#### Tone mapping

#### Digital Visual Effects, Spring 2007 Yung-Yu Chuang 2007/3/13

with slides by Fredo Durand, and Alexei Efros

#### Preliminaries

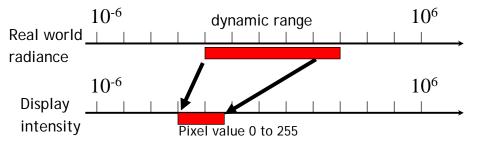
• For color images

$$\begin{bmatrix} R_d \\ G_d \\ B_d \end{bmatrix} = \begin{bmatrix} L_d \frac{K_w}{L_w} \\ L_d \frac{G_w}{L_w} \\ L_d \frac{B_w}{L_w} \end{bmatrix}$$

- Log domain is usually preferred.
- Gaussian filter. Sampling issues. Efficiency issues.

#### Tone mapping

• How can we display it? Linear scaling?, thresholding?



CRT has 300:1 dynamic range

#### Eye is not a photometer!

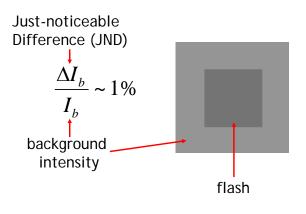


• "Every light is a shade, compared to the higher lights, till you come to the sun; and every shade is a light, compared to the deeper shades, till you come to the night."

— John Ruskin, 1879

#### We are more sensitive to contrast

#### • Weber's law



#### low key (0.18)

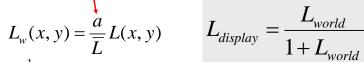
high key (0.5)

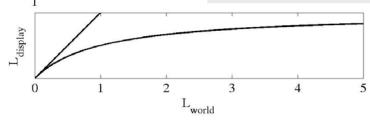
#### Global operator (Reinhart et al)

$$\overline{L} = \exp\left(\frac{1}{N}\sum_{x,y}\log(\delta + L(x, y))\right)$$

Approximation of scene's key (how light or dark it is). Map to 18% of display range for average-key scene

User-specified; high key or low key





## Frequency domain First proposed by Oppenheim in 1968!

• Under simplified assumptions,

image

= illuminance \* low-frequency attenuate more

reflectance
 high-frequency
 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 Laboratory for Computer Science Massachusetts Institute of Technology

#### A typical photo

- Sun is overexposed
- Foreground is underexposed



#### Gamma compression

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



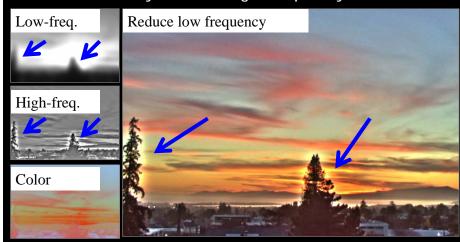
#### Gamma compression on intensity

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



#### The halo nightmare

- For strong edges
- Because they contain high frequency



#### Chiu et al. 1993

- Reduce contrast of low-frequencies
- Keep high frequencies



#### **Durand and Dorsey**

- Do not blur across edges
- Non-linear filtering



#### Edge-preserving filtering

• Blur, but not across edges

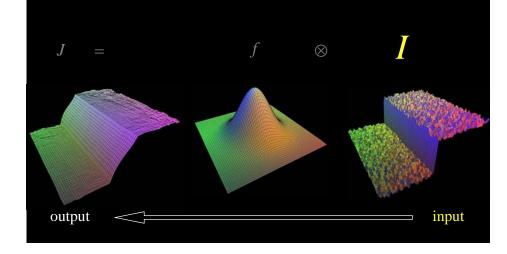
# Input



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

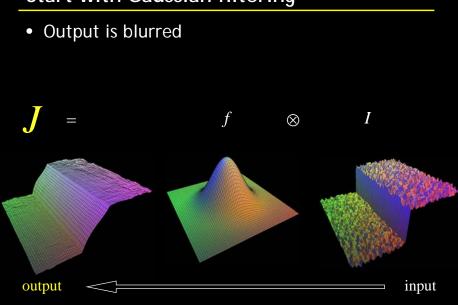
#### Start with Gaussian filtering

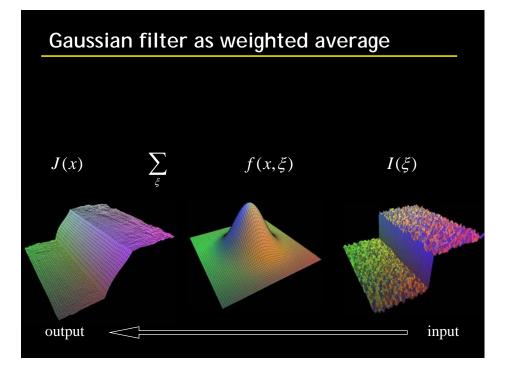
• Here, input is a step function + noise



## Start with Gaussian filtering • Spatial Gaussian f $\otimes$ Ι input output

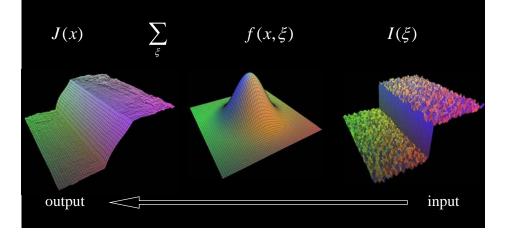
#### Start with Gaussian filtering





#### The problem of edges

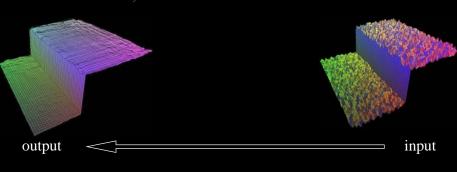
- Here,  $I(\xi)$  "pollutes" our estimate J(x)
- It is too different



