### References

- Raanan Fattal, Dani Lischinski, Michael Werman, <u>Gradient Domain High Dynamic Range Compression</u>, SIGGRAPH 2002.
- Fredo Durand, Julie Dorsey, <u>Fast Bilateral Filtering for the Display of High Dynamic Range Images</u>, SIGGRAPH 2002.
- Erik Reinhard, Michael Stark, Peter Shirley, Jim Ferwerda, <u>Photographics Tone Reproduction for Digital</u> <u>Images</u>, SIGGRAPH 2002.
- Patrick Ledda, Alan Chalmers, Tom Troscianko, Helge Seetzen, <u>Evaluation of Tone Mapping Operators using a</u> <u>High Dynamic Range Display</u>, SIGGRAPH 2005.
- Jiangtao Kuang, Hiroshi Yamaguchi, Changmeng Liu, Garrett Johnson, Mark Fairchild, <u>Evaluating HDR</u> <u>Rendering Algorithms</u>, ACM Transactions on Applied Perception, 2007.