

Ch 4 Abstract Theory of the Integral II

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(4.1) As a preparation for defining measures on product spaces, we discuss outer measures on an arbitrary space X .

Defn. Let X be a set and 2^X the set of all subsets of X .

A set function $\mu^* : 2^X \rightarrow \mathbb{R}_+$ is called an outer measure on X if

- (i) $\mu^*(\emptyset) = 0$
- (ii) μ^* is monotone (i.e. $A \subseteq B \Rightarrow \mu^*(A) \leq \mu^*(B)$)
- (iii) μ^* is countably subadditive i.e. if $\{A_i\}$ is a sequence of sets

$$\mu^*\left(\bigcup_{i=1}^{\infty} A_i\right) \leq \sum_{i=1}^{\infty} \mu^*(A_i)$$

Given an outer measure μ^* and $A \subseteq X$, we say that A is μ^* -measurable if

$$\forall E \subseteq X \quad \mu^*(E) = \mu^*(E \cap A) + \mu^*(E \cap A^c)$$

Proposition Let μ^* be an outer measure on X and $\mathcal{O}$ the class of μ^* -measurable sets.

Then $\mathcal{O}$ is a σ -algebra.

The proof is entirely similar to our procedure for Lebesgue integration.

Left to the enthusiastic student !!

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Defn. Let $\mathcal{C}$ be a collection of subsets of X s.t. $A, B \in \mathcal{C} \Rightarrow A \cap B \in \mathcal{C}$ and $A \in \mathcal{C} \Rightarrow A^c$ is a finite disjoint union of sets in $\mathcal{C}$. We call $\mathcal{C}$ a semi-algebra.

[N.B an algebra $\mathcal{O}$ of sets is defined by $A, B \in \mathcal{O} \Rightarrow A \cap B \in \mathcal{O}$ and $A \cup B \in \mathcal{O}$ and $A^c \in \mathcal{O}$]

Proposition Let $\mathcal{C}$ be a semi-algebra. Then

(i) $\exists$ a smallest algebra $\mathcal{O}$ containing $\mathcal{C}$
~~such that $\nu(\emptyset) = 0$ if $\emptyset \in \mathcal{C}$~~

(This is called the algebra generated by $\mathcal{C}$)

(ii) Suppose ν is a non-negative set function monotone on $\mathcal{C}$ such that $\nu(\emptyset) = 0$ if $\emptyset \in \mathcal{C}$

The following conditions are equivalent:

(A) ν has a unique extension to a countably additive set function π on $\mathcal{O}$

(B) If $C \in \mathcal{C}$ is the union of a finite disjoint collection $\{C_i\}_{i=1}^n \subseteq \mathcal{C}$ then $\nu(C) = \sum_{i=1}^n \nu(C_i)$

(C) If $C \in \mathcal{C}$ is the union of a countable disjoint collection $\{C_i\}_{i=1}^{\infty} \subseteq \mathcal{C}$ then $\nu(C) = \sum_{i=1}^{\infty} \nu(C_i)$

Proof. (i) The algebra $\mathcal{A}$ is defined by adding to $\mathcal{C}$ all finite disjoint unions of sets in $\mathcal{C}$.

It requires a little work to show that this defines an algebra of sets.

(ii) $(\alpha) \Rightarrow (\beta) \Rightarrow (\gamma)$ is clear from the definition

($\beta \Rightarrow \alpha$) Now, if $A = \bigcup_{i=1}^n E_i$ where $E_i \in \mathcal{C}$ are disjoint, we define

$$\mu(A) = \sum_{i=1}^n \mu(E_i). \text{ We need to check that}$$

this defines μ unambiguously.

