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Another triumph for De Giorgi's Γ -convergence

Heim Brezis, Pisa Sept. 2016

PLAN

1. Some background.
2. A suggestive computation.
Pointwise convergence of $\Lambda_\delta(u)$
3. Γ -convergence of Λ_δ .
4. Variational problems with roots
in Image processing.

1. Some background

Starting point comes from a paper Bourgain - Brezis - Mironescu, 2001.

Let $\Omega \subset \mathbb{R}^N$ smooth bounded, $N \geq 1$ (possibly $\Omega = \mathbb{R}^N$).

Let (ρ_ε) be a family of radial mollifiers i.e., $\rho_\varepsilon \in L^1_{loc}(0, \infty; \mathbb{R})$, (no smoothness required)

$$\rho_\varepsilon \geq 0 \text{ on } (0, \infty),$$
$$\int_0^\infty \rho_\varepsilon(r) r^{N-1} dr = 1,$$

and $\rho_\varepsilon \rightarrow \delta_0$, i.e. $\int_0^a \rho_\varepsilon(r) r^{N-1} dr \xrightarrow{\varepsilon \rightarrow 0} 1, \forall a > 0.$

Given $u \in L^1(\Omega)$ set

$$J_\varepsilon(u) = \int_\Omega \int_\Omega \frac{|u(x) - u(y)|}{|x - y|} \rho_\varepsilon(|x - y|) dx dy$$

Thm (BBM 2001, refinement by J. Danielli 2002)

Given $u \in L^1(\Omega)$, then

$$u \in BV(\Omega) \iff \liminf_{\varepsilon_n \rightarrow 0} J_{\varepsilon_n}(u) < \infty.$$

In that case

$$\lim_{\varepsilon \rightarrow 0} J_\varepsilon(u) = K_N \int_\Omega |Du|, \forall u \in BV(\Omega),$$

where $K_N = 2|B^{N-1}| = \int_{S^{N-1}} |\sigma \cdot e| d\sigma \quad \forall e \in S^{N-1}.$

Mode of convergence is robust.

A. Ponce (2004) proved that

$$J_\varepsilon(\cdot) \xrightarrow{\varepsilon \rightarrow 0} K_N \|\cdot\|_{BV}$$

in the sense of Γ -convergence in L^1 .

Typical examples

Ex 1 $\rho_\varepsilon(z) = \frac{N}{\varepsilon^N} \mathbb{1}_{(0, \varepsilon)}$ gives

$$\frac{1}{\varepsilon^N} \iint_{\substack{\Omega \times \Omega \\ |x-y| < \varepsilon}} \frac{|u(x) - u(y)|}{|x-y|} dx dy \rightarrow \frac{K_N}{N} \int_{\Omega} |\nabla u| \quad \forall u \in BV$$

Ex 2 $\rho_\varepsilon(z) = \frac{\varepsilon}{z^{N-\varepsilon}} \mathbb{1}_{(0, \varepsilon)}$ gives

$$\varepsilon \iint_{\Omega \times \Omega} \frac{|u(x) - u(y)|}{|x-y|^{N+1-\varepsilon}} dx dy \rightarrow K_N \int_{\Omega} |\nabla u| \quad \forall u \in BV$$

In particular if $u = \mathbb{1}_E$, then $\varepsilon \int_E \int_E \frac{dx dy}{|x-y|^{N+1-\varepsilon}} \rightarrow \frac{K_N \text{Per} E}{2}$

Appears under the name "Nonlocal perimeter" in papers by Ambrosio - De Philippis - Martinazzi (2011), Caffarelli - Valdinoci (2011), Fusco (2015) etc..

Around 2002 I played (for fun!) with another type of nonlocal functionals and I discovered that they also converge to the BV-norm.

Given $u \in L^1(\Omega)$, $\delta > 0$, set

$$L_\delta(u) = \delta \iint_{\substack{\Omega \times \Omega \\ |u(x) - u(y)| > \delta}} \frac{1}{|x-y|^{N+1}} dx dy.$$

Note that J_ε and L_δ are quite different nonlocal functionals

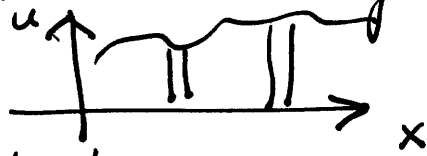
- J_ε is convex in u

- L_δ is not convex in u

(and this will be a source of tremendous difficulties).

Comment (J.M. Morel):

- J_ε is reminiscent of Riemann integral:
"slicing in x "



- L_δ is reminiscent of Lebesgue integral:
"slicing in u "



Fact (H.B. 2003). $\forall u \in C^1(\bar{\Omega})$

$$\lim_{\delta \rightarrow 0} L_\delta(u) = K_N \int_{\Omega} |\nabla u|.$$

See later a slightly more general version with a sketch of proof.

["A suggestive computation"].

