

## PCMI Lecture #14

Pedagogical Lecture: TV regularization and sets of finite perimeter as a path into geometric measure theory.

These last two lectures will look at how to integrate teaching of research centered around geometric analysis and shapes/images. First though, I will go into my philosophy in these ~~math~~ matters, to give you all a full context for the next two lectures.

### Philosophy and Perspective

The teaching of mathematics is usually discipline focused. We teach analysis, ring theory, algebraic geometry, harmonic analysis, combinatorics, ..., so on and so forth. Eventually some fraction of us will study a particular equation, some fixed mathematical object, some chosen system of inequalities... but not all, since some continue to study structure and develop tools.

What is rarely done is the teaching of mathematics through a focus on a particular equation, a particular question.

While it is proper and sometimes necessary to focus on tools, disciplines, background, I am going to suggest that much more be done through a focus on specific objects, equations, functionals. In my opinion having a balance, a point-counterpoint relation, between theory and tool based exposition and particular object/equation exploratives will bring a wholistic sense, an awareness of context and an automatic motivation to the ~~learning~~ learning.

PDE's  
is often  
taught with  
an equation  
focus

In the next two days I will follow a somewhat meandering path through subjects we have already been exposed to in this course, illustrating that a problem/equation based approach can give a very natural motivation and context to theory that is thus encountered or inspired.

This kind of teaching takes on a character much more akin to research, learning what we need as we go... or inventing it if we have to, or are inspired to do so.

Naturally allied with this style is the mode of active learning that is the inspiration behind the Moore method. (No need to get distracted by the excesses of the Moore ~~the~~ method). Simply understanding the original inspiration for R.L. Moore's method - R.L. covering up proofs to see how much he could invent on his own, therefore kicking himself out of passive, into ~~an active~~ an active frame of mind <sup>↑</sup> spirit - is enough to set us on the right path.

There are many ways to move the student into the active mode, each of which is right for some and not for others... our system has too long focused on one option: the exercises in the back of the chapter, or sometimes interspersed throughout the chapter. ~~but~~ (of course I exaggerate, but not too much)

But there are many other ways to trigger the active mode:

Examples: (nothing very new here)

Students create exercises based on the reading / theory.

Students come up with multiple examples illustrating theorems / Definitions (counter-examples too)

Students encouraged to extend / generalize ideas in course.

Students find papers relating to subject at hand and read and explain them to the class.

Students write code implementing ideas, theory.

Students write expository papers explaining what they are learning

Students give a very large latitude in the problems they solve ... say 20 out of 100 must be chosen.

etc, etc, etc ...

My claim is that the problem / research mode of teaching makes this diversity of activity natural, not artificially imposed.

The last piece of this active problem/research style of teaching is the lab environment.

## Lab environment

- open everything, open access, open source, creative commons...
- fast prototyping via scripting languages  
Matlab, Python, Octave, Sage, ...
- Place: Blackboards, Notebooks, Computers (Laptops)  
Wikipages (public & private), e-journals
- Teams: 2-3
- Jam sessions: Regular, inclusive not too long formally, open ended
- Workshops: 2-3 Days immersive, problem focused
- playfulness: a very big deal.

- \* problem/research style of teaching, point-by-point compliant to ~~the~~ theory
- \* Quasi-Moore method: principle = active mode ~~passive mode~~
- \* Lab environment: Bare-handed work with problems, code, ideas.

With that, here we go

- (A) ROF: the original paper to use total variation regularization in image analysis for the denoising problem.

After having students read the paper, I would guide them through the following explanation



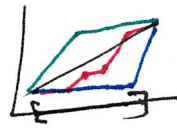
$\int |\nabla u|$  ... what does this mean

AD

i)  $C^1 u \rightarrow$  not so hard

- calculate for some easy functions  
~~monotone with fixed B values have equal  $\int |\nabla u|$~~

ii) Lipschitz  $u \rightarrow$  more interesting since we can now not get distracted by nonzero second derivatives



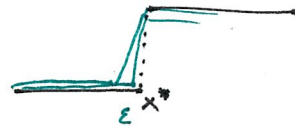
monotone with fixed B values have equal  $\int |\nabla u|$

what are you tempted to conjecture? can you prove it?

what about the fact that  $\nabla u$  not defined?  
 (case suggests these points don't matter if  $u$  is continuous)

connects: question how many points can a Lip function be discontinuous at?

iii) discontinuous  $u$



$\epsilon \rightarrow 0$

this suggests  $\nabla u = \delta(x-x^*)$

$\rightarrow$  what does this really mean?

