Tone mapping

Digital Visual Effects, Spring 2009

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2009/3/5

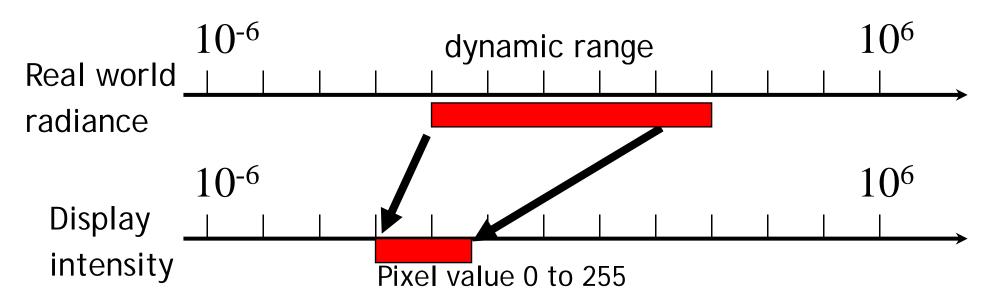
with slides by Fredo Durand, and Alexei Efros

Tone mapping

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

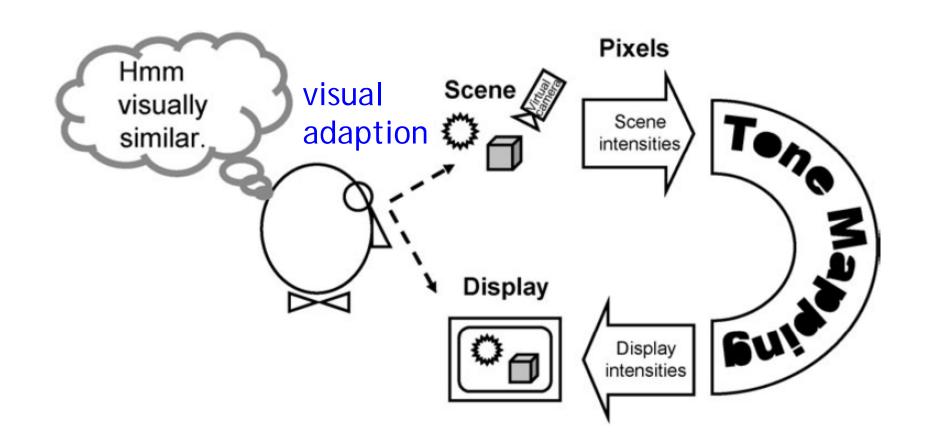
1:100) to produce a satisfactory image?

Linear scaling?, thresholding?



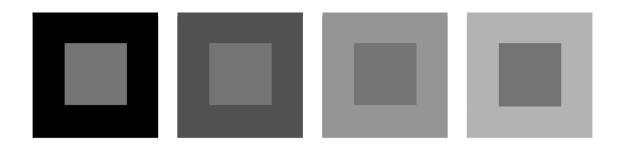
CRT has 300:1 dynamic range

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

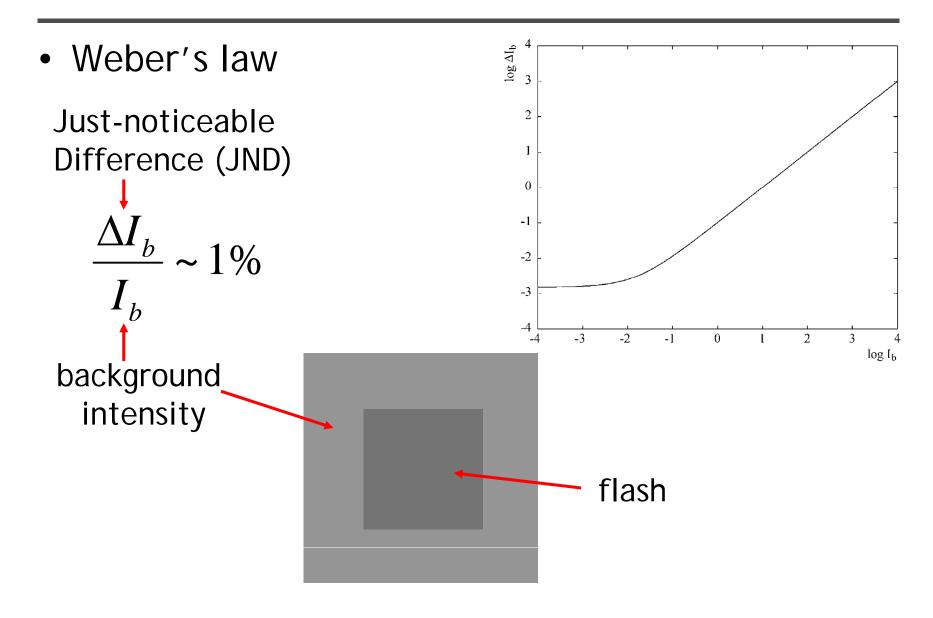
Eye is not a photometer!