In view of BBM it was natural to ask

Question (H.B. 2004) Assume $u \in L^1(\Omega)$.

Is it true that

$$u \in BV(\Omega) \iff \liminf_{\delta_n \rightarrow 0} L_{\delta_n}(u) < \infty ?$$

If so, is it true that

$$\lim_{\delta \rightarrow 0} L_\delta(u) = K_N \int_{\Omega} |\nabla u| \quad \forall u \in BV(\Omega)?$$

This question generated an abundance of results - both positive and negative!

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Pathology 1 (Ponce, 2005)

$$\exists u \in W^{1,1}(\Omega) \text{ s.t. } \lim_{\delta \rightarrow 0} L_{\delta}(u) = +\infty.$$

Hence \Rightarrow in my question has a negative answer.

How about \Leftarrow ? Turned out to be hard, but true!

Thm (Bourgain - H.M. Nguyen, 2006)

Let $u \in L^1(\Omega)$ be such that

$$\liminf_{\delta_n \rightarrow 0} L_{\delta_n}(u) < \infty$$

then $u \in BV(\Omega)$.

Moreover

$$\int_{\Omega} |Du| \leq C \liminf_{\delta_n \rightarrow 0} L_{\delta_n}(u).$$

For the second component of my question

$$\liminf_{\delta_n \rightarrow 0} L_{\delta_n}(u) < \infty \stackrel{?}{\implies} \lim_{\delta \rightarrow 0} L_{\delta}(u) = k_N \int_{\Omega} |Du|$$

the answer is ~~also~~ negative. The construction of a counterexample is new (B - Nguyen, 2016). See later Pathology 2.

In view of this "erratic" behavior of pointwise convergence of $L_\delta(u)$ [Pointwise convergence means analysis of the asymptotic behavior of $L_\delta(u)$ as $\delta \rightarrow 0$ for fixed u]

I suggested to study $\lim_{\delta \rightarrow 0} L_\delta$ in the sense of Π -convergence.

Question (H. B. 2006) Is it true that

$$L_\delta(\cdot) \xrightarrow{\delta \rightarrow 0} K_N |\cdot|_{BV}$$

in the sense of Π convergence in L^1 ?

Thm (Nguyen, Duke 2011)

$$L_\delta \xrightarrow{\delta \rightarrow 0} L_0$$

in the sense of Π -convergence in L^1 , where

$$L_0(u) = \begin{cases} K'_N \int |Du| & \text{if } u \in BV \\ +\infty & \text{otherwise} \end{cases}$$

for some constant $0 < K'_N < K_N$ (!!!)

Rem This result is much stronger than Bourgain-Nguyen, but some ingredients from the proof of BN enter here.

A very lucky coincidence happened in 2011. At the Technion I met (casually) colleagues from the Computer Science Dept. involved in Image Processing (F. Bruckstein, R. Kimmel). They mentioned to me that nonlocal functionals of the form

$$\iint_{\Omega} \int_{\Omega} \varphi\left(\frac{|u(x)-u(y)|}{\delta}\right) w(|x-y|) dx dy$$

were popular ~~in~~ Image Processing.

[$\delta > 0$ small parameter, $\varphi(t)$ function to be discussed later, $w(|x-y|)$ weight function].

They also told that J.M. Morel is a leading expert ... J.M. Morel had been my PhD student, but we had lost contact. Quickly reconnected ... This information caught my attention because $\frac{1}{\delta}$ has the form above with

$$\varphi(t) = \begin{cases} 0 & 0 < t \leq 1 \\ 1 & t > 1 \end{cases}$$

$$w(|x-y|) = \frac{1}{|x-y|^{N+1}}$$

I proposed to Nguyen that we study together such functionals.

Outcome is a long paper Aug. 2016

now on Arxiv.

[worked on/off 2012 - 2016].

Longest section concerns Γ -convergence
(~30 p.).

Non-local functionals related to the total variation and connections with Image Processing

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Abstract

We present new results concerning the approximation of the total variation, $\int_{\Omega} |\nabla u|$, of a function u by non-local, non-convex functionals of the form

$$\Lambda_{\delta} u = \int_{\Omega} \int_{\Omega} \frac{\delta \varphi(|u(x) - u(y)|/\delta)}{|x - y|^{d+1}} dx dy,$$

as $\delta \rightarrow 0$, where Ω is a domain in \mathbb{R}^d and $\varphi : [0, +\infty) \rightarrow [0, +\infty)$ is a non-decreasing function satisfying some appropriate conditions. The mode of convergence is extremely delicate and numerous problems remain open. De Giorgi's concept of Gamma-convergence illuminates the situation, but also introduces mysterious novelties. The original motivation of our work comes from Image Processing.

Key words: Total variation, bounded variation, non-local functional, non-convex functional, Gamma-convergence, Sobolev spaces.

Mathematics Subject Classification: 49, 26B30, 46E35, 28A75.

