

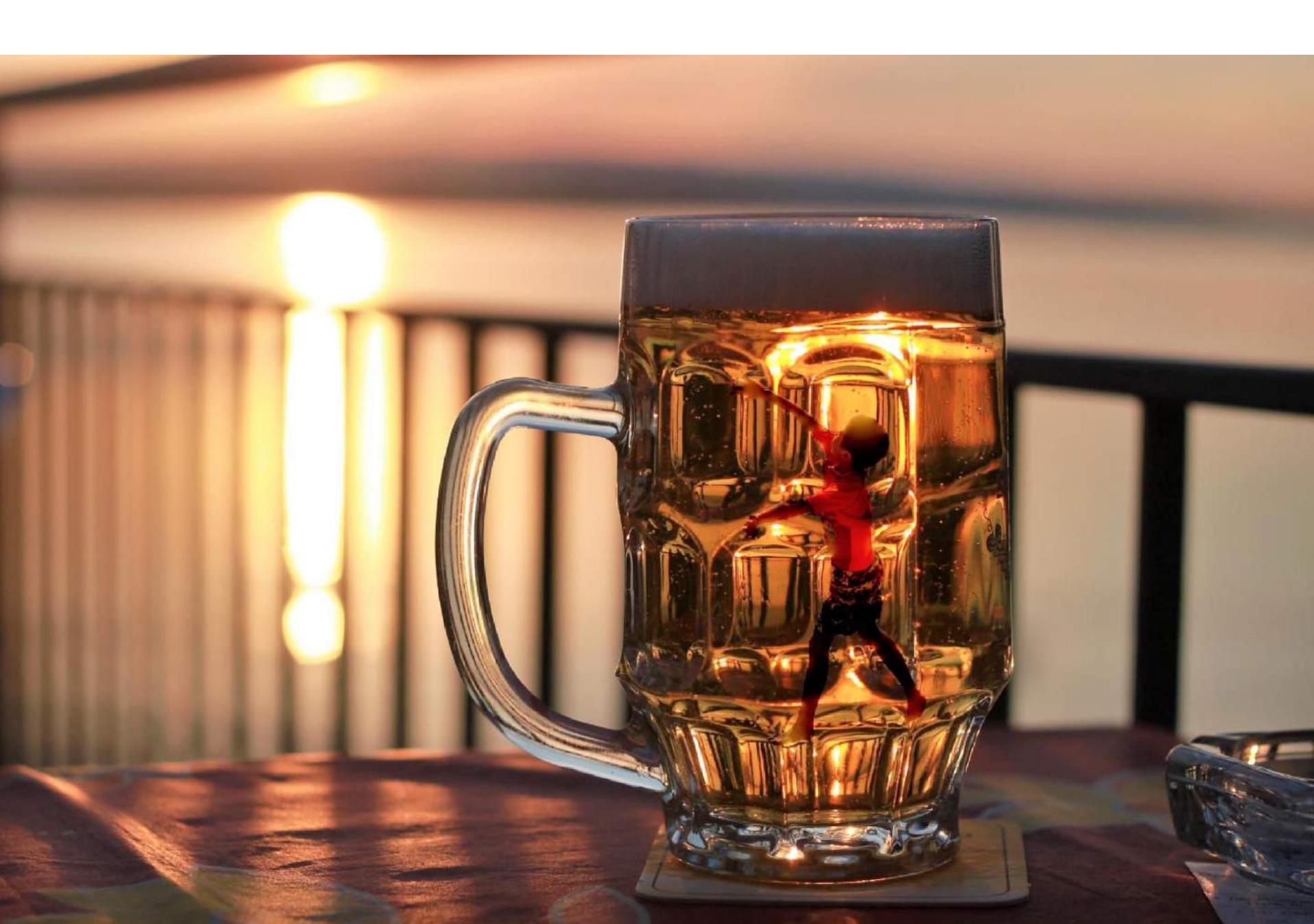


Video 8.1

Jianbo Shi

Gradient Blending





How to blend two images?

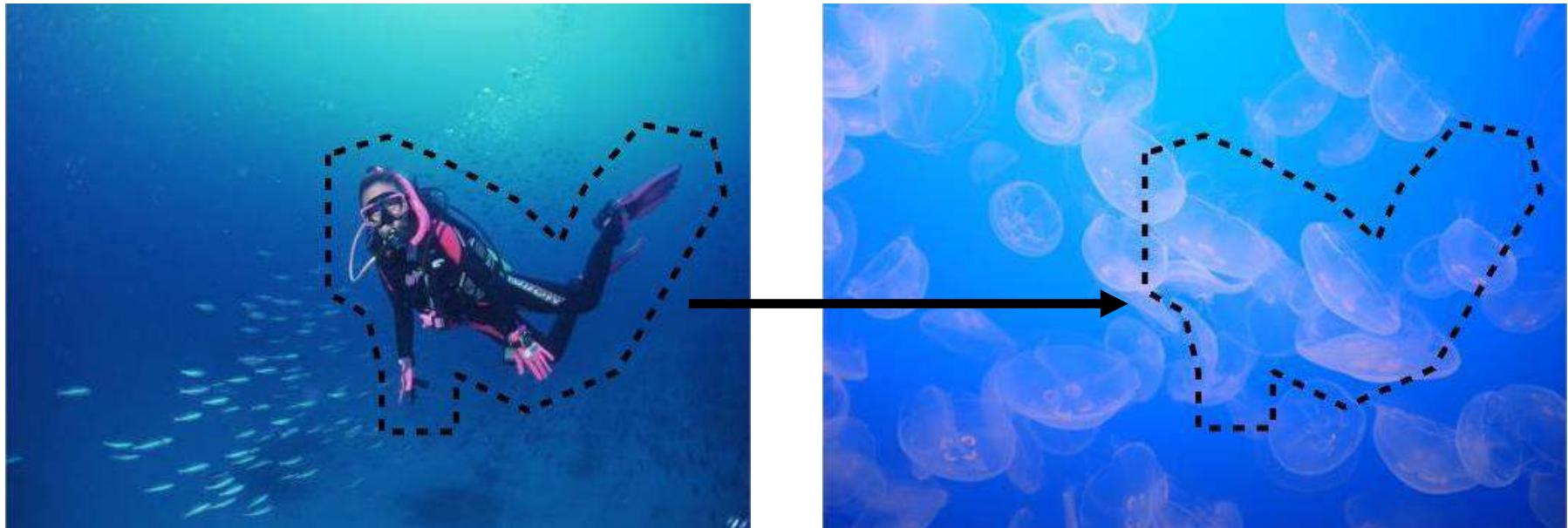


image blending: image surgery...

- cutting from one image (which we will cover in details on segmentation)
- reconstructing onto the new image

Blend = Cut and Paste images

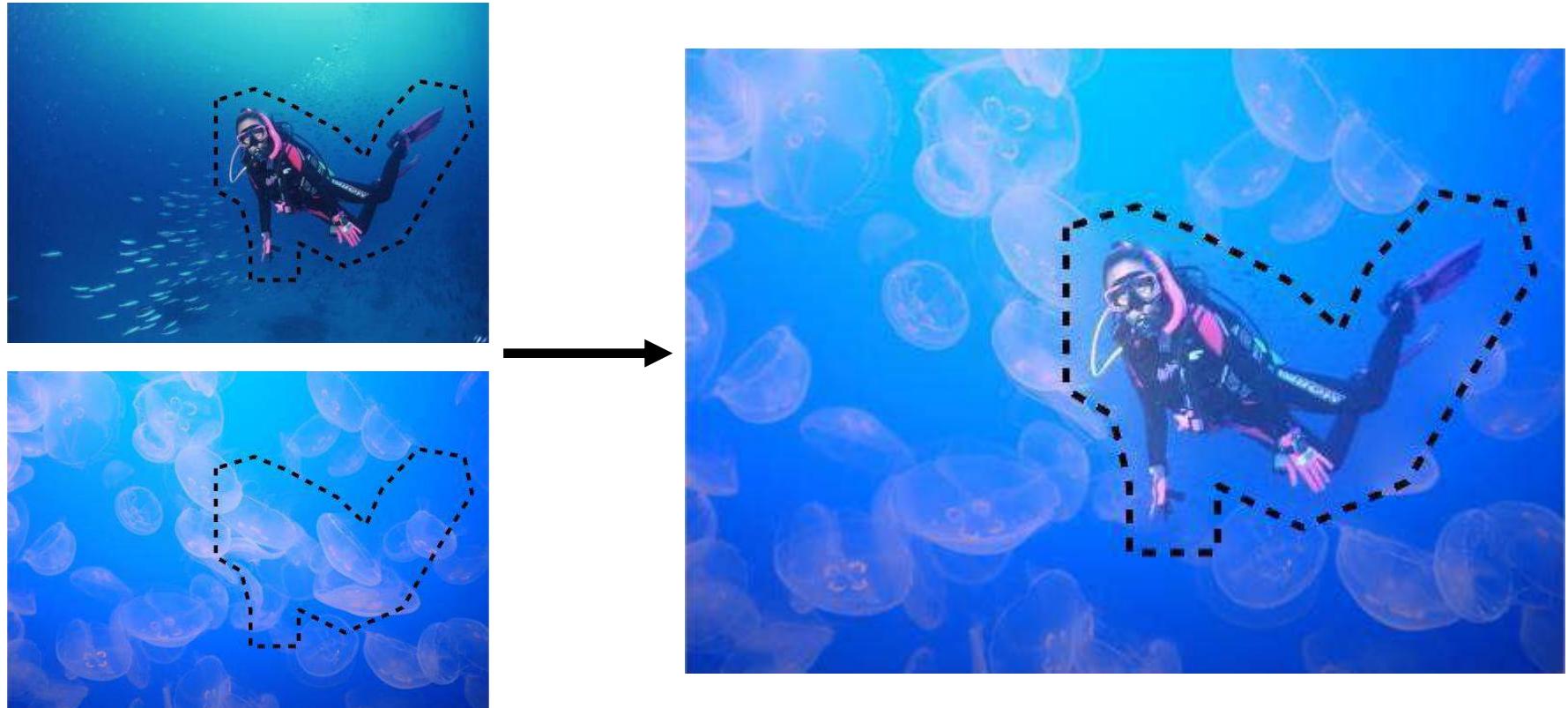
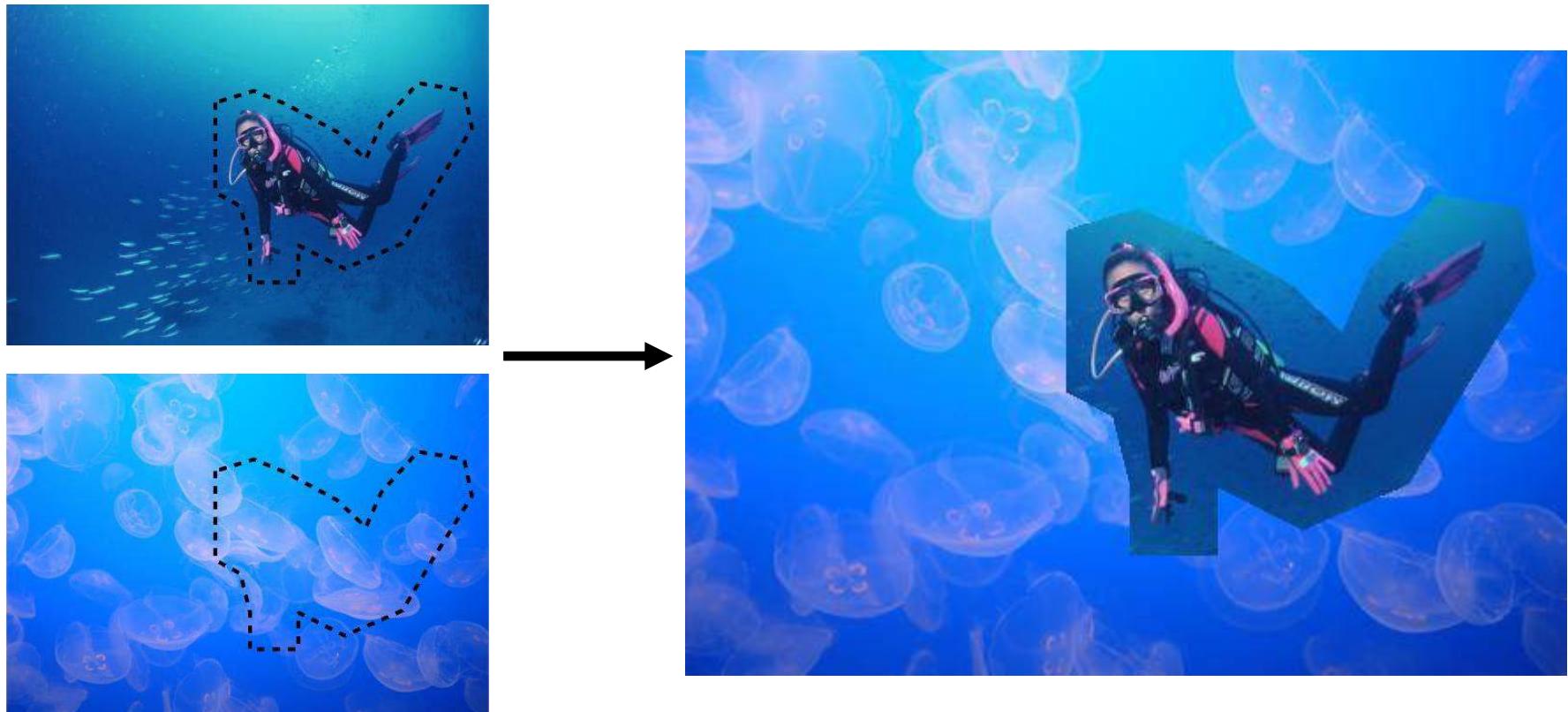


Image blending is an art of ...

faking images, hiding evidence of image surgery, making it looks natural

Direct Copy and Paste

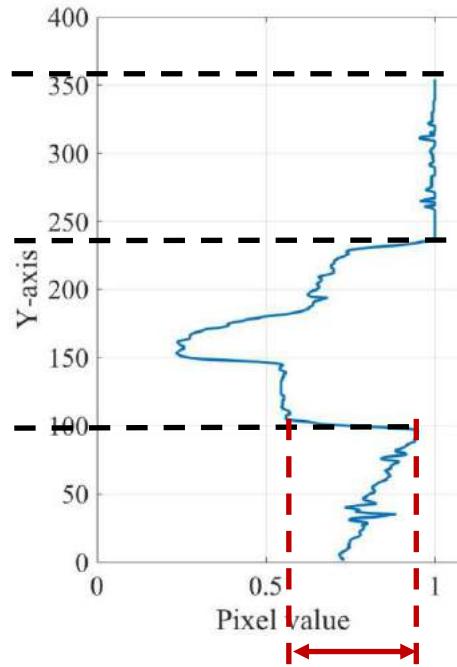


Direct attempt: not so good!

-- we created an artificial boundary between the pasted region

Challenge: color and brightness mismatch

Blue channel value of the vertical line



Big change in intensity

Big change in intensity creates new image boundary...
- any ideas on how to remove that boundary?

