

Image Compression by Partial Differential Equations

(Séminaire Doctorants IRIMAS)

Thomas Jacumin

sous la direction de Zakaria Belhachmi

Goal : reconstruct missing parts of an image $f : D \rightarrow [0, 1]$ from $K \subset D$.

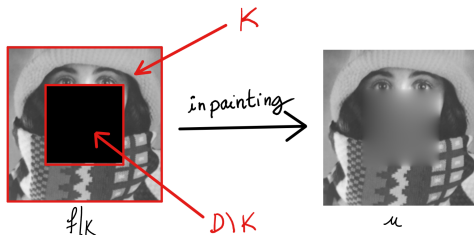


Figure – Inpainting

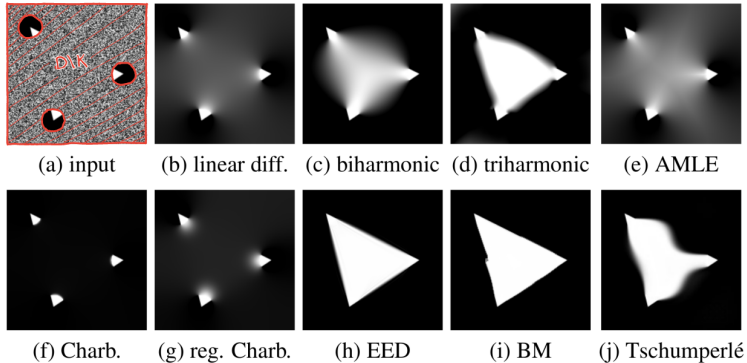


Figure – Examples [2]

JPEG

Each block 8×8 of an image is described as a function

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For compression purpose, we can neglect small $c_{u,v}$.

We do not do this : we remove parts of the image and reconstruct them by inpainting :

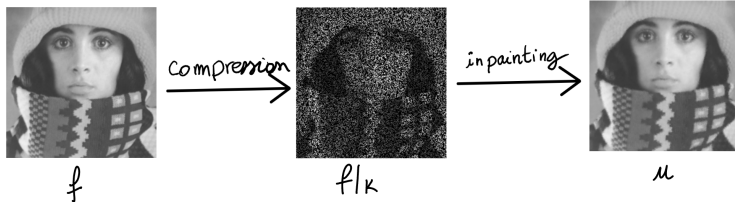


Figure – Compression by Inpainting.

Question 1

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Questions 2-3

What is a good reconstruction? How to find these pixels?

Let $D \subset \mathbb{R}^2$ the support of an image $f : D \rightarrow [0, 1]$. Some errors examples

Example 1

$$\|u - f\|_{L^2(D)} := \left(\int_D (u - f)^2 dx \right)^{1/2}.$$

Example 2

$$\|u - f\|_{H^1(D)} := \left(\int_D |\nabla u - \nabla f|^2 dx \right)^{1/2}.$$

Shape Optimisation Problem

$$\min_{K \subseteq D} \{ \mathcal{E}(u_K) \mid u_K \text{ solution of an inpainting problem} \},$$

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Compression problem

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Compression problem

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The existence of a solution depends on the inpainting problem, on the error \mathcal{E} and on m .

Problem (Homogeneous Diffusion Inpainting [1])

Find u in $H^1(D)$ such that

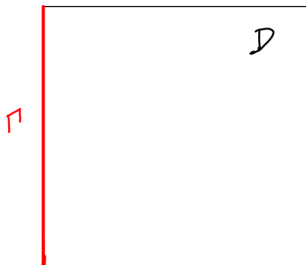
$$\begin{cases} -\Delta u = 0, & \text{in } D \setminus K, \\ u = f, & \text{in } K, \\ \frac{\partial u}{\partial n} = 0, & \text{on } \partial D. \end{cases} \quad (1)$$

Problem (Heat equation)

Find u such that

$$\begin{cases} \partial_t u - \Delta u = 0, & \text{in } [0, +\infty[\times D, \\ u = 100, & \text{on } [0, +\infty[\times \Gamma, \\ \partial_n u = 0, & \text{on } [0, +\infty[\times \partial D \setminus \Gamma. \end{cases} \quad (2)$$

and $u(0, \cdot) = 0$ in D .



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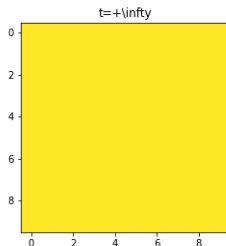
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Problem (Heat equation ($t \rightarrow +\infty$))

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and $u(0, \cdot) = 0$ in D .



Theorem

If the inpainting is the homogeneous diffusion inpainting

$$\mathcal{E}(u) = |u - f|_{H^1(D)} := \left(\int_D |\nabla u - \nabla f|^2 dx \right)^{1/2},$$

and if

$$m(K) = \text{cap}(K) := \inf \left\{ \int_D |\nabla u|^2 dx + \int_D u^2 dx \mid \right. \\ \left. u \in H_0^1(D), u \geq 1 \text{ p.p. dans } K \right\},$$

Then, the compression problem admits at least one solution.

Question

In practice, how to construct K ?

