

Weighted Energy Dissipation method for
gradient flows - Savaré. lecture 1.

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Gradient flow X ambient space, $\Psi: X \rightarrow (-\infty, +\infty]$.
functional.

In specific example: $X = \mathbb{R}^n$, $\Psi \text{ ec'}$, then G.F. is solving.

$$\begin{cases} \dot{u}(t) + D\Psi(u(t)) = 0 \\ u(0) = u_0 \end{cases}$$

$\circ X = H$, Hilbert space, Ψ is convex (\Leftrightarrow 1-convex). f. lower s.c.
 $D\Psi \rightsquigarrow \partial\Psi$ (multi-valued, sub-differential).

$$\begin{cases} \dot{u}(t) + \partial\Psi(u(t)) = 0 \\ u(0) = u_0 \end{cases}$$

Bogus, Crandall, Kruw, Pazy. '70.

$\circ X = \mathbb{R}^n$ space, Ψ is convex + c'; Ψ is diff of convex functions } requires
refined techniques

$\circ X = \text{metric space.}$

De Giorgi, Marino, Sosio, Tosques, De Giacomo ~ '90.

Anthonio-Gigli-S.

Banach Space. $\partial\Psi(u(t)) \subset X'$ (dual \Rightarrow in finite dim or Hilbert,
identify X' with H).

$\rightarrow J(u(t)) + \partial\Psi(u(t)) = 0$. (Belli; Visintin, Colliv.).
dual inclusion: $\frac{1}{2}\|u\|_X^2$.

①

(X, d) complete metric space, $\varphi: X \rightarrow (-\infty, +\infty]$ l.s.c.

In smooth situation,

$$\begin{aligned} \dot{u}(t) + D\varphi(u(t)) = 0 &\Leftrightarrow \frac{d}{dt}\varphi(u(t)) = -|\dot{u}(t)|^2 = -|D\varphi(u(t))|^2 \\ \text{vector eq}^n. & \Leftrightarrow \underbrace{\frac{d}{dt}\varphi(u(t)) + \frac{1}{2}(|\dot{u}(t)|^2 + |D\varphi(u(t))|^2)}_{\text{scalar eq}^n} \leq 0 \end{aligned}$$

Chain rule:

$$\frac{d}{dt}\varphi(u(t)) = \langle D\varphi(u(t)), \dot{u}(t) \rangle = -|\dot{u}(t)|^2 = -|D\varphi(u(t))|^2.$$

or by C.S. inequality as above!
integration.

$$\Leftrightarrow \varphi(u(t)) + \frac{1}{2} \int_0^t (|\dot{u}(r)|^2 + |D\varphi(u(r))|^2) dr \leq \varphi(u_0)$$

$$\forall t \geq 0.$$

Note, in B -space,

$$\langle D\varphi(u(t)), \dot{u}(t) \rangle \text{ is duality, } X^* \times X,$$

$$\text{so, } \langle D\varphi(u(t)), \dot{u}(t) \rangle = -\|\dot{u}(t)\|_X = -\|D\varphi(u(t))\|_{X^*}.$$

Back to (X, d) metric space:

$u: I \rightarrow X$ in AC^p ($u \in AC^p(I, X)$). if $\exists v \in L^p(I)$, $v \geq 0$.

$$d(u(s), u(t)) \leq \int_s^t v(r) dr \quad s, t \in I, s \leq t. \quad (\#).$$

$$p \in [1, \infty]. - \text{ Yes, } p = \infty \text{ note, } AC = AC^1.$$

If $u \in AC(I, X)$, then for almost every $t \in I$, then.] theorem.

$$\exists \lim_{h \rightarrow 0} \frac{d(u(t), u(t+h))}{|h|} =: |\dot{u}(t)|$$

↑ Borel function.

②

$\|u\| \leq v \Rightarrow u \in L^p$ and $\|u\|$ is the minimal v. Schiffrig (1)

If X is either a Banach space,

$\Rightarrow n \in AC(I, X)$

$\Rightarrow n$ is a.e. differentiable and

$$\exists \dot{u}(t) = \lim_{h \rightarrow 0} \frac{u(t+h)-u(t)}{h}$$

- $\|u'(t)\| = \|u(t)\|_X$.

