

# Additional Material

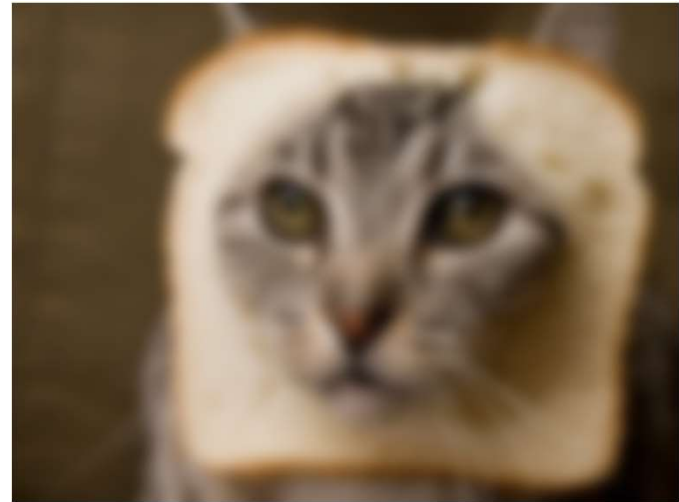
Bilateral Filter

# Smoothing an image

- Box filter leads to slightly “blocky” appearance



Box Filter



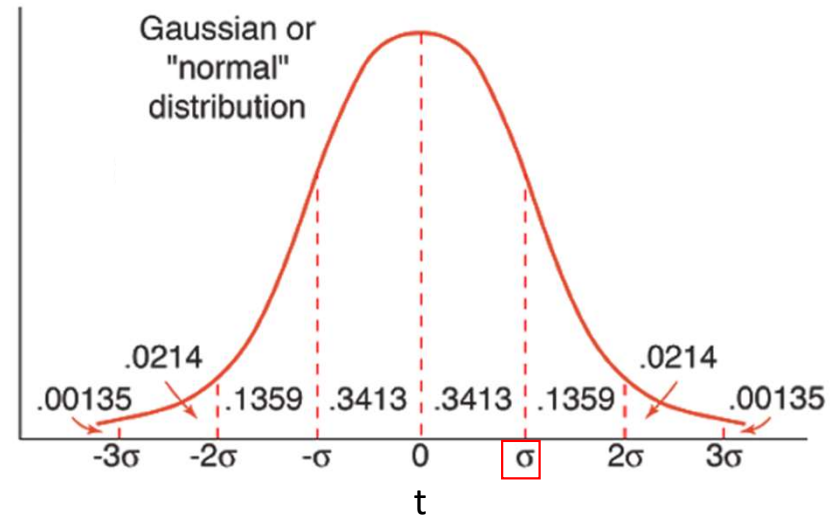
Gaussian Filter

- Reason can be explained by analyzing the image frequencies (out of scope)
- Gaussian filter is better at removing high-frequency content

# Gaussian Filter

- Gaussian Kernel

$$G(t, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{t^2}{2\sigma^2}}$$



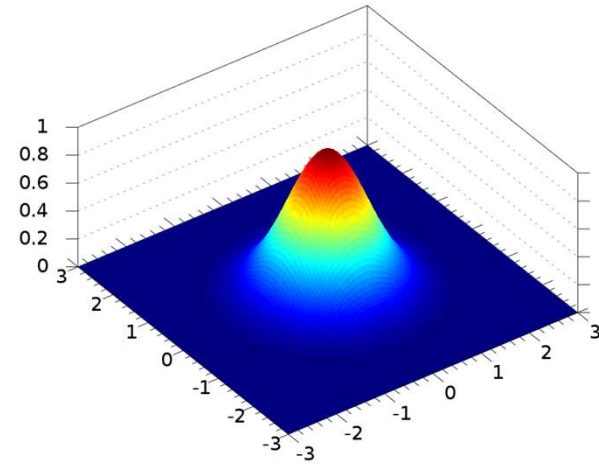
- The larger  $\sigma$ , the broader the function
- The smaller  $\sigma$ , the slimmer the function

# Gaussian Filter

- 2D Gaussian Kernel

$$G((i, j), \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(i^2+j^2)}{2\sigma^2}}$$

Normalization factor  
It makes sure that  
integrating the whole  
function results in a value  
of 1.



## Gaussian Filter – in practice

- Can you calculate a 3x3 kernel for  $G((i, j), 0.5)$  ?

$$G((i, j), \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(i^2+j^2)}{2\sigma^2}}$$

- What you will notice is that the sum of these 9 pixels is larger than 1.
- Hence, applying the filter several times, the image would become brighter in this case

# Gaussian Filter – in practice

- A simple fix to this problem is to normalize:
- Divide the kernel by the sum of all elements

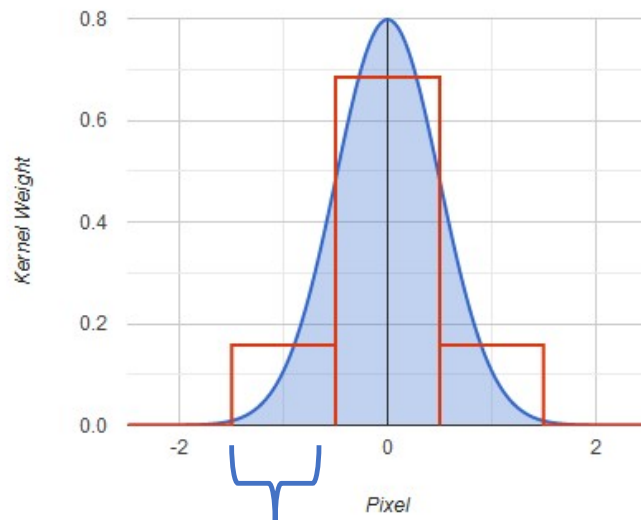
0.0	0.1	0.0
0.1	0.6	0.1
0.0	0.1	0.0

- Btw., if you normalize:

$$G((i,j), \sigma) = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(i^2+j^2)}{2\sigma^2}}$$

# Gaussian Filter – in practice

- A more precise solution:
- integrate the function over the pixels



Integrate here for  
 $G((1,0),0.5)$

<http://dev.theomader.com/gaussian-kernel-calculator/>

# Gaussian Filter – in practice

How large should the filter be?

- The number of pixels in the filter is linked to sigma
- Rule of thumb: round to odd ( $6 * \sigma$ ) as a filter size
- e.g.,  $\sigma = 0.5$  should be 3 pixels.



