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AN L^1 REGULARISATION STRATEGY TO THE INVERSE SOURCE IDENTIFICATION PROBLEM FOR ELLIPTIC EQUATIONS

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Abstract. In this paper we utilise new methods of Calculus of Variations in L^1 to provide a regularisation strategy to the ill-posed inverse problem of identifying the source of a non-homogeneous linear elliptic equation, satisfying Dirichlet data on a domain. One of the advantages over the classical Tykhonov regularisation in L^2 is that the approximated solution of the PDE is uniformly close to the noisy measurements taken on a compact subset of the domain.

1. Introduction

Let $n \geq 2$ and $\Omega \subset \mathbb{R}^n$ be a bounded domain with $C^{1,1}$ regular boundary $\partial\Omega$. Let also L be the linear non-divergence differential operator

$$(1.1) \quad L[u] := A : D^2u + b \cdot Du + cu$$

which is assumed to be uniformly elliptic with bounded continuous coefficients:

$$(1.2) \quad \begin{cases} A \in (C^0 \setminus L^1)(\Omega; \mathbb{R}_s^{n \times n}); b \in (C^0 \setminus L^1)(\Omega; \mathbb{R}^n); c \in (C^0 \setminus L^1)(\Omega); \\ \text{and exists } a_0 > 0 : A : \xi\xi \geq a_0 |\xi|^2; \text{ for all } \xi \in \mathbb{R}^n. \end{cases}$$

In the above, the notations $\cdot$ and $\cdot$ symbolise the Euclidean inner products in the space of symmetric matrices $\mathbb{R}_s^{n \times n}$ and in $\mathbb{R}^n$ respectively, whilst $Du = (D_i u)_{i=1,\dots,n}$, $D^2u = (D_{ij}^2 u)_{i,j=1,\dots,n}$ and $D_i = \partial/\partial x_i$. The direct (or forward) Dirichlet problem for the above operator has the form

$$(1.3) \quad \begin{cases} L[u] = f; & \text{in } \Omega; \\ u = g; & \text{on } \partial\Omega; \end{cases}$$

and asks to determine u , given a source f and boundary data g . This is a classical problem which is essentially textbook material, see e.g. [19, Ch. 9]. In particular, it is well-posed (in the sense of Hadamard) and, given $f \in L^1(\Omega)$ and $g \in W^{2,1}(\partial\Omega)$, there exists a unique solution u in the locally convex (Frechet) space

$$(1.4) \quad W_g^{2,1}(\Omega) := \bigcap_{1 < p < \infty} \{ u \in W^{2,p}(\Omega) \setminus W_g^{1,p}(\Omega) : L[u] \in L^1(\Omega) \}.$$

Note that due to the failure of the L^p elliptic estimates when $p = 1$ (see e.g. [18]), in general $u \notin W^{2,1}(\Omega)$. Let us also note with the assumptions (1.2) on L , the case of divergence operators with C^1 matrix coefficient A is included as a special case:

$$L^0[u] = \operatorname{div}(ADu) + b \cdot Du + cu.$$

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The *inverse problem* associated to (1.3) consists of the question of finding f , given the boundary data g and some *partial information on the solution u* , typically

can only determine a unique biharmonic function u in Ω with $\Delta^2 u = 0$. Another popular choice in the literature for the observation operator Q consist of one of the terms in the separation of variables formula (when $L = \Delta$ on rectangular domains), as e.g. in [36]. To the best of our knowledge, (1.10) has not been studied before in this generality.