Suppose that $E = \bigcup_{i=1}^n E_i = \bigcup_{j=1}^m F_j$ are two

finite disjoint collections of elements of $\mathcal{C}$ such that

$$A = \bigcup_{i=1}^n E_i = \bigcup_{j=1}^m F_j.$$

Note that $C_i = \bigcup_{j=1}^m C_i \cap F_j$, so by (B) elements of $\mathcal{C}$, so by (B)

$$\mu(C_i) = \sum_{j=1}^m \mu(C_i \cap F_j)$$

$$\text{Thus, by definition } \sum_{i=1}^n \mu(C_i) = \sum_{i=1}^n \sum_{j=1}^m \mu(C_i \cap F_j)$$

$$= \sum_{j=1}^m \sum_{i=1}^n \mu(C_i \cap F_j) = \sum_{j=1}^m \mu(F_j)$$

Thus μ is well-defined on the algebra $\mathcal{A}$

Furthermore μ is clearly monotone since μ is

We now must show that μ is countably additive

This follows from the monotone condition

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If $\{C_i\}_{i=1}^{\infty}$ is a countable disjoint collection and $C = \bigcup_{i=1}^{\infty} C_i \in \mathcal{C}$, then for all n

$$\bar{\mu}(C) = \bar{\mu}\left(\bigcup_{i=1}^{\infty} C_i\right) \geq \bar{\mu}\left(\bigcup_{i=1}^n C_i\right) = \sum_{i=1}^n \mu(C_i)$$

$$\text{Thus } \bar{\mu}(C) \geq \sum_{i=1}^{\infty} \mu(C_i)$$

But since $\bar{\mu}$ is monotone $\bar{\mu}(C) \leq \sum_{i=1}^{\infty} \mu(C_i)$

Thus we have equality and $\bar{\mu}$ is countably additive on $\mathcal{C}$

Definition Let $\mathcal{C}$ be an algebra of sets in X . A monotone set function μ on $\mathcal{C}$ s.t. $\mu(\emptyset) = 0$

is called a premeasure if $\forall$ disjoint sequence

$\{A_i\}$ in $\mathcal{C}$ s.t. $\bigcup_{i=1}^{\infty} A_i \in \mathcal{C}$

$$\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mu(A_i)$$

Now suppose we have same algebra of sets in X equipped with a premeasure μ .

For any $E \in 2^X$, define

$$\mu^*(E) = \inf_{E \subseteq \bigcup_{i=1}^{\infty} A_i} \sum_{i=1}^{\infty} \mu(A_i)$$

Proposition

μ^* is an outer measure on 2^X .

Proof

~~Lemma~~ μ^*

Lemma If ν is a premeasure on $\mathcal{O}$ and $A \in \mathcal{O}$ and A_i any sequence of sets in $\mathcal{O}$ s.t. $A \subseteq \bigcup A_i$, then $\nu(A) \leq \sum_{i=1}^{\infty} \nu(A_i)$

Proof Let $B_n = A \cap A_n \cap A_{n+1}^c \cap \dots \cap A_n^c$
Then $B_n \in \mathcal{O}$ and $B_n \subseteq A_n$ and A is a disjoint union of $\{B_n\}_{n=1}^{\infty}$. Thus

$$\nu(A) = \sum_{n=1}^{\infty} \nu(B_n) \leq \sum_{n=1}^{\infty} \nu(A_n)$$

Proof of the Propⁿ

Clearly ν^* is non-negative and monotone and $\nu^*(\emptyset) = 0$

We have to show that ν^* is countably sub-additive

Let $E \subseteq \bigcup_{i=1}^{\infty} E_i$. If $\nu^*(E_i) = \infty$ for some i , then

$\sum_{i=1}^{\infty} \nu^*(E_i) = \infty \geq \nu^*(E) \geq \nu^*(E_i) = \infty$ so we are finished.

Thus we may suppose $\nu^*(E_i) < \infty$ for all i .

Let $\varepsilon > 0$ and, for each i , choose a sequence $\{A_{ij}\}_{j=1}^{\infty}$ of sets in $\mathcal{O}$ s.t. $E_i \subseteq \bigcup_{j=1}^{\infty} A_{ij}$ and

$$\sum_{j=1}^{\infty} \nu(A_{ij}) \leq \nu^*(E_i) + \frac{\varepsilon}{2^i}.$$

$$\begin{aligned} \text{Then } \nu^*(E) &\leq \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \nu(A_{ij}) \leq \sum_{i=1}^{\infty} \left(\nu^*(E_i) + \frac{\varepsilon}{2^i} \right) \\ &= \sum_{i=1}^{\infty} \nu^*(E_i) + \varepsilon. \end{aligned}$$

