Lecture 0: Summary (part 2)

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Contents

1	Sign	ned Measures and Radon-Nikodym Derivative	2
	1.1	Signed Measure	2
	1.2	Decomposition of Signed Measure	ç
		Lebesgue-Radon-Nikodym Theorem	
2	Diff	erentiation	6
	2.1	The Lebesgue Differentiation Theorem in One Dimension	7
	2.2	The Lebesgue Differentiation Theorem in \mathbb{R}^d	8
		2.2.1 Absolute Integrable Version	8
		2.2.2 Local Integrable Version	10
	2.3	Lebesgue Density and Radon-Nikodym Derivative	11
3	The	Fundamental Theorem of Calculus for Lebesgue Integral	12
	3.1	Functions of Bounded Variations	12
	3.2	The Second Fundamental Theorem of Calculus for Lebesgue Integral	15

1 Signed Measures and Radon-Nikodym Derivative

1.1 Signed Measure

- Definition (Signed Measure)
 - Let (X, \mathcal{B}) be a measure space. A <u>signed measure</u> on (X, \mathcal{B}) is a function $\nu : \mathcal{B} \to [-\infty, +\infty]$ such that
 - 1. (**Emptyset**) $\nu(\emptyset) = 0$;
 - 2. (Finiteness in One Direction) ν assumes at most one of the values $\pm \infty$;
 - 3. (*Countable Additivity*) if $\{E_j\}$ is a sequence of disjoint sets in \mathscr{B} , then $\nu\left(\bigcup_{j=1}^{\infty} E_j\right) = \sum_{j=1}^{\infty} \nu(E_j)$, where the latter converges absolutely if $\nu\left(\bigcup_{j=1}^{\infty} E_j\right)$ is finite.
- **Definition** A measure μ is *finite*, if $\mu(X) < \infty$; μ is σ -finite, if $X = \bigcup_{k=1}^{\infty} U_k$, $\mu(U_k) < \infty$.
- Remark Every signed measure can be represented as one of these two forms
 - 1. $\nu = \mu_+ \mu_-$, where at least one of μ_+, μ_- is a finite measure;
 - 2. μ is measure on \mathscr{B} , and $f: X \to [-\infty, +\infty]$ is extended μ -integrable with at least one of f_+ and f_- finite integrable. Then $\nu(A) = \int_X f \mathbb{1} \{A\} d\mu$ is a signed measure.
- Like unsigned measure, we have monotone downward and upward convergence:

Proposition 1.1 Let ν be a **signed measure** on (X, \mathcal{B}) .

1. (Upwards monotone convergence) If $E_1 \subseteq E_2 \subseteq ...$ are \mathscr{B} -measurable, then

$$\nu\left(\bigcup_{n=1}^{\infty} E_n\right) = \lim_{n \to \infty} \nu(E_n) = \sup_{n} \nu(E_n). \tag{1}$$

2. (Downwards monotone convergence) If $E_1 \supseteq E_2 \supseteq ...$ are \mathscr{B} -measurable, and $\nu(E_n) < \infty$ for at least one n, then

$$\nu\left(\bigcap_{n=1}^{\infty} E_n\right) = \lim_{n \to \infty} \nu(E_n) = \inf_n \nu(E_n). \tag{2}$$

• Definition (Positive Measure)

If ν is a signed measure on (X, \mathcal{B}) , a **set** $E \in \mathcal{B}$ is called **positive** (resp. **negative**, **null**) for ν if $\nu(F) \geq 0$ (resp. $\nu(F) \leq 0$, $\nu(F) = 0$) for **all B-measurable subset** of E (i.e. $F \in \mathcal{B}$ such that $F \subseteq E$).

In other word, E is ν -positive, ν -negative, ν -null if and only if $\nu(E \cap M) > 0$, $\nu(E \cap M) < 0$, $\nu(E \cap M) = 0$ for any M. Thus if $\nu(E) = \int_X f \mathbb{1}\{E\} d\mu$, then it corresponds to $\underline{f \geq 0}$, $f \leq 0$ and f = 0 for μ -almost everywhere $x \in E$.

• Lemma 1.2 [Folland, 2013]

Any measureable subset of a positive set is positive, and the union of any countable positive set is positive.

• Remark For two measures μ, ν on (X, \mathcal{B}) among which at least one of them is finite, the expression $\mu \geq \nu$ on E means that for every $F \subseteq E \in \mathcal{B}$, $(\mu - nu)(F) \geq 0$. That is, E is a positive set of $(\mu - nu)$.

1.2 Decomposition of Signed Measure

- Remark Given a signed measure ν , we can *partition* the space X into positive set (i.e. all of its measurable subsets have positive measure) and negative set (i.e. all of its measurable subsets have negative measure).
- Theorem 1.3 (The Hahn Decomposition Theorem)[Folland, 2013] If ν is a signed measure on (X, \mathcal{B}) , there exists a positive set P and a negative set N for ν such that $P \cup N = X$ and $P \cap N = \emptyset$. If P', N' is another such pair, then $P\Delta P' = N\Delta N'$ is null w.r.t. ν .
- Definition [Folland, 2013, Resnick, 2013]
 The decomposition of X = P ∪ N as X is a disjoint union of a positive set and a negative set is called a Hahn decomposition for ν.
- Remark Note that the Hahn decomposition is usually **not unique** as the ν -null set can be transferred between subparts P and N. To find unique decomposition, we need the following concepts:
- **Definition** [Folland, 2013] Two signed measures μ, ν on (X, \mathcal{B}) are mutually singular, or that ν is singular w.r.t. to μ , or vice versa, if and only if there exists a partition $E, F \in \mathcal{B}$ of X such that $E \cap F = \emptyset$ and $E \cup F = X$, E is null for μ and F is null for ν . Informal speaking, mutual singular means that μ and ν "live on disjoint sets". We describe it using perpendicular sign

$$\mu \perp \nu$$

• Theorem 1.4 (The Jordan Decomposition Theorem)[Folland, 2013] If ν is a signed measure on (X, \mathcal{B}) , there exists unique positive measure ν_+ and ν_- such that

$$\nu = \nu_+ - \nu_-$$
 and $\nu_+ \perp \nu_-$.

