

Research Article

Paolo Bonicatto* and Nikolay A. Gusev

On the structure of divergence-free measures on \mathbb{R}^2

<https://doi.org/10.1515/acv-2020-0066>

Received June 19, 2020; revised October 19, 2020; accepted November 16, 2020

Abstract: We consider the structure of divergence-free vector measures on the plane. We show that such measures can be decomposed into measures induced by closed simple curves. More generally, we show that if the divergence of a planar vector-valued measure is a signed measure, then the vector-valued measure can be decomposed into measures induced by simple curves (not necessarily closed). As an application we generalize certain rigidity properties of divergence-free vector fields to vector-valued measures. Namely, we show that if a locally finite vector-valued measure has zero divergence, vanishes in the lower half-space and the normal component of the unit tangent vector of the measure is bounded from below (in the upper half-plane), then the measure is identically zero.

Keywords: Vector-valued measures, divergence-free measures, superposition Principle

MSC 2010: 49Q15, 58A25

Communicated by: Jan Kristensen

1 Introduction

In this paper we study the structure of vector-valued Borel measures μ solving the equation

$$\operatorname{div} \mu = \rho \quad (1.1)$$

in the sense of distribution on \mathbb{R}^d , where $d \geq 1$ and ρ is a given (\mathbb{R} -valued) Borel measure on \mathbb{R}^d . Many equations of the mathematical physics can be written in the form of (1.1), for instance the continuity equation. A simple (but important) example of a measure satisfying (1.1) is a *measure μ_γ induced by a Lipschitz curve $\gamma: [0, 1] \rightarrow \mathbb{R}^d$* , which is defined (via Riesz–Markov–Kakutani Theorem) by

$$\langle \mu_\gamma, \Phi \rangle \equiv \int_{\mathbb{R}^d} \Phi \cdot d\mu_\gamma := \int_0^1 \Phi(\gamma(t)) \cdot \gamma'(t) dt \quad \text{for all } \Phi \in C_0(\mathbb{R}^d; \mathbb{R}^d).$$

Here $C_0(\mathbb{R}^d; \mathbb{R}^d)$ is the closure of the set of compactly supported continuous functions $C_c(\mathbb{R}^d; \mathbb{R}^d)$ with respect to the uniform norm. It is easy to see that μ_γ solves (1.1) with $\rho := \delta_{\gamma(0)} - \delta_{\gamma(1)}$, where δ_x with $x \in \mathbb{R}^d$ denotes the Dirac measure concentrated at x .

Clearly every finite linear combination of measures of the form μ_γ still solves (1.1). More generally, let $\Gamma := \operatorname{Lip}([0, 1]; \mathbb{R}^d)$ denote the space of all Lipschitz functions $f: [0, 1] \rightarrow \mathbb{R}^d$, endowed with the sup-norm. Let $\mathcal{M}(X; \mathbb{R}^d)$ denote the set of \mathbb{R}^d -valued Borel measures on a topological space X (for $d = 1$ we will simply

*Corresponding author: Paolo Bonicatto, Mathematics Institute, University of Warwick, Coventry CV4 7AL, United Kingdom, e-mail: paolo.bonicatto@warwick.ac.uk

Nikolay A. Gusev, Moscow Institute of Physics and Technology, 9 Institutskiy per., Dolgoprudny, Moscow Region, 141700; Steklov Mathematical Institute of Russian Academy of Sciences, 8 Gubkina St, Moscow, 119991, Russia, e-mail: n.a.gusev@gmail.com

write $\mathcal{M}(X) := \mathcal{M}(X; \mathbb{R})$ and $\mathcal{M}_+(X)$ for the set of *non-negative* Borel measures). Let $|\boldsymbol{\mu}|$ denote the total variation of $\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ and recall that $\|\boldsymbol{\mu}\| := |\boldsymbol{\mu}|(\mathbb{R}^d)$ is a norm on $\mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$. Suppose that $\eta \in \mathcal{M}_+(\Gamma)$ is such that $\int_{\Gamma} \|\boldsymbol{\mu}_\gamma\| d\eta(\gamma) < \infty$. Then using Fubini’s Theorem one can show that the measure

$$\boldsymbol{\mu} := \int_{\Gamma} \boldsymbol{\mu}_\gamma d\eta(\gamma),$$

which is defined by

$$\langle \boldsymbol{\mu}, \Phi \rangle := \int_{\Gamma} \langle \boldsymbol{\mu}_\gamma, \Phi \rangle d\eta(\gamma) \quad \text{for all } \Phi \in C_0(\mathbb{R}^d; \mathbb{R}^d),$$

solves (1.1) with

$$\rho := \int_{\Gamma} (\delta_{\gamma(0)} - \delta_{\gamma(1)}) d\eta(\gamma)$$

(which is defined similarly). Therefore a natural question is whether the converse implication holds true, i.e. if any solution $\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ of (1.1) (with some $\rho \in \mathcal{M}(\mathbb{R}^d)$) can be written as $\int_{\Gamma} \boldsymbol{\mu}_\gamma d\eta(\gamma)$ for some $\eta \in \mathcal{M}_+(\Gamma)$.

Decompositions of this kind were used in [20] in order to derive the so-called superposition principle for the measure-valued solutions of the continuity equation (which was proved in [2, Theorem 12] for Euclidean spaces). In turn, such a superposition principle was used in [5] in order to obtain certain uniqueness results for solutions of the continuity equation. The main result of the present paper can be stated as follows:

Main Theorem. *Let $d = 2$. Suppose that $\rho \in \mathcal{M}(\mathbb{R}^d)$ and $\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ solve equation (1.1). Then there exists $\eta \in \mathcal{M}_+(\Gamma)$ such that*

$$\boldsymbol{\mu} = \int_{\Gamma} \boldsymbol{\mu}_\gamma d\eta(\gamma), \tag{1.2a}$$

$$|\boldsymbol{\mu}| = \int_{\Gamma} |\boldsymbol{\mu}_\gamma| d\eta(\gamma), \tag{1.2b}$$

$$|\operatorname{div} \boldsymbol{\mu}| = \int_{\Gamma} |\operatorname{div} \boldsymbol{\mu}_\gamma| d\eta(\gamma) \tag{1.2c}$$

and for η -a.e. $\gamma \in \Gamma$ there exists $\tilde{\gamma} \in \Gamma$ which is injective on $[0, 1)$ such that $\boldsymbol{\mu}_\gamma = \boldsymbol{\mu}_{\tilde{\gamma}}$.

For $d > 2$ in general such a decomposition is not possible due to examples provided in the celebrated paper [19] (in particular, one can consider $\boldsymbol{\mu}$ associated with an irrational winding of a torus). However in [19] it was proved that for any $d > 2$ the measure $\boldsymbol{\mu}$ can be decomposed into the so-called *elementary solenoids* in such a way that (1.2) hold. Recently this decomposition result was generalized for metric spaces in [15, 16]. Note that for $d > 2$ the set of elementary solenoids is strictly larger than the set of measures induced by Lipschitz curves. However, by the Main Theorem, all elementary solenoids are induced by Lipschitz curves in the case $d = 2$.

Following [15], $\boldsymbol{\sigma} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ will be called a *cycle of $\boldsymbol{\mu}$* if $\operatorname{div} \boldsymbol{\sigma} = 0$ and $\|\boldsymbol{\mu}\| = \|\boldsymbol{\mu} - \boldsymbol{\sigma}\| + \|\boldsymbol{\sigma}\|$. Moreover, $\boldsymbol{\mu}$ will be called *acyclic* if $\boldsymbol{\sigma} = 0$ is the only cycle of $\boldsymbol{\mu}$. It is known [15, 19] that any measure $\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ can be decomposed into cyclic and acyclic parts (see e.g. [15, Proposition 3.8]):

Theorem 1.1. *For any $\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ there exists a cycle $\boldsymbol{\sigma}$ of $\boldsymbol{\mu}$ such that $\boldsymbol{\mu} - \boldsymbol{\sigma}$ is acyclic.*

A curve $\gamma \in \Gamma$ will be called *simple* if γ is injective on $[0, 1)$. The acyclic part of $\boldsymbol{\mu}$ solving (1.1) can be decomposed into measures induced by simple Lipschitz curves (see e.g. [15, Theorem 5.1]):

Theorem 1.2. *If $\boldsymbol{\mu}$ is acyclic, then there exists $\eta \in \mathcal{M}_+(\Gamma)$ such that (1.2a)–(1.2c) hold and for η -a.e. $\gamma \in \Gamma$ there exists simple $\tilde{\gamma} \in \Gamma$ such that $\boldsymbol{\mu}_\gamma = \boldsymbol{\mu}_{\tilde{\gamma}}$.*

In view of Theorems 1.1 and 1.2 it is sufficient to prove the Main Theorem for $\rho = 0$. We provide two different proofs of this result. Both proofs are based on a weak version of Poincaré Lemma: every divergence-free measure $\boldsymbol{\mu}$ in \mathbb{R}^2 can be represented as $\boldsymbol{\mu} = \nabla^\perp f$, where $\nabla^\perp = (-\partial_2, \partial_1)$ and $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is a locally integrable “potential” function with finite total variation.

The first proof (inspired by a remark in [19]) exploits functional analytic tools and relies on Choquet's Theorem (see e.g. [17]), in view of which it suffices to characterize the extreme points of the unit ball in the space of divergence-free measures. Using the weak version of Poincaré Lemma mentioned above, we construct a certain space of functions with finite total variation, denoted by $FV(\mathbb{R}^2)$, which is isometrically isomorphic (via the mapping ∇^\perp) to the space of divergence-free measures. Then it remains to characterize the extreme points of the unit ball in $FV(\mathbb{R}^2)$. In order to do this we apply the Coarea Formula and a fine analysis of sets of finite perimeter using the techniques from [1]. Eventually we show that the extreme points of the unit ball in $FV(\mathbb{R}^2)$ are (normalized) characteristic functions of *simple* sets (see Definition 2.9 and Definition 2.13). Using the results from [1] and [6], we show that the divergence-free measures associated to extreme points are induced by closed simple Lipschitz curves.

In the second proof of the Main Theorem we construct the appropriate measure η directly. First we decompose the “potential” f of μ into a countable family of *monotone* functions $f_k \in FV(\mathbb{R}^2)$ using a modification of the result from [7] (which we prove in the Appendix). Then we construct the desired measure η_k for each component f_k directly using the Coarea Formula and ultimately construct η as the sum of η_k . An advantage of this approach is that it provides a more detailed description of the measure η in view of monotonicity of f_k .

1.1 Applications to rigidity properties of vector-measures

As an application of our decomposition of vector-measures into measures induced by curves, we establish a certain *rigidity property* for vector-valued measures (extending one of the results from [13]). Let $\mathcal{M}_{\text{loc}}(\mathbb{R}^d)$ denote the space of locally finite Borel measures on \mathbb{R}^d . Rigidity properties were introduced in the paper [13] to study fine properties of the *trace* (in the Anzellotti's sense [4]) of bounded, divergence-free vector fields on a class of rectifiable sets. Here we consider the following generalization of [13, Definition 1.1]: recall that, given $\mu \in \mathcal{M}_{\text{loc}}(\mathbb{R}^d; \mathbb{R}^d)$, by polar decomposition (see e.g. [3, Corollary 1.29]) there exists a unique $\tau \in L^1_{\text{loc}}(|\mu|; \mathbb{R}^d)$ with $|\tau(x)| = 1$ for $|\mu|$ -a.e. $x \in \mathbb{R}^d$ such that $\mu = \tau|\mu|$.

Definition 1.3. Let $\mathcal{F} \subset \mathcal{M}_{\text{loc}}(\mathbb{R}^d; \mathbb{R}^d)$. We say that the *linear rigidity property holds for \mathcal{F}* if for any $c > 0$ and for any $\mathbf{v} \in \mathcal{F}$ such that

- (i) $\mathbf{v}(\{x = (x_1, \dots, x_d) \in \mathbb{R}^d : x_d \leq 0\}) = 0$,
 - (ii) $\text{div } \mathbf{v} = 0$ in the distributional sense,
 - (iii) $\tau_d(x) \geq c|\tau(x)|$ for $|\mathbf{v}|$ -a.e. $x \in \mathbb{R}^d$,
- one has that $\mathbf{v} = 0$.

For \mathcal{F} consisting of locally finite vector measures which are absolutely continuous with respect to Lebesgue measure (and have uniformly bounded density) the linear rigidity property was established in [13, Theorem 1.2]. Using the decomposition of vector measure into measures induced by curves, we can prove the following result, which holds true in *every* dimension:

Theorem 1.4. *For any $d \in \mathbb{N}$, the linear rigidity property holds for $\mathcal{F} = \mathcal{M}_{\text{loc}}(\mathbb{R}^d; \mathbb{R}^d)$.*

2 Preliminaries and notation

In this section, we collect some useful and preliminary results and we set some notations that will be used throughout the paper.

2.1 General notation

The d -dimensional Euclidean space will be denoted by \mathbb{R}^d , with $d \geq 1$. Usually, $\Omega \subset \mathbb{R}^d$ stands for a generic open set. The indicator (also characteristic) function of a set A is denoted by $\mathbb{1}_A$ and the complement by A^c .

The Lebesgue measure on \mathbb{R}^d will be \mathcal{L}^d while the k -dimensional Hausdorff measure, for $k \leq d$, will be \mathcal{H}^k . If $(X, \|\cdot\|)$ is a normed space, we will denote by B_1^X the closed unit ball with center 0 and radius 1, i.e.

$$B_1^X := \{x \in X : \|x\| \leq 1\}.$$

If $U \subset X$, the notation \bar{U} will denote the closure of U .

If μ is a measure, the *restriction of μ* to some measurable subset A is $\mu_{\lfloor A}$. The space of p -integrable functions (resp. locally p -integrable functions) on Ω will be denoted in the usual way by $L^p(\Omega)$ (resp. $L^p_{\text{loc}}(\Omega)$), for $1 \leq p \leq +\infty$, and the symbol $\|\cdot\|_p$ will stand for the usual norm in the former space.

If X is a topological space, then $\mathcal{M}(X; \mathbb{R}^d)$ will denote the set of \mathbb{R}^d -valued Borel measures on X . For $d = 1$ let $\mathcal{M}(X) := \mathcal{M}(X; \mathbb{R})$ and let $\mathcal{M}_+(X)$ denote the set of $[0, +\infty]$ -valued Borel measures. For any $\mu \in \mathcal{M}(X; \mathbb{R}^d)$ let $|\mu| \in \mathcal{M}_+(X)$ denote the associated total variation measure. Recall that

$$\|\mu\|_{\mathcal{M}} := |\mu|(X)$$

is a norm on $\mathcal{M}(X; \mathbb{R}^d)$ with respect to which this space is complete (see e.g. [8]).

If X is a locally compact and separable metric space, then $\mathcal{M}(X; \mathbb{R}^d)$ can be identified (by Riesz–Markov–Kakutani Theorem) with the dual of $C_0(\mathbb{R}^d; \mathbb{R}^d)$, where $C_0(\mathbb{R}^d; \mathbb{R}^d)$ is the closure of the set of compactly supported continuous functions $C_c(\mathbb{R}^d; \mathbb{R}^d)$ with respect to the uniform norm. By default in this case we will endow $\mathcal{M}(X; \mathbb{R}^d)$ with the weak-* topology. Note that the total variation norm on $\mathcal{M}(X; \mathbb{R}^d)$ coincides with the norm induced by duality with $C_0(X; \mathbb{R}^d)$ (see e.g. [3, Theorem 1.54]).

Recall also the definition of push-forward of a measure μ on some space X through a Borel map $f: X \rightarrow Y$: we denote by $f_{\#}\mu$ the measure on Y defined by $(f_{\#}\mu)(A) := \mu(f^{-1}(A))$ for any Borel set $A \subset Y$. It is well known that the measure $f_{\#}\mu$ satisfies the following equality for every bounded Borel function $\phi: Y \rightarrow \mathbb{R}$:

$$\int_X \phi(f(x)) d\mu(x) = \int_Y \phi(y) d(f_{\#}\mu)(y).$$

The *divergence* of a vector-valued measure $\mu \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ is understood in the sense of distributions, i.e.

$$\langle \operatorname{div} \mu, \phi \rangle := - \int_{\mathbb{R}^d} \nabla \phi(x) d\mu(x) \quad \text{for all } \phi \in C_c^\infty(\mathbb{R}^d).$$

2.2 BV functions, perimeters, tangents

Let $\Omega \subset \mathbb{R}^d$ be an open set.

Definition 2.1 (BV functions, [3, Definition 3.1]). We say that a function $u \in L^1(\Omega)$ has *bounded variation in Ω* if the distributional derivative of u is representable by a finite Radon measure in Ω , i.e.

$$\int_{\Omega} u \frac{\partial \phi}{\partial x_i} dx = - \int_{\Omega} \phi d(D_i u) \quad \text{for every } \phi \in C_c^\infty(\Omega) \text{ and for every } i = 1, \dots, d$$

for some \mathbb{R}^d -valued vector measure $(D_1 u, \dots, D_d u)$ in Ω . The space of functions of bounded variation in Ω is denoted by $\operatorname{BV}(\Omega)$.

The space $\operatorname{BV}(\Omega)$ is a normed space under the norm

$$\|u\|_{\operatorname{BV}} := \|u\|_1 + \|Du\|_{\mathcal{M}}.$$

Definition 2.2 (Variation, [3, Definition 3.4]). Let $u \in L^1_{\text{loc}}(\Omega)$. The *variation $V(u, \Omega)$* of u in Ω is defined by

$$V(u, \Omega) := \sup \left\{ \int_{\Omega} u(x) \operatorname{div} \phi(x) dx : \phi \in C_c^\infty(\Omega; \mathbb{R}^d), \|\phi\|_{\infty} \leq 1 \right\}.$$

The variation enjoys several properties (see e.g. [3, Remark 3.5]): the map $u \mapsto V(u, \Omega)$ is l.s.c. in the $L^1_{\text{loc}}(\Omega)$ -topology. On the other hand, for fixed $u \in L^1_{\text{loc}}(\Omega)$, it is possible to define $V(u, A)$ for any open set $A \subset \Omega$ and then, via the Carathéodory construction, extend $V(u, \cdot)$ to a Borel measure that will still be denoted by $V(u, \cdot)$. Such measure has finite total variation in Ω if and only if $u \in \text{BV}(\Omega)$ and in this case $V(u, \Omega) = |Du|(\Omega)$ (see [3, Proposition 3.6]). For simplicity, we will simply write $V(u)$ to denote the variation in the full space $V(u, \mathbb{R}^d)$.

We recall that, as for Sobolev spaces, BV functions enjoy some higher integrability properties: these are usually expressed via embedding theorems. For our purposes, the following general result will be needed.

