

# *Two-scale Tone Management*

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- › *Two-scale Tone Management for Photographic Look*
  - › Bae, Paris, Durand
  - › SIGGRAPH 2006

## **Two-scale Tone Management for Photographic Look**

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(a) input



(b) sample possible renditions: bright and sharp, gray and highly detailed, and contrasted, smooth and grainy

# Ansel Adams

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# Recall Color Transfer (or Histogram Matching)

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- › Can learn a similar *global* look from the model photograph
- › But richness in texture is not learned
  - › Texture is *local*
- › How to improve it?

# Tone Management

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- › Photographic look
  - › To convey a mood or an aesthetic
  - › For both high- and normal-dynamic-range inputs
- › Transfer the look of a model photograph to the picture being edited
  - › Tonal balance
  - › Amount of detail
  - › Sepia toning, grainy photograph
- › To give hundreds of pictures a consistent look

# Two-scale Nonlinear Decomposition

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- › Large-scale tonal balance management
- › Spatial detail variation
- › Gradient constraint

# Using Techniques We Have Learned So Far

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- › Histogram matching
- › Bilateral filtering
  - › Joint (or cross) bilateral filtering
  - › and a technique we will learn next week: gradient domain, Poisson reconstruction



model



input

*bilateral  
filter*



base



detail

*high pass and local averaging*

*large-scale  
transfer*

*textureness  
transfer*



textureness

mo

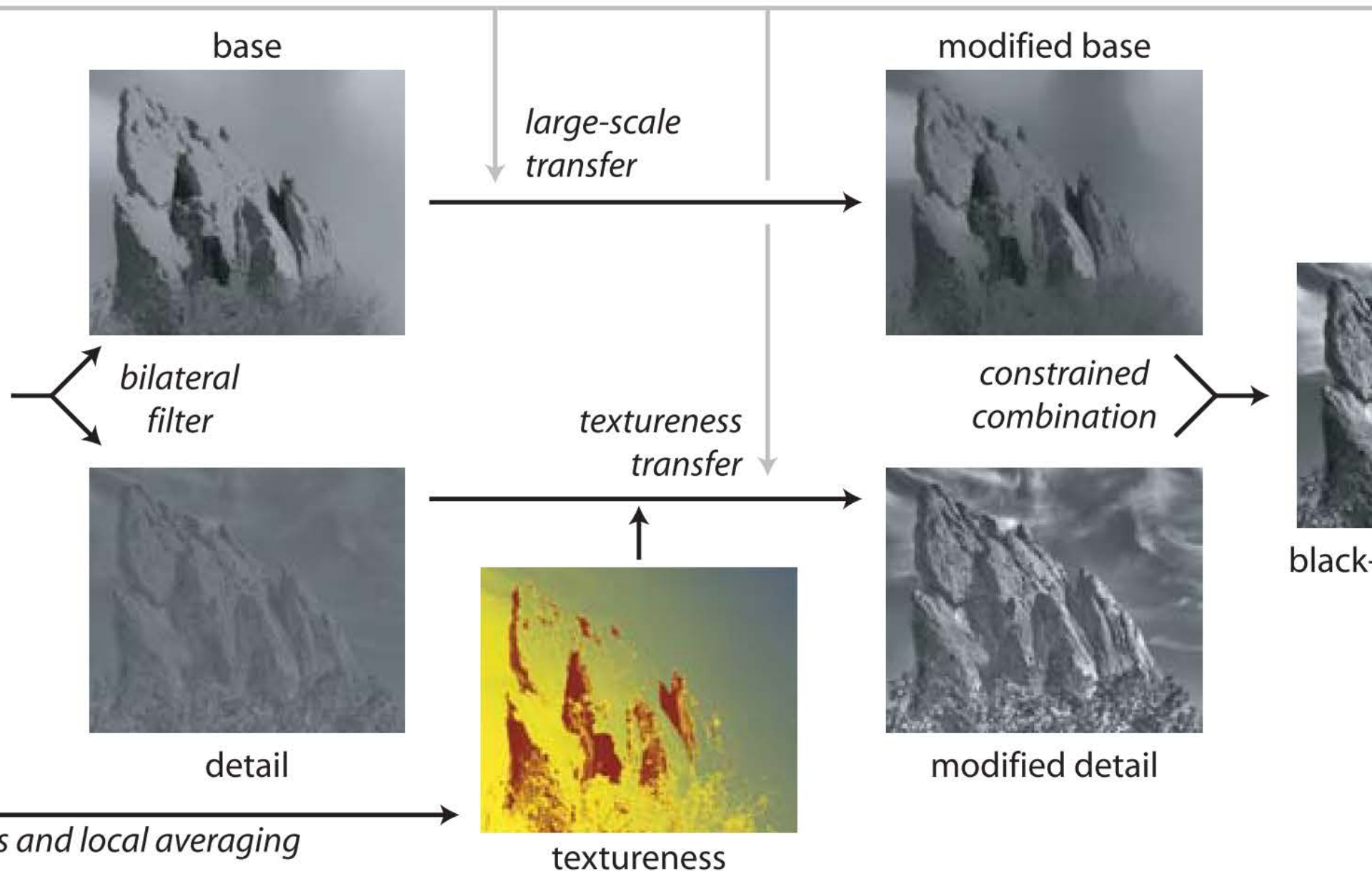


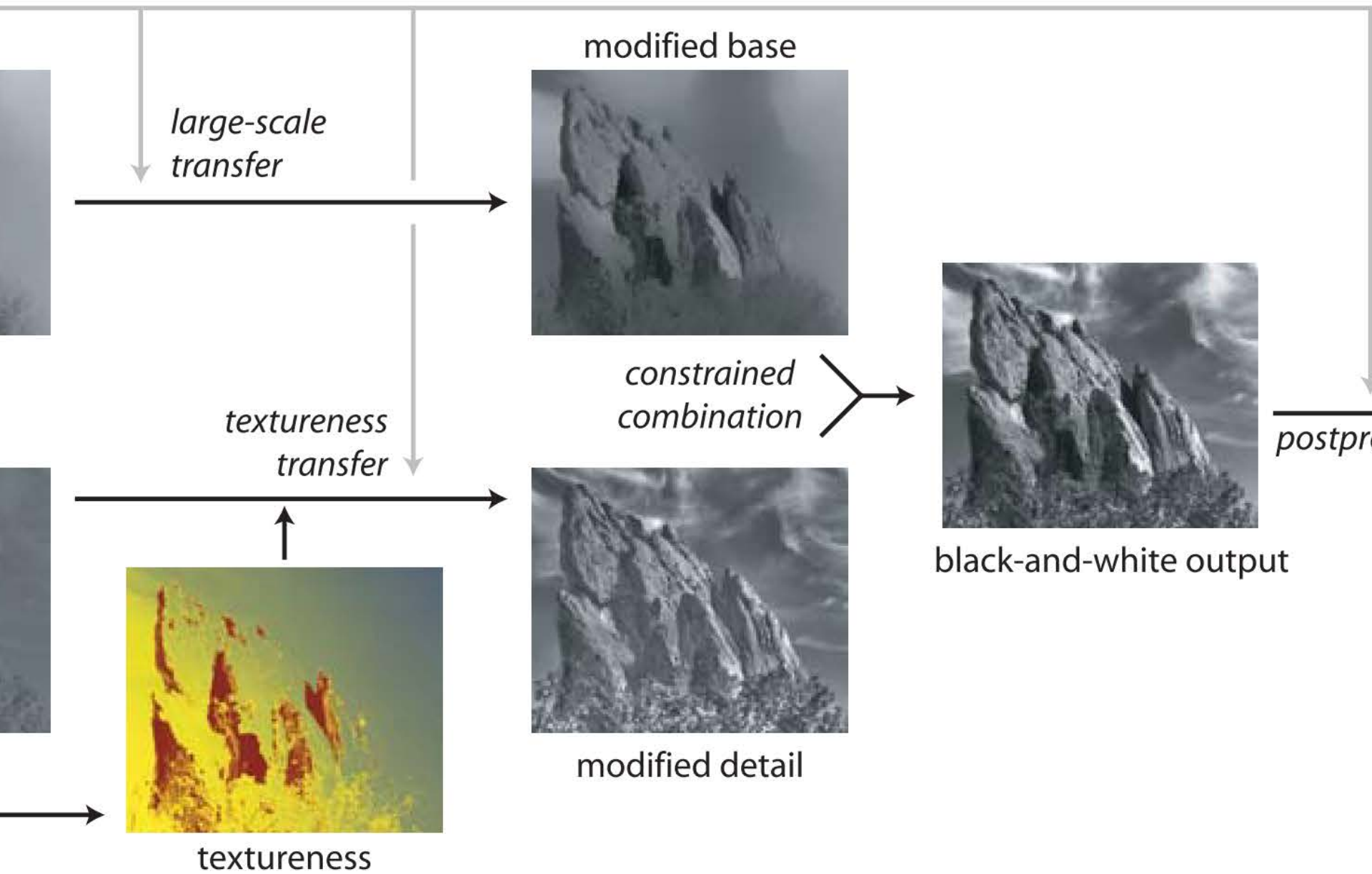
e

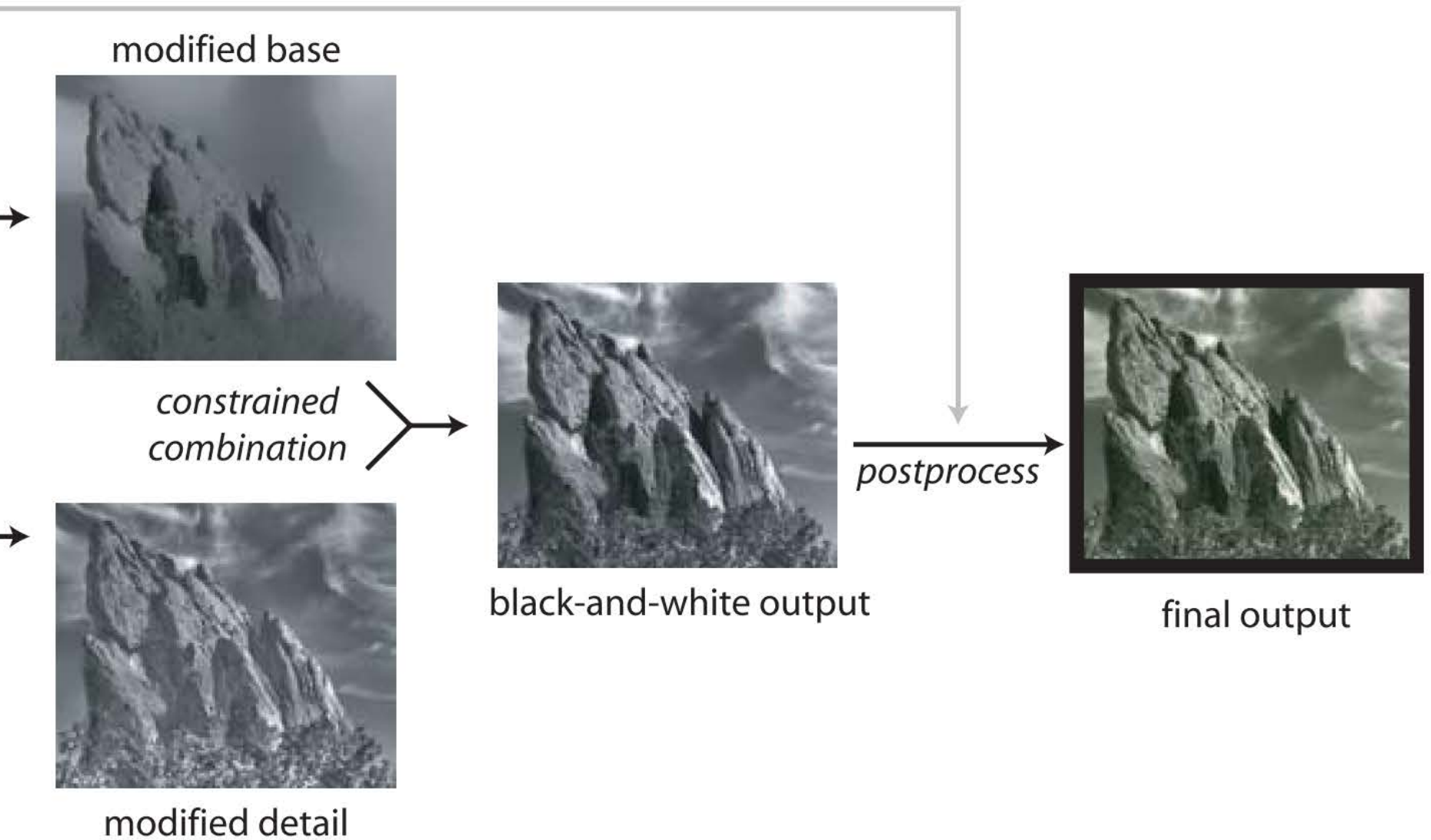


mo

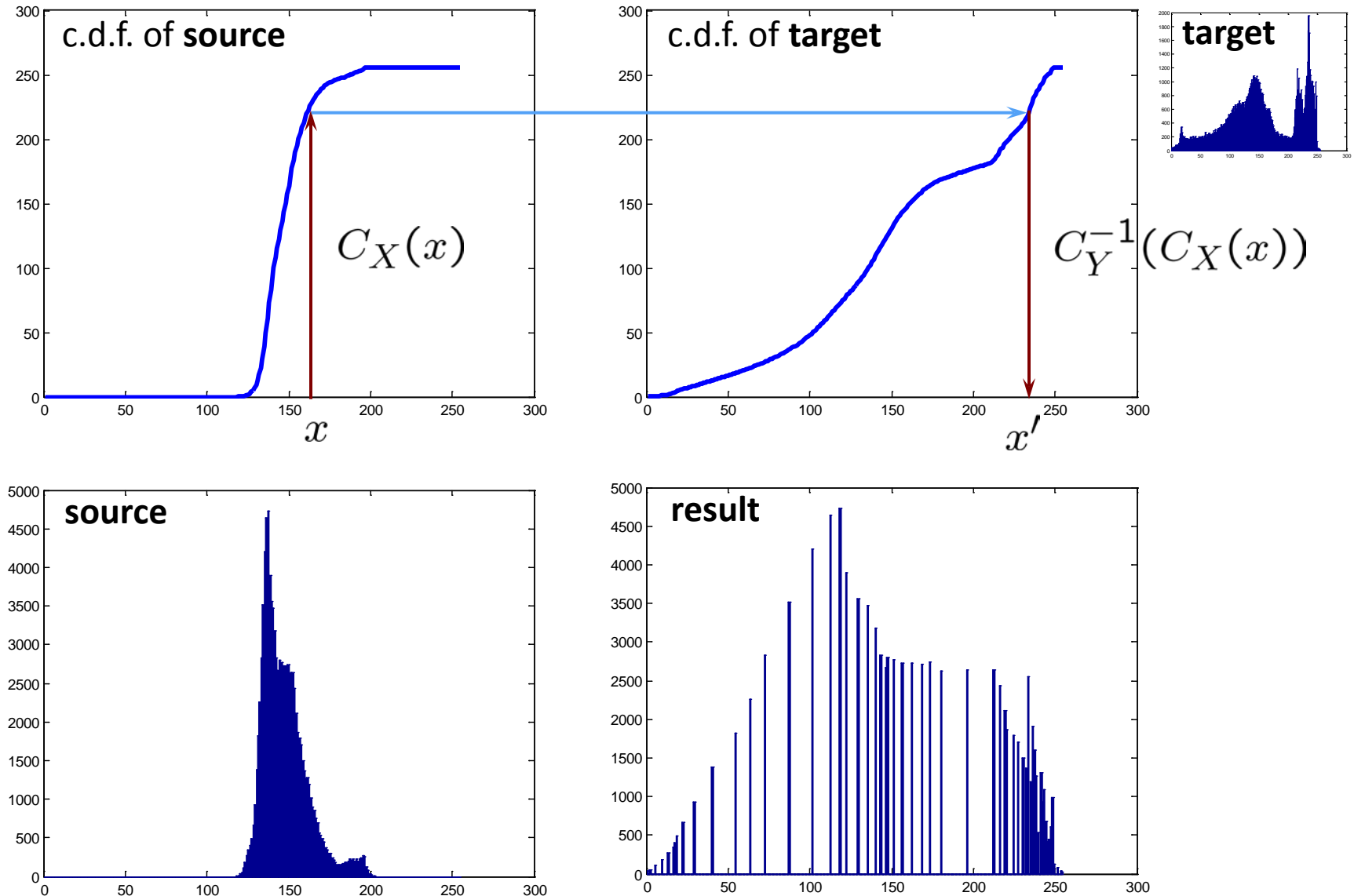




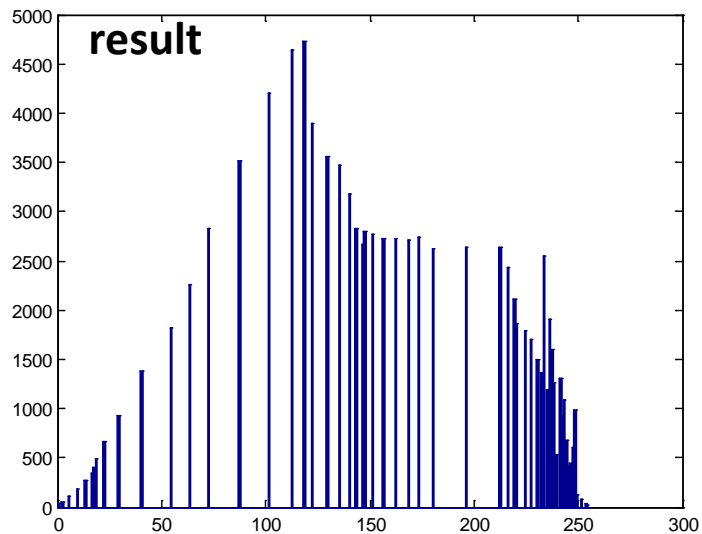
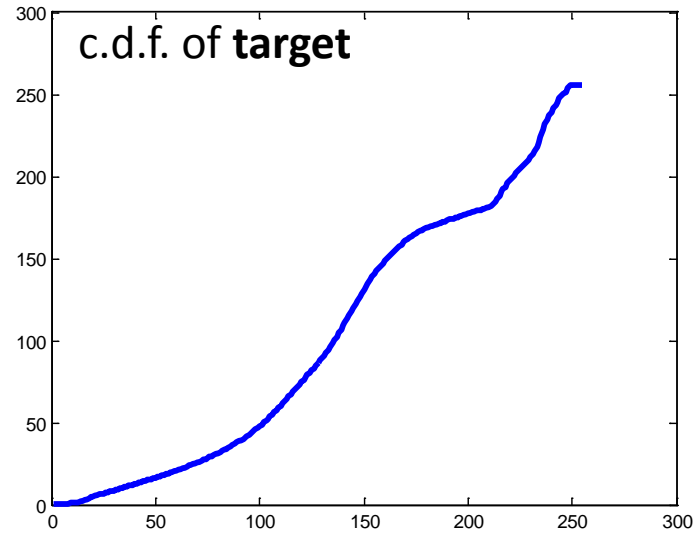
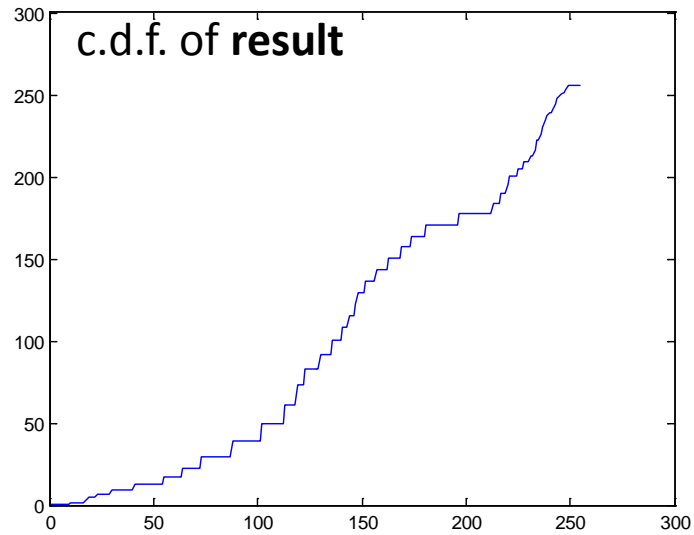




