# Lecture 14-Image Blending

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## Long history of fake images





## Long history of fake images

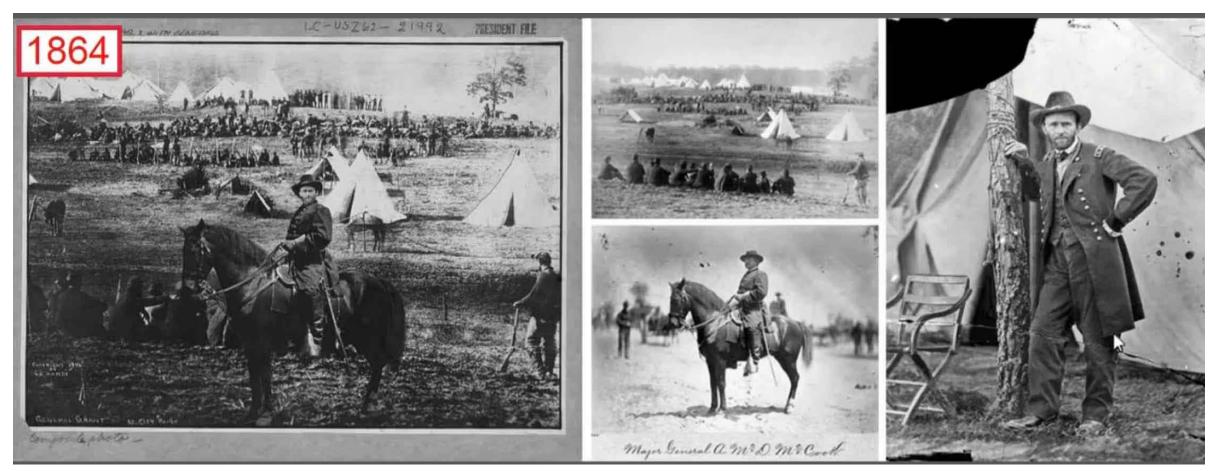






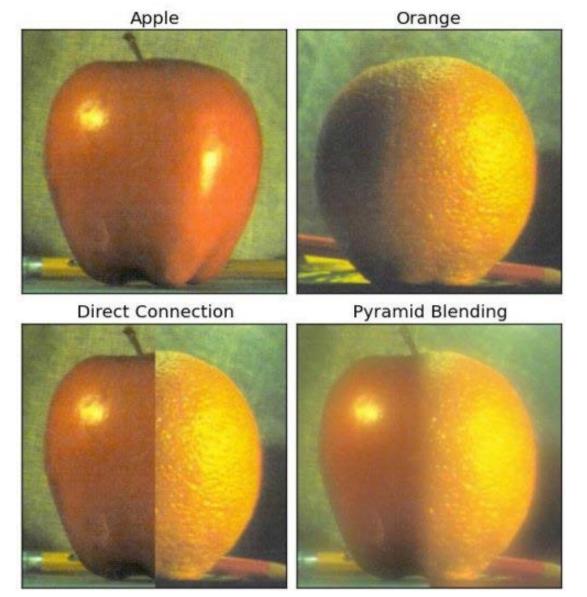


## Long history of fake images





## Hard edge composition vs Pyramid Blending





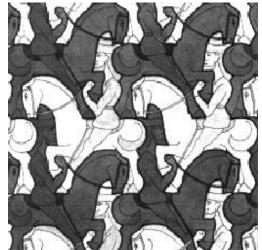
### Hard compositing

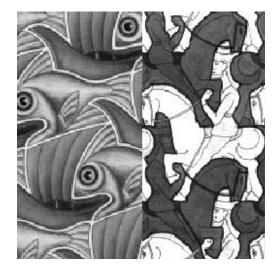
> Hard compositing:

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

> Generally bad: seam/matte line is visible



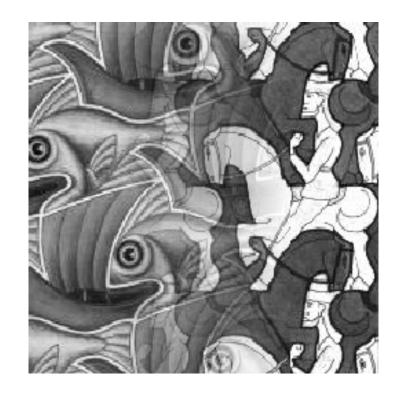


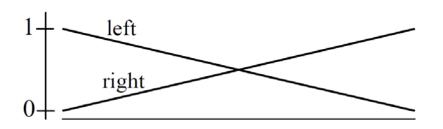


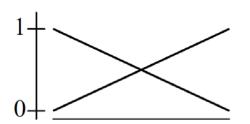


## Weighted transition region:











## Weighted transition region:



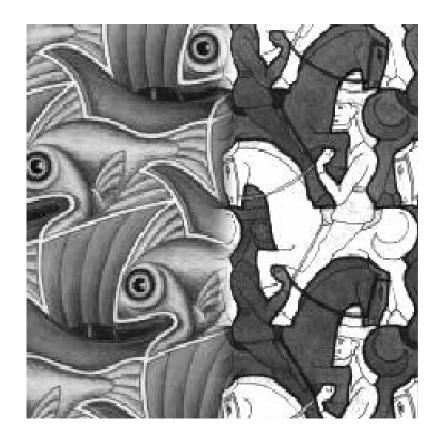








#### Good window size







- > 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)

• 
$$I_i = (G * I_{i-1}) \downarrow 2$$
 — Down-sample twice convolution

Get a series of smaller and blurry images.



## What does blurring take away?



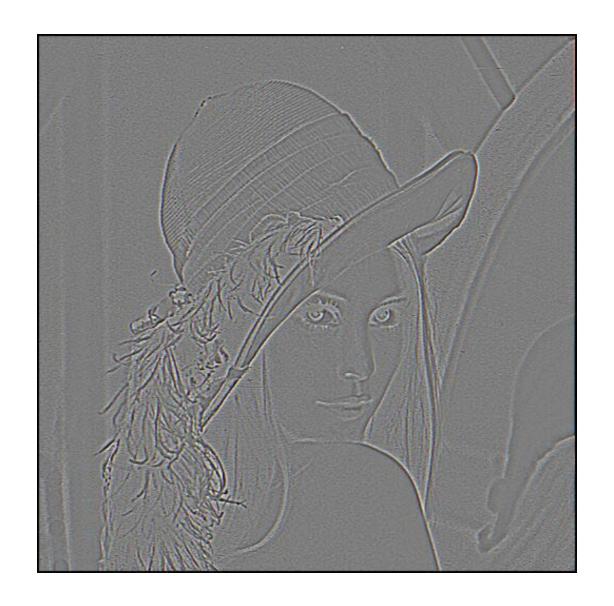


## What does blurring take away?





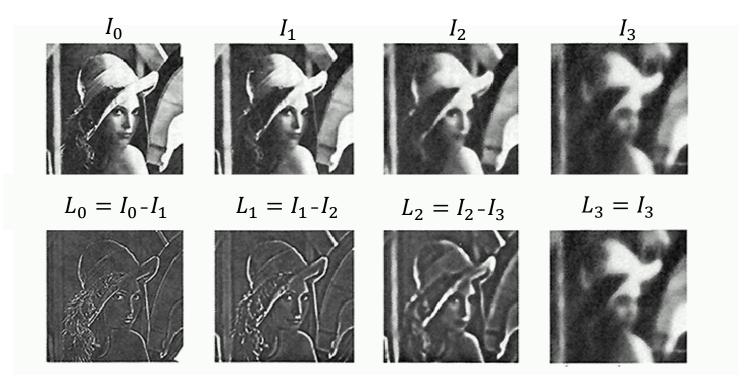
## What does blurring take away?