Theorem 2.3 (BV embeddings, [3, Theorem 3.47]). *Let $d \geq 1$. Then for any function $u \in L^1_{\text{loc}}(\mathbb{R}^d)$ satisfying $V(u) < \infty$ there exists a unique constant $m \in \mathbb{R}$ such that*

$$\|u - m\|_{L^{1^*}(\mathbb{R}^d)} \leq \gamma V(u)$$

for some universal constant $\gamma = \gamma(d)$, where

$$1^* := \begin{cases} \frac{d}{d-1}, & d > 1 \\ \infty, & d = 1. \end{cases} \tag{2.1}$$

If $u \in L^1(\mathbb{R}^d)$, then $m = 0$, $u \in \text{BV}(\mathbb{R}^d)$ and hence one has $\|u\|_{L^{1^*}(\mathbb{R}^d)} \leq \gamma V(u)$. In particular, the embedding $\text{BV}(\mathbb{R}^d) \hookrightarrow L^{1^*}(\mathbb{R}^d)$ is continuous.

Definition 2.4 (Sets of finite perimeter, [3, Definition 3.35]). A Lebesgue measurable set $E \subset \mathbb{R}^d$ is said to be of finite perimeter in $\Omega \subset \mathbb{R}^d$ if the variation of $\mathbb{1}_E$ in Ω is finite and the perimeter of E in Ω is

$$P(E, \Omega) := V(\mathbb{1}_E, \Omega).$$

We recall also the *Coarea Formula* for functions of bounded variation (see [3, Theorem 3.40]): for any function $u \in L^1(\Omega)$ it holds

$$V(u, \Omega) := \int_{\mathbb{R}} P(\{x \in \Omega : u(x) > t\}) dt,$$

where the equality is understood in the sense that the right-hand side is finite if and only if the left-hand side is finite and in this case their values coincide and $u \in \text{BV}(\Omega)$.

We also recall the *isoperimetric inequality*, see [3, Theorem 3.46]: for any set $E \subset \mathbb{R}^d$, $d > 1$, of finite perimeter either E or $\mathbb{R}^d \setminus E$ has finite Lebesgue measure and there exists a universal constant $c = c(d)$ such that

$$\min\{\mathcal{L}^d(E), \mathcal{L}^d(\mathbb{R}^d \setminus E)\} \leq c(d)P(E)^{\frac{d}{d-1}}. \tag{2.2}$$

We now recall the definition of *approximate tangent space* to a rectifiable set. Let $k \in \mathbb{N}$ with $k \geq 1$. If μ is a Radon measure on \mathbb{R}^d and $E \subset \mathbb{R}^d$ is a Borel set, we define, for $x \in \mathbb{R}^d$ and $r > 0$,

$$\mu_{x,r}(E) := (\Phi_{x,r})\# \mu(E), \quad \text{where } \Phi_{x,r}(y) := \frac{y-x}{r}.$$

If M is a locally \mathcal{H}^k -rectifiable set, then we define the *approximate tangent space to M at x* , denoted by $\text{Tan}(M, x)$, to be the set of limit points of the measures $r^{-k}\mu_{x,r}$ as $r \downarrow 0$ in the weak- $*$ topology. It is possible to prove (see, e.g. [14, Theorem 10.2]) that for \mathcal{H}^k -a.e. $x \in M$ there exists a unique k -plane π_x such that $\text{Tan}(M, x) = \{\mathcal{H}^k \llcorner \pi_x\}$. We further emphasize that the approximate tangent space to a smooth set is related to the ordinary tangent space, in the sense of differential geometry. More precisely, we have the following:

Proposition 2.5 ([3, Proposition 2.88]). *Let $\phi: \mathbb{R}^k \rightarrow \mathbb{R}^d$ be a one-to-one Lipschitz function and let $D \subset \mathbb{R}^k$ be an \mathcal{L}^k -measurable set. Then $E = \Phi(D)$ satisfies*

$$\text{Tan}(E, x) = \{\mathcal{H}^k \llcorner d\phi_{\phi^{-1}(x)}(\mathbb{R}^k)\} \quad \text{for } \mathcal{H}^k\text{-a.e. } x \in E,$$

where $d\phi$ is the usual differential of ϕ .

2.3 Fine properties of sets of finite perimeter

Given a Lebesgue measurable set $E \subseteq \mathbb{R}^d$, we define the *upper/lower densities at x* by

$$\overline{D}(E, x) := \limsup_{r \rightarrow 0} \frac{\mathcal{L}^d(E \cap B_r(x))}{\mathcal{L}^d(B_r(x))}, \quad \underline{D}(E, x) := \liminf_{r \rightarrow 0} \frac{\mathcal{L}^d(E \cap B_r(x))}{\mathcal{L}^d(B_r(x))},$$

and $D(E, x)$ denotes the common value of $\underline{D}(E, x)$ and $\overline{D}(E, x)$ whenever they are equal. In particular, we will denote by E^t , for $t \in [0, 1]$, the set of points of density t

$$E^t := \{x \in \mathbb{R}^d : D(E, x) = t\}.$$

The *essential exterior of E* is E^0 and the *essential interior of E* is E^1 . Ultimately, the *essential boundary of E* is

$$\partial^e E := \mathbb{R}^d \setminus (E^0 \cup E^1).$$

Following [3, Definition 3.54], we define the *reduced boundary* of a set $E \subset \mathbb{R}^d$ to be the set of points $x \in \text{supp } |D\mathbb{1}_E|$ such that the limit

$$v_E(x) := \lim_{r \downarrow 0} \frac{D\mathbb{1}_E(B_r(x))}{|D\mathbb{1}_E|(B_r(x))}$$

exists in \mathbb{R}^d and satisfies $|v_E(x)| = 1$. We denote by $\mathcal{F}E$ the reduced boundary and the function $v_E : \mathcal{F}E \rightarrow \mathbb{S}^{d-1}$ is called *generalized inner normal to E* .

The celebrated De Giorgi's Theorem can thus be stated as follows:

Theorem 2.6 (De Giorgi, [3, Theorem 3.59]). *Let E be a Lebesgue measurable subset of \mathbb{R}^d of finite perimeter in \mathbb{R}^d . Then $\mathcal{F}E$ is countably $(d-1)$ -rectifiable and $|D\mathbb{1}_E| = \mathcal{H}^{d-1} \llcorner_{\mathcal{F}E}$. In addition, the approximate tangent space to E at x coincide with the orthogonal hyperplane to $v_E(x)$ for \mathcal{H}^{d-1} -a.e. $x \in \mathcal{F}E$, i.e.*

$$\text{Tan}(\mathcal{F}E, x) = v_E^\perp(x).$$

The link between the reduced boundary, the essential boundary and the set of points of density $\frac{1}{2}$ is a remarkable theorem, due to Federer (see [3, Theorem 3.61]):

Theorem 2.7 (Federer). *If $E \subset \mathbb{R}^d$ has finite perimeter, then*

$$\mathcal{F}E \subset E^{\frac{1}{2}} \subset \partial^e E$$

and $\mathcal{H}^{d-1}(\partial^e E \setminus E^{\frac{1}{2}}) = 0$.

In particular, if $E \subset \mathbb{R}^d$ has finite perimeter, then $\mathcal{H}^{d-1}(E^{\frac{1}{2}}) = \mathcal{H}^{d-1}(\mathcal{F}E) < \infty$. However it is known (see e.g. [12, Theorem 6 (2)]) that the condition $\mathcal{H}^{d-1}(E^{\frac{1}{2}}) < \infty$ is not sufficient for $E \subset \mathbb{R}^d$ to have finite perimeter.

Remark 2.8. Taking into account De Giorgi's Theorem 2.6, we can write the Coarea Formula for a function $u \in \text{BV}(\mathbb{R}^d)$ in the following way (see e.g. [3, Formula (3.63)]):

$$|Du|(B) = \int_{\mathbb{R}} \mathcal{H}^{d-1}(\partial^e \{u \geq t\} \cap B) dt \quad \text{for every Borel set } B \subseteq \mathbb{R}^d. \quad (2.3)$$

2.4 Indecomposable and simple sets

From [1] we recall the following definitions.

Definition 2.9 (Decomposable and indecomposable sets). A measurable set $E \subseteq \mathbb{R}^d$ of finite perimeter is called *decomposable* if there exist two measurable sets $A, B \subseteq \mathbb{R}^d$ with strictly positive measure such that $E = A \cup B$, $A \cap B = \emptyset$ and $P(E) = P(A) + P(B)$. A set E which is not decomposable is called *indecomposable*.

The following theorem shows that any set with finite perimeter can be decomposed into at most countably many indecomposable sets [1, Theorem 1].

Theorem 2.10 (Decomposition Theorem). *Let E be a set with finite perimeter in \mathbb{R}^d . Then there exists a unique (up to permutations) at most countable family of pairwise disjoint indecomposable sets $\{E_i\}_{i \in I}$ such that $\mathcal{L}^d(E_i) > 0$, $E = \bigcup_{i \in I} E_i$ and $P(E) = \sum_i P(E_i)$. Moreover, for any indecomposable $F \subseteq E$ with $\mathcal{L}^d(F) > 0$ there exists a unique $j \in I$ such that $\mathcal{L}^d(F \setminus E_j) = 0$.*

Definition 2.11. The sets E_i defined above are called the M -connected components of E . The set $\{E_i\}_{i \in I}$ is denoted by $\mathcal{C}^M(E)$, without loss of generality $I \subseteq \{0, 1, 2, \dots\}$ and $0 \in I$.

By Theorem 2.10 the M -connected components of E are maximal in the following sense: any indecomposable $F \subseteq E$ with $\mathcal{L}^d(F) > 0$ is contained in exactly one of the M -connected components of E , up to Lebesgue negligible subsets. We refer the reader to [1] for a comparison between indecomposability and the topological notion of connectedness.

The statement of Decomposition Theorem can be slightly strengthened with the following simple result from [1, Proposition 3] (see also [1, equation (10), Remark 1]):

Proposition 2.12. *Let $E \subseteq \mathbb{R}^d$ be a set with finite perimeter. Let $\mathcal{C}^M(E) = \{E_i\}_{i \in I}$, where I is at most countable. Then $P(\bigcup_{i \in I_1 \cup I_2} E_i) = P(\bigcup_{i \in I_1} E_i) + P(\bigcup_{i \in I_2} E_i)$ for any disjoint sets $I_1, I_2 \subseteq I$.*

Definition 2.13 (Holes, saturation, simple sets). Let E be an indecomposable set. Any M -connected component of $\mathbb{R}^d \setminus E$ with finite measure is called a *hole* of E . The saturation $\text{sat}(E)$ of E is defined as union of E and all its holes. The set E is called *saturated* if $\text{sat}(E) = E$. Any indecomposable saturated subset of \mathbb{R}^d is called *simple*.

Observe that simple sets are necessarily of finite perimeter; for $d > 1$, the only simple set E with $\mathcal{L}^d(E) = \infty$ is $E = \mathbb{R}^d$.

2.5 Further facts on indecomposable and simple sets

We finally collect in this paragraph some useful, different characterization of indecomposable and of simple sets. We begin by considering indecomposable sets and we present a lemma which will be useful later.

Lemma 2.14. *Let $F \subseteq E \subset \mathbb{R}^d$ be two sets of finite perimeter. Then*

$$\partial^e F \subseteq \partial^e E \text{ mod } \mathcal{H}^{d-1} \iff \mathcal{H}^{d-1}(\partial^e F \cap E^1) = 0. \quad (2.4)$$

Furthermore, if E is indecomposable and one (hence both) of (2.4) holds true, then $\mathcal{L}^d(F) = 0$ or $\mathcal{L}^d(E \setminus F) = 0$.

Proof. Let us show first the equivalence. First notice that from $F \subset E$, together with the monotonicity of the Lebesgue measure, we deduce $E^0 \subset F^0$. Hence, the following equalities hold modulo \mathcal{H}^{d-1} :

$$\partial^e F = (\partial^e F \cap E^1) \cup (\partial^e F \cap E^0) \cup (\partial^e F \cap \partial^e E) = (\partial^e F \cap E^1) \cup (\partial^e F \cap \partial^e E). \quad (2.5)$$

From (2.5) we easily get the equivalence: on the one hand, if $\partial^e F \subseteq \partial^e E$, then we must have

$$(\partial^e F \cap E^1) \cup (\partial^e F \cap \partial^e E) = \partial^e F \subset \partial^e E \text{ mod } \mathcal{H}^{d-1}$$

and therefore the only possibility is that $\mathcal{H}^{d-1}(\partial^e F \cap E^1) = 0$. Viceversa, if $\mathcal{H}^{d-1}(\partial^e F \cap E^1) = 0$, from (2.5) we get

$$\partial^e F = (\partial^e F \cap E^1) \cup (\partial^e F \cap \partial^e E) = (\partial^e F \cap \partial^e E) \subset \partial^e E \text{ mod } \mathcal{H}^{d-1},$$

which is what we wanted. Let us now turn to prove that there are no non-trivial subsets $F \subset E$ satisfying conditions (2.4) if E is indecomposable. Let $F \subseteq E$ be a set of finite perimeter with $\partial^e F \subseteq \partial^e E \text{ mod } \mathcal{H}^{d-1}$. Then it is easy to check that

$$\partial^e(E \setminus F) \subset \partial^e E \setminus \partial^e F \text{ mod } \mathcal{H}^{d-1}. \quad (2.6)$$

Let us show (2.6): on the one hand, it is clear that $\partial^e(E \setminus F) \subset \partial^e E$. On the other hand, we show that

$$\mathcal{H}^{d-1}(\partial^e(E \setminus F) \cap \partial^e F) = 0.$$

Indeed, \mathcal{H}^{d-1} -a.e. $x \in \partial^e(E \setminus F)$ satisfies $D(E \setminus F, x) = \frac{1}{2}$ by De Giorgi–Federer’s Theorem 2.7. Similarly, \mathcal{H}^{d-1} -a.e. $x \in \partial^e F$ satisfies $D(F, x) = \frac{1}{2}$: thus, for \mathcal{H}^{d-1} -a.e. $x \in (\partial^e(E \setminus F) \cap \partial^e F)$ we have

$$D(E, x) = D(F, x) + D(E \setminus F, x) = \frac{1}{2} + \frac{1}{2} = 1,$$

which contradicts the fact that $\partial^e(E \setminus F) \cap \partial^e F \subseteq \partial^e E$. Having shown (2.6), we get, taking Hausdorff measure of both sides,

$$\mathcal{H}^{d-1}(\partial^e(E \setminus F)) \leq \mathcal{H}^{d-1}(\partial^e E \setminus \partial^e F) = \mathcal{H}^{d-1}(\partial^e E) - \mathcal{H}^{d-1}(\partial^e F)$$

or equivalently

$$P(E \setminus F) + P(F) \leq P(E).$$

The other inequality is trivial by subadditivity of the perimeter, hence

$$P(E \setminus F) + P(F) = P(E),$$

which implies the desired conclusion, being E indecomposable. □

Proposition 2.15 (Dolzmann–Müller). *A set $E \subset \mathbb{R}^d$ of finite perimeter is indecomposable if and only if for any $u \in \text{BV}_{\text{loc}}(\mathbb{R}^d)$ with $V(u) < \infty$ the following implication holds true:*

$$|Du|(E^1) = 0 \iff \exists c \in \mathbb{R}: u(x) = c \text{ for a.e. } x \in E.$$

Proof. Let E be indecomposable and $u \in \text{BV}_{\text{loc}}(\mathbb{R}^d)$ a function with $|Du|(E^1) = 0$. Set $v := u \mathbb{1}_E \in \text{BV}_{\text{loc}}(\mathbb{R}^d)$ and observe that, by Coarea Formula (2.3), we have

$$|Dv|(E_1) = \int_{\mathbb{R}} \mathcal{H}^{d-1}(\partial^e(\{u \geq t\} \cap E) \cap E^1) dt \leq \int_{\mathbb{R}} \mathcal{H}^{d-1}((\partial^e\{u \geq t\} \cup \partial^e E) \cap E^1) dt,$$

where we have used the elementary inclusion $\partial^e(\{u \geq t\} \cap E) \subset \partial^e\{u \geq t\} \cup \partial^e E$. Taking into account that $\partial^e E \cap E^1 = \emptyset$, we can continue the above chain of inequalities as follows:

$$|Dv|(E_1) \leq \int_{\mathbb{R}} \mathcal{H}^{d-1}(\partial^e\{u \geq t\} \cup \partial^e E) \cap E^1) dt = \int_{\mathbb{R}} \mathcal{H}^{d-1}(\partial^e\{u \geq t\} \cap E^1) dt = |Du|(E^1) = 0$$

by the Coarea Formula applied on u . Thus we have $\mathcal{H}^{d-1}(\partial^e(\{u \geq t\} \cap E) \cap E^1) = 0$ for a.e. $t \in \mathbb{R}$. Now we apply Lemma 2.14 to $F := \{u \geq t\} \cap E \subset E$: since E is indecomposable, we deduce

$$\mathcal{L}^d(\{u \geq t\} \cap E) = 0 \quad \text{or} \quad \mathcal{L}^d(\{u < t\} \cap E) = 0$$

for a.e. $t \in \mathbb{R}$, which is easily seen to be equivalent to u being constant in E . □

Concerning simple sets, we want to prove that simplicity for a set E with $|E| < \infty$ is equivalent to indecomposability both of E and of E^c . We need the following preliminary:

Lemma 2.16. *Let $E \subset \mathbb{R}^d$, $d > 1$, be a set of finite perimeter and assume $\mathcal{L}^d(E) = +\infty$. Let $\mathcal{C}^M(E) = \{E_i\}_{i \in I}$ be the family of its indecomposable components. Then there exists a unique $j \in I$ such that $\mathcal{L}^d(E_j) = +\infty$.*

Proof. The statement is a consequence of the convergence of the series $\sum_{i \in I} P(E_i)$ and of the isoperimetric inequality. Indeed, by contradiction, let us assume that for every $i \in I$ it holds $\mathcal{L}^d(E_i) < \infty$. In particular, for every $i \in I$ it has to be $\mathcal{L}^d(\mathbb{R}^d \setminus E_i) = +\infty$ and hence, by the isoperimetric inequality we would get

$$\mathcal{L}^d(E) = \mathcal{L}^d\left(\bigcup_{i \in I} E_i\right) \leq \sum_{i \in I} \mathcal{L}^d(E_i) \leq C_d \sum_{i \in I} P(E_i) \leq C_d P(E) < \infty,$$

which is absurd. Hence there must exist at least one element $j \in I$ such that $\mathcal{L}^d(E_j) = +\infty$. Let us now prove the uniqueness of j : assume that there exists $j_1, j_2 \in I$ such that $\mathcal{L}^d(E_{j_1}) = \mathcal{L}^d(E_{j_2}) = +\infty$. Since

$$P(E_{j_1}) + P(E_{j_2}) \leq \sum_{i \in I} P(E_i) = P(E) < +\infty,$$

we have that $P(E_{j_1}) < \infty$ and $P(E_{j_2}) < \infty$. Furthermore, by the definition of indecomposable components the sets E_{j_1} and E_{j_2} are (essentially) disjoint, i.e. $E_{j_1} \cap E_{j_2} = \emptyset \bmod \mathcal{L}^d$. In particular, we deduce that

$$E_{j_2} \subset E_{j_1}^c \implies +\infty = \mathcal{L}^d(E_{j_2}) \leq \mathcal{L}^d(E_{j_1}^c)$$

so $\mathcal{L}^d(E_{j_1}) = \mathcal{L}^d(E_{j_1}^c) = +\infty$, which is a contradiction with the fact that $P(E_{j_1}) < \infty$ (and the isoperimetric inequality). Thus $j \in I$ has to be unique and the proposition is proved. \square

We are now ready to present the following characterization of simple sets:

Proposition 2.17. *Let $E \subset \mathbb{R}^d$, $d > 1$, be a set with finite positive measure, $\mathcal{L}^d(E) \in (0, +\infty)$. The set E is simple if and only if E and E^c are indecomposable.*

Proof. Assume that E is simple. Then it is clearly indecomposable; thus it is sufficient to show that E^c is indecomposable. Since $\mathcal{L}^d(E) \in (0, +\infty)$, we have $|E^c| = +\infty$. Letting $\mathcal{C}^M(E^c) := \{U_i\}_{i \in I}$ be the indecomposable components of E^c , by Lemma 2.16 there exists one and only one $j \in I$ such that $\mathcal{L}^d(U_j) = +\infty$. So if $\#I > 1$, the other components $\{U_i\}_{i \neq j}$ of E^c must have finite measure, i.e. they are holes of E . This contradicts the simplicity of the set E : hence $\#I = 1$ and E^c is thus indecomposable.