Bring up the fact that it is often the case that an equation is used before it is well understood!!

iv) introduce weak definition  
$$\int |\nabla u| \phi x = \sup_{|\phi| \leq 1} \int \nabla u \cdot \phi$$

$$\int_{\Omega} \nabla u \cdot \phi = - \int_{\Omega} u (\nabla \phi) \quad \text{compactly supported } \phi$$

$$\Rightarrow \int |\nabla u| = - \sup_{\substack{|\phi| \leq 1 \\ \phi \in C^1}} \int u (\nabla \phi) dx$$

now define  $\int |\nabla u|$  by the RHS when  $u$  not  $C^1$ .  
upshot it all makes sense even if  $u$  is discontinuous.

This is a general pattern seen over and over in analysis. To extend some idea, some calculus beyond where it should work, do something that is defined in that "outer regime" that is ~~equivalent~~ equivalent for nice cases.

v) To understand

$$\nabla u = \vec{\sigma} d\mu$$

we need to understand

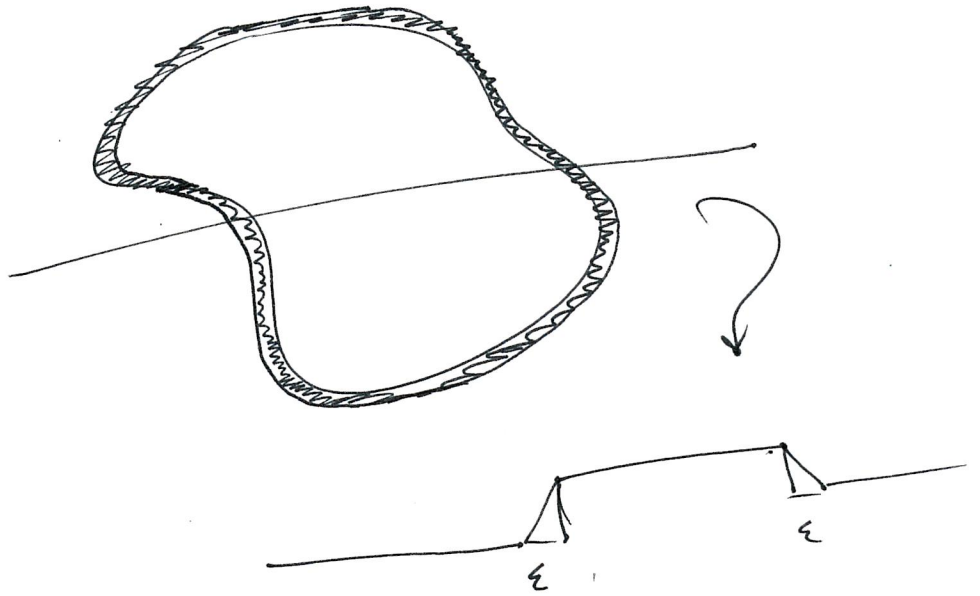
outer measure  
measurable  
regular  
Borel  
Borel regular  
Radon

... lots of theory, very well motivated...

I am suggesting lots of studied general questions and answers along the way.

(6)