Alpha blending

Alpha channel encodes the transparency of the object



Alpha blending



×



=



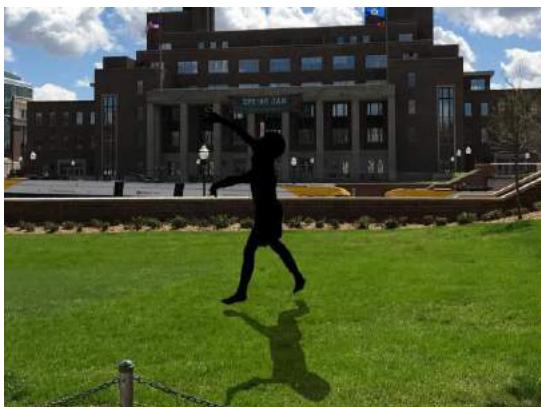
×



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Alpha blending



Alpha blending

Copy - paste is a special kind of alpha blending – binary mask



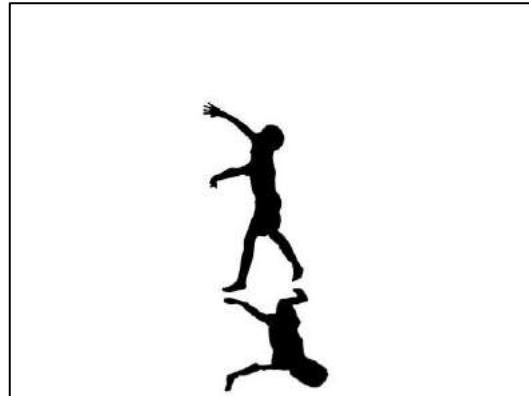
×



=

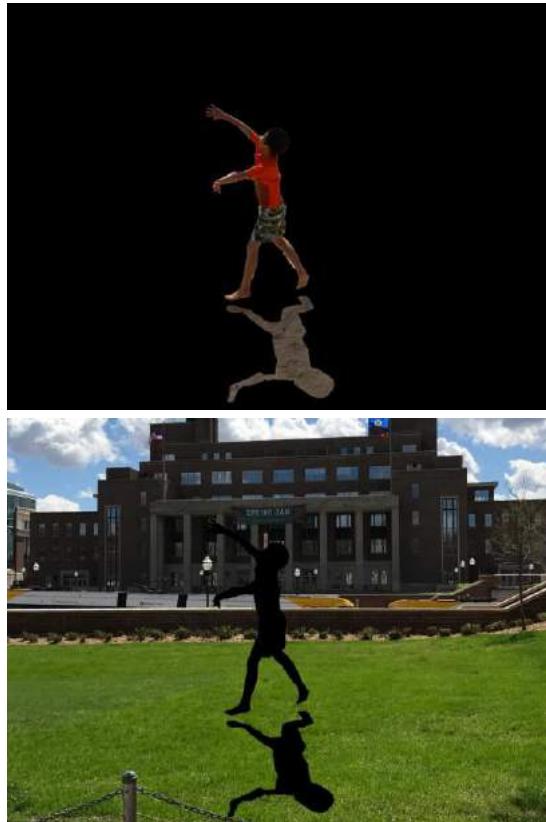


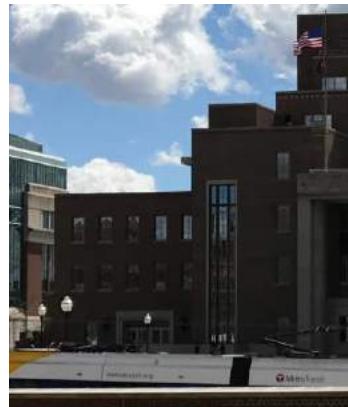
×



=









copy - paste



alpha blending

Alpha blending can deal with transparent objects

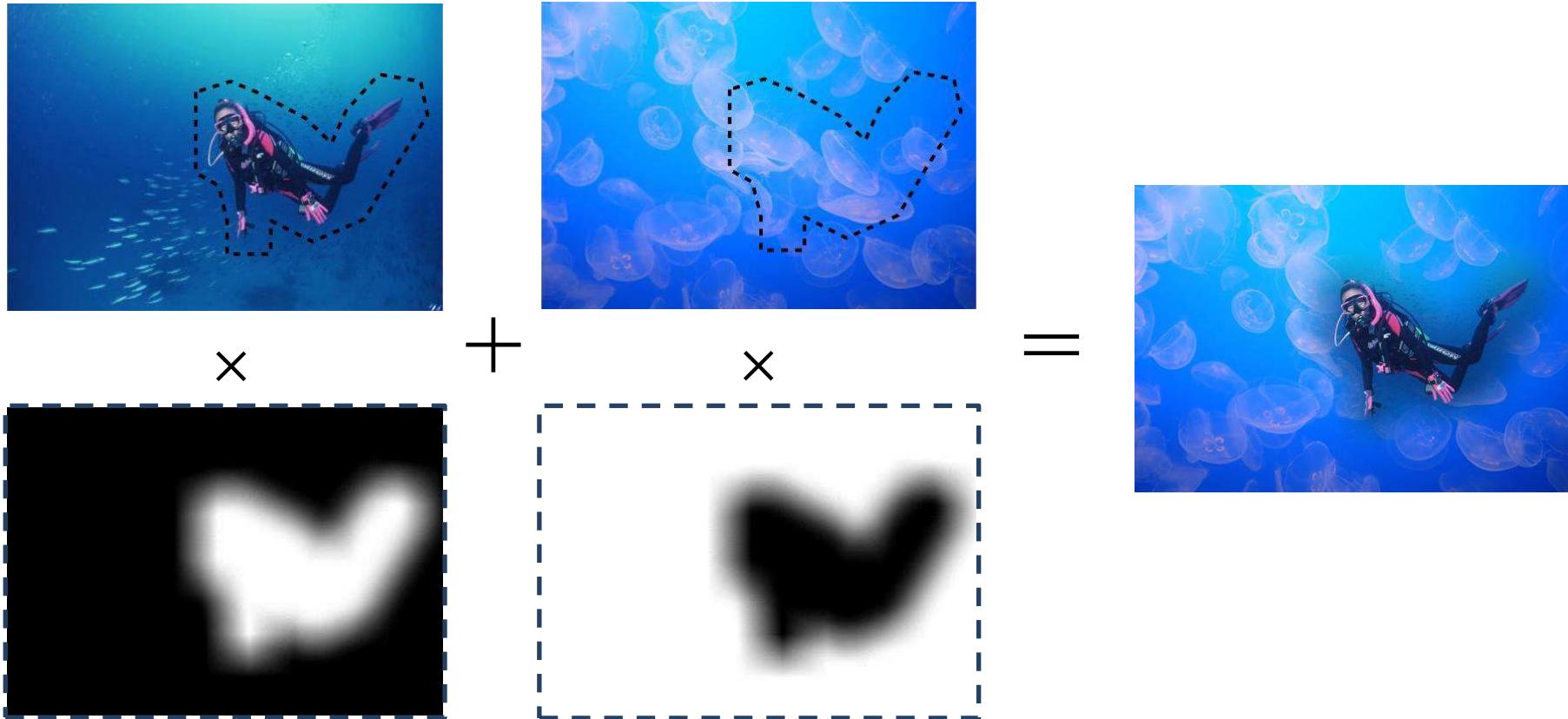


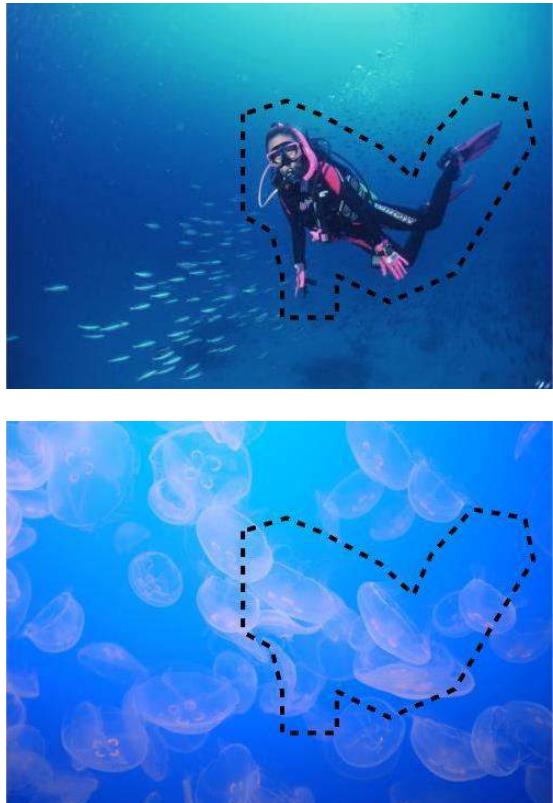
copy - paste



alpha blending

Alpha blending hacking





copy - paste



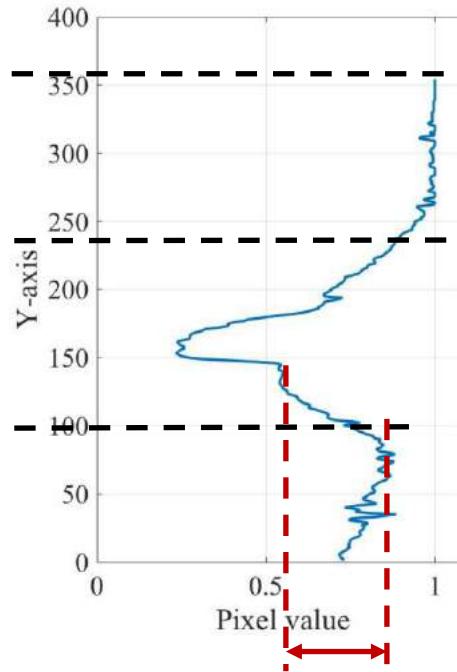
Alpha blending



Alpha Blending

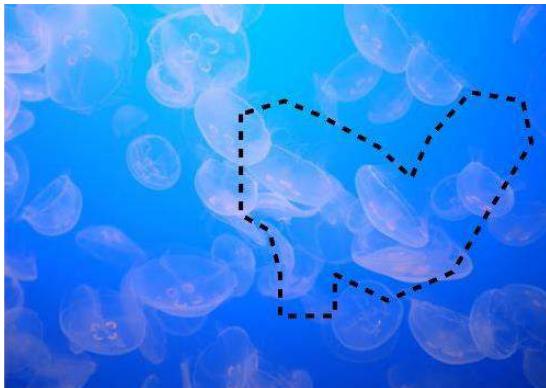
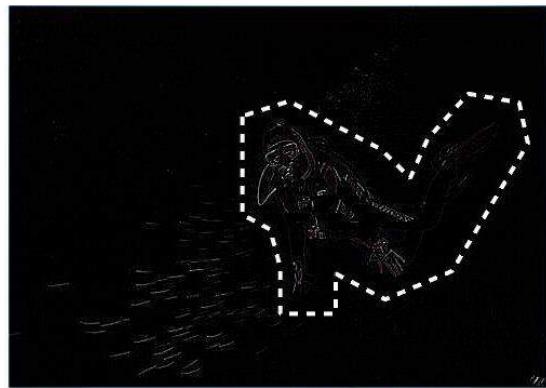


Blue channel value of the vertical line



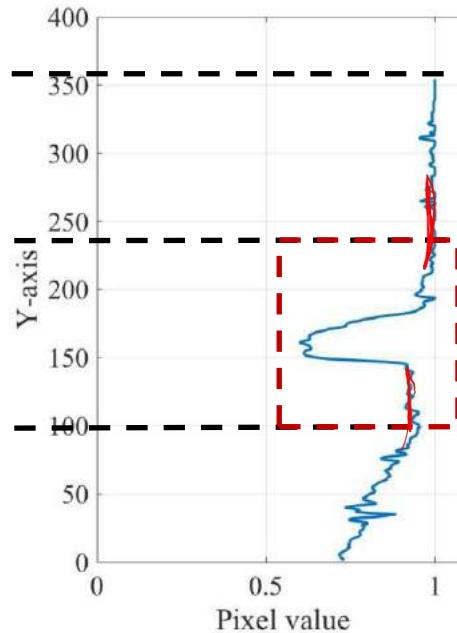
Continuous change,
but still a visually un-natural pattern in intensity

Solution: Copy only gradient + Recreate



Gradient Blending: No more intensity change!

Blue channel value of the vertical line



Smooth transition

Image Blending



copy - paste



alpha blending



Gradient blending

Results of gradient blending

Source (figure) Target (background) Result



Color of the hat blends into the background
-- successful image surgery... with interesting side-affect

Gradient Blending



Gradient Blending



Gradient Blending

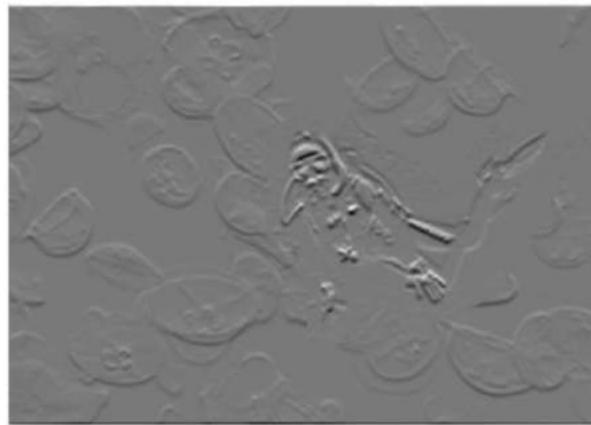
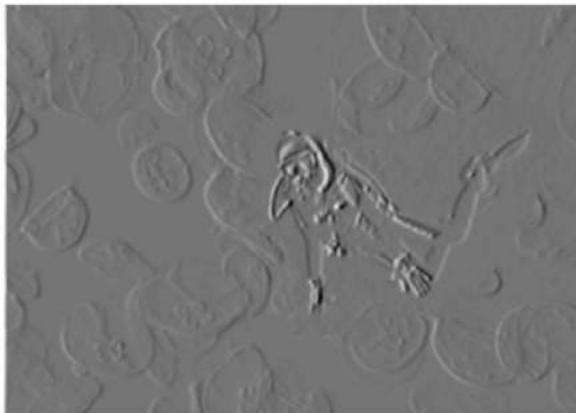




Video 8.2

Jianbo Shi

Why use the gradients to recreate images



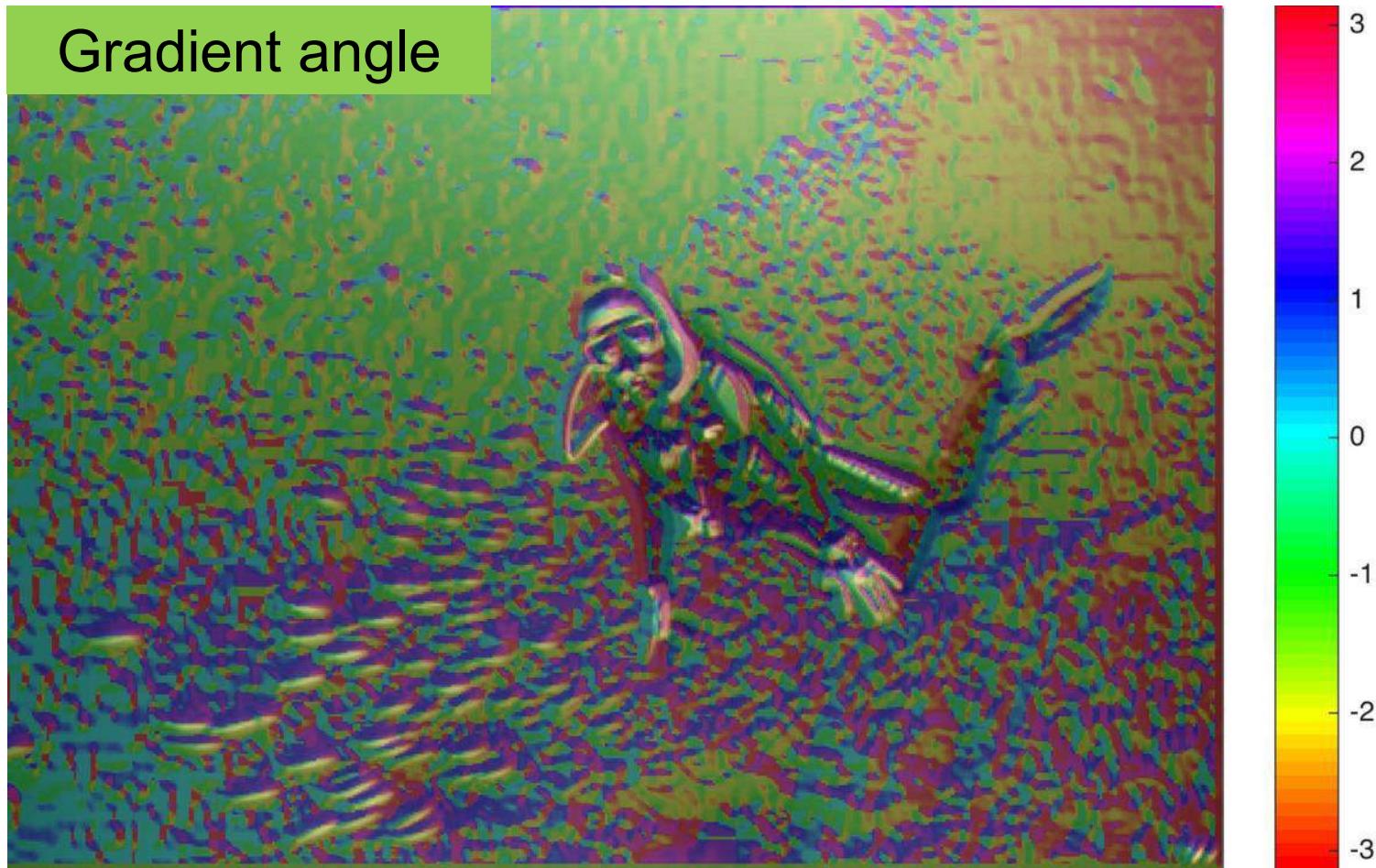
Source Image



Gradient magnitude

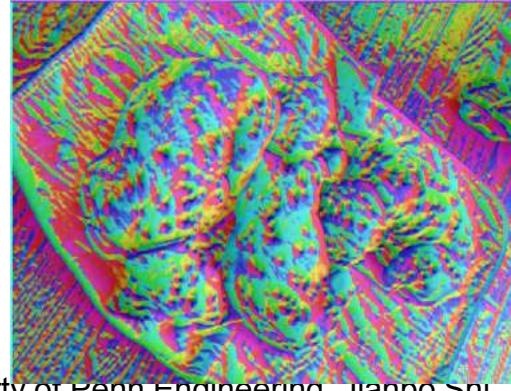


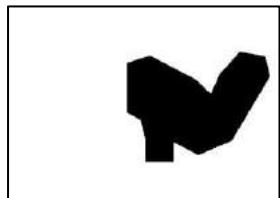
Gradient angle



Gradients domain

- Gradient captures everything important about shape and shading
- It contains the microscope texture of the object.
- It encodes subtle changes of illumination.
- In Pyramid Blending, we decomposed our image into 2nd derivatives (Laplacian) which encodes the shape perception.



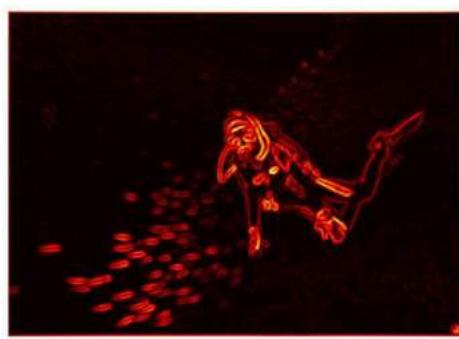
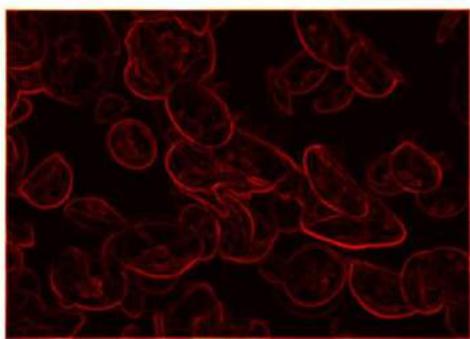


×

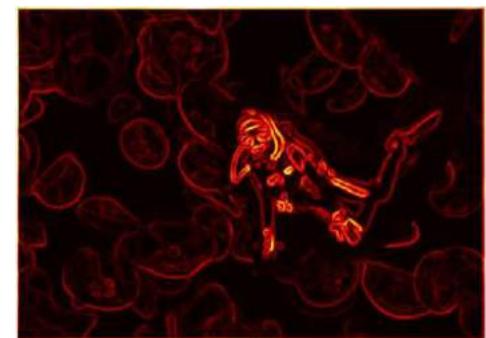
+

×

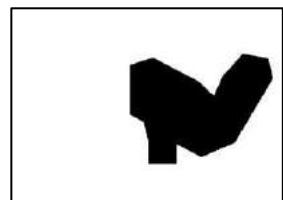
Blending Magnitude



=

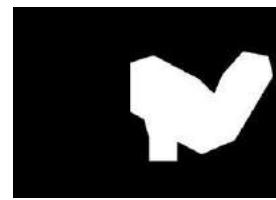


- We blend the gradient magnitude, to create a seamless ‘edge’ image



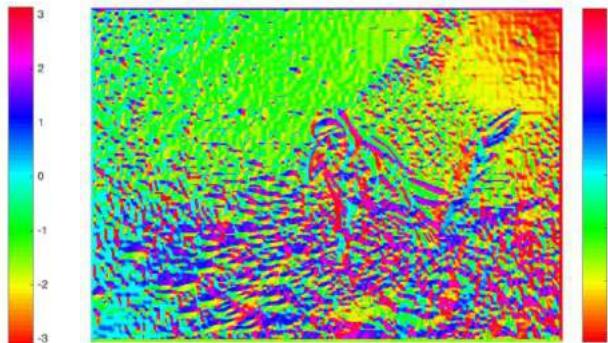
×

+

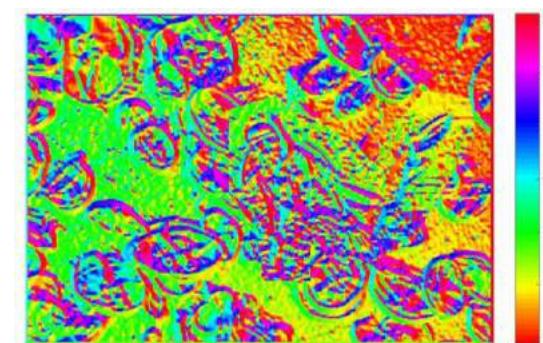


×

Blending the
Gradient angle



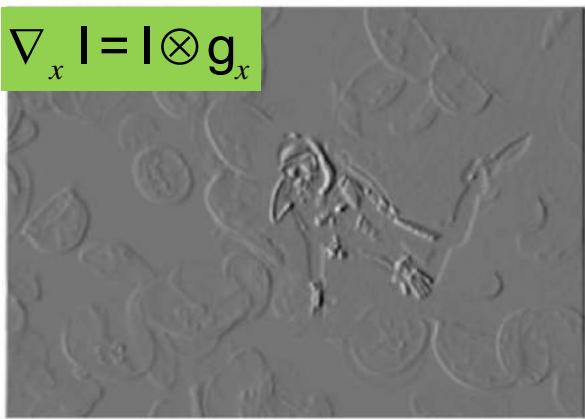
=



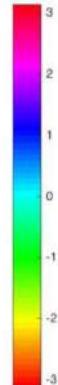
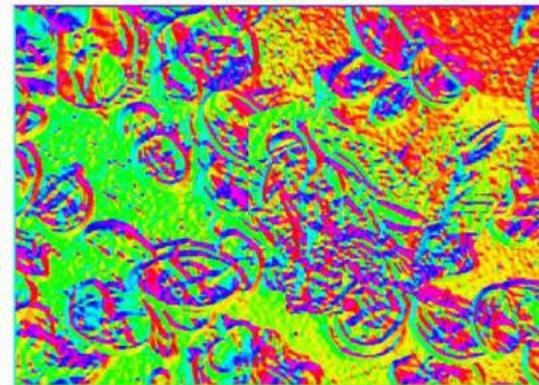
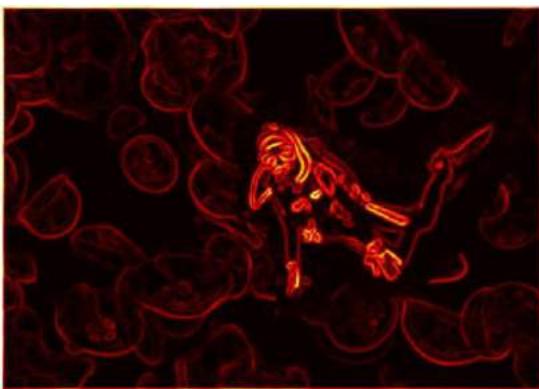
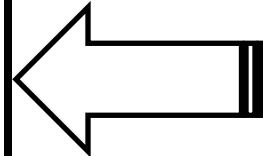
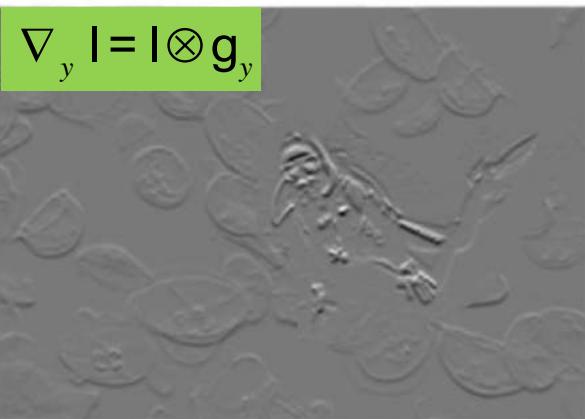
- We blend the gradient angle images, to make sure both the microscopic texture, shape of object boundary, and the illumination changes are smoothly integrated.