To prove the converse, let us now assume that $\mathcal{L}^d(E) \in (0, +\infty)$ and both E and E^c are indecomposable and we want to prove that E has no holes. By definition a hole of E is a indecomposable component of E^c of finite measure. Being indecomposable, E^c has a unique indecomposable component, which coincides with itself. But $\mathcal{L}^d(E^c) = \infty$ since E has finite measure, and this implies that E has no holes, hence it is simple. \square

Remark 2.18. The necessary condition in Proposition 2.17 holds even if $\mathcal{L}^d(E) = +\infty$ if $d > 1$: indeed, as already observed, if E is simple and $\mathcal{L}^d(E) = +\infty$, then $E = \mathbb{R}^d$, hence the claim is trivial, being the empty set indecomposable.

2.6 Jordan curves in \mathbb{R}^2

In this subsection we collect some results about Jordan curves in the plane \mathbb{R}^2 .

Definition 2.19. A set $C \subseteq \mathbb{R}^2$ is called a *Jordan curve* if $C = \gamma([a, b])$ for some $a, b \in \mathbb{R}$ (with $a < b$) and some continuous map $\gamma: [a, b] \rightarrow \mathbb{R}^2$, one-to-one on $[a, b)$ and such that $\gamma(a) = \gamma(b)$.

Remark 2.20. If $\mathcal{H}^1(C) < \infty$, then γ can be chosen in such a way that it is Lipschitz (see [1, Lemma 3]), and in this case Γ is called a *rectifiable Jordan curve*.

Without any loss of generality, when dealing with Jordan curves, we will always suppose $[a, b] = [0, 1]$. The following result, borrowed from [1], will play a crucial role in the paper.

Theorem 2.21 ([1, Theorem 7]). *Let $E \subseteq \mathbb{R}^2$ be a simple set with $\mathcal{L}^2(E) \in (0, +\infty)$. Then E is essentially bounded and $\partial^e E$ is equivalent, up to an \mathcal{H}^1 -negligible set, to a rectifiable Jordan curve. Conversely, $\text{int}(C)$ is a simple set for any rectifiable Jordan curve C .*

Here $\text{int}(C)$ denotes the bounded connected component of $\mathbb{R}^2 \setminus C$, given by the celebrated Jordan Theorem (see e.g. [10, Proposition 2B.1]).

2.7 Extreme points and Choquet theory

In this subsection we recall the main facts about extreme points of compact, convex sets in normed spaces. Standard references are [17, 18].

Let X be a topological vector space and let $K \subset X$. A point $x \in K$ is an *extreme point* of K if

$$y, z \in K: t \in [0, 1], x = (1-t)y + tz \implies x = y = z.$$

The set of extreme points of K will be denoted by $\text{ext } K$.

Remark 2.22 (The set of extreme points is a Borel set). Recall that $\text{ext } K$ is a Borel subset of K if the topology of X is induced by some metric ρ . Indeed, the set $K \setminus \text{ext } K$ can be written as $\bigcup_n C_n$, where

$$C_n := \left\{ \frac{y+z}{2} : y, z \in K, \rho(y, z) \geq \frac{1}{n} \right\} \quad \text{for every } n \in \mathbb{N} \text{ with } n \geq 1.$$

Given that each set C_n is a closed subset of X , we conclude that $\text{ext } K$ is Borel.

In the case K is a convex, compact set the (closed convex hull of the) set of extreme points of K coincides with the set K itself, as the following theorem states:

Theorem 2.23 (Krein–Milman). *If $K \subset X$ is non-empty, compact, convex set, then $K = \overline{\text{co}}(\text{ext}(K))$.*

We recall that, in a vector space X , the convex hull $\text{co}(A)$ of a set $A \subset X$ is the intersection of all convex sets containing A .

Definition 2.24 (Vector-valued integration). Let μ be a measure on a non-empty set Q . Let $f: Q \rightarrow X$ be an X -valued function such that $(\Lambda f)(q) := \Lambda(f(q))$ is μ -integrable for every $\Lambda: X \rightarrow \mathbb{R}$ linear and continuous. If there exists $y \in X$ such that

$$\Lambda y = \int_Q \Lambda f \, d\mu$$

for every $\Lambda: X \rightarrow \mathbb{R}$ linear and continuous, then we say that y is the *integral of f with respect to μ* and we write

$$\int_Q f \, d\mu := y.$$

Theorem 2.25 (Representation of the convex hull). *Let $Q \subset X$ be a compact set and let $H := \text{co}(Q)$. Assume that $\overline{H} = \overline{\text{co}}(Q)$ is compact as well. Then*

$$y \in \overline{H} \iff \exists \mu \in \mathcal{P}(Q): y = \int_Q x \, d\mu(x).$$

One of the fundamental results in functional analysis and convex analysis is the following theorem, which can be obtained combining Theorem 2.23 with Theorem 2.25:

Theorem 2.26 (Choquet [17]). *Let X be a metrizable topological vector space and let $\emptyset \neq K \subset X$ be convex and compact. Then for any point $x \in K$ there exists a Borel probability measure μ on X (possibly depending on x), which is concentrated on $\text{ext } K$ and satisfies*

$$x = \int_{\text{ext } K} y \, d\mu(y),$$

where the integral is understood in the sense of Definition 2.24, i.e. explicitly

$$\Lambda(x) = \int_{\text{ext } K} \Lambda(y) \, d\mu(y) \quad \text{for every } \Lambda: X \rightarrow \mathbb{R} \text{ linear and continuous.}$$

Remark 2.27 (Extreme points and isomorphisms). Let $(Y, \|\cdot\|_Y)$ be a normed space. Suppose that $\phi: X \rightarrow Y$ is a linear isomorphism between X and Y . Then for any set $A \subset X$ it holds

$$\text{ext } \phi(A) = \phi(\text{ext } A).$$

Indeed, consider $z \in \text{ext } \phi(A)$. Being ϕ one-to-one and onto, there exists a unique $a \in A$ such that $z = \phi(a)$. We want to prove that $a \in \text{ext } A$: let

$$f, g \in A: \lambda f + (1 - \lambda)g = a.$$

Since ϕ is linear, we have

$$\lambda\phi(f) + (1 - \lambda)\phi(g) = \phi(a) = z,$$

but z was an extreme point, hence $\phi(f) = \phi(g) = z$, which implies $f = g = a$, i.e. a is also extreme. An analogous proof shows that if $b \in \text{ext } A$, then $\phi(b)$ is also extreme of $\phi(A)$.

3 Extreme points of the unit ball of BV functions in \mathbb{R}^d

Let us consider the Banach space $X := (\text{BV}(\mathbb{R}^d), \|\cdot\|_{\text{BV}})$ and let us characterize extreme points of B_1^X , the closed unit ball.

Proposition 3.1 (Extreme points of unit ball in $\text{BV}(\mathbb{R}^d)$). *A function $f \in X$ is an extreme point of B_1^X if and only if there exists an indecomposable set $E \subset \mathbb{R}^d$ of positive, finite perimeter and positive, finite measure and a constant $\sigma \in \{\pm 1\}$ such that*

$$f(x) = \sigma \frac{\mathbb{1}_E(x)}{\|\mathbb{1}_E\|_{\text{BV}}}, \quad \mathcal{L}^d\text{-a.e. } x \in \mathbb{R}^d.$$

We will need the following auxiliary lemma.

Lemma 3.2. *Let $f \in \text{BV}_{\text{loc}}(\mathbb{R}^d)$ and let, for any $\lambda \in \mathbb{R}$,*

$$f_\lambda^+ := \max\{f - \lambda, 0\} \quad \text{and} \quad f_\lambda^- := f - f_\lambda^+ = \min\{\lambda, f\}.$$

Then for every open set $\Omega \in \mathbb{R}^d$ it holds

$$|Df|(\Omega) = |Df_\lambda^+|(\Omega) + |Df_\lambda^-|(\Omega). \quad (3.1)$$

Proof. To begin we consider the case $\lambda = 0$ and we notice that, in this case, the decomposition of f into $f_\lambda^+ + f_\lambda^-$ coincides with the standard decomposition into positive/negative part:

$$f_0^+ = f^+ \quad \text{and} \quad f_0^- := -f^-.$$

If $f \in W_{\text{loc}}^{1,1}(\mathbb{R}^d)$, then, fixed $\Omega \in \mathbb{R}^d$, it is enough to apply the Chain Rule Theorem [9, Section 4.2.2, Theorem 4 (iii)]. For the general case, consider a sequence $(f_n)_n \subset W^{1,1}(\Omega) \cap C^\infty(\Omega)$ with $f_n \rightarrow f$ strongly in $L^1(\Omega)$ and $|Df_n|(\Omega) \rightarrow |Df|(\Omega)$ (such a sequence can be obtained using Anzellotti–Giaquinta’s Theorem, see e.g. [9, Section 5.2.2]). Then for every $n \in \mathbb{N}$ it holds

$$|Df_n|(\Omega) = |Df_n^+|(\Omega) + |Df_n^-|(\Omega),$$

hence

$$\begin{aligned} |Df|(\Omega) &= \lim_n |Df_n|(\Omega) = \lim_n \inf |Df_n|(\Omega) \\ &= \lim_n \inf (|Df_n^+|(\Omega) + |Df_n^-|(\Omega)) \\ &\geq \lim_n \inf |Df_n^+|(\Omega) + \lim_n \inf |Df_n^-|(\Omega) \\ &\geq |Df^+|(\Omega) + |Df^-|(\Omega), \end{aligned}$$

where the last inequality is a consequence of the l.s.c. of the total variation, since $f_n^+ \rightarrow f^+$ and $f_n^- \rightarrow f^-$ in $L^1(\Omega)$. The statement is thus proved for $\lambda = 0$; to obtain the general case, we can apply the above claim to the function $g := f - \lambda \in \text{BV}_{\text{loc}}(\mathbb{R}^d)$, noticing that

$$g^+ = f_\lambda^+, \quad g^- = \lambda - f_\lambda^-$$

and

$$Dg = Df, \quad Dg^+ = Df_\lambda^+, \quad Dg^- = -Df_\lambda^-,$$

whence (3.1). \square

We now show the following lemma, which ensures that extreme points lie in the set of normalized indicators of sets of finite perimeter. Recall that for any set of finite perimeter $E \subset \mathbb{R}^d$, either E or E^c has finite Lebesgue measure by the isoperimetric inequality (2.2).

Lemma 3.3. *Let $f \in X$ be an extreme point of the closed unit ball B_1^X . Then there exist a set $E \subseteq \mathbb{R}^d$ with positive, finite perimeter and positive, finite measure $\mathcal{L}^d(E) < \infty$ and a constant $\sigma \in \{\pm 1\}$ such that $f = \sigma \frac{1}{\|\mathbb{1}_E\|_{\text{BV}}} \mathbb{1}_E$.*

Proof. We divide the proof into three steps.

Step 1. Any extreme function has constant sign. Let $f \in X$ be extreme of B_1^X . Then, by standard facts, we have necessarily $\|f\|_{\text{BV}} = 1$. Let us decompose f into positive and negative part as $f = f^+ - f^-$. By the very definition of Lebesgue integral for signed functions we have that

$$\|f\|_1 = \|f^+\|_1 + \|f^-\|_1,$$

while, by Lemma 3.2 with $\lambda = 0$, we have that

$$\|Df\|_{\mathcal{M}} = \|Df^+\|_{\mathcal{M}} + \|Df^-\|_{\mathcal{M}}.$$

Adding up the two equalities, we find out that

$$\|f\|_{\text{BV}} = \|f^+\|_{\text{BV}} + \|f^-\|_{\text{BV}}$$

and this can be used to decompose f into a convex linear combination of two *signed* functions with unit BV norm

$$f = \|f^+\|_{\text{BV}} \cdot \frac{f^+}{\|f^+\|_{\text{BV}}} + \|f^-\|_{\text{BV}} \cdot \frac{-f^-}{\|f^-\|_{\text{BV}}}.$$

Hence any extremal point is necessarily a function with constant sign and, without any loss of generality, we consider $f \geq 0$.

Step 2. Any extreme function attains at most one non-zero value. We now would like to prove that $f(x) \in \{0, \alpha\}$ for some $\alpha > 0$ for \mathcal{L}^d -a.e. $x \in \mathbb{R}^d$. Suppose by contradiction that it is not true: hence, there exist two points x_1, x_2 such that $f(x_1) \neq 0, f(x_2) \neq 0$ and also $f(x_1) \neq f(x_2)$. Without any loss of generality, suppose $f(x_1) < f(x_2)$. We can also assume that x_1, x_2 are Lebesgue points of f (this property being satisfied almost everywhere by standard facts). Consider an arbitrary $\lambda \in (f(x_1), f(x_2))$ and define the non-negative functions

$$f_\lambda^+ := \max\{f - \lambda, 0\} \quad \text{and} \quad f_\lambda^- := f - f_\lambda^+ = \min\{\lambda, f\}.$$

By Lemma 3.2 we deduce

$$\|Df\|_{\mathcal{M}} = \|D(f_\lambda^+)\|_{\mathcal{M}} + \|D(f_\lambda^-)\|_{\mathcal{M}},$$

while from the pointwise equality $f_\lambda^+ + f_\lambda^- = f$, together with non-negativity, we get

$$\|f\|_1 = \|f_\lambda^+\|_1 + \|f_\lambda^-\|_1,$$

and thus

$$\|f\|_{\text{BV}} = \|f_\lambda^+\|_{\text{BV}} + \|f_\lambda^-\|_{\text{BV}}.$$

In particular, we can decompose

$$f = \|f_\lambda^+\|_{\text{BV}} \cdot \frac{f_\lambda^+}{\|f_\lambda^+\|_{\text{BV}}} + \|f_\lambda^-\|_{\text{BV}} \cdot \frac{f_\lambda^-}{\|f_\lambda^-\|_{\text{BV}}}. \tag{3.2}$$

$\in B_1^X \qquad \qquad \qquad \in B_1^X$

Notice that the choice of $\lambda \in (f(x_1), f(x_2))$ together with the fact that $f(x_1) \neq 0 \neq f(x_2)$ grant that the decomposition (3.2) is non-trivial and well-posed, in the sense that:

(1) the functions f_λ^\pm are linearly independent: if $af_\lambda^+ + bf_\lambda^- = 0$ for $a, b \in \mathbb{R}$, then evaluating at x_1 we deduce

$$bf(x_1) = 0 \implies b = 0$$

and evaluation at x_2 yields

$$a(f(x_2) - \lambda) = 0 \implies a = 0,$$

(2) we have $\|f_\lambda^\pm\|_{\text{BV}} > 0$: indeed, if it were e.g. $\|f_\lambda^-\|_{\text{BV}} = 0$, then $f = f_\lambda^+$ a.e. which means

$$f(x) \geq \lambda \quad \text{for } \mathcal{L}^d\text{-a.e. } x \in \mathbb{R}^d. \tag{3.3}$$

On the other hand, x_1 is a Lebesgue point of f with Lebesgue value $f(x_1) < \lambda$, so by definition

$$\lambda > f(x_1) = \lim_{r \rightarrow 0} \int_{B_r(x_1)} f(y) dy \geq \lim_{r \rightarrow 0} \int_{B_r(x_1)} \lambda dy = \lambda$$

(where “ \geq ” follows from (3.3)), which is a contradiction.

Thus (3.2) is a non-trivial, convex decomposition of f which contradicts extremality: the contradiction stems from the assumption that there exists two points x_1, x_2 such that $f(x_1) \neq 0, f(x_2) \neq 0$ and $f(x_1) \neq f(x_2)$. So we must have $f(x) \in \{0, \alpha\}$ for a.e. x for some $\alpha > 0$.

Step 3. Any extreme function is an indicator function. From Step 2 we immediately deduce

$$f(x) = \alpha \mathbb{1}_E, \quad \text{where } E := \{x \in \mathbb{R}^d : f(x) = \alpha\}.$$

The set E has finite perimeter because $f \in BV(\mathbb{R}^d)$ and, being $\|f\|_{BV} = 1$, we deduce that necessarily $\alpha = \|\mathbb{1}_E\|_{BV}^{-1}$. This concludes the proof. \square

We can now prove the main result of this section.

Proof of Proposition 3.1. We split the proof into two steps.

