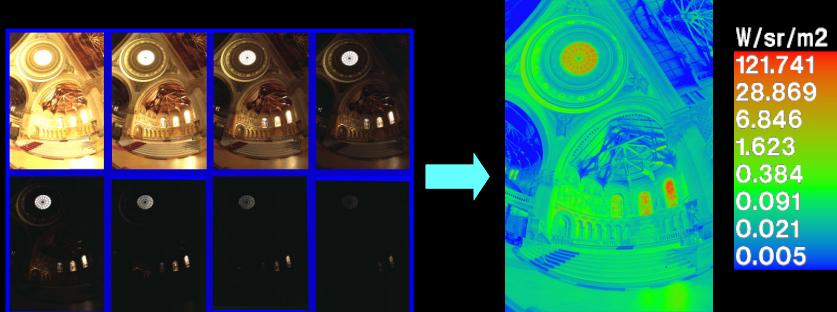


High Dynamic Range Imaging



70001 – Advanced Computer Graphics: Photographic Image Synthesis

Abhijeet Ghosh

Lecture 03, Jan. 19th 2024

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HDR Imaging

- Exposure bracketing
 - 1 stop = 2X, 2 stops = 4X, 3 stops = 8X.



- Aperture settings
 - f/2.8, f/4, f/5.6, f/8, f/11, f/16, f/22



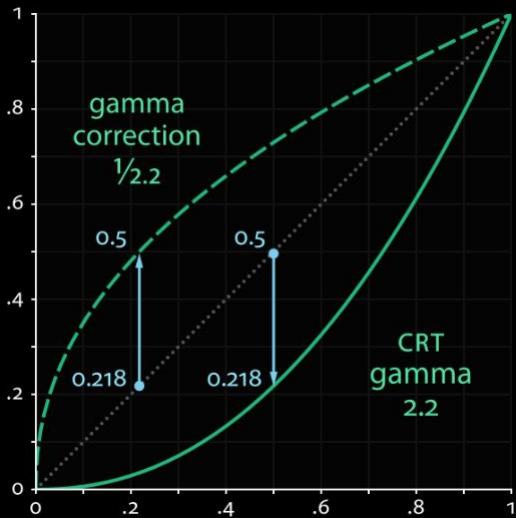
- ISO Gain
 - ISO 100 = 1X, 200 = 2X, 400 = 4X.



- Neutral Density Filter
 - 0.3 to 4.0 (log base 10 scale)

2

Gamma correction



3

Gamma 2.2 graph

Implications:

128 is less than $\frac{1}{4}$ as bright as 255

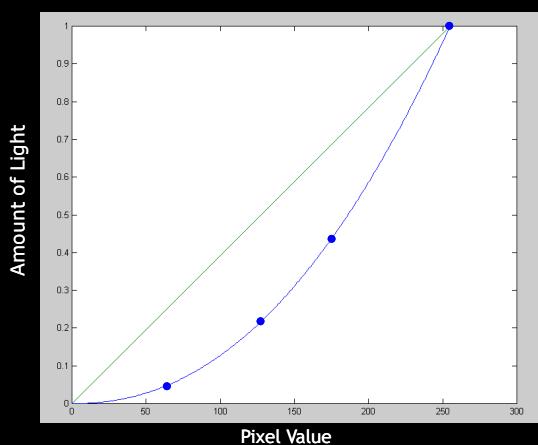
128 is more than 4 times as bright as 64

175 is twice as bright as 128

93 is half as bright as 128

$$128 + 128 = 175$$

$$128 / 2 = 93$$



See also: <http://www.poynton.com/GammaFAQ.html>

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2

Gamma correction example



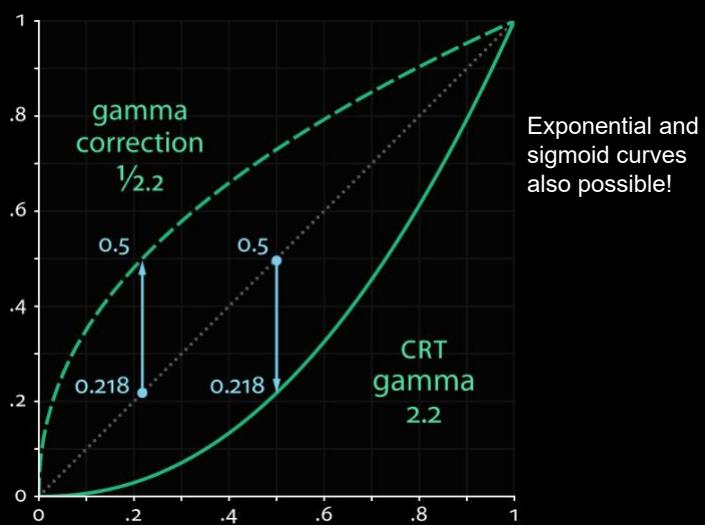
Gamma 1.0
"too saturated with dark values suppressed"



Gamma correction 2.2
"linearized display of values producing natural appearance"