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

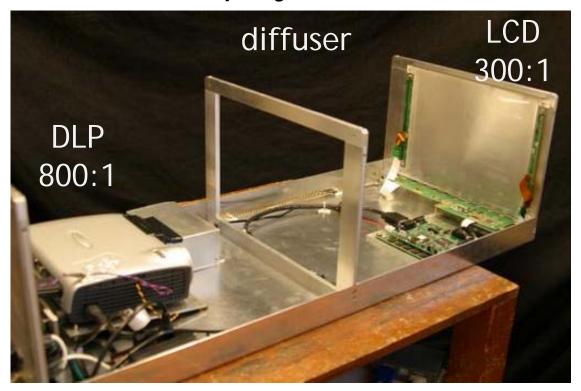
For color images

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

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

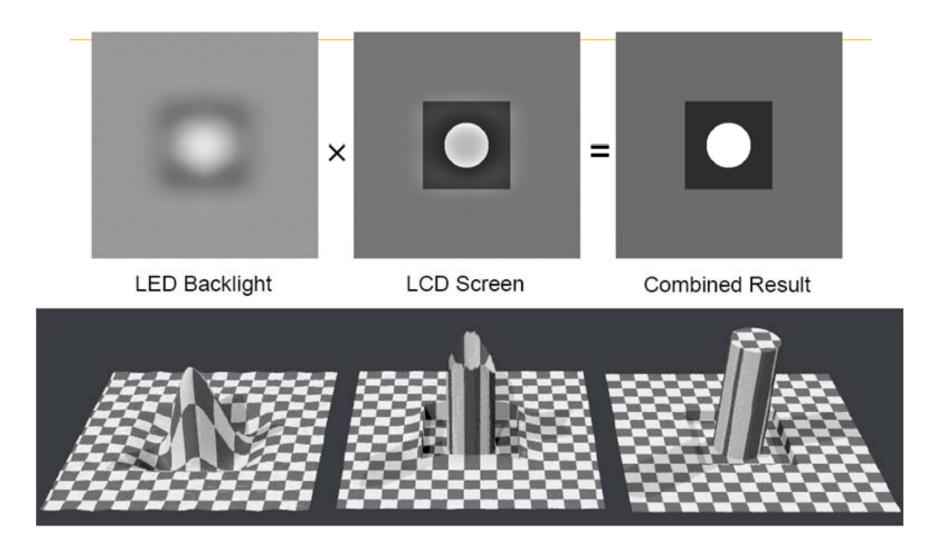
HDR display system, Sunnybrook Technology, SIGGRAPH2004

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

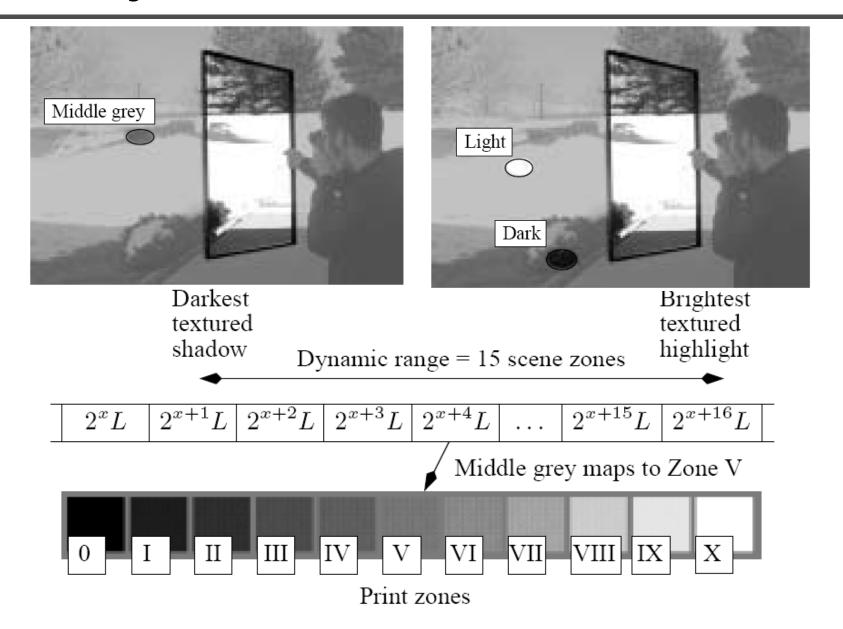
Photographic Tone Reproduction for Digital Images