Sufficiency. Let $E \subset \mathbb{R}^d$ be a set of positive, finite perimeter and assume it is indecomposable. Let $c = \frac{1}{P(E)}$ and let us prove that $f := c \mathbb{1}_E$ is an extreme point of B_1^X . Assume that for some functions $g, h \in B_1^X$ and $\lambda \in [0, 1]$ we can write

$$f = \lambda g + (1 - \lambda)h$$

and let us prove that necessarily $g = c \mathbb{1}_E$ and $h = c \mathbb{1}_E$. Since $\|f\|_{BV} = 1$, we have that

$$1 \leq \lambda \|g\|_{BV} + (1 - \lambda) \|h\|_{BV}$$

and we claim that actually equality holds. If it were

$$1 < \lambda \|g\|_{BV} + (1 - \lambda) \|h\|_{BV},$$

then we would get, being $f, g \in B_1^X$,

$$1 < \lambda \|g\|_{BV} + (1 - \lambda) \|h\|_{BV} \leq \lambda + (1 - \lambda) = 1,$$

a contradiction. In a complete similar way, one can prove that $\|g\|_{BV} = 1 = \|h\|_{BV}$. All in all, we can represent

$$\mathbb{1}_E = \phi + \psi \tag{3.4}$$

with

$$\|\mathbb{1}_E\|_{BV} = \|\phi\|_{BV} + \|\psi\|_{BV} \tag{3.5}$$

being $\phi = c^{-1} \lambda g$ and $\psi = c^{-1} (1 - \lambda) h$. Notice that ϕ, ψ have the same sign a.e., otherwise we would have

$$\int_{\mathbb{R}^d} |\phi(x) + \psi(x)| dx < \int_{\mathbb{R}^d} |\phi(x)| + |\psi(x)| dx,$$

which would yield

$$\|\phi + \psi\|_{BV} = \|\phi + \psi\|_1 + \|D(\phi + \psi)\|_{\mathcal{M}} < \|\phi\|_1 + \|\psi\|_1 + \|D\phi\|_{\mathcal{M}} + \|D\psi\|_{\mathcal{M}} = \|\phi\|_{BV} + \|\psi\|_{BV},$$

contradicting (3.5). Since $\mathbb{1}_E = \phi + \psi$, we have $\phi, \psi \geq 0$ a.e. and therefore

$$\phi = \psi = 0 \quad \text{a.e. on } E^c. \tag{3.6}$$

Notice furthermore that it holds

$$|D\mathbb{1}_E| = |D\phi| + |D\psi| \quad \text{as measures in } \mathbb{R}^d. \tag{3.7}$$

Indeed, by (3.4) and the triangle inequality we get $|D\mathbb{1}_E| \leq |D\phi + D\psi| \leq |D\phi| + |D\psi|$; the converse inequality then follows exploiting (3.5). In particular, computing (3.7) on the Borel set E^1 it follows

$$|D\phi|(E^1) + |D\psi|(E^1) = |D\mathbb{1}_E|(E^1) = 0,$$

where the last equality follows from De Giorgi’s Theorem. Hence

$$|D\phi|(E^1) = 0 = |D\psi|(E^1).$$

By Proposition 2.15 and by the indecomposability of E , there exist constants $c_1, c_2 \in \mathbb{R}$ such that

$$\phi(x) = c_1, \quad \psi(x) = c_2 \quad \text{a.e. in } E. \tag{3.8}$$

In particular, combining (3.8) together with (3.6), we obtain

$$\phi(x) = c_1 \mathbb{1}_E(x), \quad \psi(x) = c_2 \mathbb{1}_E(x) \quad \text{a.e. in } \mathbb{R}^d$$

and this in turn implies that

$$g = \alpha \mathbb{1}_E(x)$$

for some $\alpha \in \mathbb{R}$. Being $\|g\|_{\text{BV}} = 1$, we obtain that the constant has to be

$$\alpha = \frac{1}{P(E)}.$$

One can argue similarly with h and the conclusion is now achieved: we have proved that the only convex combination of elements in B_1^X representing f is the trivial one, i.e. f is an extreme point of B_1^X .

Necessity. By Lemma 3.3, we can already infer that there exist a set $E \subseteq \mathbb{R}^d$ with finite perimeter and $\sigma \in \{\pm 1\}$ such that $f = \sigma \frac{1}{\|\mathbb{1}_E\|_{\text{BV}}} \mathbb{1}_E$ a.e. with respect to the Lebesgue measure. Now we prove that E is indecomposable. Suppose by contradiction that E is a decomposable set, i.e. $E = A \cup B$ with $A \cap B = \emptyset$ and $P(E) = P(A) + P(B)$. Since by additivity of the Lebesgue measure it holds $\mathcal{L}^d(E) = \mathcal{L}^d(A) + \mathcal{L}^d(B)$, we have

$$\|\mathbb{1}_E\|_{\text{BV}} = \|\mathbb{1}_A\|_{\text{BV}} + \|\mathbb{1}_B\|_{\text{BV}}.$$

Hence

$$\frac{1}{\|\mathbb{1}_E\|_{\text{BV}}} \mathbb{1}_E = \frac{\|\mathbb{1}_A\|_{\text{BV}}}{\|\mathbb{1}_E\|_{\text{BV}}} \underbrace{\frac{\mathbb{1}_A}{\|\mathbb{1}_A\|_{\text{BV}}}}_{\in B_1^X} + \frac{\|\mathbb{1}_B\|_{\text{BV}}}{\|\mathbb{1}_E\|_{\text{BV}}} \underbrace{\frac{\mathbb{1}_B}{\|\mathbb{1}_B\|_{\text{BV}}}}_{\in B_1^X}$$

is a representation of f as a non-trivial convex combination of elements of B_1^X , contradicting extremality. Therefore if $\frac{\mathbb{1}_E}{P(E)}$ is an extreme point then E has to be indecomposable. □

4 Extreme points of the unit ball of FV functions in \mathbb{R}^d

Definition 4.1. We define the space $\text{FV}(\mathbb{R}^d)$ as the function space

$$\text{FV}(\mathbb{R}^d) := \{f \in L^{1^*}(\mathbb{R}^d) : V(f) < +\infty\}.$$

We recall $V(f) = V(f, \mathbb{R}^d)$ is the variation of a locally integrable function, see Definition 2.2, while 1^* is defined in (2.1).

Remark 4.2. It is easy to see that $\text{BV}(\mathbb{R}^d) \subset \text{FV}(\mathbb{R}^d) \subset \text{BV}_{\text{loc}}(\mathbb{R}^d)$ and both inclusion are strict. Indeed, any constant function is certainly locally integrable with zero total variation, but it is not in $L^p(\mathbb{R}^d)$ for any p . On the other hand, the function $f: \mathbb{R}^d \rightarrow \mathbb{R}$ defined by

$$f(x) = g(|x|), \quad \text{where } g(s) := \min\left\{1, \frac{1}{s^d}\right\},$$

is in $\text{FV}(\mathbb{R}^d)$ but not in $\text{BV}(\mathbb{R}^d)$. Let us verify this claim:

$$\int_{\mathbb{R}^d} f(x) \, dx = C_d \int_0^{+\infty} g(s) s^{d-1} \, ds = +\infty,$$

while taking into account that $1^* := \frac{d}{d-1}$ we have

$$\int_{\mathbb{R}^d} |f(x)|^{1^*} dx = C_d \int_0^{+\infty} g(s)^{\frac{d}{d-1}} s^{d-1} ds = \widetilde{C}_d \left(1 + \int_1^{+\infty} s^{\frac{1-2d}{d-1}} ds \right) < +\infty.$$

Notice that the variation of f is finite

$$V(f) = C_d \int_1^{+\infty} \frac{1}{s^{d+1}} s^{d-1} ds = C_d \int_1^{+\infty} \frac{1}{s^2} < +\infty.$$

We now prove that the map $\|\cdot\|_{\text{FV}} : \text{FV}(\mathbb{R}^d) \ni f \mapsto \|f\|_{\text{FV}} = V(f)$ gives to $\text{FV}(\mathbb{R}^d)$ the structure of a normed space.

Proposition 4.3. *The space $Y := (\text{FV}(\mathbb{R}^d), \|\cdot\|_{\text{FV}})$ is a normed space.*

Proof. Positivity and 1-homogeneity are clear from the definition of $\|\cdot\|_{\text{FV}}$ and the triangle inequality as well. We have to prove only definiteness: for, let $f \in \text{FV}(\mathbb{R}^d)$ with

$$\|f\|_{\text{FV}} = V(f) = 0.$$

Applying Theorem 2.3, we deduce that there exist $m \in \mathbb{R}$ and a constant $\gamma > 0$ such that

$$\|f - m\|_{1^*} \leq \gamma V(f),$$

whence $\|f - m\|_{1^*} = 0$ and $f = m$ almost everywhere. Being $f \in L^{1^*}(\mathbb{R}^d)$, the only possibility is that $m = 0$, hence the proposition is proved. \square

We now aim at characterizing extreme points of B_1^Y , the closed unit ball in Y . Observe that, if f is the characteristic function of a measurable set A , variation and perimeter coincide, i.e.

$$V(\mathbb{1}_A) = P(A, \mathbb{R}^d) = P(A).$$

Proposition 4.4 (Extreme points of unit ball in $\text{FV}(\mathbb{R}^d)$). *A function $f \in Y$ is an extreme point of B_1^Y if and only if there exist a simple set $E \subset \mathbb{R}^d$ of positive, finite perimeter and a constant $\sigma \in \{\pm 1\}$ such that*

$$f(x) = \sigma \frac{\mathbb{1}_E(x)}{P(E)}, \quad \mathcal{L}^d\text{-a.e. } x \in \mathbb{R}^d.$$

Proof. The proof is divided into two steps.

Sufficiency. Let $E \subset \mathbb{R}^d$ be a simple set. Let $c = \frac{1}{P(E)}$ and let us prove that $f := c\mathbb{1}_E$ is an extreme point of B_1^Y . Assume that for some functions $g, h \in B_1^Y$ and $\lambda \in [0, 1]$ we can write

$$f = \lambda g + (1 - \lambda)h$$

and let us prove that necessarily $g = c\mathbb{1}_E$ and $h = c\mathbb{1}_E$. Since $\|f\|_{\text{FV}} = 1$, we have that

$$1 \leq \lambda \|g\|_{\text{FV}} + (1 - \lambda)\|h\|_{\text{FV}}$$

and we claim that actually equality holds. If it were

$$1 < \lambda \|g\|_{\text{FV}} + (1 - \lambda)\|h\|_{\text{FV}},$$

then we would get, being $f, g \in B_1^Y$,

$$1 < \lambda \|g\|_{\text{FV}} + (1 - \lambda)\|h\|_{\text{FV}} \leq \lambda + (1 - \lambda) = 1,$$

a contradiction. In a complete similar way, one can prove that $\|g\|_{\text{FV}} = 1 = \|h\|_{\text{FV}}$. All in all, we can represent

$$\mathbb{1}_E = \phi + \psi \tag{4.1}$$

with

$$\|\mathbb{1}_E\|_{\text{FV}} = \|\phi\|_{\text{FV}} + \|\psi\|_{\text{FV}} \tag{4.2}$$

being $\phi = c^{-1}\lambda g$ and $\psi = c^{-1}(1 - \lambda)h$. Notice actually that it holds

$$|D\mathbb{1}_E| = |D\phi| + |D\psi| \quad \text{as measures in } \mathbb{R}^d. \tag{4.3}$$

Indeed, by (4.1) and the triangle inequality we get $|D\mathbb{1}_E| \leq |D\phi + D\psi| \leq |D\phi| + |D\psi|$; the converse inequality then follows exploiting (4.2). In particular, computing (4.3) on the Borel set E^1 , it follows

$$|D\phi|(E^1) + |D\psi|(E^1) = |D\mathbb{1}_E|(E^1) = 0,$$

where the last equality follows from De Giorgi's Theorem. Hence

$$|D\phi|(E^1) = 0 = |D\psi|(E^1).$$

By Proposition 2.15 and by the indecomposability of E , there exist constants $c_1, c_2 \in \mathbb{R}$ such that

$$\phi(x) = c_1, \quad \psi(x) = c_2 \quad \text{a.e. in } E. \tag{4.4}$$

In particular, $c_1 + c_2 = 1$. In an analogous way, we also get $|D\phi|(E^0) = 0 = |D\psi|(E^0)$: being $E^0 = (\mathbb{R}^d \setminus E)^1$, by the indecomposability of E^c (recall Proposition 2.17), we conclude again by Proposition 2.15 that there exist constants $c_3, c_4 \in \mathbb{R}$ such that

$$\phi(x) = c_3, \quad \psi(x) = c_4 \quad \text{a.e. in } E^c.$$

By the Isoperimetric Inequality (2.2), either E or E^c has finite measure and, up to rename everything, consider the case in which E has finite measure. Then E^c must have infinite Lebesgue measure and the functions ϕ, ψ are constant functions which are in $L^{1^*}(\mathbb{R}^d)$: thus it must be $c_3 = c_4 = 0$, i.e.

$$\phi(x) = 0 = \psi(x) \quad \text{a.e. in } E^c.$$

Combined with (4.4), this gives that

$$\phi(x) = c_1 \mathbb{1}_E(x), \quad \psi(x) = (1 - c_1) \mathbb{1}_E(x) \quad \text{a.e. in } \mathbb{R}^d.$$

In particular, we deduce that

$$g = \alpha \mathbb{1}_E(x)$$

and being $\|g\|_{FV} = 1$, we obtain that the constant has to be

$$\alpha = \frac{1}{P(E)}.$$

One can argue similarly with h and the conclusion is now achieved: we have proved that the only convex combination of elements in B_1^Y representing f is the trivial one, i.e. f is an extreme point of B_1^Y .

Necessity. The argument used in the proof of Lemma 3.3 can be repeated verbatim here, yielding an analogous conclusion: an extreme point f of B_1^Y has necessarily the form

$$f(x) = \sigma \frac{\mathbb{1}_E(x)}{P(E)}, \quad \mathcal{L}^d\text{-a.e. } x \in \mathbb{R}^d,$$

for some set of positive finite perimeter $E \subset \mathbb{R}^d$. It remains thus to show that E has to be simple. Let us show first that E is indecomposable. To this end, assume that it can be written as $E = A \cup B$ with $A \cap B = \emptyset$ and $P(E) = P(A) + P(B)$. Hence

$$\frac{1}{P(E)} \mathbb{1}_E = \frac{P(A)}{P(E)} \frac{1}{P(A)} \mathbb{1}_A + \frac{P(B)}{P(E)} \frac{1}{P(B)} \mathbb{1}_B$$

is a convex linear combination of indicators of sets (normalized by perimeter). Therefore if $\frac{1}{P(E)} \mathbb{1}_E$ is an extreme point of the unit ball in Y , then E has to be indecomposable.

In view of Proposition 2.17, it remains to show that E^c has to be indecomposable, too. For let us suppose that C, D are such that $E^c = C \cup D$ with $C \cap D = \emptyset$ and $P(E^c) = P(C) + P(D)$. Arguing as above, we get that

$$E = C^c \cap D^c = C' \setminus D \implies \mathbb{1}_E = \mathbb{1}_{C'} - \mathbb{1}_D,$$

with $C' = C^c$. Consequently, since $P(E) = P(E^c) = P(C) + P(D)$, it holds

$$\frac{1}{P(E)} \mathbb{1}_E = \frac{P(C)}{P(E)} \frac{1}{P(C)} \mathbb{1}_{C'} + \frac{P(D)}{P(E)} \frac{-1}{P(D)} \mathbb{1}_D$$

is a convex linear combination of indicators of sets (normalized by perimeter). Therefore if $\frac{1}{P(E)} \mathbb{1}_E$ is an extremal point of the unit ball in Y , then E, E^c have to be indecomposable, hence E is simple and this concludes the proof. \square

5 Hamiltonian potential of divergence-free vector measures in \mathbb{R}^2

5.1 Divergence-free measures and FV

We now define the space of vector-valued divergence-free measures.

Definition 5.1. We will denote by $\mathcal{J}(\mathbb{R}^d)$ the following set of vector-valued measures:

$$\mathcal{J}(\mathbb{R}^d) := \{\boldsymbol{\mu} \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d) : \operatorname{div} \boldsymbol{\mu} = 0\},$$

where the divergence operator is understood in the sense of distributions.

The space \mathcal{J} is a real vector space under the usual operations of additions of measures and multiplication by real numbers and it can be equipped with a norm given by the total variation:

$$\|\boldsymbol{\mu}\|_{\mathcal{J}} := |\boldsymbol{\mu}|(\mathbb{R}^2).$$

Remark 5.2. As already observed in the Introduction, an important (somehow paradigmatic) example of a measure belonging to \mathcal{J} is the one associated to a Lipschitz closed curve: if $\gamma: [0, 1] \rightarrow \mathbb{R}^2$ is a Lipschitz map, injective on $[0, 1)$ and with $\gamma(0) = \gamma(1)$, we can define the measure $\boldsymbol{\mu}_\gamma \in \mathcal{M}(\mathbb{R}^2; \mathbb{R}^2)$ to be

$$\langle \Phi, \boldsymbol{\mu}_\gamma \rangle := \int_{\mathbb{R}^2} \Phi(z) d\mathcal{H}^1_{\gamma([0,1])}(z) \quad \text{for all } \Phi \in C^0(\mathbb{R}^2)^2,$$

which, by the Area Formula, can also be written as

$$\langle \Phi, \boldsymbol{\mu}_\gamma \rangle = \int_0^1 \Phi(\gamma(t)) \cdot \gamma'(t) dt.$$

Notice that this definition is well-posed, in the sense that it does not depend on the parametrization γ of the curve. It is easy to see that $\operatorname{div} \boldsymbol{\mu}_\gamma = 0$ in the sense of distributions, as a consequence of the fact that $\gamma(0) = \gamma(1)$, so $\boldsymbol{\mu}_\gamma \in \mathcal{J}(\mathbb{R}^2)$.

The following proposition establishes a functional analytic connection between $\mathcal{J}(\mathbb{R}^2)$ and $\operatorname{FV}(\mathbb{R}^2)$.

Proposition 5.3. *The map*

$$\nabla^\perp : \operatorname{FV}(\mathbb{R}^2) \rightarrow \mathcal{J}(\mathbb{R}^2), \quad f \mapsto \boldsymbol{\mu} := \nabla^\perp f = (-\partial_y f, \partial_x f)$$

is an isometric isomorphism.

Proof. We divide the proof into four steps.

Well-posedness and linearity. The map ∇^\perp is well-posed, because $\operatorname{div} \nabla^\perp f = 0$ for any $f \in \operatorname{FV}(\mathbb{R}^2)$: indeed, for any test function $\phi \in C_c^\infty(\mathbb{R}^2)$,

$$\begin{aligned} \langle \operatorname{div} \nabla^\perp f, \phi \rangle &= \int_{\mathbb{R}^2} \nabla \phi(z) \cdot d(\nabla^\perp f)(z) \\ &= \int_{\mathbb{R}^2} (\partial_x \phi(z), \partial_y \phi(z)) \cdot d((-\partial_y f, \partial_x f))(z) \\ &= \int_{\mathbb{R}^2} \partial_y \partial_x \phi(z) f(z) dz - \int_{\mathbb{R}^2} \partial_x \partial_y \phi(z) f(z) dz = 0. \end{aligned}$$

Linearity of ∇^\perp is trivial.