#### Principle of Bilateral filtering

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

$$J(x) = \frac{1}{k(x)} \sum_{\xi} f(x,\xi) \qquad g(I(\xi) - I(x)) \qquad I(\xi)$$



#### **Bilateral filtering**

- [Tomasi and Manduchi 1998]
- Spatial Gaussian f

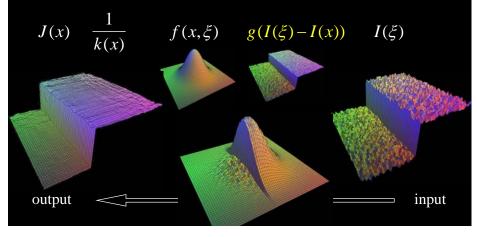
output

$$J(x) \quad \frac{1}{k(x)} \qquad f(x,\xi) \quad g(I(\xi) - I(x)) \quad I(\xi)$$

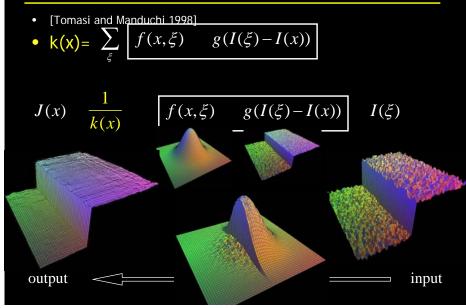
input

#### Bilateral filtering

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

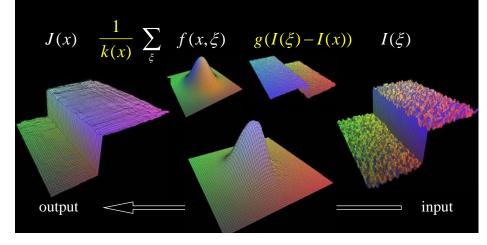


#### Normalization factor

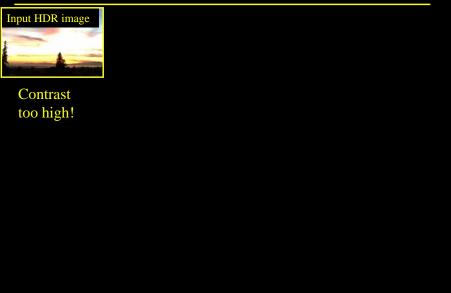


#### Bilateral filtering is non-linear

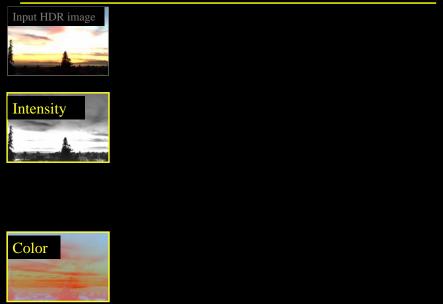
- [Tomasi and Manduchi 1998]
- The weights are different for each output pixel

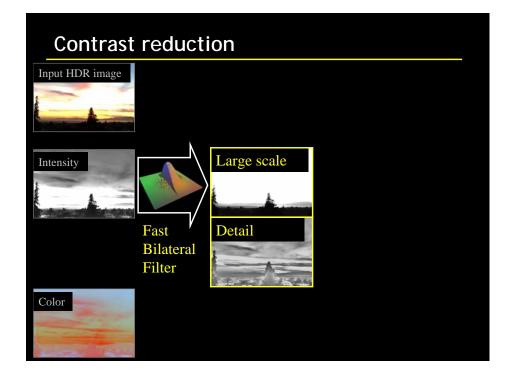


#### **Contrast reduction**

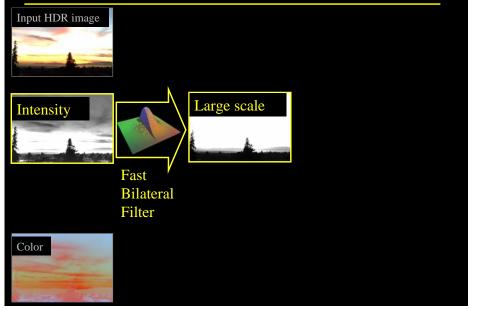


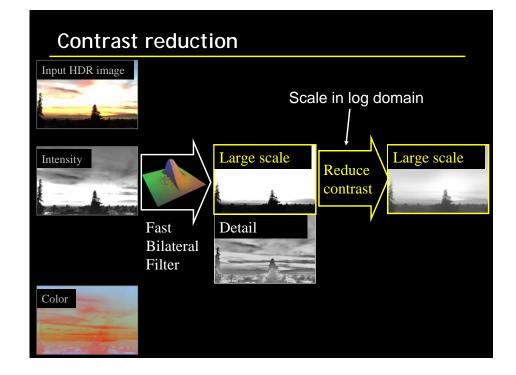
#### **Contrast reduction**

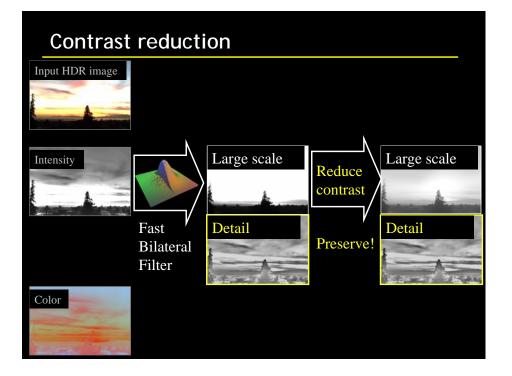




#### Contrast reduction









Oppenheim

bilateral

