

# Folland Measure Theory Sections 2.4-2.6

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# Review of Uniform and Pointwise Convergence

Let  $X$  be a non-empty set. Consider a sequence of functions  $\{f_n\}$  with each  $f_n : X \rightarrow \mathbb{C}$  that converges (uniformly or pointwise) to a function  $f : X \rightarrow \mathbb{C}$ .

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**Pointwise Convergence:** We say that  $\{f_n\}$  converges pointwise to  $f$  if for all  $\varepsilon > 0$  and  $x \in X$ , there is an integer  $N(x)$  (note the dependency on  $x$ ) such that

$$n \geq N(x) \implies |f_n(x) - f(x)| < \varepsilon.$$

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**Uniform Convergence:** We say that  $\{f_n\}$  converges uniformly to  $f$  if for all  $\varepsilon > 0$ , there exists an integer  $N$  (note that it is independent of  $x$ ) such that

$$n \geq N \implies |f_n(x) - f(x)| < \varepsilon,$$

for all  $x \in X$ .

# Convergence in Measure

- We say a sequence  $\{f_n\}$  of measurable functions with  $f_n : X \rightarrow \mathbb{C}$  is **Cauchy in measure** if for every  $\varepsilon > 0$ ,

$$\mu(\{x \in X : |f_n(x) - f_m(x)| \geq \varepsilon\}) \rightarrow 0 \text{ as } m, n \rightarrow \infty.$$

- The sequence **converges in measure** to  $f : X \rightarrow \mathbb{C}$  if for every  $\varepsilon > 0$ ,

$$\mu(\{x \in X : |f_n(x) - f(x)| \geq \varepsilon\}) \rightarrow 0 \text{ as } n \rightarrow \infty.$$

## Theorem

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*$\{f_n\}$  Cauchy in measure. Then there is a measurable function  $f$  such that  $f_n \rightarrow f$  in measure, and there is a subsequence  $f_{n_j}$  that converges to  $f$  a.e. If  $f_n \rightarrow g$  in measure, then  $g = f$  a.e.*

# Egoroff's Theorem

## Theorem (Egoroff's Theorem)

*Suppose that  $(X, \mathcal{M}, \mu)$  is a finite measure space, and  $\{f_n\}$  and  $f$  are complex-valued measurable functions on  $X$  such that  $f_n \rightarrow f$  a.e. For every  $\varepsilon > 0$ , there exists  $E \subset X$  such that  $\mu(E) < \varepsilon$  and  $f_n \rightarrow f$  uniformly on  $E^C$ . This is called **almost uniform convergence**, and it implies p.w. convergence a.e. and convergence in measure.*



# Egoroff's Theorem (Proof)

## Proof.

WLOG let  $f_n \rightarrow f$  on  $X$ . For  $k, n \in \mathbb{N}$ , define

$$E_n(k) = \bigcup_{m=n}^{\infty} \{x \in X : |f_m(x) - f(x)| \geq \frac{1}{k}\}.$$

For fixed  $k$ ,  $E_n(k)$  decreases as  $n \rightarrow \infty$ , and  $\bigcap_n E_n(k) = \emptyset$ , so since  $\mu(X) < \infty$ , we have that  $\mu(E_n(k)) \rightarrow 0$  as  $n \rightarrow \infty$ . Given  $\varepsilon > 0$  and  $k \in \mathbb{N}$ , choose  $n_k$  so large enough so that

$$\mu(E_{n_k}(k)) < \frac{\varepsilon}{2^k},$$

and let  $E = \bigcup_k E_{n_k}(k)$ . Then  $\mu(E) < \varepsilon$ , and we have

$$|f_n(x) - f(x)| < \frac{1}{k}$$

for  $n > n_k$  and  $x \in E^c$ . Thus  $f_n \rightarrow f$  uniformly on  $E^c$ . □

# Pointwise, Uniform, and $L^1$ Convergence

We saw from **Problem 2** in our last HW that  $L^1$  convergence implies pointwise convergence almost everywhere for a subsequence. Let us summarize the relationships between different modes of convergence.

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We saw from **Problem 2** in our last HW that  $L^1$  convergence implies pointwise convergence almost everywhere for a subsequence. Let us summarize the relationships between different modes of convergence.

- Uniform convergence  $\implies$  pointwise convergence  $\implies$  pointwise convergence a.e.
- $L^1$  convergence  $\implies$  pointwise convergence a.e. for a subsequence
- Uniform convergence  $\not\iff L^1$  convergence (unless the measure space is finite, then  $\text{unif} \implies L^1$ )
- Pointwise convergence (a.e.)  $\not\iff L^1$  convergence

# The Product Measure

Recall from last HW that we can construct a product measure on the product measurable space  $(X, \mathcal{M})$  with

$$X = \prod_{j=1}^n X_j, \quad \mathcal{M} = \bigotimes_{j=1}^n \mathcal{M}_j.$$

This measure  $\mu$  is unique on  $\mathcal{M}$  and when the components  $\mu_j$  are  $\sigma$ -finite, then  $\mu$  is  $\sigma$ -finite. We want to define integration for this product space  $(X, \mathcal{M}, \mu)$  and relate it to integration on the individual component spaces  $(X_j, \mathcal{M}_j, \mu_j)$ .

