MONOTONICITY METHODS IN PDE

ZUJIN ZHANG

ABSTRACT. In this paper, we renormalize the huts 5.1.3 and 6.1.1 in [1], so as to be more accessible, see more details in [4]. Roughly speaking, monotonicity is the natural substitution of convexity in building solutions of *PDE*.

CONTENTS

- 1. Minty-Browder method in L^2 12. Minty-Browder method in L^{∞} 3Acknowledgements9REFERENCES9
- 1. **Minty-Browder method in** L^2 **.** In this hut, we introduce the **monotonicity method** due to Minty and Browder. As as illustrative problem, we consider the following quasi-linear PDE:

$$\begin{cases}
-\operatorname{div}\left(\mathbf{E}(Du)\right) = f, & \text{in } U, \\
u = 0, & \text{on } \partial U,
\end{cases}$$
(1)

where $E: \mathbb{R}^n \to \mathbb{R}^n$ is given.

Observe that (1) can be solved by calculations of variations in case E = DF for some convex $F : \mathbb{R}^n \to \mathbb{R}$.

Key words and phrases. Monotonicity method, weak convergence method.

Our problem is then what natural conditions on E so that (1) may be directly tackled, when E is no longer the gradient of a convex function.

This is the work of Minty and Browder, who give

Definition 1. A vector field E on \mathbb{R}^n is called monotone if

$$(E(p) - E(q)) \cdot (p - q) \ge 0, \ \forall \ p, q \in \mathbb{R}^n, \tag{2}$$

and show (1) can be tacitly worked out as

Theorem 2. Assume E is monotone and satisfies the growth condition $|E(p)| \le C(1+|p|), p \in \mathbb{R}^n$.

Let $\{u_k\} \in H_0^1(U)$ be weak solutions of the approximating problems

$$\begin{cases}
-div \ (\mathbf{E}(Du_k)) = f_k, & \text{in } U, \\
u_k = 0, & \text{on } \partial U,
\end{cases}$$
(3)

with $f_k \to f$ in $L^2(U)$.

Suppose $u_k \to u$ in $H_0^1(U)$. Then u is a weak solution of (1).

Proof. We first write down

$$0 \leq \int_{U} [E(Du_{k}) - E(Dv)] [Du_{k} - Dv] dx \quad (Monotonicity)$$

$$= \int_{U} [f_{k}u_{k} - f_{k}v - E(Dv)(Du_{k} - Dv)] dx, \quad \forall v \in H_{0}^{1}(U)$$
(integration by parts and weak formulation).

Then taking $k \to \infty$ yields

$$0 \le \int_{U} \left[f(u - v) - E(Dv) \cdot (Du - Dv) \right] dx.$$

Choosing $v = u + \lambda w$, with $\lambda \in \mathbb{R}$, $w \in H_0^1(U)$ furthermore gives

$$0 \le sgn(\lambda) \int_{U} \left[\mathbf{E} (Du + \lambda Dw) \cdot Dw - fw \right] dx.$$

Passing $\lambda \to 0$ finally, we have as desired

$$0 = \int_{U} \left[E(Du) \cdot Dv - fw \right] dx, \ \forall \ w \in H_0^1(U).$$

2. **Minty-Browder method in** L^{∞} . We consider the strong solutions of *PDE*, instead of weak solutions in (1). Hence the Minty-Browder method moves from L^2 to L^{∞} .

To illustrate how it works, let us consider the following fully non-linear *PDE*:

$$\begin{cases} F(D^2 u) = f, & \text{in } U, \\ u = 0, & \text{on } \partial U, \end{cases}$$
 (4)

where $F: S^{n \times n} \to \mathbb{R}$ is given. Here $S^{n \times n}$ is the space of real, symmetric $n \times n$ matrices.