# Gaussian Filter – in practice

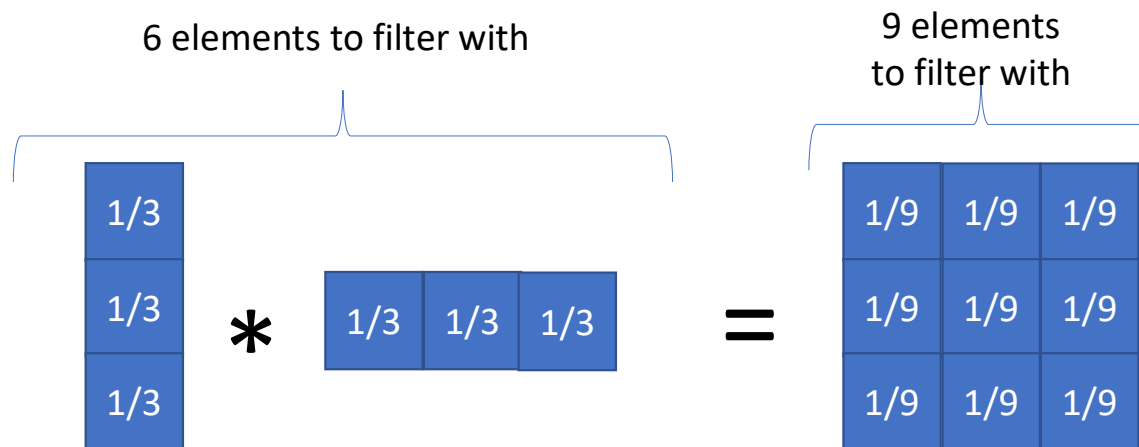
- Gaussian filter (and Box filter as well) is separable
- Separable means that 2 convolutions with a 1D kernel can be performed instead of 1 convolution with a 2D kernel:

Gaussian along the X axis  $G_x$ , followed by a Gaussian along Y axis  $G_y$

$$((I * G_x) * G_y) = (I * (G_x * G_y)) = I * G$$

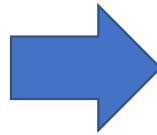
- This is efficient because of linear cost instead of quadratic cost

Example of separability  
for the Box filter:

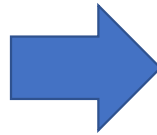


# Gaussian Filter

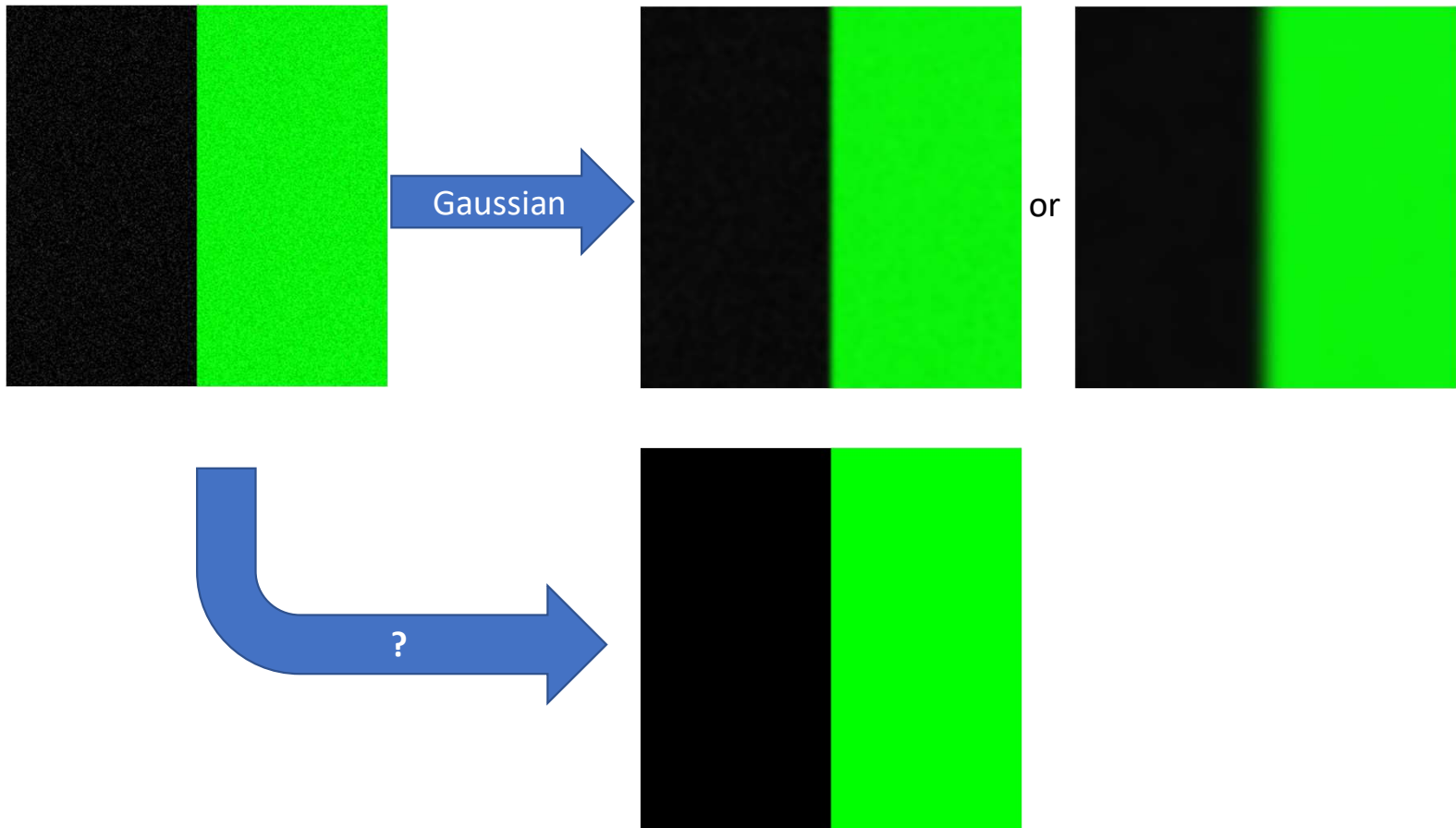
- Rem



## Smarter Filters?

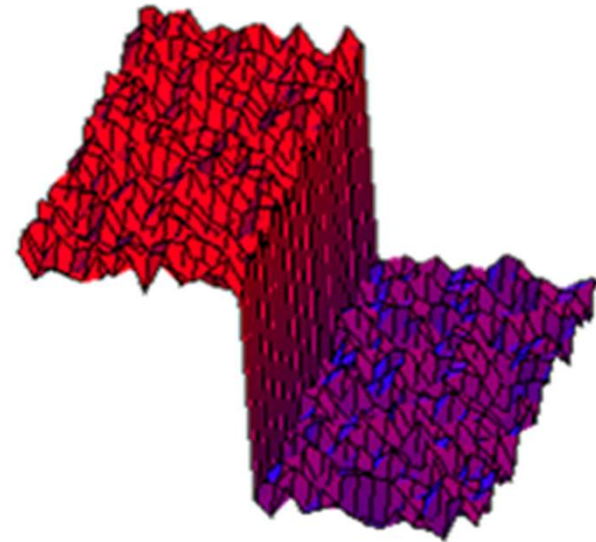
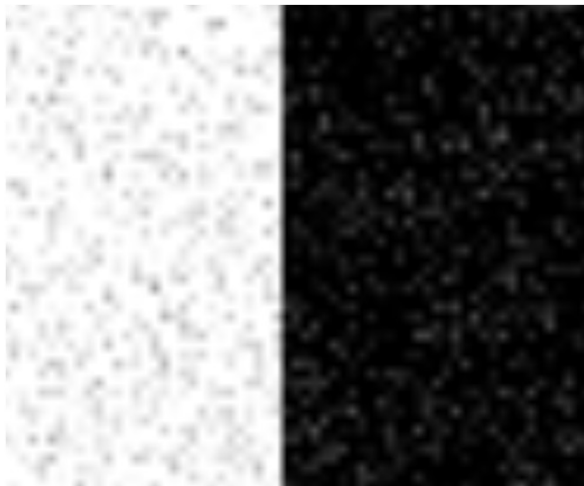


# Smarter Filters?



# Visualization

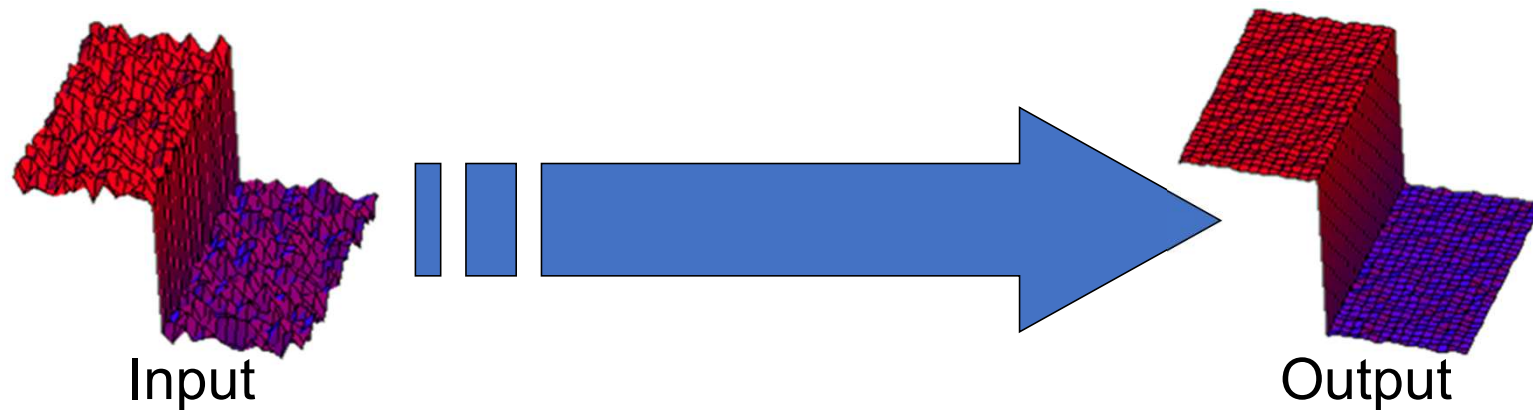
- To ease understanding, we will illustrate images as a “terrain”



# Bilateral Filter

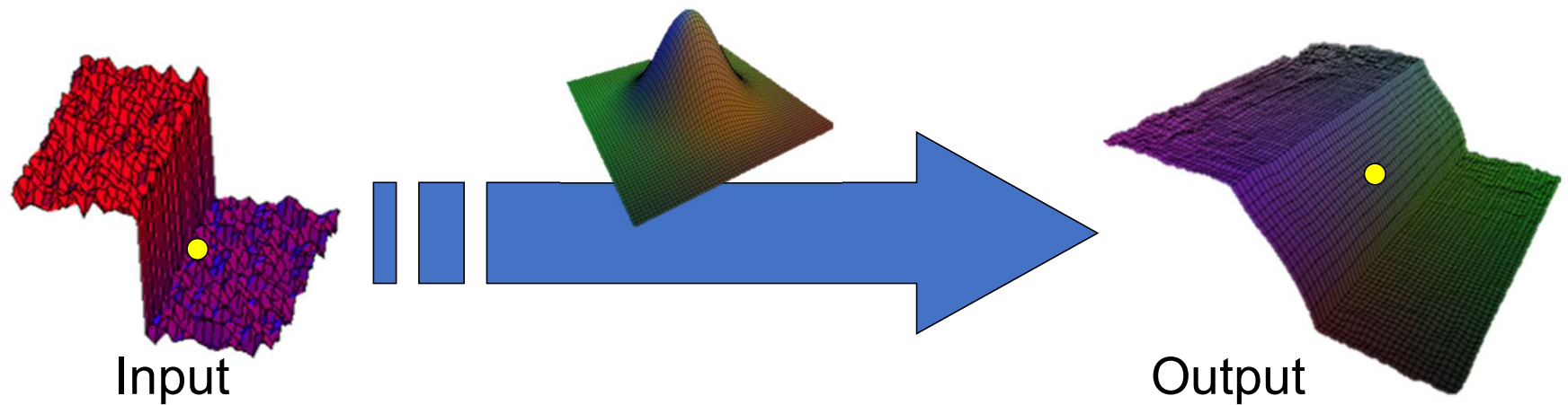
- **Bilateral filter** – edge preserving filter

