Lecture 15-1-Image Blending

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Course piazza link: piazza.com/shanghaitech.edu.cn/spring2021/cs270spring2021



Long history of fake images





Long history of fake images

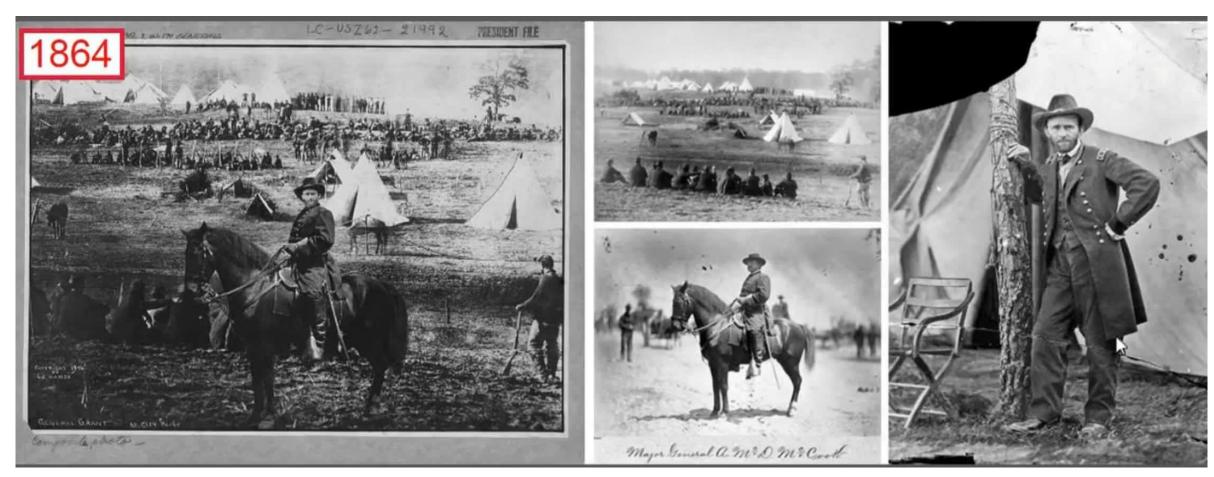






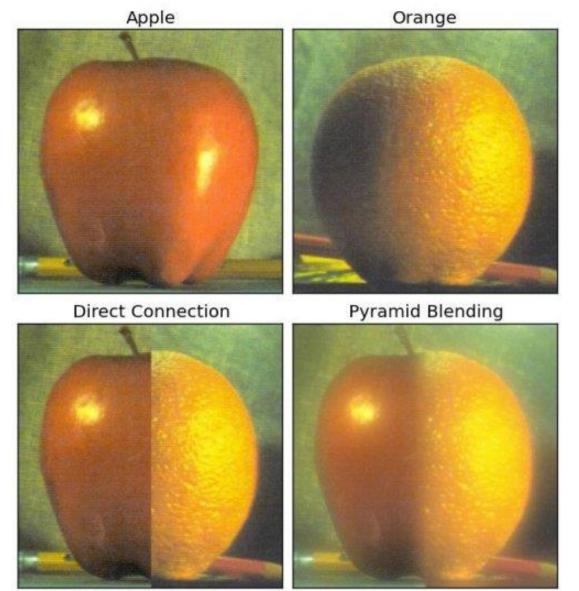


Long history of fake images





Hard edge composition vs Pyramid Blending





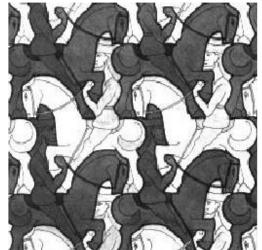
Hard compositing

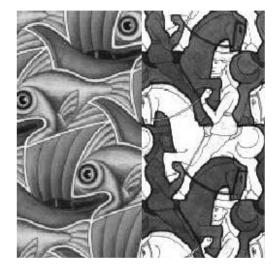
Hard compositing:

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

> Generally bad: seam/matte line is visible



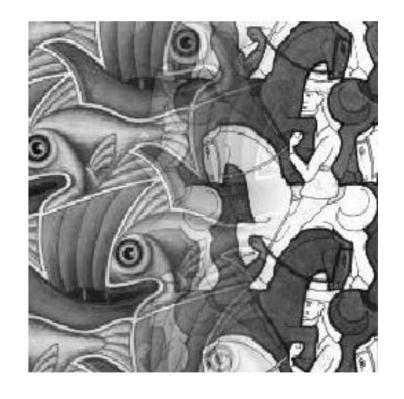


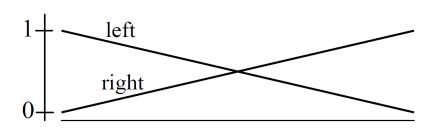


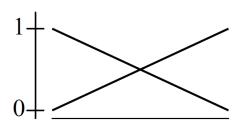


Weighted transition region:



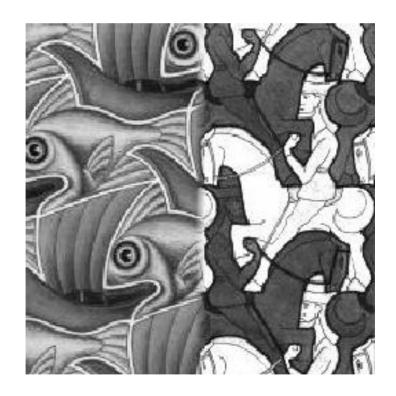








Weighted transition region:



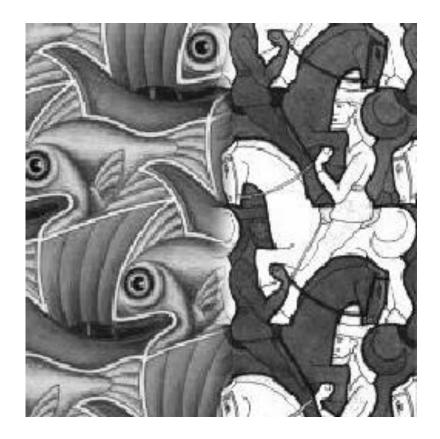








Good window size







- > 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?



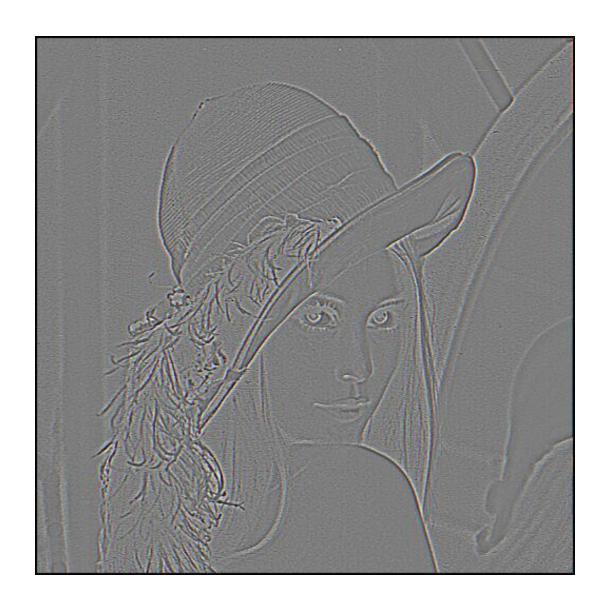


What does blurring take away?





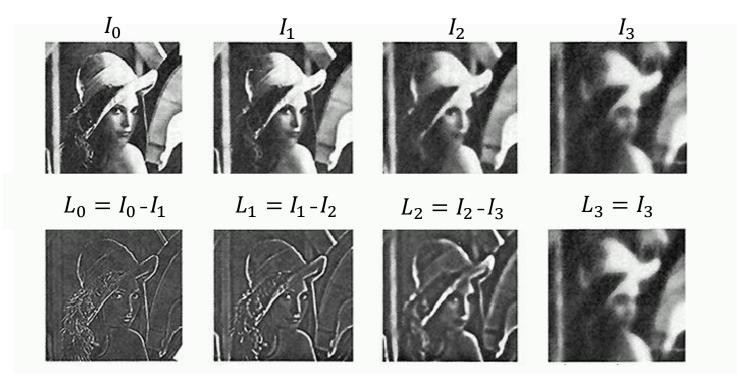
What does blurring take away?