Herein we follow an approach based on recent advances in Calculus of Variations in the space L^1 (see [22, 23, 24, 25]) developed recently for functionals involving higher order derivatives. The field has been initiated in the 1960s by Gunnar Aronsson (see e.g. [3, 4, 5, 6, 7]) and is still a very active area of research; for a review of the by-now classical theory involving scalar first order functionals we refer to [21]. To this end, we provide a *regularisation strategy* inspired by the classical Tykhonov regularisation strategy in L^2 (see e.g. [27, 30]), but for the next L^1 "error" functional:

$$(1.11) \quad E_1(u) := \|Q[u] - q\|_{L^1(\cdot; H)} + \lambda \|L[u]\|_{L^1(\cdot)}; \quad u \in W_g^{2,1}(\cdot);$$

where $\lambda > 0$ is a fixed regularisation parameter for the penalisation term $\lambda \|L[u]\|$. In the variational language, it serves to make the functional coercive in the space. The benefit of finding a best fitting solution in L^1 is apparent: we can keep the error term $\|Q[u] - q\|$ due to the noise effects *uniformly small, not merely small on average*, which would happen if one chose to minimise the integral of the error instead of the supremum.

As it is well known to the experts of Calculus of Variations in L^1 , mere (global) minimisers of supremal functionals, albeit typically easy to obtain with standard direct minimisation methods ([13, 16]), they are not truly optimal and they do not share the nice "local" minimality properties of minimisers of their integral counterparts ([10, 32]). A popular method is to use minimisers of L^p approximating functionals as $p \rightarrow 1$ and prove appropriate convergence of such L^p minimisers to a limiting L^1 minimiser. This method is fairly standard nowadays and provides a selection principle of L^1 minimisers with additional favourable properties (see e.g. [9, 11, 12, 17, 22, 23]). This idea is inspired by the simple measure-theoretic fact that the L^p norm (of a fixed $L^1 \setminus L^1$ function) converges to the L^1 norm as $p \rightarrow 1$. Accordingly, we will obtain *special* minimisers of (1.11) as limits of minimisers of

$$(1.12) \quad E_p(u) := \|Q[u] - q\|_{L^p(\cdot; H)} + \lambda \|L[u]\|_{L^p(\cdot)}; \quad u \in W_g^{2,1}(\cdot);$$

Theorem 1 (L^1 and L^p regularisations of the inverse source identification problem). *Let $\Omega \subset \mathbb{R}^n$ be a bounded $C^{1,1}$ domain and let also g be in $W^{2,1}_g(\Omega)$. Suppose also the operators (1.1) and (1.6) are given, satisfying the assumptions (1.2), (1.7), (1.8). Suppose further a function $q \in L^1(\Omega; H)$ is given which satisfies (1.9) for $\gamma > 0$. Let finally $\epsilon > 0$ be fixed. Then, we have the next results in relation to the problem (1.10):*

(i) [Existence] *There exist a global minimiser $u_1 = u_1^\epsilon \in W^{2,1}_g(\Omega)$ of the functional E_1 defined in (1.11). In particular, we have $E_1(u_1) = E_1(v)$ for all $v \in W^{2,1}_g(\Omega)$ and*

$$f_1 = f_1^\epsilon := L[u_1^\epsilon] \in L^1(\Omega);$$

In addition, there exist signed Radon measures

$$\mu_1 = \mu_1^\epsilon \in M(\Omega); \quad \nu_1 = \nu_1^\epsilon \in M(\Omega)$$

such that the divergence PDE

$$(1.13) \quad K_r(\cdot; u_1; Du_1) = \operatorname{div} K_p(\cdot; u_1; Du_1) = 1 + L[f_1] = 0;$$

is satisfied by the triplet $(u_1; \mu_1; \nu_1)$ in the distributional sense. In (1.13), the operator L is the formal adjoint of L , defined through duality, i.e.