Smith and Brady 1997; Tomasi and Manducci 1998; Durand et al. 2002



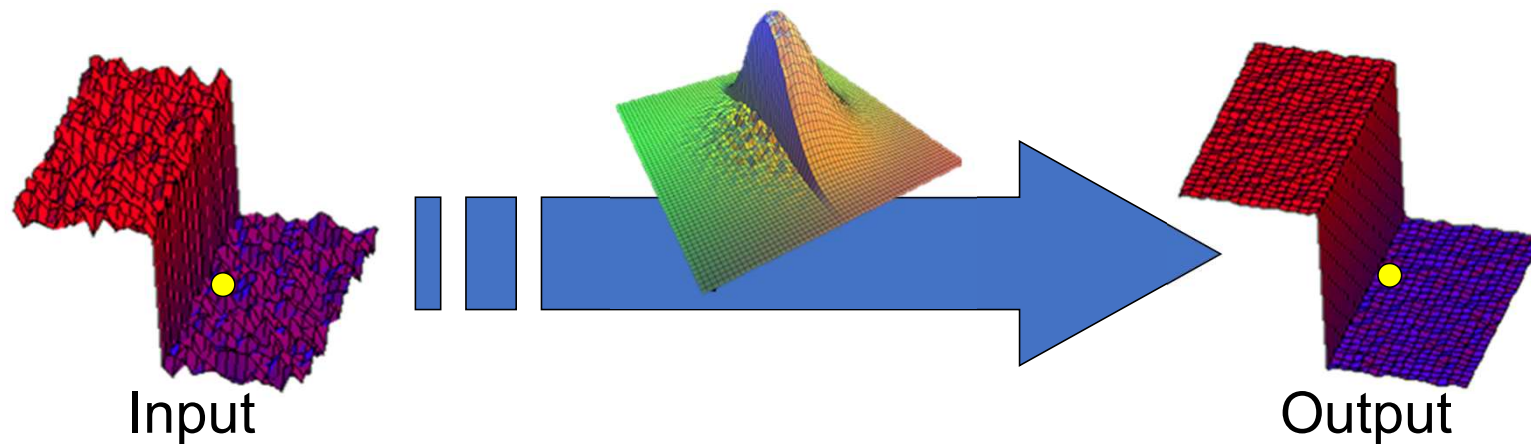
# Bilateral Filter: Motivation

- Gaussian



# Bilateral Filter: Illustration

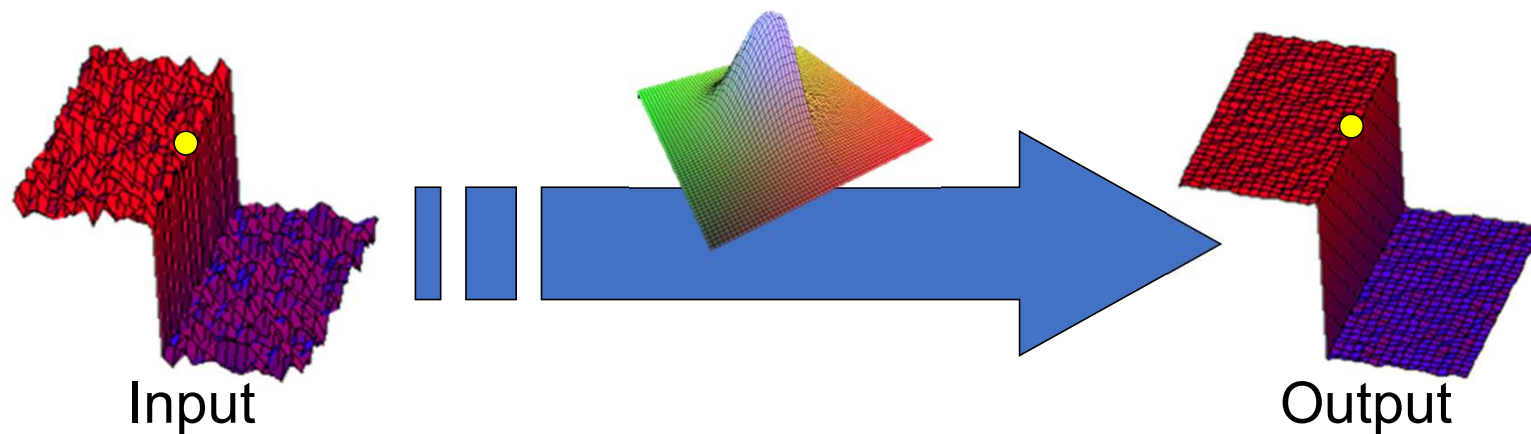
- **Bilateral filter** – edge preserving filter  
depending on the center pixel, the filter looks different



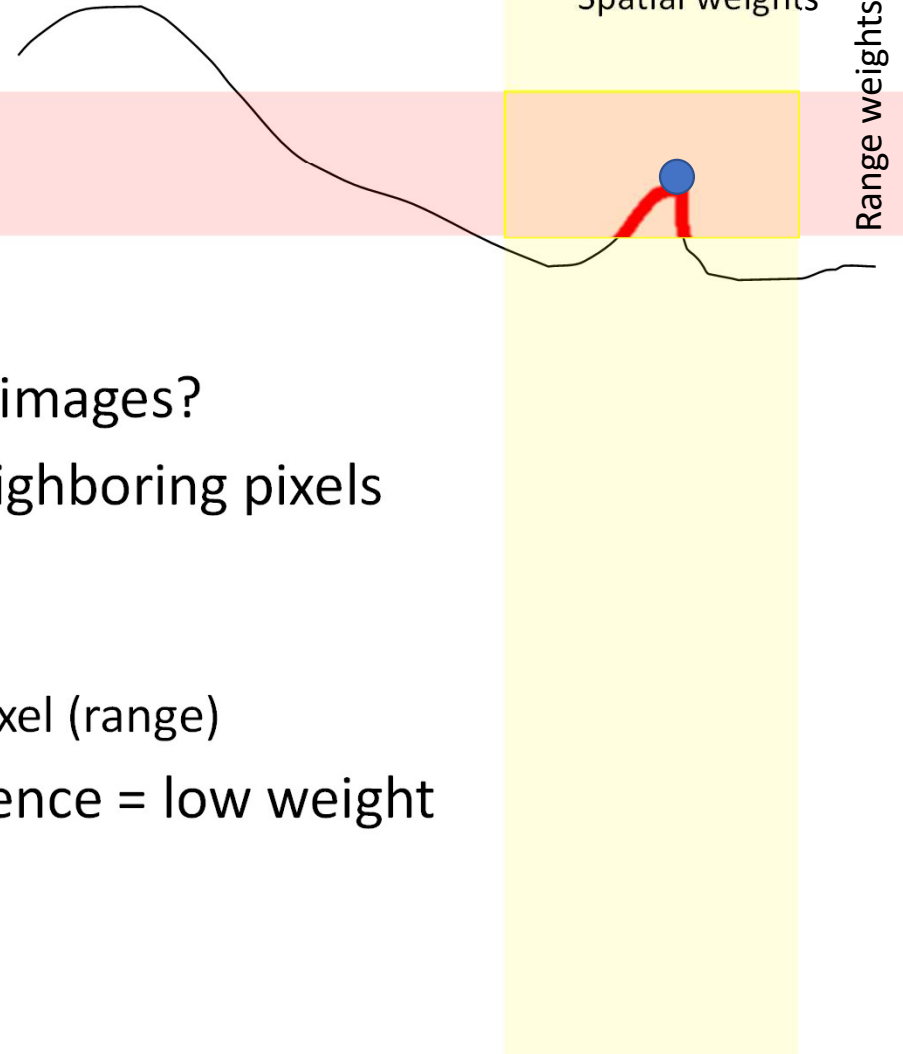


# Bilateral Filter: Illustration

- **Bilateral filter** – edge preserving filter  
depending on the center pixel, the filter looks different



