

Lecture 0: Summary (part 2)

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Nov. 16th., 2022

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1 σ -Algebra, Measure and Integration on Product Space

1.1 Product σ -Algebra

- **Definition (*Product Topology*)**

Let X and Y be topological spaces. **The product topology** on $X \times Y$ is the topology having as basis the collection \mathcal{B} of all sets of the form $U \times V$, where U is an open subset of X and V is an open subset of Y .

Definition Let $\pi_X : X \times Y \rightarrow X$ be defined by the equation

$$\pi_X(x, y) = x;$$

$\pi_Y : X \times Y \rightarrow Y$ be defined by the equation

$$\pi_Y(x, y) = y.$$

The maps π_X and π_Y are called *the projections of $X \times Y$ onto its first and second factors*, respectively.

- **Definition (*Product σ -Algebra*)**

Suppose that (X, \mathcal{B}_X) and (Y, \mathcal{B}_Y) are measurable spaces. We can form the **pullback σ -algebras**

$$\begin{aligned}\pi_X^*(\mathcal{B}_X) &:= \{\pi_X^{-1}(E) : E \in \mathcal{B}_X\} = \{E \times Y : E \in \mathcal{B}_X\} \\ \pi_Y^*(\mathcal{B}_Y) &:= \{\pi_Y^{-1}(F) : F \in \mathcal{B}_Y\} = \{X \times F : F \in \mathcal{B}_Y\}\end{aligned}$$

We then define **the product σ -algebra $\mathcal{B}_X \times \mathcal{B}_Y$** to be the σ -algebra generated by the union of these two pull-back σ -algebras:

$$\mathcal{B}_X \times \mathcal{B}_Y := \langle \pi_X^*(\mathcal{B}_X) \cup \pi_Y^*(\mathcal{B}_Y) \rangle.$$

This definition has several equivalent formulations:

- **Proposition 1.1 (*Equivalent Definition of Product σ -Algebra*)** [Tao, 2011]

Let (X, \mathcal{B}_X) and (Y, \mathcal{B}_Y) be measurable spaces.

1. $\mathcal{B}_X \times \mathcal{B}_Y$ is the σ -algebra generated by the sets $E \times F$ with $E \in \mathcal{B}_X$, $F \in \mathcal{B}_Y$. In other words, $\mathcal{B}_X \times \mathcal{B}_Y$ is the **coarsest σ -algebra** on $X \times Y$ with the property that **the product** of a \mathcal{B}_X -measurable set and a \mathcal{B}_Y -measurable set is always $\mathcal{B}_X \times \mathcal{B}_Y$ measurable.
2. $\mathcal{B}_X \times \mathcal{B}_Y$ is the **coarsest σ -algebra** on $X \times Y$ that makes the **projection maps** π_X, π_Y both measurable.

- **Proposition 1.2 (*Property of Product σ -Algebra*)** [Tao, 2011]

Let (X, \mathcal{B}_X) and (Y, \mathcal{B}_Y) be measurable spaces.

1. (**Slices of Measurable Set**) If $E \in \mathcal{B}_X \times \mathcal{B}_Y$, then sets

$$E_x := \{y \in Y : (x, y) \in E\} \in \mathcal{B}_Y$$

for every $x \in X$, and similarly that the sets

$$E^y := \{x \in X : (x, y) \in E\} \in \mathcal{B}_X$$

for every $y \in Y$.

2. (**Slices of Measurable Function**) If $f : X \times Y \rightarrow [0, +\infty]$ is measurable (with respect to $\mathcal{B}_X \times \mathcal{B}_Y$), then the function

$$f_x : y \rightarrow f(x, y)$$

is \mathcal{B}_Y -measurable for every $x \in X$, and similarly that the function

$$f^y : x \rightarrow f(x, y)$$

is \mathcal{B}_X -measurable for every $y \in Y$.

3. The product of two **trivial** σ -algebras (on two different spaces X, Y) is again **trivial**.
4. The product of two **atomic** σ -algebras is again **atomic**.
5. The product of two **finite** σ -algebras is again **finite**.
6. The product of two **Borel** σ -algebras (on two Euclidean spaces $\mathbb{R}^d, \mathbb{R}^{d'}$ with $d, d' \geq 1$) is again the **Borel** σ -algebra (on $\mathbb{R}^d \times \mathbb{R}^{d'} \equiv \mathbb{R}^{d+d'}$).
7. The product of two **Lebesgue** σ -algebras (on two Euclidean spaces $\mathbb{R}^d, \mathbb{R}^{d'}$ with $d, d' \geq 1$) is **not** the **Lebesgue** σ -algebra. (Hint: argue by **contradiction** and use slices of measurable set as above proposition.)
8. However, the Lebesgue σ -algebra on $\mathbb{R}^{d+d'}$ is the **completion** of the product of the Lebesgue σ -algebras of \mathbb{R}^d and $\mathbb{R}^{d'}$ with respect to $(d+d')$ -dimensional Lebesgue measure.