Definition 3. The problem (4) is **elliptic**, if F is monotone decreasing with respect to matrix ordering on $S^{n\times n}$, and so

$$F(S) \le F(R), \text{ if } S \ge R, S, R \in S^{n \times n}.$$
 (5)

Remark 4. This very definition of ellipticity, coincides with the classical ones. In fact, we say PDE

$$Tr[A \cdot Du] = f$$

is **elliptic** if A is a non-positive definite symmetric matrix. One then readily verifies

$$S \ge R \implies S - R \text{ non-negative definite}$$

$$\Rightarrow Tr[A \cdot (S - R)] \le 0$$

$$\Rightarrow Tr[A \cdot S] \le Tr[A \cdot R], S, R \in S^{n \times n}.$$

Now, suppose $f_k \to f$ uniformly, and consider the approximating problems

$$\begin{cases} F(Du_k) = f_k, & \text{in } U, \\ u_k = 0, & \text{on } \partial U. \end{cases}$$
 (6)

Assume (6) has a smooth solution u_k , a priori bounded in $W^{2,\infty}(U)$. Then, up to a subsequence,

$$u_k \to u$$
 uniformly, $D^2 u_k \stackrel{*}{\rightharpoonup} D^2 u$ in $L^{\infty}(U; S^{n \times n})$,

for some *u*.

Our **problem** is then: does u satisfies (4)?

If F is uniformly elliptic and convex, then strong estimates are available and passing to limit is simple, see [3]. The main interest is consequently for the nonconvex F, as in hut 1.

Recall that in hut 1, the main assumption leading to the existence of a weak solution is the monotonicity inequality (2). We shall then furnish a similar monotonicity in this current circumstance, replacing the ellipticity of F.

For this purpose, we need

Proposition 5. Let $(X, \|\cdot\|)$ be a Banach space. Then the limit

$$[f,g] = \lim_{\lambda \to 0_+} \frac{\|g + \lambda f\|^2 - \|g\|^2}{2\lambda}$$
 (7)

exists for all $f, g \in X$.

Proof. Writing

$$\frac{\|g + \lambda f\|^2 - \|g\|^2}{2\lambda} = \frac{\|g + \lambda f\| + \|g\|}{2} \cdot \frac{\|g + \lambda f\| - \|g\|}{\lambda},$$

we need only show that $\left\{\frac{\|g + \lambda f\| - \|g\|}{\lambda}\right\}_{\lambda > 0}$ is bounded from below and increasing in λ . In fact, we have

1.
$$\frac{\|g + \lambda f\| - \|g\|}{\lambda} \ge \frac{-\lambda \|f\|}{\lambda} = -\|f\|;$$
2. for $0 < \lambda < \tilde{\lambda}$,
$$\frac{\|g + \lambda f\| - \|g\|}{\lambda} - \frac{\|g + \tilde{\lambda} f\| - \|g\|}{\tilde{\lambda}}$$

$$= \frac{\|\tilde{\lambda} g + \lambda \tilde{\lambda} f\| - \tilde{\lambda} \|g\| - \|\lambda g + \lambda \tilde{\lambda} f\| + \lambda \|g\|}{\lambda \tilde{\lambda}}$$

$$\le \frac{\|(\tilde{\lambda} - \lambda)g\| - (\tilde{\lambda} - \lambda) \|g\|}{\lambda \tilde{\lambda}} = 0.$$

Remark 6. In case X is a Hilbert space, [f, g] is simply the inner product.

We now give an useful property of $[\cdot, \cdot]$ as

Proposition 7. The map $X \times X \ni \{f, g\} \mapsto [f, g]$ is upper semicontinous, that is,

$$\limsup_{n \to \infty} [f_n, g_n] \le [f, g], \tag{8}$$

for all $f, g \in X$, $f_n \to f$, $g_n \to g$ in X.

Proof. Observe that in the proof of (7), we have $\left\{\frac{\|g + \lambda f\| - \|f\|}{\lambda}\right\}_{\lambda > 0}$ is increasing in λ , for $f, g \in X$ fixed.