- $u(t) = u(s) + \int_s^t u'(r) dr.$

Upper gradient and slope

G: $X \rightarrow [0, \infty]$ is an upper gradient for φ . If

for every $n \in AC([0, T], X)$. a.t.

$$(uG1) \quad \int_0^T G(u(t)) |u'(t)| dt < +\infty$$

$$\xrightarrow{(uG2)} |\varphi(u(T)) - \varphi(u(0))| \leq \int_0^T G(u(t)) |u'(t)| dt.$$

Now, in \mathbb{R} -space.

$$\left| \frac{d}{dt} \varphi(u(t)) \right| = |\langle D\varphi(u(t)), u'(t) \rangle| \leq \|D\varphi(u(t))\| |u'(t)|.$$

In this situation, replace $D\varphi$ by G .

- Upper Gradi. implies $\left| \frac{d}{dt} \varphi(u(t)) \right| \leq G(u(t)) \cdot |u'(t)|$ a.e.

(uG3)

- Candidate for upper bnd: slope (descending).

$$|\partial\varphi|(u) := \limsup_{v \rightarrow u} \frac{(\varphi(u) - \varphi(v))_+}{\alpha(u, v)} \quad (\alpha)_+ = \text{positive part of } \alpha.$$

Remark. If $\varphi(u(t))$ differentiable at t , then .

$$|\frac{d}{dt} \varphi(u(t))| \leq |\partial\varphi|(u(t)) |u'(t)|.$$

i.e., slope is always a candidate "upper gradient". (u6.3)
However, you don't always get (u6.2).

When $|\partial\varphi|$ is an upper gradient., then

$$|\varphi(u(t)) - \varphi(u(s))| \leq \int_s^t |\partial\varphi|(u(r)) |u'(r)| dr.$$

let (X, d) be a complete metric space, $\varphi: X \rightarrow (-\infty, +\infty]$
be a l.s.c. function, $G: X \rightarrow [0, +\infty]$ an upper gradient.
Given $u_0 \in X$, a curve $u: [0, +\infty) \rightarrow X$ is a G.F.
of φ . Starting from u_0 if $u \in AC_{loc}([0, +\infty); X)$ and

$$\varphi(u(t)) + \frac{1}{2} \int_0^t (|u'(r)|^2 + G^2(u(r))) dr \leq \varphi(u_0) \quad \forall t \geq 0.$$

$$\mathcal{S}(\varphi) = \{u \in X : \varphi(u) < \infty\}.$$

• From the defⁿ and the UG property,

$$\forall u \in AC_{loc}([0, +\infty)) \text{ and}$$

$$\frac{d}{dt} \varphi(u(t)) = -|u'(t)|^2 = -G^2(u(t)) \quad t-a.e.$$

• Assume that: $\exists \varphi(u) \geq -a - b d(u, u_0)$. for some $a, b \geq 0, u_0 \in X$.

If $(u_n) \in X$; $d(u_m, u_n) \leq c$, $\varphi(u_m) \leq c$. $\forall n, m \in \mathbb{N}$.
 $\Rightarrow \exists u_{n_k} \rightarrow u$.

(III) ~~$G \leq |\partial\varphi|$~~ $G \leq |\partial\varphi|$ is lower semicontinuous

④

Then (I) - (III)

$\Rightarrow \exists \bar{u} \in D(\varphi), \exists u \in AC_{loc}^2([0,+\infty), X)$. a G.F. for φ starting at \bar{u} .

Variational (WED) formulation:

$\varepsilon > 0, \mu_\varepsilon = \frac{e^{-t\varepsilon}}{\varepsilon} L^1$, probability measure on $[0, +\infty)$.

$$J_\varepsilon(u) := \int_0^{+\infty} \left(\frac{\varepsilon}{2} |u'|^2(t) + \varphi(u(t)) \right) d\mu_\varepsilon(t).$$

$$= \int_0^{+\infty} e^{-t\varepsilon} \left(\frac{1}{2} |u'|^2(t) + \frac{1}{\varepsilon} \varphi(u(t)) \right) dt, \quad u \in AC_{loc}^2([0, +\infty), X)$$

• let u_ε be a minimiser of J_ε among all curves in $AC_{loc}^2([0, +\infty), X)$ with $u(0) = \bar{u}$. (u_ε exists assum (I) & (II))

• $J_{\varepsilon u} \downarrow 0 \cdot u_\varepsilon(t) \rightarrow u(t) \cdot N_t$.

• u is a gradient flow for φ (and G) (you need 3)
lower semicontinuity
of G .