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1 Introduction

Throughout this paper, we assume that $\varphi : [0, +\infty) \rightarrow [0, +\infty)$ is continuous on $[0, +\infty)$ except at a finite number of points in $(0, +\infty)$ where it admits a limit from the left and from the right. We also assume that $\varphi(0) = 0$ and that $\varphi(t) = \min\{\varphi(t+), \varphi(t-)\}$ for all $t > 0$, so that φ is lower semi-continuous. We assume that the domain $\Omega \subset \mathbb{R}^d$ is either bounded and smooth, or that $\Omega = \mathbb{R}^d$. The case $d = 1$ is already of great interest; many difficulties (and open problems!) occur even when $d = 1$.

Given a measurable function u on Ω , and a small parameter $\delta > 0$, we define the following non-local functionals:

$$\Lambda(u) := \int_{\Omega} \int_{\Omega} \frac{\varphi(|u(x) - u(y)|)}{|x - y|^{d+1}} dx dy \quad \text{and} \quad \Lambda_\delta(u) := \delta \Lambda(u/\delta). \quad (1.1)$$

Sometimes, it is convenient to be more specific and to write $\Lambda_\delta(u, \varphi, \Omega)$ or $\Lambda_\delta(u, \Omega)$ instead of $\Lambda_\delta(u)$.

2. A suggestive computation. Pointwise convergence of $\Lambda_\delta(u)$.

Let $\Omega \subset \mathbb{R}^N$ smooth bounded, $N \geq 1$.
The case $N=1$ is already very interesting
[still many problems open when $N=1$].
The case $N=2$ is important because
it appears in Image Processing.

The basic notations / assumptions :

Fix a function $\varphi : [0, \infty) \rightarrow [0, \infty)$.

Given $u \in L^1(\Omega)$ consider the nonlocal functional

$$\Lambda(u) = \iint_{\Omega \times \Omega} \varphi(|u(x) - u(y)|) \frac{1}{|x-y|^{N+1}} dx dy < \infty$$

and the scaled version

$$\Lambda_\delta(u) = \delta \iint_{\Omega \times \Omega} \varphi\left(\frac{|u(x) - u(y)|}{\delta}\right) \frac{1}{|x-y|^{N+1}} dx dy = \delta \Lambda\left(\frac{u}{\delta}\right),$$

with $\delta > 0$ small parameter.

Class A consists of all functions $\varphi : [0, \infty) \rightarrow [0, \infty)$ satisfying:

- (1) φ is nondecreasing on $[0, \infty)$.
- (2) φ is continuous on $[0, \infty)$ except at a finite number of points where $\varphi(t) = \varphi(t-)$.
- (3) $\varphi(t) \leq at^2 \quad \forall t \in [0, 1]$ and some $a > 0$
- (4) $\varphi(t) \leq b \quad \forall t \geq 0$ and some $b > 0$
- (5) $K_N \int_0^\infty \frac{\varphi(t)}{t^2} dt = 1$.

Assumption (5) is a normalization condition - its role will become clear later.

Typical examples are

$$\varphi_1(t) = C_1 \tilde{\varphi}_1(t)$$

$$\varphi_2(t) = C_2 \tilde{\varphi}_2(t)$$

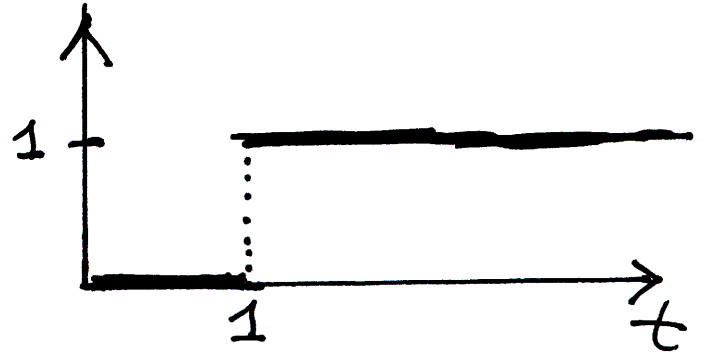
$$\varphi_3(t) = C_3 \tilde{\varphi}_3(t)$$

where $\tilde{\varphi}_1, \tilde{\varphi}_2, \tilde{\varphi}_3$ are defined below and C_1, C_2, C_3

are adjusted so that $\varphi_1, \varphi_2, \varphi_3$ satisfy (5).

