

# **Image Inpainting**

**2019. 11. 25**

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## ■ Image segmentation

- Concept of the clustering (unsupervised)
- Complete your implementation regarding K-means clustering
- SLIC algorithm
  - Spatially-limited K-means clustering
  - Apply SLIC segmentation to the face region

## ■ Image inpainting

### • Reconstruction problem from small damaged images

- Fill out some defected parts in a given image without any additional materials
  - ※ It can be regarded as an ill-posed problem
  - ※ Applications : removing texts and logos, reconstruction from scan, etc.



Original input



Inpainted result

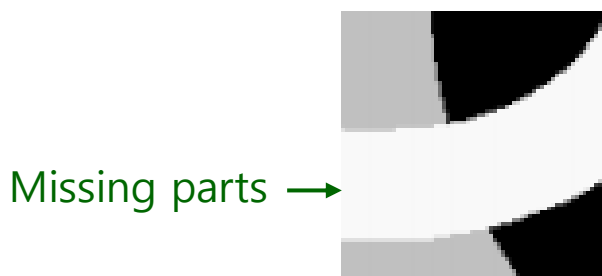
## ■ Algorithm flow

- Most inpainting methods follows the two steps :

- 1) Inpainted regions are selected (manually or automatically)
- 2) Color information is propagated inward from the region boundaries  
i.e., Known image patches are used to fill missing parts

※ To produce a perceptually plausible reconstruction ...

- Continue the *isophotes* as smoothly as possible inside the recon. region



Original input



Which one is more desirable ?

## ■ Two main strategies

### • Partial differential equation (PDE) based approaches

- Color propagation inside the missing region  
while preserving the isophotes' direction (but not practical)  
→ Diffusion for solving PDE, however, yielding the blurring effect !

### • Convolution-based approaches

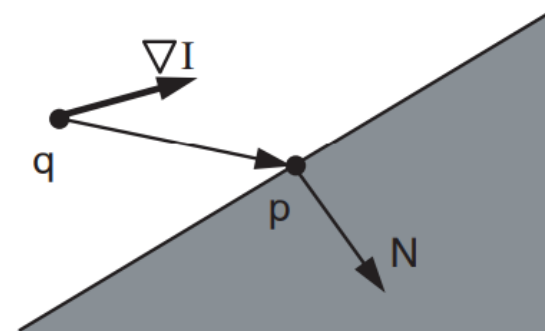
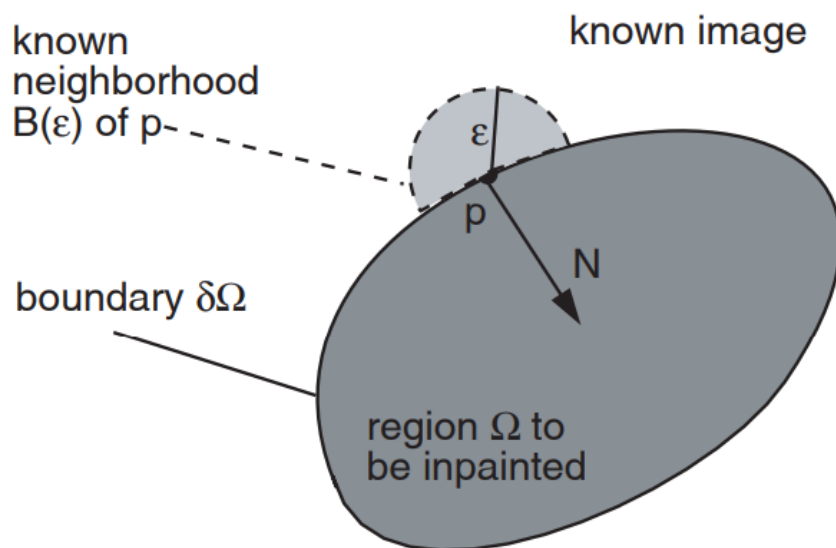
- Repeatedly convolving a simple 3x3 filter over missing regions  
using the known image patches (i.e., other image regions)  
→ No provisions for preserving the isophotes' directions !

※ In this class, a simple and fast method, so-called FMM, is introduced

## ■ Mathematical formulation [1]

### • Propagation with known image patches

- We will inpaint the point  $p$  on the boundary  $\partial\Omega$



[ Inpainting principle ]

1. A. Telea, "An image inpainting technique based on the fast matching method," Journal of Graphics Tools, vol. 9, no. 1, pp. 25-36, 2004.