# Histogram Matching



# Histogram Matching



# Poisson Reconstruction

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Given a 2D field of 2D vectors  $\mathbf{v}$ , one can build an image  $I$  with a gradient  $\nabla I$  as close as possible to  $\mathbf{v}$ , in the least square sense.

$$\Delta I = \operatorname{div}(\mathbf{v})$$

There might not exist an image  $I$  such that  $\nabla I = \mathbf{v}$

# Bilateral Filtering

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$$bf(I)_p = \frac{1}{k} \sum_{q \in I} g_{\sigma_s}(\|p - q\|) g_{\sigma_r}(\|I_p - I_q\|) I_q$$

$$k = \sum_{q \in I} g_{\sigma_s}(\|p - q\|) g_{\sigma_r}(\|I_p - I_q\|)$$

$$g_{\sigma}(x) = \exp(-x^2/\sigma^2)$$

$$\sigma_s = \min(\text{width}, \text{height})/16$$

$$\sigma_r = p_{90}(\|\nabla I\|)$$

<sup>1</sup>For an image  $I$ ,  $p_n(I)$  is the intensity value such that  $n\%$  of the values of  $I$  are under it, *e.g.*  $p_{50}(I)$  is the median. Percentiles are robust to outliers.

# Large-Scale Tonal Distribution

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# Bilateral Decomposition (log-Domain)

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- › Base layer

$$B = bf(I)$$

- › Detail layer

$$D = I - B$$

# Uncertainty

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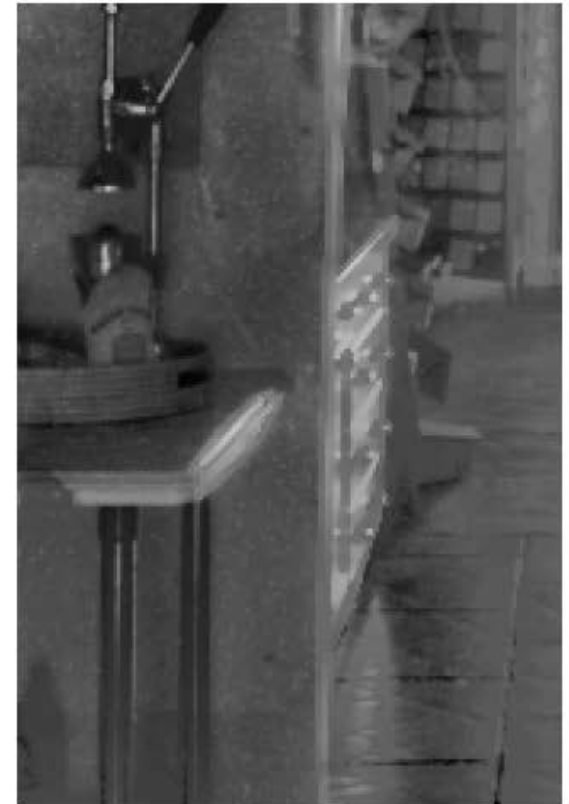
- › The bilateral filter can cause gradient reversals in the detail layer near smooth edges.



(a) input image



(b) uncorrected detail



(c) corrected detail

# Gradient Reversal Removal

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- › Uncertainty
- › We want a reliable halo-free detail layer

build the gradient field  $\mathbf{v} = (x_{\mathbf{v}}, y_{\mathbf{v}})$

$$x_{\mathbf{v}} = \begin{cases} 0 & \text{if } \text{sign}(\partial D / \partial x) \neq \text{sign}(\partial I / \partial x) \\ \partial I / \partial x & \text{if } |\partial D / \partial x| > |\partial I / \partial x| \\ \partial D / \partial x & \text{otherwise} \end{cases}$$