Erik Reinhard Mike Stark
Peter Shirley Jim Ferwerda
SIGGRAPH 2002

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

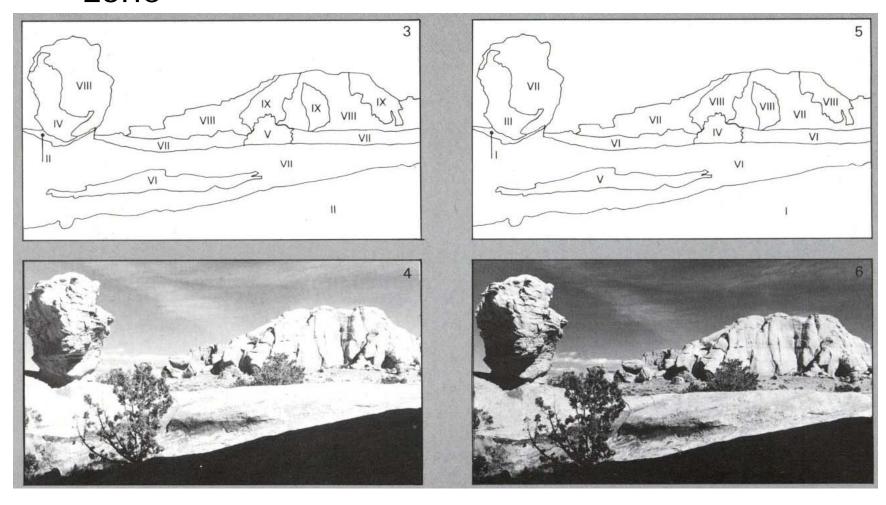
and buildings

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

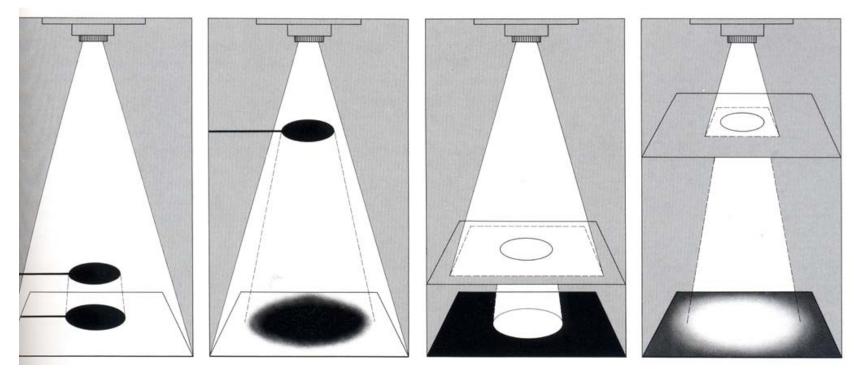
IX Pure untextured white

The Zone system

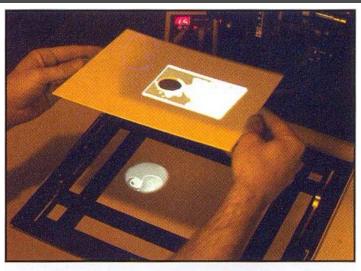
 You decide to put part of the system in a given zone



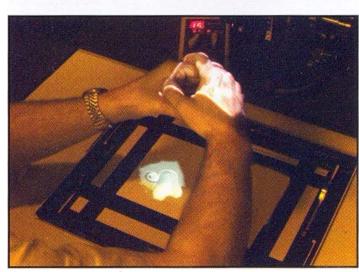
- During the print
- Hide part of the print during exposure
 - Makes it brighter









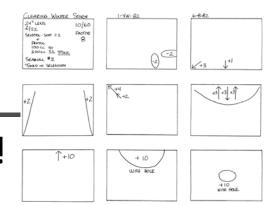


dodging

burning

From Photography by London et al.

• Must be done for every single print!







Straight print

After dodging and burning

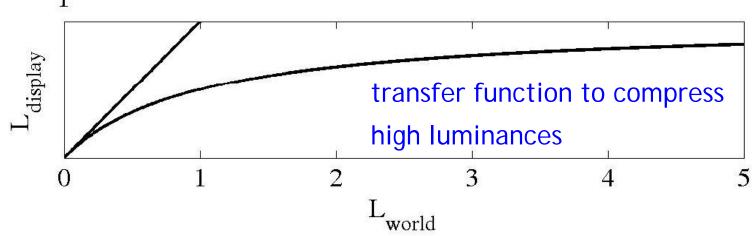
Global operator

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

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

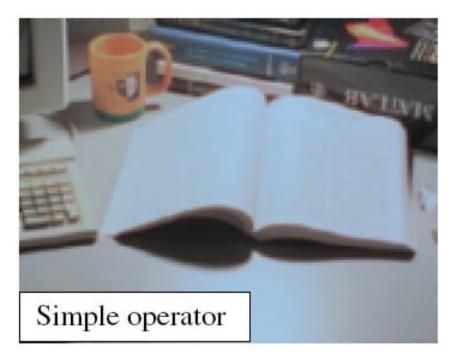
$$V_{s}(x, y) = \frac{L_{s}^{blur}(x, y) - L_{s+1}^{blur}(x, y)}{2^{\phi} a/s^{2} + L_{s}^{blur}}$$

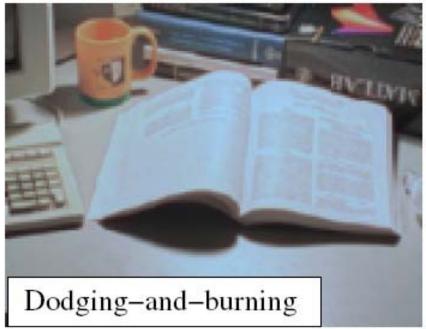
$$s_{\max}: |V_{s_{\max}}(x, y)| < \varepsilon$$

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





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

Frédo Durand & Julie Dorsey

SIGGRAPH 2002

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

A typical photo

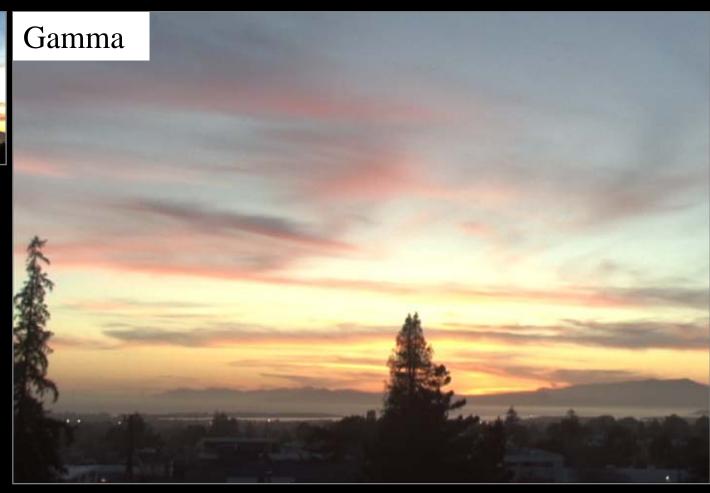
- Sun is overexposed
- Foreground is underexposed