• **Exercise 1.3** [Tao, 2011]

If $E \in \mathcal{B}_X \times \mathcal{B}_Y$, show that the slices $E_x := \{y \in Y : (x, y) \in E\}$ lie in a countably generated σ -algebra. In other words, show that there exists an at most countable collection $\mathcal{A} = \mathcal{A}_E$ of sets (which can depend on E) such that $\{E_x : x \in X\} \subseteq \langle \mathcal{A} \rangle$. Conclude in particular that the number of **distinct** slices E_x is at most c , the **cardinality** of the continuum.

• **Exercise 1.4** [Tao, 2011]

Give an example to show that the product of two **discrete** σ -algebras is **not** necessarily **discrete**.

On the other hand, show that the product of two **discrete** σ -algebras $2^X, 2^Y$ is again a **discrete** σ -algebra if at least one of the domains X, Y is **at most countably infinite**.

1.2 Product Measure

• **Definition** (σ -Finite).

A measure space (X, \mathcal{B}, μ) is σ -**finite** if X can be expressed as the **countable union** of sets of **finite** measure, i.e. $X = \bigcup_n X_n$, $\mu(X_n) < \infty$ for all n .

• **Example** (\mathbb{R}^d)

\mathbb{R}^d with **Lebesgue measure** is σ -finite, as \mathbb{R}^d can be expressed as the union of (for instance) the balls $B(0, n)$ for $n = 1, 2, 3, \dots$, each of which has finite measure.

On the other hand, \mathbb{R}^d with **counting measure** is **not** σ -finite.

• **Proposition 1.5** (**Existence and Uniqueness of Product Measure**) [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be σ -finite measure spaces. Then there exists a **unique**

measure $\mu_X \times \mu_Y : \mathcal{B}_X \times \mathcal{B}_Y \rightarrow [0, \infty]$ on product σ -algebra $\mathcal{B}_X \times \mathcal{B}_Y$ that obeys

$$\mu_X \times \mu_Y(E \times F) = \mu_X(E) \mu_Y(F)$$

whenever $E \in \mathcal{B}_X$ and $F \in \mathcal{B}_Y$.

- **Remark** When X, Y are **not both** σ -finite, then one can still construct **at least one product measure**, but it will, in general, **not be unique**.

- **Remark** (*Product Measure of Lebesgue Measures*)

The **product** $m^d \times m^{d'}$ of the **Lebesgue measures** $m^d, m^{d'}$ on $(\mathbb{R}^d, \mathcal{L}[\mathbb{R}^d])$ and $(\mathbb{R}^{d'}, \mathcal{L}[\mathbb{R}^{d'}])$ respectively will **agree** with Lebesgue measure $m^{d+d'}$ on the product space $\mathcal{L}[\mathbb{R}^d] \times \mathcal{L}[\mathbb{R}^{d'}]$, which is a **subalgebra** of $\mathcal{L}[\mathbb{R}^{d+d'}]$. After taking the **completion** $\overline{m^d \times m^{d'}}$ of this product measure, one obtains the **full Lebesgue measure** $m^{d+d'}$.

- **Proposition 1.6** [Tao, 2011]

Let (X, \mathcal{B}_X) and (Y, \mathcal{B}_Y) be measurable spaces.

1. The product of two **Dirac measures** on $(X, \mathcal{B}_X), (Y, \mathcal{B}_Y)$ is a **Dirac measure** on $(X \times Y, \mathcal{B}_X \times \mathcal{B}_Y)$.
2. If X, Y are **at most countable**, the product of the two **counting measures** on $(X, \mathcal{B}_X), (Y, \mathcal{B}_Y)$ is the **counting measure** on $(X \times Y, \mathcal{B}_X \times \mathcal{B}_Y)$.