# The Monotone Class Lemma

## Definition

A **monotone class** on a space  $X$  is a collection of subsets  $\mathcal{C}$  that's closed under countable increasing unions and countable decreasing unions. In other words, if  $E_j \in \mathcal{C}$  and

$$E_1 \subset E_2 \subset \cdots, \text{ then } \bigcup_{j=1}^{\infty} E_j \in \mathcal{C}$$

and similarly for intersections. We denote the monotone class generated by a collection of subsets  $\mathcal{A}$  by  $\mathcal{M}(\mathcal{A})$ .

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## Theorem (Monotone Class Lemma)

*If  $\mathcal{A}$  is an algebra of subsets of  $X$ , then  $\mathcal{M}(\mathcal{A}) = \sigma(\mathcal{A})$ .*

# The Monotone Class Lemma (Proof)

## Proof.

Since  $\sigma(\mathcal{A})$  is a monotone class, we have that

$$\mathcal{M}(\mathcal{A}) \subset \sigma(\mathcal{A}),$$

so if we can show that  $\mathcal{M}(\mathcal{A})$  is a  $\sigma$ -algebra, we are done. For  $E \in \mathcal{M}(\mathcal{A})$ , define

$$\mathcal{C}(E) = \{F \in \mathcal{M}(\mathcal{A}) : E \setminus F, F \setminus E, E \cap F \in \mathcal{M}(\mathcal{A})\}.$$

Clearly,  $\emptyset, E \in \mathcal{C}(E)$  and  $E \in \mathcal{C}(F)$  iff  $F \in \mathcal{C}(E)$ .  $\mathcal{C}(E)$  is a monotone class. If  $E \in \mathcal{A}$ , then  $F \in \mathcal{C}(E)$  for all  $F \in \mathcal{A}$  because it is an algebra. Hence,  $\mathcal{M}(\mathcal{A}) \subset \mathcal{C}(E)$ . This means that  $E \in \mathcal{C}(F)$  for all  $E \in \mathcal{A}$ , so  $\mathcal{M}(\mathcal{A}) \subset \mathcal{C}(F)$ . Hence,  $\mathcal{M}(\mathcal{A})$  is an algebra. But if  $\{E_j\} \subset \mathcal{M}(\mathcal{A})$ , we have that  $\bigcup_1^n E_j \in \mathcal{M}(\mathcal{A})$  for every  $n$ , and since it is a monotone class,  $\bigcup_1^\infty E_j \in \mathcal{M}(\mathcal{A})$ . □

## Definition

Suppose  $(X, \mathcal{M}, \mu)$  and  $(Y, \mathcal{N}, \nu)$  are measure spaces, and  $f$  a function on  $X \times Y$ . If  $E \subset X \times Y$ , then for fixed  $x \in X, y \in Y$ :

- The **x-section**  $E_x$  of  $E$

$$E_x = \{y \in Y : (x, y) \in E\}.$$

- The **y-section**  $E^y$  of  $E$  is

$$E^y = \{x \in X : (x, y) \in E\}.$$

- The **x and y sections**  $f_x$  and  $f^y$  of  $f$  are

$$f_x(y) = f^y(x) = f(x, y).$$



## Theorem

*The  $x$  and  $y$  sections are measurable sets in their respective  $\sigma$ -algebra, and if  $f$  is  $\mathcal{M} \otimes \mathcal{N}$ -measurable, then  $f_x$  is  $\mathcal{N}$ -measurable and  $f_y$  is  $\mathcal{M}$ -measurable for all  $x \in X$  and  $y \in Y$  respectively.*

# Preliminary Results

## Theorem

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## Theorem (2.36)

*Suppose  $(X, \mathcal{M}, \mu)$  and  $(Y, \mathcal{N}, \nu)$  are  $\sigma$ -finite measure spaces. If  $E \in \mathcal{M} \otimes \mathcal{N}$ , then  $x \mapsto \nu(E_x)$  and  $y \mapsto \mu(E^y)$  are measurable  $X$  and  $Y$  respectively, and*

$$(\mu \times \nu)(E) = \int \nu(E_x) d\mu(x) = \int \mu(E^y) d\nu(y).$$

## Preliminary Result Theorem (2.36) (Proof)

### Proof.

We shall not present a full proof as it is quite long.