#### • Difference of Gaussian at each scale:

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

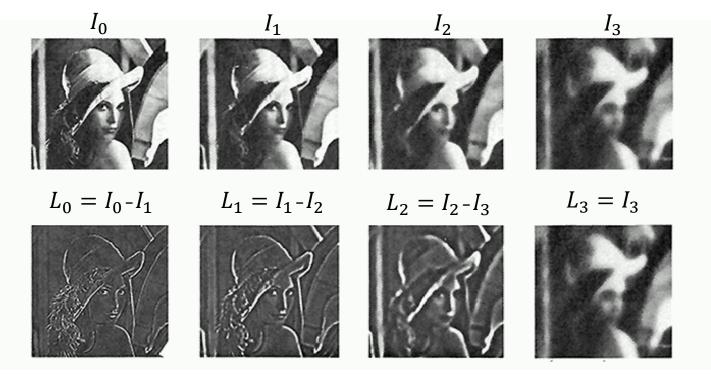


 $\{L_i\}$  = the set of  $L_i$  form. A Laplacian pyramid  $L_1$ ,  $L_2$ ,  $L_3$ ...,  $L_n$ 



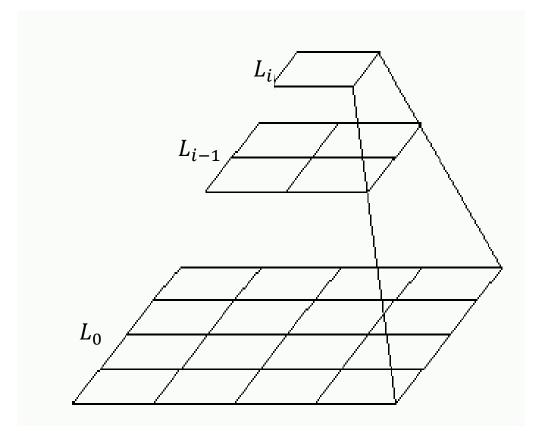
**➤** We can recover the original as:

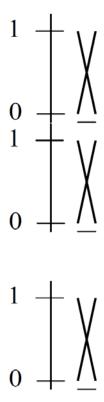
$$I = \sum_{i=0}^{N} (L_i) \uparrow$$

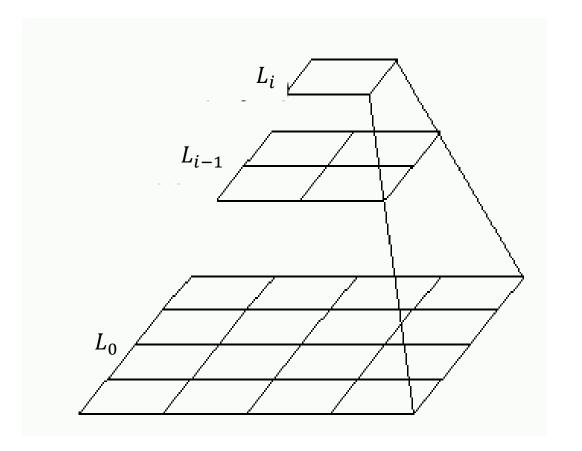


 $\{L_i\}$  = the set of  $L_i$  form. A Laplacian pyramid  $L_1$ ,  $L_2$ ,  $L_3$ ...,  $L_n$ 