- **Proposition 1.7** (*Associativity of Product*). [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X), (Y, \mathcal{B}_Y, \mu_Y), (Z, \mathcal{B}_Z, \mu_Z)$ be σ -**finite sets**. We may identify the Cartesian products $(X \times Y) \times Z$ and $X \times (Y \times Z)$ with each other in the obvious manner. If we do so, then

$$(\mathcal{B}_X \times \mathcal{B}_Y) \times \mathcal{B}_Z = \mathcal{B}_X \times (\mathcal{B}_Y \times \mathcal{B}_Z)$$

and

$$(\mu_X \times \mu_Y) \times \mu_Z = \mu_X \times (\mu_Y \times \mu_Z).$$

1.3 Tonelli's Theorem

- **Definition** (*Monotone Class*)

Define a **monotone class** in X is a collection \mathcal{B} of subsets of X with the following two **closure properties**:

1. If $E_1 \subset E_2 \subset \dots$ are a **countable increasing** sequence of sets in \mathcal{B} , then $\bigcup_{n=1}^{\infty} E_n \in \mathcal{B}$.
2. If $E_1 \supset E_2 \supset \dots$ are a **countable decreasing** sequence of sets in \mathcal{B} , then $\bigcap_{n=1}^{\infty} E_n \in \mathcal{B}$.

- **Lemma 1.8** (*Monotone Class Lemma*). [Tao, 2011]

Let \mathcal{A} be a **Boolean algebra** on X . Then $\langle \mathcal{A} \rangle$ is the **smallest monotone class** that contains \mathcal{A} .

- **Theorem 1.9** (*Tonelli's Theorem, Incomplete Version*). [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be σ -**finite measure spaces**, and let $f : X \times Y \rightarrow [0, +\infty]$ be measurable with respect to $\mathcal{B}_X \times \mathcal{B}_Y$. Then:

1. The functions

$$\begin{aligned} x &\rightarrow \int_Y f(x, y) d\mu_Y(y) \\ y &\rightarrow \int_X f(x, y) d\mu_X(x) \end{aligned}$$

(which are well-defined) are **measurable** with respect to \mathcal{B}_X and \mathcal{B}_Y respectively.

2. We have

$$\begin{aligned} \int_{X \times Y} f(x, y) d(\mu_X \times \mu_Y)(x, y) &= \int_X \left(\int_Y f(x, y) d\mu_Y(y) \right) d\mu_X(x) \\ &= \int_Y \left(\int_X f(x, y) d\mu_X(x) \right) d\mu_Y(y) \end{aligned} \quad (1)$$

• **Corollary 1.10 (Slice of Null Set)** [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be **σ -finite measure spaces**, and let $E \in \mathcal{B}_X \times \mathcal{B}_Y$ be a **null set** with respect to $\mu_X \times \mu_Y$. Then for μ_X -almost every $x \in X$, the set $E_x := \{y \in Y : (x, y) \in E\}$ is a μ_Y -**null set**; and similarly, for μ_Y -almost every $y \in Y$, the set $E^y := \{x \in X : (x, y) \in E\}$ is a μ_X -**null set**.

With this corollary, we can extend *Tonelli's theorem* to the completion $(X \times Y, \overline{\mathcal{B}_X \times \mathcal{B}_Y}, \overline{\mu_X \times \mu_Y})$ of the product space $(X \times Y, \mathcal{B}_X \times \mathcal{B}_Y, \mu_X \times \mu_Y)$.

• **Theorem 1.11 (Tonelli's Theorem, Complete Version)**. [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be **complete σ -finite measure spaces**, and let $f : X \times Y \rightarrow [0, +\infty]$ be measurable with respect to $\mathcal{B}_X \times \mathcal{B}_Y$. Then:

1. For μ_X -almost every $x \in X$, the function

$$y \rightarrow f(x, y)$$

is \mathcal{B}_Y -**measurable** and in particular, $\int_Y f(x, y) d\mu_Y(y)$ exists. Furthermore, the (μ_X -almost everywhere defined) map

$$x \rightarrow \int_Y f(x, y) d\mu_Y(y)$$

is \mathcal{B}_X -**measurable**.

2. For μ_Y -almost every $y \in Y$, the function

$$x \rightarrow f(x, y)$$

is \mathcal{B}_X -**measurable** and in particular, $\int_X f(x, y) d\mu_X(x)$ exists. Furthermore, the (μ_Y -almost everywhere defined) map

$$y \rightarrow \int_X f(x, y) d\mu_X(x)$$

is \mathcal{B}_Y -**measurable**.