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

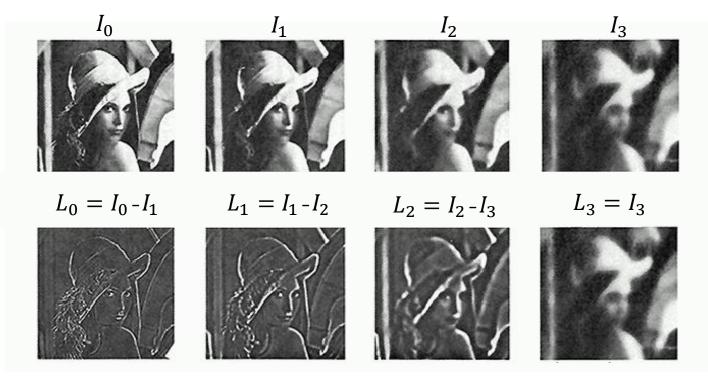






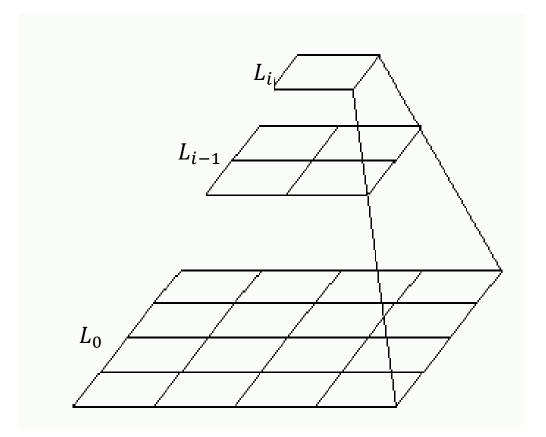
➤ We can recover the original as:

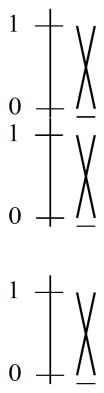
$$I = \sum_{i=0}^{N} (L_i) \uparrow$$

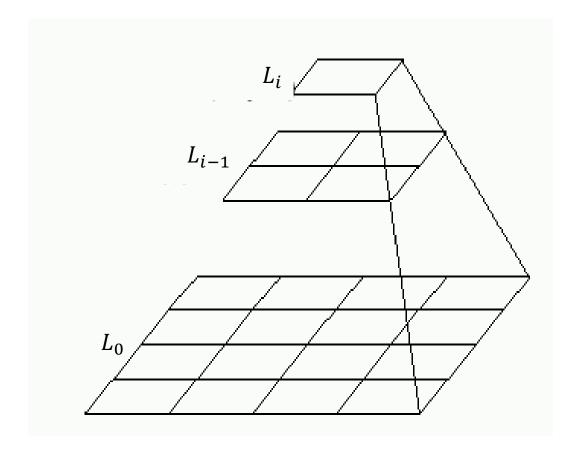


 $\{L_i\}$ = the set of L_i form. A Laplacian pyramid L_1 , L_2 , L_3 ..., L_n







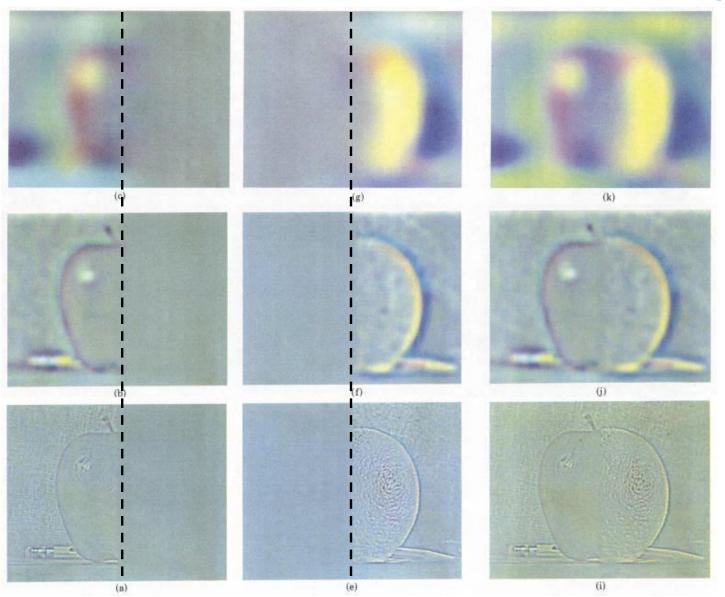


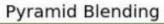
Left pyramid

blend

Right pyramid











Season Blending









Season Blending









Target image



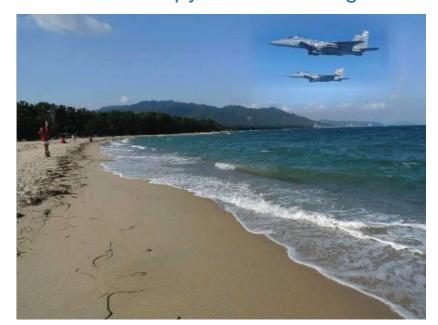
Source image



Target image with editing region



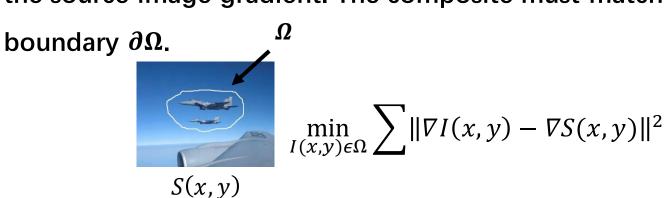
Result of pyramid blending





Poisson image editing

- ➤ A even better idea: to reduce the color mismatch between source and target, create composite in gradient domain.
- We want the gradient of the composite inside Ω to look as close as possible to the source image gradient. The composite must match target image on the T(x,y)







I(x,y)

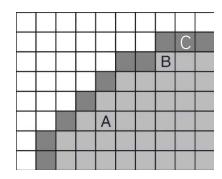
$$s.t.I(x,y) = T(x,y) \text{ on } \partial\Omega$$

ightharpoonup We want the gradient of the composite inside Ω 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 Ω ,

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

$$\uparrow \qquad \uparrow$$

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

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

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



Poisson image editing

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

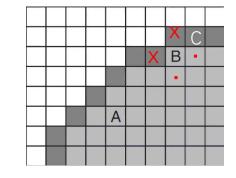
$$\nabla^{2}I(x,y) = \nabla^{2}S(x,y)$$

$$\uparrow \qquad \uparrow$$

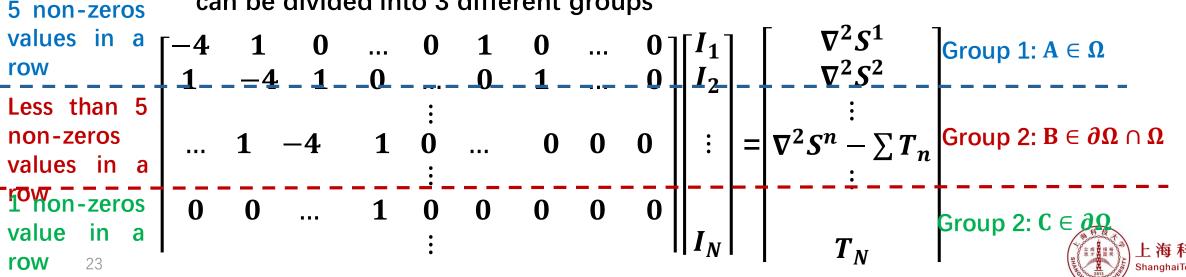
$$I(x+1,y)+I(x,y+1)+ (xx) \qquad S(x+1,y)+S(x,y+1)+$$

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

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



• Big linear system: so in all there will be N unknowns and N equations that can be divided into 3 different groups



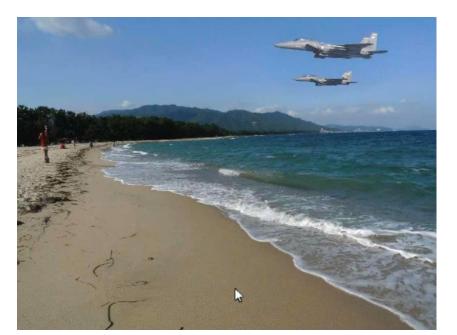
Source image



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.