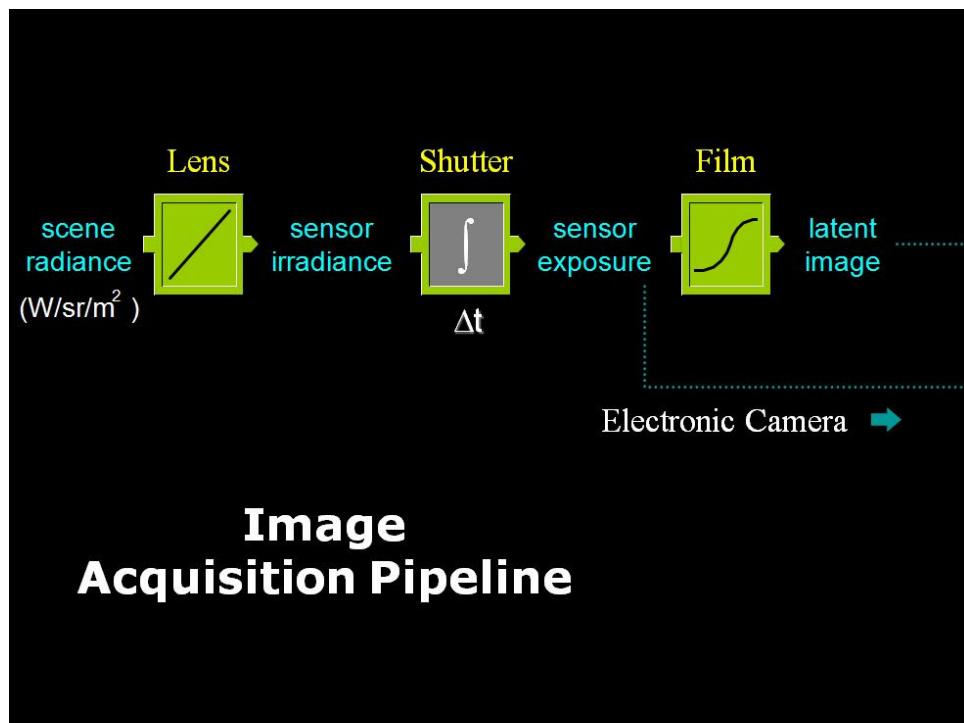
5

Gamma correction

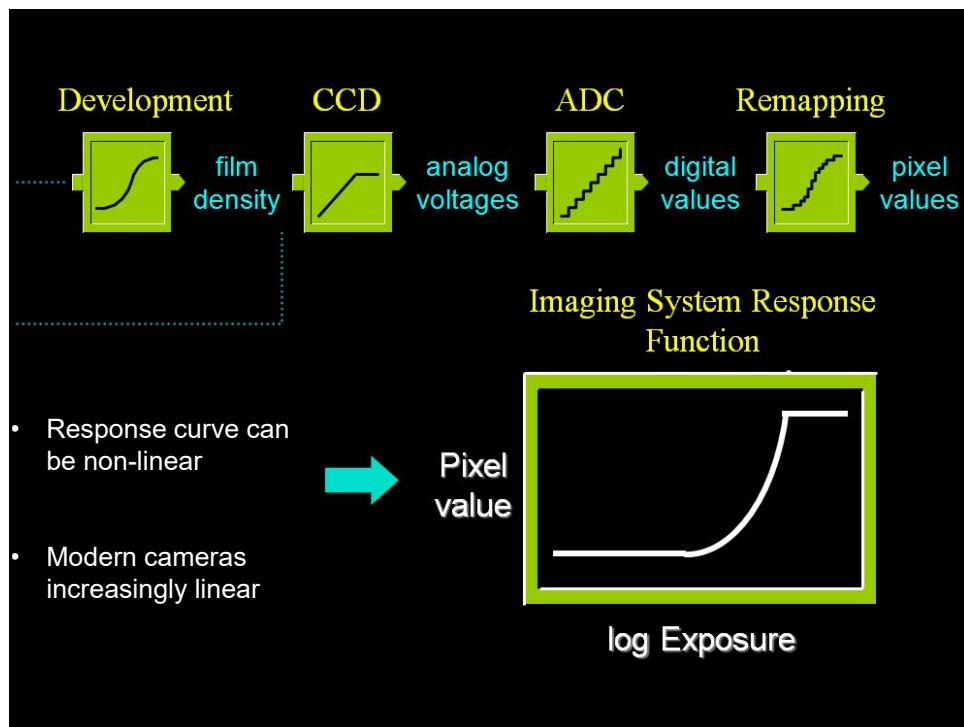


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3



7



8

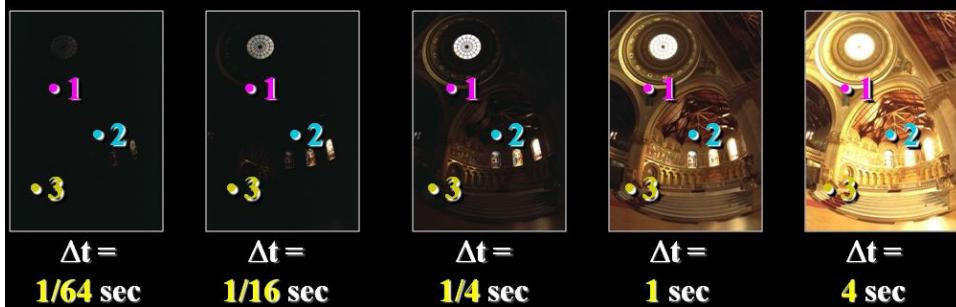
Response Curve Recovery

Mann and Picard SPIE 95:

Track one pixel value across series and fit a gamma-like curve

Debevec and Malik SIGGRAPH 97:

Derive detailed curve from many pixels



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

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

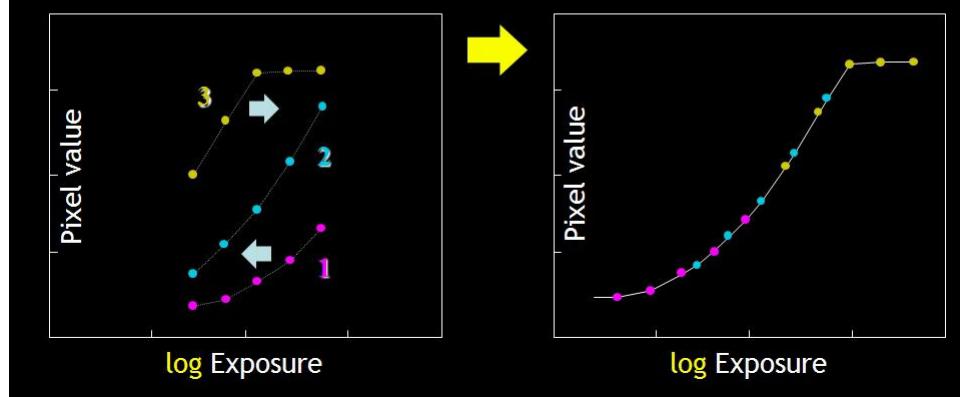
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Recovering the Response Curve

HDRShop www.hdrshop.com

Assuming unit radiance
for each pixel

After adjusting radiances to
obtain a smooth curve



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Image formation model

$$Z_{ij} = f(E_i \Delta t_j)$$

Camera pixels Camera Irradiance \times exposure
 response
(possibly nonlinear)

$$f^{-1}(Z_{ij}) = E_i \Delta t_j$$

$$\ln f^{-1}(Z_{ij}) = \ln E_i + \ln \Delta t_j$$

[Debevec & Malik 97]

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Image formation model

$$Z_{ij} = f(E_i \Delta t_j)$$

Camera pixels Camera Irradiance \times exposure
 response
(possibly nonlinear)

More details:

<http://www.pauldebevec.com/Research/HDR/debevec-siggraph97.pdf>

[Debevec & Malik 97]

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Linear response simplification

$$Z_{ij} = f(E_i \Delta t_j)$$

Camera pixels

Irradiance \times exposure

$$Z_{ij} = E_i \Delta t_j$$

Unknown

$$E_i = Z_{ij} / \Delta t_j$$

Note that each exposure provides $j = 1$ to N different estimates of E_i ,

Final step is to merge the multiple E_i estimates into one HDR value using weighted averaging.

Typically higher weights for values in the middle of [0 - 1] range than values at the ends of the range, i.e., hat function or a curve.

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HDR Imaging application – motion blur