Image gradients blending

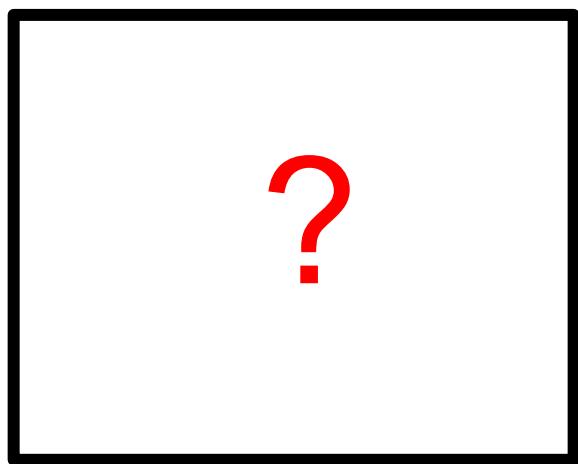
$$\nabla_x I = I \otimes g_x$$



$$\nabla_y I = I \otimes g_y$$



How to recreate the original image from gradient?



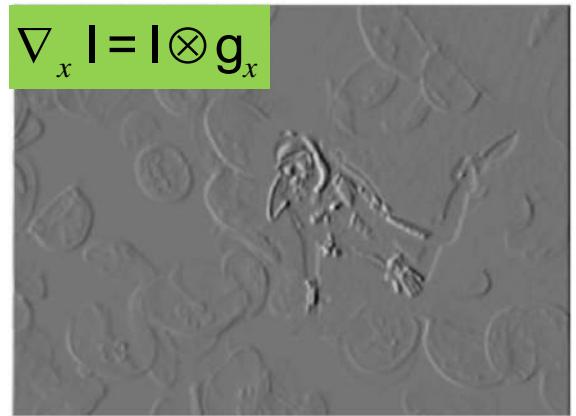
$$\begin{array}{|c|c|} \hline 1 & -1 \\ \hline \end{array}$$



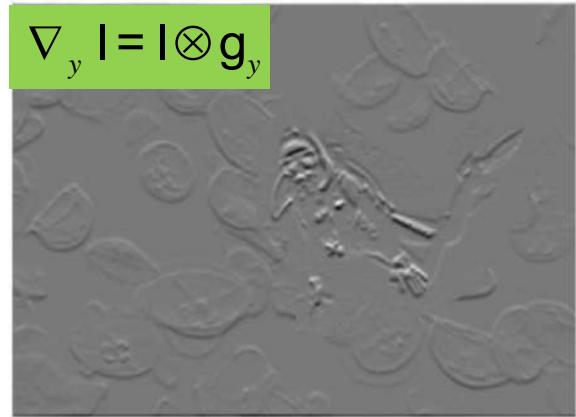
$$\begin{array}{|c|c|} \hline 1 \\ \hline -1 \\ \hline \end{array}$$

=

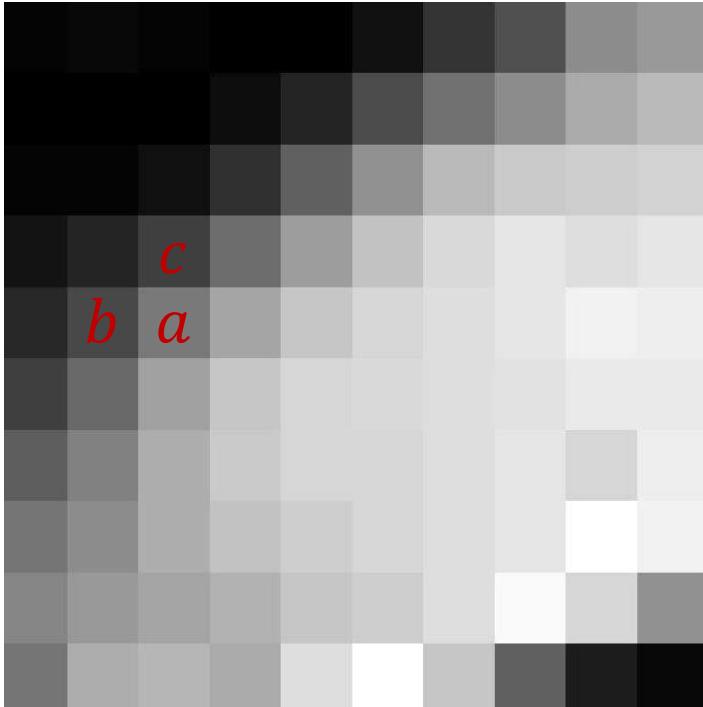
$$\nabla_x I = I \otimes g_x$$



$$\nabla_y I = I \otimes g_y$$



zoom in a small patch



g

$$a - b = \text{gray}$$

$$a - c = \text{gray}$$

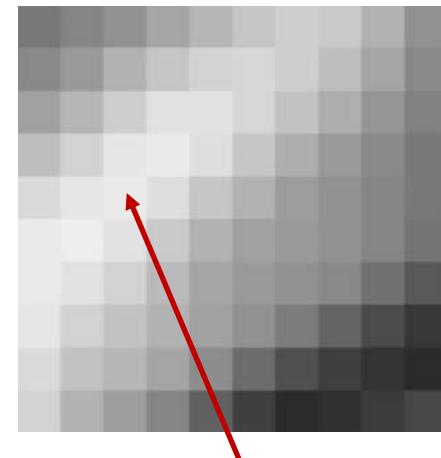
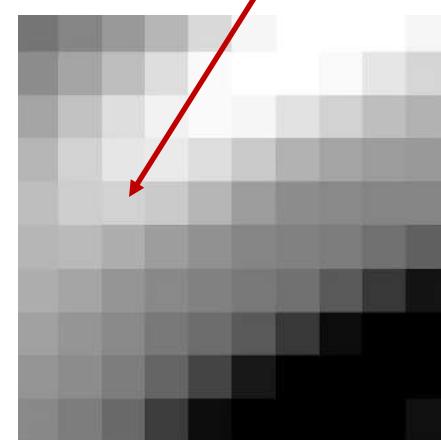


$$\begin{array}{|c|c|} \hline 1 & -1 \\ \hline \end{array}$$



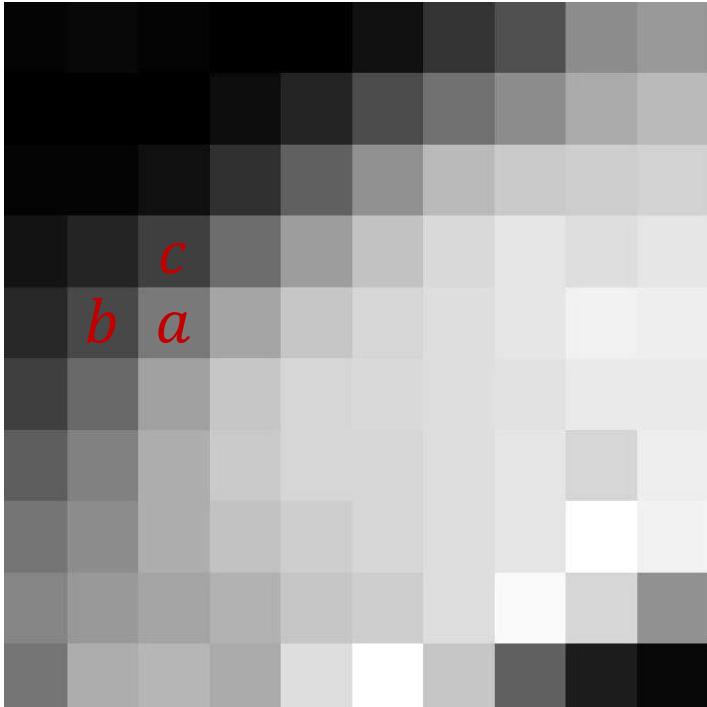
$$\begin{array}{|c|} \hline 1 \\ \hline -1 \\ \hline \end{array}$$

$$\nabla_x I(a)$$



$$\nabla_y I(a)$$

2 equations with 3 unknowns,
need constraints



g

$$\begin{matrix} a & - & b \end{matrix}$$

$$=$$



$\nabla_x I(a)$

$$\begin{matrix} a & - & c \end{matrix}$$

$$=$$



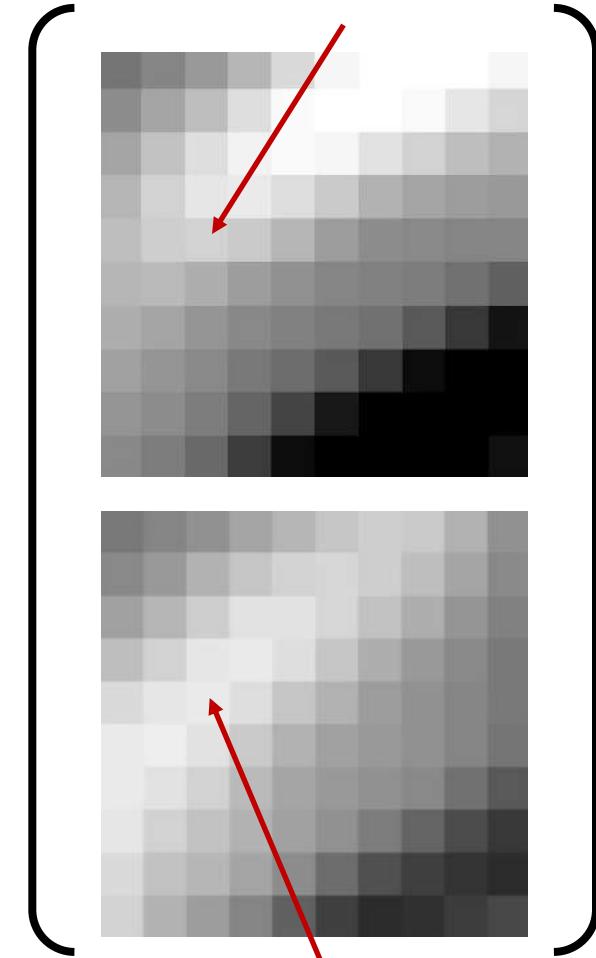
$\nabla_y I(a)$



$$\begin{matrix} 1 & -1 \end{matrix}$$

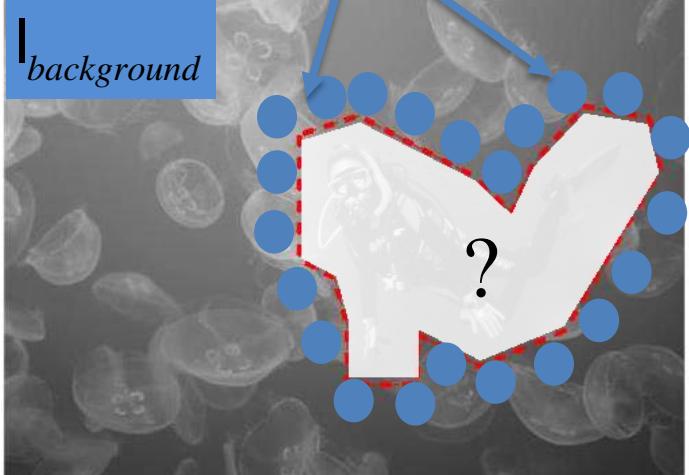


$$\begin{matrix} 1 \\ -1 \end{matrix}$$

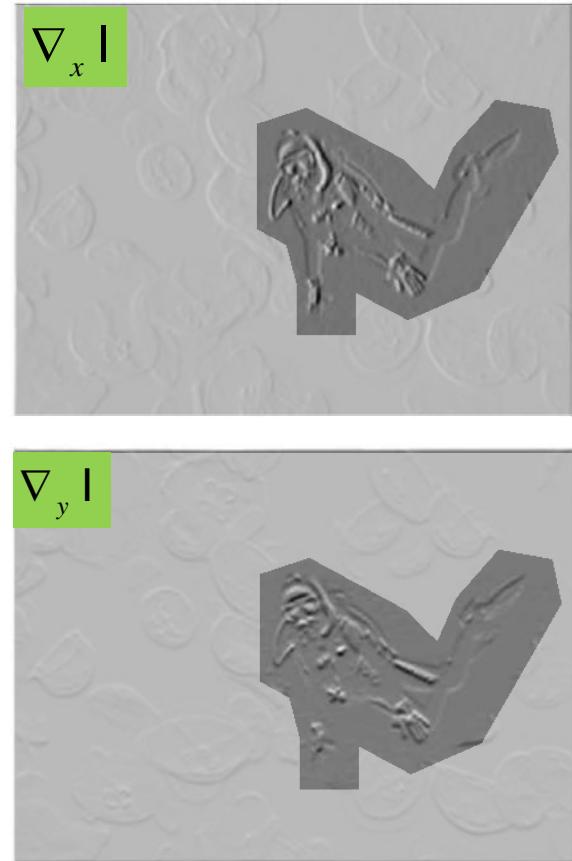


$\nabla_y I(a)$

Boundary condition

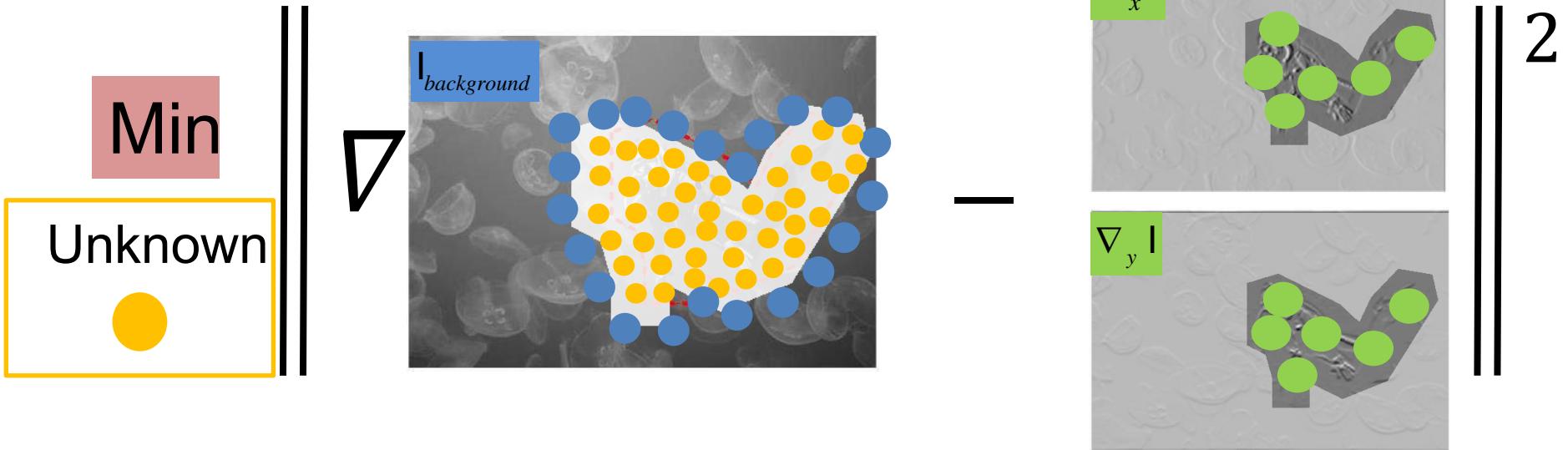


$$\otimes \begin{array}{|c|c|} \hline 1 & -1 \\ \hline \end{array} = \otimes \begin{array}{|c|c|} \hline 1 \\ \hline -1 \\ \hline \end{array}$$

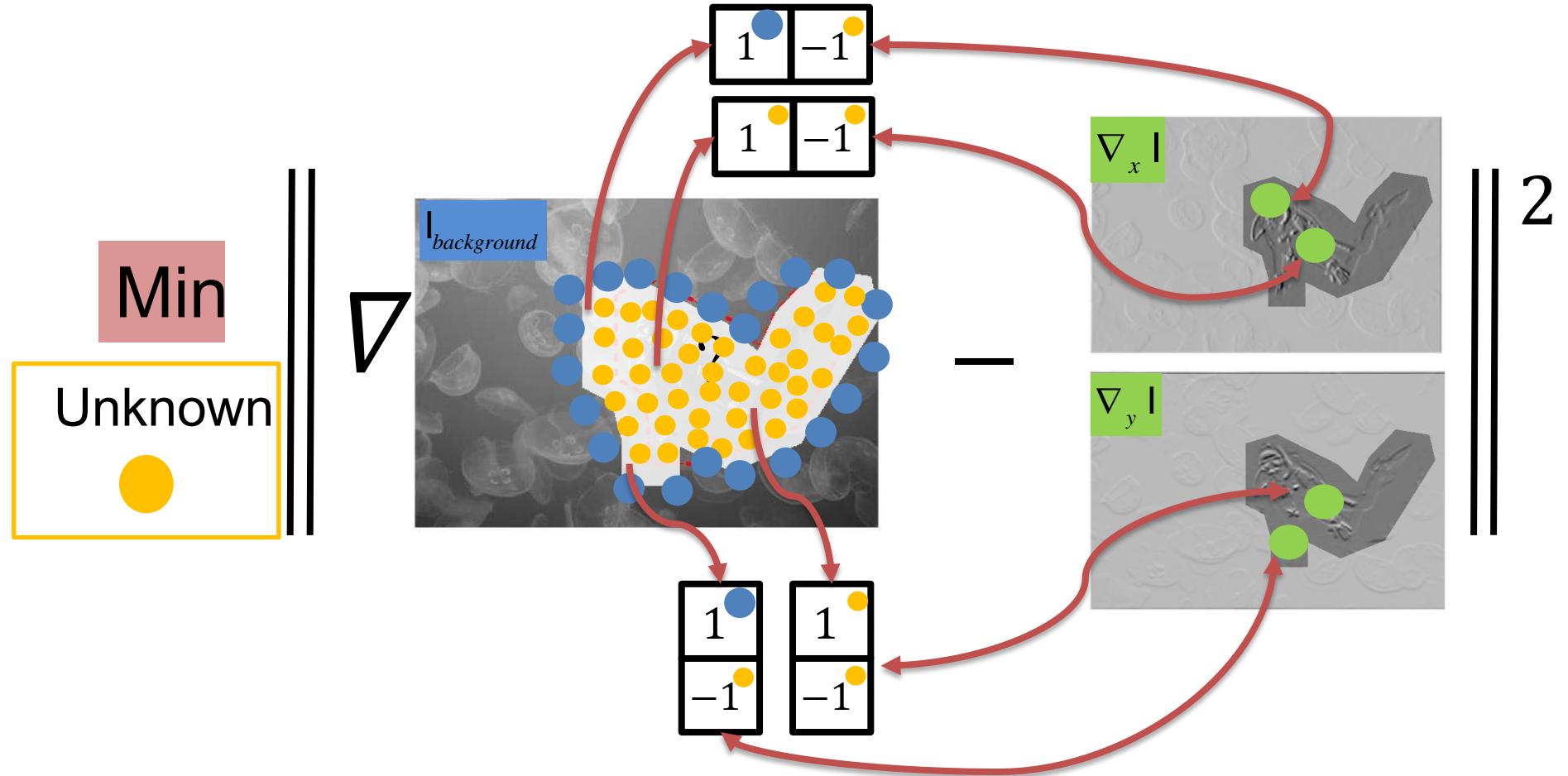


- Keep the value on boundary $\partial\Omega$ the same

Least Square Problem



- Minimize the loss with respect to all pixels in the region Ω
- Keep the boundary $\partial\Omega$ the same

