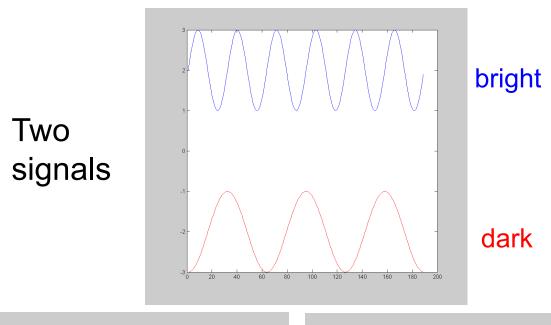
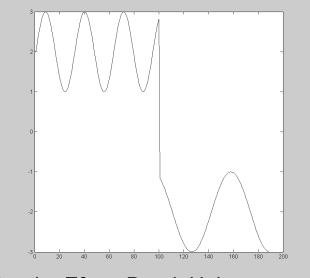
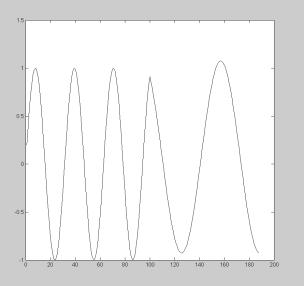
Gradient Domain blending (1D)





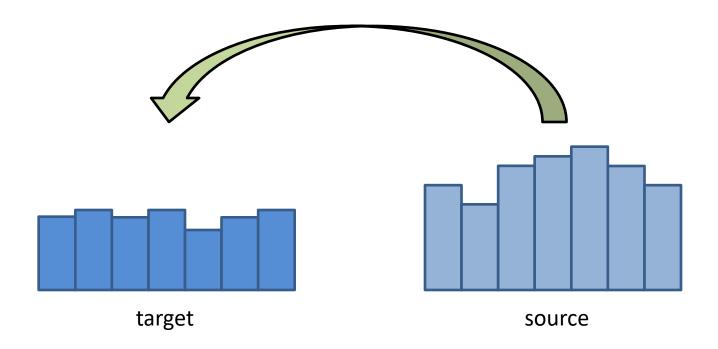


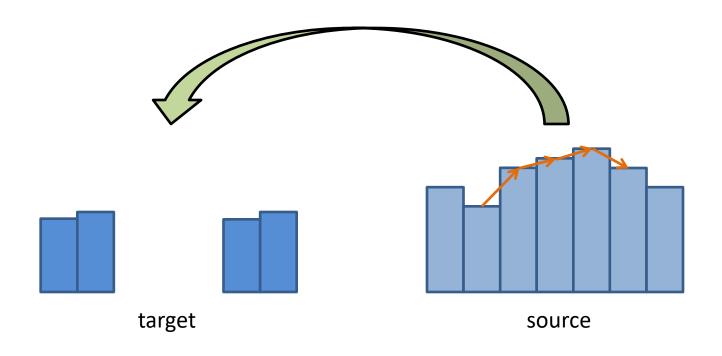


Blending derivatives

Slide credit: Alyosha Efros, Derek Hoiem

Gradient hole-filling (1D)

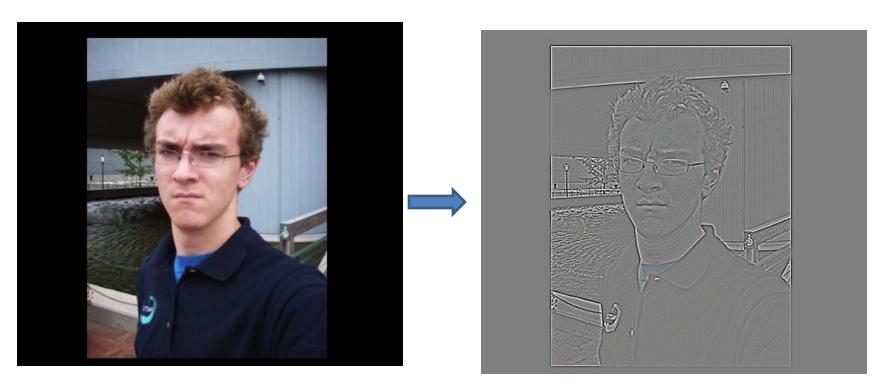




It is impossible to faithfully preserve the gradients

Slide credit: Alyosha Efros, Derek Hoiem

Example



Gradient Visualization

Source: Evan Wallace







Specify object region



Slide credit: Alyosha Efros, Derek Hoiem

Source: Evan Wallace

Poisson Blending Algorithm

A good blend should preserve gradients of source region without changing the background