This holds for any ε , so $\nu^*(E) \leq \sum_{i=1}^{\infty} \nu^*(E_i)$

This shows that ν^* is countably sub-additive and hence is an outer measure □

Properties of this extension

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If $A \in \mathcal{O}$ and A is measurable wrt ν^* , then

$$\nu^*(A) = \nu(A)$$

Proof. We want to show that for $E \in 2^X$

$$\nu^*(E) = \nu^*(E \cap A) + \nu^*(E \cap A^c)$$

This is clear if $\nu^*(E) = \infty$, so let's assume $\nu^*(E) < \infty$

Then we have $E \subseteq \bigcup_{i=1}^{\infty} A_i$ for some sequence $\{A_i\}_{i=1}^{\infty}$ of disjoint sets in $\mathcal{A}$ st. $\sum \nu(A_i) \leq \nu^*(E) + \epsilon$

Then $E \cap A \subseteq \bigcup_{i=1}^{\infty} (A_i \cap A)$ and $E \cap A^c \subseteq \bigcup_{i=1}^{\infty} (A_i \cap A^c)$ are both coverings by sets in $\mathcal{O}$. Thus

$$\begin{aligned} \nu^*(E) &\leq \nu^*(E \cap A) + \nu^*(E \cap A^c) \quad (\text{subadditivity}) \\ &\leq \sum_{i=1}^{\infty} \nu(A_i \cap A) + \sum_{i=1}^{\infty} \nu(A_i \cap A^c) \quad (\text{defn of } \nu^*) \\ &= \sum_{i=1}^{\infty} \nu(A_i \cap A) + \nu(A_i \cap A^c) \\ &= \sum_{i=1}^{\infty} \nu(A_i) \quad (\text{finite additivity of } \nu) \\ &\leq \nu^*(E) + \epsilon. \end{aligned}$$

Since this holds for any $\epsilon > 0$, it is true for $\epsilon = 0$ $\square$

Definition If $\mathcal{O}$ is an algebra of sets, let

$\mathcal{O}_\sigma =$ the sets formed by taking countable unions of elements of $\mathcal{O}$

$\mathcal{O}_{\sigma\delta} =$ the sets formed by taking countable intersections of elements of $\mathcal{O}_\sigma$.

Proposition A Let ν be a premeasure on an algebra $\mathcal{O}$ and let ν^* be the induced outer measure.

For every $E \in 2^X$ and for every $\epsilon > 0 \exists A \in \mathcal{O}_\sigma$ such that $E \subseteq A$ and $\nu^*(A) = \nu^*(E) + \epsilon$.

Proof. By defⁿ of ν^* there is a sequence $\{A_i\}_{i=1}^\infty$ of sets in $\mathcal{O}$ st. $E \subseteq \bigcup_{i=1}^\infty A_i$ and

$$\sum_{i=1}^\infty \nu(A_i) \leq \nu^*(E) + \epsilon.$$

Clearly $A = \bigcup_{i=1}^\infty A_i \in \mathcal{O}_\sigma$ and $\nu^*(A) \leq \sum_{i=1}^\infty \nu(A_i) \leq \nu^*(E) + \epsilon$

Proposition B Let ν be a premeasure on an algebra $\mathcal{O}$ and ν^* the induced outer measure on 2^X . For every set E with $\nu^*(E) < \infty$ there is $A \in \mathcal{O}_\sigma$ with $E \subseteq A$ and $\nu^*(E) = \nu^*(A)$

Proof For all n , use Proposition A to get

$A_n \in \mathcal{O}_\sigma$ st. $E \subseteq A_n$ and $\nu^*(E) \leq \nu^*(A_n) \leq \nu^*(E) + \frac{1}{n}$

Now let $A = \bigcap_{n=1}^\infty A_n$. Then $A \in \mathcal{O}_\sigma$ and

$$\nu^*(E) = \nu^*(A)$$

□

Theorem (Carathéodory Extension Theorem)

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Let ν be a premeasure on the algebra $\mathcal{A}$ and ν^* the induced outer measure on 2^X .