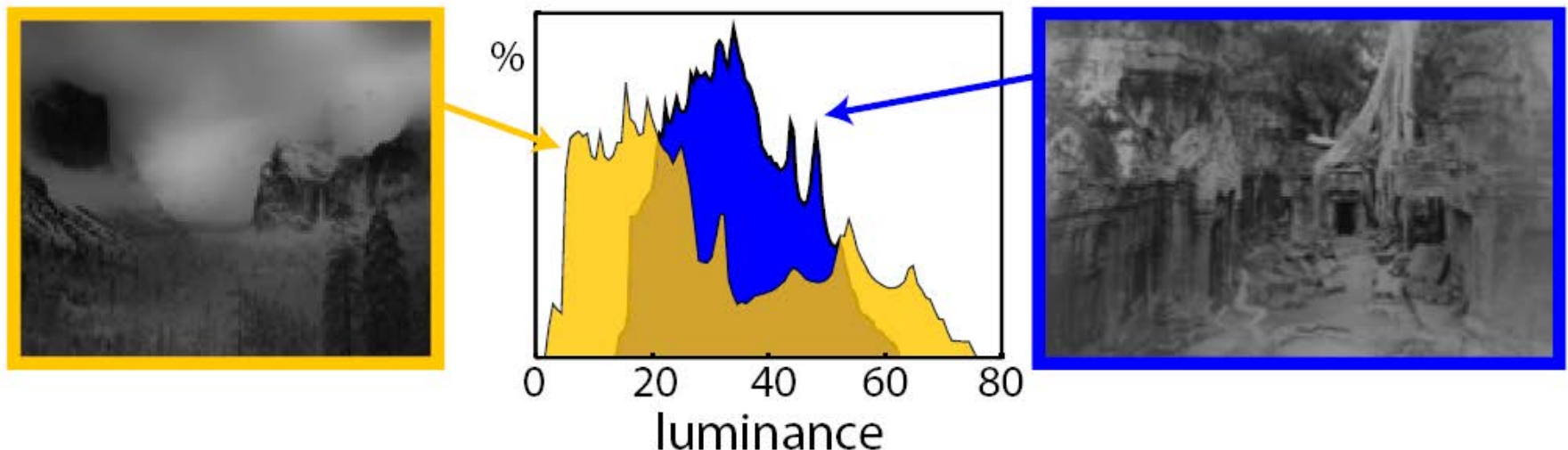
the corrected detail layer is obtained by solving the corresponding Poisson equation.

update the base layer  $B = I - D$

# Tonal Balance

- › Perform histogram matching and transfer the histogram of model base layer onto the new base layer

the luminance histogram of the base component is a good indicator of tonal balance



# Detail and Texture Management

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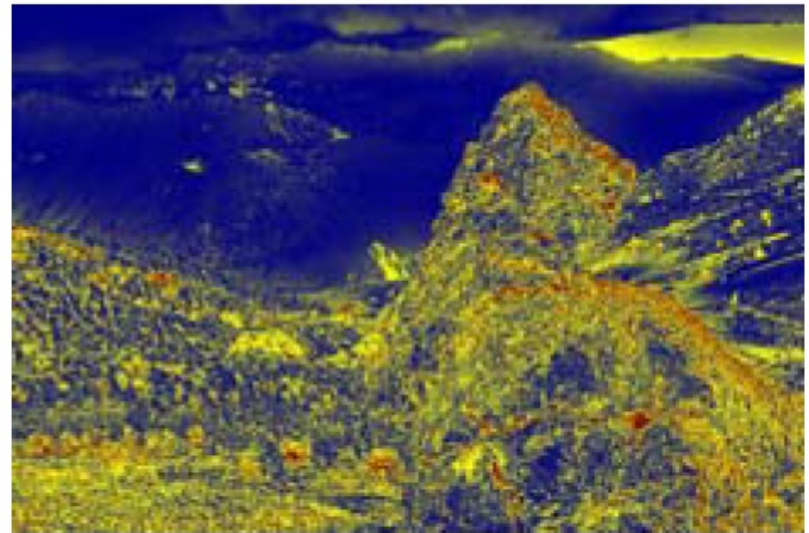
# High Frequencies of Input