Scene



LDR simulation



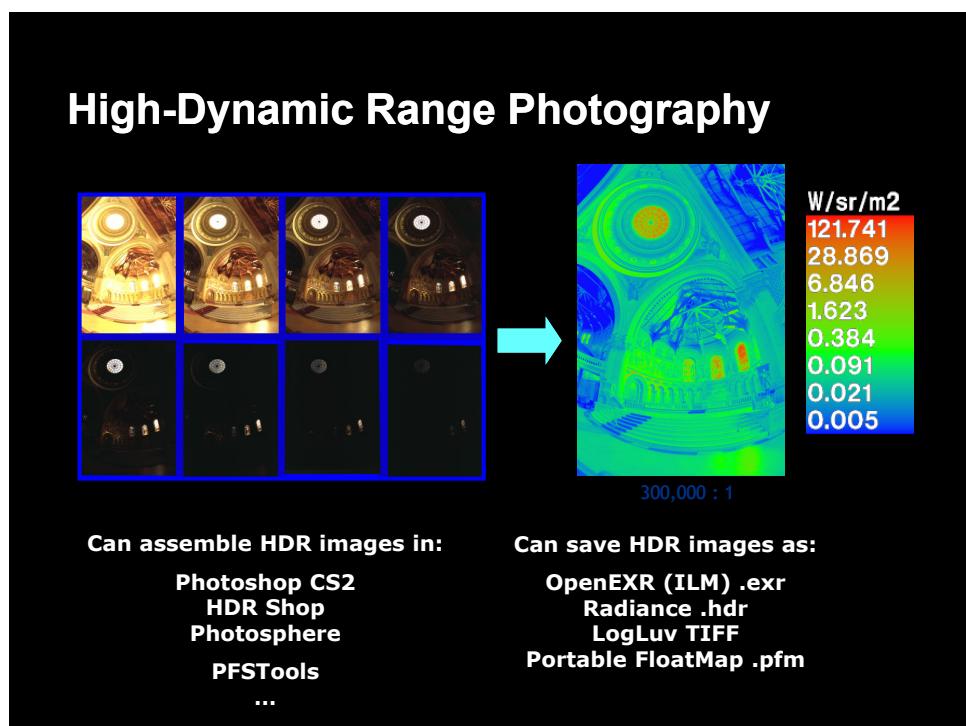
HDR simulation



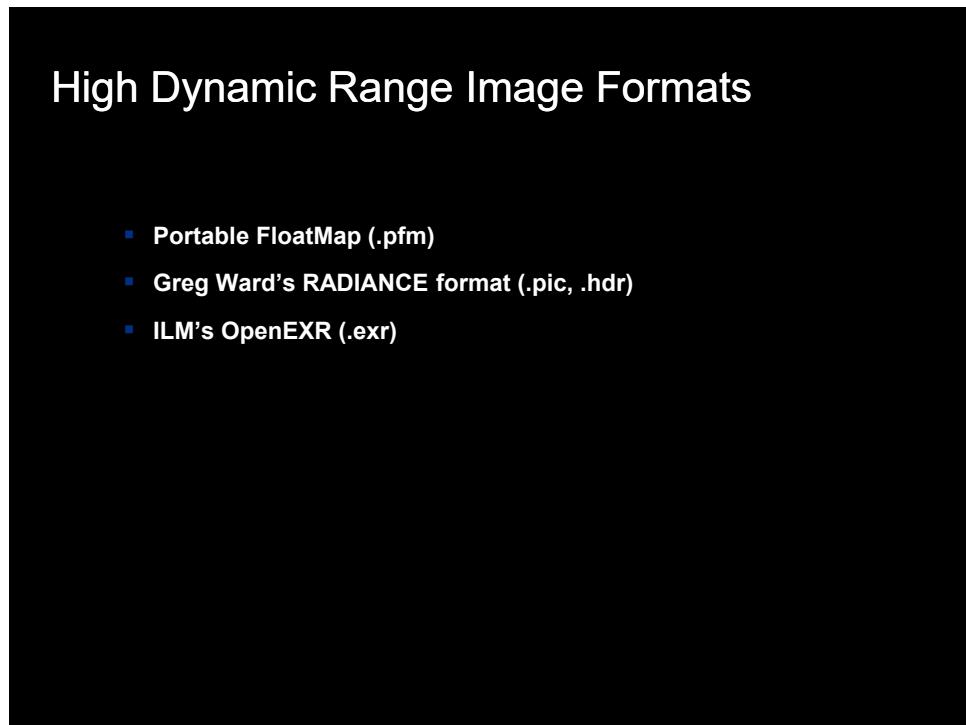
Real photograph

[Debevec & Malik 97]

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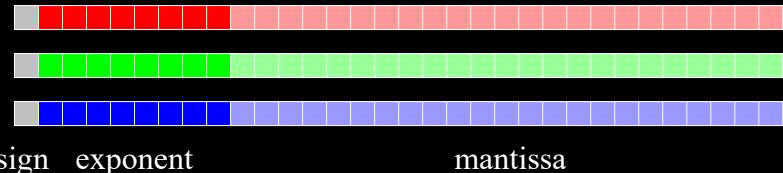
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Portable FloatMap

- 12 bytes per pixel, 4 for each channel



Text header similar to Jeff Poskanzer's .ppm image format:

Floating Point TIFF similar

PF
768 512
3
<binary image data>

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RADIANCE Format – Greg Ward



$(145, 215, 87, 149) =$ $(145, 215, 87) * 2^{(149-128)} =$ $(1190000, 1760000, 713000)$	$(145, 215, 87, 103) =$ $(145, 215, 87) * 2^{(103-128)} =$ $(0.00000432, 0.00000641, 0.00000259)$
---	---

Ward, Greg. "Real Pixels," in Graphics Gems IV, edited by James Arvo, Academic Press, 1994

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ILM OpenEXR Format

- **Purpose:** HDR lighting and compositing
- 16-bit/primary floating point
(sign-e5-m10)
- 9.6 orders of magnitude in 0.1% steps
- Wavelet compression of about 40%
- Negative colors and full gamut RGB
- Open Source I/O library released Fall 2002

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ILM's OpenEXR (.exr)