Gamma compression

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

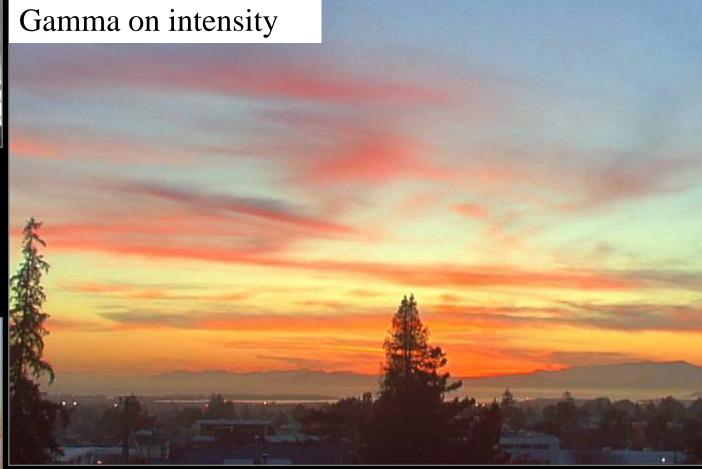




Gamma compression on intensity

 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



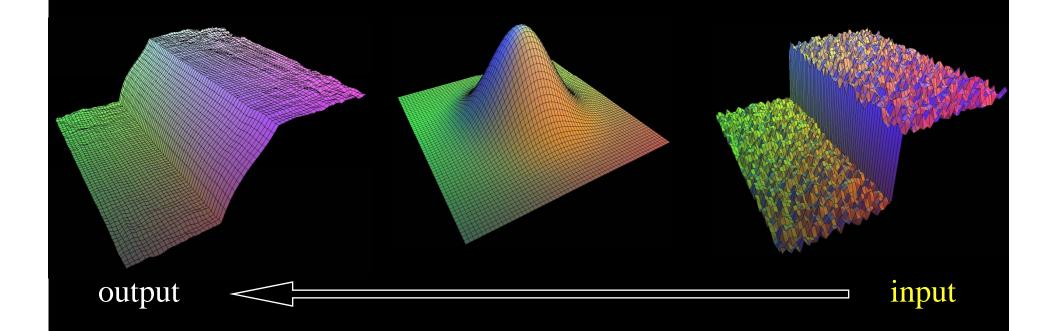




- Anisotropic diffusion [Perona & Malik 90]
 - Blurring as heat flow
 - LCIS [Tumblin & Turk]
- Bilateral filtering [Tomasi & Manduci, 98]

