# Lecture 13 Image Blending

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# Long history of fake images





# Long history of fake images

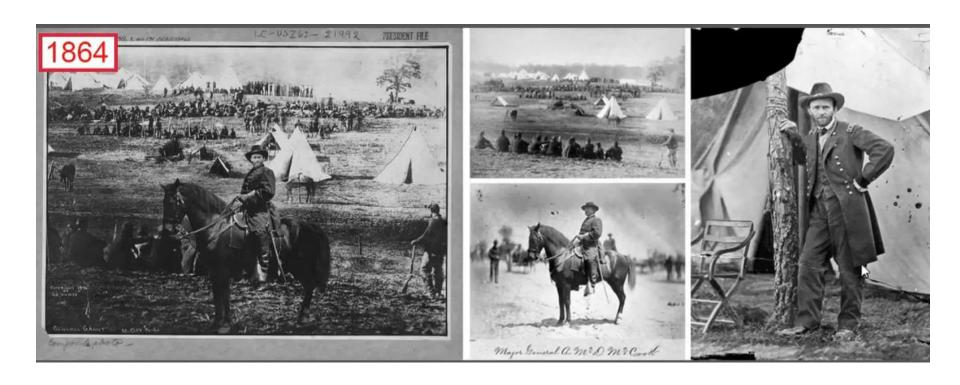






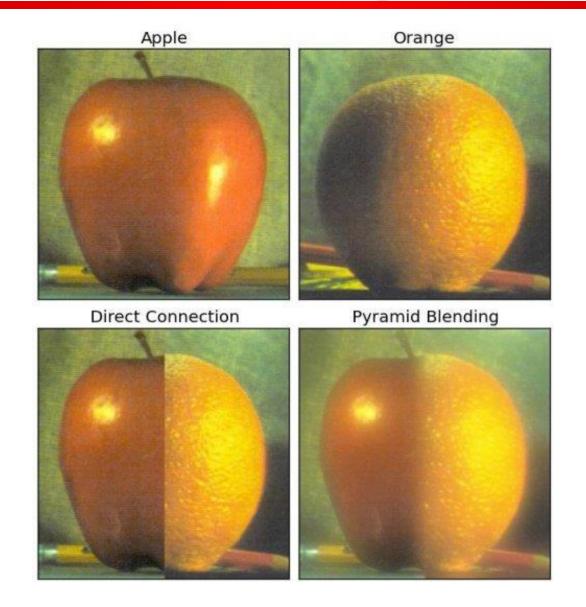


# Long history of fake images





# Hard edge composition vs Pyramid Blending





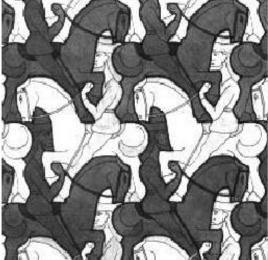
### **Hard compositing**

#### **☐** Hard compositing:

$$I(x,y) = M(x,y)S(x,y) + (1 - M(x,y))T(x,y)$$
$$= \begin{cases} S(x,y) & M(x,y) = 1 \\ T(x,y) & M(x,y) = 0 \end{cases}$$

☐ Generally bad: seam/matte line is visible

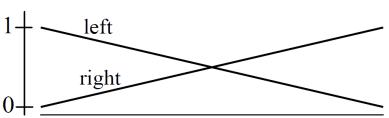




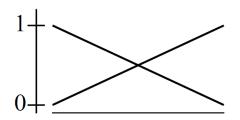


# Weighted transition region





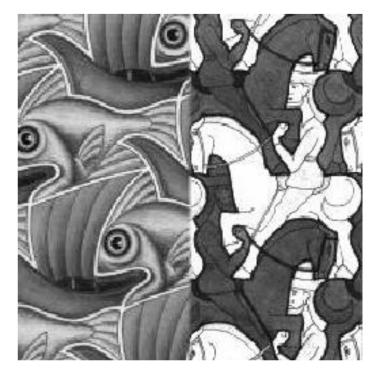






# Weighted transition region

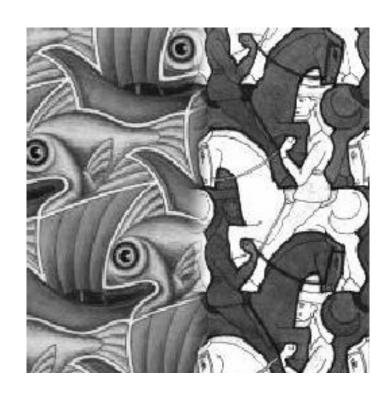








#### **Good window size**







- ☐ Better idea: Multi-resolution blending with a Laplacian pyramid.
- ➤ Idea: wide transition regions for low-frequency component, narrow transition regions for high-frequency component (edges).
- Gaussian pyramid:

G = 5x5 Gaussian filter

 $I_0$  = original image (full resolution)

Convolution

Get a series of smaller and blurry images.