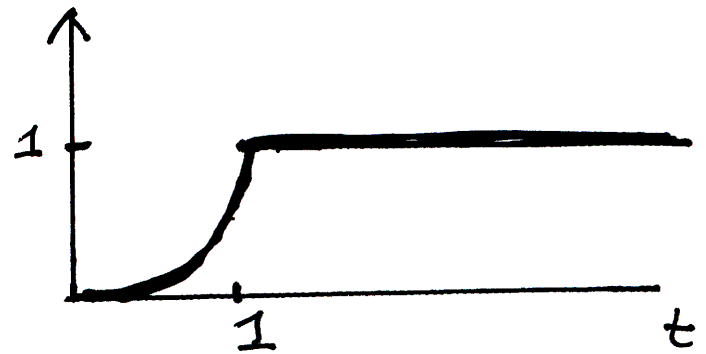
Ex 1

$$\tilde{\varphi}_1(t) = \begin{cases} 0 & t \leq 1 \\ 1 & t > 1 \end{cases}$$



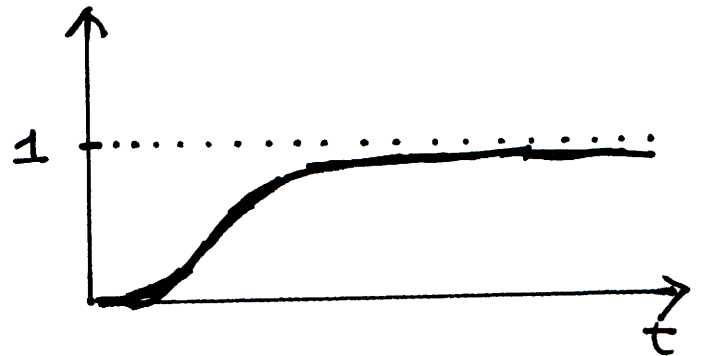
Ex 2

$$\tilde{\varphi}_2(t) = \begin{cases} t^2 & t \leq 1 \\ 1 & t > 1 \end{cases}$$



Ex 3

$$\tilde{\varphi}_3(t) = 1 - e^{-t^2}$$



$\tilde{\varphi}_1$ is the case studied by Nguyen.

$\tilde{\varphi}_2, \tilde{\varphi}_3$ occur in Image Processing.

Prop Assume $\varphi \in \mathcal{Q}$. Then

$$(6) \quad \lim_{\delta \rightarrow 0} \Lambda_\delta(u) = \int_{\Omega} |\nabla u| \quad \forall u \in C^1(\bar{\Omega}).$$

[Normalization condition enters to have 1 here]

More generally

$$(7) \quad \lim_{\delta \rightarrow 0} \Lambda_\delta(u) = \int_{\Omega} |\nabla u| \quad \text{if } u \in W^{1,p}(\Omega) \text{ for some } 1 < p < \infty.$$

(8) If $u \in W^{1,1}(\Omega)$ one can only assert that

$$\liminf_{\delta \rightarrow 0} \Lambda_\delta(u) \geq \int_{\Omega} |\nabla u|.$$

Proof of (6) is fairly easy (see below).

Proof of (7) is much more complicated.

(Uses theory of maximal functions).

Proof of (8) relies on Fatou.

Sketch of the proof of (6)
 "A suggestive computation".

$$\Lambda_\delta(u) = \delta \iint_{\substack{\dots \\ |x-y| < \delta^\alpha \\ = A_\delta}} \dots + \delta \iint_{\substack{\dots \\ |x-y| \geq \delta^\alpha \\ = B_\delta}}$$

where $0 < \alpha < \frac{1}{N+1}$ is fixed.

Clearly $B_\delta \rightarrow 0$ as $\delta \rightarrow 0$.

Changing variables $y \mapsto x+h$

$$A_\delta = \delta \iint_{|h| < \delta^\alpha} \varphi\left(\frac{|u(x+h) - u(x)|}{\delta}\right) \frac{1}{|h|^{N+1}} dx dh.$$

Taylor's expansion (OK since $|h| < \delta^\alpha \rightarrow 0$)

$$A_\delta \approx \delta \iint_{|h| < \delta^\alpha} \varphi\left(\frac{|h \cdot \nabla u(x)|}{\delta}\right) \frac{1}{|h|^{N+1}} dx dh.$$

[Cheating but this can be fixed]

Use polar coordinates $r = |h|$ $\sigma = \frac{h}{|h|} \in S^{N-1}$

Then

$$A_\delta \approx \delta \int_{\Omega} dx \int_{S^{N-1}} d\sigma \int_0^{\delta^\alpha} \varphi\left(\frac{r |\sigma \cdot \nabla u(x)|}{\delta}\right) \frac{1}{r^2} dr$$

Change of variable $r \mapsto s = \frac{r |\sigma \cdot \nabla u(x)|}{\delta}$

$$A_\delta \approx \int_{\Omega} dx \int_{S^{N-1}} d\sigma \int_0^{\frac{|\sigma \cdot \nabla u(x)|}{\delta^{1-\alpha}}} \frac{\varphi(s)}{s^2} |\sigma \cdot \nabla u(x)| ds,$$

As $\delta \rightarrow 0$,

$$A_\delta \rightarrow \int_{\Omega} dx \int_{S^{N-1}} |\sigma \cdot \nabla u(x)| d\sigma \int_0^\infty \frac{\varphi(s)}{s^2} ds.$$

