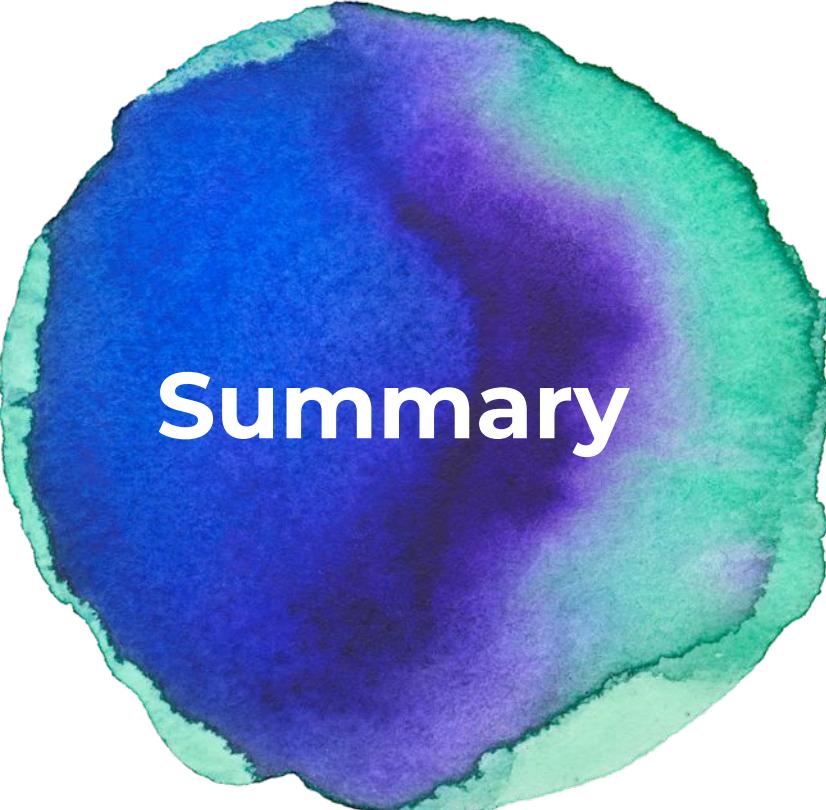
The background of the slide features a large, organic, blue ink or paint splatter centered behind the main title. Smaller blue specks are scattered across the bottom of the slide.

Image Inpainting

Digital signal and image processing Project

Cantarini Giorgio

a.a. 2017/2018



Summary

- 1. Problem**
- 2. Algorithm**
- 3. Results**



1

Problem

What is image inpainting?

Inpainting is the process of **reconstructing** lost or deteriorated parts of images and videos



Goals and applications

Some examples:

- restoration of damaged paintings and photographs
- removal/replacement of selected objects
- film restoration
- removal of red-eye
- removal of the stamped date from photographs
- removal of logos in videos



2

Algorithm

Notation

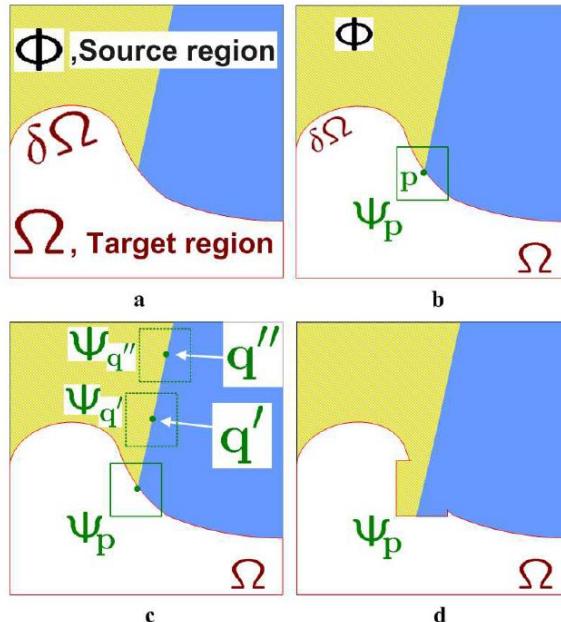


Image I

Region to be inpainted Ω

Source region $(I - \Omega) = \Phi$

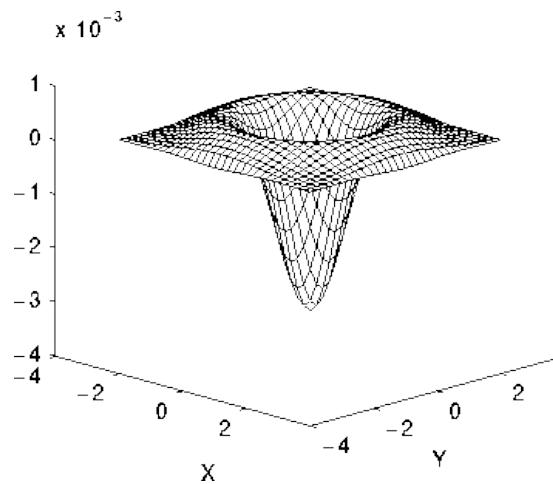
Boundary of target region $\delta\Omega$

The patch centred on the point p Ψ_p

Finding boundary of target region

Convolving the image with Laplacian filter we find edge

1	1	1
1	-8	1
1	1	1



“Region Filling and Object Removal by Exemplar-Based Image Inpainting” A.Criminisi, P. Pérez

Given an input image, the user selects a target region to be removed and filled.