3. We have

$$\begin{aligned} \int_{X \times Y} f(x, y) d(\overline{\mu_X \times \mu_Y})(x, y) &= \int_X \left(\int_Y f(x, y) d\mu_Y(y) \right) d\mu_X(x) \\ &= \int_Y \left(\int_X f(x, y) d\mu_X(x) \right) d\mu_Y(y) \end{aligned} \quad (2)$$

- Specialising to the case when f is an indicator function $f = \mathbb{1}_E$, we conclude

Corollary 1.12 (Tonellis Theorem for Sets). [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be **complete σ -finite measure spaces**, and let $E \in \overline{\mathcal{B}_X \times \mathcal{B}_Y}$. Then:

1. For μ_X -almost every $x \in X$, the set

$$E_x := \{y \in Y : (x, y) \in E\} \in \mathcal{B}_Y$$

and the (μ_X -almost everywhere defined) map

$$x \rightarrow \mu_Y(E_x)$$

is \mathcal{B}_X -measurable.

2. For μ_Y -almost every $y \in Y$, the set

$$E^y := \{x \in X : (x, y) \in E\} \in \mathcal{B}_X$$

and the (μ_Y -almost everywhere defined) map

$$y \rightarrow \mu_X(E^y)$$

is \mathcal{B}_Y -measurable.

3. We have

$$\begin{aligned} \overline{\mu_X \times \mu_Y}(E) &= \int_X \mu_Y(E_x) d\mu_X(x) \\ &= \int_Y \mu_X(E^y) d\mu_Y(y) \end{aligned} \quad (3)$$

- **Remark** Tonellis theorem can **fail** if **the σ -finite hypothesis is removed**, and also that **product measure need not be unique**.

1.4 Fubini's Theorem

- **Remark** Tonelli's theorem is for the **unsigned integral**, but it leads to an important analogue for the **absolutely integral**, known as *Fubini's theorem*:
- **Theorem 1.13 (Fubinis Theorem).** [Tao, 2011]
Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be **complete σ -finite measure spaces**, and let $f : X \times Y \rightarrow \mathbb{C}$ be **absolutely integrable with respect to $\overline{\mathcal{B}_X \times \mathcal{B}_Y}$** . Then:

1. For μ_X -almost every $x \in X$, the function

$$y \rightarrow f(x, y)$$

is **absolutely integrable with respect to μ_Y** and in particular, $\int_Y f(x, y) d\mu_Y(y)$ exists. Furthermore, the (μ_X -almost everywhere defined) map

$$x \rightarrow \int_Y f(x, y) d\mu_Y(y)$$

is **absolutely integrable with respect to μ_X** .

2. For μ_Y -almost every $y \in Y$, the function

$$x \rightarrow f(x, y)$$

is **absolutely integrable with respect to μ_X** and in particular, $\int_X f(x, y) d\mu_X(x)$ exists. Furthermore, the (μ_Y -almost everywhere defined) map

$$y \rightarrow \int_X f(x, y) d\mu_X(x)$$

is **absolutely integrable with respect to μ_Y** .

3. We have

$$\begin{aligned} \int_{X \times Y} f(x, y) d(\overline{\mu_X \times \mu_Y})(x, y) &= \int_X \left(\int_Y f(x, y) d\mu_Y(y) \right) d\mu_X(x) \\ &= \int_Y \left(\int_X f(x, y) d\mu_X(x) \right) d\mu_Y(y) \end{aligned} \quad (4)$$

- **Remark** Fubini's theorem fails when one drops the hypothesis that f is **absolutely integrable with respect to the product space**.
- **Remark** Despite the failure of Tonelli's theorem in the σ -finite setting, it is possible to (carefully) extend Fubini's theorem to **the non- σ -finite setting**, as the **absolute integrability hypotheses**, when combined with *Markov's inequality*, can provide a substitute for the σ -finite property.
- **Remark** Informally, *Fubini's theorem* allows one to **always interchange the order of two integrals**, as long as the **integrand is absolutely integrable in the product space** (or its completion). In particular, specialising to Lebesgue measure, we have

$$\int_{\mathbb{R}^{d+d'}} f(x, y) d(x, y) = \int_{\mathbb{R}^d} \left(\int_{\mathbb{R}^{d'}} f(x, y) dy \right) dx = \int_{\mathbb{R}^{d'}} \left(\int_{\mathbb{R}^d} f(x, y) dx \right) dy$$

whenever $f : \mathbb{R}^{d+d'} \rightarrow \mathbb{C}$ is **absolutely integrable**. In view of this, we often write $dx dy$ (or $dy dx$) for $d(x, y)$.