$$L[v] := \operatorname{div}(\operatorname{div}(Av)) = \operatorname{div}(bv) + cv$$

and $K_r; K_p$ denote the partial derivatives of $K(x; r; p)$ with respect to $(r; p) \in \mathbb{R} \times \mathbb{R}^n$. Additionally, the error measure μ_1 is supported in the closure of the subset of Ω of maximum noise, that is

$$(1.14) \quad \operatorname{supp}(\mu_1) \subset \bigcap_n \{Q[u_1] - q\}^F = \{Q[u_1] - q\}^F \cap L^1(\Omega; H);$$

where $\{ \cdot \}^F$ symbolises the "essential limsup" with respect to the Radon measure $H \llcorner \Omega$ on Ω , see Proposition 6 that follows. If additionally the measurement function q is continuous on Ω , (1.14) improves to

$$(1.15) \quad \operatorname{supp}(\mu_1) \subset \bigcap_n \{Q[u_1] - q\} = \{Q[u_1] - q\} \cap L^1(\Omega; H);$$

(ii) [Convergence] *For any $\epsilon > 0$, the minimiser u_1 can be approximated by a family of minimisers $(u_p)_{p>n} = (u_p^\epsilon)_{p>n}$ of the respective L^p functionals (1.12) and the pair of measures $(\mu_1; \nu_1) \in M(\Omega) \times M(\Omega)$ can be approximated by respective absolutely continuous signed measures $(\mu_p; \nu_p)_{p>n} = (\mu_p^\epsilon; \nu_p^\epsilon)_{p>n}$, as follows:*

For any $p > n$, the functional (1.12) has a global minimiser $u_p = u_p^\epsilon$ in the space $(W^{2,p}_g \setminus W^{1,p}_g)(\Omega)$ and there exists a sequence $p_j \rightarrow \infty$ as $j \rightarrow \infty$, such that

$$(1.16) \quad u_{p_j} \rightarrow u_1; \quad \text{in } C^1(\bar{\Omega}); \quad \mu_{p_j} \rightarrow \mu_1; \quad \nu_{p_j} \rightarrow \nu_1$$

as $p \rightarrow 1$ along the sequence. Further, for each $p > n$, the triplet $(u_p; \rho_p; \rho_p)$ solves the equation

$$(1.18) \quad K_r(\cdot; u_p; Du_p)_{\rho_p} - \operatorname{div} K_p(\cdot; u_p; Du_p)_{\rho_p} + L[u_p] = 0;$$

in the distributional sense.

(iii) [L^1 error estimates] For any exact solution $u^0 \in W_g^{2;1}(\cdot)$ of (1.10) (with $f = L[u^0]$ and $Q[u^0] = q^0$) corresponding to measurements with zero noise, we have the estimate:

$$(1.19) \quad \|Q[u_p] - Q[u^0]\|_{L^1(\cdot; H)}^2 \leq k \|L[u^0]\|_{L^1(\cdot)};$$

for any $\epsilon > 0$.

(iv) [L^p error estimates] For any exact solution $u^0 \in (W^{2;p} \setminus W_g^{1;p})(\cdot)$ of (1.10) (with $f = L[u^0]$ and $Q[u^0] = q^0$) corresponding to measurements with zero noise and for $p > n$, we have the estimate:

$$(1.20) \quad \|Q[u_p] - Q[u^0]\|_{L^p(\cdot; H)}^2 \leq k \|L[u^0]\|_{L^p(\cdot)};$$

for any $\epsilon > 0$.

The estimate in part (iv) above is useful if we have merely that $L[u^0] \in L^p(\cdot)$ for $p < 1$ (namely when perhaps $L[u^0] \in$

$Q[u] := u(x; c)$, for $n = 2$ and $\Omega = (a; b) \times (c; d)$ being a rectangular domain (i.e., one of the products in the separation of variables when $L = \Omega$). This implies that (1.19) simplifies to

$$\|u_1^j(\cdot; c) - u^0(\cdot; c)\|_{L^1((a; b); H^1)}^2 + \|kL[u^0]\|_{L^1((a; b) \times (c; d))}^2 \text{ as } \varepsilon \rightarrow 0;$$

and similarly for its L^p -counterpart.