Then the user select the size of the template window.

The algorithm iterates the following **three steps** until all pixels have been filled:

- 1)** Computing patch priorities
- 2)** Propagating texture and structure information
- 3)** Updating confidence values

1) Computing patch priorities

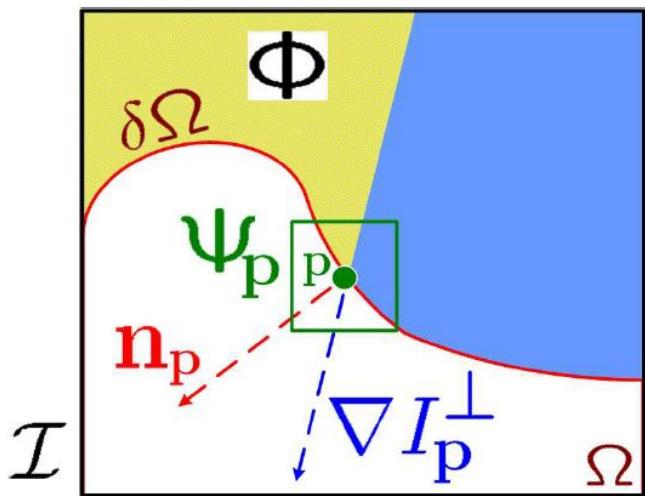
Given a patch Ψ_p centred at the point p for some $p \in \delta\Omega$, we define its priority $P(\mathbf{p})$

$$P(\mathbf{p}) = C(\mathbf{p})D(\mathbf{p}) \quad \text{where}$$

$$C(\mathbf{p}) = \frac{\sum_{\mathbf{q} \in \Psi_p \cap (\mathcal{I} - \Omega)} C(\mathbf{q})}{|\Psi_p|}, \quad D(\mathbf{p}) = \frac{|\nabla I_p^\perp \cdot \mathbf{n}_p|}{\alpha}$$

\mathbf{n}_p is unit vector orthogonal to the front $\delta\Omega$

∇I_p^\perp is the isophote(direction and intensity)



2) Propagating texture and structure information

Once all priorities on the fill front have been computed, the patch Ψ_p with highest priority is found.

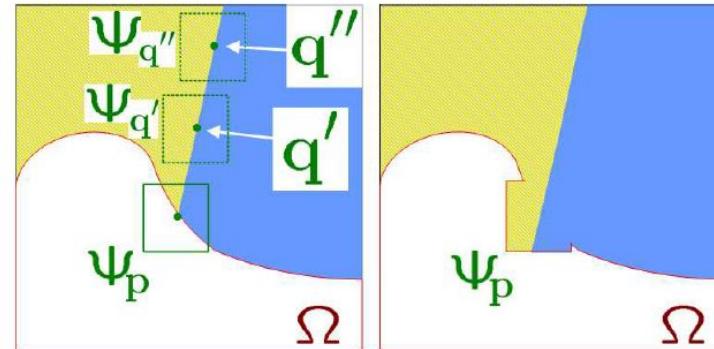
We search in the source region for that patch which is most similar to Ψ_p .

$$\Psi_{\hat{q}} = \arg \min_{\Psi_q \in \Phi} d(\Psi_{\hat{p}}, \Psi_q)$$

2) Propagating texture and structure information

Once found the source exemplar $\Psi_{\hat{q}}$, the value of each pixel-to-be-filled, $p' \in \Psi_{\hat{p} \cap \Omega}$ is copied from its corresponding position inside $\Psi_{\hat{q}}$.

This suffices to achieve the **propagation of both structure and texture information** from the source Φ to the target region Ω , one patch at a time.



3) Updating confidence values

After the patch $\Psi_{\hat{p}}$ has been filled with new pixel values, the confidence $C(p)$ is updated in the area delimited by $\Psi_{\hat{p}}$ as follows:

$$C(p) = C(\hat{p}) \quad \forall p \in \Psi_{\hat{p}} \cap \Omega$$

As filling proceeds, confidence values decay, indicating that we are less sure of the colour values of pixels near the centre of the target region.