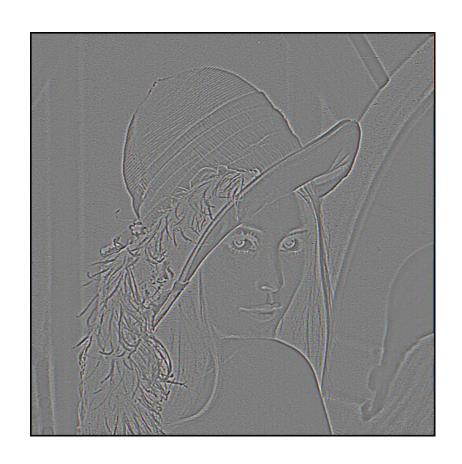
# What does blurring take away?



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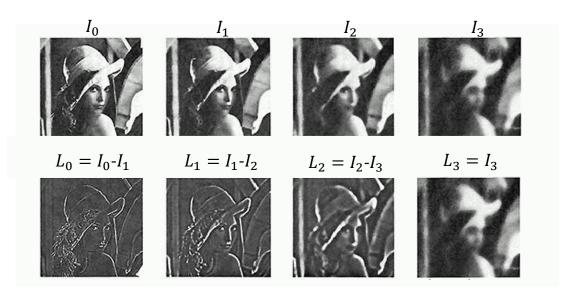


# What does blurring take away?



#### **□** Difference of Gaussian at each scale:

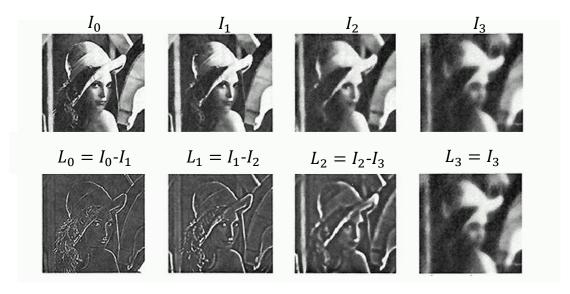
High-pass image at scale i  $\longrightarrow$   $L_i = I_i - (G * I_i) \downarrow 2$  Blurred version of level i Gaussian pyramid image at scale i



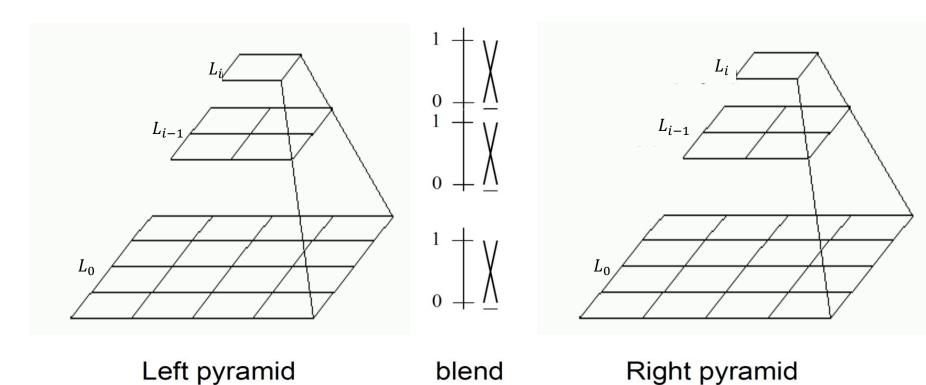
 $\{L_i\}$  = the set of  $L_i$  form. A Laplacian pyramid  $L_1$ ,  $L_2$ ,  $L_3$ ...,  $L_n$ 

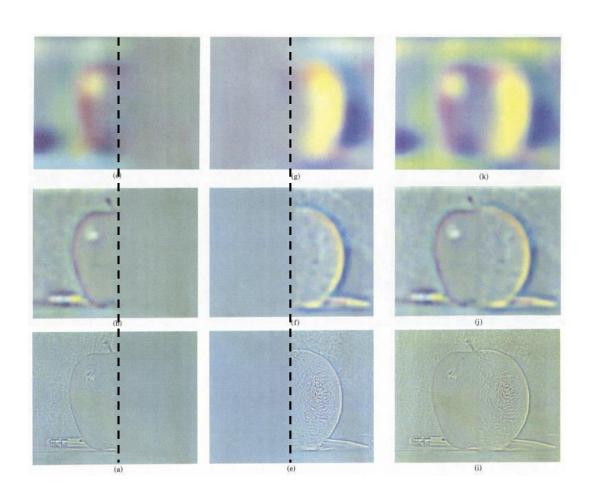
#### **☐** We can recover the original as:

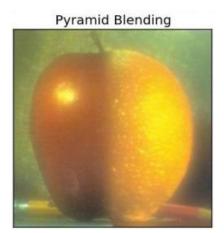
$$I = \sum_{i=0}^{N} (L_i) \uparrow$$



 $\{L_i\}$  = the set of  $L_i$  form. A Laplacian pyramid  $L_1$ ,  $L_2$ ,  $L_3$ ...,  $L_n$ 