- 6 bytes per pixel, 2 for each channel, compressed



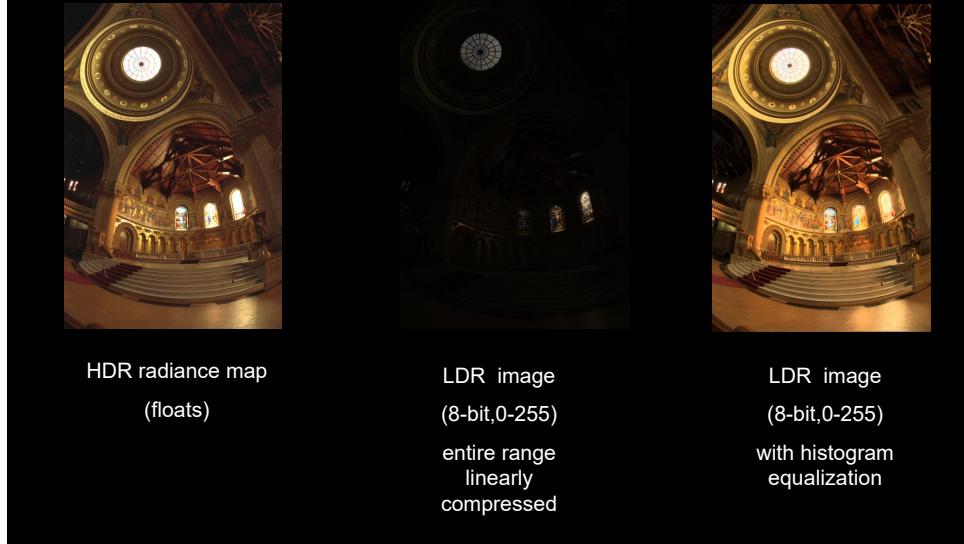
sign exponent mantissa

- Several lossless compression options, 2:1 typical
- Compatible with the “half” datatype in NVidia’s Cg
- Supported natively on GeForce FX and Quadro FX

- Available at <http://www.openexr.net/>

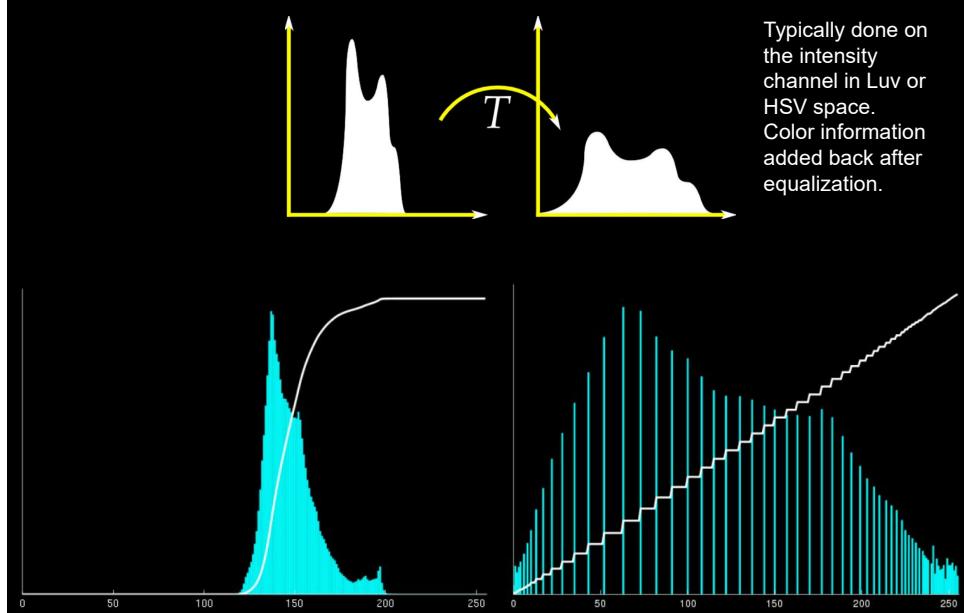
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Tone mapping – automatic remapping



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Histogram equalization



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Global tone mapping – user driven



HDR radiance map
(floats)



LDR image
(8-bit,0-255)
Exposure bracketing
(1 stop = x2 scaling)



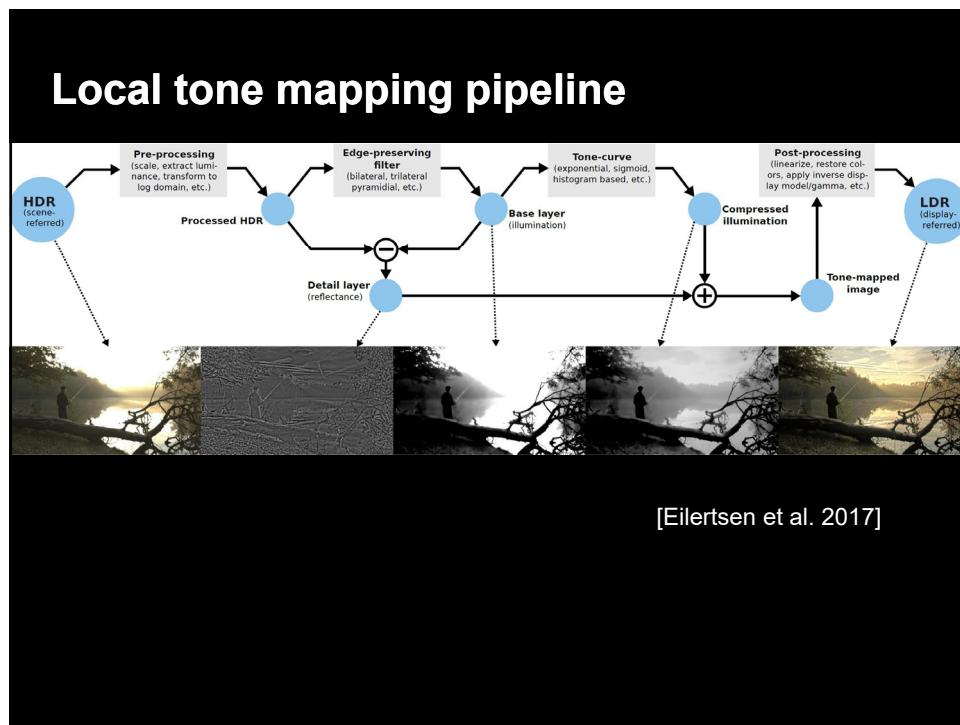
LDR image
(8-bit,0-255)
Exposure
bracketing +
gamma control