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(a) input

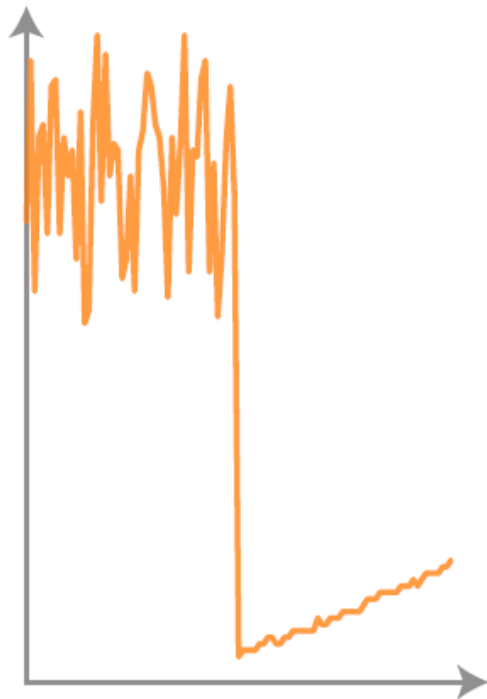


(b) high frequencies of input

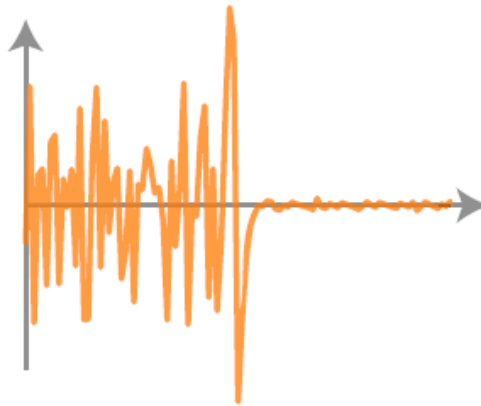


# Detail and Texture Management

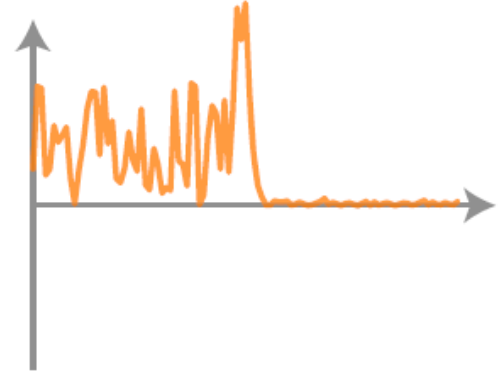
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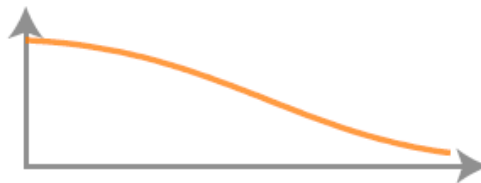
(a) input  $I$



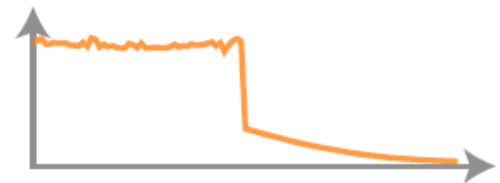
(b) high frequencies  $H$



(c) absolute values  $|H|$



(d) activity/power map:  
low pass of  $|H|$



(e) texture: cross  
bilateral filter of  $|H|$  and  $I$

# Textureiness:

## Local Amount of High Frequency Content

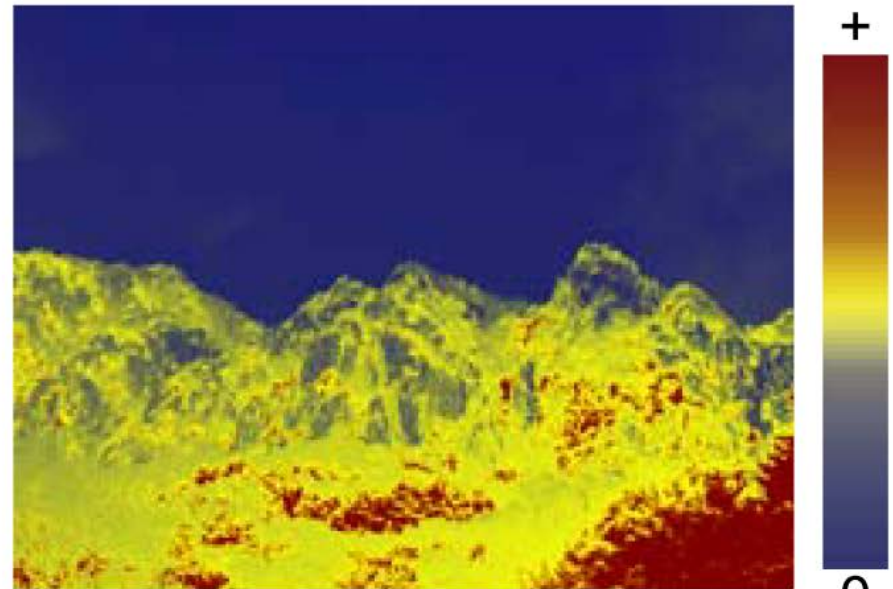
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$$T(I)_p = \frac{1}{k} \sum_{\mathbf{q} \in |H|} g_{\sigma_s}(\|\mathbf{p} - \mathbf{q}\|) g_{\sigma_r}(|I_p - I_q|) |H|_q$$

with:  $k = \sum_{\mathbf{q} \in I} g_{\sigma_s}(\|\mathbf{p} - \mathbf{q}\|) g_{\sigma_r}(|I_p - I_q|)$