Pseudocode

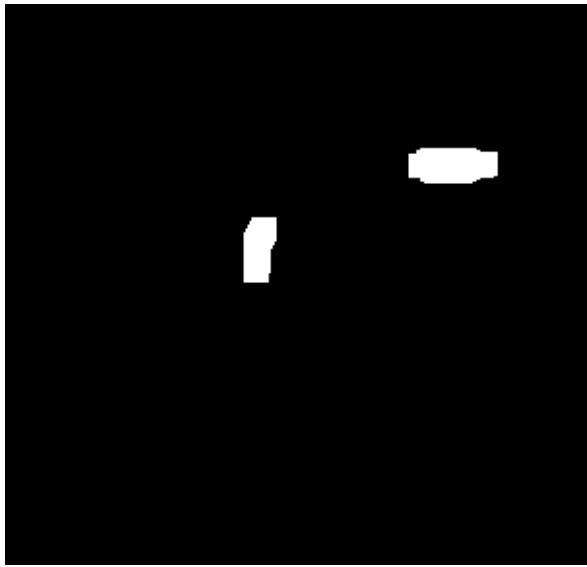
- Extract the manually selected initial front $\delta\Omega^0$.
- Repeat until done:
 - 1a.** Identify the fill front $\delta\Omega^t$. If $\Omega^t = \emptyset$, exit.
 - 1b.** Compute priorities $P(p) \quad \forall p \in \delta\Omega^t$.
 - 2a.** Find the patch $\Psi_{\hat{p}}$ with the maximum priority,
i.e., $\hat{p} = \arg \max_{p \in \delta\Omega^t} P(p)$.
 - 2b.** Find the exemplar $\Psi_{\hat{q}} \in \Phi$ that minimizes $d(\Psi_{\hat{p}}, \Psi_{\hat{q}})$.
 - 2c.** Copy image data from $\Psi_{\hat{q}}$ to $\Psi_{\hat{p}} \quad \forall p \in \Psi_{\hat{p}} \cap \Omega$.
 - 3.** Update $C(p) \quad \forall p \in \Psi_{\hat{p}} \cap \Omega$



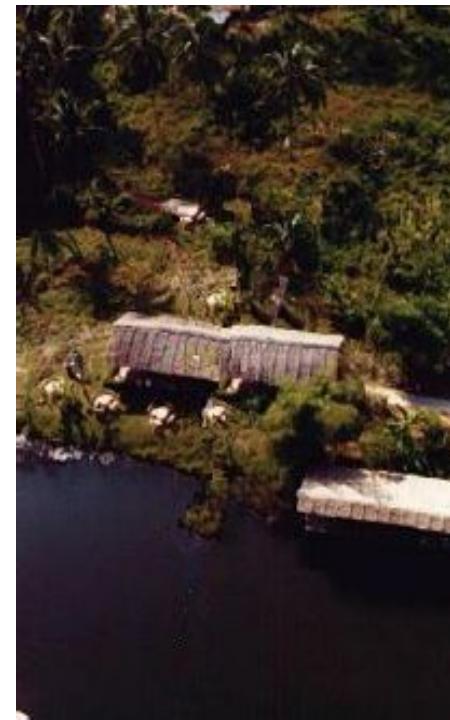
3

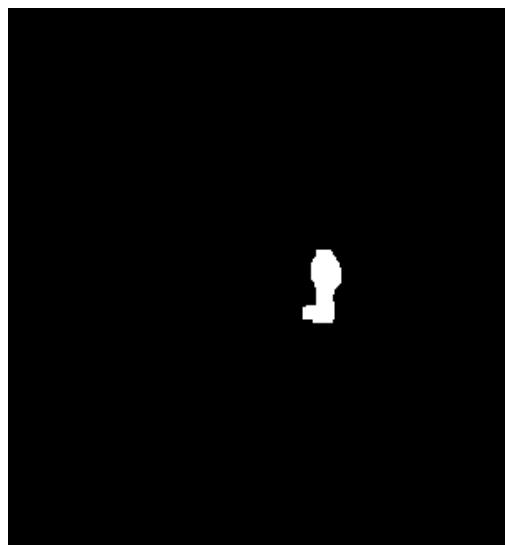
Results

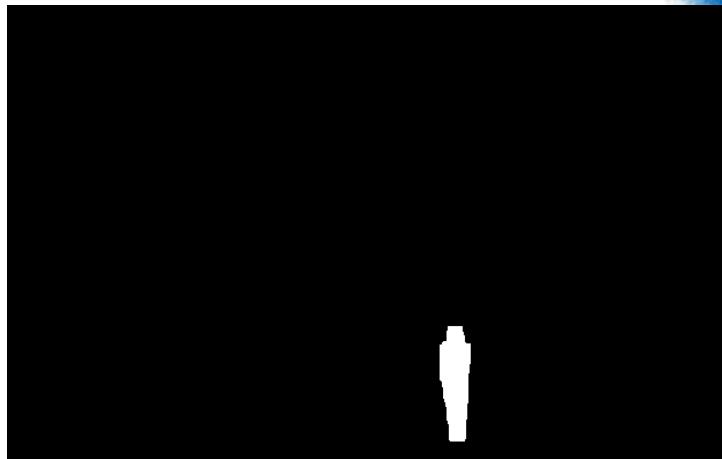












Thanks!

Any questions?