- By combining *Fubini's theorem* with *Tonelli's theorem*, we can recast the **absolute integrability hypothesis**:

Corollary 1.14 (Fubini-Tonelli Theorem). [Tao, 2011]

Let $(X, \mathcal{B}_X, \mu_X)$ and $(Y, \mathcal{B}_Y, \mu_Y)$ be **complete σ -finite measure spaces**, and let $f : X \times Y \rightarrow \mathbb{C}$ be **measurable with respect to $\overline{\mathcal{B}_X \times \mathcal{B}_Y}$** . If

$$\int_X \left(\int_Y |f(x, y)| d\mu_Y(y) \right) d\mu_X(x) < \infty$$

then f is **absolutely integrable** with respect to $\overline{\mathcal{B}_X \times \mathcal{B}_Y}$, and in particular the conclusions of **Fubini's theorem** hold.

Similarly if we use $\int_Y \left(\int_X |f(x, y)| d\mu_X(x) \right) d\mu_Y(y)$ instead of $\int_X \left(\int_Y |f(x, y)| d\mu_Y(y) \right) d\mu_X(x)$.

- **Proposition 1.15 (Area Interpretation of Integral).** [Tao, 2011]

Let (X, \mathcal{B}, μ) be a σ -finite measure space, and let \mathbb{R} be equipped with Lebesgue measure m and the Borel σ -algebra $\mathcal{B}[\mathbb{R}]$. Then $f : X \rightarrow [0, +\infty]$ is **measurable if and only if its subgraph**

$$\{(x, t) \in X \times \mathbb{R} : 0 \leq t \leq f(x)\}$$

is **measurable** in $\mathcal{B} \times \mathcal{B}[\mathbb{R}]$, in which case we have

$$(\mu \times m) \{(x, t) \in X \times \mathbb{R} : 0 \leq t \leq f(x)\} = \int_X f(x) d\mu(x).$$

Similarly if we replace $\{(x, t) \in X \times \mathbb{R} : 0 \leq t \leq f(x)\}$ by $\{(x, t) \in X \times \mathbb{R} : 0 \leq t < f(x)\}$.

- **Proposition 1.16 (Distribution Formula).** [Tao, 2011]

Let (X, \mathcal{B}, μ) be a σ -finite measure space, and let $f : X \rightarrow [0, +\infty]$ be **measurable**. Then

$$\int_X f(x) d\mu(x) = \int_{[0, \infty]} \mu \{x \in X : f(x) \geq \lambda\} d\lambda \quad (5)$$

(Note that the integrand on the right-hand side is monotone and thus Lebesgue measurable.)

Similarly if we replace $\{x \in X : f(x) \geq \lambda\}$ by $\{x \in X : f(x) > \lambda\}$.

2 Signed Measures and Radon-Nikodym Derivative

2.1 Signed Measure

- **Definition (Signed Measure)**

Let (X, \mathcal{B}) be a measure space. A **signed measure** on (X, \mathcal{B}) is a function $\nu : \mathcal{B} \rightarrow [-\infty, +\infty]$ 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 \mathcal{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 \mathcal{B} , and $f : X \rightarrow [-\infty, +\infty]$ is *extended μ -integrable* with at least one of f_+ and f_- finite integrable. Then $\nu(A) = \int_X f \mathbf{1}_{\{A\}} d\mu$ is a signed measure.
- Like unsigned measure, we have monotone downward and upward convergence:

Proposition 2.1 *Let ν be a **signed measure** on (X, \mathcal{B}) .*

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

$$\nu \left(\bigcup_{n=1}^{\infty} E_n \right) = \lim_{n \rightarrow \infty} \nu(E_n) = \sup_n \nu(E_n). \quad (6)$$

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

$$\nu \left(\bigcap_{n=1}^{\infty} E_n \right) = \lim_{n \rightarrow \infty} \nu(E_n) = \inf_n \nu(E_n). \quad (7)$$

- **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 \mathcal{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 $\nu(E) = \int_X f \mathbf{1}_{\{E\}} d\mu$, then it corresponds to $f \geq 0$, $f \leq 0$ and $f = 0$ for **μ -almost everywhere** $x \in E$.

- **Lemma 2.2** [Folland, 2013]

*Any **measurable 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 $(\mu - \nu)$.