Let $\mathcal{B}$ be the smallest σ -algebra containing $\mathcal{A}$

(Note: Intersections of σ -algebras are σ -algebras!)

and ν' any extension of ν to $\mathcal{B}$

If $\nu^*(B) < \infty$ for a set $B \in \mathcal{B}$, then $\nu'(B) = \nu^*(B)$

If ν is a σ -finite measure, then the extension ν' to $\mathcal{B}$ is unique.

Proof. Let $A \in \mathcal{A}_\sigma$: it is a countable disjoint union of sets in $\mathcal{A}$ and so ν' is uniquely defined on $\mathcal{A}_\sigma$ and must coincide with ν^*

Let $B \in \mathcal{B}$ with $\nu^*(B) < \infty$. Choose $A \in \mathcal{A}_\sigma$ with $B \subseteq A$ and $\nu^*(A) \leq \nu^*(B) + \varepsilon$

Then $\nu'(B) \leq \nu'(A) = \nu^*(A) \leq \nu^*(B) + \varepsilon$

Thus it follows that $\nu'(B) \leq \nu^*(B) \quad \forall B \in \mathcal{B}$

Since the ν^* measurable sets form a σ -algebra containing $\mathcal{A}$, they must contain $\mathcal{B}$.

Let B be measurable, $A \in \mathcal{A}_\sigma$ s.t. $B \subseteq A$ with $\nu^*(A) \leq \nu^*(B) + \varepsilon$

Then $\nu'(A) = \nu^*(A) = \nu^*(B) + \nu^*(A \setminus B)$

Hence $\mu^*(A \setminus B) < \epsilon$ if $\mu^*(B) < \infty$.

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$$\begin{aligned}\text{Thus } \mu^*(B) &\leq \mu^*(A) = \mu'(A) = \mu'(B) + \mu'(A \setminus B) \\ &\leq \mu'(B) + \epsilon\end{aligned}$$

From this, it follows that $\mu^*(B) \leq \mu'(B)$.

Now suppose μ is σ -finite. Let $\{X_i\}$ be a countable collection of disjoint sets in $\mathcal{C}$ so that $X = \bigcup_{i=1}^{\infty} X_i$ and $\mu(X_i) < \infty \quad \forall i$.

Now any $B \in \mathcal{B}$ can be written $B = \bigcup_{i=1}^{\infty} (X_i \cap B)$

$$\text{So we have } \mu'(B) = \sum_{i=1}^{\infty} \mu'(X_i \cap B)$$

If $\bar{\mu}$ is a second measure ~~with these on~~ $\mathcal{B}$ which coincides with μ' on $\mathcal{C}$, by the previous part

$$\bar{\mu}(X_i \cap B) = \mu^*(X_i \cap B) = \mu'(X_i \cap B).$$

$$\text{Thus } \bar{\mu}(B) = \sum \bar{\mu}(X_i \cap B) = \sum \mu'(X_i \cap B) = \mu'(B) \quad \square$$

Summary of this section

Let $\mathcal{C}$ be a semi-algebra with a non-negative monotone set function ν on $\mathcal{C}$ such that

$$\nu(c) = \sum_{i=1}^n \nu(c_i) \quad \text{if } c = \bigcup_{i=1}^n c_i \text{ is}$$

a finite ^{union of disjoint sets} ~~disjoint union of sets~~ in $\mathcal{C}$

THEN

• ν extends to a countably additive set function (also called ν) on $\mathcal{O}\mathcal{L}$ — the algebra generated by $\mathcal{C}$

• This extended ν is a premeasure on $\mathcal{O}\mathcal{L}$ which we use to define a ~~non-measure~~ center measure ν^* on X

• The measurable sets of ν^* include $\mathcal{O}\mathcal{L}$ and $\nu^*|_{\mathcal{O}\mathcal{L}} = \nu$.

• Let $\mathcal{B}$ be the smallest σ -algebra containing $\mathcal{O}\mathcal{L}$ (so it is contained in the measurable set)

Then there is a unique measure on $\mathcal{B}$ which coincides with ν on $\mathcal{O}\mathcal{L}$, viz ν^* .

Product Measures

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Let $(X, \mathcal{B}, \mu)$ and $(Y, \mathcal{C}, \nu)$ be two measure spaces.