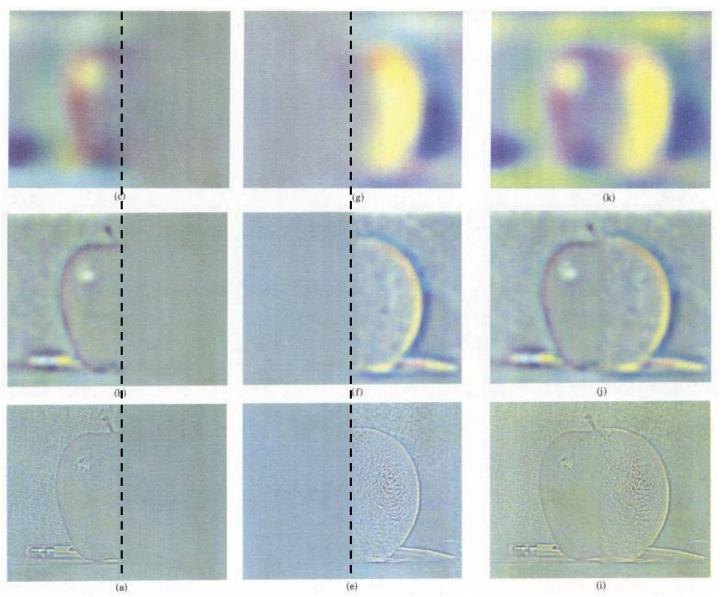


Left pyramid

blend

Right pyramid





Pyramid Blending





## Season Blending









## Season Blending









Target image



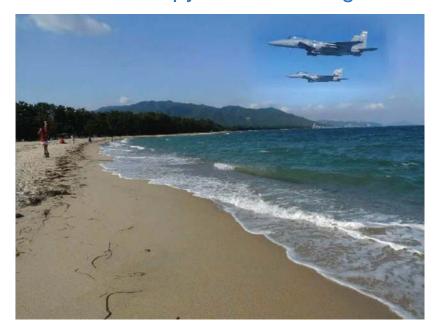
Source image



Target image with editing region



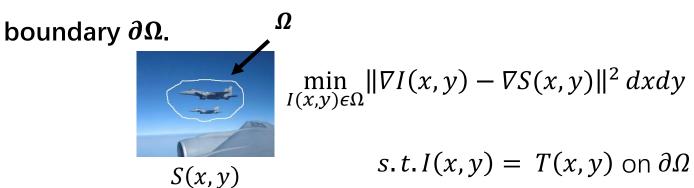
Result of pyramid blending

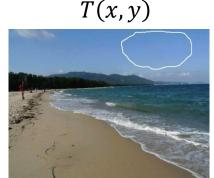


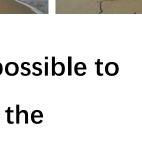


#### Poisson image editing

- > A even better idea: to reduce the color mismatch between source and target, create composite in gradient domain.
- $\succ$  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







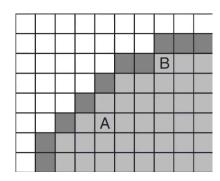
I(x,y)

 $\triangleright$  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  $\partial \Omega$ .

### Poisson image editing

> Solution for this Pb:

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



- Poisson equation
- Discretizing and solving the problem:
- 1) For a pixel A inside  $\Omega$ ,

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

$$\uparrow \qquad \uparrow$$

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

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

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



### Poisson image editing

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

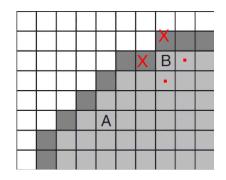
$$\nabla^{2}I(x,y) = \nabla^{2}S(x,y)$$

$$\uparrow \qquad \uparrow$$

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

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

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



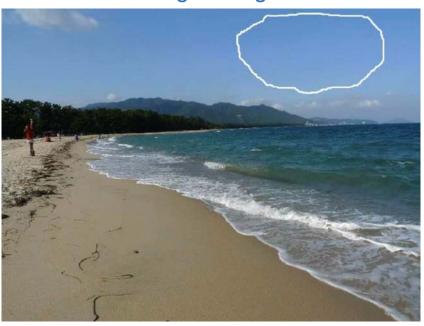
Big linear system



#### Source image



Target image



Poisson image editing result





## Take home message

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