2.2 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 2.3** (**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 $X = P \cup N$ 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 2.4 (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

$$\nu = \nu_+ - \nu_- \quad \text{and} \quad \nu_+ \perp \nu_-.$$

- **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 2.5** 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 2.6** If ν_1, ν_2 are signed measures that both omit $\pm\infty$, then $|\nu_1 + \nu_2| \leq |\nu_1| + |\nu_2|$

- **Exercise 2.7** 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

$$\left| \int_X f d\nu \right| \leq \int_X |f| d|\nu|$$

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

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

- **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 $\nu(E) = \int_E f d\mu$ 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.

2.3 Lebesgue-Radon-Nikodym Theorem

- **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 $E \in \mathcal{B}$ for which $\mu(E) = 0$.
- **Proposition 2.8** 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 2.9 (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 2.10** If $f \in L^1(X, \mu)$, for every $\epsilon > 0$, there exists a $\delta > 0$, such that $|\int_E f d\mu| < \epsilon$ 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 2.11** [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 2.12 (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

$$\lambda \perp \mu, \quad \text{and} \quad \rho \ll \mu, \quad \text{and} \quad \nu = \lambda + \rho.$$

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

$$d\nu = f d\mu, \quad \text{for some } f.$$

- **Definition** The decomposition $\nu = \rho + \lambda$, where $\lambda \perp \mu$ and $\rho \ll \mu$, is called the **Lebesgue decomposition** 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

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

- **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 singular** 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 positive 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 2.13** [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 2.14** 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 2.15** If μ_1, \dots, μ_n are measures on (X, \mathcal{B}) , then there exists a measure μ such that $\mu_i \ll \mu$ for all $i = 1, \dots, n$, namely, $\mu = \sum_{i=1}^n \mu_i$.

- **Exercise 2.16 (Conditional Expectation)**

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

3 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*.

3.1 The Lebesgue Differentiation Theorem in One Dimension

- **Theorem 3.1 (Lebesgue differentiation theorem, one-dimensional case).**

Let $f : \mathbb{R} \rightarrow \mathbb{C}$ be an **absolutely integrable function**, and let $F : \mathbb{R} \rightarrow \mathbb{C}$ 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 3.2** (*Lebesgue differentiation theorem, second formulation*).

Let $f : \mathbb{R} \rightarrow \mathbb{C}$ be an **absolutely integrable** function. Then

$$\lim_{h \rightarrow 0+} \frac{1}{h} \int_{[x, x+h]} f(t)dt = f(x) \quad (8)$$

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

$$\lim_{h \rightarrow 0+} \frac{1}{h} \int_{[x-h, x]} f(t)dt = f(x) \quad (9)$$

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

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

The conclusion (8) 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* $f : \mathbb{R} \rightarrow \mathbb{R}$), 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.

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

3.2.1 Absolute Integrable Version

- **Theorem 3.3** (*Lebesgue Differentiation Theorem (Absolute Integrable version)*)
[Tao, 2011]

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

$$\lim_{r \rightarrow 0} \frac{1}{m(B(x, r))} \int_{B(x, r)} |f(z) - f(x)| dz = 0 \quad (10)$$

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

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 (10) 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

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

- **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 *measurable*.
 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 *Lebesgue differentiation theorem*:

Theorem 3.4 (The Hardy-Littlewood Maximal Theorem) [Stein and Shakarchi, 2009, Folland, 2013]

Suppose f is integrable, then

1.

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

is measurable.

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

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

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

Note that $H^*f \geq |f|$, a.e., 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 Vitali covering 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 3.5 (Lebesgue number lemma)**

For any **open covering** \mathcal{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 \mathcal{A} containing it.

- **Lemma 3.6 (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

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

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

Suppose E is a set of finite measure and \mathcal{B} is a Vitali covering of E . For any $\delta > 0$, we can find **finitely many balls** B_1, \dots, B_N in \mathcal{B} that are disjoint and so that

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

- **Corollary 3.8** [Stein and Shakarchi, 2009, Royden and Fitzpatrick, 1988]
Following the setting above, we can arrange the choice of balls so that

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

3.2.2 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 $f(x)\mathbf{1}_B$ is integrable.