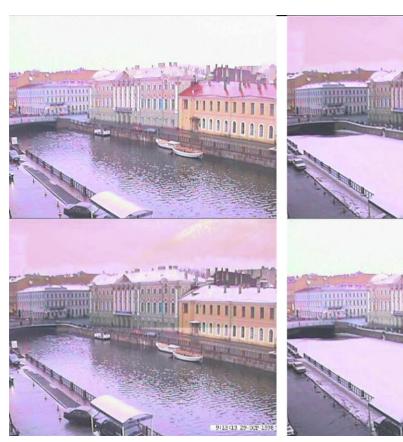


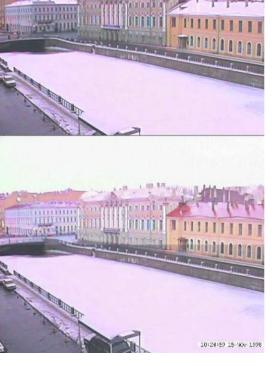






# Season blending





# Season blending





**Target image** 



**Source image** 



Target image with editing region



**Result of pyramid blending** 

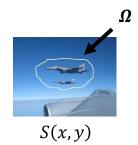




# Poisson image editing

- A better idea: to reduce the color mismatch between source and target, create composite in gradient domain.
- lacksquare We want the gradient of the composite inside  $\Omega$  to look as close as possible to the source image gradient. The composite must match target image on the boundary  $\partial \Omega$ .

$$\min_{I(x,y)\in\Omega} \|\nabla I(x,y) - \nabla S(x,y)\|^2 dxdy$$



$$s.t.I(x,y) = T(x,y) \text{ on } \partial\Omega$$



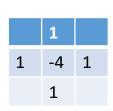
## Poisson image editing

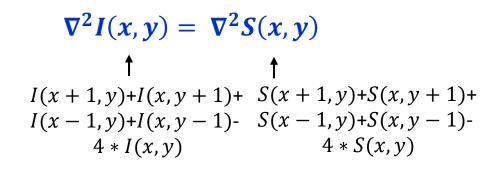
Solution for this Pb:

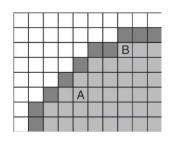
$$abla^2 I(x,y) = 
abla^2 S(x,y) \ in \Omega$$
 $I(x,y) = T(x,y) \ on \partial \Omega$ 

- Poisson equation
- Discretizing and solving the problem:









## Poisson image editing

 $\square$  For a pixel B not inside  $\Omega$  (whose neighbor is  $\Omega$ ).

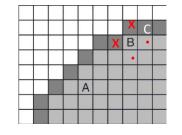
$$\nabla^{2}I(x,y) = \nabla^{2}S(x,y)$$

$$\uparrow \qquad \qquad \uparrow \qquad \qquad \uparrow$$

$$I(x+1,y)+I(x,y+1)+ (..) \qquad \qquad S(x+1,y)+S(x,y+1)+$$

$$T(x-1,y)+T(x,y-1)- (xx) \qquad \qquad S(x-1,y)+S(x,y-1)-$$

$$4*I(x,y) \qquad \qquad 4*S(x,y)$$



Big linear system: so in all there will be N unknowns and N

equations that can be divided into 3 different groups

row

| 5 non-zeros<br>values in a<br>row | [-4<br>1 | 1<br>-4          | 0<br>1 | <br>0       | <b>0</b><br> | 1<br>0 | 0<br>1 |      | 0<br>0 | $\begin{bmatrix} I_1 \\ I_2 \end{bmatrix}$ |   | $\begin{bmatrix} & \nabla^2 S_1 \\ & \nabla^2 S_2 \end{bmatrix}$ | Group 1: $A \in \Omega$                               |
|-----------------------------------|----------|------------------|--------|-------------|--------------|--------|--------|------|--------|--|---|--|---|
| 3 non-zeros<br>values in a<br>row |          | 1 -              | -4     | 1           | :            |        |        |      |        | :  | = | $ abla^2 S_n - \sum_{i=1}^{n} T_n $                              | Group 2: $\mathbf{B} \in \partial \Omega \cap \Omega$ |
| 1 non-zeros<br>value in a         | 0        | - <del>o</del> - |        | <b>-1</b> - | <b>70</b> -  | 0      | 0      | TO T |        | $I_N$                                      |   | $T_N$  | Group 3: $C \in \partial \Omega$                      |

Source image



Target image



Poisson image editing result





### Take home message

- ☐ Pyramid image blending is able to merge two images with similar background, but it is not robust for color mismatch.
- □ Poisson image edit is more powerful on image blending Pbs with variations on background color.