Consider the direct product $X \times Y$.

We want to make it into a ~~prod~~ measure space.

Defn. If $A \subseteq X$ and $B \subseteq Y$ are measurable sets then $A \times B \subseteq X \times Y$ is called a measurable rectangle. Let $\mathcal{R}$ denote the collection of all measurable rectangles.

Define a set function λ on $\mathcal{R}$ by

$$\lambda(A \times B) = \mu(A) \nu(B)$$

• Notice that $(A \times B) \cap (C \times D) = (A \cap C) \times (B \cap D)$

$$\text{and } (A \times B)^c = (A^c \times B) \cup (A \times B^c) \cup (A^c \times B^c)$$

so $\mathcal{R}$ is a semi-algebra

Lemma Let $\{A_i \times B_i\}$ be a countable collection of disjoint measurable rectangles st.

$$A \times B = \bigcup_{i=1}^{\infty} A_i \times B_i \quad \text{is a measurable rectangle}$$

$$\text{Then } \lambda(A \times B) = \sum_{i=1}^{\infty} \lambda(A_i \times B_i)$$

Proof Let $x \in A$ be fixed. Then for all $y \in B$

(x, y) is in exactly one rectangle $A_i \times B_i$

(as they are disjoint). Hence B is the disjoint union of all the B_i such that $x \in A_i$ corresponds to B_i

$$\text{Thus } \sum \nu(B_i) \chi_{A_i}(x) = \nu(B) \chi_A(x)$$

$$\text{so } \int \nu(B_i) \chi_{A_i}(x) d\mu(x) = \nu(B) \int \chi_A d\mu$$

$$\text{so } \sum \nu(B_i) \mu(A_i) = \nu(B) \mu(A) = \lambda(A \times B) \quad \square$$

Thus we've shown that λ is a premeasure on the semi-algebra $\mathcal{R}$. Using the Carathéodory theorem we may extend λ to obtain a complete measure on $X \times Y$. We denote this by $\mu \times \nu$.

Note that $\mu \times \nu$ is defined on the smallest σ -algebra containing $\mathcal{R}$ (called the product σ -algebra) and may be extended uniquely to the class of $\mu \times \nu$ measurable sets.

Examples If $X = Y = \mathbb{R}$, then $\mu \times \nu$ is Lebesgue measure on $\mathbb{R}^2$.

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Definition Given $x \in X$ and $E \subseteq X \times Y$, define

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

as the section of E at x .

Note that $\chi_{E_x}(y) = \chi_E(x, y)$

Consider the $\mu \times \nu$ msble sets in $X \times Y$

Lemma - Let $x \in X$ and $E \in \mathcal{R}_{\sigma\delta}$. Then

E_x is a msble subset of Y .

Proof. This is true if $E \in \mathcal{R}$ by definition of $\mathcal{R}$

Let $E \in \mathcal{R}_{\sigma}$, so that $E = \bigcup_{i=1}^{\infty} E_i$, where

$E_i \in \mathcal{R}$ for all i

$$\text{Then } \chi_{E_x}(y) = \chi_E(x, y) = \sup_i \chi_{E_i}(x, y)$$

$$= \sup_i \chi_{(E_i)_x}(y)$$

Since $E_i \in \mathcal{R}$, $(E_i)_x$ is msble, so $\chi_{(E_i)_x}$ is μ -msble function. Thus the sup. is measurable. It follows that χ_{E_x} is a msble fn, which implies that E_x is a msble set

Now suppose $E \in \mathcal{R}_{\sigma\delta}$ - we may suppose $E = \bigcap_{i=1}^{\infty} E_i$ where $E_i \in \mathcal{R}_{\sigma}$.