• **Definition** The two positive measures ν_+, ν_- are called the **positive** and **negative variations** of ν , and $\nu = \nu_+ - \nu_-$ is called the **Jordan decomposition** of ν .

Furthermore, define the **total variations** of ν as the measure $|\nu|$ such that

$$|\nu| = \nu_+ + \nu_-.$$

- Proposition 1.5 Let ν, μ be signed measures on (X, \mathcal{B}) and $|\nu|$ is the total variations of ν . Then
 - 1. $E \in \mathcal{B}$ is ν -null if and only if $|\nu|(E) = 0$
 - 2. $\nu \perp \mu$ if and only if $|\nu| \perp \mu$ if and only if $(\nu_+ \perp \mu) \wedge (\nu_- \perp \mu)$.

- Proposition 1.6 If ν_1, ν_2 are signed measures that both omit $\pm \infty$, then $|\nu_1 + \nu_2| \leq |\nu_1| + |\nu_2|$
- Exercise 1.7 Let ν be a signed measure on (X, \mathcal{B}) .
 - 1. $L^1(\nu) = L^1(|\nu|);$
 - 2. If $f \in L^1(\nu)$, then

$$\left| \int_X f d\nu \right| \le \int_X |f| \, d|\nu|$$

3. If $E \in \mathcal{B}$, then

$$\left|\nu\right|(E) = \sup\left\{\left|\int_{E} f d\nu\right| : \left|f\right| \le 1\right\}$$

- Remark We recall that ν assume at most one of values on $\pm \infty$:
 - 1. If ν does not take $+\infty$, then $\nu_+(X) = \nu(P) < \infty$ is a finite measure;
 - 2. if ν does not take $-\infty$, then $\nu_{-}(X) = -\nu(N) < \infty$ is a finite measure.

In particular, if the range of ν is contained in \mathbb{R} , then ν is bounded.

- Remark We observe that ν is of form $\underline{\nu(E)} = \int_E f d\mu$ where $|\nu| = \mu$ and $f = \mathbb{1}_P \mathbb{1}_N$ and $X = P \cup N$ being a Hahn decomposition for ν .
- Remark (Integration with respect to Signed Measure) Let ν be signed measures on (X, \mathcal{B}) and $\nu = \nu_+ - \nu_-$ is the Jordan decomposition of ν then

$$\int_X f d\nu = \int_X f d\nu_+ - \int_X f d\nu_-$$

for all $f \in L^1(X, \nu)$.

• **Definition** A signed measure ν is called σ -finite if $|\nu|$ is σ -finite.

1.3 Lebesgue-Radon-Nikodym Theorem

ullet **Definition** [Folland, 2013]

Suppose ν is a signed measure on (X, \mathcal{B}) and μ is a positive measure on (X, \mathcal{B}) . Then ν is said to be absolutely continuous w.r.t. μ and write

$$\nu \ll \mu$$

if $\nu(E) = 0$ for every $E \in \mathscr{B}$ for which $\mu(E) = 0$.

- Proposition 1.8 Suppose ν is a signed measure on (X, \mathcal{B}) , ν_+, ν_- are positive and negative variation of ν and $|\nu|$ is the total variation. Then $\nu \ll \mu$ if and only if $|\nu| \ll \mu$ if and only if $(\nu_+ \ll \mu) \wedge (\nu_- \ll \mu)$.
- Remark Absolutly continuity is in a sense antithesis (i.e. direct opposite) of mutual singularity. More precisely, if $\nu \perp \mu$ and $\nu \ll \mu$, then $\nu = 0$, since E, F are disjoint sets such that $E \cup F = X$, and $\mu(E) = |\nu|(F) = 0$, then $\nu \ll \mu$ implies that $|\nu|(E) = 0$. One can extend the notion of absolute continuity to the case where μ is a signed measure (namely, $\nu \ll \mu$ iff $\nu \ll |\mu|$), but we shall have no need of this more general definition.

- Theorem 1.9 (ϵ - δ Language of Absolute Continuity of Measures) Let ν is a finite signed measure and μ is a positive measure on (X, \mathcal{B}) . Then $\nu \ll \mu$ if and only if for every $\epsilon > 0$, there exists a $\delta > 0$ such that $|\nu(E)| < \epsilon$, whenever $\mu(E) < \delta$.
- Remark If μ is a measure and f is extended μ -integrable, then the signed measure ν defined via $\nu(E) = \int_E f d\mu$ is absolutely continuous w.r.t. μ ; it is finite if and only if f is absolutely integrable. For any complex-valued $f \in L^1(\mu)$, the preceding theorem can be applied to $\Re(f)$ and $\Im(f)$.
- Corollary 1.10 If $f \in L^1(X, \mu)$, for every $\epsilon > 0$, there exists a $\delta > 0$, such that $\left| \int_E f d\mu \right| < \epsilon$ whenever $\mu(E) < \delta$.
- **Definition** For a signed measure ν defined via $\nu(E) = \int_E f d\mu$ for all $E \in \mathcal{B}$, we use the notation to express the relationship

$$d\nu = f d\mu$$
.