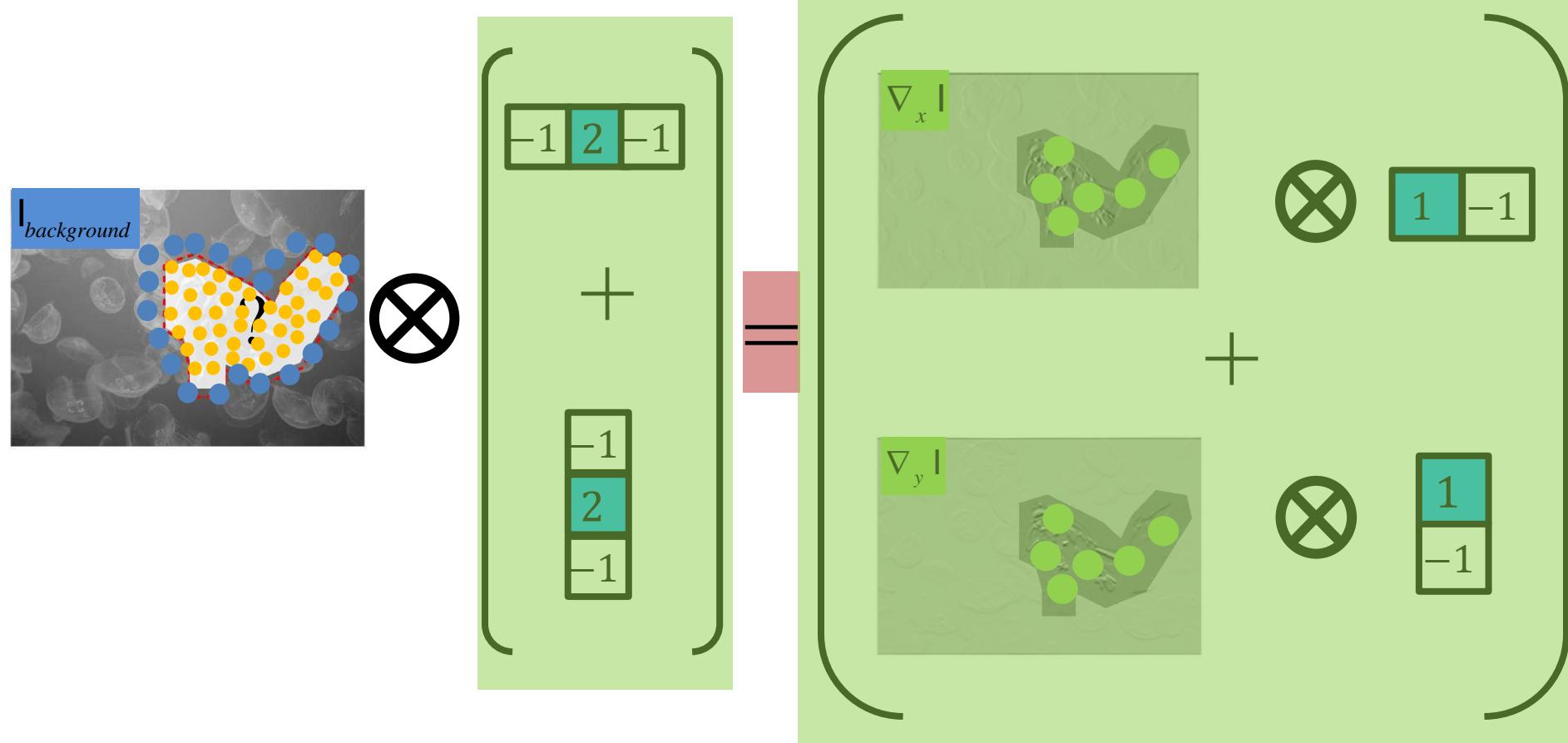


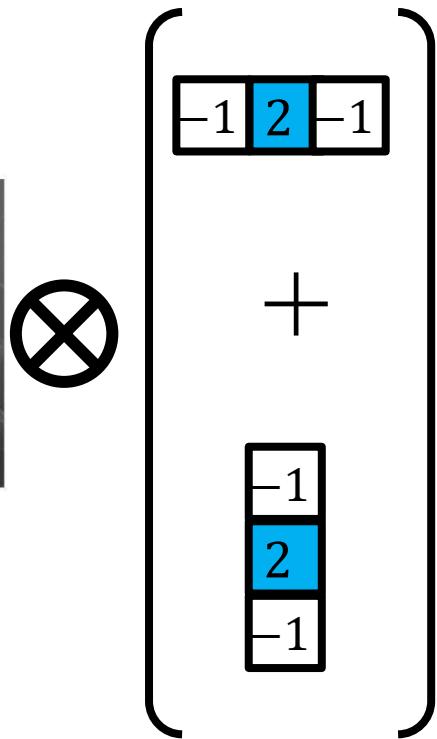
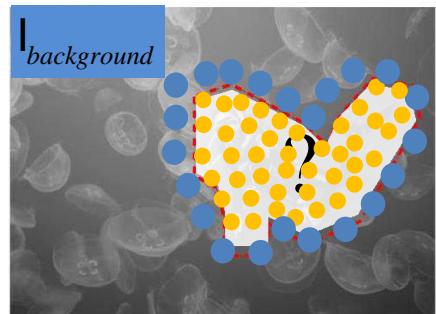


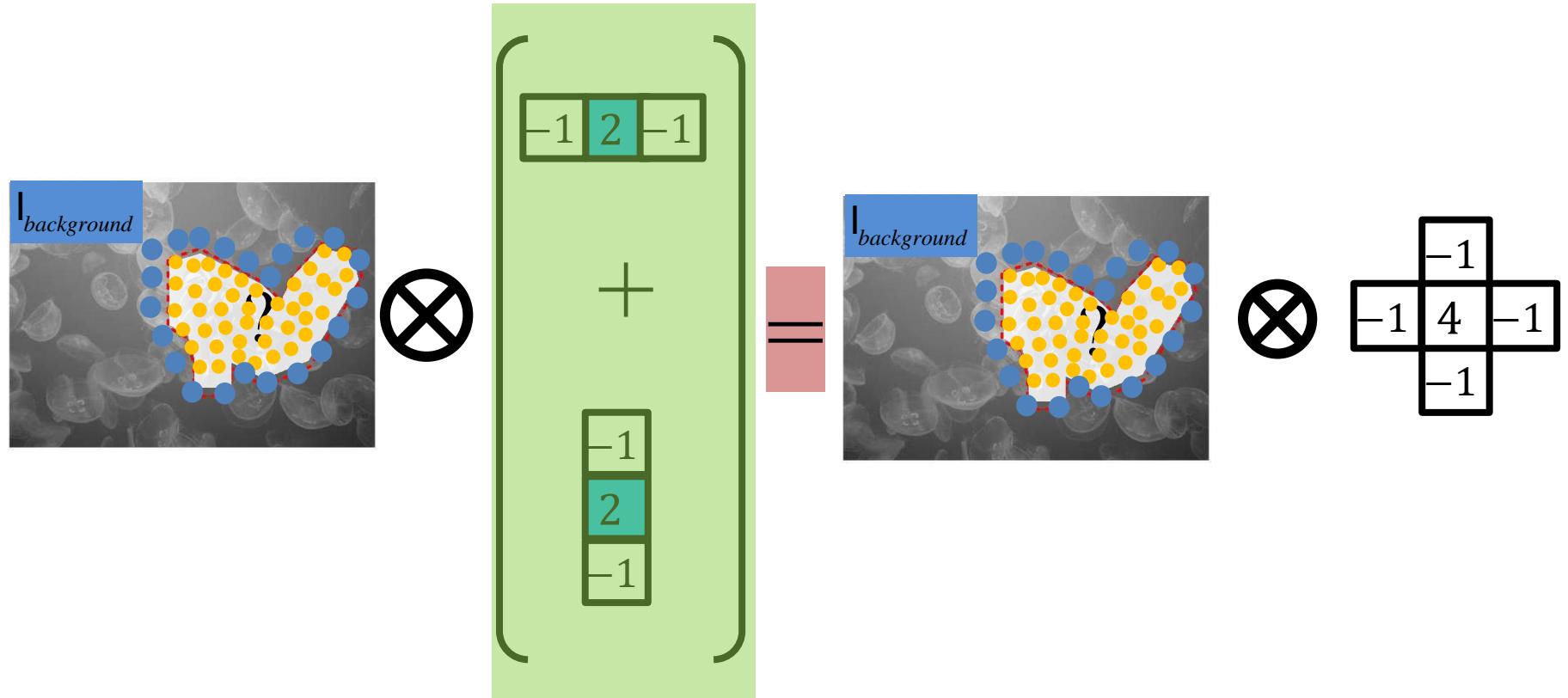
Video 8.3

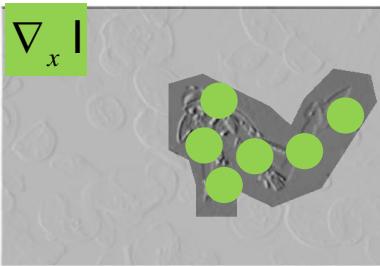
Jianbo Shi

Least square solution



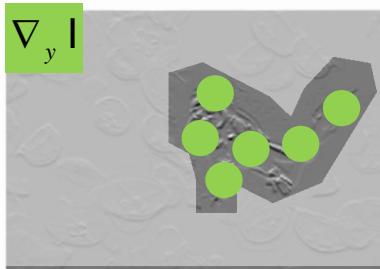






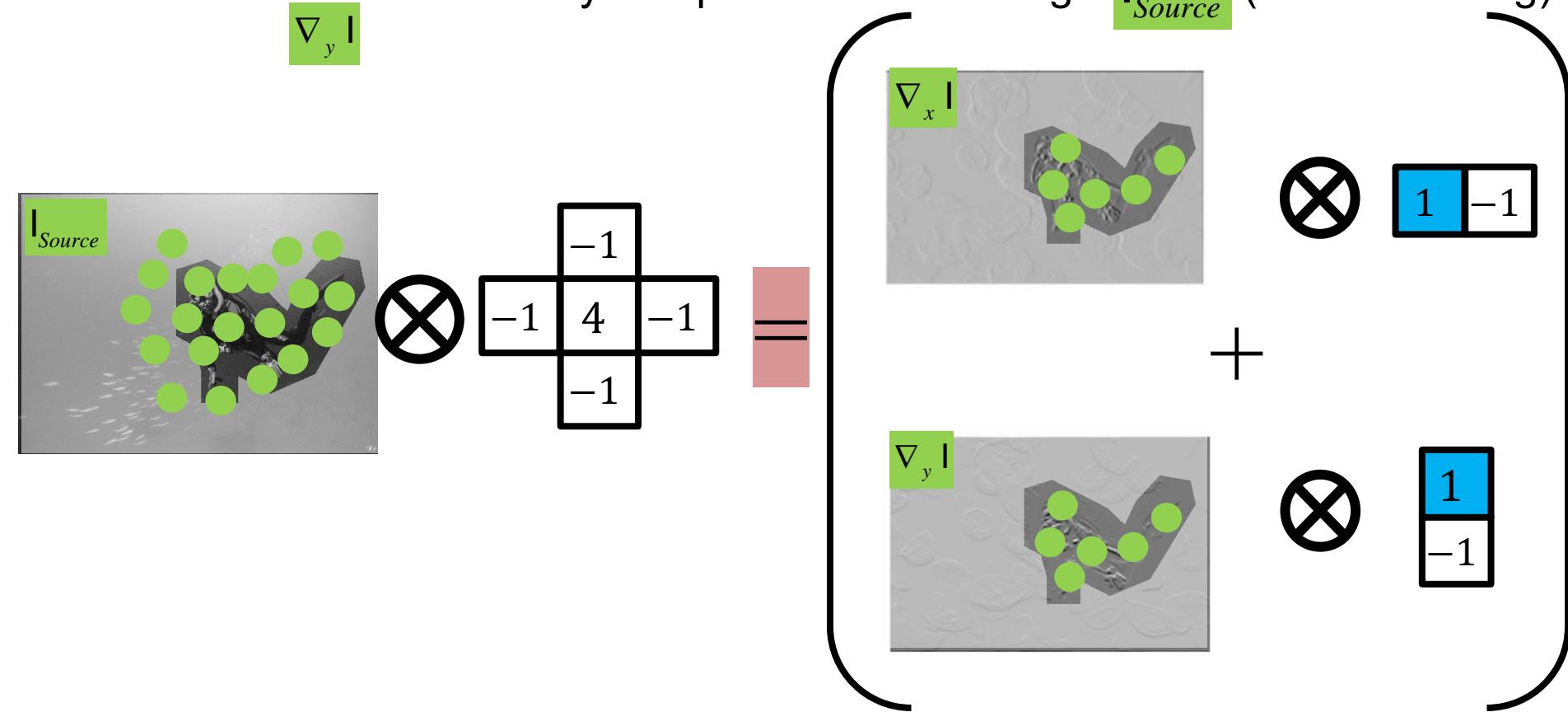
| | |
|---|----|
| 1 | -1 |
|---|----|

+



| | |
|---|----|
| 1 | -1 |
|---|----|

- When the $\frac{\nabla_x |}{\nabla_y |}$ are directly computed from an image $|_{Source}$ (without editing)

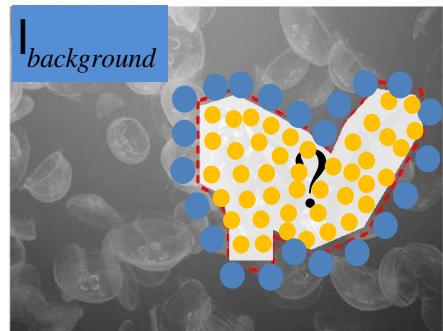


Solution on Pixels

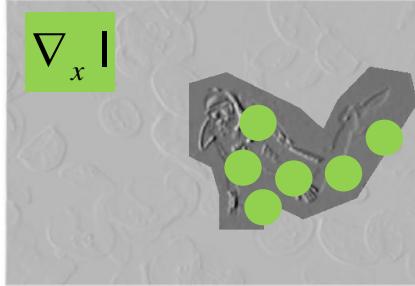
Given



Unknown



$$\begin{array}{c} -1 \\ \hline -1 & 4 & -1 \\ -1 \end{array}$$



$$\begin{array}{c} 1 \\ -1 \end{array}$$

+



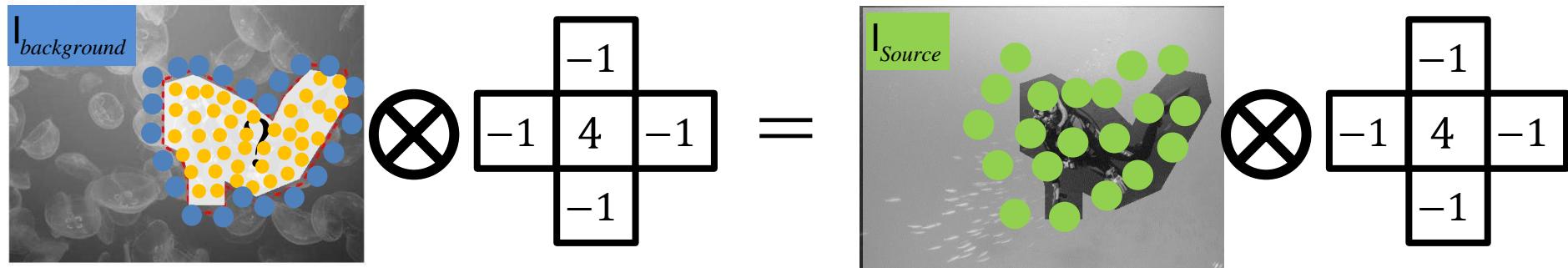
$$\begin{array}{c} 1 \\ -1 \end{array}$$

$$\begin{array}{c} -1 \\ \hline -1 & 4 & -1 \\ -1 \end{array}$$

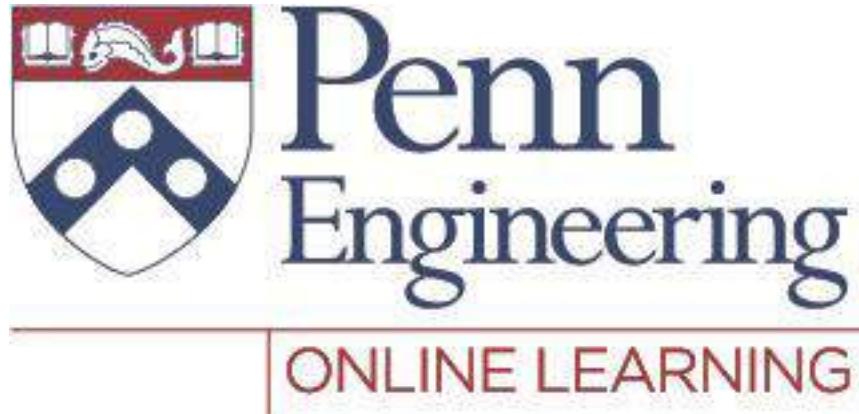
- The convolutional kernel for Laplacian operator

Special case

- When the $\nabla_x I$ and $\nabla_y I$ are directly computed from an image I_{Source} (without editing)



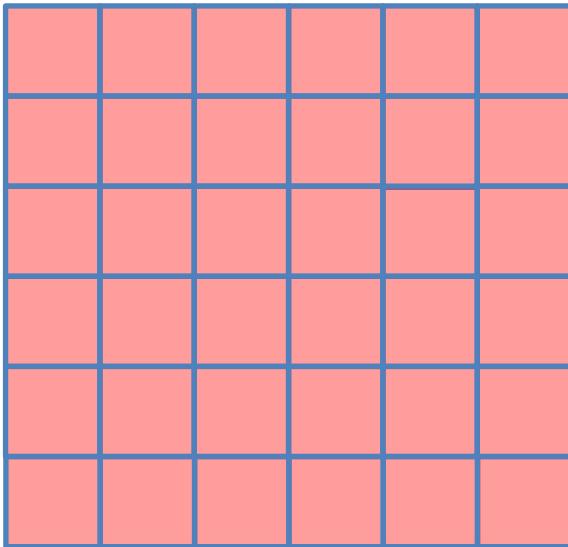
- Keep the value on the boundary $\partial\Omega$ the same
- Solve the equation for each channel (RGB) separately
- There is one equation matching the Laplacian values from the source to the target



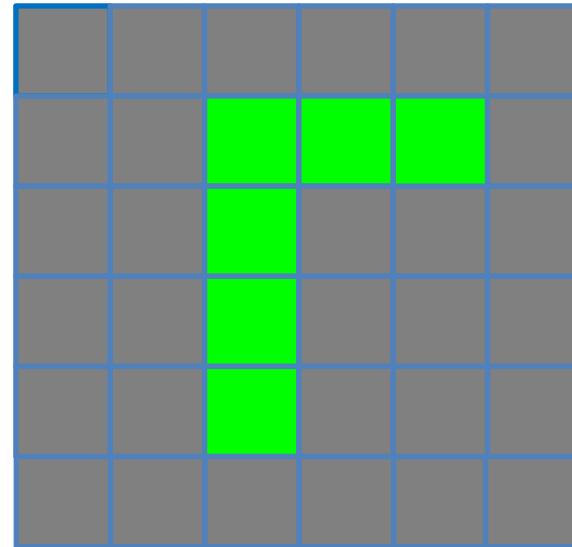
Video 8.4

Jianbo Shi

An example for gradient blending



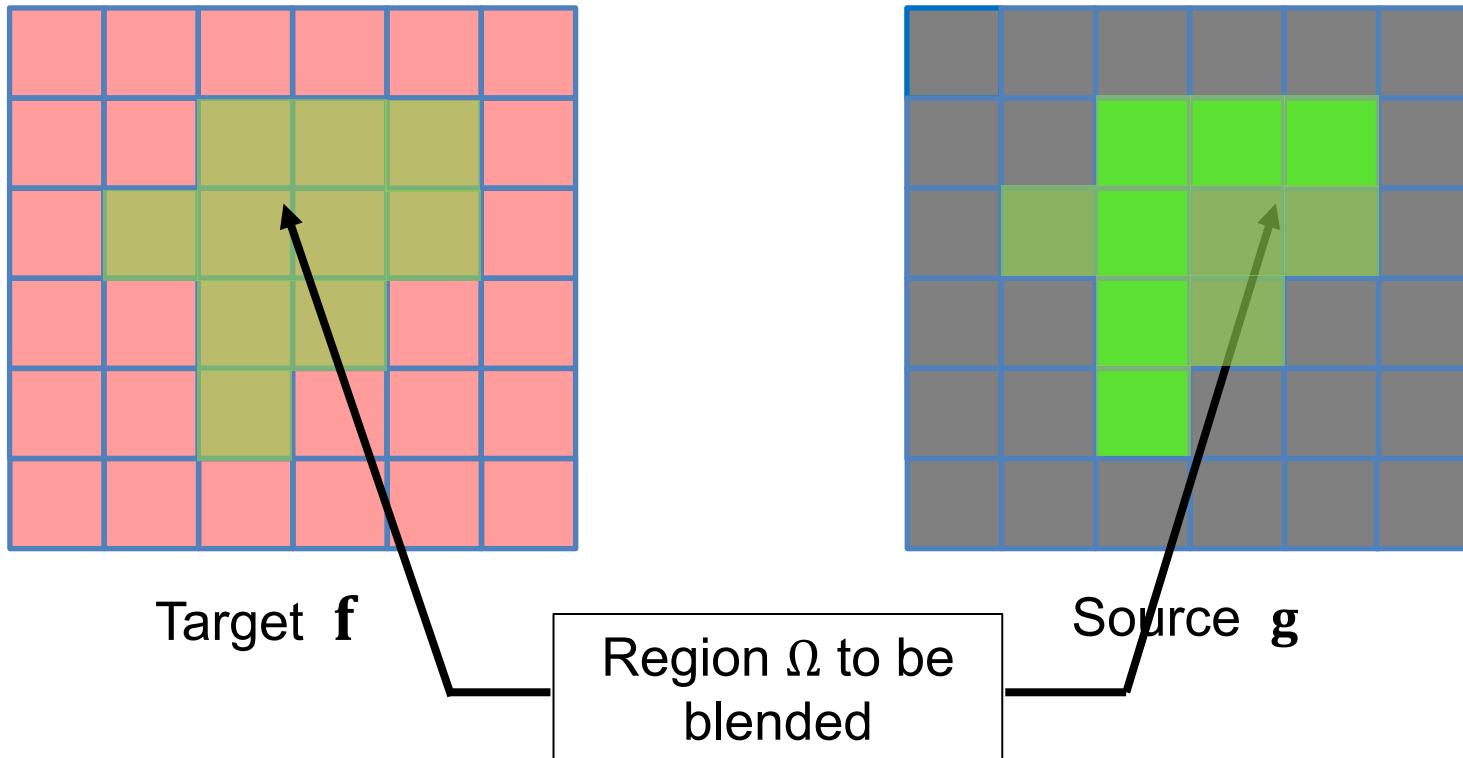
Target **f**



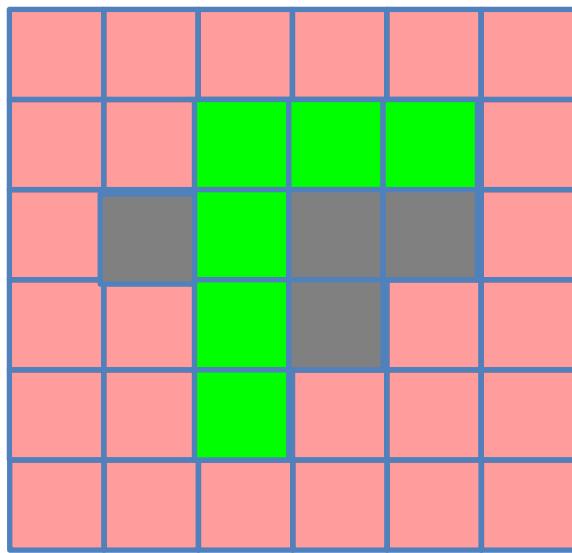
Source **g**

The RGB value for **f** is (255,0,0), background of **g** is (0,0,0)