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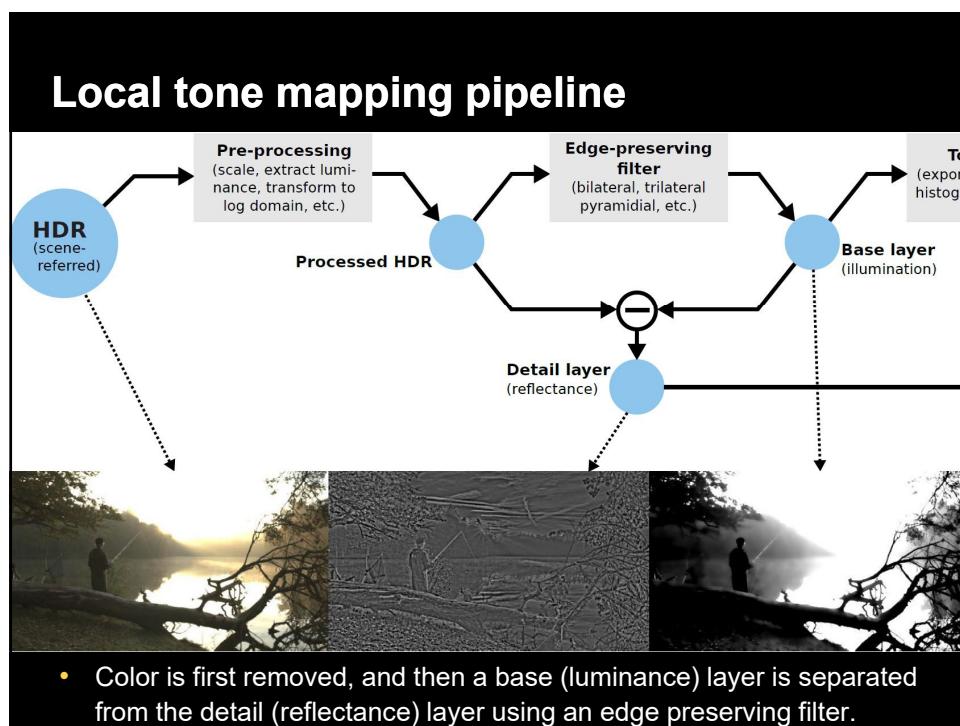
Global vs local tone mapping

- Global mapping – compress global dynamic range (scaling, gamma, sigmoid, histogram equilization, etc.)
 - Reduced local contrast, bad for Human visual perception ☺
- Local mapping – Edge or detail (local contrast) preservation
 - Mimic human visual system's perception
 - Secondary step along with some global mapping