Thus

$$\lim_{n \to \infty} \sup [f_n, g_n] = \lim_{n \to \infty} \sup_{\lambda \to 0_+} \frac{\|g_n + \lambda f_n\|^2 - \|g_n\|^2}{2\lambda}
= \lim_{n \to \infty} \sup \left\{ \lim_{\lambda \to 0_+} \left[\frac{\|g_n + \lambda f_n\| + \|g_n\|}{2} \cdot \frac{\|g_n + \lambda f_n\| - \|g_n\|}{\lambda} \right] \right\}
= \lim_{n \to \infty} \sup \left[\|g_n\| \cdot \lim_{\lambda \to 0_+} \frac{\|g_n + \lambda f_n\| - \|g_n\|}{\lambda} \right]
\leq \|g\| \cdot \lim_{n \to \infty} \sup \frac{\|g_n + \lambda f_n\| - \|g_n\|}{\lambda}$$

$$\leq \ \|g\|\cdot \frac{\|g+\lambda f\|-\|g\|}{\lambda}, \ \forall \lambda>0.$$

Taking $\lambda \to 0_+$, we obtain

$$\limsup_{n \to \infty} [f_n, g_n] = \|g\| \cdot \lim_{\lambda \to 0_+} \frac{\|g + \lambda f\| - \|g\|}{\lambda}$$
$$= \lim_{\lambda \to 0_+} \frac{\|g + \lambda f\|^2 - \|g\|^2}{2\lambda}$$
$$= [f, g].$$

Then an explicit formula in case $X = C(\bar{U})$ as

Proposition 8. *Let* $X = C(\bar{U})$ *, then*

$$[f,g] = \max \left\{ f(x_0)g(x_0); \ x_0 \in \bar{U}, |g(x_0)| = ||g||_{C(\bar{U})} \right\}, \ f,g \in C(\bar{U}). \tag{9}$$

Proof. Denote by

$$M_h = \left\{ x \in \bar{U}; \ |h(x)| = ||h|| \right\}, \ h \in C(\bar{U}).$$

Then

1. due to

$$\frac{\|g+\lambda f\|^2-\|g\|^2}{2\lambda}\geq \frac{(g(x_0)+\lambda f(x_0))^2-g(x_0)^2}{2\lambda}=g(x_0)f(x_0),\ \forall\ x_0\in M_g,$$

we have

$$[f,g] \ge RHS$$
 of (9).

2. for any sequence $\{\lambda_n\} \setminus 0$, $x_n \in M_{g+\lambda_n f}$,

$$\frac{\|g + \lambda_n f\|^2 - \|g\|^2}{2\lambda_n} \leq \frac{(g(x_n) + \lambda_n f(x_n))^2 - g(x_n)^2}{2\lambda_n}$$

$$= f(x_n)g(x_n) + \frac{\lambda_n}{2}f(x_n)^2$$

$$\to f(x_\infty)g(x_\infty), \text{ as } n \to \infty,$$
(10)

for some $\bar{U} \ni x_{\infty} \leftarrow x_n$.

Meanwhile, taking $n \to \infty$ in

$$|g(x_n) + \lambda_n f(x_n)| = ||g + \lambda_n f||,$$

gives

$$|g(x_{\infty})| = ||g||.$$

This together with (10) shows that

$$[f,g] \leq RHS$$
 of (9).

The proof is then completed.

With this explicit formula for [f, g], we show that monotonicity is a consequence of ellipticity as

Proposition 9. *If F is convex, then the operator* $A[u] \equiv F(D^2u)$ *satisfies*

$$0 \le [A[u] - A[v], u - v], \ \forall \ u, v \in C_0^2(\bar{U}).$$
 (11)

Here $C_0^2(\bar{U})$ is the subspace of $C^2(\bar{U})$, with vanishing boundary data.

Proof. Suppose $(u - v)(x_0) = ||u - v||_{C(\bar{U})}, x_0 \in U$, then

$$D^2(u-v)(x_0) \le 0$$

$$\Rightarrow$$
 $F(D^2u)(x_0) \ge F(D^2v)(x_0)$ (by ellipticity)

$$\Rightarrow [A[u] - A[v], u - v] = (F(D^2u) - F(D^2v))(x_0) \cdot (u - v)(x_0) \ge 0,$$

by invoking (9).