Injectivity. The kernel of ∇^\perp is given by the functions f for which

$$\nabla^\perp f = 0,$$

which means f is constant in \mathbb{R}^2 , in particular $f = 0$ in $\operatorname{FV}(\mathbb{R}^2)$: injectivity follows.

Surjectivity. Pick $\mu \in \mathcal{J}(\mathbb{R}^2)$ and let $\{\rho_\varepsilon\}_{\varepsilon>0}$ be a standard family of mollifiers in \mathbb{R}^2 . Set

$$\Phi_\varepsilon(x) := \mu * \rho_\varepsilon(x) = \int_{\mathbb{R}^2} \rho_\varepsilon(x-y) d\mu(y)$$

and observe that by standard facts $\Phi_\varepsilon \in C_c^\infty(\mathbb{R}^2; \mathbb{R}^2)$ with $\operatorname{div} \Phi_\varepsilon = 0$. By the Poincaré Lemma, for every $\varepsilon > 0$, there exists $f_\varepsilon \in C_c^\infty(\mathbb{R}^2)$ such that $\nabla^\perp f_\varepsilon = \Phi_\varepsilon$. Notice that for any $\varepsilon > 0$,

$$V(f_\varepsilon) = \|\Phi_\varepsilon\|_1 \leq \|\mu\|_{\mathcal{J}},$$

hence $(f_\varepsilon)_{\varepsilon>0} \subset \operatorname{FV}(\mathbb{R}^2)$. By Theorem 2.3 there exist $\{m_\varepsilon\}_{\varepsilon>0} \subset \mathbb{R}$ and a universal constant $\gamma > 0$ such that

$$\|f_\varepsilon - m_\varepsilon\|_{L^1(\mathbb{R}^2)} \leq \gamma V(f_\varepsilon) \leq \gamma \|\mu\|_{\mathcal{J}}.$$

In particular, if we now fix any open $\Omega \in \mathbb{R}^2$, we have using Hölder inequality

$$\|f_\varepsilon - m_\varepsilon\|_{L^1(\Omega)} \leq \mathcal{L}^d(\Omega)^{\frac{1}{d}} \|f_\varepsilon - m_\varepsilon\|_{L^1(\mathbb{R}^2)} \leq \gamma \mathcal{L}^d(\Omega)^{\frac{1}{d}} \|\mu\|_{\mathcal{J}}.$$

On the other hand,

$$V(f_\varepsilon - m_\varepsilon, \Omega) \leq V(f_\varepsilon - m_\varepsilon) = V(f_\varepsilon) \leq \|\mu\|_{\mathcal{J}}$$

and hence we are in a position to apply the Compactness Theorem [3, Theorem 3.23]: there exists a function $f \in L^1_{\operatorname{loc}}(\mathbb{R}^2)$ such that, up to a subsequence, $(f_\varepsilon - m_\varepsilon) \rightarrow f$ strongly in $L^1_{\operatorname{loc}}(\mathbb{R}^2)$ as $\varepsilon \rightarrow 0$. In particular, f is also in $\operatorname{FV}(\mathbb{R}^2)$ by the l.s.c. of the total variation

$$V(f) \leq \liminf_{\varepsilon \downarrow 0} V(f_\varepsilon) \leq c \|\mu\|_{\mathcal{J}}.$$

It remains now to check that $\nabla^\perp f = \mu$: for any smooth, compactly supported test function $\Psi \in C_c^\infty(\mathbb{R}^2, \mathbb{R}^2)$ it holds

$$\begin{aligned} \int_{\mathbb{R}^2} f(x) \operatorname{div} \Psi(x) dx &= \lim_{\varepsilon \rightarrow 0} \int_{\mathbb{R}^2} f_\varepsilon(x) \operatorname{div} \Psi(x) dx \\ &= \lim_{\varepsilon \rightarrow 0} \int_{\mathbb{R}^2} \Psi(x) \cdot \Phi_\varepsilon^\perp(x) dx \\ &= \int_{\mathbb{R}^2} \Psi(x) d\mu^\perp(x), \end{aligned}$$

where in the last passage we have used that $\Phi_\varepsilon \rightarrow \mu$ as $\varepsilon \rightarrow 0$ (see e.g. [3, Theorem 2.2]).

∇^\perp is an isometry. It remains thus to show that ∇^\perp is an isometry: taken $f \in \operatorname{FV}(\mathbb{R}^2)$, by definition

$$\begin{aligned} \|f\|_{\operatorname{FV}} = V(f, \mathbb{R}^2) &= \sup \left\{ \int_{\mathbb{R}^2} f(x) \operatorname{div} \Phi(x) dx : \Phi \in C_c^\infty(\mathbb{R}^2), \|\Phi\|_\infty \leq 1 \right\} \\ &= \sup \{ \langle \Phi, \nabla f \rangle : \Phi \in C_c^\infty(\mathbb{R}^2), \|\Phi\|_\infty \leq 1 \} \\ &= \sup \{ \langle \Phi, \nabla^\perp f \rangle : \Phi \in C_c^\infty(\mathbb{R}^2), \|\Phi\|_\infty \leq 1 \} \\ &= \sup \left\{ \int_{\mathbb{R}^2} \operatorname{div} \Phi(x) d(\nabla^\perp f)(x) : \Phi \in C_c^\infty(\mathbb{R}^2), \|\Phi\|_\infty \leq 1 \right\} \\ &= \|\nabla^\perp f\|_{\mathcal{J}}. \end{aligned} \quad \square$$

6 Simple sets and closed curves

Aim of this section is to give a detailed description of the extreme points of the unit ball of \mathcal{J} . Since ∇^\perp is an isometry we have

$$B_1^{\mathcal{J}} = \nabla^\perp(B_1^Y)$$

and hence, by Remark 2.27, we have

$$\begin{aligned} \text{ext}(B_1^{\mathcal{J}}) &= \text{ext}(\nabla^{\perp}(B_1^{\mathcal{Y}})) \\ &= \nabla^{\perp}(\text{ext}(B_1^{\mathcal{Y}})) \\ &= \left\{ \sigma \frac{\nabla^{\perp} \mathbb{1}_E}{P(E)} : E \subset \mathbb{R}^2 \text{ simple set, } P(E) > 0 \text{ and } \sigma \in \{\pm 1\} \right\}. \end{aligned} \tag{6.1}$$

Let us introduce the following notation:

$$\Gamma := \{ \gamma : [0, 1] \rightarrow \mathbb{R}^2 : \text{Lipschitz on } [0, 1], \text{ injective on } [0, 1] \text{ and } \gamma(0) = \gamma(1) \}.$$

For any $\gamma \in \Gamma$, we define its length to be

$$\ell(\gamma) := \int_0^1 |\gamma'(t)| dt \in (0, +\infty).$$

Notice, in particular, that any $\gamma \in \Gamma$ induces a rectifiable Jordan curve $C := \gamma([0, 1])$, and viceversa every rectifiable Jordan curve can be parametrized by some $\gamma \in \Gamma$. Being a subset of $(\text{Lip}[0, 1])^2$, the space Γ can be thought as a normed space, being the norm the (restriction of the) uniform one $\| \cdot \|_{\infty}$.

We now want to prove the following proposition.

Proposition 6.1. *The following equality holds true:*

$$\text{ext}(B_1^{\mathcal{J}}) = \left\{ \frac{1}{\ell(\gamma)} \mu_{\gamma} : \gamma \in \Gamma \right\}.$$

Proof. Let $\mu \in \text{ext}(B_1^{\mathcal{J}})$. From (6.1) we have that

$$\mu = \sigma \frac{1}{P(E)} \nabla^{\perp} \mathbb{1}_E$$

for some simple set $E \subset \mathbb{R}^2$ with $P(E) > 0$ and $\sigma \in \pm 1$. From Theorem 2.21, the essential boundary $\partial^e E$, is equivalent, up to an \mathcal{H}^1 -negligible set, to a rectifiable Jordan curve. Using Theorem 2.7, we can conclude that also $\mathcal{F}E$ can be parametrized by some Jordan curve, which can be taken to be Lipschitz (see [1, Lemma 3]). All in all, we have that there exists $\gamma \in \Gamma$ such that

$$\gamma([0, 1]) = \mathcal{F}E,$$

up to a \mathcal{H}^1 -null set.

On the one hand, by De Giorgi’s Theorem 2.6, for \mathcal{H}^1 -a.e. $x \in \mathcal{F}E$ we have

$$\text{Tan}(\mathcal{F}E, x) = \text{span}(v_E^{\perp}(x)), \tag{6.2}$$

where $v_E(x)$ is the generalized inner normal to E and $\text{span}(v_E^{\perp}(x))$ denotes the orthogonal line to $v_E(x)$.

On the other hand, since $\mathcal{F}E = \gamma([0, 1])$, we have using Proposition 2.5,

$$\text{Tan}(\gamma([0, 1]), x) = \text{span}(\gamma'(y^{-1}(x))). \tag{6.3}$$

Since the approximate tangent space is a one-dimensional vector space and since $v_E(x)$ is unit vector for \mathcal{H}^1 -almost every $x \in \mathcal{F}E$, equalities (6.2) and (6.3) force that for \mathcal{H}^1 -a.e. $x \in \mathcal{F}E$,

$$v_E^{\perp}(x) = \sigma(x) \frac{\gamma'(y^{-1}(x))}{|\gamma'(y^{-1}(x))|} \quad \text{for } \sigma(x) \in \{\pm 1\}.$$

This means that the vector $v_E^{\perp}(x)$ is tangent to the curve γ at the point $\gamma(y^{-1}(x))$ for \mathcal{H}^1 -a.e. $x \in \gamma([0, 1])$. Since $\text{div}(v_E^{\perp} \mathcal{H}^1 \llcorner_{\gamma([0,1])}) = 0$, we can apply [6, Theorem 4.9], obtaining that

$$\exists \bar{\sigma} \in \{\pm 1\} : v_E^{\perp}(\gamma(t)) = \bar{\sigma} \cdot \frac{\gamma'(t)}{|\gamma'(t)|} \text{ for } \mathcal{L}^1\text{-a.e. } t \in [0, 1].$$

Reversing the parametrization of γ , if necessary, one can achieve that $\bar{\sigma} = 1$. Then for any test function $\Phi \in C_c^\infty(\mathbb{R}^2; \mathbb{R}^2)$, using the Area Formula, we obtain

$$\begin{aligned} \langle \mu, \Phi \rangle &= \left\langle \frac{1}{P(E)} \nabla^\perp \mathbb{1}_E, \Phi \right\rangle \\ &= \left\langle \frac{1}{P(E)} \nu_E^\perp \mathcal{H}^1 \llcorner_{\mathcal{F}E}, \Phi \right\rangle \\ &= \frac{1}{P(E)} \int_{\mathbb{R}^2} \Phi(x) \cdot \nu_E^\perp(x) \, d\mathcal{H}^1 \llcorner_{\mathcal{F}E}(x) \\ &= \frac{1}{P(E)} \int_{\mathbb{R}^2} \Phi(x) \cdot \frac{y'(y^{-1}(x))}{|y'(y^{-1}(x))|} \, d\mathcal{H}^1 \llcorner_{\gamma([0,1])}(x) \\ &= \frac{1}{P(E)} \int_0^1 \Phi(\gamma(t)) \cdot \frac{y'(t)}{|y'(t)|} |y'(t)| \, dt \\ &= \frac{1}{\ell(\gamma)} \int_0^1 \Phi(\gamma(t)) \cdot y'(t) \, dt \\ &= \left\langle \frac{1}{\ell(\gamma)} \mu_\gamma, \Phi \right\rangle, \end{aligned} \tag{6.4}$$

where we have also used the fact that

$$P(E) = V(\mathbb{1}_E) = \|\nu_E \mathcal{H}^1 \llcorner_{\mathcal{F}E}\|_{\mathcal{M}} = \mathcal{H}^1(\gamma([0, 1])) = \ell(\gamma)$$

(which also follows from the Area Formula).

Thus we have shown that any extreme point μ of B_1^d has necessarily the form $\frac{1}{\ell(\gamma)} \mu_\gamma$. The converse implication, namely that normalized measures μ_γ are extreme, follows immediately from the second part of Theorem 2.21: any $\gamma \in \Gamma$ induces a rectifiable Jordan curve $C := \gamma([0, 1])$, hence $\text{int}(\Gamma) =: E$ is a simple set by Theorem 2.21. Extremality follows from (6.1), noticing that

$$\frac{1}{\ell(\gamma)} \mu_\gamma = \frac{1}{P(E)} \nabla^\perp \mathbb{1}_E$$

as above, and the proof is thus complete. □

7 Measures as superposition of curves I: A proof using Choquet theory

In this section we prove the Main Theorem with $\rho = 0$.

Theorem 7.1. *Let $\mu \in \mathcal{J}(\mathbb{R}^2)$, where $\mathcal{J}(\mathbb{R}^2)$ is as in Definition 5.1. Then there exists a σ -finite, non-negative measure $\eta \in \mathcal{M}_+(\Gamma)$ such that (1.2a) and (1.2b) hold.*

Consider the maps $p: \Gamma \rightarrow \mathcal{J}(\mathbb{R}^d)$ and $F: p(\Gamma) \rightarrow \mathcal{J}(\mathbb{R}^d)$ defined by

$$p(\gamma) := \mu_\gamma, \quad F(v) := \begin{cases} \frac{v}{\|v\|}, & v \neq 0, \\ 0, & v = 0. \end{cases} \tag{7.1}$$

For any $m \in \mathbb{N}$ let

$$\Gamma_m := \{\gamma \in \Gamma : |\gamma(0)| + \|\gamma'\|_\infty \leq m\}.$$

In view of the Arzelà–Ascoli Theorem, Γ_m is a compact subset of Γ (with respect to the topology of the uniform convergence).

The lemma below works in any dimension d (not only $d = 2$).

Lemma 7.2. *The maps p and F defined in (7.1) have the following properties:*

- (1) *For any $m \in \mathbb{N}$ the map $p: \Gamma_m \rightarrow \mathcal{J}(\mathbb{R}^d)$ defined in (7.1) is continuous (with respect to uniform topology on Γ_m and weak- $*$ topology on $\mathcal{J}(\mathbb{R}^d)$).*
- (2) *The map $F: \mathcal{J}(\mathbb{R}^d) \rightarrow \mathcal{J}(\mathbb{R}^d)$ is Borel.*
- (3) *The sets $p(\Gamma)$ and $F(p(\Gamma))$ are Borel.*
- (4) *The $F: p(\Gamma) \rightarrow F(p(\Gamma))$ has Borel inverse F^{-1} .*

Proof. It is sufficient to verify sequential continuity of p . Let $(\gamma_k)_{k \in \mathbb{N}} \subset \Gamma_m$ be a sequence with $\gamma_n \rightarrow \gamma$ for a certain $\gamma \in \Gamma_m$. Let us show that $\mu_{\gamma_n} \xrightarrow{*} \mu_\gamma$ first in the sense of distributions: let $\Phi \in C_c^\infty(\mathbb{R}^d; \mathbb{R}^d)$. Then

$$\begin{aligned} |\langle \mu_{\gamma_n}, \Phi \rangle - \langle \mu_\gamma, \Phi \rangle| &= \left| \int_0^1 \Phi(\gamma_n(t)) \cdot \gamma_n'(t) dt - \int_0^1 \Phi(\gamma(t)) \cdot \gamma'(t) dt \right| \\ &= \left| \int_0^1 (\Phi(\gamma_n(t)) - \Phi(\gamma(t))) \cdot \gamma_n'(t) dt - \int_0^1 \Phi(\gamma(t)) \cdot (\gamma'(t) - \gamma_n'(t)) dt \right| \\ &\leq m \int_0^1 |\Phi(\gamma_n(t)) - \Phi(\gamma(t))| dt + \left| \int_0^1 \frac{d}{dt} \Phi(\gamma(t)) \cdot (\gamma(t) - \gamma_n(t)) dt \right| \\ &\leq m \int_0^1 |\Phi(\gamma_n(t)) - \Phi(\gamma(t))| dt + \|\nabla \Phi\|_\infty \int_0^1 |\gamma(t) - \gamma_n(t)| dt \rightarrow 0 \end{aligned}$$

as $n \rightarrow +\infty$. Moreover, $\sup_{n \in \mathbb{N}} \|\mu_{\gamma_n}\| \leq m$. Hence the functionals $\mu_{\gamma_n} \in C_0(\mathbb{R}^d; \mathbb{R}^d)^*$ are uniformly bounded and converge to μ pointwise on the set $C_c^\infty(\mathbb{R}^d; \mathbb{R}^d)$ which is dense in $C_0(\mathbb{R}^d; \mathbb{R}^d)$. Therefore $\mu_{\gamma_n} \xrightarrow{*} \mu$ as $n \rightarrow \infty$.

Since for any $m \in \mathbb{N}$ the set $p(\Gamma_m)$ is compact (being an image of a compact under a continuous map), the set $p(\Gamma) = \bigcup_{m \in \mathbb{N}} p(\Gamma_m)$ is Borel.

For any $\Phi \in C_0(\mathbb{R}^d; \mathbb{R}^d)$ the map $\mathbf{v} \mapsto \frac{\langle \mathbf{v}, \Phi \rangle}{\|\mathbf{v}\|}$ is Borel. Indeed, the numerator is a continuous function of \mathbf{v} and the denominator is lower semicontinuous (hence Borel). Therefore F is a Borel map from $\mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ to $\mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ (with respect to weak- $*$ topologies). Since for every $m \in \mathbb{N}$ the set $p(\Gamma_m)$ is contained in a closed ball of $\mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ (note that this ball is Polish with respect to weak- $*$ topology) and F is injective on $p(\Gamma_m)$, it follows that $F(p(\Gamma_m))$ is Borel (see e.g. [8, Theorem 6.8.6]). Therefore $F(p(\Gamma)) = \bigcup_{m=1}^\infty F(p(\Gamma_m))$ is Borel. Similarly, the image of any Borel subset of $p(\Gamma)$ under F is Borel, and by injectivity of F on $p(\Gamma)$ this means that $F: p(\Gamma) \rightarrow F(p(\Gamma))$ has Borel inverse. □

Lemma 7.3. *Suppose that $\mu \in \mathcal{J}(\mathbb{R}^d)$ and there exists a finite measure $\xi \in \mathcal{M}_+(\mathcal{J}(\mathbb{R}^d))$ concentrated on $F(p(\Gamma))$ such that*

$$\mu = \int_{\mathcal{J}(\mathbb{R}^d)} \mathbf{v} d\xi(\mathbf{v}), \quad |\mu| = \int_{\mathcal{J}(\mathbb{R}^d)} |\mathbf{v}| d\xi(\mathbf{v}). \tag{7.2}$$

Then there exists σ -finite $\eta \in \mathcal{M}_+(\Gamma)$ such that (1.2a) and (1.2b) hold for μ and η .