Topological Gradient

Goal : Determine the influence of make a hole in K on the error.
Let $K_\varepsilon := K \setminus B(x_0, \varepsilon)$ such that $B(x_0, \varepsilon) \subset K$.

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$$j : A \subset D \mapsto \min_{v \in H^1(D), v=f \text{ in } A} \int_D |\nabla v|^2 dx.$$

Proposition

The compression problem is equivalent to $\max_{K \subset D, m(K) \leq c} j(K)$.

Topological Gradient

$$\begin{aligned} \max_{K \subset D, m(K) \leq c} j(K) &= \max_{K \subset D, m(K) \leq c} \min_{v \in H^1(D), v=f \text{ in } K} \int_D |\nabla v|^2 dx \\ &= \max_{K \subset D, m(K) \leq c} \left(\underbrace{\left(\min_{v \in H^1(D), v=f \text{ in } K} \int_{D \setminus K} |\nabla v|^2 dx \right)}_{\geq 0} + \int_K |\nabla f|^2 dx \right). \end{aligned}$$

Then,

$$\max_{K \subset D, m(K) \leq c} \int_K |\nabla f|^2 dx \leq \max_{K \subset D, m(K) \leq c} j(K).$$

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When ε tends to 0,

$$j(K_\varepsilon) - j(K) = -|\Delta f(x_0)|^2 \frac{\pi}{2} \varepsilon^4 + o(\varepsilon^4).$$

Topological Gradient



Figure – f et $|\Delta f|$.

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Answer

Keep points x_0 where the quantity $|\Delta f(x_0)|$ is maximal.

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Fat Pixels

For $m > 0$ and $n \in \mathbb{N}^*$, we set

$$\mathcal{A}_{m,n} := \left\{ \bigcup_{i=1}^n \overline{B(x_i, r)} \mid x_i \in D, r = mn^{-1/2} \right\}.$$

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Answer

Built K such that the pixels density increase with the quantity $|\Delta f|$.

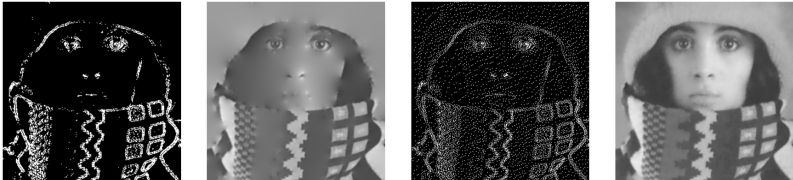


Figure – Topological Gradient vs Fat Pixels.

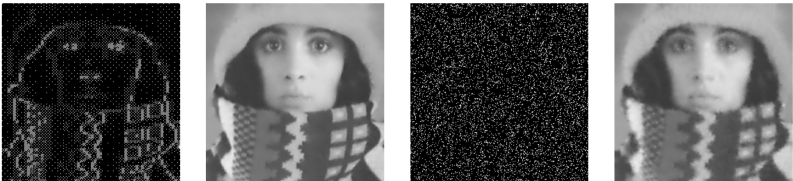


Figure – B-Tree vs Random.



Figure – Image with Gaussian Noise.

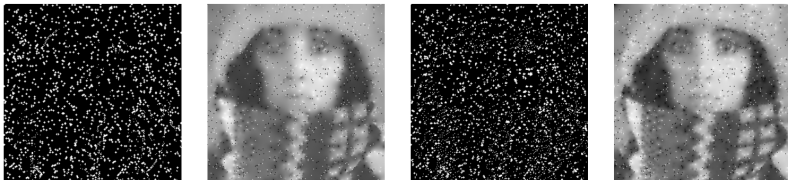


Figure – Image with Salt and Pepper Noise.

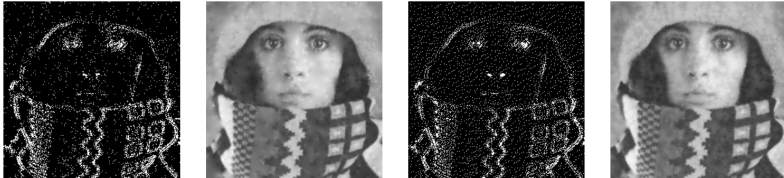


Figure – Image with Gaussian Noise.

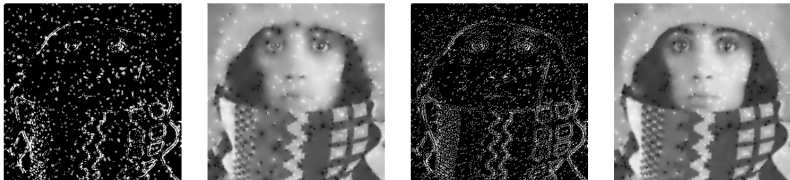


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Thanks for your attention !



Z. Belhachmi, D. Bucur, B. Burgeth, and J. Weickert, *How to choose interpolation data in images*, 70, pp. 333–352.



C. Schmaltz, P. Peter, M. Mainberger, F. Huth, J. Weickert, and A. Bruhn, *Understanding, optimising, and extending data compression with anisotropic diffusion*, 108.