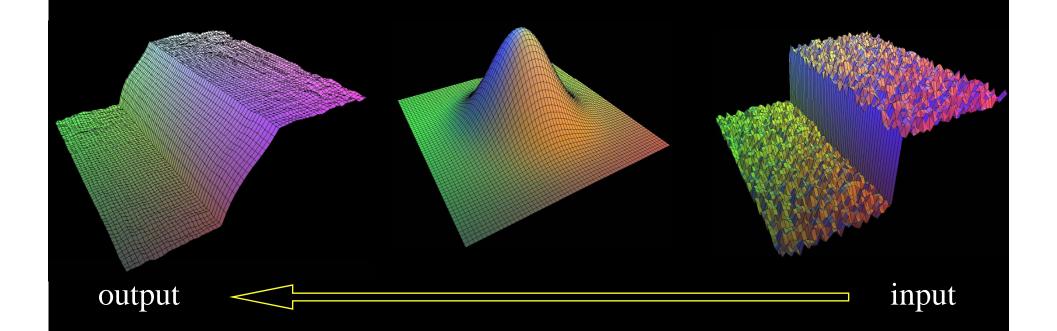
Start with Gaussian filtering

• Here, input is a step function + noise



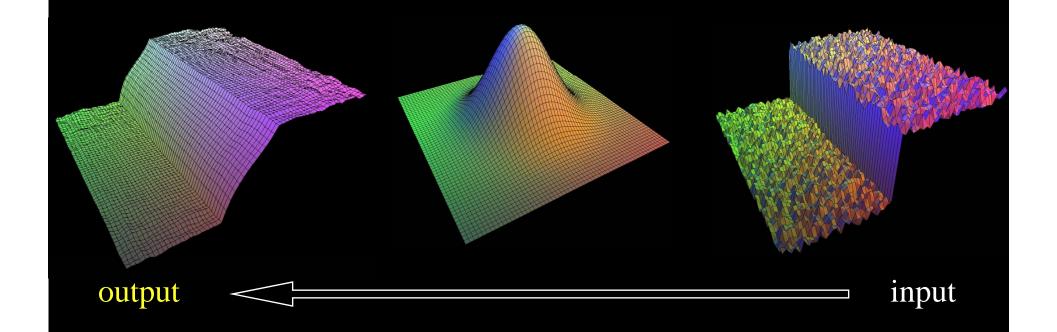
Start with Gaussian filtering

Spatial Gaussian f

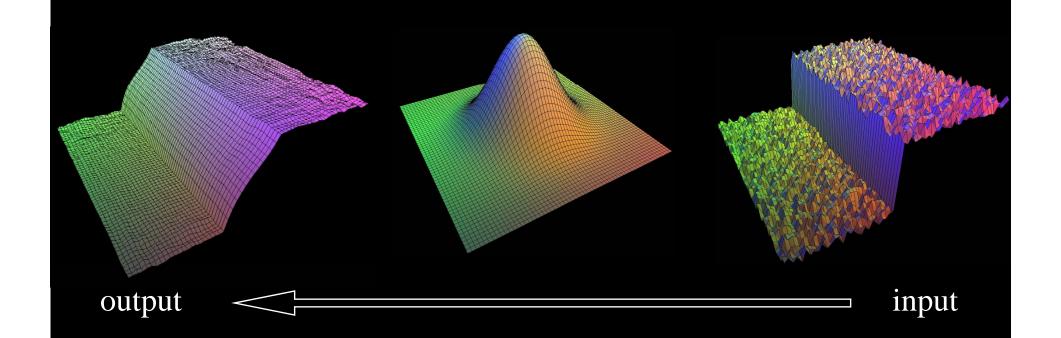


Start with Gaussian filtering

Output is blurred

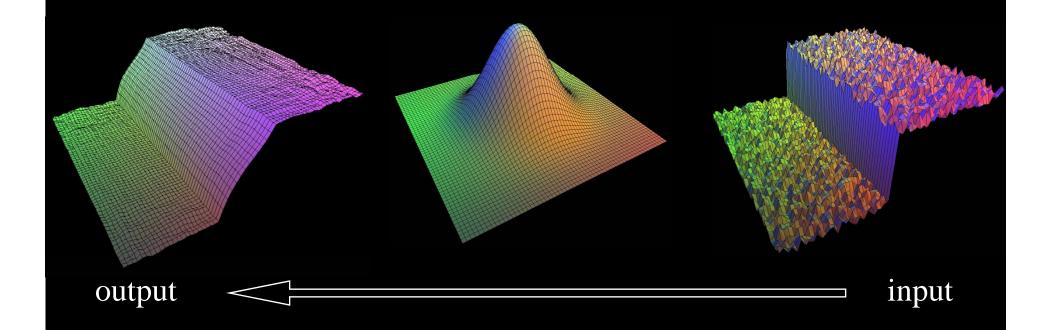


Gaussian filter as weighted average



The problem of edges

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



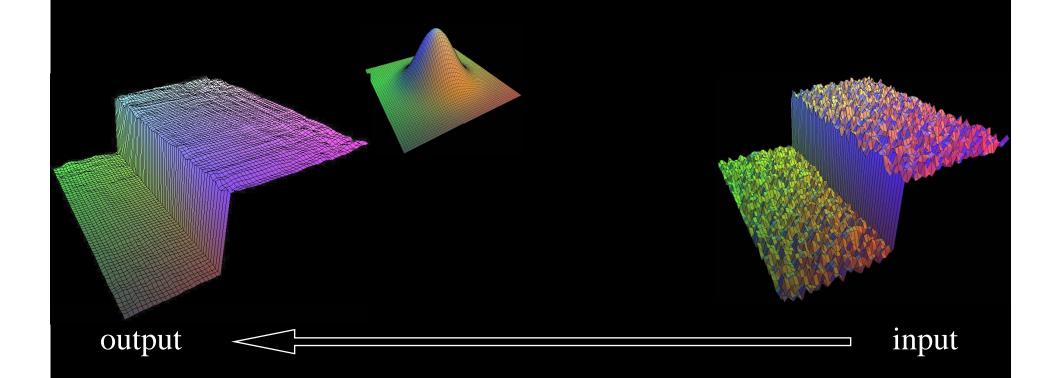
Principle of Bilateral filtering

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



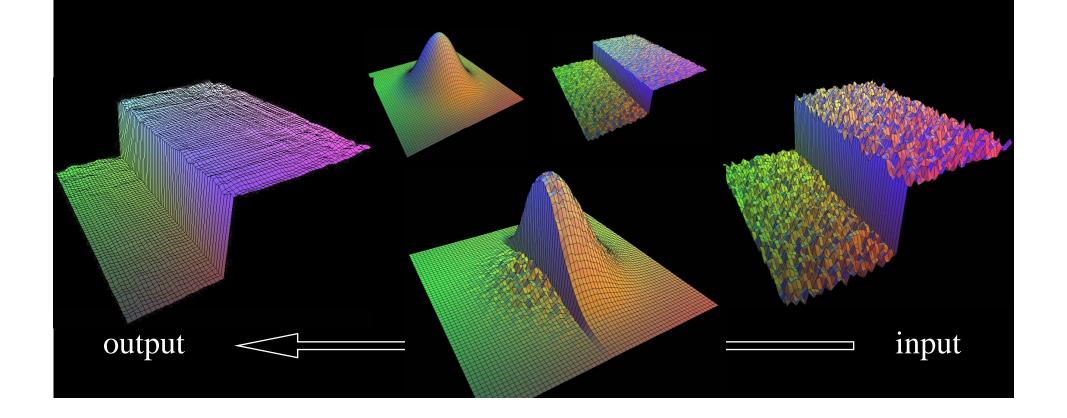
Bilateral filtering