# Bilateral Filter: Definition



- How does the filter work on grayscale images?
- Weighted average over intensity of neighboring pixels
- Weights based on
  - Distance from center (spatial)
  - Difference in intensity from the center pixel (range)
- Large distance or large intensity difference = low weight

## Bilateral Filter: Example



# Bilateral Filter: Formal Definition

- $f, g$  are filters (spatial and range respectively), typically Gaussians
- $I$  is the image to be filtered

$$J(x) = \frac{1}{k(x)} \sum_y \overbrace{f(x-y)}^{\text{Penalizes spatial distance (distance of position y to position x)}} \underbrace{g(I(x) - I(y))}_{\text{Penalizes value distance (distance of value at y to value at x)}} I(y)$$

$$k(x) := \sum_y f(x-y)g(I(x) - I(y))$$

- $k$  is a normalization factor (compare discussion for Gaussian)

# Bilateral Filter Application: Tone Mapping

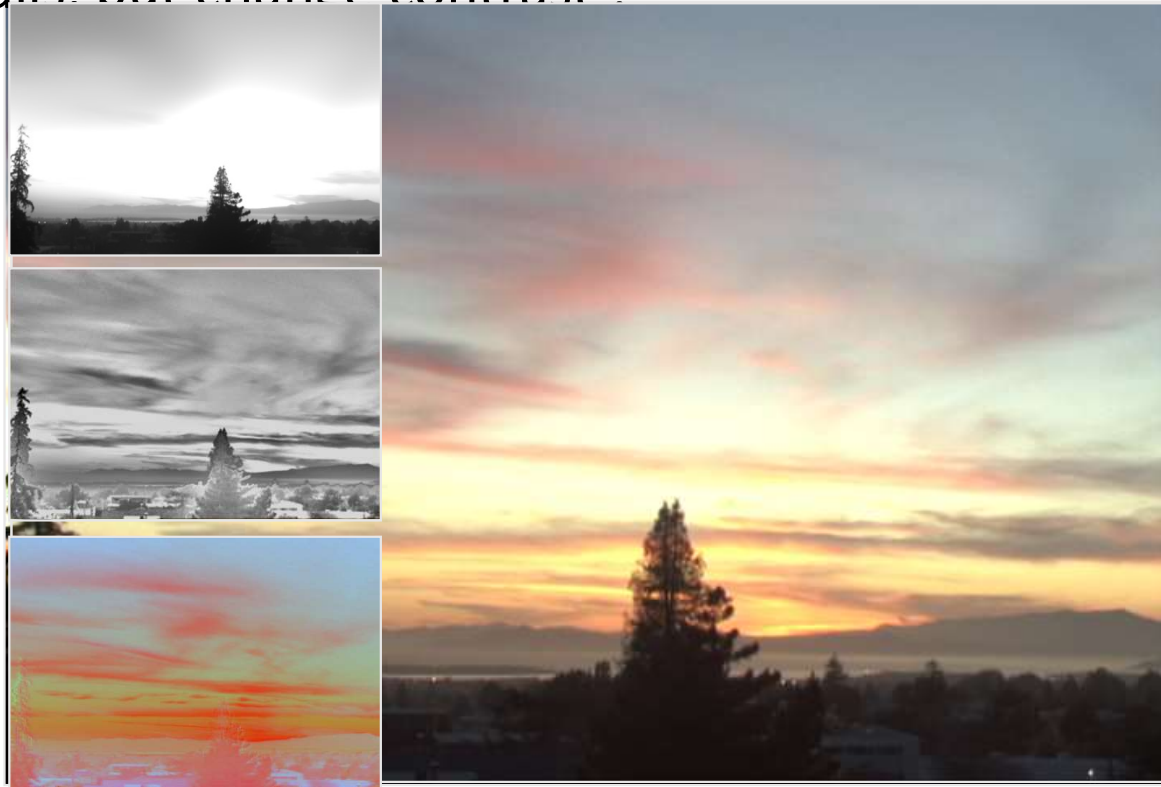
Contrast Reduction!

How to preserve details, but change contrast ?

Large-scale

Detail

Color



# HDR Contrast Reduction [Durand & Dorsey 2002]



Contrast too high!

# HDR Contrast Reduction [Durand & Dorsey 2002]



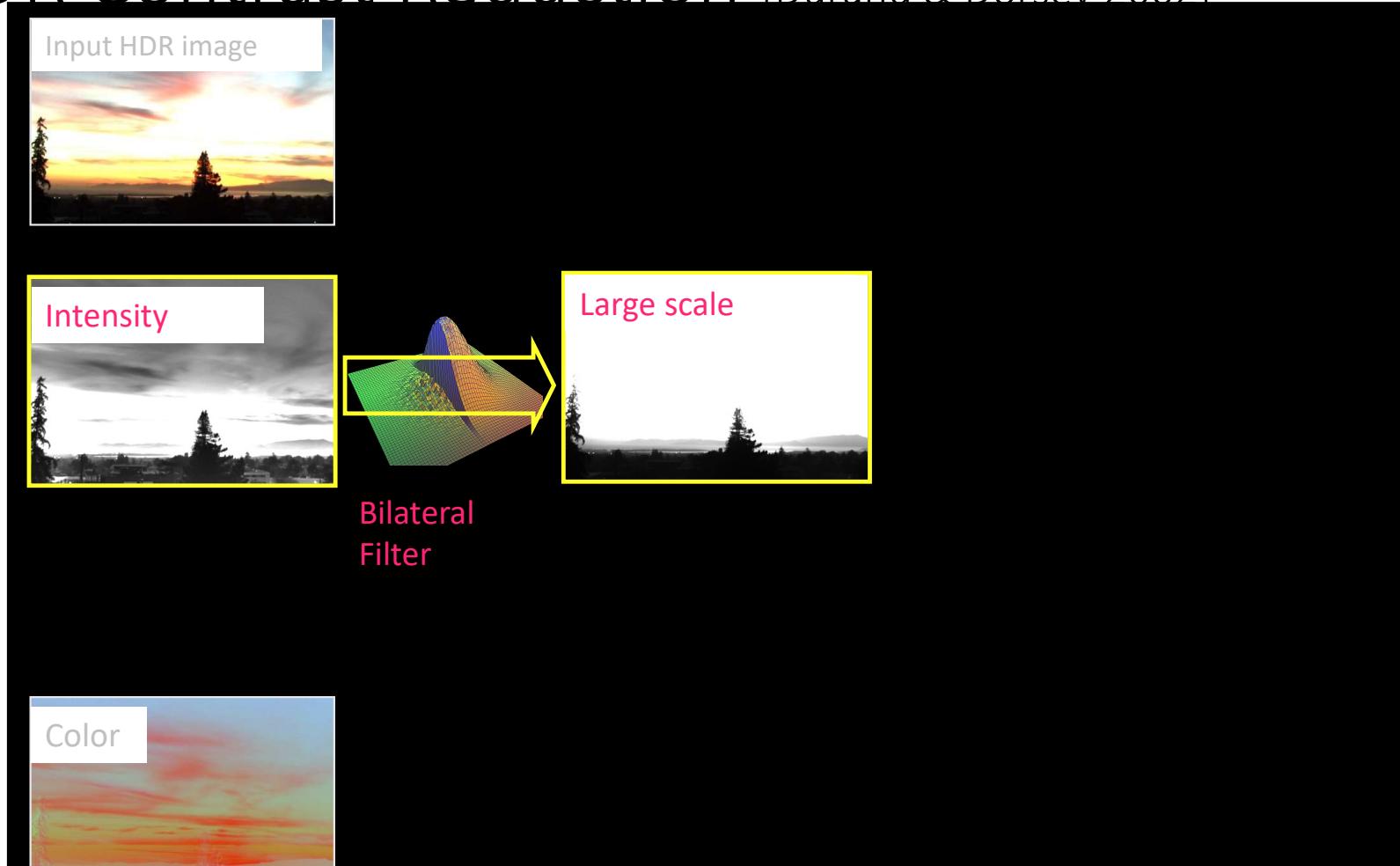
# HDR Contrast Reduction: Intensity/Color