Sometimes, by a slight abuse of language, we shall refer to "the signed measure $f d\mu$ "

- Lemma 1.11 [Folland, 2013] Suppose that ν and μ are finite measures on (X, \mathcal{B}) . Either $\nu \perp \mu$, or there exists $\epsilon > 0$ and $E \in \mathcal{B}$ such that $\mu(E) > 0$ and $\nu \geq \epsilon \mu$ on E, i.e. E is a **positive set for** $\nu - \epsilon \mu$.
- Theorem 1.12 (Lebesgue-Radon-Nikodym Theorem)[Folland, 2013] Let ν be a σ -finite signed measure and μ be a σ -finite positive measure on (X, \mathcal{B}) . There exists unique σ -finite signed measure λ , ρ on (X, \mathcal{B}) such that

$$\lambda \perp \mu$$
, and $\rho \ll \mu$, and $\nu = \lambda + \rho$.

In particular, if $\nu \ll \mu$, then

$$d\nu = f d\mu$$
, for some f .

- **Definition** The decomposition $\nu = \rho + \lambda$, where $\lambda \perp \mu$ and $\rho \ll \mu$, is called the <u>Lebesgue</u> <u>decomposition</u> of ν with respect to μ .
- **Definition** If $\nu \ll \mu$, then according to the Lebesgue-Radon-Nikodym theorem, $d\nu = f d\mu$ for some f, where f is called the **Radon-Nikodym derivative** of ν w.r.t. μ and is denoted as

$$f := \frac{d\nu}{d\mu} \quad \Rightarrow \quad d\nu = \frac{d\nu}{d\mu}d\mu.$$

• Remark By Lebesgue decomposition, a signed measure ν can be represented as

$$d\nu = d\lambda + fd\mu$$

- Remark (Jordan Decomposition vs. Lebesgue Decomposition)
 We see two unique decompositions: the Jordan decomposition and the Lebesgue decomposition. We can make a comparison:
 - 1. Both of these two are decompositions of a signed measure ν .

- 2. Both of these two decompositions separate ν into two *mutually signular* sub-measures of ν .
- 3. Both of these two decompositions are *unique*

On the other hand,

- 1. The Jordan decomposition is to split a signed measure ν itself into two positive measures, i.e. ν_+ and ν_- that are mutually singular $(\nu_+ \perp \nu_-)$.
- 2. The Lebesgue decomposition is to split a signed measure ν with respect to a postive measure μ . The result is two-fold: 1) two mutually singular sub-measures $\lambda \perp \rho$ 2) their relationship with μ is opposite: $\lambda \perp \mu$, i.e. their support do not overlap; $\rho \ll \mu$, i.e. its support lies within support of μ .
- 3. Note that λ, ρ from the Lebesgue decomposition is **not** necessarily **positive**. But both ν and μ need to be σ -finite which is not required for the Jordan decomposition.
- Proposition 1.13 [Folland, 2013]

Suppose ν is σ -finite signed measure and λ , μ are σ -finite measure on (X, \mathcal{B}) such that $\nu \ll \mu$ and $\mu \ll \lambda$.

1. If $g \in L^1(X, \nu)$, then $g\left(\frac{d\nu}{d\mu}\right) \in L^1(X, \mu)$ and

$$\int g d\nu = \int g \, \frac{d\nu}{d\mu} \, d\mu$$

2. We have $\nu \ll \lambda$, and

$$\frac{d\nu}{d\lambda} = \frac{d\nu}{d\mu} \frac{d\mu}{d\lambda}, \quad \lambda \text{-a.e.}$$

- Corollary 1.14 If $\mu \ll \lambda$ and $\lambda \ll \mu$, then $(d\lambda/d\mu)(d\mu/d\lambda) = 1$ a.e. (with respect to either λ or μ).
- Proposition 1.15 If μ_1, \ldots, μ_n are measures on (X, \mathcal{B}) , then there exists a measure μ such that $\mu_i \ll \mu$ for all $i = 1, \ldots, n$, namely, $\mu = \sum_{i=1}^n \mu_i$.
- Exercise 1.16 (Conditional Expectation)

Let (X, \mathcal{B}, μ) be a **finite measure space**, \mathscr{F} is a sub- σ -algebra of \mathscr{B} , and $\nu = \mu|_{\mathscr{F}}$. Show that if $f \in L^1(X, \mu)$, there exists $g \in L^1(X, \nu)$ (thus g is \mathscr{F} -measureable) such that $\int_E f d\mu = \int_E g d\nu$ for all $E \in \mathscr{F}$; if g' is another such function then $g = g' \nu$ -a.e.

In probability theory, where $(X, \mathcal{B}) \equiv (\Omega, \mathcal{A})$, $f \equiv X$ is a random variable, then $g \equiv \mathbb{E}[X|\mathcal{F}]$ is called **the conditional expectation of** X **on** \mathcal{F} , which is \mathcal{F} -measure random variable.

2 Differentiation

• Remark In these notes we explore the question of the extent to which these theorems continue to hold when the differentiability or integrability conditions on the various functions F, F', f are relaxed. Among the results proven in these notes are

- 1. The Lebesgue differentiation theorem, which roughly speaking asserts that the Fundamental Theorem of Calculus continues to hold for almost every x if f is merely absolutely integrable, rather than continuous;
- 2. A number of differentiation theorems, which assert for instance that monotone, Lipschitz, or bounded variation functions in one dimension are almost everywhere differentiable; and
- 3. The Second Fundamental Theorem of Calculus for absolutely continuous functions.

2.1 The Lebesgue Differentiation Theorem in One Dimension

- Theorem 2.1 (Lebesgue differentiation theorem, one-dimensional case). Let $f : \mathbb{R} \to \mathbb{C}$ be an absolutely integrable function, and let $F : \mathbb{R} \to \mathbb{C}$ be the definite integral $F(x) := \int_{[-\infty,x]} f(t)dt$. Then F is continuous and almost everywhere differentiable, and F'(x) = f(x) for almost every $x \in \mathbb{R}$.
- Theorem 2.2 (Lebesgue differentiation theorem, second formulation). Let $f : \mathbb{R} \to \mathbb{C}$ be an absolutely integrable function. Then

$$\lim_{h \to 0+} \frac{1}{h} \int_{[x,x+h]} f(t)dt = f(x)$$
 (3)

for almost every $x \in \mathbb{R}$, and

$$\lim_{h \to 0+} \frac{1}{h} \int_{[x-h,x]} f(t)dt = f(x)$$
 (4)

for almost every $x \in \mathbb{R}$.