But $\int_{S^{N-1}} |\sigma \cdot \nabla u(x)| d\sigma = |\nabla u(x)| \int_{S^{N-1}} |\sigma \cdot e| d\sigma$
 $e \in S^{N-1}$
 $= K_N |\nabla u(x)|.$

From the normalization condition (5)

we have

$$A_\delta \xrightarrow{\delta \rightarrow 0} \int_{\Omega} |\nabla u(x)| dx.$$

Some pathologies

Pathology 1 (similar to Ponce, 2005)

$$\forall \varphi \in \mathcal{A} \quad \exists u \in W^{1,1}(\Omega) \text{ s.t.}$$

$$\lim_{\delta \rightarrow 0} \Lambda_{\delta}(u) = +\infty$$

Pathology 2 $\forall \varphi \in \mathcal{A}, \exists u \in W^{1,1}(\Omega) \text{ s.t.}$

$$\limsup_{\delta \rightarrow 0} \Lambda_{\delta}(u) = +\infty$$

$$\liminf_{\delta \rightarrow 0} \Lambda_{\delta}(u) = \int_{\Omega} |\nabla u|$$

Pathology 3 $\forall \varphi \in \mathcal{A}, \exists u \in BV(\Omega) (u \notin W^{1,1})$

s.t.

$$\liminf_{\delta \rightarrow 0} \Lambda_{\delta}(u) < \int_{\Omega} |\nabla u|.$$

Seems hopeless to have a satisfactory theory of pointwise convergence for Λ_{δ} .

Γ -convergence saves the situation!!

3. Γ -convergence of Λ_δ

The main result:

THEOREM 1 $\forall \varphi \in \mathcal{Q}, \exists$ constant $k = k(\varphi)$

(k is independent of Ω) s.t.

(9) $0 < k \leq 1,$

and

(10) $\Lambda_\delta \xrightarrow{\delta \rightarrow 0} \Lambda_0$ in the sense of Γ -convergence in $L^1(\Omega)$

where

$$\Lambda_0(u) = \begin{cases} k \int_{\Omega} |\nabla u| & \text{if } u \in BV(\Omega), \\ +\infty & \text{if } u \in L^1(\Omega), u \notin BV(\Omega). \end{cases}$$

Recall definition of Γ -convergence

(G1) $\forall (u_\delta) \subset L^1(\Omega)$ s.t. $u_\delta \xrightarrow{\delta \rightarrow 0} u$ in $L^1(\Omega)$

we have

$$\liminf_{\delta \rightarrow 0} \Lambda_\delta(u_\delta) \geq \Lambda_0(u).$$

(G2) $\forall u \in L^1(\Omega), \exists (\tilde{u}_\delta) \subset L^1(\Omega)$ s.t.

$$\tilde{u}_\delta \rightarrow u \text{ in } L^1(\Omega)$$

and

$$\limsup_{\delta \rightarrow 0} \Lambda_\delta(\tilde{u}_\delta) \leq \Lambda_0(u).$$

Comments

- 1) Proof of Thm 1 is long and intricate.
Total ≈ 50 p. [30 p. in new paper
+ 20 p. of results coming from Nguyen, Duke 2011]
Please help to simplify it !!
- 2) The fact that $k > 0$ is already nontrivial.
A special consequence is a stronger version
of the question I raised in 2004 :

Cor Assume $(u_{\delta_n}) \subset L^1(\Omega)$ and $u \in L^1(\Omega)$ satisfy

$$u_{\delta_n} \rightarrow u \quad \text{in } L^1(\Omega) \text{ as } \delta_n \rightarrow 0,$$

and

$$\liminf_{\delta_n \rightarrow 0} \Lambda_{\delta_n}(u_n) < \infty$$

Then

$$u \in BV(\Omega) \text{ and } \int_{\Omega} |\nabla u| \leq \frac{1}{k} \liminf_{\delta_n \rightarrow 0} \Lambda_{\delta_n}(u_n).$$

My original question was for $u_{\delta_n} \equiv u \in L^1$
and $\varphi = \varphi_1$. Settled positively by
Bourgain - Nguyen (2006).

3) The occurrence of the constant $0 < k \leq 1$ in Thm 1 is still mysterious!

Not clear why it may happen that $k(\varphi) < 1$ for some φ 's (possibly every $\varphi \in \mathcal{A}$?)

Open problem Is it true that
 $k(\varphi) < 1 \quad \forall \varphi \in \mathcal{A}$

Open even for $N=1$, $\varphi = \varphi_2$ or $\varphi = \varphi_3$.

The existence of k is established via a "semi-explicit" construction (see later) which yields very little information about k . In fact the exact value of k is not known even when $N=1$, $\varphi = \varphi_1$

At this time, all we know is

Prop 2 (Nguyen, Duke, 2011). For any $N \geq 1$

$$k(\varphi_1) < 1.$$

Sketch of proof (bizarre!)