$Q[u] := Du \cdot n$, where n is the outer normal vector on $\partial\Omega$. In this case, (1.21) simplifies to

$$\|n \cdot Du_1^j - Du^0\|_{L^1(\partial\Omega; H^{n-1})}^2 + \|kL[u^0]\|_{L^1(\Omega)}^2 \text{ as } \varepsilon \rightarrow 0;$$

and similarly for its L^p -counterpart.

We would like to note again that, due to the ill-posed nature of the problem, in general it is not possible to obtain an estimate on $\|n\|$.

We now provide some clarifications regarding Theorem 1.

Remark 5. (i) We note that in (1.13) the distributional meaning of this PDE is

$$K_r(\cdot; u_1; Du_1) \, d\Omega + K_p(\cdot; u_1; Du_1) \, D \, d\Omega + L[\cdot] \, d\Omega;$$

measure respectively, the above is in fact equivalent to

$$\begin{aligned} K_r(\cdot; u_p; Du_p) + K_p(\cdot; u_p; Du_p) &= \frac{Q[u_p] - q}{jQ[u_p] - q} \frac{j^{p-2} Q[u_p] - q}{j^{p-1} L^p(\cdot; H)} dH \\ &+ \frac{jL[u_p]j^{p-2} L[u_p]}{jL[u_p]j^{p-1} L^p(\cdot)} dL^n = 0; \end{aligned}$$

for all $\gamma \in C_c^2(\cdot)$.

(iii) Since we only prescribe boundary conditions $u = g$ on ∂ but impose no condition on the gradient (as opposed to e.g. [22], wherein an L^1 minimisation problem was considered by imposing $Du = Dg$ on ∂ additionally to $u = g$ on ∂), we therefore have "natural boundary conditions" for the gradient on ∂ . We will make no particular further use of this observation.

The following two results are of independent interest and are utilised in the proof of Theorem 1 that follows. We state and prove them in considerably greater generality than that needed herein, as they have their own merits in the Calculus of Variations in L^1 .

Proposition 6 (The essential limsup). *Let $X \subset \mathbb{R}^n$ be a Borel set, endowed with the induced Euclidean topology and let also $\mu \in M^+(X)$ be a positive finite Radon measure on X . For any $f \in L^1(X; \mu)$, we define the function $f^F \in L^1(X; \mu)$ by setting*

$$f^F(x) := \lim_{r \downarrow 0} \operatorname{ess\,sup}_{y \in B^-(x, r)} f(y)$$

(i) *There exists a subsequence $(k_i)_{i=1}^\infty$ and a limit measure $\mu \in M(X)$ such that*

$$\mu_k \xrightarrow{*} \mu \text{ in } M(X);$$

as $k_i \rightarrow \infty$.

(ii) *If there exists $f \in L^1(X; \mu)$ such that*

$$\sup_X |f_k - f| \rightarrow 0 \text{ as } k \rightarrow \infty;$$

then the limit measure is supported in the set where (the

by our assumptions on L and the Hölder inequality we have

$$\begin{aligned} E_p(v) &= kL[v]k_{L^p(\cdot)} \\ &\leq \frac{C(p; A; b; c)}{C(p; A; b; c)} kvk_{W^{2;p}(\cdot)} kgk_{W^{2;p}(\cdot)} \\ &\leq \frac{C(p; A; b; c)}{C(p; A; b; c)} kvk_{W^{2;p}(\cdot)} kgk_{W^{2;1}(\cdot)} \end{aligned}$$

for some $C = C(p; A; b; c) > 0$ and any $v \in (W^{2;p} \setminus W_g^{1;p})(\cdot)$. Let $(u_p^m)_1^\infty$ be a minimising sequence of E_p :

$$E_p(u_p^m) \rightarrow \inf_{v \in (W^{2;p} \setminus W_g^{1;p})(\cdot)} E_p(v) = 0;$$

as $m \rightarrow \infty$. Then, by the above estimates, we have the uniform bound