- This theorem follows from *the Hardy-Littlewood maximal inequality*

Theorem 3.9 [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

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

- **Definition** [Stein and Shakarchi, 2009]

If $f \in L^1_{loc}(\mathbb{R}^d)$, the **Lebesgue set** of f consists of all points $\bar{x} \in \mathbb{R}^d$ for which $f(\bar{x})$ is **finite** and

$$\lim_{\substack{m(B) \rightarrow 0 \\ \bar{x} \in B}} \frac{1}{m(B)} \int_B |f(z) - f(\bar{x})| dz = 0.$$

or equivalently, [Folland, 2013],

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

- These results are from *the Hardy-Littlewood maximal inequality*

Corollary 3.10 Suppose E is a measurable 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 3.11** 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 \bar{x} or has **bounded eccentricity** at \bar{x} if there is a constant $c > 0$ such that for each U_α there is a ball B with

$$\bar{x} \in B, \quad U_\alpha \subset B, \quad m(U_\alpha) \geq c m(B).$$

- **Theorem 3.12 (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

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

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

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

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

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

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

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

- **Definition** A Borel measure ν on \mathbb{R}^d will be called **regular** if

1. $\nu(K) < \infty$ for every **compact** K ;
2. $\nu(E) = \inf\{\nu(U) : U \text{ open}, E \subseteq U\}$ 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 3.13 (Lebesgue Density from Radon-Nikodym derivative)** [Folland, 2013]
Let ν be a **regular signed measure** on \mathbb{R}^d , and let $d\nu = d\lambda + f dm$ be its Lebesgue-Radon-Nikodym decomposition, where $\lambda \perp m$. Then for m -almost every $x \in \mathbb{R}^d$,

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

where E_r **shrinks regularly** to x .

4 The Fundamental Theorem of Calculus for Lebesgue Integral

4.1 Functions of Bounded Variations

- **Theorem 4.1 (Monotone Differentiation Theorem)**. [Tao, 2011]

Any function $F : \mathbb{R} \rightarrow \mathbb{R}$ 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 $0 \leq \theta \leq 1$; 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, \dots is any enumeration of the *rational numbers*, 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 4.2 (*Continuous-singular decomposition for monotone functions*).

Let $F : \mathbb{R} \rightarrow \mathbb{R}$ 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 $\lim_{y \rightarrow x^-} F(y)$ and $\lim_{y \rightarrow x^+} F(y)$ both exist, but are **unequal**, with $\lim_{y \rightarrow x^-} F(y) < \lim_{y \rightarrow x^+} F(y)$.
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 4.3** 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 $F : \mathbb{R} \rightarrow \mathbb{R}$ be a function. **The total variation** $\|F\|_{TV(\mathbb{R})}$ (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, \dots, 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

$$\|F\|_{TV([a, b])} := \sup_{a \leq x_0 < \dots < x_n \leq b} \sum_{i=1}^n |F(x_i) - F(x_{i-1})|$$

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

- **Proposition 4.4** If $F : \mathbb{R} \rightarrow \mathbb{R}$ 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 4.5** For any functions $F, G : \mathbb{R} \rightarrow \mathbb{R}$, 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**): $\|F + G\|_{TV(\mathbb{R})} \leq \|F\|_{TV(\mathbb{R})} + \|G\|_{TV(\mathbb{R})}$

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

- **Exercise 4.6** (Bounded Variation is **Stronger** than Bounded)
 1. Show that every function $f : \mathbb{R} \rightarrow \mathbb{R}$ of **bounded variation** is **bounded**, and that the limits $\lim_{x \rightarrow +\infty} f(x)$ and $\lim_{x \rightarrow -\infty} f(x)$, are well-defined.
 2. Give a counterexample of a **bounded, continuous, compactly supported** function f that is **not of bounded variation**.
- **Proposition 4.7** A function $F : \mathbb{R} \rightarrow \mathbb{R}$ 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

$$F^+(x) = \sup_{x_0 < \dots < x_n \leq x} \sum_{i=1}^n \max\{F(x_i) - F(x_{i-1}), 0\}$$

$$F^-(x) = \sup_{x_0 < \dots < x_n \leq x} \sum_{i=1}^n \max\{-F(x_i) + F(x_{i-1}), 0\}$$

We have

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

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

- **Corollary 4.8** (*Bounded Variation Differentiation Theorem*).
Every bounded variation function is differentiable almost everywhere.
- **Definition** (*Locally Bounded Variation*)
A function is locally of bounded variation if it is of bounded variation on every compact interval $[a, b]$.
- **Corollary 4.9** (*Locally Bounded Variation Differentiation Theorem*).
Every locally bounded variation function is differentiable almost everywhere.
- **Definition** (*Lipschitz Continuous Function*)
A function $f : \mathbb{R} \rightarrow \mathbb{R}$ is said to be Lipschitz continuous if there exists a constant $C > 0$ such that

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

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

Corollary 4.10 (*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 $f : \mathbb{R} \rightarrow \mathbb{R}$ is said to be **convex** if one has $f((1-t)x + ty) \leq (1-t)f(x) + tf(y)$ for all $x < y$ and $0 < t < 1$.