## ■ Mathematical formulation – cont'd

### • We consider the first order approximation of $p$ using Taylor's series

1) The pixel value of  $p$  can be represented as follows :

$$\begin{aligned} I(p) &= I(q + \Delta x) = I(q) + \nabla I(q) \Delta x + \frac{1}{2!} \Delta x^T \nabla^2 I(q) \Delta x + \dots \\ &\approx I(q) + \nabla I(q) \Delta x \end{aligned}$$

where  $\Delta x$  denotes  $p - q$  (difference of pixel positions)

2) We inpaint  $p$  as a function of all points  $q$  in  $p$ 's neighbor region :

$$I(p) = \frac{\sum_{q \in B_\varepsilon(p)} w(p, q) [I(q) + \nabla I(q)(p - q)]}{\sum_{q \in B_\varepsilon(p)} w(p, q)}$$

where the weighting factor is  $w(p, q) = \text{dir}(p, q) \text{dst}(p, q) \text{lev}(p, q)$

## ■ Detailed procedure

- Iteratively compute  $I(p)$  using the equation in the previous slide

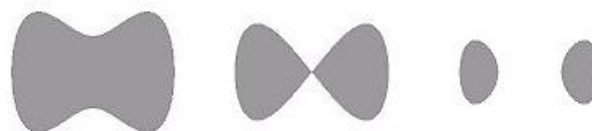
- That is, propagates  $\partial\Omega$  into  $\Omega$  by advancing pixels in order of their distances

➡ Fast marching method solving the Eikonal equation :

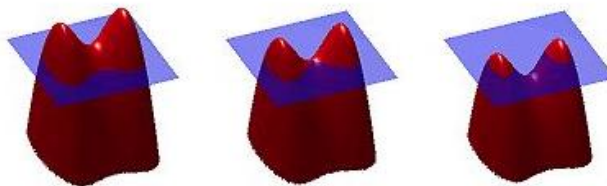
$$|\nabla T| = 1 \text{ on } \Omega \text{ with } T = 0 \text{ on } \partial\Omega$$

※ What is the basic solution for  $T$  in the image processing ?

※ FMM is originally used in the level set computation



: Results in images



: Results in level sets



## ■ Algorithm details – cont'd

### • Weight computation

- A product of three factors :  $w(p, q) = dir(p, q)dst(p, q)lev(p, q)$

where  $dir(p, q) = \frac{p - q}{\|p - q\|} \cdot N(p)$  (inner product)

$$dst(p, q) = \frac{d_0^2}{\|p - q\|^2}$$

$$lev(p, q) = \frac{T_0}{1 + |T(p) - T(q)|}$$

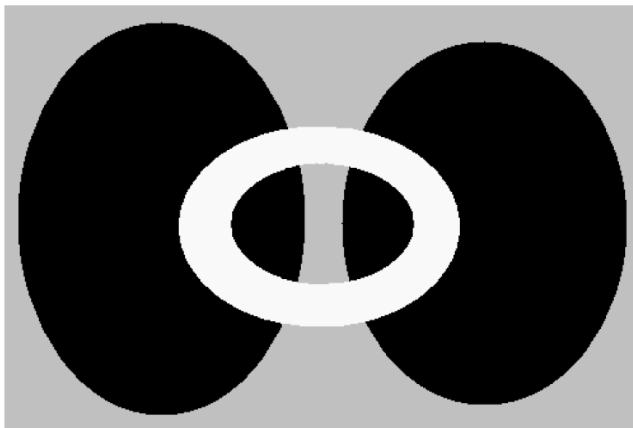
- \* Similar to the direction of normal at p
- \* Close to the point p in a geometric manner
- \* Close to the contour through the point p

**Weight increases !**

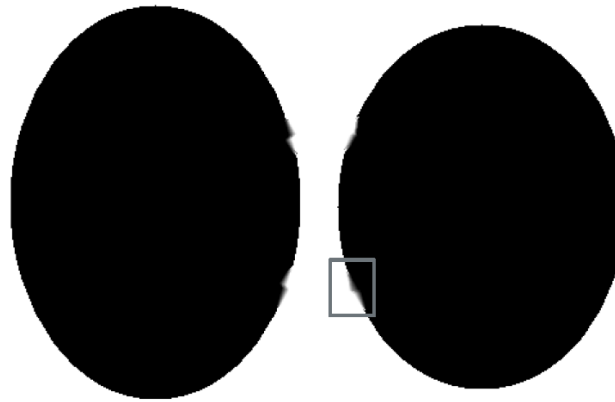
## ■ Algorithm details – cont'd

### • Effect of each weight component

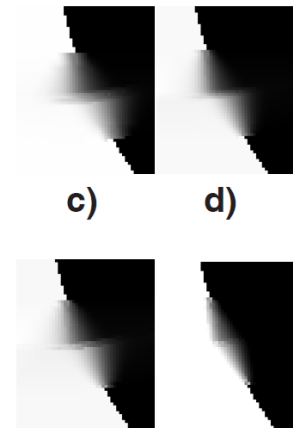
- A product of three factors :  $w(p,q) = dir(p,q)dst(p,q)lev(p,q)$



a)