Treat pixels as variables to be solved

- Minimize squared difference between gradients of foreground region and gradients of target region
- Keep background pixels constant

Target (background)

$$\mathbf{v} = \arg\min_{i \in S, j \in N_i \cap S} \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - v_j) - (s_i - s_j))^2 + \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - t_j) - (s_i - s_j))^2$$
Output
Source (foreground)

i current pixel's index

 N_i Current pixel's neighbors neighbor pixel index

S = S foreground/background mask

v: output pixels

s: source pixels

t: background (target) pixels

Slide credit: Alyosha Efros, Derek Hoiem

Perez et al. 2003

Examples

Gradient domain processing

 $\mathbf{v} = \arg\min_{i \in S, j \in N_i \cap S} \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - v_j) - (s_i - s_j))^2 + \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - t_j) - (s_i - s_j))^2$ Output Source (foreground)

source image

¹ 20	⁵ 20	⁹ 20	¹³ 20
² 20	⁶ 80	¹⁰ 20	¹⁴ 20
³ 20	⁷ 20	¹¹ 80	¹⁵ 20
⁴ 20	⁸ 20	¹² 20	¹⁶ 20

background image

¹ 10	⁵ 10	⁹ 10	¹³ 10
² 10	⁶ 10	¹⁰ 10	¹⁴ 10
³ 10	⁷ 10	¹¹ 10	¹⁵ 10
⁴ 10	⁸ 10	¹² 10	¹⁶ 10

target image

Target (background)

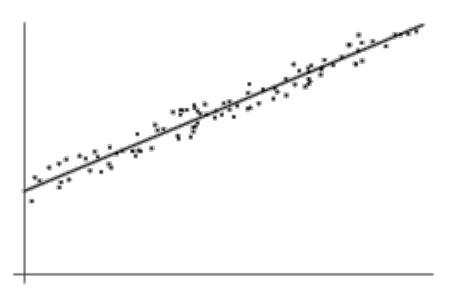
¹ 10	⁵ 10	⁹ 10	¹³ 10
² 10	6 $\mathbf{v_1}$	10 v ₃	¹⁴ 10
³ 10	7 \mathbf{v}_{2}		¹⁵ 10
⁴ 10	⁸ 10	¹² 10	¹⁶ 10

e.g., pixel
$$v_1$$
 left $(v_1 - 10) - (80 - 20)^2 + ((v_1 - 10) - (80 - 20))^2$

right bottom
$$((v_1 - v_3) - (80 - 20))^2 + ((v_1 - v_2) - (80 - 20))^2$$

Gradient-domain editing

Creation of image = least squares problem in terms of: 1) pixel intensities; 2) differences of pixel intensities



$$\hat{\mathbf{v}} = \underset{\mathbf{v}}{\operatorname{arg\,min}} \sum_{i} (\mathbf{a}_{i}^{T} \mathbf{v} - b_{i})^{2}$$

$$\hat{\mathbf{v}} = \underset{\mathbf{v}}{\operatorname{arg\,min}} (\mathbf{A} \mathbf{v} - \mathbf{b})^{2}$$