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$$\begin{aligned}\text{Now } \chi_{E_x} &= \chi_E(x, y) = \inf_i \chi_E(x, y) \\ &= \inf_i \chi_{(E_i)_x}(y)\end{aligned}$$

Since $(E_i)_x$ is a measurable set, it follows as above $\square$

that E_x is measurable

Lemma Let $E \in \mathcal{R} \otimes \mathcal{S}$ and suppose that

$$\mu \times \nu(E) < \infty. \text{ Define } g(x) = \nu(E_x)$$

Then g is a measurable function of x and

$$\int_X g d\mu = \mu \times \nu(E)$$

Proof. The lemma is clearly true if $E \in \mathcal{R}$

If $E \in \mathcal{R}_\sigma$ has the form $E = \bigcup_{i=1}^{\infty} E_i$ with $E_i \in \mathcal{R}$ pairwise disjoint, let $g_i(x) = \nu((E_i)_x)$.

Then $g_i \geq 0$, g_i is measurable and $g = \sum_{i=1}^{\infty} g_i$

Letting $g^n = \sum_{i=1}^n g_i$, we see that $g^n \uparrow g$. Then

g is measurable and so by the monotone convergence theorem

$$\int g d\mu = \sum_{i=1}^{\infty} \int g_i d\mu = \sum_{i=1}^{\infty} \mu \times \nu(E_i) = \mu \times \nu(E) \quad \square$$

Thus the lemma holds if $E \in \mathcal{R}_\sigma$

Now suppose $E \in \mathcal{R}_\sigma \otimes \mathcal{S}$. Choose a sequence

$\{E_i\}_{i=1}^{\infty}$ in $\mathcal{R}_\sigma$ with $E_i \subseteq E_{i+1}$ and $E = \bigcup_{i=1}^{\infty} E_i$

and $(\mu \times \nu)(E_i) < \infty$ for all i .

It follows that $\mu \times \nu(E) < \infty$, so there exists (96)

$A \in \mathcal{R}_0$ with

$$\mu \times \nu(A) \leq \mu \times \nu(E) + \epsilon$$

Let $g_i(x) = \nu((E_i)_x)$. Then $g(x) = \lim_{i \rightarrow \infty} g_i(x)$, so

g is measurable. Furthermore $\nu((E_{i+1})_x) \leq \nu((E_i)_x)$

So again by the MCT we have

$$\begin{aligned} \int g(x) d\mu(x) &= \lim_i \int g_i(x) d\mu(x) = \lim_i \mu \times \nu(E_i) \\ &= \mu \times \nu(E). \end{aligned}$$

We now come to the two big theorems on product measures.

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Theorem (Fubini) Let $(X, \mathcal{B}, \mu)$ and $(Y, \mathcal{C}, \nu)$ be two measure spaces, and f a $\mu \times \nu$ integrable function on $X \times Y$. Then

(i) For ν a.e. x , f_x defined by $f_x(y) = f(x, y)$

is ν integrable on Y , and for μ a.e. y , f_y

defined by $f_y(x) = f(x, y)$ is μ -integrable on X .

(ii) The function defined a.e. on X by

$$x \mapsto \int_Y f_x(y) d\nu(y) \quad \text{is } \mu\text{-integrable}$$

and the function defined a.e. on Y by

$$y \mapsto \int_X f_y(x) d\mu(x) \quad \text{is } \nu\text{-integrable}$$

$$(iii) \quad \int_X \left(\int_Y f(x, y) d\nu(y) \right) d\mu(x) = \int_Y \left(\int_X f(x, y) d\mu(x) \right) d\nu(y)$$

$$= \int_{X \times Y} f(x, y) d\mu \times \nu(x, y)$$

Proof Clear in the case where $f(x, y) = \chi_C(x, y)$

where C is a measurable subset of $X \times Y$.

Hence, it's clear for simple functions.

Since f is integrable on $X \times Y$ we can find a sequence $\{f_n\}$ of simple functions s.t.

$$\int |f - f_n| d\mu \times \nu \rightarrow 0 \quad \text{and} \quad f - f_n \geq 0 \text{ in measure}$$

and $f_n \leq f$.