$$ku_p^m k_{W^{2;p}(\cdot)} \leq C$$

for some $C > 0$ depending on p but independent of $m \in \mathbb{N}$. By standard weak and strong compactness arguments in Sobolev spaces, there exists a subsequence $(u_p^{m_k})_1^\infty$ and a function $u_p \in (W^{2;p} \setminus W_g^{1;p})(\cdot)$ such that, along this subsequence we have

$$\begin{aligned} u_p^{m_k} &\rightharpoonup u_p; & \text{in } L^p(\cdot); \\ Du_p^{m_k} &\rightharpoonup Du_p; & \text{in } L^p(\cdot; \mathbb{R}^n); \\ D^2 u_p^{m_k} &\rightarrow D^2 u_p; & \text{in } L^p(\cdot; \mathbb{R}_s^{n \times n}); \end{aligned}$$

as $m_k \rightarrow \infty$. Additionally, since $p > n$, by the regularity of the boundary we have the compact embedding $W^{2;p}(\cdot) \hookrightarrow C^{1;k}(\bar{\cdot})$ as a consequence of the Morrey estimate. Hence,

$$u_p^{m_k} \rightarrow u_p \text{ in } C^{1;k}(\bar{\cdot}); \text{ for } 0 \leq k \leq \frac{n}{p};$$

as $m_k \rightarrow \infty$. The above modes of convergence and the continuity of the function K defining the operator Q imply that $Q[u_p^{m_k}] \rightarrow Q[u_p]$ uniformly on Γ as $m_k \rightarrow \infty$. Therefore,

$$jQ[u_p^{m_k}] - q|_{j(p) \in L^p(\cdot; H)} \rightarrow jQ[u_p] - q|_{j(p) \in L^p(\cdot; H)}$$

as $m_k \rightarrow \infty$. Additionally, by the linearity of the operator L and because its coefficients are L^1 , we have that

$$L[u_p^{m_k}] \rightarrow L[u_p] \text{ in } L^p(\cdot);$$

as $m_k \rightarrow \infty$. Since the functional

$$j|_{j(p) \in L^p(\cdot)} : L^p(\cdot) \rightarrow \mathbb{R}$$

is convex on this reflexive space and also it is strongly continuous, it is weakly lower semi-continuous and therefore

$$jL[u_p]j_{j(p) \in L^p(\cdot)} = \liminf_{k \rightarrow \infty} jL[u_p^{m_k}]j_{j(p) \in L^p(\cdot)}.$$

By putting all the above together, we see that

$$E_p(u_p) = \liminf_{k \rightarrow \infty} E_p(u_p^{m_k}) = \inf_{v \in (W^{2;p} \setminus W_g^{1;p})(\cdot)} E_p(v) = 0;$$

which concludes the proof.

Lemma 9. *For any $\varepsilon > 0$, there exists a (global) minimiser $u_1 \in W^{2,1}_g(\Omega)$ and a sequence of minimisers $(u_{p_i})_1^\infty$ of the respective E_p -functionals constructed in Lemma 8, such that (1.16) holds true.*

Proof. For each

Then, the triplet $(u_p; p; \rho)$ satisfies the PDE (1.18) in the distributional sense. In fact, the following stronger assertion holds: we have

$$\begin{aligned} K_r(\cdot; u_p; Du_p) + K_p(\cdot; u_p; Du_p) &= D \frac{Q[u_p] - q}{jQ[u_p] - q} \frac{j^{p-2} Q[u_p] - q}{j^{p-1} L^p(\cdot; H)} dH \\ &+ L[\cdot] \frac{jL[u_p] j^{p-2} L[u_p]}{jL[u_p] j^{p-1} L^p(\cdot)} dL^n = 0; \end{aligned}$$

for all $\varphi \in W_0^{2,p}(\cdot)$.

for any $\gamma \in W_0^{2,p}(\Omega) \cap C^1(\overline{\Omega})$, because of the continuity of $K(x; r; p)$ in x and the C^1 regularity in $(r; p)$.