Take e.g. $N=1$, $\Omega = (0,1)$, $\varphi = \varphi_1$

Set $U(x) = x$, so that by Prop. 1

$$\lim_{\delta \rightarrow 0} \Lambda_\delta(U) = \int_0^1 |U'| = 1.$$

A brute force construction yields a family (U_δ) s.t.

$$U_\delta \rightarrow U \quad \text{in } L^1 \text{ (even in } L^\infty)$$

and

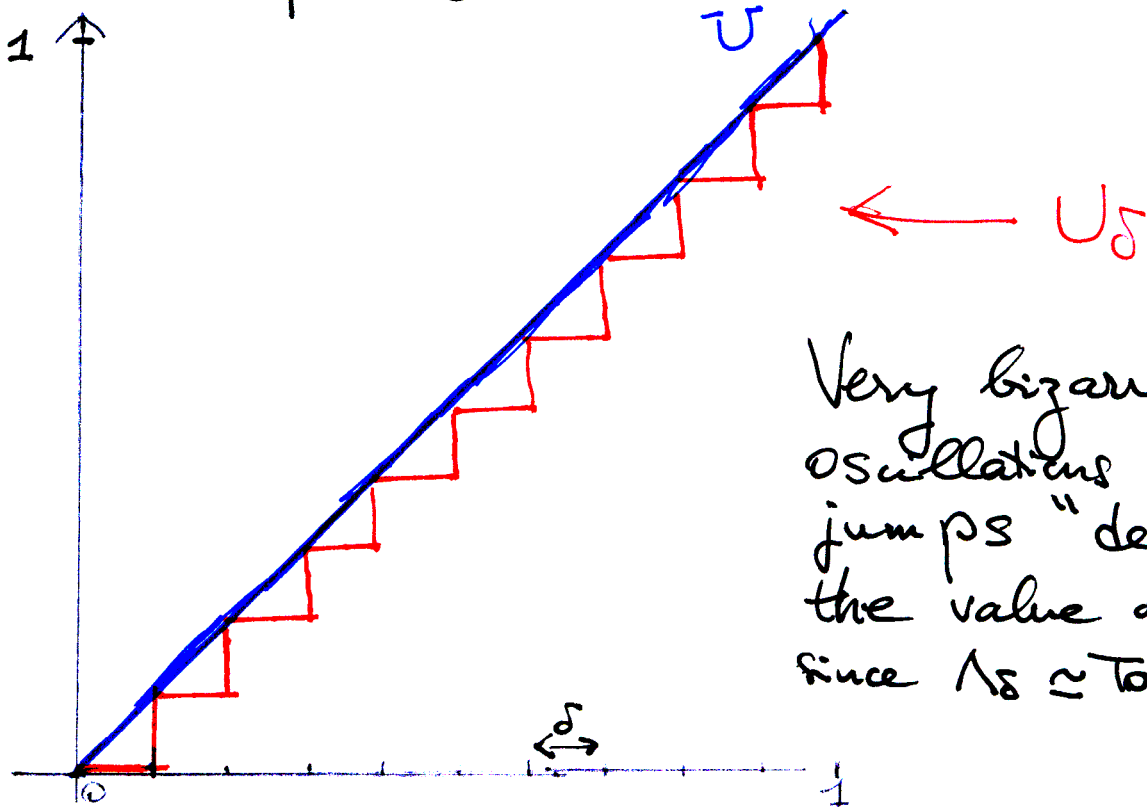
$$\liminf_{\delta \rightarrow 0} \Lambda_\delta(U_\delta) < 1$$

From (G1) we have

$$\underbrace{k \int |U'|}_{=1} \leq \liminf_{\delta \rightarrow 0} \Lambda_\delta(U_\delta) < 1$$

and thus $k < 1$.

Construction of U_δ



Very bizarre that oscillations and jumps "decrease" the value of Λ_δ since $\Lambda_\delta \approx$ Total variation

So far this argument works only for $\varphi = \varphi_1$.
 We are unable to adapt it to φ_2 or φ_3 .
 It might well be that $k(\varphi_2) = 1$ or $k(\varphi_3) = 1$?

Structure of the proof of Thm 1.

Take $\Omega = Q =$ unit cube in \mathbb{R}^N .

Fix a special simple function U :

$$U(x_1, x_2, \dots, x_N) = \frac{1}{\sqrt{N}} (x_1 + x_2 + \dots + x_N)$$

so that $\int_Q |\nabla U| = 1$.

Define

$$k(\varphi) = \inf \left\{ \liminf_{\delta \rightarrow 0} \Lambda_\delta(u_\delta); u_\delta \xrightarrow{\delta \rightarrow 0} U \text{ in } L^1 \right\}.$$

↑
 taken over all sequence u_δ s.t. $u_\delta \rightarrow U$ in L^1

Prove that (G.1) and (G.2) in Thm 1 hold ~~with~~ with this k !

4. Variational problems with roots in Image Processing

A fundamental challenge in Image Processing is to improve images of poor quality.

Denoising is an immense subject. Excellent survey by Buades-Coll-Morel SIAM Review 2010.