Since f_n is increasing to f , $f_n(x, y) \leq f(x, y)$

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So f_x is the pointwise increasing limit of $(f_n)_x$

Thus f_x is measurable and by the MCT

$$\lim_{n \rightarrow \infty} \int (f_n)_x dx = \int f_x dx \text{ and thus } \int f_x dx \text{ is an}$$

increasing sequence

Since the f_n are simple functions, each of the

functions $x \mapsto \int (f_n)_x dx$ is measurable and integrable

(in $\mathbb{R}$). It follows that $\int f_n dx$ is also measurable

and integrable in $\mathbb{R}$, and

$$\iint f_n dx dv = \lim_{n \rightarrow \infty} \iint (f_n)_x dx dv = \iint f_x dx dv$$

$$= \lim_{n \rightarrow \infty} \int f_n dx = \int f dx$$

It follows that

$$\iint f(x, y) dv(y) d\mu(x) = \int f d\mu(x)$$

The other ~~inequality~~ equality follows by symmetry $\square$

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Theorem (Tonelli) Let $(X, \mathcal{B}, \mu)$ and $(Y, \mathcal{C}, \nu)$ be two σ -finite measure spaces, and let f be a nonnegative fn on $X \times Y$

Suppose that

(i) For a.e. x , $f_x(y) = f(x, y)$ is an integrable fn on Y and for a.e. y , $f_y(x) = f(x, y)$ is an integ. fn. on X

(ii) The functions $x \mapsto \int f(x, y) d\nu(y)$ and

$y \mapsto \int f(x, y) d\mu(x)$ are both integrable on X (resp. Y)

Then f is integrable and

$$\begin{aligned} \int_X \int_Y f(x, y) d\nu(y) d\mu(x) &= \int_Y \int_X f(x, y) d\mu(x) d\nu(y) \\ &= \int_{X \times Y} f(x, y) d\mu \times \nu(x, y) \end{aligned}$$

Proof Since X, Y are σ -finite, $X \times Y$ is σ -finite

Write $X \times Y = \bigcup_{i=1}^{\infty} Z_i$ with $Z_i \cap Z_j = \emptyset$ for $i \neq j$ and $(\mu \times \nu)(Z_i) < \infty$ for all i .

It will suffice to show that f is integrable on each of the Z_i 's.

To see this, let $f_n = f \vee n = \begin{cases} f(x) & \text{if } f(x) \leq n \\ n & \text{if } f(x) > n \end{cases}$

Then $f_n \uparrow f$. By the Bounded Convergence Theorem

f_n is integrable, and by the MCT, the theorem

follows. □

Infinite product spaces

Let $\{(X_i, \mathcal{B}_i, \mu_i)\}_{i=1}^{\infty}$ be a sequence of measure spaces.

Define $X = \prod_{i=1}^{\infty} X_i = \{x = (x_1, x_2, \dots) : x_i \in X_i\}$

Let $\mathcal{C} = \text{set of all rectangles } \{B_1 \times \dots \times B_n : n \in \mathbb{N}\}$
 $B_i \in \mathcal{C}_i, i=1, \dots, n$

where $B_1 \times \dots \times B_n = \{x \in X : x_1 \in B_1, x_2 \in B_2, \dots, x_n \in B_n\}$

Then $\mathcal{C}$ is a semi-algebra and

$N = \bigotimes_{i=1}^{\infty} \mu_i$ may be defined on $\mathcal{C}$ by

$$N(B_1 \times \dots \times B_n) = \mu_1(B_1) \mu_2(B_2) \dots \mu_n(B_n)$$

Using the Carathéodory theorem, we may define

$\mathcal{B}$ = the smallest σ -algebra containing $\mathcal{C}$ and extend μ to a measure on $\mathcal{B}$

$(X, \mathcal{B}, \mu)$ is called the infinite product space

Example : We already saw an example when

$X_i = \{0, 1\}$. Choose $\mu_i(0) = \mu_i(1) = 1/2$ for $i=1, 2, \dots$

Then X = all sequences of 0's and 1's

$$N(\{0\} \times \{1\} \times \{0\} \times \{0, 1\} \times \dots \times \{0, 1\} \times \dots)$$

$$= \mu_1(0) \mu_2(1) \mu_3(0) \dots \quad (\text{finite product})$$

Let $(X, \mathcal{A}, \mu)$ be a probability space, i.e. a

measure space with $\mu(X) = 1$.