Lemma 11. *For any $\gamma \in W_0^{2,p}(\Omega) \cap C^1(\overline{\Omega})$, consider the minimiser u_1 of E_1 constructed in Lemma 9 as sequential limit of minimisers $(u_p)_{p>n}$ of the functionals $(E_p)_{p>n}$ as $p_i \rightarrow 1$. Then, there exist signed Radon measures $\mu_1 \in M(\Omega)$ and $\nu_1 \in M(\Omega)$ such that the triplet $(u_1; \mu_1; \nu_1)$ satisfies the PDE (1.13) in the distributional sense, that is*

$$K_r(\cdot; u_1; Du_1) + K_p(\cdot; u_1; Du_1) \cdot D \gamma + L[\gamma] du_1 = 0;$$

for all $\gamma \in C_c^2(\Omega)$. Additionally, there exists a further subsequence along which the weak* modes of convergence of (1.17) hold true as $p \rightarrow 1$.

Proof. As noted in the beginning of the proof of Lemma 10, we have the p -uniform total variation bounds $k_p k(\cdot) \leq 1$ and $k_p k(\cdot) \leq 1$. Hence, by the sequential weak* compactness of the spaces of Radon measures

$$M(\Omega) = C_0^0(\Omega); \quad M(\Omega) = C^0(\Omega);$$

there exists a further subsequence denoted again by $(p_i)_{i=1}^\infty$ such that $\mu_p \rightharpoonup^* \mu_1$ in $M(\Omega)$ and $\nu_p \rightharpoonup^* \nu_1$ in $M(\Omega)$, as $p_i \rightarrow 1$. Fix now $\gamma \in C_c^2(\Omega)$. By Lemma 10, we have that the triplet $(u_p; \mu_p; \nu_p)$ satisfies (1.18), that is

$$K_r(\cdot; u_p; Du_p) + K_p(\cdot; u_p; Du_p) \cdot D \gamma + L[\gamma] du_p = 0;$$

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the $((\cdot; \cdot))$ -dependent) minimiser u_p of E_p (constructed in Lemmas 8-11), satisfies the error bounds (1.20), that is:

$$Q[u_p] - Q[u^0]_{L^1(\cdot; H)} \leq 2 + kL[u^0]k_{L^p(\cdot)}:$$

If additionally $u^0 \in W_g^{2;1}(\cdot)$, then the $((\cdot; \cdot))$ -dependent) minimiser u_1 of E_1 (constructed in Lemmas 8-11), satisfies the error bounds (1.19), that is:

$$Q[u_1] - Q[u^0]_{L^1(\cdot; H)} \leq 2 + kL[u^0]k_{L^1(\cdot)}:$$

Proof. Let us use the symbolisation $q^0 := Q[u^0]$, noting also that $q^0 \in C^0(\cdot)$ and that we have the estimate

$$kq^0k_{L^1(\cdot; H)} \leq 2:$$

For any $p \geq 2$ ($n \geq 1$), the function u_p is a global minimiser of E_p in $(W^{2;p} \setminus W_g^{1;p})(\cdot)$. Therefore,

$$E_p(u_p) = E_p(u^0):$$

This implies the estimate

$$\begin{aligned} Q[u_p] - q^0_{L^p(\cdot; H)} &+ L[u_p]_{L^p(\cdot)} \\ &= Q[u^0] - q^0_{L^p(\cdot; H)} + L[u^0]_{L^p(\cdot)}: \end{aligned}$$