Proof. By Lemma 7.2 the map $F: p(\Gamma) \rightarrow F(p(\Gamma))$ has Borel inverse F^{-1} , hence we can change variables using the map F^{-1} :

$$\int_{F(p(\Gamma))} y d\xi(y) = \int_{F(p(\Gamma))} F(F^{-1}(y)) d\xi(y) = \int_{p(\Gamma)} F(\mathbf{v}) d(F_{\#}^{-1}\xi)(\mathbf{v}) = \int_{p(\Gamma)} \mathbf{v} d\hat{\xi}(\mathbf{v}),$$

where $\hat{\xi}$ denotes the measure on $\mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ defined by

$$\hat{\xi}(A) := \int_{A \setminus \{0\}} \frac{1}{\|\mathbf{v}\|} d(F_{\#}^{-1}\xi)(\mathbf{v}),$$

$A \subset \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ being an arbitrary Borel subset (clearly $\hat{\xi}$ is concentrated on $p(\Gamma)$).

Since $p(\Gamma) = \bigcup_{m \in \mathbb{N}} p(\Gamma_m)$, we can write $\hat{\xi}$ as a sum of its restrictions $\hat{\xi}_m$ on the sets $p(\Gamma_{m+1}) \setminus p(\Gamma_m)$, where $m \in \mathbb{N}$.

By Lemma 7.2 the map $p: \Gamma_m \rightarrow \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ is continuous and the set Γ_m is compact, hence there exists a Borel set $B_m \subset \Gamma_m$ such that the restriction of p to B_m is injective and $p(B_m) = p(\Gamma_m)$ (see e.g. [8, Theorem 6.9.7]). Therefore the inverse map $q_m: p(\Gamma_m) \rightarrow \Gamma_m$ is Borel. Now we change variables using q_m :

$$\int_{p(\Gamma)} \mathbf{v} \, d\tilde{\xi}_m(\mathbf{v}) = \int_{p(\Gamma_m)} p(q_m(\mathbf{v})) \, d\tilde{\xi}_m(\mathbf{v}) = \int_{\Gamma_m} p(\gamma) \, d((q_m)_\# \tilde{\xi}_m)(\gamma) = \int_{\Gamma} \boldsymbol{\mu}_\gamma \, d(\eta_m)(\gamma),$$

where $\eta_m := (q_m)_\# \tilde{\xi}_m$. Denoting $\eta := \sum_{m=1}^\infty \eta_m$, we ultimately obtain

$$\boldsymbol{\mu} = \int_{\Gamma} \boldsymbol{\mu}_\gamma \, d\eta(\gamma).$$

Equality holds for total variations as well: indeed, by the triangle inequality,

$$|\boldsymbol{\mu}| \leq \int_{\Gamma} |\boldsymbol{\mu}_\gamma| \, d\eta(\gamma) \quad \text{as measures on } \mathbb{R}^d.$$

If the inequality above were strict, then by evaluating it on the whole \mathbb{R}^d we would get a contradiction:

$$\|\boldsymbol{\mu}\| = |\boldsymbol{\mu}|(\mathbb{R}^d) < \int_{\Gamma} |\boldsymbol{\mu}_\gamma|(\mathbb{R}^d) \, d\eta(\gamma) = \int_{\mathcal{J}(\mathbb{R}^d)} |\mathbf{v}|(\mathbb{R}^d) \, d\xi(\mathbf{v}) = \|\boldsymbol{\mu}\|.$$

Since $\|\boldsymbol{\mu}\| = \int \|\boldsymbol{\mu}_\gamma\| \, d\eta(\gamma)$ and for any $k \in \mathbb{N}$ the set $\{\gamma \in \Gamma : \|\boldsymbol{\mu}_\gamma\| > k^{-1}\}$ is Borel, it is clear that η is σ -finite. \square

We are now ready to prove the Main Theorem.

Proof of Theorem 7.1. By Proposition 6.1 we have

$$\text{ext}(B_1^{\mathcal{J}}) = \left\{ \frac{1}{\ell(\gamma)} \boldsymbol{\mu}_\gamma : \gamma \in \Gamma \right\} \subset F(p(\Gamma)).$$

By Remark 2.22 the set $\text{ext}(B_1^{\mathcal{J}})$ is Borel.

Let $0 \neq \boldsymbol{\mu} \in \mathcal{J}(\mathbb{R}^2)$ and consider the normalized measure

$$\frac{\boldsymbol{\mu}}{\|\boldsymbol{\mu}\|} \in B_1^{\mathcal{J}}.$$

By Choquet's Theorem 2.26 there exists a Borel probability measure $\pi \in \mathcal{P}(\text{ext } B_1^{\mathcal{J}})$ such that

$$\frac{\boldsymbol{\mu}}{\|\boldsymbol{\mu}\|} = \int_{\text{ext } B_1^{\mathcal{J}}} y \, d\pi(y),$$

the integral being understood in the sense of Definition 2.24. By the triangle inequality we deduce from the equality above that

$$|\boldsymbol{\mu}| \leq \|\boldsymbol{\mu}\| \int_{\text{ext } B_1^{\mathcal{J}}} |y| \, d\pi(y).$$

Note that the latter inequality is in fact an equality, since otherwise by evaluating it on \mathbb{R}^2 we would get a contradiction:

$$\|\boldsymbol{\mu}\| < \|\boldsymbol{\mu}\| \int_{\text{ext } B_1^{\mathcal{J}}} \|\boldsymbol{y}\| \, d\pi(y) = \|\boldsymbol{\mu}\|.$$

In order pass to integration over Γ instead of $\text{ext}(B_1^{\mathcal{J}}) \subset \mathcal{M}(\mathbb{R}^2; \mathbb{R}^2)$, it remains to change variables by applying Lemma 7.3 with $\xi := \|\boldsymbol{\mu}\|\pi$. This concludes the proof. \square

Note that the elements of Γ are not necessarily simple. However, since the measure π is concentrated on a set of measures induced by simple curves, it is easy to see from the proof of Theorem 7.1 that for η -a.e. $\gamma \in \Gamma$ there exists a simple $\tilde{\gamma} \in \Gamma$ such that $\boldsymbol{\mu}_\gamma = \boldsymbol{\mu}_{\tilde{\gamma}}$.

8 Measures as superposition of curves II: A proof using decomposition of FV functions

In this section, we present an alternative proof of Theorem 7.1. This proof does not rely on Choquet theory, but it is based on the following decomposition result for FV functions.

Theorem 8.1. *Let $f \in \text{FV}(\mathbb{R}^d)$. There exists an at most countable family $\{f_i\}_{i \in I} \subset \text{FV}(\mathbb{R}^d)$ of monotone functions such that the series*

$$\sum_{i \in I} f_i$$

converges as an element of $\text{FV}(\mathbb{R}^d)$ and

$$f = \sum_{i \in I} f_i \quad \text{with} \quad \|f\|_{\text{FV}} = \sum_{i \in I} \|f_i\|_{\text{FV}}. \quad (8.1)$$

For the definition of monotone function and for a proof of Theorem 8.1 as well we refer the reader to Appendix A.

Proof of Theorem 7.1 using Theorem 8.1. Let $\mu \in \mathcal{J}(\mathbb{R}^2)$ and let $H \in \text{FV}(\mathbb{R}^2)$ be the function such that $\mu = \nabla^\perp H$, whose existence and uniqueness are granted by Proposition 5.3.

Case 1. Suppose first that H is monotone. Let $E_t := \{H > t\}$. Since the function H lies in $\text{FV}(\mathbb{R}^2)$, we have $H \in L^1(\mathbb{R}^2)$: using Chebyshev's inequality, this integrability property implies that for a.e. $t \in \mathbb{R}$ it holds $\mathcal{L}^2(E_t) < \infty$. Combined with the Coarea Formula, this observation yields the existence of a set $N \subset \mathbb{R}$ such that $\mathcal{L}^1(N) = 0$ and E_t has finite measure and finite perimeter for every $t \in \mathbb{R} \setminus N$. Consider now the function $g: \mathbb{R} \rightarrow \mathcal{J}(\mathbb{R}^2)$ defined by

$$g(t) := \begin{cases} \frac{\nabla^\perp \mathbb{1}_{E_t}}{P(E_t)} & \text{if } \mathcal{L}^2(E_t) > 0 \text{ and } t \notin N \text{ with } P(E_t) > 0, \\ 0 & \text{otherwise,} \end{cases}$$

and the measure $\rho \in \mathcal{M}_+(\mathbb{R})$

$$\rho(dt) := P(E_t) \mathcal{L}^1(dt).$$

By the Coarea Formula, we have

$$\nabla^\perp H = \int_{\mathbb{R}} g(t) d\rho(t) \quad (8.2)$$

and Fubini's Theorem further ensures that f is a measurable measure-valued map (see [3, Definition 2.25]). In particular, from (8.2) we deduce for any $\Psi \in C_c(\mathbb{R}^2)^2$,

$$\langle \nabla^\perp H, \Psi \rangle = \int_{\mathbb{R}} \langle g(t), \Psi \rangle d\rho(t) = \int_{f(\mathbb{R})} \langle y, \Psi \rangle d\eta(y), \quad (8.3)$$

where we have set

$$\xi := g\#\rho.$$

From (8.3) and from the arbitrariness of test function Ψ , we deduce the sought formula

$$\nabla^\perp H = \int_{\mathcal{J}(\mathbb{R}^2)} y d\eta(y).$$

Observe that, for every $t \in \mathbb{R} \setminus N$ such that $\mathcal{L}^2(E_t) > 0$ the computations in (6.4) yield the equality

$$g(t) = \frac{\mu_{\gamma_t}}{P(E_t)},$$

where γ_t is the parametrization of $\partial^* E_t$ given by Theorem 2.21. Thus, by the very definition of g , the measure ξ is concentrated on $F(p(\Gamma))$ (see (7.1)). Moreover,

$$\|\xi\| = \|g\#\rho\| = \|\rho\| = \int_{\mathbb{R}} P(E_t) \mathcal{L}^1(dt) = \|\nabla^\perp H\|. \quad (8.4)$$

Case 2. If H is not monotone, apply Theorem 8.1 to the function H and let $\{H_i\}_{i \in I}$ be at most countable family of monotone functions satisfying (8.1) (without loss of generality we may assume that $I = \mathbb{N}$). Let ξ_i be a measure representing $\nabla^\perp H_i$ (obtained as in Case 1, since H_i is monotone). Then it is easy to see that

$$\xi := \sum_{i=1}^{\infty} \xi_i$$

defines a measure on $\mathcal{J}(\mathbb{R}^2)$, as the series converges strongly: indeed, by (8.4) and (8.1) we get

$$\sum_{i=1}^{\infty} \|\xi_i\| = \sum_{i=1}^{\infty} \|\nabla^\perp H_i\| = \|\nabla^\perp H\| < \infty.$$

Since the series above converge, we can pass to the limit as $n \rightarrow \infty$ in the equalities

$$\sum_{i=1}^n \nabla^\perp H_i = \int_{\mathcal{J}(\mathbb{R})} y d\left(\sum_{i=1}^n \xi_i\right)(y), \quad \sum_{i=1}^n |\nabla^\perp H_i| = \int_{\mathcal{J}(\mathbb{R})} |y| d\left(\sum_{i=1}^n \xi_i\right)(y).$$

We thus see that μ and ξ defined above satisfy (7.2). It remains to change variables using Lemma 7.3. □

9 Linear rigidity for vector-valued measures

In this section we give a proof of Theorem 1.4, which is inspired by (and generalizes) one of the results from [13] (see Theorem 1.2 therein).

Lemma 9.1. *Let $\mu \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ and consider its polar decomposition $\mu = \tau|\mu|$. Suppose that there exists $\eta \in \mathcal{M}_+(\Gamma)$ such that (1.2a) and (1.2b) hold. Then for η -a.e. $\gamma \in \Gamma$,*

$$\gamma'(t) = \tau(\gamma(t))|\gamma'(t)|$$

for a.e. $t \in [0, 1]$.

Proof. Since $|\mu|$ is a finite measure and $C_c(\mathbb{R}^d)$ is dense in $L^1(|\mu|)$, we can use τ as a test function in the distributional formulation of (1.2a), obtaining

$$\int_{\mathbb{R}^d} \tau \cdot d\mu = \int_{\Gamma} \tau \cdot \mu_\gamma d\eta(\gamma) = \int_{\Gamma} \int_0^1 \tau(\gamma(t)) \cdot \gamma'(t) dt d\eta(\gamma).$$

On the other hand,

$$\int_{\mathbb{R}^d} \tau \cdot d\mu = |\mu|(\mathbb{R}^d) = \int_{\Gamma} |\mu_\gamma| d\eta(\gamma) = \int_{\Gamma} \int_0^1 |\gamma'(t)| dt d\eta(\gamma).$$

Therefore

$$\int_{\Gamma} \int_0^1 (\tau(\gamma(t)) \cdot \gamma'(t) - |\gamma'(t)|) dt d\eta(\gamma) = 0. \tag{9.1}$$

The integrand is non-positive, since

$$\tau(\gamma(t)) \cdot \gamma'(t) \leq |\tau(\gamma(t))| \cdot |\gamma'(t)| = |\gamma'(t)|,$$

hence by (9.1) for η -a.e. γ ,

$$\tau(\gamma(t)) \cdot \gamma'(t) = |\gamma'(t)|$$

for a.e. $t \in [0, 1]$. □

Recall the following definition: $\sigma \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ is called a *subcurrent* of $\mu \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ if

$$\|\mu\| = \|\mu - \sigma\| + \|\sigma\|.$$

Proposition 9.2. Let $\mu \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$. Then $\sigma \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ is a subcurrent of μ if and only if

$$\sigma = g\mu,$$

where $g \in L^1(|\mu|)$ satisfies $0 \leq g(x) \leq 1$ for $|\mu|$ -a.e. $x \in \mathbb{R}^d$.

Proof. We split the proof into two parts.

Sufficiency. If $g \in L^1(|\mu|)$ satisfies $0 \leq g(x) \leq 1$ and $\sigma = g\mu$, then

$$|\mu - \sigma| + |\sigma| = (1 - g)|\mu| + g|\mu| = |\mu|,$$

and it remains to evaluate the equality above on \mathbb{R}^d .

Necessity. By the Radon–Nikodym Theorem there exist mutually singular $\sigma^a, \sigma^s \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ such that $\sigma^a \ll |\mu|$, $\sigma^s \perp |\mu|$ and $\sigma = \sigma^a + \sigma^s$. Then by the definition of subcurrent

$$\|\mu\| = \|\mu - \sigma\| + \|\sigma\| = \|\mu - \sigma^a\| + \|\sigma^a\| + 2\|\sigma^s\| \geq \|\mu\| + 2\|\sigma^s\|$$

by the triangle inequality, hence $\|\sigma^s\| = 0$. Therefore $\sigma = \theta|\mu|$ and $\mu = \tau|\mu|$ for some $\theta, \tau \in L^1(|\mu|; \mathbb{R}^d)$ (by polar decomposition). Writing again the definition of subcurrent we obtain

$$\int (|\tau| - |\tau - \theta| - |\theta|) d|\mu| = 0,$$

which implies (in view of the triangle inequality) that

$$|\tau(x)| - |\tau(x) - \theta(x)| - |\theta(x)| = 0 \tag{9.2}$$

for $|\mu|$ -a.e. $x \in \mathbb{R}^d$. In particular, for $|\mu|$ -a.e. $x \in \mathbb{R}^d$ if $\tau(x) = 0$, then $\theta(x) = 0$. Since the two vectors $a = \theta(x)$ and $b = \tau(x) - \theta(x)$ with $a \neq 0$ satisfy $|a + b| - |a| - |b| = 0$ if and only if $b = |b| \frac{a}{|a|}$, we conclude that there exists $g = g(x) \in \mathbb{R}$ such that $\theta(x) = g(x)\tau(x)$. Substituting this into equation (9.2), we conclude that $0 \leq g(x) \leq 1$ for $|\mu|$ -a.e. $x \in \mathbb{R}^d$. \square

Corollary 9.3. Suppose that $\mathbf{v} \in \mathcal{M}_{\text{loc}}(\mathbb{R}^d; \mathbb{R}^d)$ has polar decomposition $\mathbf{v} = \tau|\mathbf{v}|$. If $\tau_1(x) > 0$ for $|\mathbf{v}|$ -a.e. $x \in \mathbb{R}^d$, then \mathbf{v} is acyclic.

Proof. For any $\mu \in \mathcal{M}(\mathbb{R}^d; \mathbb{R}^d)$ satisfying (1.1) with some $\rho \in \mathcal{M}(\mathbb{R}^d)$ the distributional formulation of (1.1) holds for any test function $\varphi \in C^\infty(\mathbb{R}^d)$ such that $\|\varphi_\infty\| + \|\nabla\varphi\|_\infty < \infty$. In order to prove this, it is sufficient to consider $\omega \in C_c^\infty(\mathbb{R}^d)$ such that $\omega(x) = 1$ if $|x| \leq 1$ and $\omega(x) = 0$ if $|x| \geq 2$ and pass to the limit in

$$- \int_{\mathbb{R}^d} \nabla(\varphi(x)\omega(R^{-1}x)) \cdot d\mu(x) = \int_{\mathbb{R}^d} \varphi(x)\omega(R^{-1}x) d\rho(x)$$

as $R \rightarrow \infty$ using the Dominated Convergence Theorem.

In particular, if σ is a cycle of \mathbf{v} , then by Proposition 9.2 there exists $g \in L^1(|\mathbf{v}|)$ such that $0 \leq g(x) \leq 1$ for $|\mathbf{v}|$ -a.e. $x \in \mathbb{R}^d$ and $\sigma = g\mathbf{v}$. Writing the distributional formulation of $\text{div}(\sigma) = 0$ with the test function $\varphi(x) = \text{atan}(x_1)$, we get

$$- \int_{\mathbb{R}^d} \frac{g(x)\tau_1(x)}{1 + x_1^2} d|\mathbf{v}|(x) = 0,$$

hence $g(x) = 0$ for $|\mathbf{v}|$ -a.e. $x \in \mathbb{R}^d$. Therefore $\sigma = 0$ is the only cycle of \mathbf{v} . \square

Proof of Theorem 1.4. Suppose that $\mathbf{v} \in \mathcal{M}_{\text{loc}}(\mathbb{R}^d; \mathbb{R}^d)$ satisfies (i)–(iii) from Definition 1.3.

Let $\omega \in C_c^\infty(\mathbb{R}^d)$ be a nonnegative function such that $\omega(x) = 1$ if $|x| \leq 1$ and $\omega(x) = 0$ if $|x| \geq 2$. Let $h > 0$ and $r > 0$ and let $R > 0$ be such that $r^2 + h^2 < R^2$ and $r + c^{-1}h < R$, where $c > 0$ is the constant from Definition 1.3.