• Remark (*Density Argument*) [Tao, 2011]

The conclusion (3) we want to prove is a **convergence theorem** - an assertion that for all functions f in a given class (in this case, the class of absolutely integrable functions $f: \mathbb{R} \to \mathbb{R}$), a certain sequence of linear expressions $T_h f$ (in this case, the right averages $T_h f(x) = \frac{1}{h} \int_{[x,x+h]} f(t)dt$) converge in some sense (in this case, pointwise almost everywhere) to a specified limit (in this case, f).

There is a general and very useful argument to prove such convergence theorems, known as **the density argument**. This argument requires **two ingredients**, which we state informally as follows:

- 1. A *verification* of the convergence result for some "*dense subclass*" of "*nice*" functions f, such as *continuous functions*, *smooth functions*, *simple functions*, etc.. By "*dense*", we mean that a *general function* f in the *original class* can be *approximated to arbitrary accuracy* in a suitable sense by a function *in the nice subclass*.
- 2. A quantitative estimate that upper bounds the maximal fluctuation of the linear expressions $T_h f$ in terms of the "size" of the function f (where the precise definition of "size" depends on the nature of the approximation in the first ingredient).

Once one has these two ingredients, it is usually not too hard to put them together to obtain the desired convergence theorem for general functions f (not just those in the dense subclass).

• Remark One drawback with *the density argument* is it gives convergence results which are *qualitative* rather than *quantitative* - there is no explicit bound on the rate of convergence.

2.2 The Lebesgue Differentiation Theorem in \mathbb{R}^d

2.2.1 Absolute Integrable Version

• Theorem 2.3 (Lebesgue Differentiation Theorem (Absolute Integrable version))

[Tao, 2011]

Suppose $f: \mathbb{R}^d \to \mathbb{C}$ is **absolutly integrable**. Then for almost every x, we have

$$\lim_{r \to 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(z) - f(x)| dz = 0$$
and
$$\lim_{r \to 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} f(z) dz = f(x),$$
(5)

where $B(x,r) := \{ y \in \mathbb{R}^d : ||x-y|| < r \}$ is the open ball of radius r centred at x.

- **Definition** A point x for which (5) holds is called **a** Lebesgue point of f; thus, for an **absolutely integrable function** f, almost every point in \mathbb{R}^d will be a Lebesgue point for \mathbb{R}^d .
- The *quantitative estimate* we will need is the Hardy-Littlewood maximal inequality. First, we need to introduce the Hardy-Littlewood maximal function:

Definition [Folland, 2013]

If $f \in L^1_{loc}(\mathbb{R}^d)$, the **Hardy-Littlewood maximal function** Hf(x) is defined as

$$Hf(x) \equiv \sup_{r>0} \frac{1}{m(B(r,x))} \int_{B(r,x)} |f(z)| dz$$

where $B(r,x) = \{y : ||y-x|| < r\}$, and the **average value** of f on B(r,x) is

$$A_r f(x) = \frac{1}{m(B(r,x))} \int_{B(r,x)} f(z) dz.$$

• Remark A useful variant of Hf(x) (see [Stein and Shakarchi, 2009]) as

$$H^*f(x) \equiv \sup \left\{ \frac{1}{m(B)} \int_B |f(z)| dz, B \text{ is a ball } , x \in B \right\}.$$

- Remark The Hardy-Littlewood maximal function is an important function in the field of (real-variable) harmonic analysis.
- Remark The Hardy-Littlewood maximal function has the following properties:
 - 1. $(Hf)^{-1}(a,\infty) = \bigcup_{r>0} (A_r f)^{-1}(a,\infty)$ is open for any $a \in \mathbb{R}$, so the Hardy-Littlewood maximal function is *measureable*.
 - 2. Moreover, $Hf(x) < \infty$, a.e.x is **essentially bounded**.
 - 3. Note that $Hf \leq H^*f \leq 2^d Hf$

• We need to prove the following theorem for Lebesque differentiation theorem:

Theorem 2.4 (The Hardy-Littlewood Maximal Theorem) [Stein and Shakarchi, 2009, Folland, 2013]

Suppose f is integrable, then

1.

$$H^*f(x) \equiv \sup \left\{ \frac{1}{m(B)} \int_B |f(z)| \, dz, \ B \text{ is a ball } , x \in B \right\}.$$

is measurable.

- 2. $H^*f(x) < \infty$ for a.e. x.
- 3. H^*f satisfies the Hardy-Littlewood maximal inequality:

$$m(\{x: H^*f(x) > \alpha\}) \le \frac{A}{\alpha} \|f\|_{L^1(\mathbb{R}^d)}$$

for $\alpha > 0$, where $A = 3^d$, and $||f||_{L^1(\mathbb{R}^d)} = \int_{\mathbb{R}^d} |f(x)| dx$.

Note that $H^*f \ge |f|$, a.e.x, but the above expression indicates that H^*f is not much larger than |f|. However, we may not be able to assume H^*f integrable for any f.

- **Remark** In order to show the Hardy-Littlewood maximal inequality, we need to first find a dense covering called Vitali covering:
 - Definition (Vitali Covering) [Royden and Fitzpatrick, 1988, Stein and Shakarchi, 2009]

A collection \mathcal{B} of balls $\{B\}$ is said to be a <u>Vitali covering</u> of a set E, (covers E in **Vitali sense**,) if for every $x \in E$, any $\eta > 0$, there is a ball $B \in \mathcal{B}$, such that $x \in B$ and $m(B) < \eta$. Thus every point is covered by balls of arbitrary small measure.