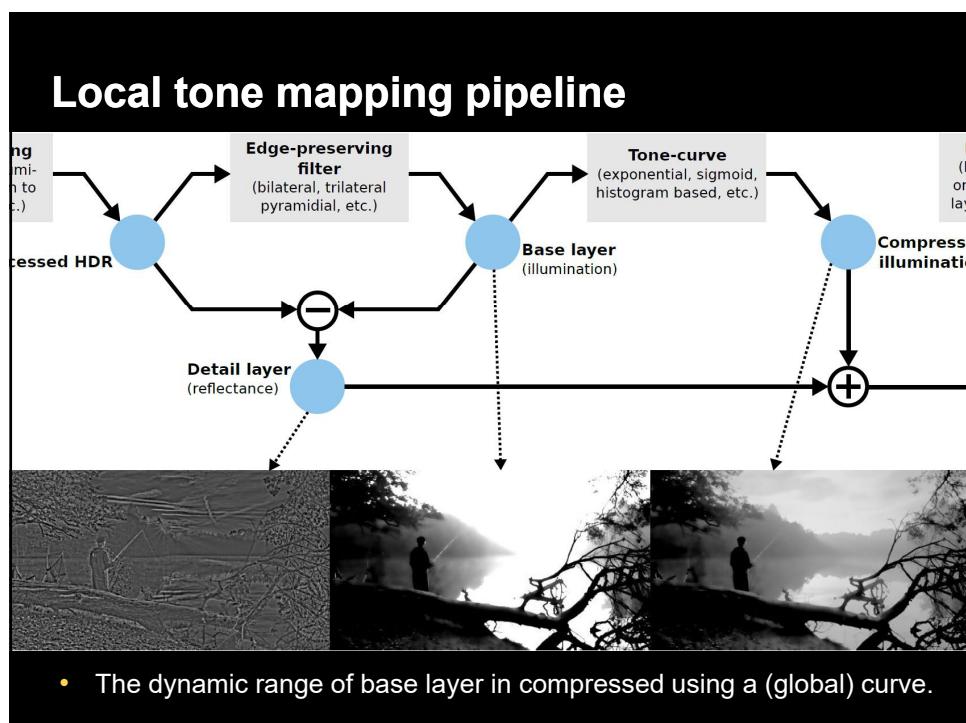
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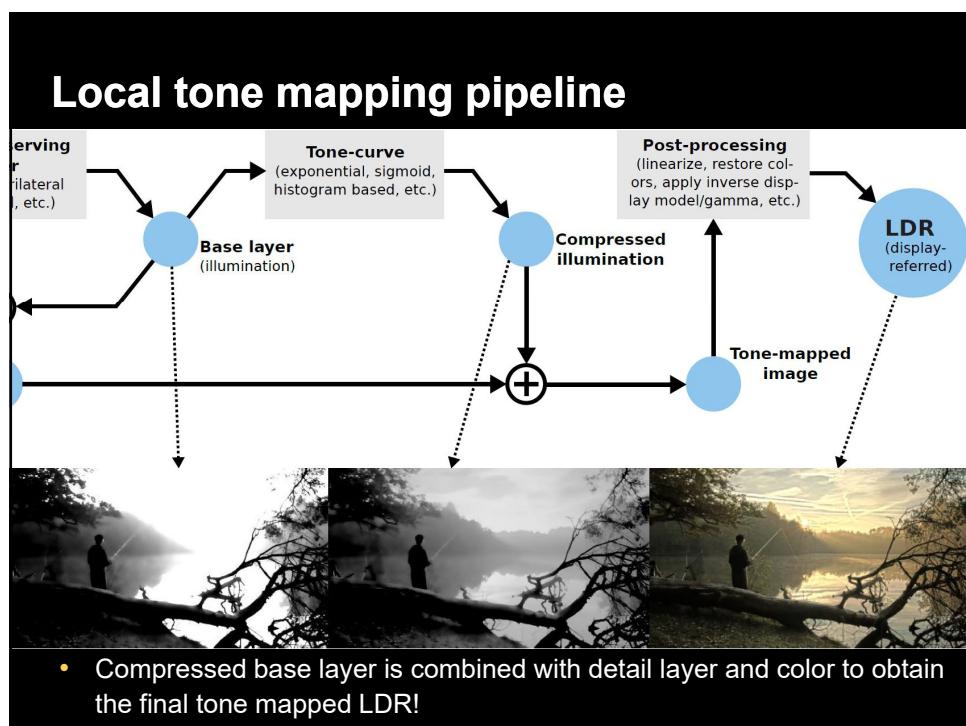
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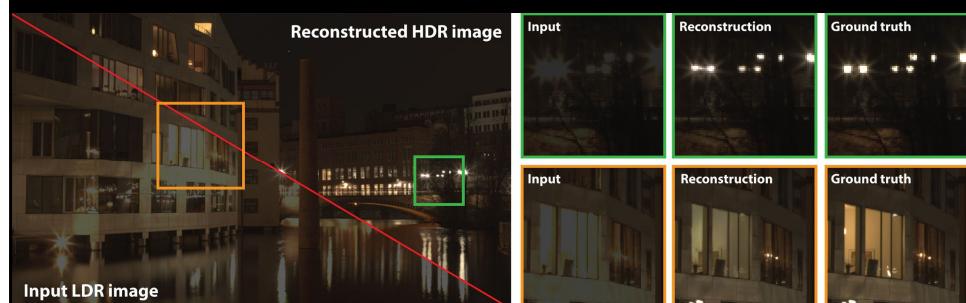
Local tone mapping

- Tricky to get the local contrast preservation step correct
 - Halos and other artifacts such as image noise can arise if not done right!
- Various edge preserving filters considered
 - Bilateral
 - Filtering in the image gradient domain instead of image intensity

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Deep HDR from LDR

[Eilertsen et al. 2017]