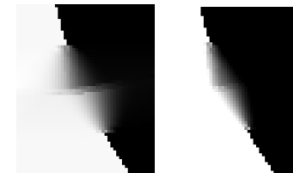


b)



c)

d)



e)

f)

a) Thick region to inpaint      b) Inpainting result of a)

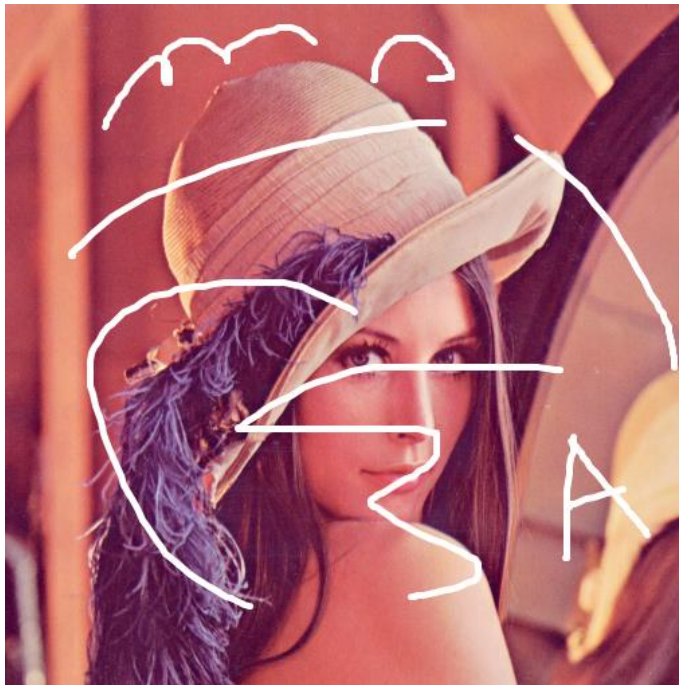
### Effect of weighting functions :

c) direction      d) direction and geometric distance

e) Direction and level set distance      d) direction, geometric, and level set distance

# Examples of Inpainting (1/2)

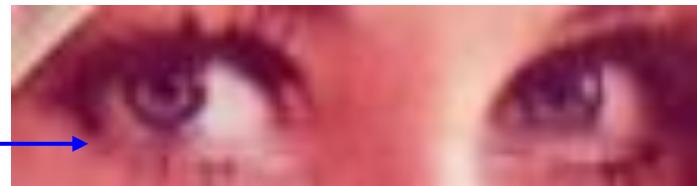
- Examples of segmentation results
  - Weakness : blurring effect when thickness is more than 10~15 pixels



Original input



Blurring artifact →



# Examples of Inpainting (2/2)

## ■ State-of-the-art results

- Patch refinement in MRF framework [2]



Original  
input

Previous approaches

[2]'s result

2. M. Ghorai , S. Mandal, and B. Chanda, "A group-based image inpainting using patch refinement in MRF framework," IEEE Transactions on Image Processing, vol. 27, no. 2, pp. 556-567, Feb. 2018.

## ■ Make a your code for inpainting

- Key header file : `#include <opencv2/photo.hpp>`

```
void inpaint(InputArray src, InputArray inpaintMask, OutputArray dst, double inpaintRadius, int flags)
```

### Parameters:

src : Input 8-bit 1-channel or 3-channel image.

inpaintMask : Inpainting mask, 8-bit 1-channel image.

(Non-zero pixels indicate the area that needs to be inpainted)

dst : Output image with the same size and type as src .

inpaintRadius : Radius of a circular neighborhood of each point inpainted.

flags : Inpainting method that could be one of the following:

INPAINT\_NS Navier-Stokes based method.

INPAINT\_TELEA Method by Alexandru Telea [Telea04].

Make your main function !

# Implementation Task (2/2)

- Inpaint the overlay text in a given scene
  - Try to make your own algorithm to remove the overlay text
    - ⌘ You can use color, corner information (and more...)

## ■ Image inpainting

- It is very useful for various applications
  - Reconstruction from old and damaged images
  - Remove texts and logs in a given scene, etc.
- Fast marching method based inpainting
  - PDE-based solution
- Try to complete implementation