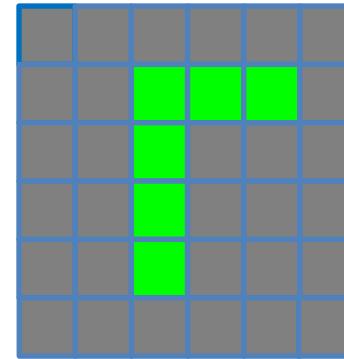
Mask for blending



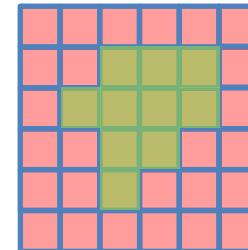
Direct copy and paste



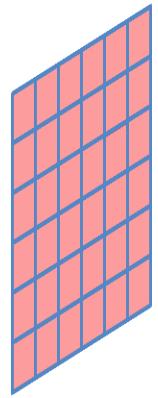
Blended f



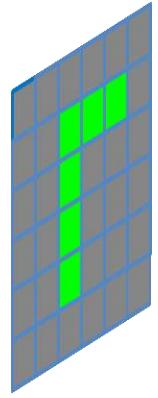
Source g



Mask

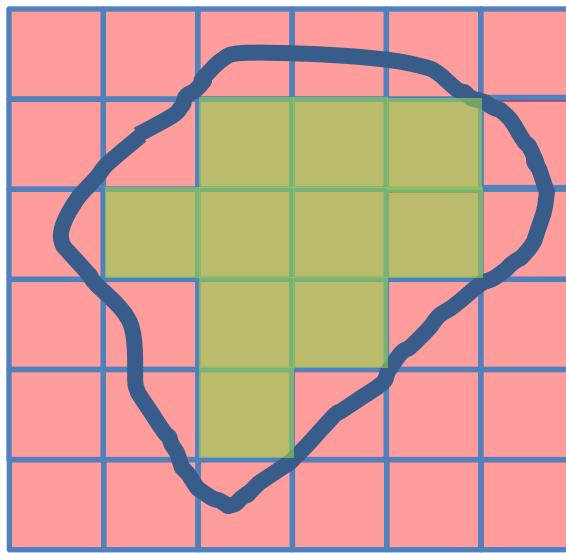


Target \mathbf{f}



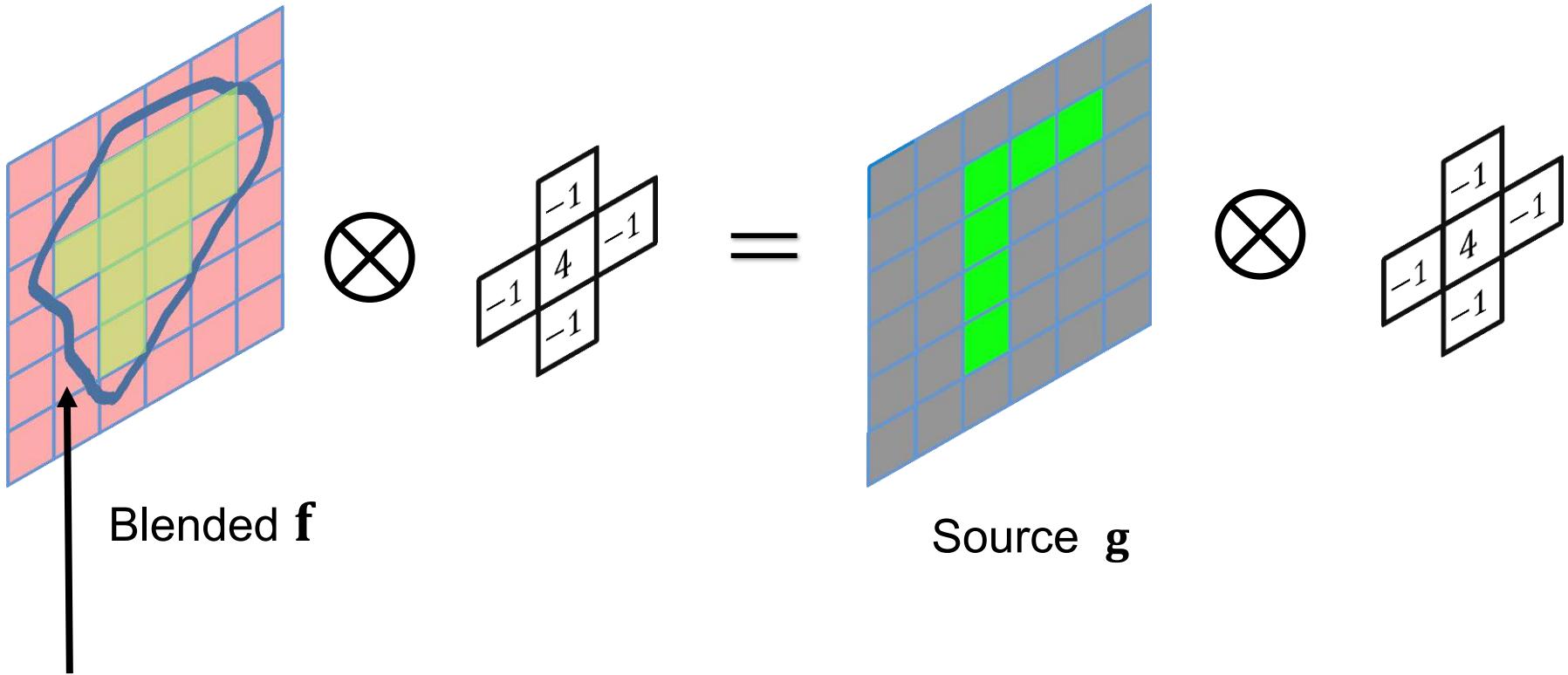
Source \mathbf{g}

Goal: recreate values in the mask with the two constraints



Mask Ω

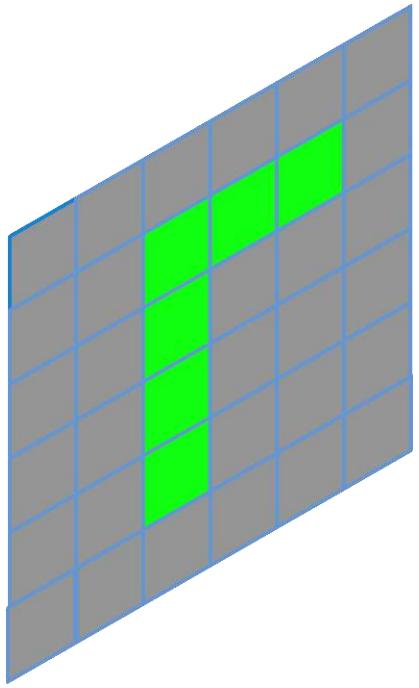
$$\left\{ \begin{array}{l} \Delta \mathbf{f} = \mathbf{b} \text{ in } \Omega \\ \mathbf{f}|_{\partial\Omega} \text{ keeps same} \end{array} \right.$$



Region boundary $\partial\Omega$

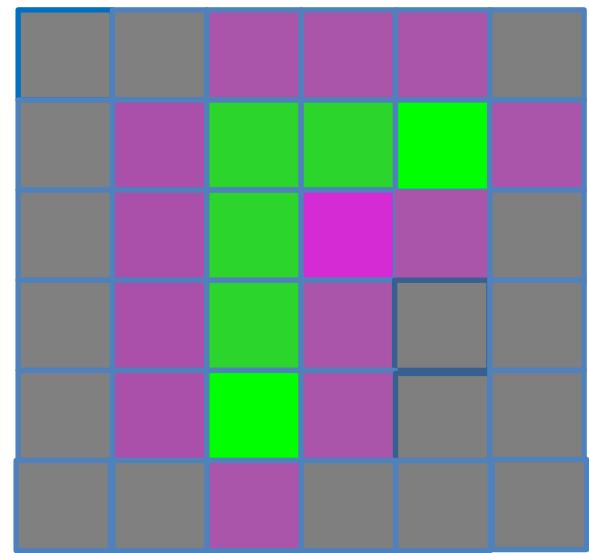
Solution: recreate values in the mask +
Keep f the same on the boundary $\partial\Omega$

Laplacian of the
source

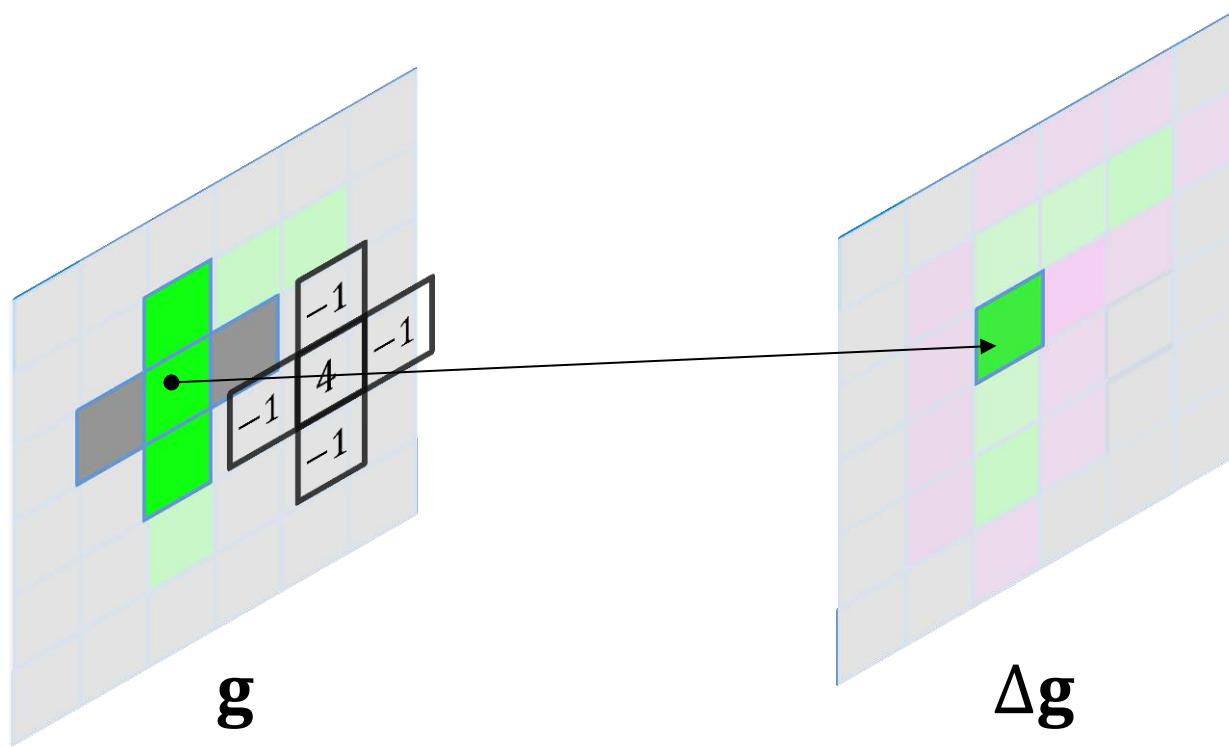


$$\begin{matrix} -1 & & -1 \\ -1 & 4 & -1 \\ -1 & & -1 \end{matrix}$$

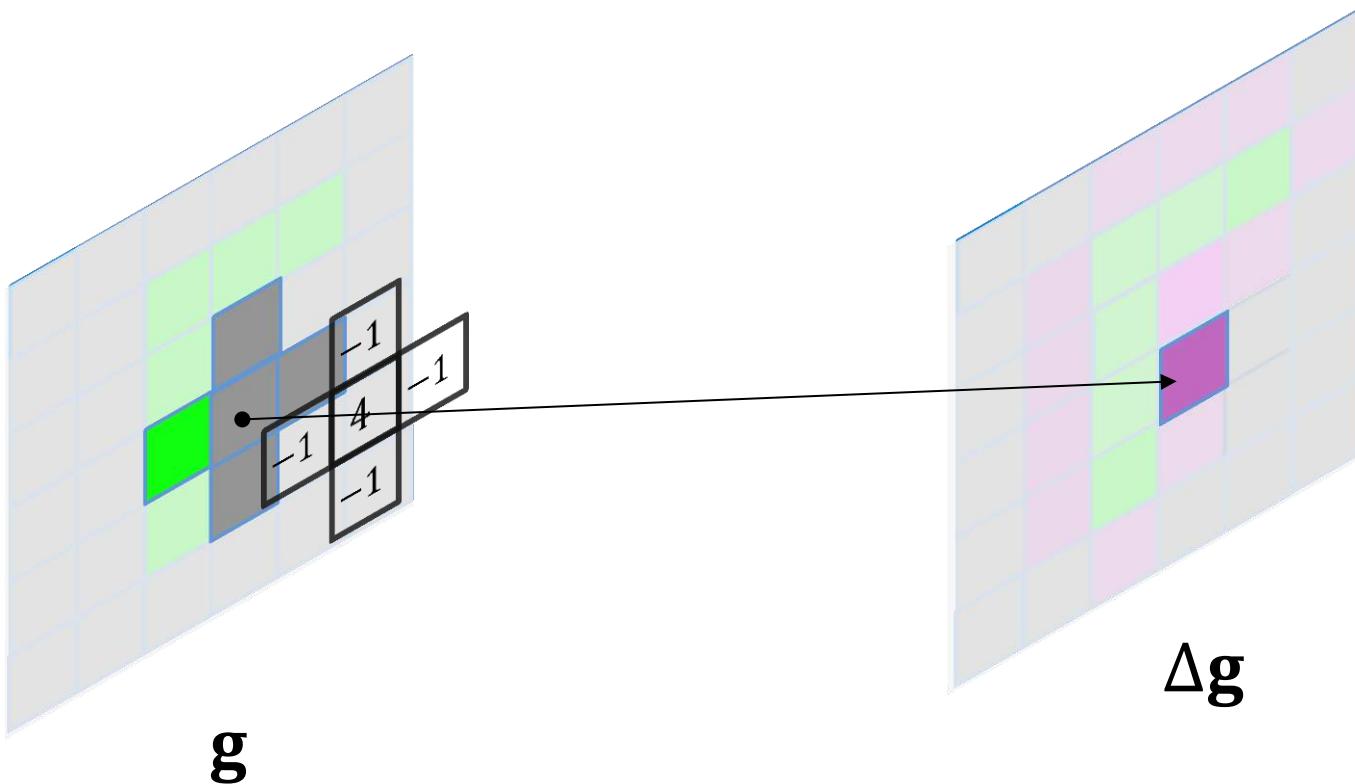
=

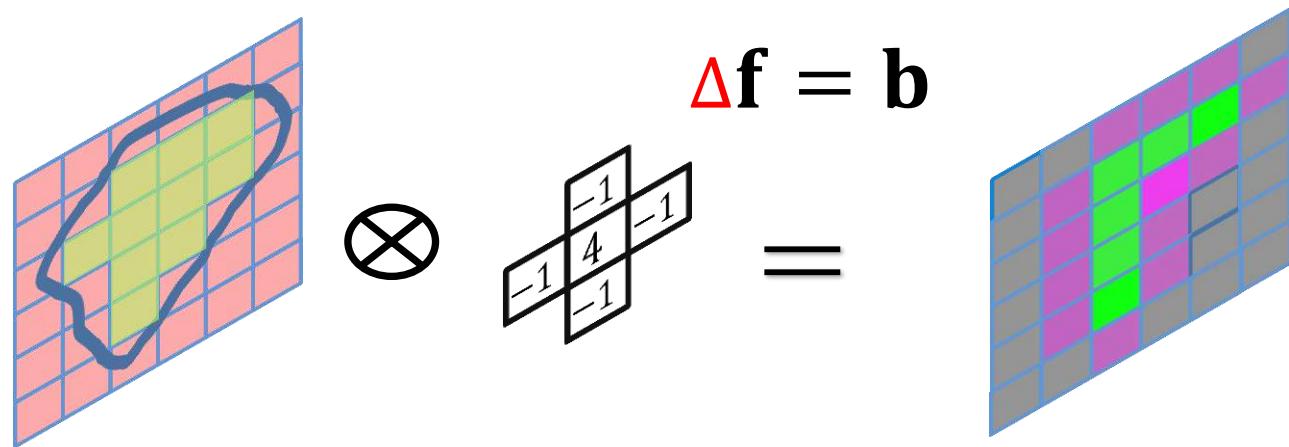


Laplacian of the source

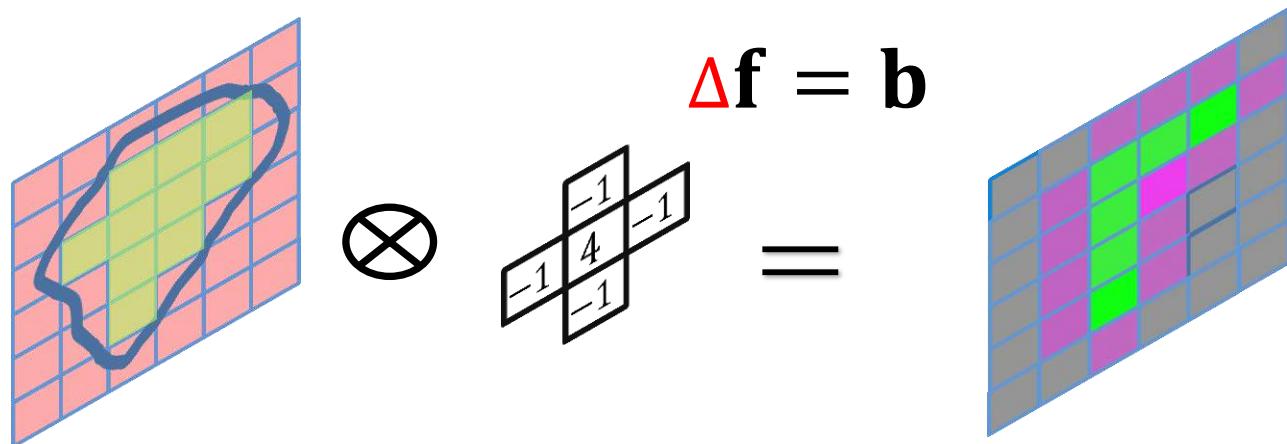


Laplacian of the source



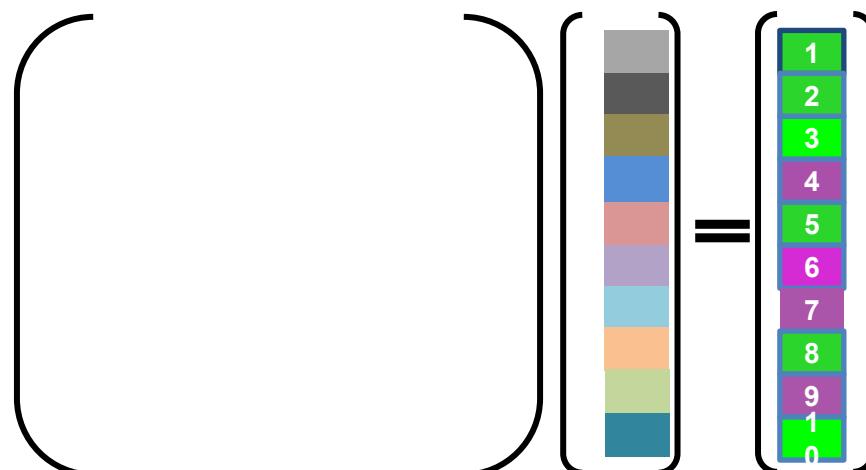


we have per pixel equation for the blended 2D image



we have per pixel equation for the blended 2D image
...will rewrite them as Matrix-Vector equation