The latter estimate together with the Minkowski and Hölder inequalities, in turn yield

$$\begin{aligned} Q[u_p] - Q[u^0]_{L^p(\cdot; H)} &= Q[u^0] - q^0_{L^p(\cdot; H)} \\ &+ Q[u^0] - q^0_{L^p(\cdot; H)} + L[u^0]_{L^p(\cdot)} \\ &= 2kq^0k_{L^1(\cdot; H)} + L[u^0]_{L^p(\cdot)} \\ &\leq 2 + kL[u^0]k_{L^p(\cdot)}; \end{aligned}$$

as claimed. To obtain the corresponding estimate for u_1 in the case that additionally $u^0 \in W_g^{2;1}(\cdot)$, we may pass to the limit as $p \rightarrow 1$ in the last estimate above: indeed, consider the subsequence $p_i \rightarrow 1$ along which we have the strong convergence $u_{p_i} \rightarrow u_1$ in $C^1(\cdot)$ and therefore $Q[u_{p_i}] \rightarrow Q[u_1]$ uniformly on $\cdot$. Since by assumption $L[u^0] \in L^1(\cdot)$, the conclusion follows by letting $i \rightarrow 1$ in the last estimate.

We now establish Proposition 6.

Proof of Proposition 6. (i) Let $B^n(x)$ be the open n -ball of $\mathbb{R}^n$ centred at x . By the Lebesgue differentiation theorem (see e.g. [16]) applied to the measure $\mathbf{X}_X$ (namely to $\mathbf{X}_X$ extended to $\mathbb{R}^n$ by zero on $\mathbb{R}^n \setminus X$) and by recalling that $B(x)$ symbolises the open ball in X , we have

$$\begin{aligned} f(x) &= \lim_{r \rightarrow 0} \frac{1}{|B^n(x)|} \int_{B^n(x)} f d(\mathbf{X}_X) \\ &= \lim_{r \rightarrow 0} \frac{1}{|B(x)|} \int_{B(x)} f d \end{aligned}$$

and therefore

$$f(x) = \lim_{t \downarrow 0} \frac{1}{(B_t(x))} \int_{B_t(x)} f \, d\mu$$

$$= \lim_{t \downarrow 0} \operatorname{ess\,sup}_{B_t(x)} f$$

By the Lebesgue-Besicovitch differentiation theorem (see e.g. [16]), -a.e. point $x \in X$ has density 1, namely

$$\lim_{r \rightarrow 0} \frac{|X \cap B_r(x)|}{|B_r(x)|} = 1;$$

where $B_r(x)$ is the open r -ball centred at x with respect to $\mathbb{R}^n$. Hence, since

$$B_r(x) \cap X = X \cap B_r(x);$$

for any $r > 0$, there exists $x \in X(r)$ such that

$$|X \cap B_r(x)| = |X \cap B_r(x)| > 0.$$

Therefore, since

$$\operatorname{ess\,sup}_{y \in X} f(y) \leq f(x); \quad \text{a.e. } x \in X(r);$$

we deduce

$$\begin{aligned} \operatorname{ess\,sup}_{y \in X} f(y) &\leq \operatorname{ess\,sup}_{y \in B_r(x) \cap X} f(y) \\ &\leq \operatorname{ess\,sup}_{y \in B_r(x)} f(y); \end{aligned}$$

By letting $r \rightarrow 0$ in the above inequality, we infer that

$$\begin{aligned} \operatorname{ess\,sup}_{x \in X} f(x) &\leq \lim_{r \rightarrow 0} \operatorname{ess\,sup}_{y \in B_r(x)} f(y) \\ &= f^*(x) \\ &\leq \sup_{x \in X} f^*(x); \end{aligned}$$

for any $r > 0$. By letting $r \rightarrow 0$, we obtain

$$\operatorname{ess\,sup}_{x \in X} f(x) \leq \sup_{x \in X} f^*(x);$$

as desired. This inequality completes the proof.