(a) input



(b) textureiness



# Textureness Transfer

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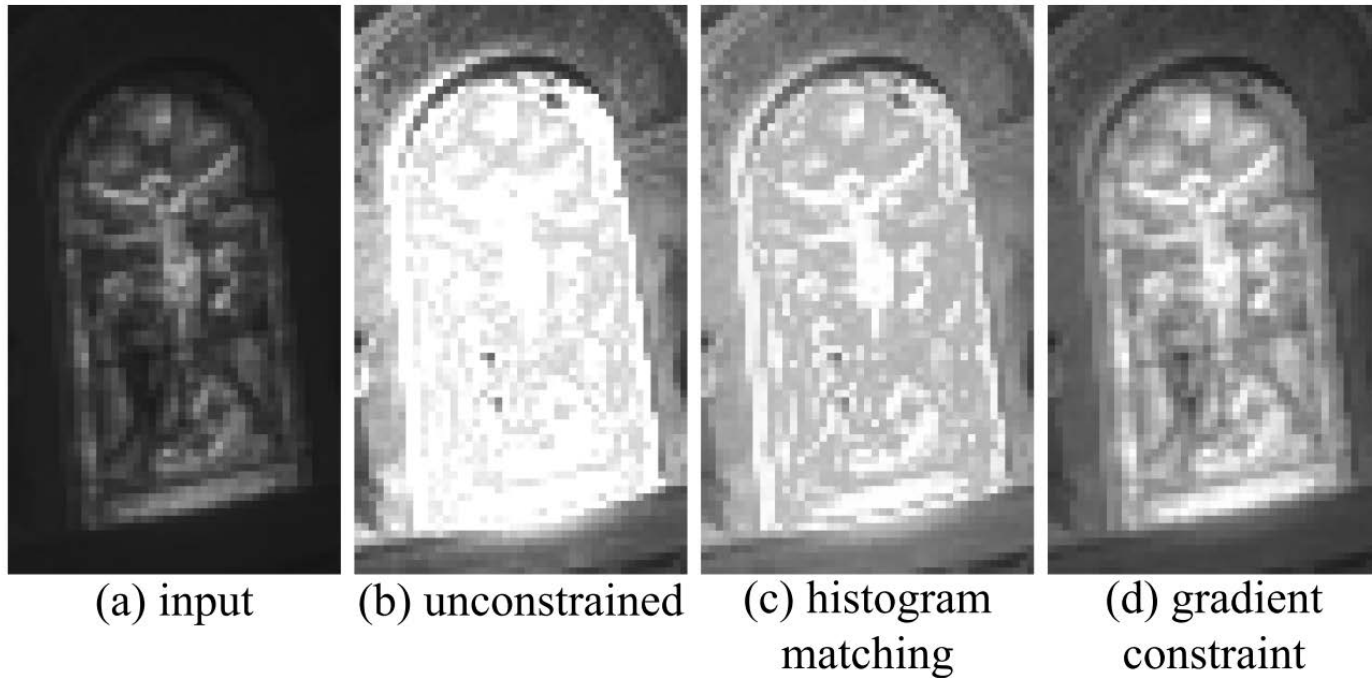
- › The input  $I$  and model  $M$  have textureness maps  $T(I)$  and  $T(M)$ . Using histogram transfer to enforce the histogram of  $T(M)$  onto  $T(I)$  to build the desired textureness map  $T'$
- › Modify the detail layer
  - › Scale the values of the detail layer by a ratio

$$\rho_{\mathbf{p}} = \max \left( 0, \frac{T'_{\mathbf{p}} - T(B')_{\mathbf{p}}}{T(D)_{\mathbf{p}}} \right)$$

output  $O = B' + \rho D$ .

# Fine Tuning

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$$x_v = \begin{cases} \alpha \partial I / \partial x & \text{if } |\partial O / \partial x| < \alpha |\partial I / \partial x| \\ \beta \partial I / \partial x & \text{if } |\partial O / \partial x| > \beta |\partial I / \partial x| \\ \partial O / \partial x & \text{otherwise} \end{cases}$$

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Film grain: extract the grain from the model; generate a grain layer by texture synthesis

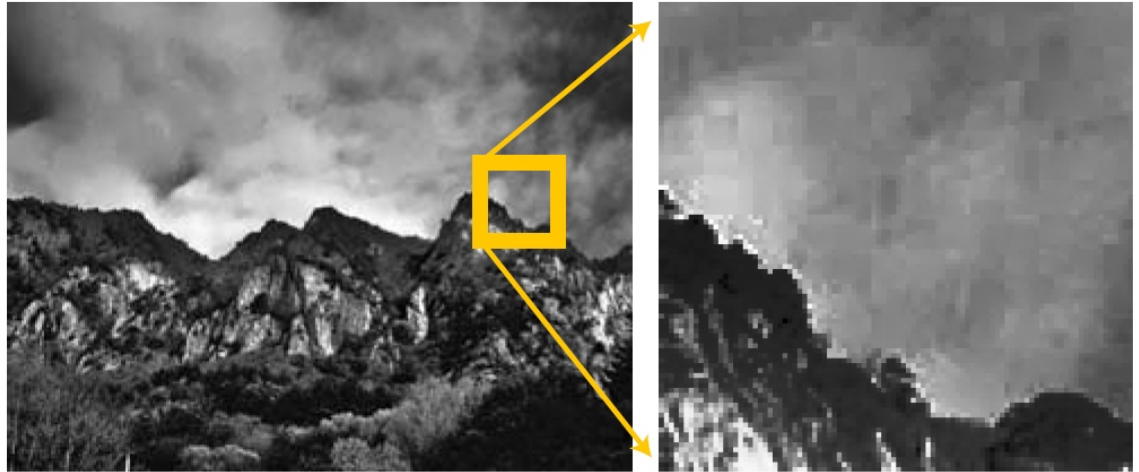
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# Discussion

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- › JPEG artifacts



- › A one megapixel image takes six seconds on a 2.6GHz PC
- › Portraits
  - › Skin defects are emphasized