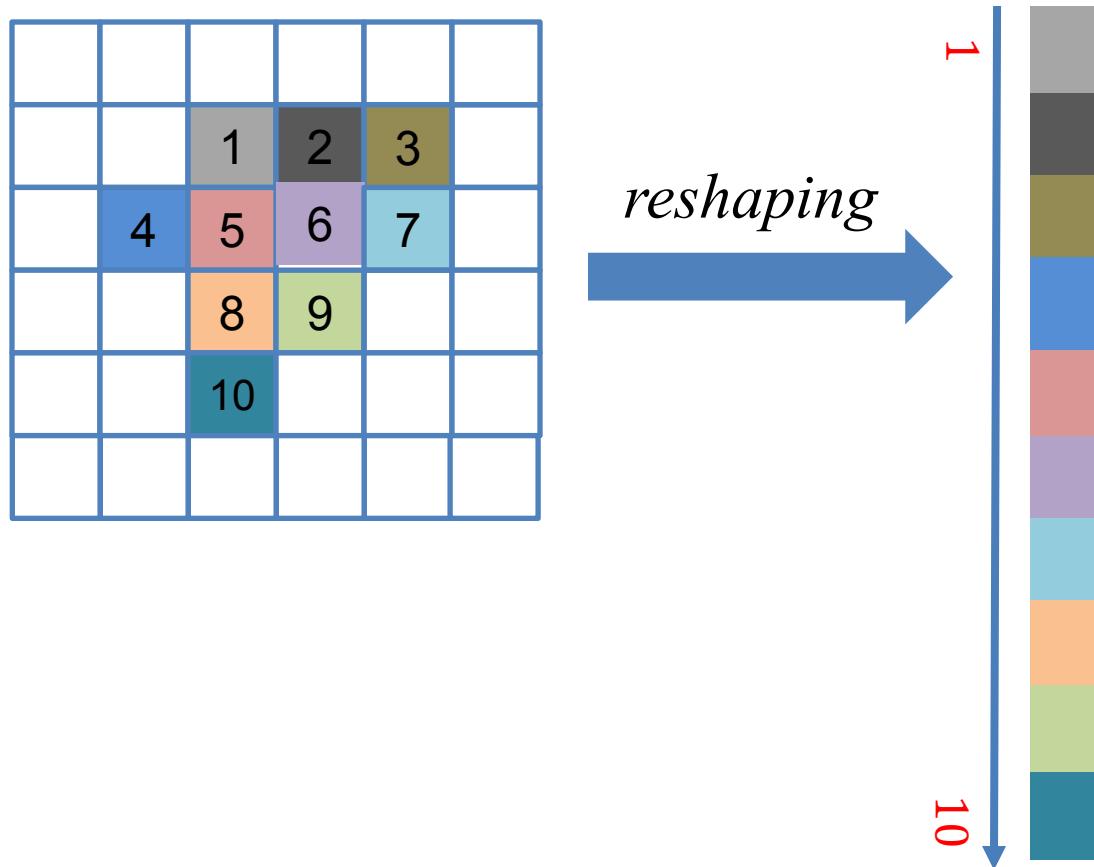
$$Af = b$$



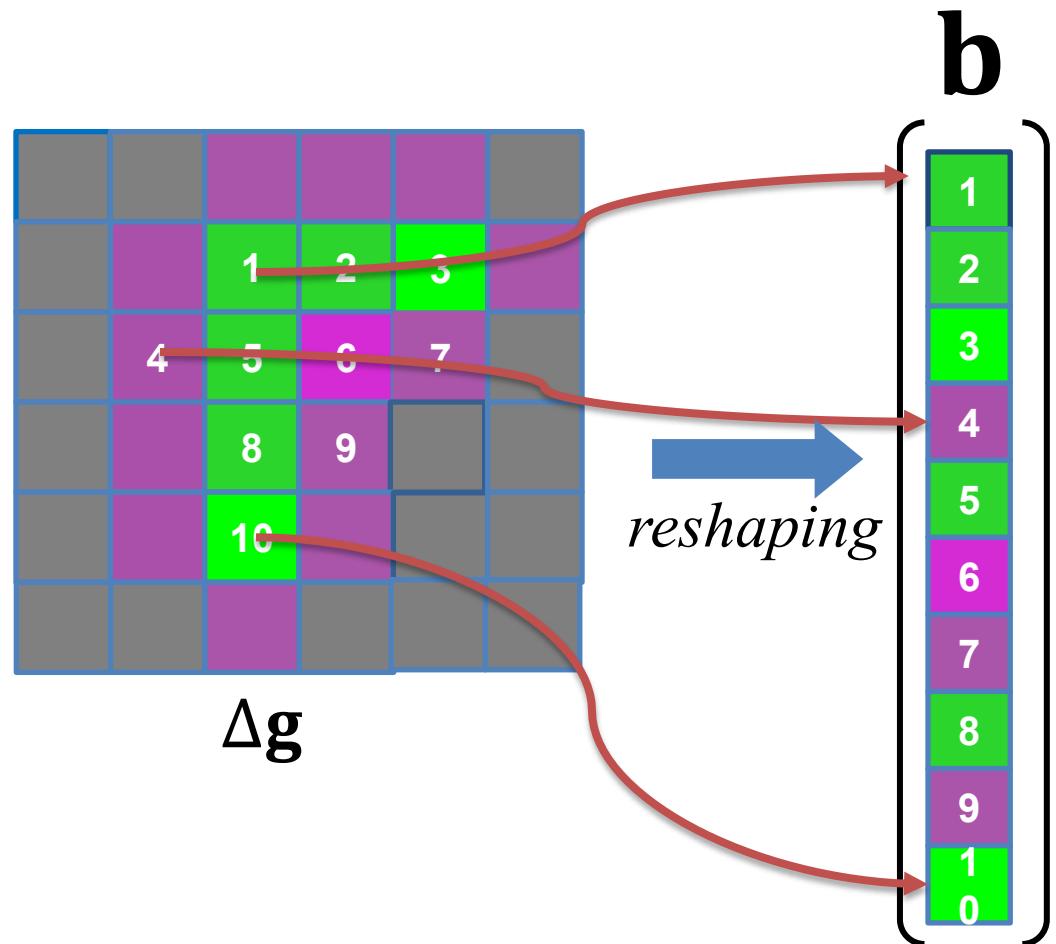
$$\mathbf{A}\mathbf{f} = \mathbf{b}$$

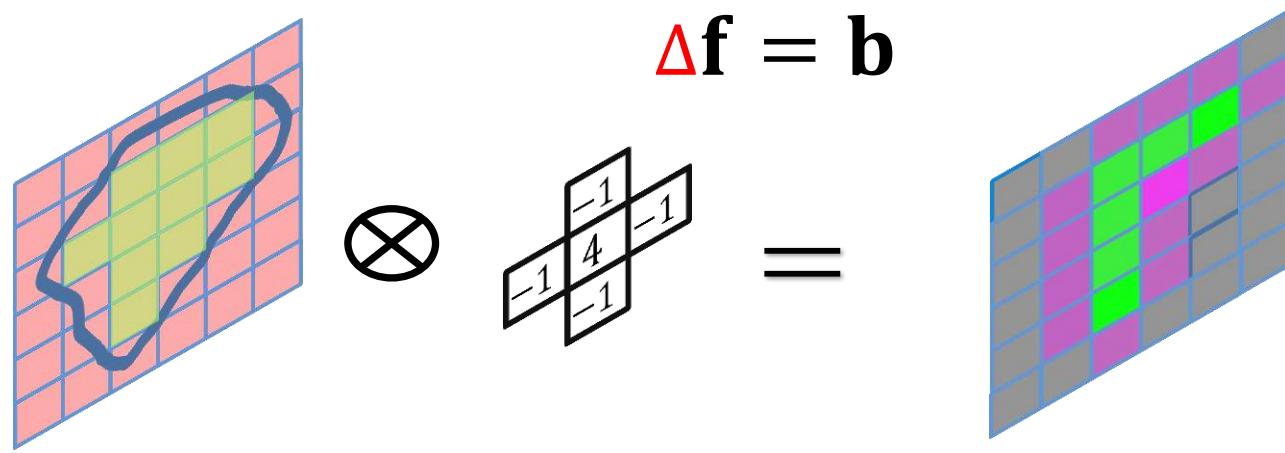
Think 2D image as a 1D vector

For each pixel in the mask,
we first create a 1D indexing vector



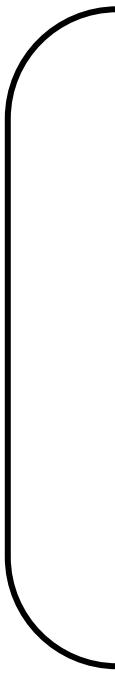
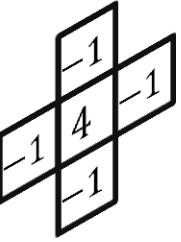
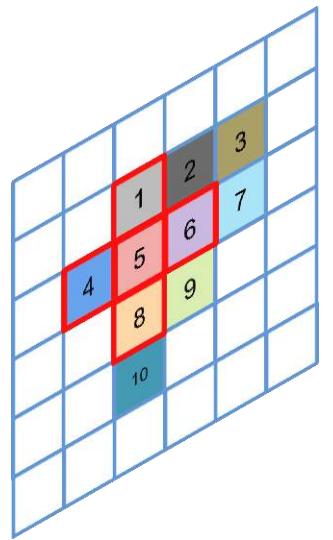
Copying and *reshaping* the Laplacian of source



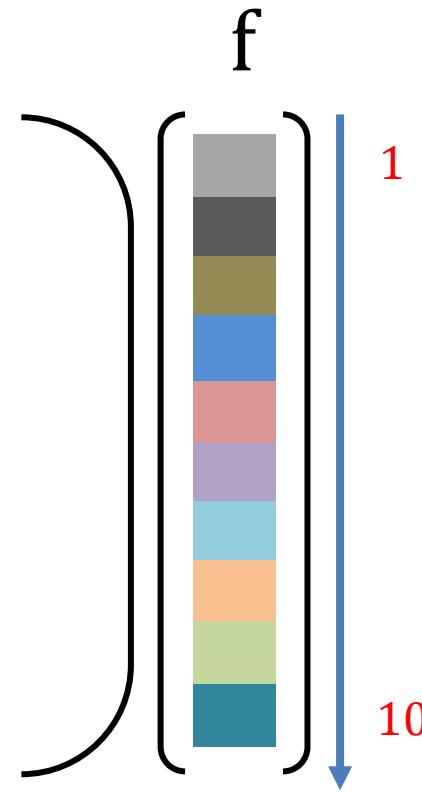


We can use a matrix \mathbf{A} to encode 2D Convolution!

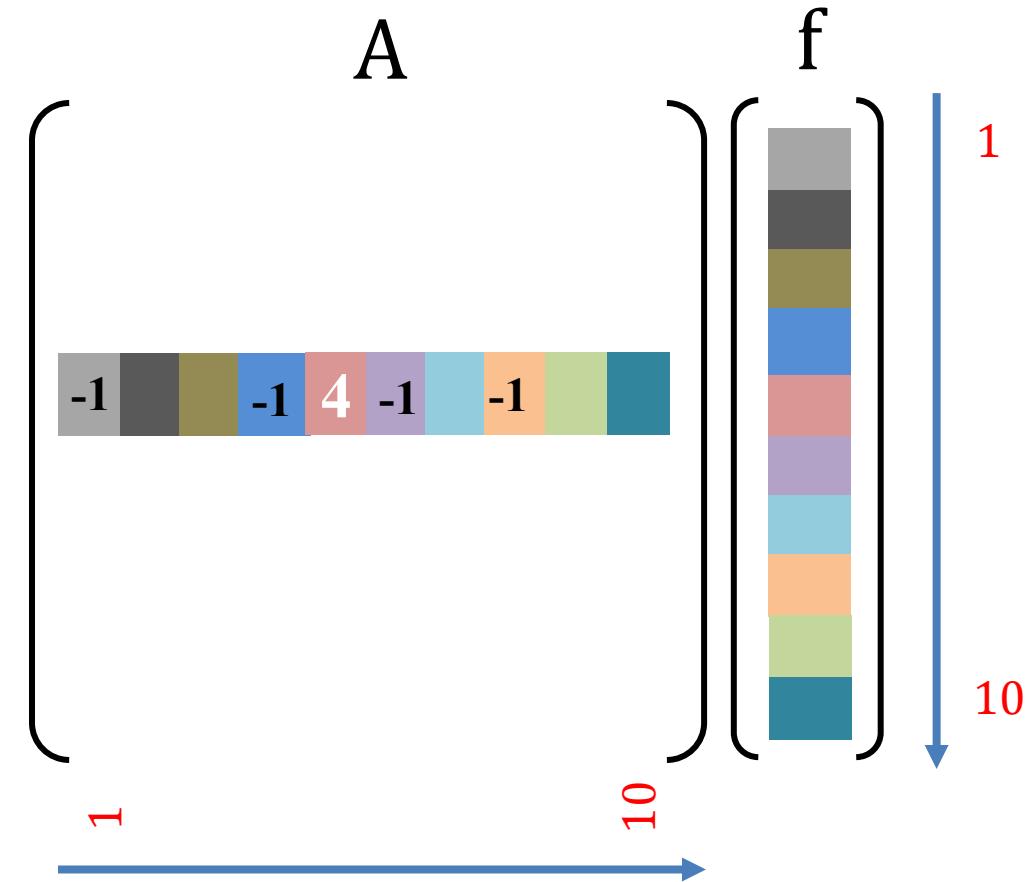
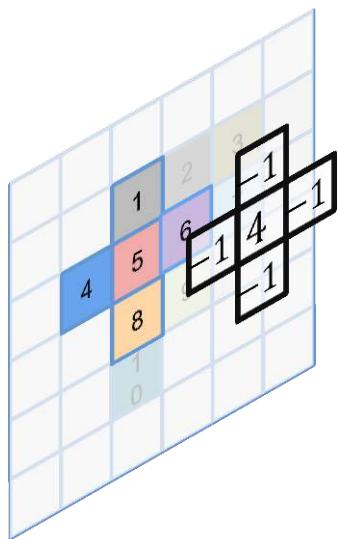
$$\mathbf{Af} = \mathbf{b}$$



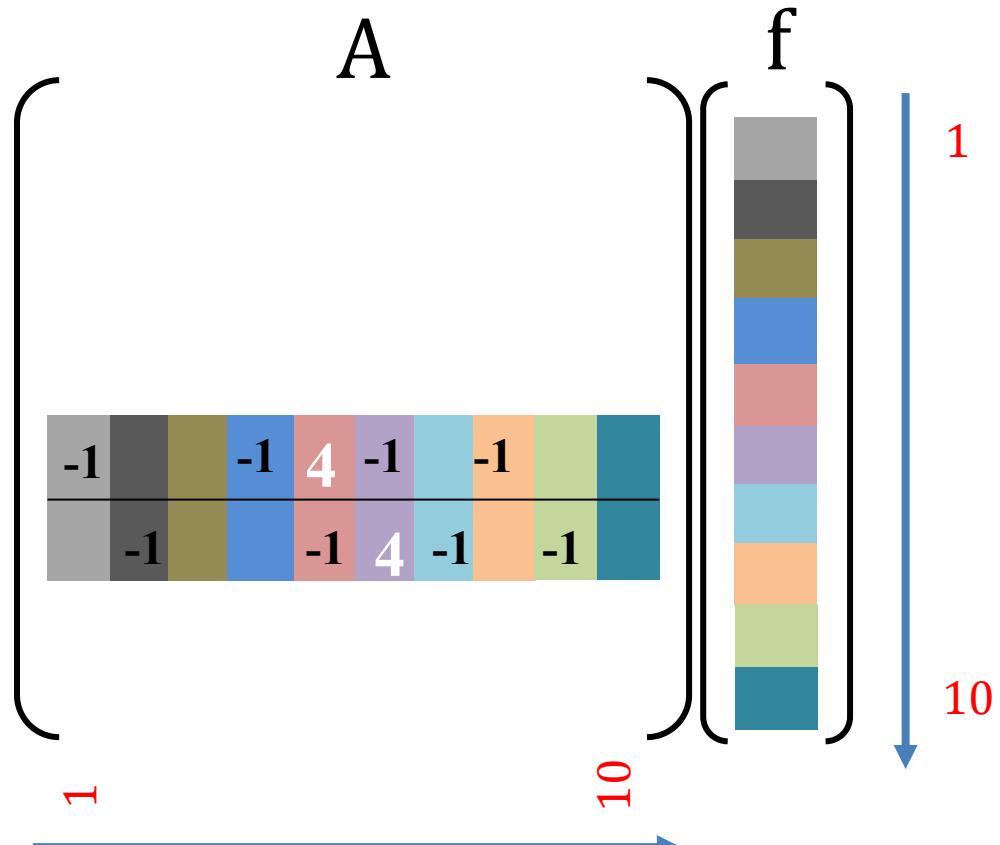
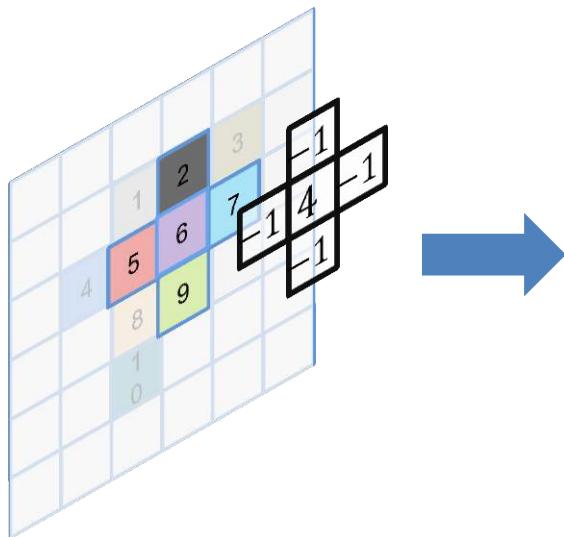
A?