- Lemma 2.5 (Lebesgue number lemma) For any open covering A of the metric space (X,d). If X is compact, there exists a number $\delta > 0$ such that for any subset of X having diameter $< \delta$, there exists an element of A containing it.
- Lemma 2.6 (Vitali Covering Lemma in elementary form) [Stein and Shakarchi, 2009]

Suppose $\mathcal{B} \equiv \{B_1, \dots, B_N\}$ is a finite collection of open balls in \mathbb{R}^d . Then there exists a disjoint sub-collection $B_{i_1}, B_{i_2}, \dots, B_{i_k}$ of \mathcal{B} that satisfies

$$m\left(\bigcup_{s=1}^{N} B_s\right) \le 3^d \sum_{j=1}^{k} m(B_{i_j})$$

Loosely speaking, we may always find a disjoint sub-collection of balls that covers a fraction of the region covered by the original collection of balls.

Lemma 2.7 (Vitali Covering Lemma in general) [Stein and Shakarchi, 2009, Folland, 2013]

Suppose E is a set of finite measure and B is a Vitali covering of E. For any $\delta > 0$, we

can find finitely many balls B_1, \ldots, B_N in \mathcal{B} that are disjoint and so that

$$\sum_{i=1}^{N} m(B_i) \ge m(E) - \delta$$

- Corollary 2.8 [Stein and Shakarchi, 2009, Royden and Fitzpatrick, 1988] Follwing the setting above, we can arrange the choice of balls so that

$$m\left(E - \bigcup_{i=1}^{N} B_i\right) < 2\delta$$

2.2.2 Local Integrable Version

- **Definition** [Stein and Shakarchi, 2009] A measurable function f on \mathbb{R}^d is **locally integrable**, i.e. $f \in L^1_{loc}(\mathbb{R}^d)$, if for every ball B the function $f(x)\mathbb{1}_B$ is integrable.
- This theorem follows from the Hardy-Littlewood maximal inequality

Theorem 2.9 [Stein and Shakarchi, 2009] If $f \in L^1_{loc}(\mathbb{R}^d)$ is **locally integrable**, then for the **average** of f, i.e.

$$A_r f(x) = \frac{1}{m(B(r,x))} \int_{B(r,x)} f(z) dz,$$

we have

$$A_r f(x) \stackrel{a.e.}{\to} f(x), \quad r \to 0.$$

• **Definition** [Stein and Shakarchi, 2009] If $f \in L^1_{loc}(\mathbb{R}^d)$, the **Lebesgue set** of f consists of all points $\overline{x} \in \mathbb{R}^d$ for which $f(\overline{x})$ is **finite** and

$$\lim_{\substack{m(B)\to 0\\ \overline{x}\in P}}\frac{1}{m\left(B\right)}\int_{B}\left|f(z)-f(\overline{x})\right|dz=0.$$

or equivalently, [Folland, 2013],

$$Lf \equiv \left\{ x \in \mathbb{R}^d : \lim_{r \to 0} \frac{1}{m(B(r,x))} \int_{B(r,x)} |f(z) - f(x)| \, dz = 0 \right\}.$$

 \bullet These results are from the Hardy-Littlewood maximal inequality

Corollary 2.10 Suppose E is a measureable set in \mathbb{R}^d . Then

- 1. Almost every $x \in E$ is a point of Lebesgue density of E;
- 2. Almost every $x \notin E$ is **not a point of Lebesgue density** of E.

- Corollary 2.11 If f is locally integrable on \mathbb{R}^d , then almost every point belongs to the Lebesgue set of f.
- Definition A collection of sets $\{U_{\alpha}\}$ is said to **shrink regularly** to \overline{x} or has **bounded eccentricity** at \overline{x} if there is a constant c > 0 such that for each U_{α} there is a ball B with

$$\overline{x} \in B, \qquad U_{\alpha} \subset B, \qquad m(U_{\alpha}) \ge c \, m(B).$$

• Theorem 2.12 (Lebesgue Differentiation Theorem (Local Integrable version)) [Stein and Shakarchi, 2009, Folland, 2013]

Suppose f is **locally integrable** on \mathbb{R}^d . For every x in the Lebesgue set of f, i.e. for almost every x, we have

$$\lim_{\substack{m(U_{\alpha})\to 0\\x\in U_{\alpha}}}\frac{1}{m(U_{\alpha})}\int_{U_{\alpha}}|f(z)-f(x)|\,dz=0$$
and
$$\lim_{\substack{m(U_{\alpha})\to 0\\x\in U_{\alpha}}}\frac{1}{m(U_{\alpha})}\int_{U_{\alpha}}f(z)dz=f(x),$$

for every family $\{U_{\alpha}\}$ that shrinks regularly to x.

2.3 Lebesgue Density and Radon-Nikodym Derivative

• Now we turn to consequences of the Lebesgue differentiation theorem.

Definition [Stein and Shakarchi, 2009]

If E is a measureable set in \mathbb{R}^d , $x \in \mathbb{R}^d$ is a **point of Lebesgue density** of E if

$$\lim_{\substack{m(B)\to 0\\x\in B}}\frac{m(B\cap E)}{m(B)}=1.$$

Loosely speaking, it says that a small ball contains x are almost entirely covered by E. Then for any $\alpha < 1$ close to 1, and every ball of sufficiently small radius containing x, we have

$$m(E \cap B) \ge \alpha m(B)$$
.

- **Definition** A Borel measure ν on \mathbb{R}^d will be called **regular** if
 - 1. $\nu(K) < \infty$ for every **compact** K;
 - 2. $\nu(E) = \inf \{ \nu(U) : U \text{ open}, E \subseteq U \} \text{ for every } E \in \mathcal{B}[\mathbb{R}^d].$

(Condition (2) is actually implied by condition (1). A signed or complex Borel measure ν will be called regular if $|\nu|$ is regular.