- [Tomasi and Manduchi 1998]
- Spatial Gaussian f



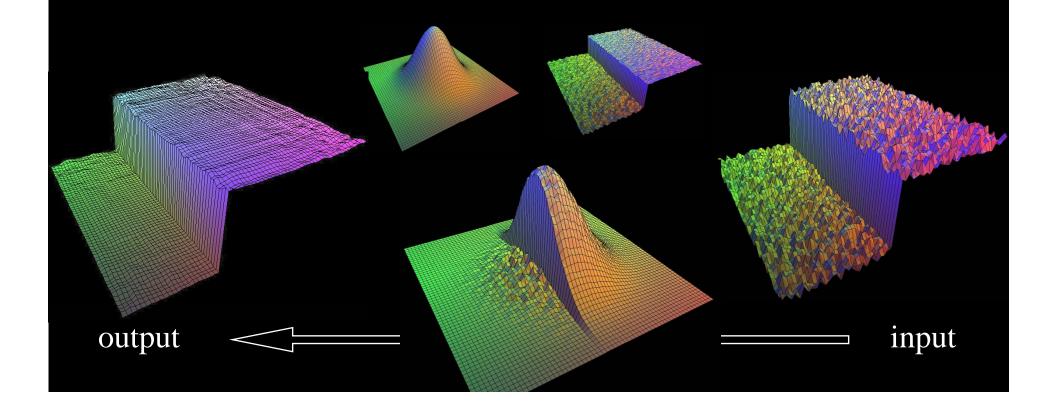
Bilateral filtering

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



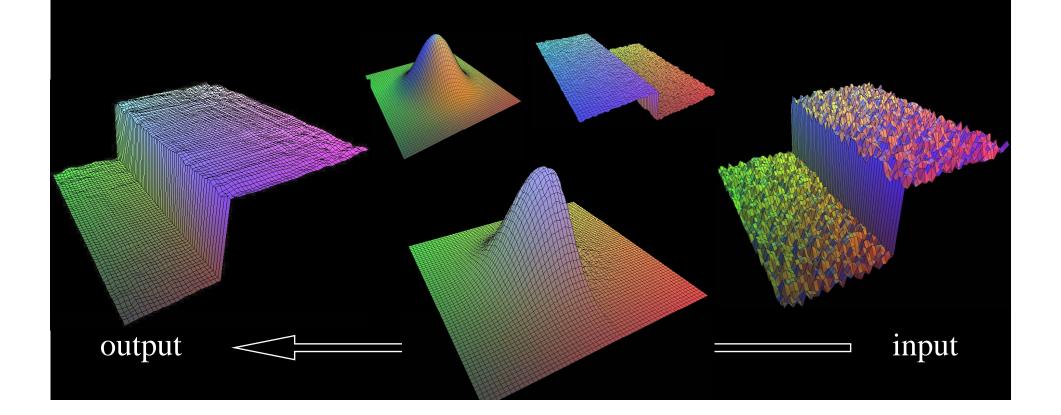
Normalization factor

- [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!

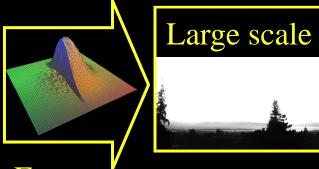












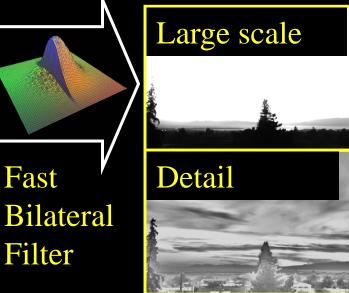
Fast
Bilateral
Filter



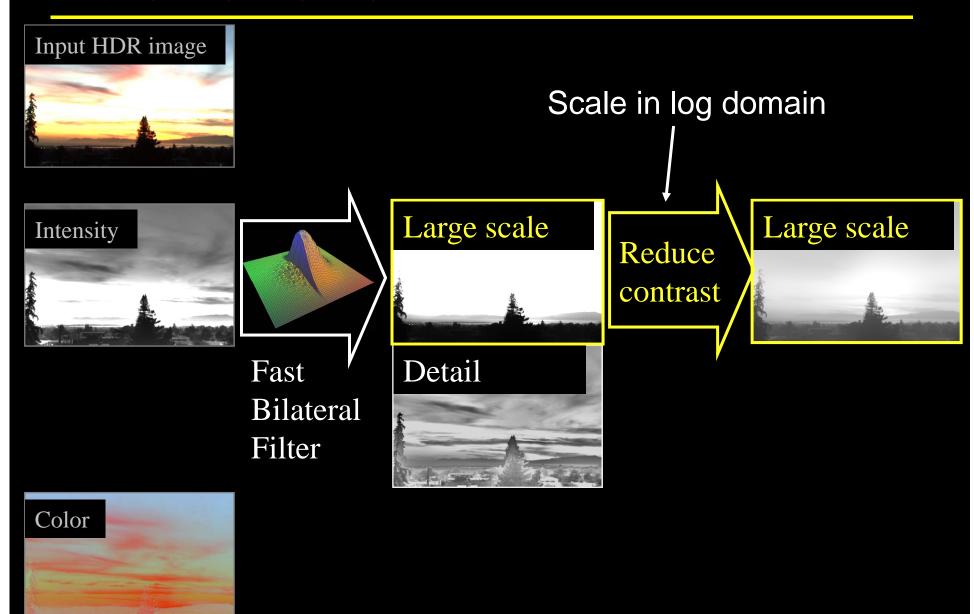






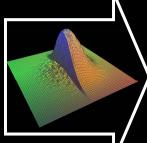












Fast
Bilateral
Filter





Reduce contrast

Preserve!

Large scale



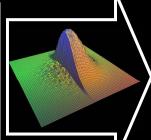






















Preserve!