A prevailing strategy is to use a variational formulation associated with a filter F . Given an image $f(x)$ of poor quality, consider

$$(11) \quad \min_u \left\{ \lambda \int_{\Omega} |f-u|^2 + F(u) \right\}$$

where $\Omega \subset \mathbb{R}^2$ (we'll work with $\Omega \subset \mathbb{R}^N$)

$\lambda > 0$ is the fidelity parameter (at our disposal). F is an appropriate functional; many filters are being tested - still an empirical science!

The minimizers of (11) are the denoised images.

Two popular filters:

A) Rudin - Osher - Fatemi (ROF, 1992)

$$F(u) = \int_{\Omega} |\nabla u|$$

$$(ROF) \quad \min_u \left\{ \lambda \int_{\Omega} |f - u|^2 + \int_{\Omega} |\nabla u| \right\}$$

Convex functional. $\forall f \in L^2$ Unique
-minimizer $u_0 \in BV \cap L^2$.

B) Neighborhood filters

(Lee, Yaroslavsky, 1985, SUSAN filter, 1997
Bilateral filter, 2008, Non-local means, 2011 ...)

uses
$$F(u) \approx \int_{\Omega} \int_{\Omega} \varphi\left(\frac{|u(x) - u(y)|}{\delta}\right) w(|x - y|) dx dy$$

where $\delta > 0$ is a small parameter,

$\varphi = \varphi_3$ for example and $w(\cdot)$ is
weight function (many choices have been
used).

problem (2.2) for a given value of λ . The Euler-Lagrange equation associated with the minimization problem is given by

$$(u(\mathbf{x}) - v(\mathbf{x})) - \frac{1}{2\lambda} \text{curv}(u)(\mathbf{x}) = 0$$

(see [78]). Thus, we have the following theorem.

THEOREM 2.5. *The method noise of the total variation minimization (2.2) is*

$$u(\mathbf{x}) - TVF_\lambda(u)(\mathbf{x}) = -\frac{1}{2\lambda} \text{curv}(TVF_\lambda(u))(\mathbf{x}).$$

As in the anisotropic case, straight edges are maintained because of their small curvature. However, details and texture can be oversmoothed if λ is too small, as is shown in Figure 2.

2.4. Neighborhood Filters. The previous filters are based on a notion of spatial neighborhood or proximity. Neighborhood filters instead take into account grayscale values to define neighboring pixels. In the simplest and more extreme case, the denoised value at pixel i is an average of values at pixels which have a grayscale value close to $u(i)$. The grayscale neighborhood is therefore

$$B(i, h) = \{j \in I \mid u(i) - h < u(j) < u(i) + h\}.$$

This is a fully nonlocal algorithm, since pixels belonging to the whole image are used for the estimation at pixel i . This algorithm can be written in a more continuous form,

$$NF_h u(\mathbf{x}) = \frac{1}{C(\mathbf{x})} \int_\Omega u(\mathbf{y}) e^{-\frac{|u(\mathbf{y}) - u(\mathbf{x})|^2}{h^2}} d\mathbf{y},$$



(at the level of the Euler eq. $2\lambda(u-f) + F'(u) = 0$)

where $\Omega \subset \mathbb{R}^2$ is an open and bounded set, and $C(\mathbf{x}) = \int_\Omega e^{-\frac{|u(\mathbf{y}) - u(\mathbf{x})|^2}{h^2}} d\mathbf{y}$ is the normalization factor.

The Yaroslavsky neighborhood filters [97, 95] consider mixed neighborhoods $B(i, h) \cap B_\rho(i)$, where $B_\rho(i)$ is a ball of center i and radius ρ . So the method takes an average of the values of pixels which are both close in grayscale and spatial distance. This filter can be easily written in a continuous form as

$$YNF_{h,\rho}(\mathbf{x}) = \frac{1}{C(\mathbf{x})} \int_{B_\rho(\mathbf{x})} u(\mathbf{y}) e^{-\frac{|u(\mathbf{y}) - u(\mathbf{x})|^2}{h^2}} d\mathbf{y},$$

where $C(\mathbf{x}) = \int_{B_\rho(\mathbf{x})} e^{-\frac{|u(\mathbf{y}) - u(\mathbf{x})|^2}{h^2}} d\mathbf{y}$ is the normalization factor. More recent versions, namely, the *SUSAN filter* [83] and the *bilateral filter* [86], weigh the distance to the reference pixel \mathbf{x} instead of considering a fixed spatial neighborhood,

In the next theorem we compute the asymptotic expansion of the Yaroslavsky neighborhood filter when $\rho, h \rightarrow 0$.