Corollary 4.11 (*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

4.2 The Second Fundamental Theorem of Calculus for Lebesgue Integral

- **Proposition 4.12** (*Upper bound for second fundamental theorem*).

Let $F : [a, b] \rightarrow \mathbb{R}$ be **monotone non-decreasing** (so that, as discussed above, F' is defined almost everywhere, is unsigned, and is measurable). Then

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

In particular, F' is **absolutely integrable**.

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

Proposition 4.13 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 4.14 (*Second fundamental theorem for Lipschitz functions*).

Let $F : [a, b] \rightarrow \mathbb{R}$ 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 fundamental 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 4.15 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 $F'(x) := \lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$ 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 $F : \mathbb{R} \rightarrow \mathbb{R}$ is **continuous** if, for every $\epsilon > 0$ and $x_0 \in \mathbb{R}$, there exists a $\delta > 0$ such that $|F(b) - F(a)| \leq \epsilon$ whenever (a, b) is an interval of length at most δ that contains x_0 .

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

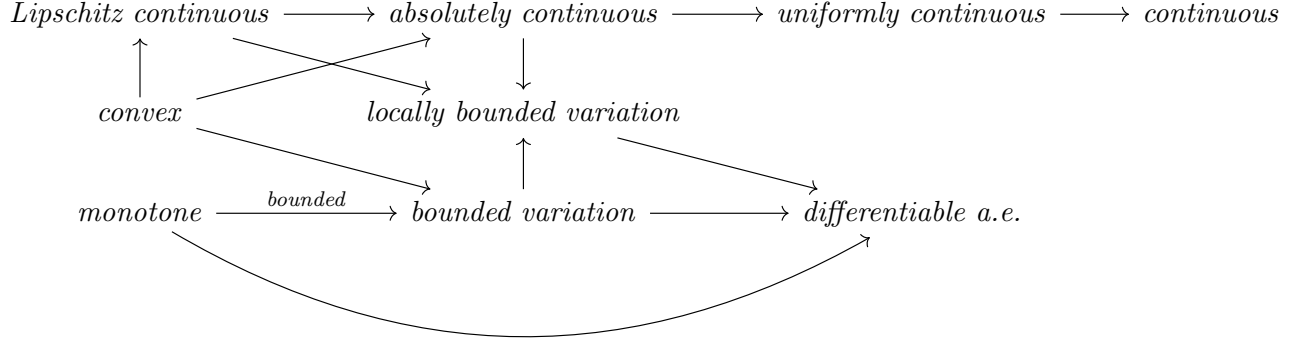
- **Definition (Absolute Continuity)**

A function $F : \mathbb{R} \rightarrow \mathbb{R}$ is said to be **absolutely continuous** if, for every $\epsilon > 0$, there exists a $\delta > 0$ such that $\sum_{j=1}^n |F(b_j) - F(a_j)| \leq \epsilon$ whenever $(a_1, b_1), \dots, (a_n, b_n)$ is a **finite collection of disjoint intervals of total length $\sum_{j=1}^n |b_j - a_j|$ at most δ** .

- **Proposition 4.16** *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 $f : \mathbb{R} \rightarrow \mathbb{R}$ 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** \nrightarrow **absolutely continuous**: See Cantor function example [Tao, 2011].

– **absolutely continuous** \nrightarrow **Lipschitz continuous**: $x \mapsto \sqrt{x}$

- **Proposition 4.17** *Absolutely continuous functions map **null sets** to **null sets**, i.e. if $F : \mathbb{R} \rightarrow \mathbb{R}$ is **absolutely continuous** and E is a null set then $F(E) := \{F(x) : x \in E\}$ is also a null set.*

Exercise 4.18 *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 4.19 (Second Fundamental Theorem for Absolutely Continuous Functions).

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

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

- **Proposition 4.20 (Classification of Absolute Continuous Function)**
A function $F : [a, b] \rightarrow \mathbb{R}$ is **absolutely continuous** if and only if it takes the form

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

for some **absolutely integrable** $f : [a, b] \rightarrow \mathbb{R}$ 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 4.21 (Second Fundamental Theorem of Calculus, again).**
Let $[a, b]$ be a compact interval of positive length, let $F : [a, b] \rightarrow \mathbb{R}$ 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 4.22** Let $F : [-1, 1] \rightarrow \mathbb{R}$ 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 derivative 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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