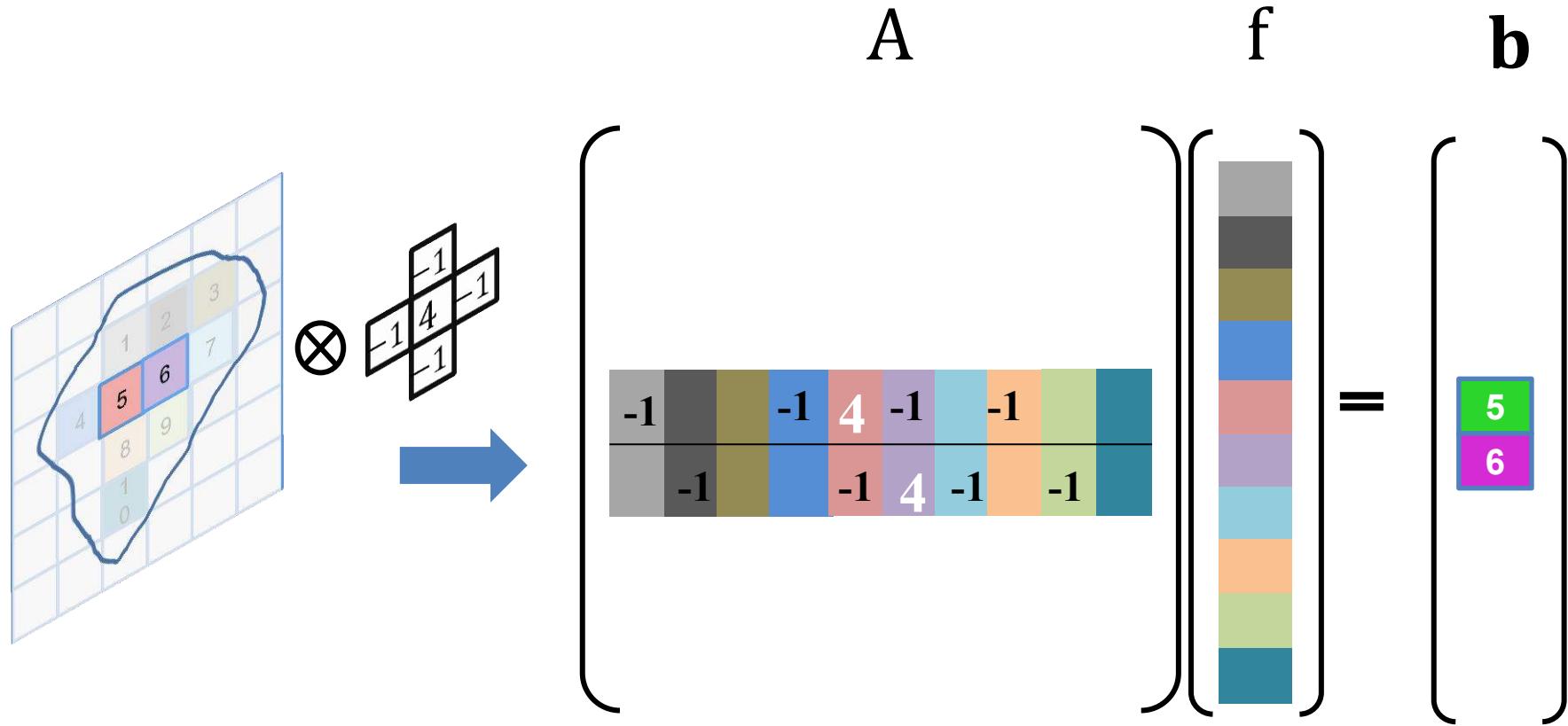
Matrix A encoding the Laplacian Convolution



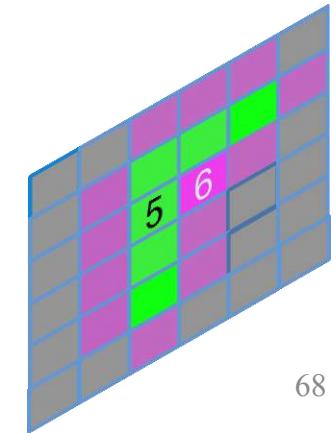
Matrix A encoding the Laplacian Convolution



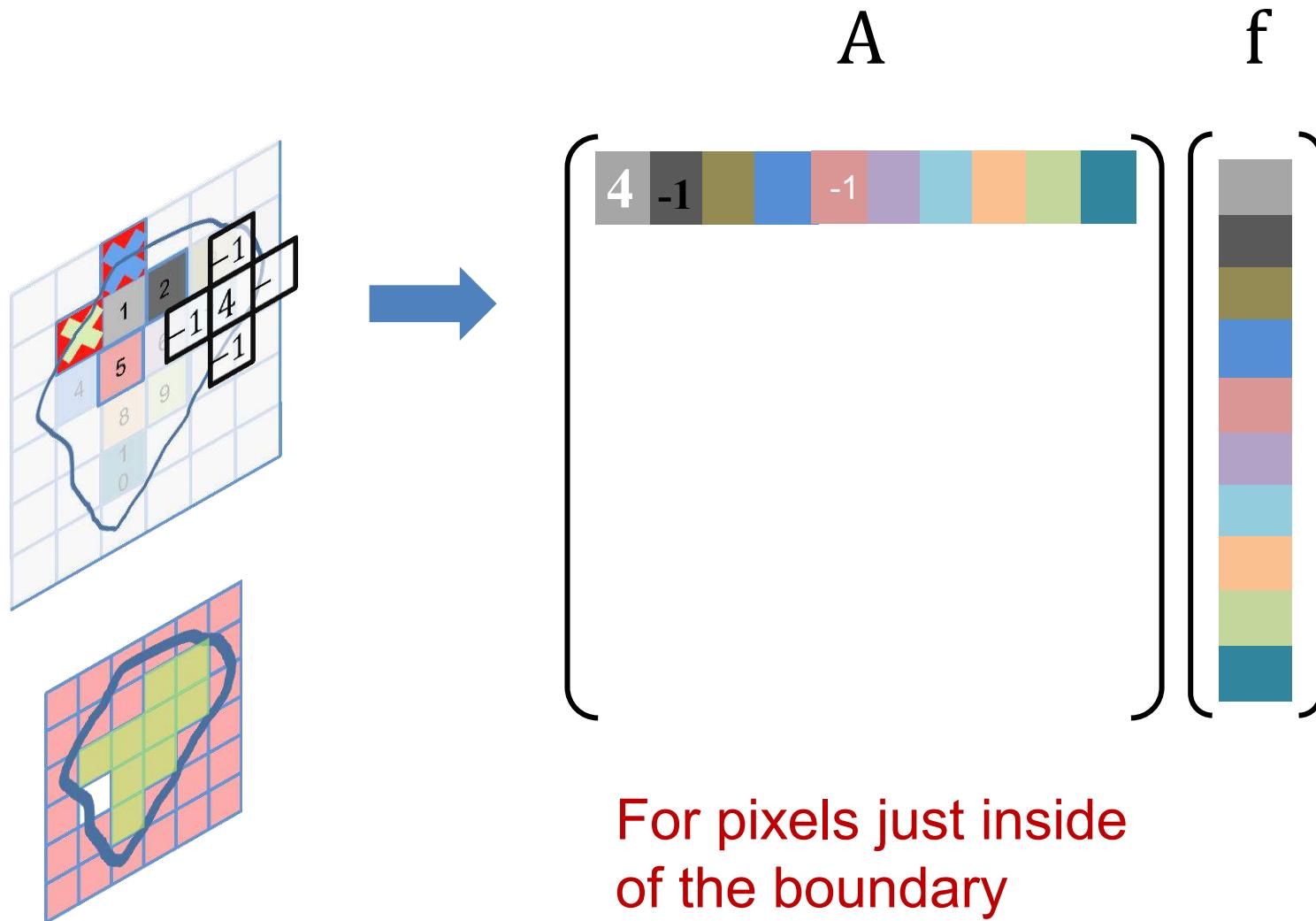
For interior pixels



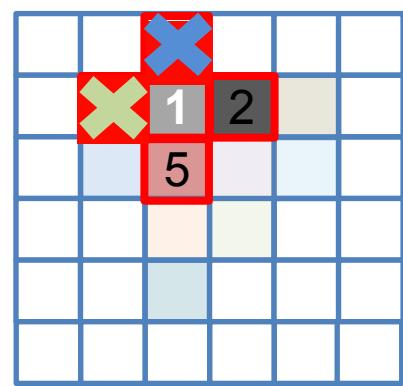
For all interior pixels



How to deal with the knowns Boundary Values?



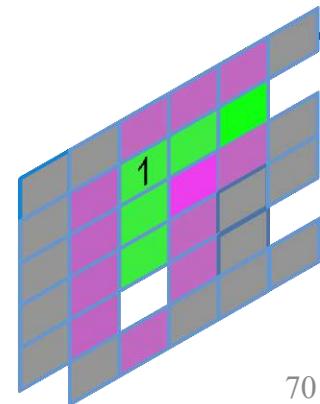
Move the boundary value of knowns to **b** side!



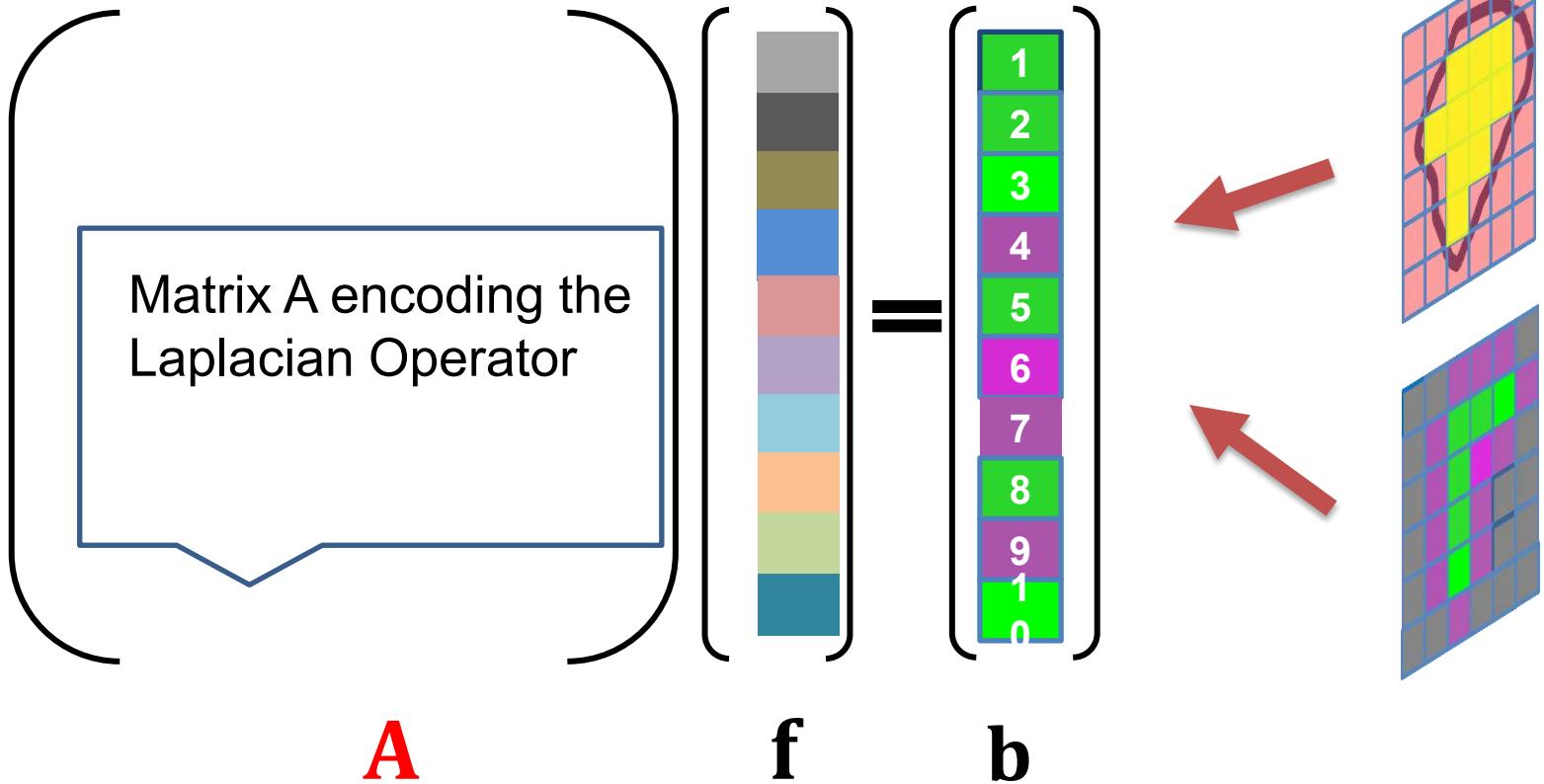
$$\left[\begin{array}{c} A \\ f \end{array} \right] = \left[\begin{array}{c} b \end{array} \right]$$

The diagram illustrates the matrix equation $A\mathbf{f} = \mathbf{b}$. On the left, the matrix A is shown as a row vector with entries: 4, -1, -1, 1, 1. The vector \mathbf{f} is shown as a column vector with entries: 1, 2, 5, 1, 1. An equals sign separates the left side from the right side, which contains the vector \mathbf{b} as a column vector with entries: 1, X , X .

$$f(1) \times 4 - f(2) - f(5) - \text{X} - \text{X} = b(1)$$

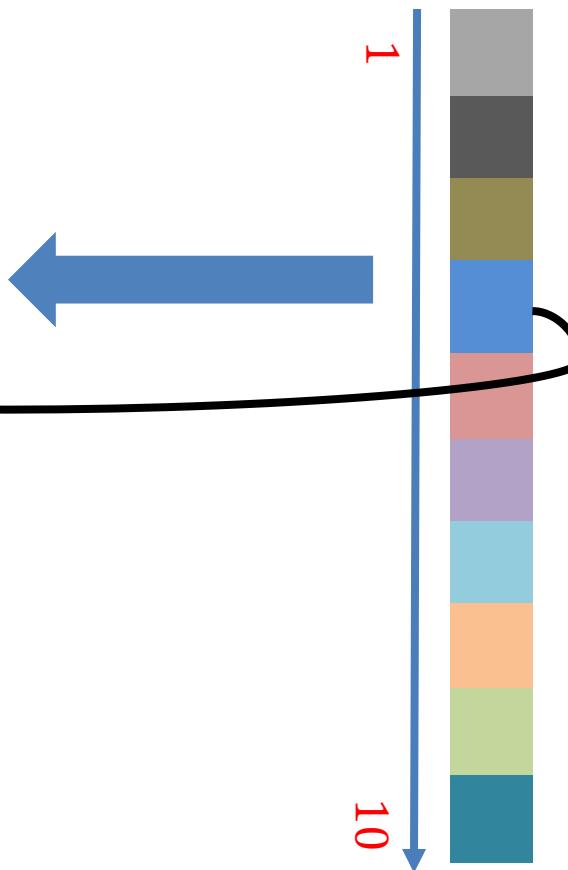
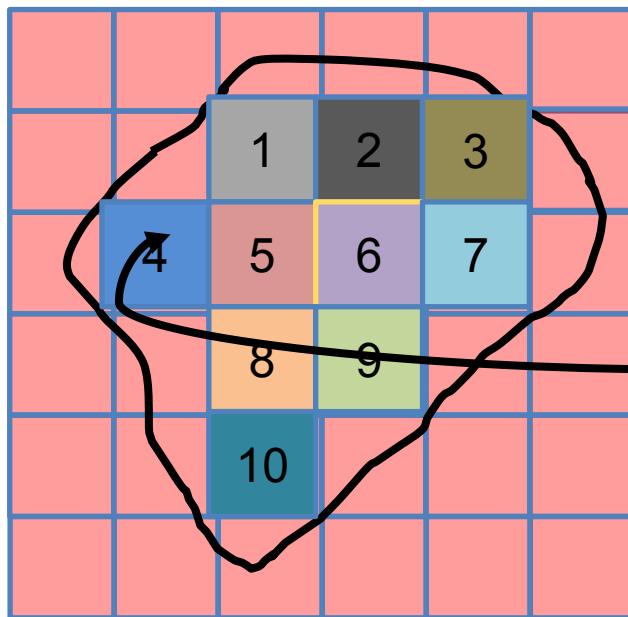


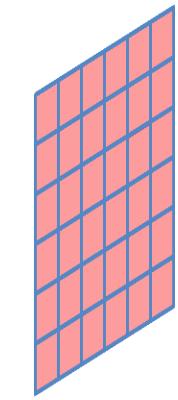
$$\mathbf{A}\mathbf{f} = \mathbf{b}$$



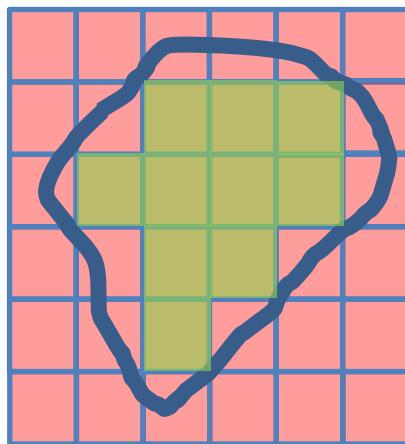
For each pixel in the mask,
put f back to the blended image

$$\mathbf{f} = \mathbf{A} \backslash \mathbf{b}$$

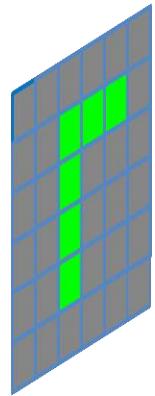




Target f



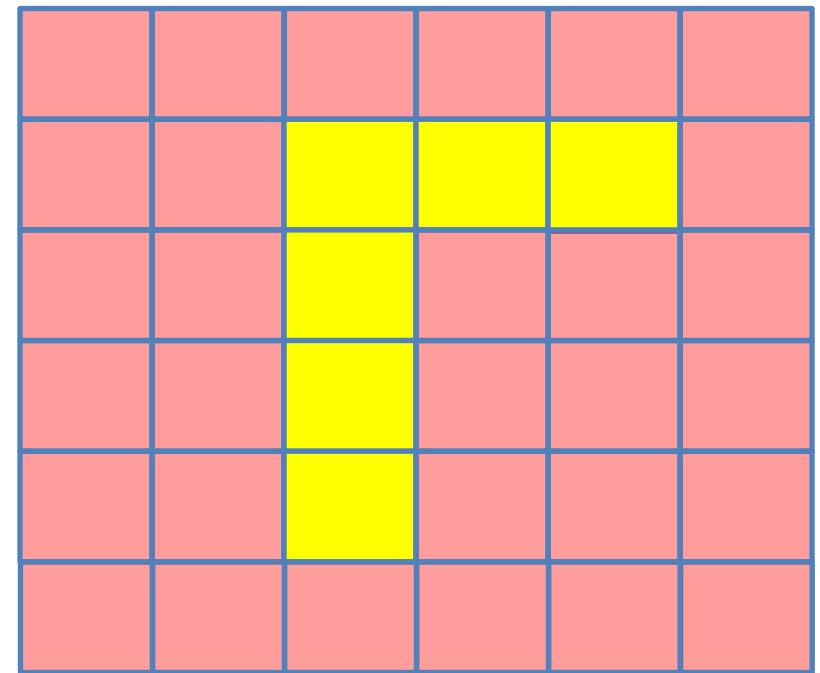
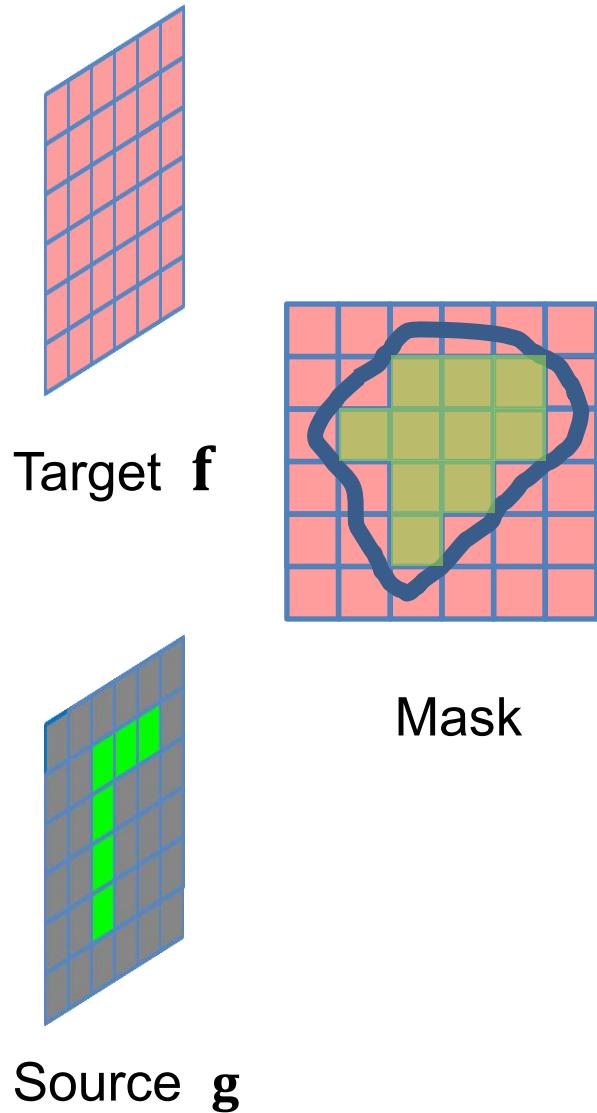
Mask

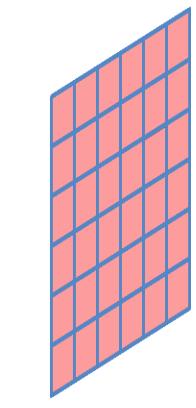


Source g

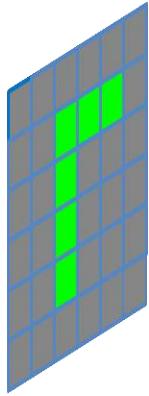
Guess what's our
blended image?