• Theorem 2.13 (Lebesgue Density from Radon-Nikodym derivative) [Folland, 2013] Let ν be a regular signed measure on \mathbb{R}^d , and let $d\nu = d\lambda + fdm$ be its Lebesgue-Radon-Nikodym decomposition, where $\lambda \perp m$. Then for m-almost every $x \in \mathbb{R}^d$,

$$\lim_{r \to 0} \frac{\nu(E_r)}{m(E_r)} = f(x),$$

where E_r shrinks regularly to x.

3 The Fundamental Theorem of Calculus for Lebesgue Integral

3.1 Functions of Bounded Variations

- Theorem 3.1 (Monotone Differentiation Theorem). [Tao, 2011] Any function $F : \mathbb{R} \to \mathbb{R}$ which is monotone (either monotone non-decreasing or monotone non-increasing) is differentiable almost everywhere.
- **Definition** (*Jump function*). [Tao, 2011] A basic jump function J is a function of the form

$$J(x) := \begin{cases} 0 & \text{when } x < x_0 \\ \theta & \text{when } x = x_0 \\ 1 & \text{when } x > x_0 \end{cases}$$

for some real numbers $x_0 \in \mathbb{R}$ and $0 \le \theta \le 1$; we call x_0 the point of discontinuity for J and θ the fraction. Observe that such functions are monotone non-decreasing, but have a discontinuity at one point.

A jump function is any absolutely convergent combination of basic jump functions, i.e. a function of the form $F = \sum_{n} c_n J_n$, where n ranges over an at most countable set, each J_n is a basic jump function, and the c_n are positive reals with $\sum_{n} c_n < \infty$. If there are only finitely many n involved, we say that F is a piecewise constant jump function.

Example If $q_1, q_2, q_3, ...$ is any enumeration of the *rationals*, then $\sum_{n=1}^{\infty} 2^{-n} \mathbb{1}_{[q_n, +\infty)}$ is a jump function.

• Remark All jump functions are monotone non-decreasing.

From the absolute convergence of the c_n we see that **every jump function is the uniform limit of piecewise constant jump functions**, for instance $\sum_{n=1}^{\infty} c_n J_n$ is the uniform limit of $\sum_{n=1}^{N} c_n J_n$. One consequence of this is that the points of discontinuity of a jump function $\sum_{n=1}^{\infty} c_n J_n$ are precisely those of the individual summands $c_n J_n$, i.e. of the points x_n where each J_n jumps.

• The key fact is that these Jump functions, together with the continuous monotone functions, essentially generate all monotone functions, at least in the bounded case:

Lemma 3.2 (Continuous-singular decomposition for monotone functions). Let $F : \mathbb{R} \to \mathbb{R}$ be a monotone non-decreasing function.

- 1. The only discontinuities of F are jump discontinuities. More precisely, if x is a point where F is discontinuous, then the limits $\lim_{y\to x^-} F(y)$ and $\lim_{y\to x^+} F(y)$ both exist, but are unequal, with $\lim_{y\to x^-} F(y) < \lim_{y\to x^+} F(y)$.
- 2. There are at most countably many discontinuities of F.
- 3. If F is bounded, then F can be expressed as the sum of a continuous monotone non-decreasing function F_c and a jump function F_{pp} .
- Exercise 3.3 Show that the decomposition of a bounded monotone non-decreasing function F into continuous F_c and jump components F_{pp} given by the above lemma is unique.
- Remark As positive measures on \mathbb{R} are related to increasing functions, complex measures

on \mathbb{R} are related to so-called functions of bounded variation.

- Remark Just as the integration theory of unsigned functions can be used to develop the integration theory of the absolutely convergent functions, the differentiation theory of monotone functions can be used to develop a parallel differentiation theory for the class of functions of bounded variation:
- **Definition** (Bounded variation). Let $F: \mathbb{R} \to \mathbb{R}$ be a function. <u>The total variation</u> $||F||_{TV(\mathbb{R})}$ (or $||F||_{TV}$ for short) of F is defined to be the **supremum**

$$||F||_{TV(\mathbb{R})} := \sup_{x_0 < \dots < x_n} \sum_{i=1}^n |F(x_i) - F(x_{i-1})|$$

where the supremum ranges over all *finite increasing sequences* x_0, \ldots, x_n of real numbers with $n \geq 0$; this is a quantity in $[0, +\infty]$. We say that F has bounded variation (on \mathbb{R}) if $||F||_{TV(\mathbb{R})}$ is *finite*. (In this case, $||F||_{TV(\mathbb{R})}$ is often written as $||F||_{BV(\mathbb{R})}$ or just $||F||_{BV}$.)

• Remark Given any interval [a, b], we define the total variation $||F||_{TV([a,b])}$ of F on [a, b] as

$$||F||_{TV(\mathbb{R})} := \sup_{a \le x_0 < \dots < x_n \le b} \sum_{i=1}^n |F(x_i) - F(x_{i-1})|$$

We say that a function F has **bounded variation on** [a,b] if $||F||_{BV([a,b])}$ is finite. Note that $||F||_{TV(\mathbb{R})} = \lim_{N \to \infty} ||F||_{TV([-N,N])}$.