- Input Image pixel (R,G,B)
- $\text{Intensity} = \text{ColorToGrayscale}(R,G,B)$
- $\text{Color} = (R,G,B) / \text{Intensity}$

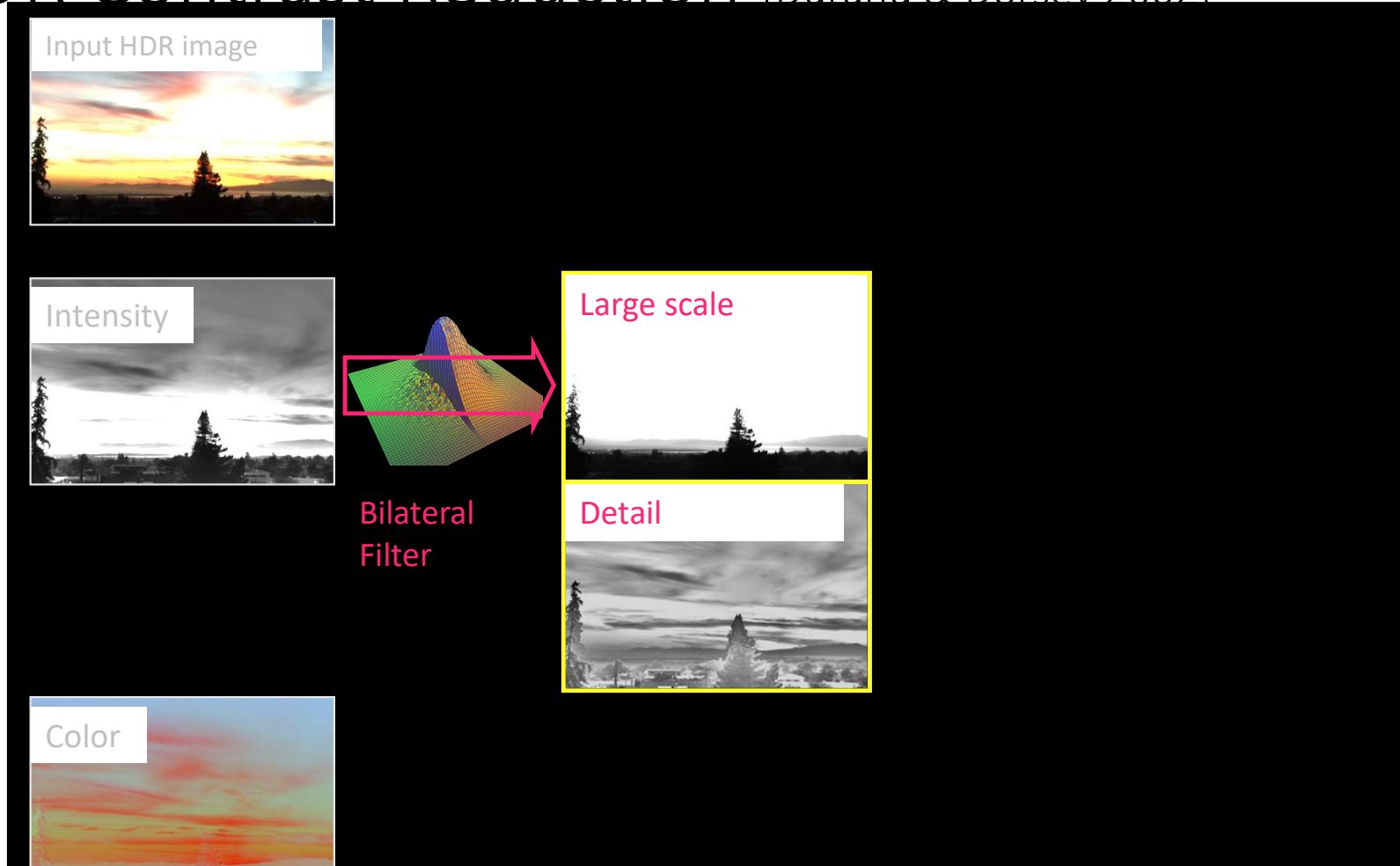




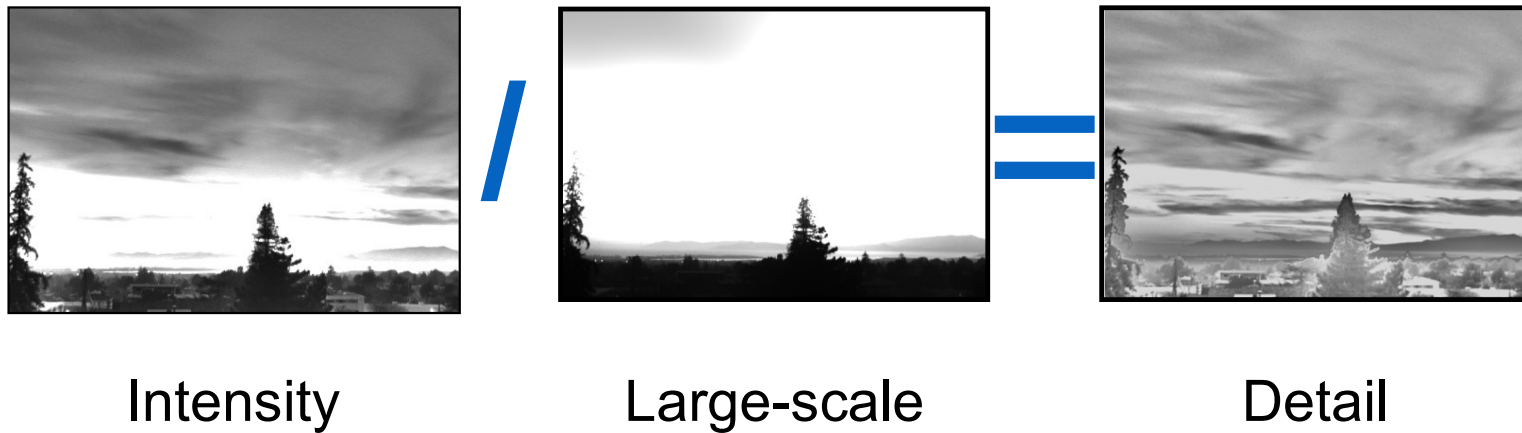
# HDR Contrast Reduction [Durand & Dorsey 2002]