Use sparse linear equation solver in Python and MATLAB

Examples

Gradient domain processing

 $\mathbf{v} = \arg\min_{i \in S, j \in N_i \cap S} \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - v_j) - (s_i - s_j))^2 + \sum_{i \in S, j \in N_i \cap \neg S} ((v_i - t_j) - (s_i - s_j))^2$ Output Source (foreground)

source image

¹ 20	⁵ 20	⁹ 20	¹³ 20
² 20	⁶ 80	¹⁰ 20	¹⁴ 20
³ 20	⁷ 20		¹⁵ 20
⁴ 20	⁸ 20	¹² 20	¹⁶ 20

background image

	0		0
¹ 10	⁵ 10	⁹ 10	¹³ 10
² 10	⁶ 10	¹⁰ 10	¹⁴ 10
³ 10	⁷ 10	'	¹⁵ 10
⁴ 10	⁸ 10	¹² 10	¹⁶ 10

target image

Target (background)

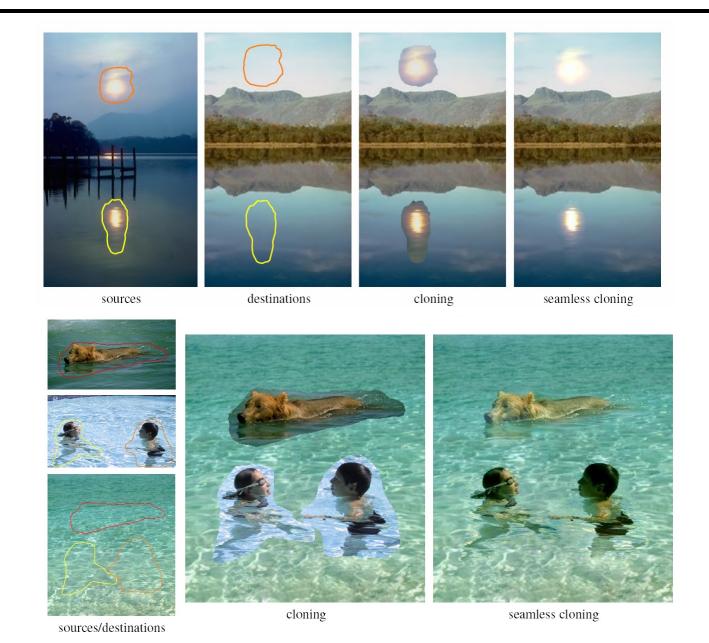
¹ 10	⁵ 10	⁹ 10	¹³ 10
² 10	6 $\mathbf{v_1}$	¹⁰ v ₃	¹⁴ 10
³ 10	7 \mathbf{v}_{2}	$\mathbf{v_4}$	¹⁵ 10
⁴ 10	⁸ 10	¹² 10	¹⁶ 10

e.g., pixel
$$v_1 = ((v_1 - 10) - (80 - 20))^2 + ((v_1 - 10) - (80 - 20))^2$$

Least squares:
$$((v_1 - v_3) - (80 - 20))^2 + ((v_1 - v_2) - (80 - 20))^2$$

Linear equation: $4v_1 - 10 - 10 - v_3 - v_2 = (80 - 20) \times 4$

Perez et al., 2003









target source mask





no blending

gradient domain blending

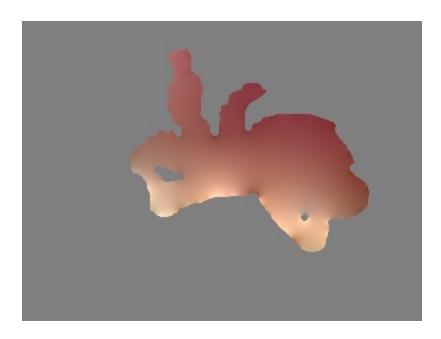
What's the difference?



gradient domain blending



no blending



Perez et al, 2003











Local color changes

Limitations:

- Can't do contrast reversal (gray on black -> gray on white)
- Colored backgrounds "bleed through"
- Images need to be very well aligned

Drawing in Gradient Domain

Real-Time Gradient-Domain Painting

James McCann* Carnegie Mellon University Nancy S. Pollard[†] Carnegie Mellon University



James McCann & Nancy Pollard **Real-Time Gradient-Domain Painting**,

SIGGRAPH 2009

(CMU paper)

Drawing in Gradient Domain



James McCann & Nancy Pollard

Real-Time Gradient-Domain Painting,

SIGGRAPH 2009