- Suppose the measures are finite. Let  $\mathcal{C}$  be the set of all  $E \in \mathcal{M} \otimes \mathcal{N}$  such that the conclusions of the thm are true. It can be proved that this collection contains the algebra of finite disjoint unions of rectangles.
- Show that  $\mathcal{C}$  is a monotone class, which shows that  $\mathcal{C}$  is the  $\sigma$ -algebra generated by the algebra of finite disjoint unions of rectangles by the **MCL**, using the **MCT**, showing that the result is true for finite measure spaces.
- Rewrite  $X \times Y$  as a disjoint union of  $X_j \times Y_j$  with finite measure. Use the **MCT** to conclude that this works for  $\sigma$ -finite measure spaces.



# The Fubini-Tonelli Theorem

## Theorem (Fubini-Tonelli)

Let  $(X, \mathcal{M}, \mu)$  and  $(Y, \mathcal{N}, \nu)$  be  $\sigma$ -finite measure spaces.

1. **(Tonelli)** If  $f \in L^+(X \times Y)$ , then  $g(x) = \int f_x \, d\nu$  and  $h(y) = \int f^y \, d\mu$  are in  $L^+(X)$  and  $L^+(Y)$  respectively, and

$$\begin{aligned} \int f \, d(\mu \times \nu) &= \int \left[ \int f(x, y) \, d\nu(y) \right] d\mu(x) \\ &= \int \left[ \int f(x, y) \, d\mu(x) \right] d\nu(y). \end{aligned}$$

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2. **(Fubini)** If  $f \in L^1(\mu \times \nu)$ , then  $f_x \in L^1(\nu)$  for a.e.  $x \in X$ ,  $f^y \in L^1(\mu)$  for a.e.  $y \in Y$ , the a.e.-defined functions  $g(x) = \int f_x \, d\nu$  and  $h(y) = \int f^y \, d\mu$  are in  $L^1(\mu)$  and  $L^1(\nu)$  respectively, and the formulas in 1. are valid.

# The Fubini-Tonelli Theorem (Proof)

## Proof.

Tonelli's theorem is **Theorem 2.36** when  $f$  is a characteristic function, and since simple functions are linear combinations of char functions, the result holds for nonnegative simple functions by linearity. If  $f \in L^+(X \times Y)$ , let  $\{f_n\}$  be a sequence of simple functions that increase pointwise to  $f$  (recall **Theorem 2.10**). The **MCT** implies that the corresponding  $g_n$  and  $h_n$  increase to  $g$  and  $h$  and

$$\int g \, d\mu = \lim \int g_n \, d\mu = \lim \int f_n \, d(\mu \times \nu) = \int f \, d(\mu \times \nu),$$

$$\int h \, d\nu = \lim \int h_n \, d\nu = \lim \int f_n \, d(\mu \times \nu) = \int f \, d(\mu \times \nu).$$

Hence,  $f \in L^+(X \times Y)$  and  $\int f \, d(\mu \times \nu) < \infty$ , then  $g, h < \infty$  a.e. meaning that  $f_x \in L^1(\nu)$  a.e. and  $f^y \in L^1(\mu)$  a.e. For Fubini, just take the positive/negative/real/imaginary parts of  $f$ .  $\square$

# Fubini-Tonelli for Complete Measures

Even if  $\mu$  and  $\nu$  are complete measures, then  $\mu \times \nu$  is usually never complete. Hence, we must take extra care when dealing with the completion of product spaces.

## Theorem

*Let  $(X, \mathcal{M}, \mu)$  and  $(Y, \mathcal{N}, \nu)$  be complete,  $\sigma$ -finite measure spaces, and let  $(X \times Y, \mathcal{L}, \lambda)$  be the completion of  $(X \times Y, \mathcal{M} \otimes \mathcal{N}, \mu \times \nu)$ . If  $f$  is  $\mathcal{L}$ -measurable and either*

- *(a)  $f \geq 0$*
- *(b)  $f \in L^1(\lambda)$*

*then  $f_x$  and  $f_y$  are measurable in their respective  $\sigma$ -algebras for a.e.  $x$  and  $y$  (if (a) holds), and if (b) holds, then  $f_x$  and  $f_y$  are also integrable for a.e.  $x$  and  $y$ . Moreover,  $x \mapsto \int f_x \, d\nu$  and  $y \mapsto \int f_y \, d\mu$  are measurable, and in case (b) also integrable, and*

$$\int f \, d\lambda = \iint f(x, y) \, d\mu(x) \, d\nu(y) = \iint f(x, y) \, d\nu(y) \, d(\mu(x)).$$

As described so far, we know how to construct measures on the product of measure spaces. Hence, we are ready to construct the Lebesgue measure on  $\mathbb{R}^n$ .