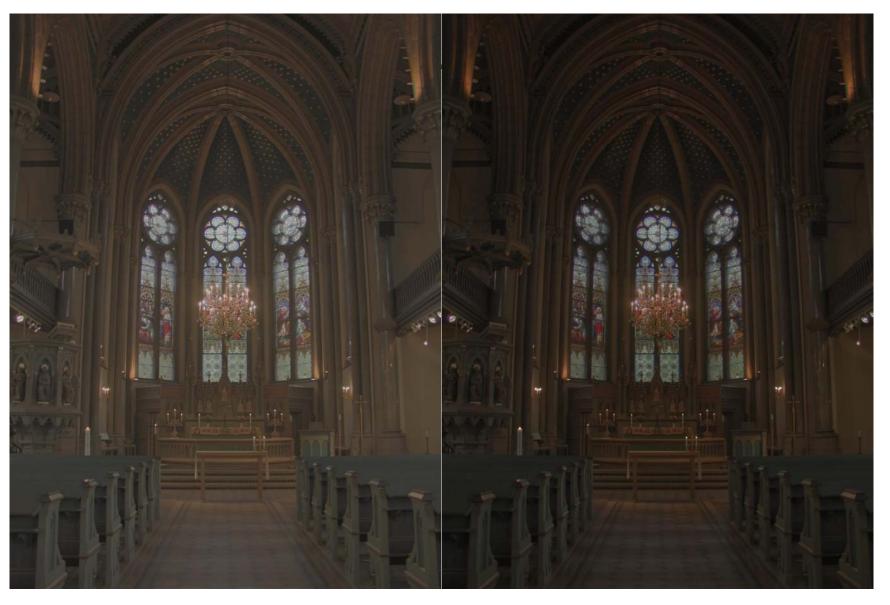






Bilateral filter is slow!

- 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

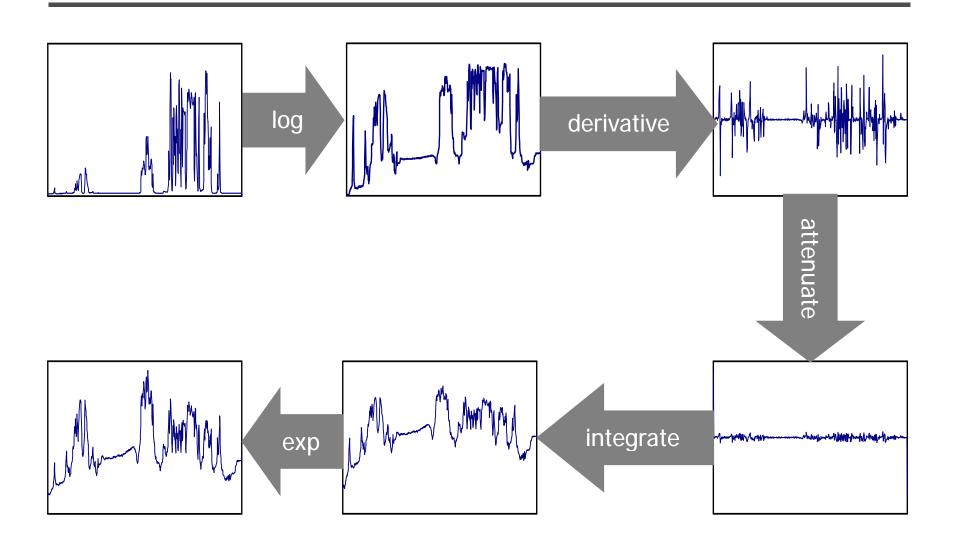
Raanan Fattal Dani Lischinski Michael Werman

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 is not 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

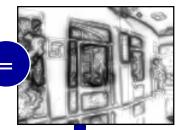
Multiscale gradient attenuation



interpolate







interpolate







Final gradient attenuation map

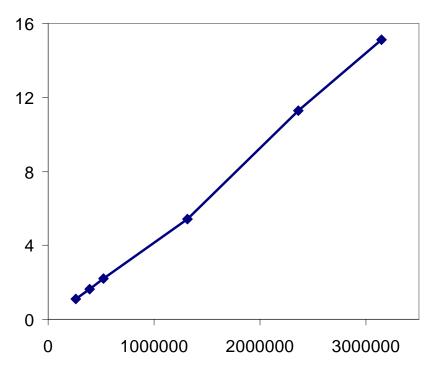


Performance

• Measured on 1.8 GHz Pentium 4:

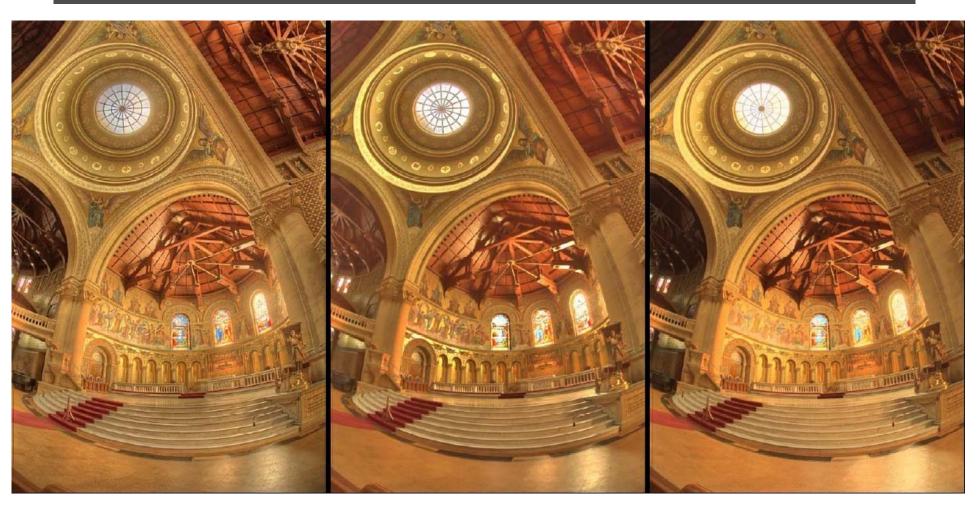
- 512 x 384: 1.1 sec

- 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

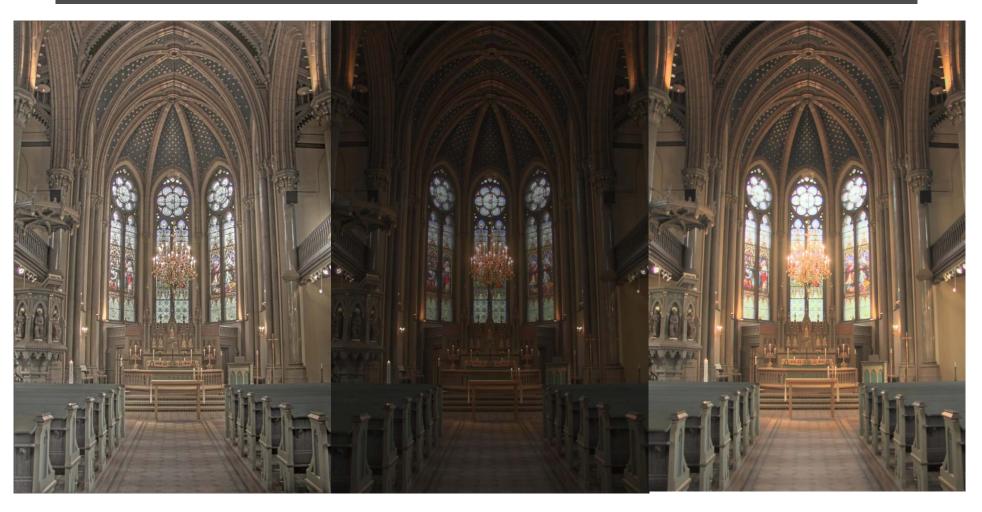


Gradient domain [Fattal et al.]

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

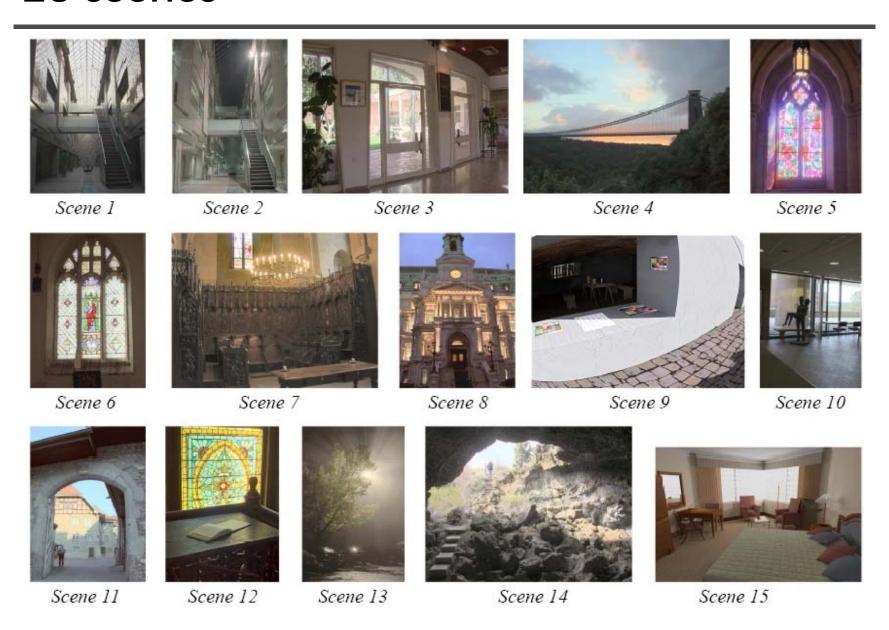
Patrick Ledda Alan Chalmers Tom Troscinko Helge Seetzen

SIGGRAPH 2005

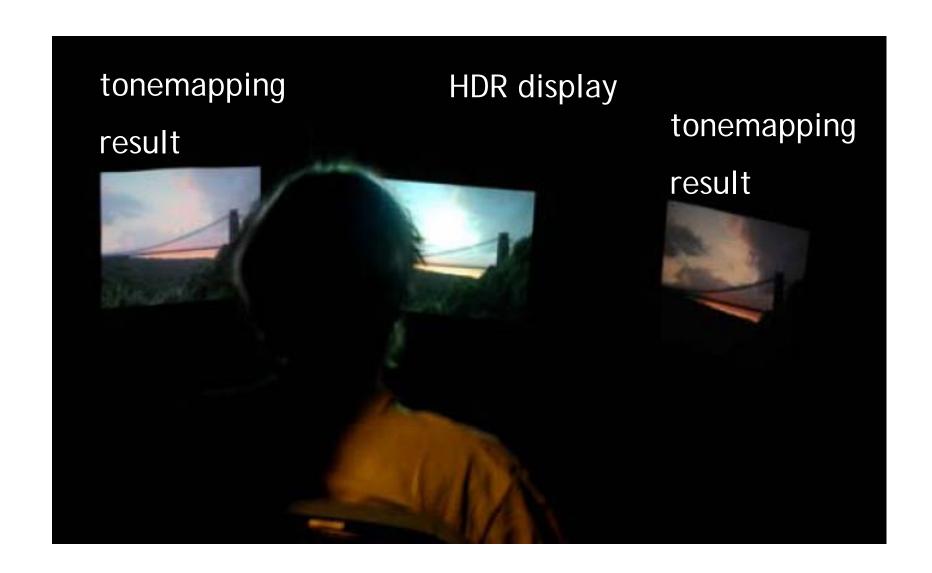
Six operators

- 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 ₃	tmo_4	tmo ₅	tmo ₆	Score
tmo_1	_	1	0	0	1	1	3
tmo_2	0	-	0	1	1	0	2
tmo ₃	1	1	-	1	1	1	5
tmo ₄	1	0	0	-	0	0	1
tmo ₅	0	0	0	1	-	1	2
tmo ₆	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.