By invoking Proposition 7 whose proof follows, we readily obtain (1.14)-(1.15) by choosing

$$X = \mathbb{R}^n; \quad \mu = H^n; \quad f_k = Q[u_{p_k}] \quad q; \quad f_1 = Q[u_1] \quad q;$$

Proof of Proposition 7. (i) By the definition of f_k , we have for any continuous function $\varphi \in C^0(X)$ with $\|\varphi\|_1 = 1$ that

$$\begin{aligned} \int_X \varphi(x) dx &\leq \frac{1}{\|f_k\|_{L^k(X)}} \int_X \varphi(x) f_k(x)^{k-1} dx \\ &\leq \frac{1}{\|f_k\|_{L^k(X)}} \int_X \varphi(x) f_k(x)^{k-1} dx; \end{aligned}$$

Hence, by Hölder inequality, we have the total variation bound

$$\begin{aligned} \|f_k\|_{L^k(X)} &\leq \left(\int_X \varphi(x) f_k(x)^{k-1} dx \right)^{\frac{1}{k-1}} \left(\int_X \varphi(x) f_k(x)^k dx \right)^{\frac{1}{k}} \\ &= 1; \end{aligned}$$

By the sequential weak* compactness of the space $M(X) = C^0(X)$

By the above, for any $\epsilon > 0$ small enough (recall that $f_1 \neq 0$) and for any $k = k(\epsilon)$, we have the estimate

$$\frac{d_k}{d} \leq \frac{1}{(X)} \left(\frac{1}{k f_1 k_{L^1(X; \cdot)}} + \frac{\epsilon}{2} \right) \leq \frac{1}{k f_1 k_{L^1(X; \cdot)}} + \frac{\epsilon}{2}; \quad \text{a.e. on } X;$$

By choosing $k(\epsilon)$ even larger if needed, we can arrange

$$\frac{d_k}{d} \leq \frac{1}{(X)} \left(\frac{2 f_1 j + \epsilon}{2 k f_1 k_{L^1(X; \cdot)}} \right) \leq \frac{1}{k f_1 k_{L^1(X; \cdot)}} + \frac{\epsilon}{2}; \quad \text{a.e. on } X;$$

Since by Proposition 6 we have $j f_1 j = j f_1 j^F$ -a.e. on X , we obtain

$$\frac{d_k}{d} \leq \frac{1}{(X)} \left(\frac{2 j f_1 j^F + \epsilon}{2 k f_1 k_{L^1(X; \cdot)}} \right) \leq \frac{1}{k f_1 k_{L^1(X; \cdot)}} + \frac{\epsilon}{2}; \quad \text{a.e. on } X;$$

Consider now for any $\epsilon > 0$ the ϵ -measurable set

$$X_\epsilon := \{ j f_1 j^F < k f_1 k_{L^1(X; \cdot)} + \epsilon \};$$

Notice also that X_ϵ is in fact open in X because $j f_1 j^F$ is upper semicontinuous (Proposition 6). Additionally, we have the estimate

$$\frac{d_k}{d} \leq \frac{1}{(X)} \left(\frac{2 k f_1 k_{L^1(X; \cdot)} + \epsilon}{2 k f_1 k_{L^1(X; \cdot)}} \right) \leq \frac{1}{k f_1 k_{L^1(X; \cdot)}} + \frac{\epsilon}{2}; \quad \text{a.e. on } X_\epsilon;$$

The above estimate together with the Lebesgue Dominated Convergence theorem imply that for any $\epsilon > 0$ small enough we have

$$\frac{d_k}{d} \rightarrow 0 \quad \text{in } L^1(X_\epsilon; \cdot); \quad \text{as } k \rightarrow \infty;$$

Consider now the sequence of nonnegative data $\{f_n\}_{n=1}^\infty$ in $L^1(\mathbb{R}^d)$ such that $f_n \rightarrow f$ in $L^1(\mathbb{R}^d)$ and $f \neq 0$. For any $\epsilon > 0$ small enough, we can find N such that for all $n \geq N$, $f_n \neq 0$ and f_n is ϵ -measurable. For such n , we have

Therefore, since X^* is open in X

23. N. Katzourakis, T. Pryer, *2nd order L^1*