- Proposition 3.4 If $F : \mathbb{R} \to \mathbb{R}$ is a monotone function, $||F||_{TV([a,b])} = |F(b) F(a)|$ for any interval [a,b]. Thus F has bounded variation on \mathbb{R} if and only if it is bounded.
- Proposition 3.5 For any functions $F, G : \mathbb{R} \to \mathbb{R}$, the total variation $\|\cdot\|_{TV(\mathbb{R})}$ satisfies the following property:
 - 1. (Non-Negativity): $||F||_{TV(\mathbb{R})} \geq 0$;
 - 2. (Positive Definiteness): $||F||_{TV(\mathbb{R})} = 0$ if and only if F is constant.
 - 3. (Homogeneity): $||cF||_{TV(\mathbb{R})} = |c| ||F||_{TV(\mathbb{R})}$ for any $c \in \mathbb{R}$.
 - 4. (Triangle Inequality): $||F + G||_{TV(\mathbb{R})} \le ||F||_{TV(\mathbb{R})} + ||G||_{TV(\mathbb{R})}$

Thus $\|\cdot\|_{TV(\mathbb{R})}$ is a **norm**.

- Exercise 3.6 (Bounded Variation is Stronger than Bounded)
 - 1. Show that every function $f : \mathbb{R} \to \mathbb{R}$ of **bounded variation** is **bounded**, and that the $\lim_{x \to +\infty} f(x)$ and $\lim_{x \to -\infty} f(x)$, are well-defined.
 - 2. Give a counterexample of a **bounded**, **continuous**, **compactly supported** function f that is **not** of **bounded variation**.
- Proposition 3.7 A function $F : \mathbb{R} \to \mathbb{R}$ is of bounded variation if and only if it is the difference of two bounded monotone functions.
- Remark Much as an absolutely integrable function can be expressed as the difference of its positive and negative parts, a bounded variation function can be expressed as the

difference of two bounded monotone functions. Let

$$F^{+}(x) = \sup_{x_{0} < \dots < x_{n} \le x} \sum_{i=1}^{n} \max\{F(x_{i}) - F(x_{i-1}), 0\}$$
$$F^{-}(x) = \sup_{x_{0} < \dots < x_{n} \le x} \sum_{i=1}^{n} \max\{-F(x_{i}) + F(x_{i-1}), 0\}$$

We have

$$F(x) = F(-\infty) + F^{+}(x) - F^{-}(x)$$
$$||F||_{TV([a,b])} = F^{+}(b) - F^{+}(a) + F^{-}(b) - F^{-}(a)$$
$$||F||_{TV(\mathbb{R})} = F^{+}(+\infty) + F^{-}(+\infty)$$

for every interval [a, b], where $F(-\infty) := \lim_{x \to -\infty} F(x)$, $F^+(+\infty) := \lim_{x \to +\infty} F^+(x)$, and $F^-(+\infty) := \lim_{x \to +\infty} F^-(x)$.

- Corollary 3.8 (Bounded Variation Differentiation Theorem).

 Every bounded variation function is differentiable almost everywhere.
- Definition (Locally Bounded Variation)
 A function is <u>locally of bounded variation</u> if it is of bounded variation on every compact interval [a, b].

Corollary 3.9 (Locally Bounded Variation Differentiation Theorem). Every locally bounded variation function is differentiable almost everywhere.

• Definition (Lipschitz Continuous Function) A function $f: \mathbb{R} \to \mathbb{R}$ is said to be <u>Lipschitz continuous</u> if there exists a constant C > 0 such that

$$|f(x) - f(y)| \le C |x - y|$$

for all $x, y \in \mathbb{R}$; the smallest C with this property is known as **the Lipschitz constant** of f.

Corollary 3.10 (Lipschitz Differentiation Theorem, one-dimensional case).

Every Lipschitz continuous function F is **locally** of **bounded variation**, and hence **differentiable almost everywhere**. Furthermore, the **derivative** F', when it exists, is **bounded** in magnitude by the Lipschitz constant of F.

Remark The same result is true in *higher dimensions*, and is known as *the Rademacher differentiation theorem*.

• Definition (Convex Function)

A function $f : \mathbb{R} \to \mathbb{R}$ is said to be **convex** if one has $f((1-t)x + ty) \le (1-t)f(x) + tf(y)$ for all x < y and 0 < t < 1.

Corollary 3.11 (Convex Differentiation Theorem, one-dimensional case)

If f is convex, then it is continuous and almost everywhere differentiable, and its derivative f' is equal almost everywhere to a monotone non-decreasing function, and so is itself almost everywhere differentiable.

(Hint: Drawing the graph of f, together with a number of chords and tangent lines, is likely to be very helpful in providing visual intuition.)

Remark Thus we see that in some sense, convex functions are "almost everywhere twice differentiable". Similar claims also hold for concave functions, of course.

- Remark From above, we see that the class of functions of locally bounded variations contains the following sub-classes:
 - 1. Bounded Monotone Functions
 - 2. Lipschitz Continuous Functions
 - 3. Convex (Concave) Function
 - 4. Absolute Continuous Function thus includes Uniformly Continuous Function too

3.2 The Second Fundamental Theorem of Calculus for Lebesgue Integral

Proposition 3.12 (Upper bound for second fundamental theorem).
 Let F: [a, b] → ℝ be monotone non-decreasing (so that, as discussed above, F' is defined almost everywhere, is unsigned, and is measurable). Then

$$\int_{[a,b]} F'(x)dx \le F(b) - F(a).$$

In particular, F' is absolutely integrable.

• For function of bounded variation, the derivative is also absolutely integrable

Proposition 3.13 Any function of bounded variation has an (almost everywhere defined) derivative that is absolutely integrable.

• For Lipschitz continuous function, we can directly prove the second fundamental theorem of calculus:

Theorem 3.14 (Second fundamental theorem for Lipschitz functions). Let $F : [a, b] \to \mathbb{R}$ be Lipschitz continuous.

$$\int_{[a,b]} F'(x)dx = F(b) - F(a).$$

(Hint: Argue as in the proof of Proposition above, but use *the dominated convergence theorem* in place of *Fatous lemma*)

• Remark One of the main *challenge* to show the second fundament theorem of calculus for *all monotone function* (i.e. to show the equality condition holds above) is that *all the variation* of F may be *concentrated in a set of measure zero*, and thus *undetectable* by the *Lebesgue integral* of F'. The following is one of example

Example The Heaviside function is defined as $F := \mathbb{1}\{[0, +\infty)\}$. It is clear that F' vanishes almost everywhere, but F(b)F(a) is not equal to $\int_{[a,b]} F'(x)dx$ if b and a lie on **opposite** sides of the discontinuity at 0.