vi) more: 2D  $\int |\nabla \chi_{\Omega}| = \text{Per}(\Omega)$



- Q: how wiggly can  $\partial\Omega$  be and still  
 have finite length?  $\Rightarrow$  intro to Rademacher's Theorem.  
 $\Rightarrow$  Rectified boundary,  $\Rightarrow$  struct. thm. Cauchy's sol.
- Q: what other ways might we find  $\text{Per}(\Omega)$ ?  
 what about  $\frac{\text{Vol}(\Omega + \epsilon B) - \text{Vol}(\Omega)}{\epsilon}$ ?

~~~~~~~~~

vii) BV coarea  $\int |\nabla u| = \int H'(u=y) dy$   
 is 2D picture !!

$\Downarrow$

usual coarea formula (works for Lipschitz)

$\Downarrow$

Fubini: special case of coarea

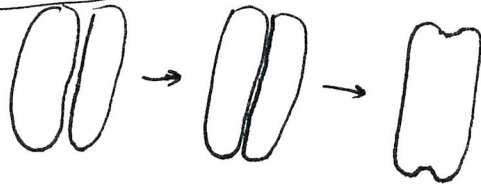
... a transition now

- (B) Q: Can you find a sequence of shapes  $\Omega_i \xrightarrow{L'} \Omega$   
 such that  $\text{Per}(\Omega_i) \rightarrow L > \text{Per}(\Omega)$ ?  
 Q: how about " " "  $L < \text{Per}(\Omega)$ ?

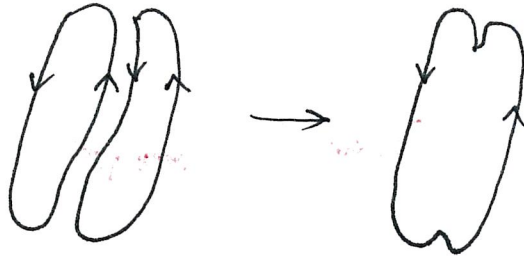
(7)

$\Rightarrow$  lower semicontinuity for BV.

$$\int f \operatorname{div} g = \lim_{i \rightarrow \infty} \int f_i \operatorname{div} g \leq \liminf_{i \rightarrow \infty} \int |Df_i|$$

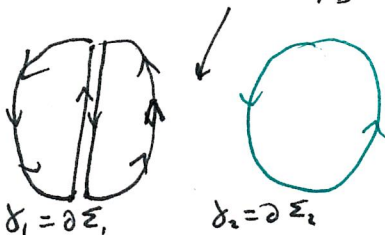


Boundaries make the natural



... somehow it seems natural that close, oppositely oriented boundaries should cancel.

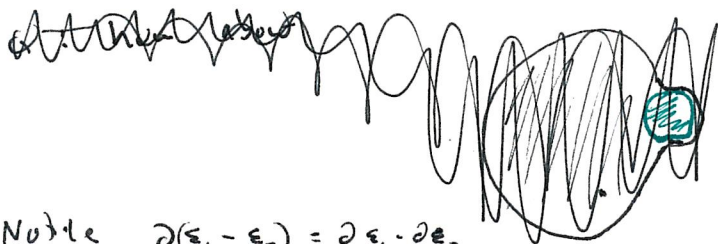
Q: how can we do this ... construct a distance that says these shapes are close?



Q: how about comparing the shapes that these are boundaries at?



distance = area of  $(\Sigma_1 \Delta \Sigma_2)$



Note  $\partial(\Sigma_1 - \Sigma_2) = \partial \Sigma_1 - \partial \Sigma_2$  is the boundary of a ~~small~~ small area... key observation.

8



Inspiration: measure size of  $\gamma_1 - \gamma_2 = \partial(\varepsilon_1 - \varepsilon_2)$   
 by area of  $\varepsilon_1 - \varepsilon_2$ . that is, for  
 boundaries  $T$  (in our case  $T = \partial(\varepsilon_1 - \varepsilon_2)$ )  
 where  $T = \partial S$

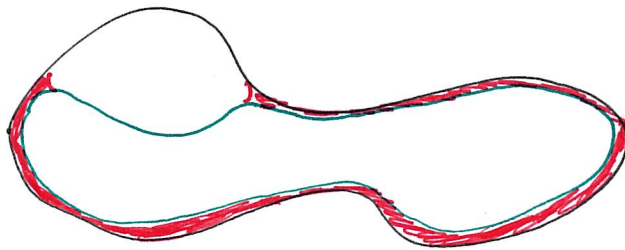
$$F(T) = \mathcal{H}^2(S)$$

Q: what if we want to apply this to  $T \neq \partial S$   
 for any  $S$ ? well, try

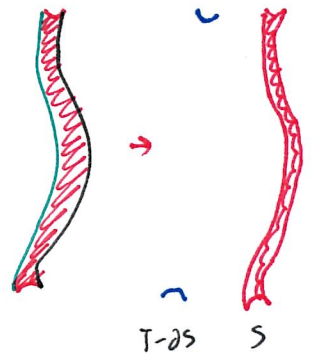
$$F(T) = \underbrace{\mathcal{H}^1(T - \partial S)}_{\text{length of miss}} + \underbrace{\mathcal{H}^2(S)}_{\text{area at approx diff.}}$$

... minimize over  $S$ .

→ Binsu: Flat norm... generalized to any closed  
 not-just boundaries.



Even if  $T = \partial S$  that might  
 not be optimal decomposition



... move on to more boundaries next time.