The case
$$(v - u)(x_0) = ||u - v||_{C(\bar{U})}$$
, $x_0 \in U$ is similarly treated. \square

With all the above preparations above, we now state and prove our main result in this hut. **Theorem 10.** Consider problem (4) and its approximating problems (6). If $A[u] \equiv F(D^2u)$ satisfies the monotonicity inequality:

$$0 \le [A[u] - A[v], u - v], \ \forall \ u, v \in C_0^2(\bar{U}).$$
 (12)

Then u solves (4) a.e..

Proof. 1. For the approximating solution $\{u_k\}$, we have

$$0 \le [A[u_k] - A[v], u_k - v]$$

$$\le [f_k - A[v], u_k - v], \ \forall \ v \in C_0^2(\bar{U}).$$

Taking $k \to \infty$ upon a subsequence, we obtain by invoking (8) that

$$0 \le [f - A[v], u - v], \ \forall \ v \in C_0^2(\bar{U}).$$
 (13)

2. Our strategy to prove *u* solves (4) is then to choose appropriate *v* in (13).

In fact, since $u \in W^{2,\infty}(U)$, Rademacher's theorem (see [2, 5]) implies then u is C^2 a.e.. Fix any $x_0 \in U$ where $D^2u(x_0)$ exists. We **handcraft** a C^2 function v having the form

$$v(x) \begin{cases} = u(x_0) + Du(x_0)(x - x_0) \\ + \frac{1}{2}D^2u(x_0)(x - x_0, x - x_0) + \varepsilon |x - x_0|^2 - 1, & x \text{ near } x_0; \\ = 0, & x \in \partial U; \\ \in \left(u(x) - \frac{1}{2}, u(x) + \frac{1}{2}\right), & \text{otherwise.} \end{cases}$$
(14)

(The $\varepsilon > 0$ is chosen so that u - v looks like a parabola for x near x_0 .) Then |u - v| attains its maximum over \bar{U} only at x_0 . But then (13) and (9) say $(f - A[u])(x_0) \ge 0$, that is,

$$f(x_0) \ge F\left(D^2 u(x_0) + 2\varepsilon I\right).$$

Sending $\varepsilon \to 0_+$, we find

$$f(x_0) \ge F(D^2 u(x_0)).$$

The opposite inequality follows by replacing $\varepsilon |x - x_0|^2 - 1$ by $-\varepsilon |x - x_0|^2 + 1$ in (13). Consequently, we have

$$F(D^2u(x_0)) = f(x_0), \ a.e.x_0 \in U.$$

Acknowledgements. Thanks are due to the discussion group of Professor Yin at Sun Yat-sen University, in particular Dr. Liu's lectures on the monotone property of $\left\{\frac{\|g+\lambda f\|-\|g\|}{\lambda}\right\}_{\lambda>0}$ in the proof of (7), setting forth the simple observation of the proof of (8) by the author through suffering two misleading applications of L' Hospital's law in calculus.

REFERENCES

- L.C. Evans, Weak convergence methods for nonlinear partial differential equations. CBMS Regional Conference Series in Mathematics, 74. American Mathematical Society, 1990.
- [2] H. Federer, Geometric measure theory. Springer-Verlag New York Inc., New York, 1969.
- [3] D. Gilbarg, N.S. Trudinger, Elliptic partial differential equations of second order. Reprint of the 1998 edition. Classics in Mathematics. Springer-Verlag, Berlin, 2001.
- [4] J.L. Lions, Quelques mthodes de ré solution des problmes aux limites non linaires. Gauthier-Villars, Paris, 1969.
- [5] L. Simon, Lectures on geometric measure theory. Australian National University, Centre for Mathematical Analysis, Canberra, 1983.

Department of Mathematics, Sun Yat-sen University, Guangzhou, 510275, P.R. China

E-mail address: uia.china@gmail.com