Let $\mu = f \cdot \mathbf{v}$, where $f(x) = \omega(\frac{x}{R})$. Clearly $\text{div} \mu$ belongs to $\mathcal{M}(\mathbb{R}^d)$ and is concentrated on

$$A := \{x \in \mathbb{R}^d : x_d \geq 0, R \leq |x| \leq 2R\}.$$

Moreover, μ is acyclic by Corollary 9.3.

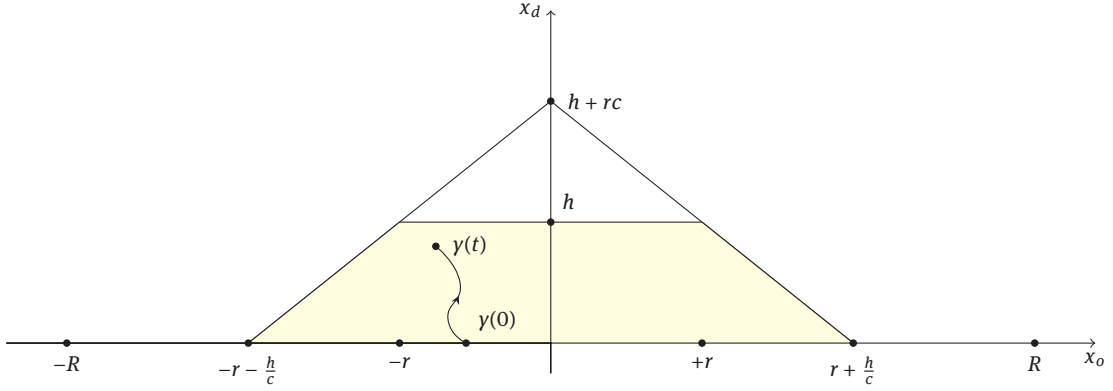


Figure 1: The region depicted in yellow is the set T defined in (9.3) (in the case $d = 2$). In the proof of Theorem 1.4, we show that $\eta(\Gamma_T) = 0$, i.e. the set of curves γ such that $\mu_\gamma(T) > 0$ is η -negligible. From this we deduce that $\mu(T) = 0$.

For any $x \in \mathbb{R}^d$ let $x_o := (x_1, \dots, x_{d-1})$. Let

$$T := \{y \in \mathbb{R}^d : 0 < y_d < h, |y_o| < r + c^{-1}(h - y_d)\} \tag{9.3}$$

(see Figure 1). Let $\eta \in \mathcal{M}_+(\Gamma)$ be given by Theorem 1.2 applied to μ (in particular, (1.2a)–(1.2c) hold). Let

$$\Gamma_T := \{\gamma \in \Gamma : |\mu_\gamma|(T) > 0\}.$$

Note that $\mu = \tau f|v| = \tau|\mu|$ is the polar decomposition of μ . Hence by Lemma 9.1 for η -a.e. $\gamma \in \Gamma_T$ for a.e. $z \in [0, 1]$ we have

$$\gamma'(z) = \tau(\gamma(z))|\gamma'(z)|.$$

Writing this equation for γ_d and γ_o separately and using condition (iii) from Definition 1.3, we get

$$|\gamma'_o(z)| = |\tau_o(\gamma(z))| \cdot |\gamma'(z)| \leq |\tau(\gamma(z))| \cdot |\gamma'(z)| \leq \frac{1}{c} \tau_d(\gamma(z)) \cdot |\gamma'(z)| = \frac{1}{c} \gamma'_d(z). \tag{9.4}$$

For η -a.e. $\gamma \in \Gamma_T$ there exists $t \in (0, 1)$ such that $\gamma(t) \in T$. Then by inequality (9.4) we obtain

$$|\gamma_o(0)| \leq |\gamma_o(t)| + \left| \int_0^t \gamma'_o(z) dz \right| \leq |\gamma_o(t)| + \frac{1}{c} (\gamma_d(t) - \gamma_d(0)),$$

hence $\gamma(0) \in \bar{T}$. Clearly $\gamma'_d \geq 0$ a.e., so $\gamma_d(0) \leq \gamma_d(t) \leq h$, since $\gamma(t) \in T$.

Note that for η -a.e. $\gamma \in \Gamma_T$ we have $\gamma(1) \neq \gamma(0)$. Indeed, otherwise the measure

$$\sigma := \int_{\{\gamma \in \Gamma : \gamma(0) = \gamma(1)\}} \mu_\gamma d\eta(\gamma)$$

would be a nonzero cycle of μ , which is not possible since μ is acyclic. Therefore for η -a.e. $\gamma \in \Gamma_T$,

$$|\operatorname{div} \mu_\gamma|(\bar{T}) = \delta_{\gamma(0)}(\bar{T}) + \delta_{\gamma(1)}(\bar{T}) \geq 1$$

since $\gamma(0) \in \bar{T}$. But $|\operatorname{div} \mu|$ is concentrated on A and $A \cap T = \emptyset$, hence

$$\eta(\Gamma_T) = \int_{\Gamma_T} 1 d\eta(\gamma) \leq \int_{\Gamma_T} |\operatorname{div} \mu_\gamma|(\bar{T}) d\eta(\gamma) = |\operatorname{div} \mu|(\bar{T}) = 0$$

and therefore

$$|\mu|(T) = \int_{\Gamma_T} |\mu_\gamma|(T) d\eta(\gamma) = 0.$$

By the arbitrariness of h and r we conclude that $\mu = 0$. □

A Decomposition Theorem for FV functions

We begin with the following definition.

Definition A.1. A function $f \in \text{FV}(\mathbb{R}^d)$ is said to be *monotone* if the sets $\{f > t\}$ and $\{f \leq t\}$ are indecomposable for a.e. $t \in \mathbb{R}$.

Notice that, by Remark 2.18, a function f such that the superlevel sets $\{f > t\}$ are *simple* for a.e. $t \in \mathbb{R}$ is necessarily monotone.

The goal of this appendix is to give a self-contained proof of the following theorem (see also [7]).

Theorem A.2. For any $f \in \text{FV}(\mathbb{R}^d)$ there exists an at most countable family $\{f_i\}_{i \in I} \subset \text{FV}(\mathbb{R}^d)$ of monotone functions such that

$$f = \sum_{i \in I} f_i \quad \text{and} \quad |Df| = \sum_{i \in I} |Df_i|. \quad (\text{A.1})$$

In particular,

$$\|f\|_{\text{FV}} = \sum_{i \in I} \|f_i\|_{\text{FV}}.$$

Remark A.3. Observe that from the embeddings of FV (see Theorem 2.3) the first series in (A.1) converges also in $L^{1^*}(\mathbb{R}^d)$ but, in general, we cannot improve this to convergence in $L^1(\mathbb{R}^d)$. Secondly, we remark that the decomposition provided in Theorem A.2 is not unique: we refer the reader to the counterexample presented in the paper [7].

The proof of Theorem A.2 will be presented at the end of the appendix and it requires some preliminary lemmas.

Lemma A.4. Let $\varphi, \psi \in \text{FV}(\mathbb{R}^d)$ and assume $0 \leq \psi \leq \varphi$.

(1) If for a.e. $t \in \mathbb{R}$ it holds

$$P(\{\varphi > t\}) = P(\{\varphi > t\} \setminus \{\psi > t\}) + P(\{\psi > t\}), \quad (\text{A.2})$$

then

$$\|\varphi\|_{\text{FV}} = \|\varphi - \psi\|_{\text{FV}} + \|\psi\|_{\text{FV}}.$$

(2) If for a.e. $t \in \mathbb{R}$ it holds

$$P(\{\psi > t\}) = P(\{\varphi > t\} \setminus \{\psi > t\}) + P(\{\varphi > t\}), \quad (\text{A.3})$$

then

$$\|\psi\|_{\text{FV}} = \|\varphi\|_{\text{FV}} + \|\varphi - \psi\|_{\text{FV}}.$$

Proof. We present the proof of the two claims.

(1) Concerning the first point, it suffices to show

$$\|D\varphi\|_{\mathcal{M}} \geq \|D(\varphi - \psi)\|_{\mathcal{M}} + \|D\psi\|_{\mathcal{M}},$$

because the other inequality is trivial by the triangle inequality. Using the layer cake representation and Fubini's Theorem we get

$$\begin{aligned} \|D(\varphi - \psi)\|_{\mathcal{M}} &= \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} (\varphi(x) - \psi(x)) \cdot \text{div } \omega(x) \, dx = \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} \int_0^{\infty} (\mathbb{1}_{\{\varphi > t\}}(x) - \mathbb{1}_{\{\psi > t\}}(x)) \cdot \text{div } \omega(x) \, dt \, dx \\ &= \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} \int_0^{\infty} \mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}(x) \cdot \text{div } \omega(x) \, dt \, dx = \sup_{\|\omega\|_{\infty} \leq 1} \int_0^{\infty} \int_{\mathbb{R}^d} \mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}(x) \cdot \text{div } \omega(x) \, dx \, dt \\ &\leq \sup_{\|\omega\|_{\infty} \leq 1} \int_0^{\infty} \langle D\mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}, \omega \rangle \, dt \leq \int_0^{\infty} P(\{\varphi > t\} \setminus \{\psi > t\}) \, dt = \int_0^{\infty} P(\{\varphi > t\}) - P(\{\psi > t\}) \, dt, \end{aligned}$$

where the last equality follows from (A.2). Applying again the Coarea Formula, we obtain the conclusion.

(2) The proof of the second claim is similar to the proof of the first one. Notice that $|Dw| = |D(-w)|$ as measures for any $w \in \text{FV}(\mathbb{R}^d)$ hence

$$\|D\psi\|_{\mathcal{M}} \leq \|D\varphi\|_{\mathcal{M}} + \|D(\varphi - \psi)\|_{\mathcal{M}},$$

which is equivalent to

$$\|D(\varphi - \psi)\|_{\mathcal{M}} \geq \|D\psi\|_{\mathcal{M}} - \|D\varphi\|_{\mathcal{M}}.$$

It thus remains to show

$$\|D(\varphi - \psi)\|_{\mathcal{M}} \leq \|D\psi\|_{\mathcal{M}} - \|D\varphi\|_{\mathcal{M}}.$$

By layer cake representation and Fubini, as in Point (1), we have

$$\begin{aligned} \|D(\varphi - \psi)\|_{\mathcal{M}} &= \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} (\varphi(x) - \psi(x)) \cdot \text{div } \omega(x) \, dx = \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} \int_0^{\infty} (\mathbb{1}_{\{\varphi > t\}}(x) - \mathbb{1}_{\{\psi > t\}}(x)) \cdot \text{div } \omega(x) \, dt \, dx \\ &= \sup_{\|\omega\|_{\infty} \leq 1} \int_{\mathbb{R}^d} \int_0^{\infty} \mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}(x) \cdot \text{div } \omega(x) \, dt \, dx = \sup_{\|\omega\|_{\infty} \leq 1} \int_0^{\infty} \int_{\mathbb{R}^d} \mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}(x) \cdot \text{div } \omega(x) \, dx \, dt \\ &\leq \sup_{\|\omega\|_{\infty} \leq 1} \int_0^{\infty} \langle D\mathbb{1}_{\{\varphi > t\} \setminus \{\psi > t\}}, \omega \rangle \, dt \leq \int_0^{\infty} P(\{\varphi > t\} \setminus \{\psi > t\}) \, dt = \int_0^{\infty} P(\{\psi > t\}) - P(\{\varphi > t\}) \, dt, \end{aligned}$$

where the last equality follows from (A.3). Again the application of the Coarea Formula yields the desired conclusion. \square

Lemma A.5 (From superlevel sets to function). *Let $I \subset [0, +\infty)$ be an interval and let $(A_t)_{t \in I}$ be a family of sets such that $t, s \in I$ with $s < t$ implies $A_t \subset A_s$. Then there exists a measurable function $w: \mathbb{R}^d \rightarrow [0, +\infty)$ such that $\{w > t\} = A_t$ (up to Lebesgue negligible subsets) for a.e. $t \in I$.*

Proof. Due to monotonicity of the family $(A_t)_{t \in I}$, the function $h(t) := |A_t|$ is non-increasing on I . Therefore there exists a Lebesgue negligible set $N \subset I$ such that h is continuous at every $t \in I \setminus N$. Let $Q \subseteq I \setminus N$ be a countable set, which is dense in I . For any $x \in \mathbb{R}^d$ we define

$$w(x) := \sup_{t \in Q} (t \cdot \mathbb{1}_{A_t}(x)).$$

Clearly w is Lebesgue measurable. By definition of w for any $s \in I \setminus N$,

$$\{w > s\} = \bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t.$$

Since for any $s < t$ it holds $|A_t \setminus A_s| = 0$ and Q is countable, it follows that

$$\left| \left(\bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t \right) \setminus A_s \right| = 0.$$

On the other hand, let $\varepsilon := |A_s \setminus \bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t|$. For any $t \in Q \cap (s, +\infty) \cap I$ we have $A_t \subset A_s$, hence

$$\bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t \subset A_s.$$

In particular, we can estimate

$$|A_s| = \left| A_s \setminus \bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t \right| + \left| \bigcup_{t \in Q \cap (s, +\infty) \cap I} A_t \right| \geq \varepsilon + |A_t|.$$

Since Q is dense in $(s, +\infty) \cap I$ and h is continuous at s , the only possible case is $\varepsilon = 0$. We have thus proved that $|\{g > s\} \Delta A_s| = 0$ for a.e. $s \in I$ and this concludes the proof. \square

The following lemma is a building block of the proof of Theorem A.2. It allows to “extract” from a non-negative FV function (whose superlevel sets in general are not indecomposable) a non-trivial function with indecomposable superlevel sets.

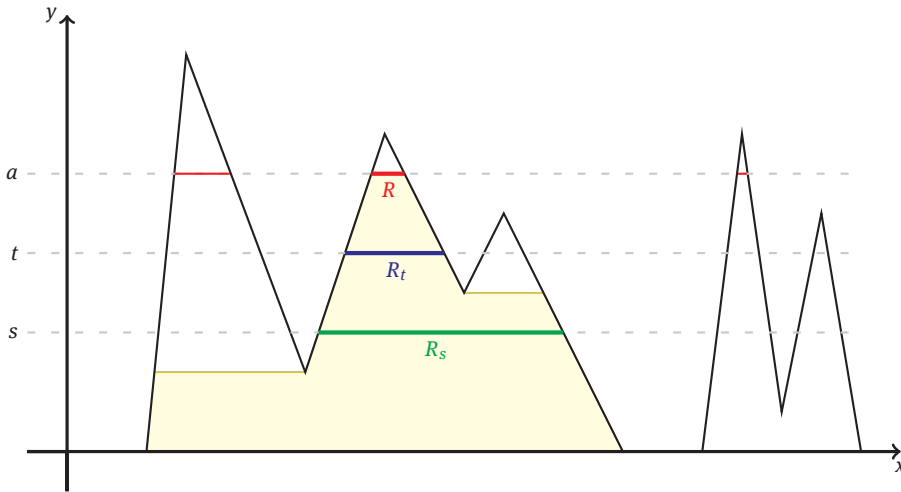


Figure 2: Situation described in the proof of Lemma A.6. The black curve represents the graph of a generic function $f \in \text{FV}(\mathbb{R}^d)$. The red segments make up the level set E_a . The red, thick segment is the component R and the blue and green ones are respectively R_t and R_s . The area depicted in yellow is the subgraph of the function g , whose superlevel sets are indecomposable.

Lemma A.6 (Extraction Lemma I). *Let $f \in \text{FV}(\mathbb{R}^d)$ and assume f is not identically zero and non-negative. Then there exists $g \in \text{FV}(\mathbb{R}^d)$ with $0 \leq g \leq f$ and $g \neq 0$ such that:*

- (i) *for a.e. $t \geq 0$ the set $\{g > t\}$ is indecomposable,*
- (ii) *it holds $\|f\|_{\text{FV}} = \|f - g\|_{\text{FV}} + \|g\|_{\text{FV}}$.*

Proof. For any $t \geq 0$ let $E_t := \{f > t\}$. Since $f \in \text{FV}(\mathbb{R}^d)$, there exists a Lebesgue negligible set $N \subseteq (0, +\infty)$ such that for any $t \in (0, +\infty) \setminus N$ the set E_t has finite perimeter. Let E_t^k denote the k -th M -connected component of E_t , $t \in (0, +\infty) \setminus N$.

Fix some $a > 0$ such that $|E_a| > 0$. Let R be some M -connected component of E_a . For any $t \in (0, a) \setminus N$ we have $E_t \supseteq E_a \supseteq R$, and R is indecomposable, hence by Theorem 2.10 there exists a unique $j = j(t)$ such that

$$|R \setminus E_t^{j(t)}| = 0.$$

Let $R_t := E_t^{j(t)}$, $t \in (0, a) \setminus N$. Note that for any $s, t \in (0, a) \setminus N$ with $s < t$ it holds that

$$|R_t \setminus R_s| = 0.$$

Indeed, $E_s \supseteq E_t \supseteq R_t$ and R_t is indecomposable, hence again by Theorem 2.10 there exists a unique k such that $|R_t \setminus E_s^k| = 0$. But $|R \setminus R_t| = 0$, hence

$$E_s^k \setminus R = (E_s^k \cap R_t \cap R^c) \cup (E_s^k \cap R_t^c \cap R^c) \subseteq (R_t \setminus R) \cup (E_s^k \setminus R_t)$$

is Lebesgue negligible. Therefore $k = j(s)$ by the uniqueness of $j(s)$. Applying now Lemma A.5, we can construct a function $g: \mathbb{R}^d \rightarrow [0, a]$ such that $\{g > s\} = R_s$ (up to Lebesgue negligible subsets) for a.e. $s \in (0, a)$. (See Figure 2.)