Gradient Blending

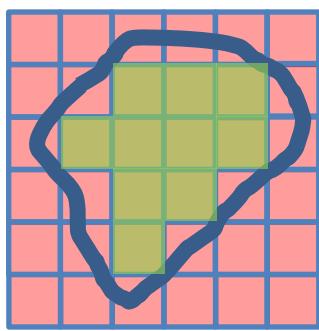




Target f

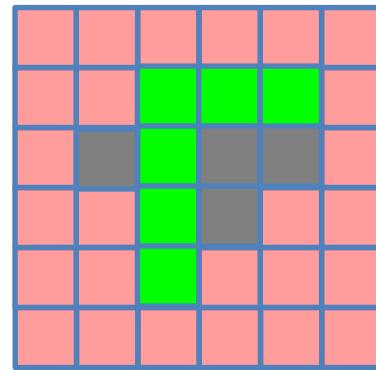


Source g

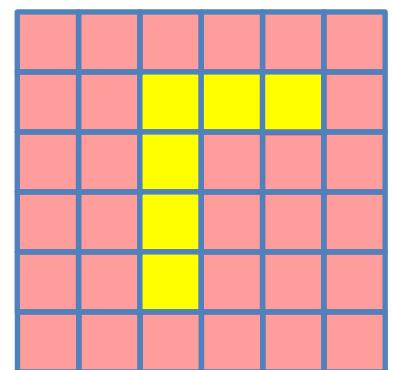


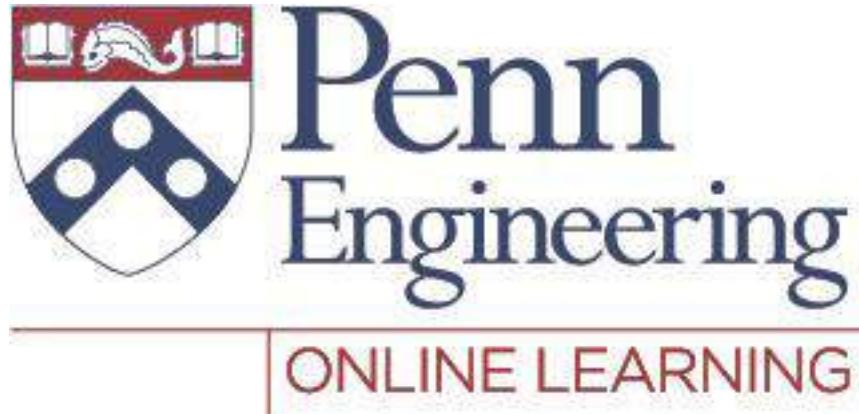
Mask

Direct Copy
and Paste



Gradient
Blending





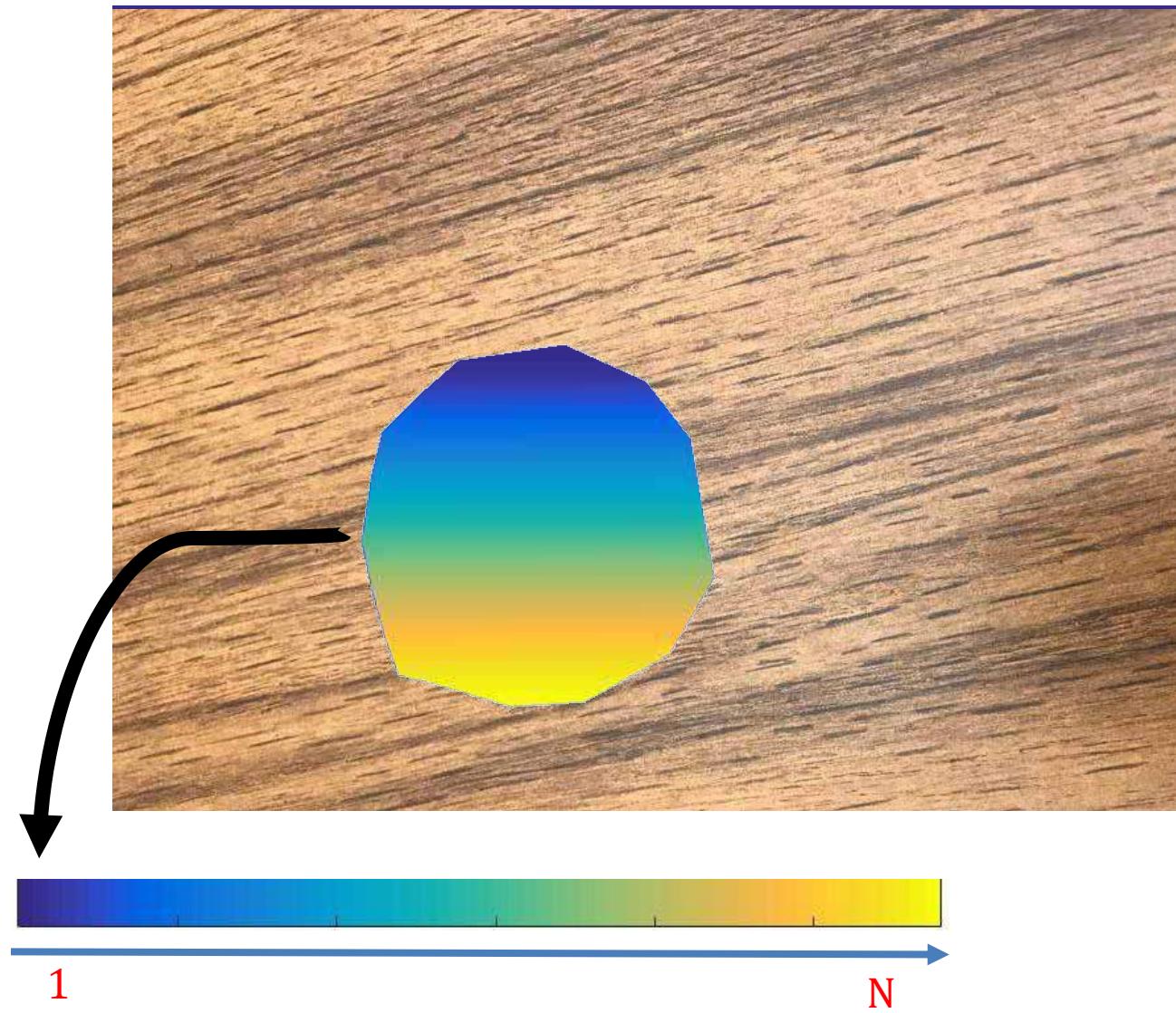
Video 8.5

Jianbo Shi

Let's summarize with a real image



Step1: Indexing the unknowns in masked pixels

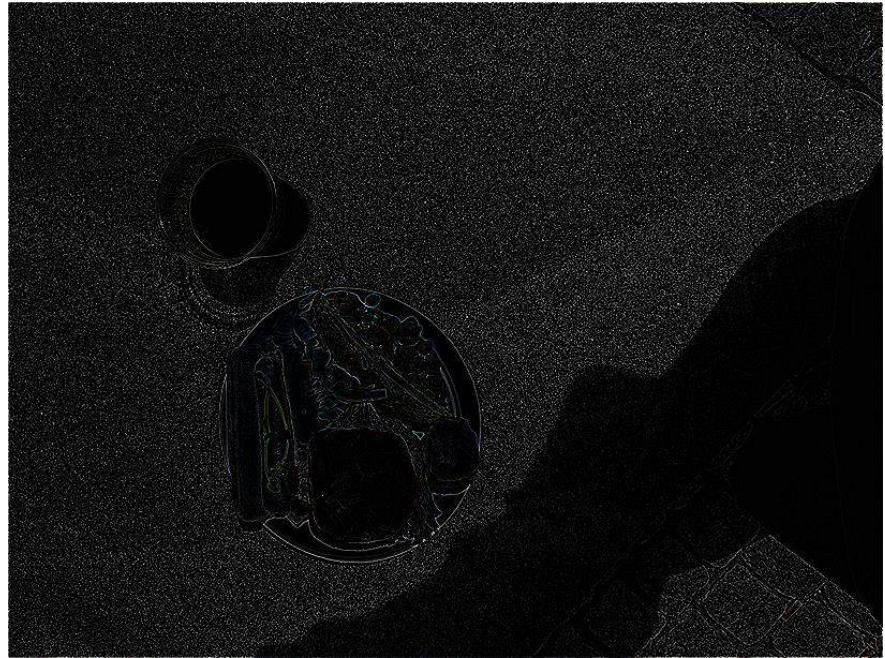


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Step2: Compute the Laplacian from source

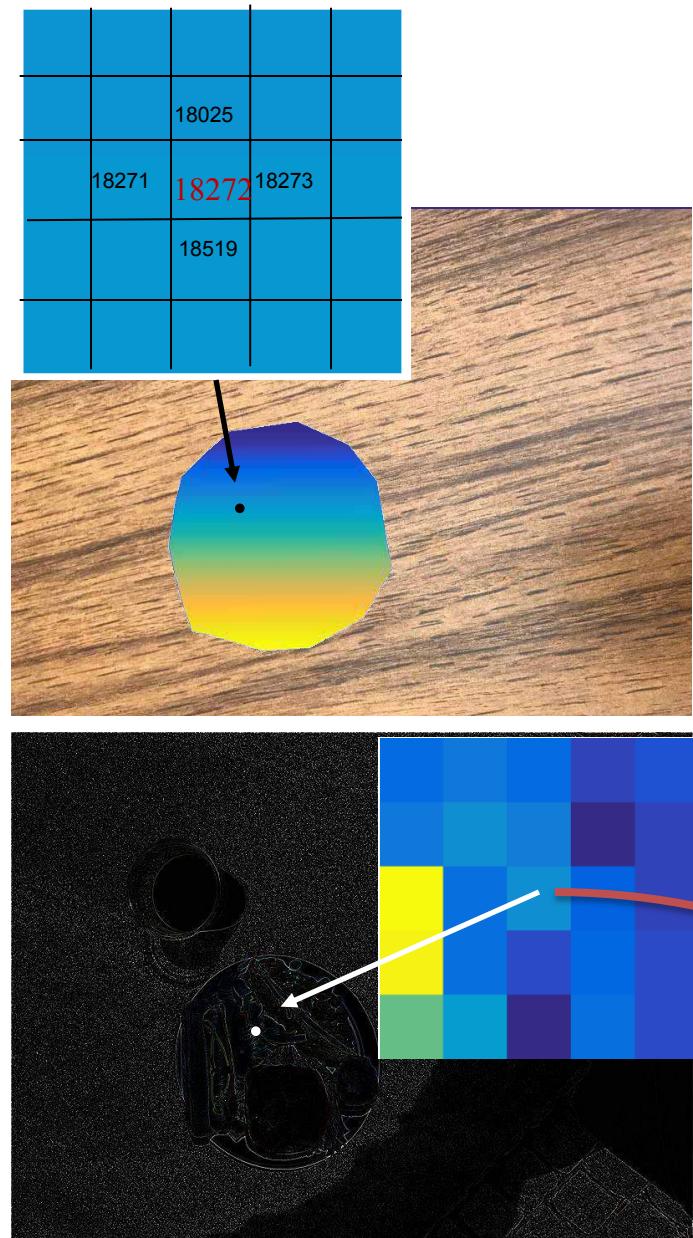


Source image



Laplacian

Step3: Constructing matrix A for the Laplacian Operator



For pixel in the region

A

$b(18272) =$

$A(18272, 18272) = 4$

$A(18272, 18025) = -1$

$A(18272, 18519) = -1$

$A(18272, 18271) = -1$

$A(18272, 18273) = -1$

f

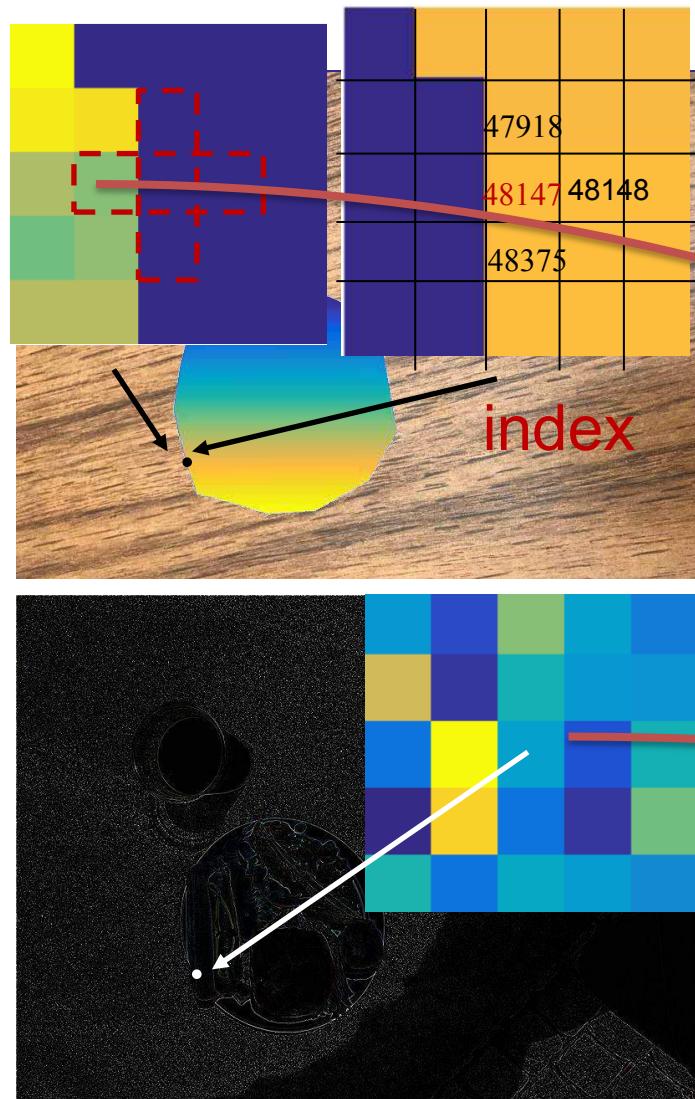
?

=

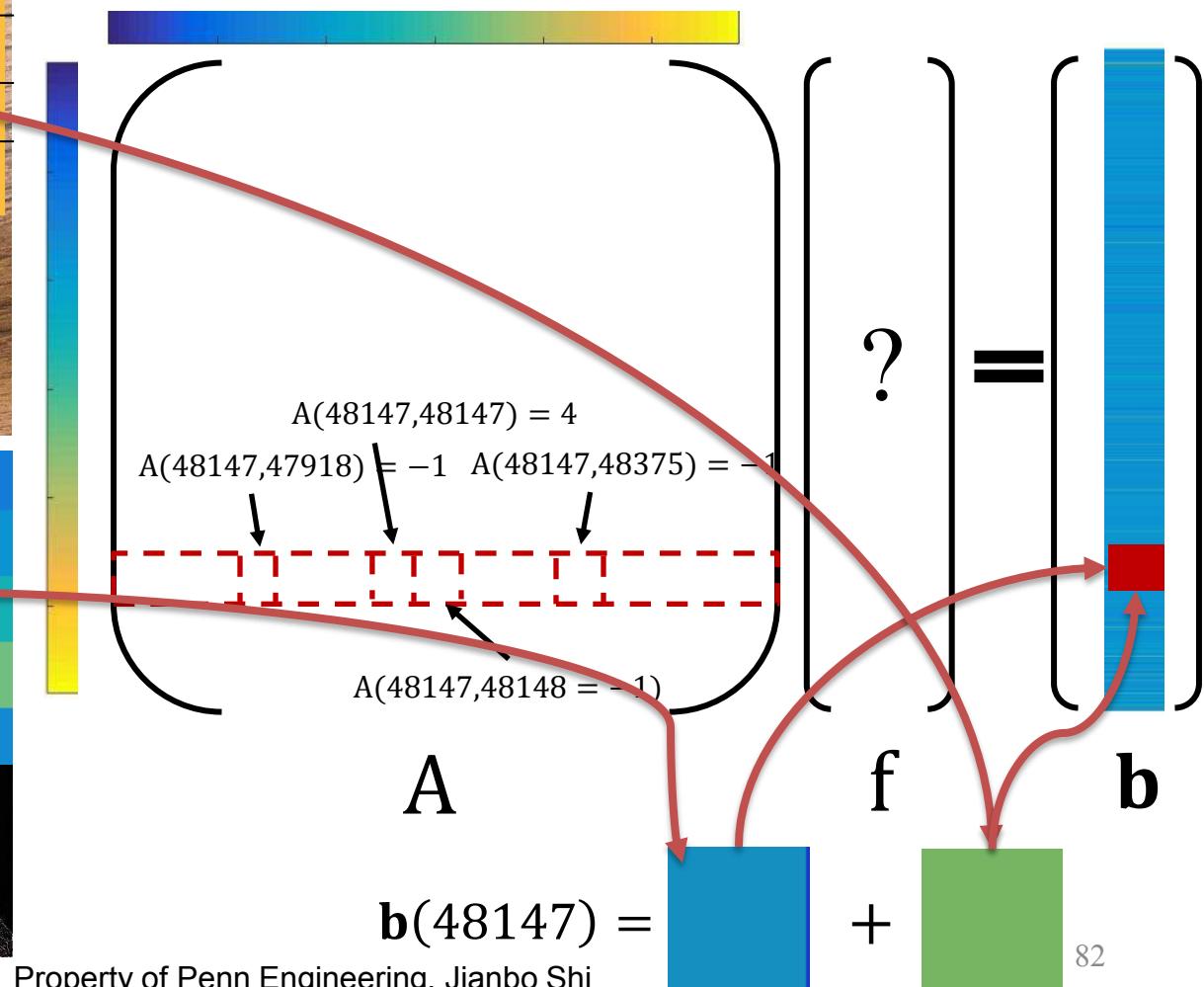
b

Step3. Constructing matrix A for the Laplacian Operator

boundary value



For pixel on the boundary



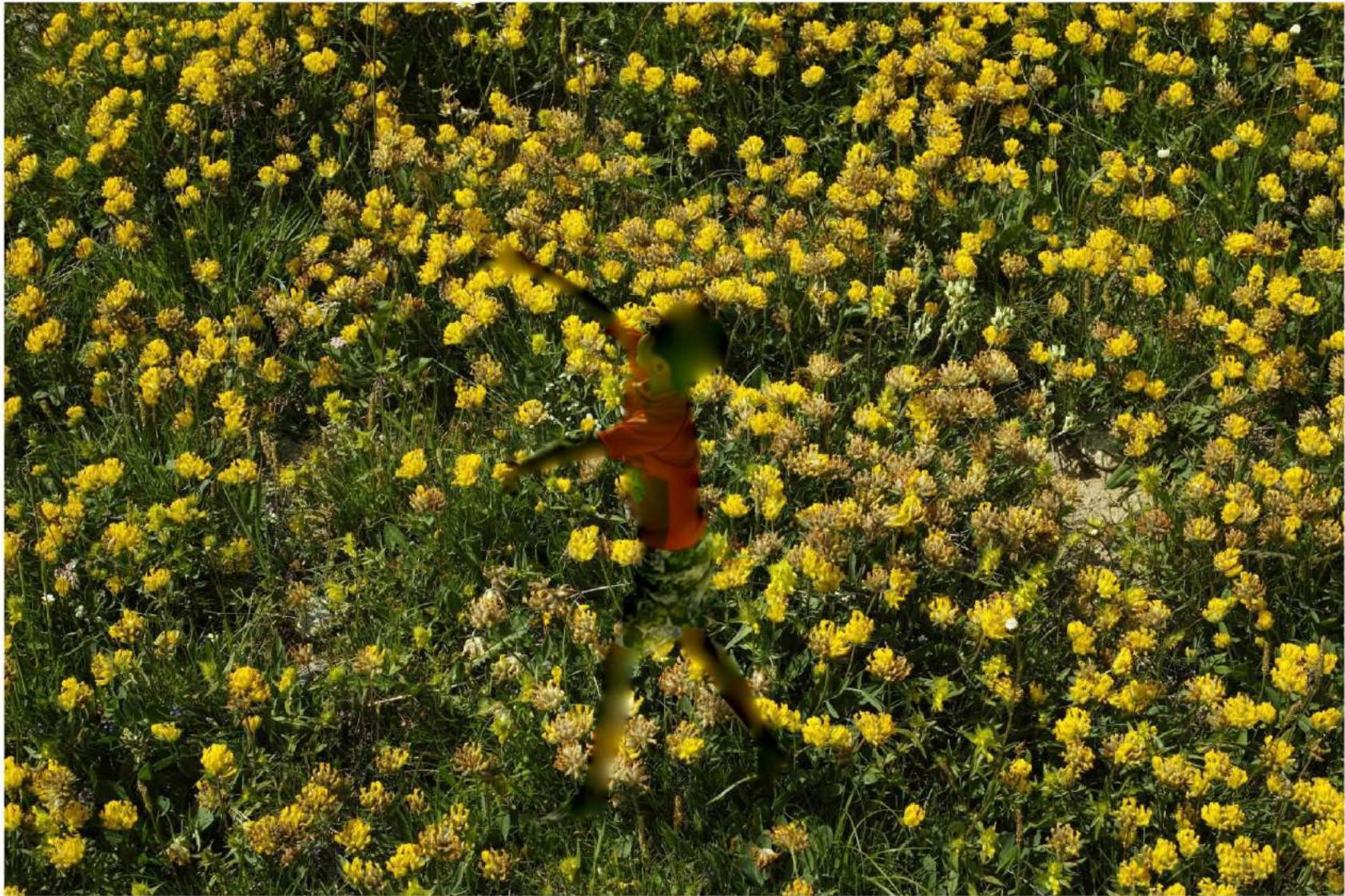
Solving the linear system and
copy the values back to blended image







Where is waldo



Where is waldo



Where is waldo



Where is waldo