# HDR Contrast Reduction [Durand & Dorsey 2002]

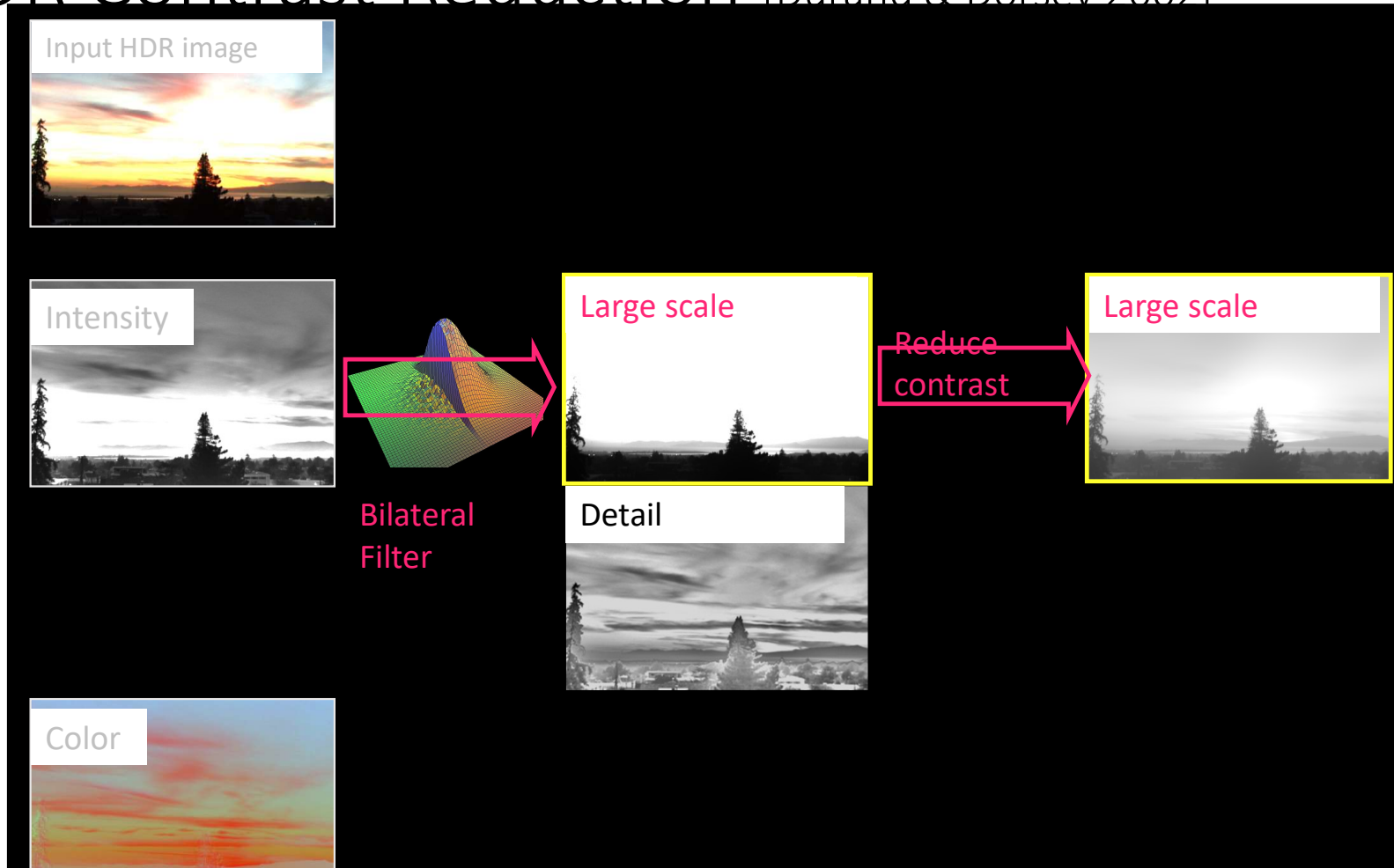


# HDR Contrast Reduction: Detail

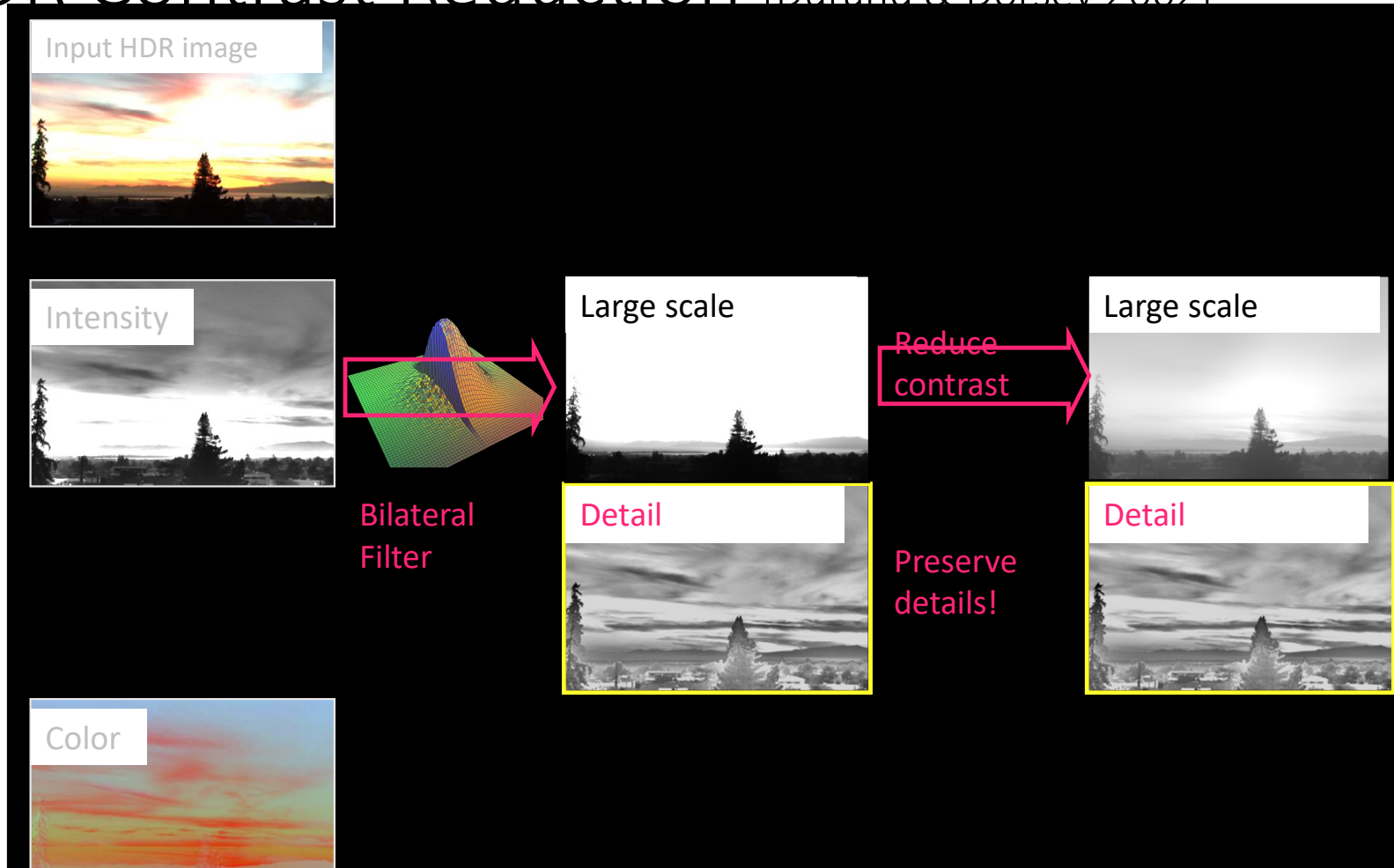