• Moreover, we have

Proposition 3.15 If F is a jump function, then F' vanishes almost everywhere.

Thus the second fundamental theorem of calculus does not hold for any jump functions.

• Remark Even only consider the continuous monotone function, it is still possible for all the fluctuation to now be concentrated, not in a countable collection of jump discontinuities, but instead in an uncountable set of zero measure, such as the middle thirds Cantor set. This can be illustrated by the key counterexample of the Cantor function, also known as the Devil's staircase function.

This example shows that the classical derivative $F'(x) := \lim_{h\to 0} \frac{F(x+h)F(x)}{h}$ of a function has some defects; it cannot "see" some of the variation of a continuous monotone function such as the Cantor function.

- Remark In view of this counterexample, we see that we need to add an additional hypothesis to **the continuous monotone** non-increasing function F before we can recover the second fundamental theorem. One such hypothesis is **absolute continuity**.
- **Definition** A function $F : \mathbb{R} \to \mathbb{R}$ is **continuous** if, for every $\epsilon > 0$ and $x_0 \in \mathbb{R}$, there exists a $\delta > 0$ such that $|F(b) F(a)| \le \epsilon$ whenever (a, b) is an interval of length at most δ that contains x_0 .

Definition A function $F: \mathbb{R} \to \mathbb{R}$ is *uniformly continuous* if, for every $\epsilon > 0$, there exists a $\delta > 0$ such that $|F(b) - F(a)| \le \epsilon$ whenever (a, b) is an interval of length at most δ .

- Definition (Absolute Continuity) A function $F: \mathbb{R} \to \mathbb{R}$ is said to be **absolutely continuous** if, for every $\epsilon > 0$, there exists a $\delta > 0$ such that $\sum_{j=1}^{n} |F(b_j) - F(a_j)| \le \epsilon$ whenever $(a_1, b_1), \ldots, (a_n, b_n)$ is a **finite collection of disjoint intervals** of **total length** $\sum_{j=1}^{n} |b_j - a_j|$ **at most** δ .
- Proposition 3.16 The followings statements are true:
 - 1. Every absolutely continuous function is uniformly continuous and therefore continuous.
 - 2. Every absolutely continuous function is of bounded variation on every compact interval [a, b]. (Hint: first show this is true for any sufficiently small interval.) Thus, by the Local Bounded Variation Differentiation Theorem, absolutely continuous functions are differentiable almost everywhere.
 - 3. Every Lipschitz continuous function is absolutely continuous.
 - 4. The function $x \mapsto \sqrt{x}$ is absolutely continuous, but not Lipschitz continuous, on the interval [0,1].
 - 5. The Cantor function is continuous, monotone, and uniformly continuous, but not absolutely continuous, on [0,1].
 - 6. If $f: \mathbb{R} \to \mathbb{R}$ is **absolutely integrable**, then the indefinite integral $F(x) := \int_{[-\infty,x]} f(y) dy$ is **absolutely continuous**, and F is differentiable almost everywhere with F'(x) = f(x) for almost every x.
 - 7. The **sum** or **product** of two absolutely continuous functions on an interval [a, b] remains absolutely continuous.

• Remark We can draw the relative strength of different concepts on a compact interval [a, b].

- uniformly continuous \rightarrow absolutely continuous: See Cantor function example [Tao, 2011].
- absolutely continuous \neq Lipschitz continuous: $x \mapsto \sqrt{x}$
- Proposition 3.17 Absolutely continuous functions map null sets to null sets, i.e. if F:
 ℝ → ℝ is absolutely continuous and E is a null set then F(E) := {F(x) : x ∈ E} is also a null set.

Exercise 3.18 Show that the Cantor function does not have this property above.

• For absolutely continuous functions, we can recover the second fundamental theorem of calculus:

Theorem 3.19 (Second Fundamental Theorem for Absolutely Continuous Functions).

Let $F:[a,b] \to \mathbb{R}$ be absolutely continuous. Then

$$\int_{[a,b]} F'(x)dx = F(b) - F(a).$$

• Proposition 3.20 (Classification of Absolute Continuous Function)

A function $F: [a,b] \to \mathbb{R}$ is absolutely continuous if and only if it takes the form

$$F(x) = \int_{[a,x]} f(y)dy + C$$

for some **absolutely integrable** $f : [a, b] \to \mathbb{R}$ and a constant C.

- Remark We see that the *absolute continuity* was used primarily in *two ways*:
 - 1. firstly, to ensure the almost everywhere existence of F'
 - 2. to control an exceptional null set E.

It turns out that one can achieve the latter control by making a different hypothesis, namely that the function F is everywhere differentiable rather than merely almost everywhere differentiable. More precisely, we have

• Theorem 3.21 (Second Fundamental Theorem of Calculus, again). Let [a,b] be a compact interval of positive length, let $F:[a,b] \to \mathbb{R}$ be a differentiable function, such that F' is absolutely integrable. Then the Lebesgue integral

$$\int_{[a,b]} F'(x)dx = F(b) - F(a).$$

• Exercise 3.22 Let $F: [-1,1] \to \mathbb{R}$ be the function defined by setting

$$F(x) := x^2 \sin\left(\frac{1}{x^3}\right)$$

when x is non-zero, and F(0) := 0. Show that F is everywhere differentiable, but the deriative F' is not absolutely integrable, and so the second fundamental theorem of calculus does not apply in this case (at least if we interpret $\int_{[a,b]} F'(x) dx$ using the absolutely convergent Lebesgue integral).

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