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## Definition

The **Lebesgue measure on  $\mathbb{R}^n$**  is denoted  $\lambda^n$  and is defined as the completion of the product of the one-dimensional Lebesgue measure  $\lambda$  with itself  $n$  times on  $(\mathbb{R}^n, \mathcal{L}^n)$  where

$$\mathbb{R}^n = \mathbb{R} \times \cdots \times \mathbb{R}, \quad \mathcal{L}^n = \mathcal{L} \otimes \cdots \otimes \mathcal{L}.$$

# Properties of the $n$ -dimensional Lebesgue Measure

## Theorem (Approximations)

$\lambda^n$  has the following properties (note the similarities to the 1-dim case):

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- We may write  $E \in \mathcal{L}^n$  in the form

$$E = A_1 \cup N_1 = A_2 \setminus N_2$$

where  $A_1$  is  $F_\sigma$  and  $A_2$  is  $G_\delta$  and  $N_1, N_2$  are null sets.

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- If  $\lambda^n(E) < \infty$ , for any  $\varepsilon > 0$ , there is a finite collection  $\{R_j\}_1^N$  of disjoint rectangles whose sides are intervals such that

$$\lambda^n \left( E \triangle \bigcup_{j=1}^N R_j \right) < \varepsilon.$$

# Properties of the $n$ -dimensional Lebesgue Measure (cont.)

## Theorem (Invariance)

$\lambda^n$  has the following properties (note the similarities to the 1-dim case):

1. **Translation invariance:** for  $a \in \mathbb{R}^n$  define the addition map  $\tau_a : \mathbb{R}^n \rightarrow \mathbb{R}^n$  by  $\tau_a(x) = x + a$ . Then

$$E \in \mathcal{L}^n \implies \tau_a(E) \in \mathcal{L}^n \text{ and } \lambda^n(\tau_a(E)) = \lambda^n(E).$$

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2. **Scaling under linear maps:** if  $E \in \mathcal{L}^n$ , then  $T(E) \in \mathcal{L}^n$  and

$$\lambda^n(T(E)) = |\det T| \lambda^n(E).$$

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2. **Scaling under linear maps:** if  $E \in \mathcal{L}^n$ , then  $T(E) \in \mathcal{L}^n$  and

$$\lambda^n(T(E)) = |\det T| \lambda^n(E).$$

3. **Rotation invariance:** if  $E \in \mathcal{L}^n$ , then rotations are linear maps  $T$  such that  $TT^T = I$ . Let  $T$  be one such map. Then

$$\lambda^n(T(E)) = \lambda^n(E).$$

# Change of Variables

Now that we know some properties of the Lebesgue measure, we want to investigate how it interacts with integration in general. One such result from the 1-dim case that we are interested in generalizing is  $u$ -substitution, or change of variables.



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## Theorem (CoV for Linear Maps)

*Suppose  $T \in GL(n, \mathbb{R})$  ( $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ ). If  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is a Lebesgue measurable function on  $\mathbb{R}^n$ , so is  $f \circ T : \mathbb{R}^n \rightarrow \mathbb{R}$ . If  $f \geq 0$  or  $f \in L^1(\lambda^n)$ , then*

$$\int f(x) \, dx = |\det T| \int (f \circ T)(x) \, dx.$$

It can be proved using the Fubini-Tonelli theorem.

## Change of Variables (cont.)

The previous theorem applies only to linear maps, a very restrictive condition. We want to generalize this result, and it turns out  $C^1$  diffeomorphisms are functions "nice" enough that our formula holds similarly.

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### Definition

Let  $G = (g_1, \dots, g_n)$  be a map from an open set  $\Omega \subset \mathbb{R}^n$  into  $\mathbb{R}^n$ , whose components  $g_j$  are  $C^1$  (continuous first order partials). Let  $D_x(G)$  be the linear map defined by the Jacobian matrix of  $G$  at  $x$ .  $G$  is a  $C^1$  **diffeomorphism** if  $G$  is **injective** and  $D_x G$  is **invertible** for all  $x \in \Omega$ .

## Change of Variables (cont.)

We are now finally ready to present the change of variables theorem.

### Theorem (Change of Variables)

*Let  $G : \Omega \rightarrow \mathbb{R}^n$  be a  $C^1$  diffeomorphism. If  $f$  is Lebesgue measurable on  $G(\Omega)$ , then  $f \circ G$  is Lebesgue measurable on  $\Omega$ . If  $f \geq 0$  or  $f \in L^1(G(\Omega), \lambda^n)$ , then*

$$\int_{G(\Omega)} f(x) \, dx = \int_{\Omega} (f \circ G)(x) |\det D_x G| \, dx.$$

*Furthermore, if  $\Omega \supset E \in \mathcal{L}^n$ , then  $G(E) \in \mathcal{L}^n$  and*

$$\lambda^n(G(E)) = \int_E |\det D_x G| \, dx.$$

The proof is long and difficult, so it shall be omitted.