Operation can be inverted:  
Large scale \* Detail = Intensity

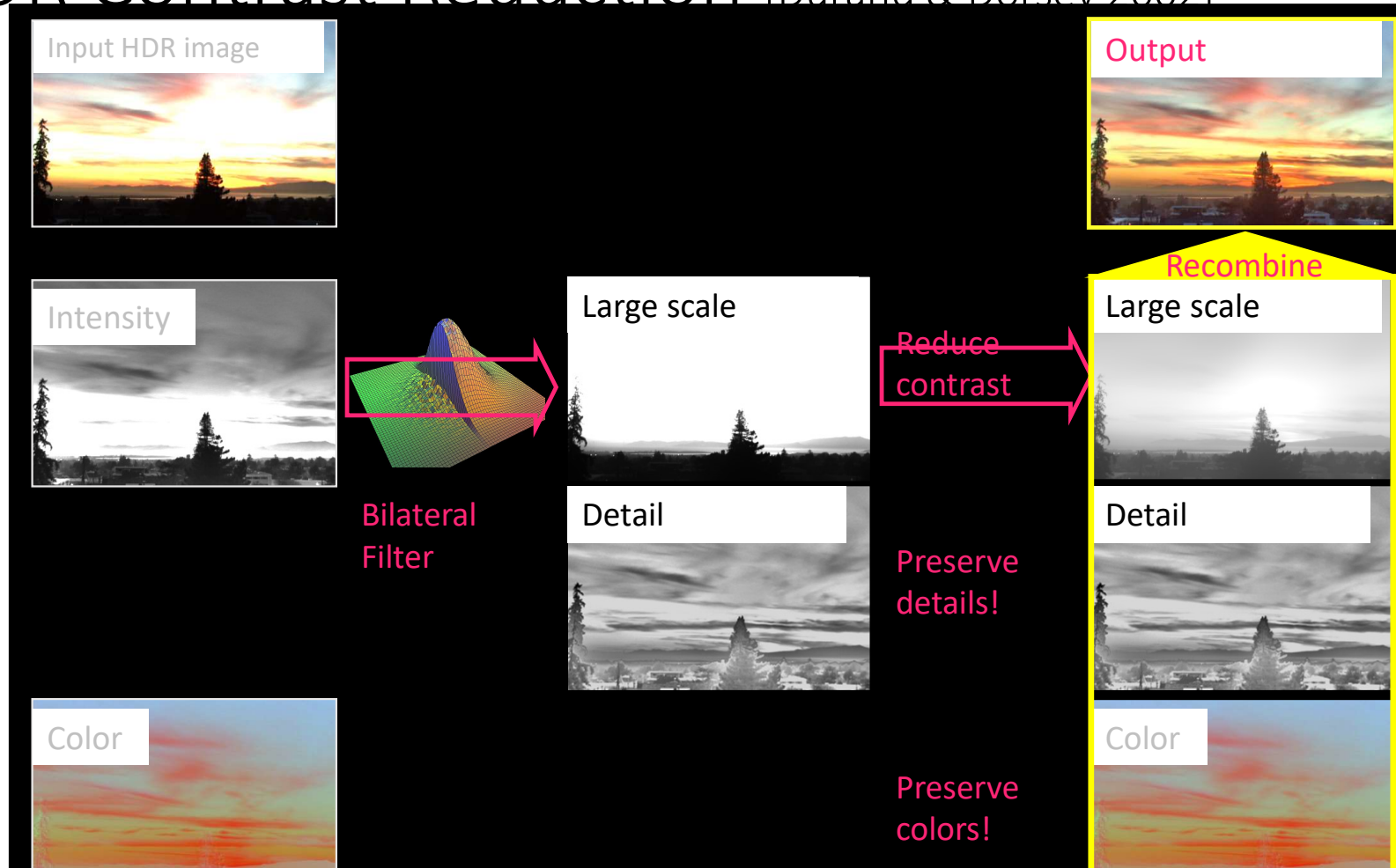
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