THEOREM 2.6. *Suppose $u \in C^2(\Omega)$, and let $\rho, h, \alpha > 0$ such that $\rho, h \rightarrow 0$ and $h = O(\rho^\alpha)$. Let us consider the continuous function \tilde{g} defined by $\tilde{g}(t) = \frac{1}{3} \frac{te^{-t^2}}{E(t)}$, for $t \neq 0$, $\tilde{g}(0) = \frac{1}{6}$, where $E(t) = 2 \int_0^t e^{-s^2} ds$. Let \tilde{f} be the continuous function defined by $\tilde{f}(t) = 3\tilde{g}(t) + \frac{3\tilde{g}(t)}{t^2} - \frac{1}{2t^2}$, $\tilde{f}(0) = \frac{1}{6}$. Then, for $\mathbf{x} \in \Omega$,*

1. *if $\alpha < 1$, $YNF_{h,\rho}u(\mathbf{x}) - u(\mathbf{x}) \simeq \frac{\Delta u(\mathbf{x})}{6} \rho^2$;*

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We suggest to use

$$\lambda \approx \frac{1}{\delta}, \quad \varphi \in \mathcal{A}, \quad w(|x-y|) = \frac{1}{|x-y|^{N+1}}$$

Then denoising mechanism becomes

$$(NF)_\delta \quad \text{Min}_{u \in L^2} \left\{ \int_{\Omega} |f-u|^2 + \Lambda_\delta(u) \right\}.$$

THEOREM 2 Assume $\varphi \in \mathcal{A}$ and also

$$(11) \quad \varphi(t) > 0 \quad \forall t > 0 \quad [\varphi = \varphi_2 \text{ or } \varphi_3 \text{ OK}]$$

Then, given any $f \in L^2$

(i) $\forall \delta > 0 \exists u_\delta$ minimizer of $(NF)_\delta$
(but no uniqueness)

(ii) As $\delta \rightarrow 0$, $u_\delta \rightarrow u_0$ in L^2 where u_0 is the unique minimizer of

$$(ROF)_R \quad \text{Min}_{u \in L^2} \left\{ \int_{\Omega} |f-u|^2 + R \int_{\Omega} |Du| \right\}$$

and $R > 0$ is the constant which comes from the Γ -convergence Thm 1

Comments

1) The convergence part (ii) is totally new and came as a surprise to the experts in Image Processing!!
Concretely they still prefer to use $(NF)_\delta$ with $\delta > 0$ small (and not $\delta \rightarrow 0$) but it is gratifying to know that $(NF)_\delta \rightarrow (ROF)$ as $\delta \rightarrow 0$.

[Compare with various convergence results in Fluid Mechanics. Boltzmann, Navier-Stokes, Euler..]

Part (ii) does not require assumption (11); it suffices to assume that u_δ is an "almost minimizer".

2) The existence of a minimizer - part (i) - for $(NF)_\delta$ with fixed $\delta > 0$ is also totally new. [But it is of limited interest in Image Processing because people are satisfied with almost minimizers].

Recall that Λ_δ is not convex and standard tools of Functional Analysis cannot be applied. Use instead new compactness results.

Basic ingredients for the proof of Thm 2

Two compactness lemmas:

- For fixed $\delta > 0$.

- For $\delta \rightarrow 0$.

Lemma 1 Assume $\varphi \in \mathcal{A}$ and (11).

For every fixed $\delta > 0$ (e.g. $\delta = 1$) the set

$$\left\{ u \in L^1(\Omega); \|u\|_{L^1} \leq C \text{ and } \Lambda_\delta(u) \leq C \right\}$$

is compact in $L^1(\Omega)$.

Lemma 1 is used to prove part (i) in Thm 2.

Lemma 2 Assume $\varphi \in \mathcal{A}$. Let (u_{δ_n}) be a sequence such that $\delta_n \rightarrow 0$,

$$\|u_{\delta_n}\|_{L^1} \leq C \text{ and } \Lambda_{\delta_n}(u_{\delta_n}) \leq C.$$

Then (u_{δ_n}) is relatively compact in $L^1(\Omega)$.

Part (ii) in Thm 2 is an immediate consequence of Thm 1 (Γ -convergence of Λ_δ) and Lemma 2.

The proofs of Lemmas 1 and 2 rely on some new unusual inequality due to H.M. Nguyen (CV PDE, 2011) with roots in Bourgain-Nguyen (2006).

Let $B_1 =$ unit ball in \mathbb{R}^N .

Lemma 3 $\exists C_N$ such that $\forall u \in L^1(B_1)$

$$\iint_{B_1, B_1} |u(x) - u(y)| dx dy \leq C_N \left[1 + \iint_{\substack{B_1, B_1 \\ [|u(x) - u(y)| > 1]}} \frac{1}{|x - y|^{N+1}} dx dy \right].$$

Proof of Lemma 3 is non-trivial. Better proof?

Better inequality? e.g.

$$\iint_{\substack{B_1, B_1 \\ [|u(x) - u(y)| > 1]}} |u(x) - u(y)| dx dy \leq C_N \iint_{\substack{B_1, B_1 \\ [|u(x) - u(y)| > 1]}} \frac{1}{|x - y|^{N+1}} dx dy ?$$