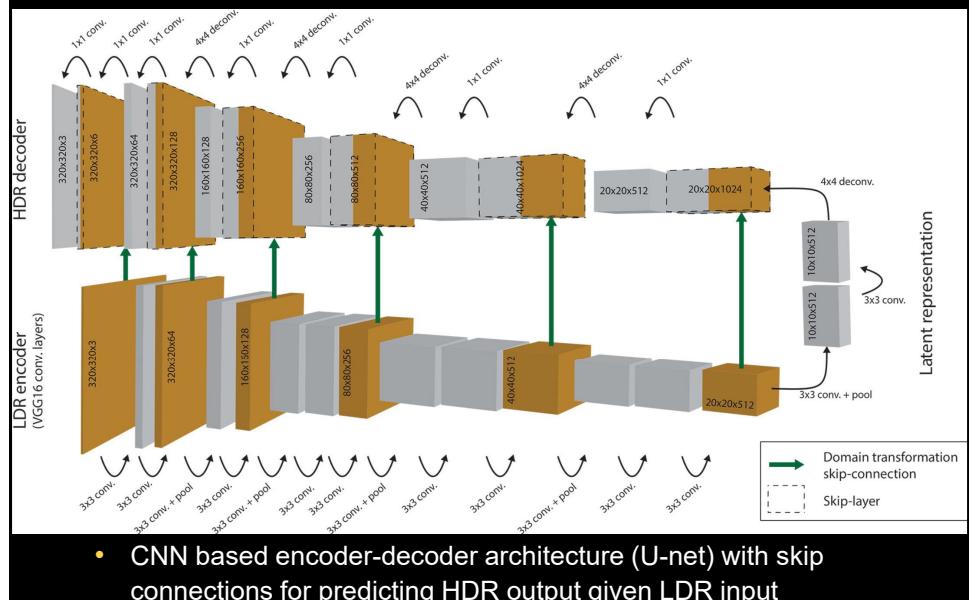


- CNN based encoder-decoder architecture for predicting HDR output given LDR input
- Deep learning for inverse tone-mapping!

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Deep HDR from LDR

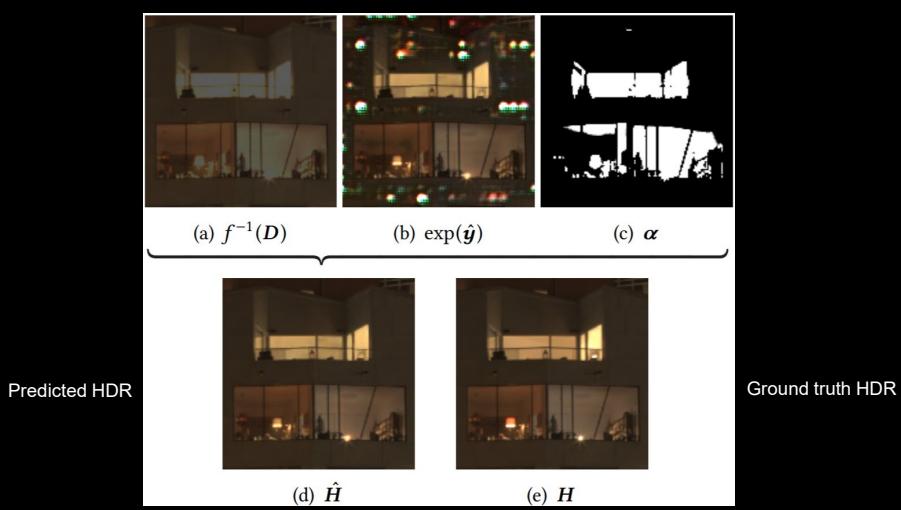
[Eilertsen et al. 2017]



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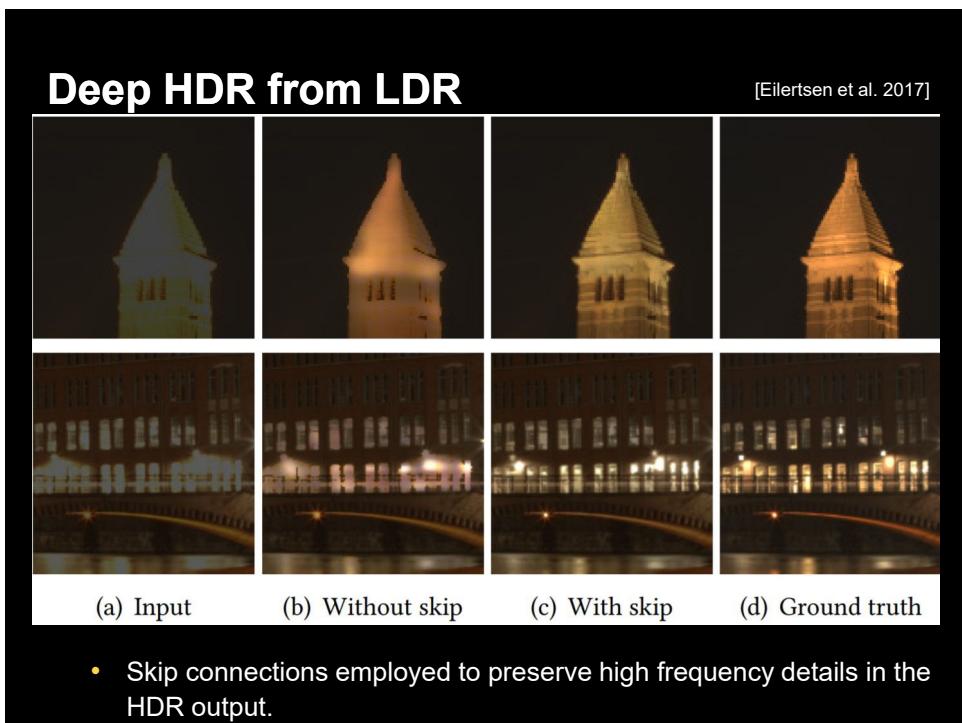
Deep HDR from LDR

[Eilertsen et al. 2017]



- CNN prediction only used for bright saturated regions of LDR images. Mask used to ignore the prediction for dark regions.

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