History:

φ : quadratic, $X = \text{Hilbert} \rightarrow$ Gauss-Marguerite.

Ishii '94, De Giorgi '96. (Minimal surface).

Milner, Ortiz, Stefanelli - Hilbert case, φ convex.

Smooth (finite dom case): u_ε is a minimiser.

$$\Rightarrow -\varepsilon u_\varepsilon''(t) + u_\varepsilon'(t) + D\varphi(u_\varepsilon(t)) = 0. \quad (\star\star)$$

gradient regularized by 2nd order term.

(5)

(*) There is a elliptic problem, because we have $-\varepsilon u''(t)$, ie, sign is -ve.

Fix $\sigma \in C_c^\infty([0, \infty); \mathbb{R}^d)$, $m = \delta v$, δ small.

$I_\varepsilon(m + \delta v) \geq I_\varepsilon(m)$, ie check.

$$\begin{aligned} & \frac{\partial}{\partial \delta} \int_0^{+\infty} \left(\frac{\varepsilon}{2} |u_\varepsilon + \delta v|^2 + \Psi(m_\varepsilon + \delta v) \right) d\mu_\varepsilon \Big|_{\delta=0} \\ &= \int_0^{+\infty} \varepsilon \langle u_\varepsilon(t), \dot{v}(t) \rangle + \langle D\Psi(m_\varepsilon(t)), v(t) \rangle d\mu_\varepsilon. \\ & \quad \forall v \in C_c^\infty([0, \infty), \mathbb{R}^d) \end{aligned}$$

Integration by parts: $\int_0^{+\infty} \varepsilon \langle u(t), \dot{v}(t) \rangle d\mu_\varepsilon = \int_0^{+\infty} (-\varepsilon \dot{u}(t) + w(t), v(t)) d\mu_\varepsilon$

$$+ \underbrace{4\varepsilon \langle u(0), v(0) \rangle}_{\text{if } w(0)=0}$$

Since $\frac{d}{dt} \circ c \leqslant w$.
in gen,

So, $\int_0^{+\infty} \underbrace{\langle -\varepsilon(u_\varepsilon'' + u_\varepsilon' + D\Psi(m)), v(t) \rangle}_{\text{comes for letting } m = m_\varepsilon} d\mu_\varepsilon = 0$

Why gen approximation?
 $D\Psi$ averages up.

Marginal functionals, $\mathcal{X} = \text{Hausdorff},$ not convex or f convex
 $\varphi(n) = \min \Phi(n, x)$, $x \in Y \subset \text{convex space } Y$, Φ nice func.

$$\Phi = H^{-1}(x), x \in H_0^1(\Omega), \Phi(n, x) = \int_{\Omega} \frac{1}{2} n \cdot x^2 dx + \frac{1}{2} \int_{\Omega} |n - x|^2 dx + \frac{1}{2} \int_{\Omega} |Dx|^2 dx$$

Lindelöf diff. $\mathcal{D}\Psi(n) = \{x(n-x)\},$ where x solves $\int_{\Omega} W(x) dx.$

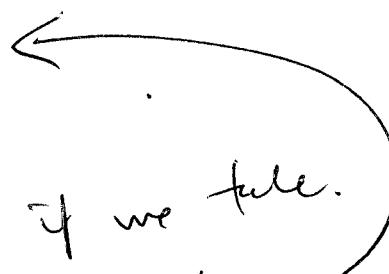
$-\Delta x + W'(x) + x = n$ < more than one sol^b.
cub diff multi-valued b.

$\mathcal{W} \leftarrow \mathcal{W}$ (6)

Hmm

The equatiⁿ to solve is

$$\begin{cases} \partial_x n - \Delta(n-x) = 0 \\ -\Delta x + W'(x) + x = n \end{cases}$$



b, slope is not good, because if we take
the minimizer, it may flip b/w solns to
b/c ∂_x is multi-valued.

~~Algorithm~~ G, F : $\varphi(n) = \inf_x (E(n) + n^T x + G(x))$.

$$= E(n) - G^*(n).$$

and $\begin{cases} \partial_x n + DE(n) = x. \\ D G(x) = n. \end{cases}$

~~Then b/e good w/ a G/F combination.
Then this choice of G is the "right" G, b/c
it is global's minimum of the x 's.~~

$$\partial_x n + DE(n) + \underbrace{(DG)'(n)}_{} = 0$$

If next G.F, is convex then
this is multi valued, which is
the problem in prev. example.