Observe that $\|\bar{f}\|_{\text{FV}} = \|g\|_{\text{FV}} + \|\hat{f}\|_{\text{FV}}$, where $\bar{f}(x) = \min(a, f(x))$ and $\hat{f} := f - \bar{f}$. For a.e. $t \in (0, a)$ we have

$$\{\bar{f} > t\} = \{f > t\} = E_t^{j(t)} \cup \bigcup_{k \neq j(t)} E_t^k,$$

hence by the construction of g and Proposition 2.12,

$$P(\{\bar{f} > t\}) = P(\{g > t\}) + P(\{f > t\} \setminus \{g > t\}).$$

Hence by Lemma A.4 we have

$$\|\bar{f}\|_{\text{FV}} = \|g\|_{\text{FV}} + \|\bar{f} - g\|_{\text{FV}}.$$

Then by the triangle inequality

$$\|f\|_{\text{FV}} = \|\tilde{f}\|_{\text{FV}} + \|\hat{f}\|_{\text{FV}} = \|g\|_{\text{FV}} + \|\tilde{f} - g\|_{\text{FV}} + \|\hat{f}\|_{\text{FV}} \geq \|g\|_{\text{FV}} + \|\tilde{f} - g + \hat{f}\|_{\text{FV}}$$

and $\|f\|_{\text{FV}} = \|g + \tilde{f} - g + \hat{f}\|_{\text{FV}} \leq \|g\|_{\text{FV}} + \|\tilde{f} - g + \hat{f}\|_{\text{FV}}$, hence property (ii) follows. \square

Lemma A.7 (Extraction Lemma II). *Let $f \in \text{FV}(\mathbb{R}^d)$ and assume f is not identically zero and non-negative. Then there exists $h \in \text{FV}(\mathbb{R}^d)$ with $h \neq 0$ such that:*

- (i) *for a.e. $t \geq 0$ the set $\{h > t\}$ is simple,*
- (ii) *it holds $\|f\|_{\text{FV}} = \|f - h\|_{\text{FV}} + \|h\|_{\text{FV}}$.*

Proof. First of all, we apply Lemma A.6 and we obtain a function $g \in \text{FV}(\mathbb{R}^d)$ such that $G_t := \{g > t\}$ is indecomposable for a.e. $t \geq 0$ and

$$\|f\|_{\text{FV}} = \|f - g\|_{\text{FV}} + \|g\|_{\text{FV}}. \quad (\text{A.4})$$

Let us now work on the function g . By the construction of g , for a.e. $t \geq 0$ the set G_t is indecomposable. Fix some $a > 0$ such that $|G_a| > 0$ and G_a is not simple (otherwise there is nothing to prove): let us denote by $\{F_t^i\}_{i \in I_t}$ the non-empty family of holes of G_t (i.e. $\mathcal{CC}^M(\mathbb{R}^d \setminus G_t) = \{F_t^i\}_{i \in I_t}$).

Observe that, if H is an hole of G_a , for any $t \in (a, +\infty) \setminus N$, we have $G_t \subseteq G_a$, and hence $G_t^c \supseteq G_a^c \supseteq H$: this means that H is an hole of G_t for any $t \in (a, +\infty) \setminus N$: by the uniqueness claim in Theorem 2.10 there exists a unique $j = j(t)$ such that $|H \setminus F_t^{j(t)}| = 0$.

For any $t \in (0, a)$ define $S_t := \text{sat}(G_t)$. Observe that the sequence $(S_t)_{t \in (0, a)}$ is monotone [1, Proposition 6 (iii)] and thus, applying Lemma A.5, we obtain a function $h: \mathbb{R}^d \rightarrow \mathbb{R}$ such that $\{h > r\} = S_r$ (up to Lebesgue negligible subsets) for a.e. $r \in (0, a)$. By construction the function h is non-negative and $\{h > r\}$ is simple for a.e. $r \in (0, a)$, because the saturation of an indecomposable set is simple. It thus remains to show property (ii) of the statement. For, notice preliminarily, that $h - g \geq 0$ by the construction of h ; by [1, Proposition 9], it holds for any $t \in (a, +\infty) \setminus N$,

$$P(G_t) = P(\text{sat}(G_t)) + P\left(\bigcup_{i \in I_t} F_t^i\right),$$

which can be also written as

$$P(\{g > t\}) = P(\{h > t\}) + P(\{h > t\} \setminus \{g > t\}).$$

We are now in a position to apply Lemma A.4 (ii), choosing $\varphi := h$ and $\psi := g$ (which is possible since $h \geq g$): we obtain

$$\|g\|_{\text{FV}} = \|h\|_{\text{FV}} + \|g - h\|_{\text{FV}}. \quad (\text{A.5})$$

It is now easy to check that property (ii) follows combining (A.4) with (A.5) – and the triangle inequality:

$$\begin{aligned} \|f\|_{\text{FV}} &\leq \|f - h\|_{\text{FV}} + \|h\|_{\text{FV}} \\ &\leq \|f - g\|_{\text{FV}} + \|g - h\|_{\text{FV}} + \|h\|_{\text{FV}} \\ &\stackrel{(\text{A.4})}{=} \|f\|_{\text{FV}} - \|g\|_{\text{FV}} + \|g - h\|_{\text{FV}} + \|h\|_{\text{FV}} \\ &\stackrel{(\text{A.5})}{=} \|f\|_{\text{FV}} - \|g\|_{\text{FV}} + \|g\|_{\text{FV}} - \|h\|_{\text{FV}} + \|h\|_{\text{FV}} = \|f\|_{\text{FV}}, \end{aligned}$$

and this completes the proof. \square

Lemma A.8 (Extraction Lemma III). *Let $f \in \text{FV}(\mathbb{R}^d)$ and assume f is not identically zero. Then there exists $m \in \text{FV}(\mathbb{R}^d)$ with $m \neq 0$ such that:*

- (i) *m is monotone and $\text{sign } m = \text{constant a.e.}$,*
- (ii) *it holds $\|f\|_{\text{FV}} = \|f - m\|_{\text{FV}} + \|m\|_{\text{FV}}$.*

Proof. Let us decompose $f = f^+ - f^-$. Suppose $\|f^+\|_{\text{FV}} > 0$. Since $f^+ \geq 0$, we can apply Lemma A.6 to f^+ , thus obtaining a function $u \geq 0$ such that $\{u > t\}$ is indecomposable for a.e. $t > 0$ and it holds

$$\|f^+\|_{\text{FV}} = \|f^+ - u\|_{\text{FV}} + \|u\|_{\text{FV}}. \quad (\text{A.6})$$

Applying now Lemma A.7 to $u \geq 0$, we obtain a function $m \in \text{FV}(\mathbb{R}^d)$ such that for a.e. $t \geq 0$ the set $\{m > t\}$ is simple and it holds

$$\|u\|_{\text{FV}} = \|u - m\|_{\text{FV}} + \|m\|_{\text{FV}}. \quad (\text{A.7})$$

By the triangle inequality,

$$\begin{aligned} \|f\|_{\text{FV}} &\leq \|f - m\|_{\text{FV}} + \|m\|_{\text{FV}} \\ &\leq \|f^+ - m\|_{\text{FV}} + \|f^-\|_{\text{FV}} + \|m\|_{\text{FV}} \\ &\leq \|f^+ - u\|_{\text{FV}} + \|u - m\|_{\text{FV}} + \|f^-\|_{\text{FV}} + \|m\|_{\text{FV}} \\ &\stackrel{(\text{A.7})}{=} \|f^+ - u\|_{\text{FV}} + \|u\|_{\text{FV}} + \|f^-\|_{\text{FV}} \\ &\stackrel{(\text{A.6})}{=} \|f^+\|_{\text{FV}} + \|f^-\|_{\text{FV}} = \|f\|_{\text{FV}}, \end{aligned}$$

hence property (ii) holds true. Since the function m is monotone, this concludes the proof in the case $\|f^+\|_{\text{FV}} > 0$. It remains to consider the case in which $f^+ \equiv 0$. If $f^- \equiv 0$, there is nothing to prove; if $\|f^-\|_{\text{FV}} > 0$, then we repeat the same argument above for the function $\tilde{f} := -f \in \text{FV}(\mathbb{R}^d)$. We end up with a monotone function \tilde{m} of constant sign such that

$$\|\tilde{f}\|_{\text{FV}} = \|\tilde{f} - \tilde{m}\|_{\text{FV}} + \|\tilde{m}\|_{\text{FV}},$$

which is clearly equivalent to property (ii) (renaming $-\tilde{m}$ as m). \square

Now we prove Theorem A.2 using Lemma A.8 and transfinite induction:

Proof of Theorem A.2. Let $X := \{g \in \text{FV}(\mathbb{R}^d) : g \text{ is monotone and } \|g\|_{\text{FV}} > 0\}$. For any $h \in \text{FV}(\mathbb{R}^d)$ let

$$Y(h) := \{g \in X : \|h\|_{\text{FV}} = \|h - g\|_{\text{FV}} + \|g\|_{\text{FV}}\}.$$

Note that by Lemma A.8 $Y(h) = \emptyset$ if and only if $h \equiv 0$. Ultimately, let $\mathfrak{s} : \mathcal{P}(\text{FV}(\mathbb{R}^d)) \rightarrow \text{FV}(\mathbb{R}^d)$ denote a choice function (given by the Axiom of Choice).

Let us define, for any ordinal $\alpha < \omega_1$ (where ω_1 is the first uncountable ordinal) and any transfinite sequence $\{g_\xi\}_{\xi < \alpha} \subset X \cup \{\infty\}$,

$$E(\{g_\xi\}_{\xi < \alpha}) := \begin{cases} \infty, & \text{if } \infty \in \{g_\xi\}_{\xi < \alpha} \text{ or if } \sum_{\xi < \alpha} \|g_\xi\|_{\text{FV}} = \infty, \\ \mathfrak{s}(Y(f - \sum_{\xi < \alpha} g_\xi)), & \text{if } \sum_{\xi < \alpha} \|g_\xi\|_{\text{FV}} < \infty \text{ and } Y(f - \sum_{\xi < \alpha} g_\xi) \neq \emptyset, \\ 0, & \text{if } \sum_{\xi < \alpha} \|g_\xi\|_{\text{FV}} < \infty \text{ and } Y(f - \sum_{\xi < \alpha} g_\xi) = \emptyset. \end{cases}$$

By transfinite recursion (see e.g. [11, p. 21]) there exists a transfinite sequence $\{g_\alpha\}_{\alpha < \omega_1}$ such that

$$g_\alpha = E(\{g_\xi\}_{\xi < \alpha})$$

for any $\alpha < \omega_1$.

Note that for any $\alpha < \omega_1$ the following properties hold:

$$\infty \notin \{g_\xi\}_{\xi < \alpha}, \quad (\text{A.8a})$$

$$\sum_{\xi < \alpha} \|g_\xi\|_{\text{FV}} \leq \|f\|_{\text{FV}}, \quad (\text{A.8b})$$

$$\|f\|_{\text{FV}} = \left\| f - \sum_{\xi < \alpha} g_\xi \right\|_{\text{FV}} + \sum_{\xi < \alpha} \|g_\xi\|_{\text{FV}}. \quad (\text{A.8c})$$

Observe that (A.8b) follows from (A.8c), but without (A.8b) the term $\sum_{\xi < \alpha} g_\xi$ in (A.8a) is not well-defined. Indeed, these properties trivially hold for $\alpha = 0$. Let $\beta < \omega_1$ and suppose that these properties hold for any $\alpha < \beta$. In order to show that (A.8a)–(A.8c) hold with $\alpha = \beta$, we consider two cases.

First, if β is not a limit ordinal, then $\beta = \gamma + 1$ for some ordinal γ , so by the definition of $\{g_\xi\}_{\xi < \omega_1}$ we have

$$g_{\gamma+1} = \mathfrak{s}\left(Y\left(f - \sum_{\xi < \gamma} g_\xi\right)\right).$$

Hence

$$\|f\|_{\text{FV}} = \left\| f - \sum_{\xi < \gamma} g_\xi \right\|_{\text{FV}} + \sum_{\xi < \gamma} \|g_\xi\|_{\text{FV}} = \left\| f - \sum_{\xi < \gamma} g_\xi - g_\gamma \right\|_{\text{FV}} + \|g_\gamma\|_{\text{FV}} + \sum_{\xi < \gamma} \|g_\xi\|_{\text{FV}}$$

and it follows that (A.8a)–(A.8c) hold with $\alpha = \gamma + 1$.

Second, if β is a limit ordinal, then $\beta = \bigcup_{\alpha < \beta} \alpha$. Consequently,

$$\{g_\xi\}_{\xi < \beta} = \bigcup_{\alpha < \beta} \{g_\xi\}_{\xi < \alpha},$$

hence condition (A.8a) holds with $\alpha = \beta$. Furthermore, since β is at most countable, we can enumerate it as $\beta = \{\alpha_n\}_{n \in \mathbb{N}}$. Let $A_n := \alpha_1 \cup \dots \cup \alpha_n$ (note that for any $n \in \mathbb{N}$ there exists $m \in \{1, \dots, n\}$ such that $A_n = \alpha_m$). Since $\beta = \bigcup_{\alpha < \beta} \alpha = \bigcup_{n \in \mathbb{N}} A_n$, we have

$$\sum_{\xi < \beta} \|g_\xi\|_{\text{FV}} = \sum_{\xi < \beta} \left(\sup_{n \in \mathbb{N}} \mathbb{1}_{A_n}(\xi) \right) \|g_\xi\|_{\text{FV}} = \sup_{n \in \mathbb{N}} \sum_{\xi \in A_n} \|g_\xi\|_{\text{FV}} \leq \sup_{\alpha < \beta} \|g_\alpha\|_{\text{FV}} \leq \|f\|_{\text{FV}},$$

hence (A.8b) holds with $\alpha = \beta$. Consequently,

$$\sum_{\xi < \beta} g_\xi = \lim_{n \rightarrow \infty} \sum_{\xi \in A_n} g_\xi$$

and

$$\sum_{\xi < \beta} \|g_\xi\|_{\text{FV}} = \lim_{n \rightarrow \infty} \sum_{\xi \in A_n} \|g_\xi\|_{\text{FV}}.$$

Writing (A.8c) with $\alpha = A_n$ and passing to the limit as $n \rightarrow \infty$, we conclude that (A.8c) holds with $\alpha = \beta$. We have thus shown that (A.8a)–(A.8c) hold with $\alpha = \beta$. Hence by transfinite induction (A.8a)–(A.8c) hold for any $\alpha < \omega_1$.

By (A.8b) for any $\varepsilon > 0$ the set $\{\alpha < \omega_1 : \|g_\alpha\|_{\text{FV}} > \varepsilon\}$ is finite and thus the set $A := \{\alpha < \omega_1 : \|g_\alpha\|_{\text{FV}} > 0\}$ is at most countable. Setting $\gamma := \sup A$, we have $g_{\gamma+1} = 0$. As already noted above, by Lemma A.8 this means that

$$f = \sum_{\xi < \gamma} g_\xi$$

and

$$\|f\|_{\text{FV}} = \sum_{\xi < \gamma} \|g_\xi\|_{\text{FV}}$$

by (A.8c).

By the triangle inequality,

$$|Df| \leq \sum_{\xi < \gamma} |Dg_\xi|.$$

If this inequality were strict, we would have

$$\|f\|_{\text{FV}} = |Df|(\mathbb{R}^d) < \sum_{\xi < \gamma} |Dg_\xi|(\mathbb{R}^d) = \sum_{\xi < \gamma} \|g_\xi\|_{\text{FV}} = \|f\|_{\text{FV}},$$

which is a contradiction. □

Acknowledgment: The first author thanks Graziano Crasta and Annalisa Malusa for interesting discussions during the preparation of the paper; he also kindly acknowledges Gian Paolo Leonardi for introducing him to the problem of rigidity of divergence-free vector fields.

Funding: The first author was supported by ERC Starting Grant 676675 FLIRT. This project has received funding from the European Research Council (ERC) under the European Union’s Horizon 2020 research and innovation programme under grant agreement No. 757254 (SINGULARITY). The work of the second author was supported by RFBR Grant 18-31-00279.

References

- [1] L. Ambrosio, V. Caselles, S. Masnou and J.-M. Morel, Connected components of sets of finite perimeter and applications to image processing, *J. Eur. Math. Soc. (JEMS)* **3** (2001), no. 1, 39–92.
- [2] L. Ambrosio and G. Crippa, Existence, uniqueness, stability and differentiability properties of the flow associated to weakly differentiable vector fields, in: *Transport Equations and Multi-D Hyperbolic Conservation Laws*, Lect. Notes Unione Mat. Ital. 5, Springer, Berlin (2008), 3–57.
- [3] L. Ambrosio, N. Fusco and D. Pallara, *Functions of Bounded Variation and Free Discontinuity Problems*, Oxford Math. Monogr., The Clarendon, New York, 2000.
- [4] G. Anzellotti, Pairings between measures and bounded functions and compensated compactness, *Ann. Mat. Pura Appl. (4)* **135** (1983), 2933–318.
- [5] S. Bianchini, P. Bonicatto and N. A. Gusev, Renormalization for autonomous nearly incompressible BV vector fields in two dimensions, *SIAM J. Math. Anal.* **48** (2016), no. 1, 1–33.
- [6] S. Bianchini and N. A. Gusev, Steady nearly incompressible vector fields in two-dimension: Chain rule and renormalization, *Arch. Ration. Mech. Anal.* **222** (2016), no. 2, 451–505.
- [7] S. Bianchini and D. Tonon, A decomposition theorem for BV functions, *Commun. Pure Appl. Anal.* **10** (2011), no. 6, 1549–1566.
- [8] V. I. Bogachev, *Measure Theory*, Springer, Berlin, 2006.
- [9] L. C. Evans and R. F. Gariepy, *Measure Theory and Fine Properties of Functions*, Stud. Adv. Math., CRC Press, Boca Raton, 1992.
- [10] A. Hatcher, *Algebraic Topology*, Cambridge University, Cambridge, 2002.
- [11] T. Jech, *Set Theory*, Springer Monogr. Math., Springer, Berlin, 2003.
- [12] V. I. Kolyada, On the metric Darboux property, *Anal. Math.* **9** (1983), no. 4, 291–312.
- [13] G. P. Leonardi and G. Saracco, Rigidity and trace properties of divergence-measure vector fields, *Adv. Calc. Var.* (2020), DOI 10.1515/acv-2019-0094.
- [14] F. Maggi, *Sets of Finite Perimeter and Geometric Variational Problems*, Cambridge Stud. Adv. Math. 135, Cambridge University, Cambridge, 2012.
- [15] E. Paolini and E. Stepanov, Decomposition of acyclic normal currents in a metric space, *J. Funct. Anal.* **263** (2012), no. 11, 3358–3390.
- [16] E. Paolini and E. Stepanov, Structure of metric cycles and normal one-dimensional currents, *J. Funct. Anal.* **264** (2013), no. 6, 1269–1295.
- [17] R. R. Phelps, *Lectures on Choquet’s Theorem*, 2nd ed., Lecture Notes in Math. 1757, Springer, Berlin, 2001.
- [18] W. Rudin, *Functional Analysis*, Int. Ser. Pure Appl. Math., McGraw-Hill, New York, 2006.
- [19] S. K. Smirnov, Decomposition of solenoidal vector charges into elementary solenoids, and the structure of normal one-dimensional flows, *Algebra i Analiz* **5** (1993), no. 4, 206–238.
- [20] E. Stepanov and D. Trevisan, Three superposition principles: Currents, continuity equations and curves of measures, *J. Funct. Anal.* **272** (2017), no. 3, 1044–1103.