Probabilist interpret the elements in $\mathcal{A}$ as events

(X is the sample space) and elements of $\mathcal{A}$ as collections

of events, which may be the outcome of some

observations. If $A \in \mathcal{A}$, then $\mu(A)$ is thought of as

the probability that this collection of events occurs

eg - infinite product space $\prod_{i=1}^{\infty} \Omega_i \leftrightarrow$ coin tossing

- unit interval $[0, 1] \leftrightarrow$ probability that an atom

emits an α -particle within unit time

Random variables are measurable functions

Given a probability space, a sub σ -algebra $\mathcal{B} \subseteq \mathcal{A}$

corresponds heuristically to partial information

(eg info. only up to n tosses of the coin, or up to

time t)

Often the space X is derived by Ω - and if a

$\omega \in \Omega$ is chosen and the probability that it lies in

$B \subseteq \mathcal{B}$ is $\mu(B)$, which we denote by $P(B)$.

The observer, not knowing ω , might only know the

information that $\omega \in B$ for a given $B \in \mathcal{B}$ - but

not for all $A \in \mathcal{A}$.

How can we model this?

Of course, given two sets A, B , the probability of ω belongs to A w.r.t. a smaller set B

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$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \text{unless } P(B)=0, \text{ in which case it isn't defined}$$

So suppose we know $P(A|B_i)$ for $B_1, B_2, \dots$
and suppose $\mathcal{B}$ is the σ -algebra generated by the disjoint sets $B_1, B_2, \dots$

$$\text{Let } f(\omega) = \frac{P(A \cap B_i)}{P(B_i)} \quad \text{if } \omega \in B_i, \quad i=1, 2, \dots$$

This random variable is denoted $P(A|\mathcal{B})$ — the conditional probability of A given $\mathcal{B}$.

If $P(B_i)=0$ for $B_i \neq \emptyset$, this definition is not well-made. In that case, we take $P(A|\mathcal{B})$ to be an arbitrary constant value, so actually $P(A|\mathcal{B})$ is a whole family of measurable functions defined a.e. Choosing one of these is called choosing a version of $P(A|\mathcal{B})$.

— | —
In a more general setting, let $\mathcal{B}$ be any sub σ -algebra of $\mathcal{A}$. Imagine an observer who knows, for each $A \in \mathcal{A}$ whether $\omega \in A$ or not.

Think of $\mathcal{B}$ as a collection of experiments.

For $A \in \mathcal{A}$, fixed, we define a finite measure on $\mathcal{B}$

$$\text{by } \nu(B) = \mu(A \cap B) \quad \forall B \in \mathcal{B}$$

Then $\nu(B)=0 \Rightarrow \nu(B)=0$. By the Radon-Nikodym theorem applied to ν and ν , $\exists$ a $\mathcal{B}$ -measurable function $f = \frac{d\nu}{d\nu}$ s.t. $\forall B \in \mathcal{B}$

$$\nu(A \cap B) = \nu(B) = \int_B f d\nu$$

The function f is denoted $P(A|B)$

It's a random variable with two properties

(i) $P(A|B)$ is $\mathcal{B}$ -measurable and integrable

(ii) $P(A|B)$ satisfies

$$\int_B P(A|B) dP = \nu(B) \quad \forall B \in \mathcal{B}.$$

There will in general be many such random variables but any two are equal with probability 1 (i.e. almost surely) A specific such random variable is called a version of the conditional probability

Example. Take $X = [0, 1]$, $\mathcal{B}_n = \sigma$ -alg. generated by

$$\left[\frac{j}{2^n}, \frac{j+1}{2^n} \right) : j=0, \dots, 2^n-1$$

Then for a set A

$$P(A|B)(\omega) = \frac{1}{2^n} \nu(A \cap \left[\frac{j}{2^n}, \frac{j+1}{2^n} \right)) \text{ if } \frac{j}{2^n} \leq \omega < \frac{j+1}{2^n}$$

Note that $0 \leq P(A|B) \leq 1$ a.s.

and if $A = \bigcup_{n=1}^{\infty} A_n$ is a disjoint union, then

$$P(A|B) = \sum_n P(A_n|B).$$

Conditional Expectations

Defn Let f be an integrable random variable on $(X, \mathcal{F}, \mu)$ and $\mathcal{B} \subseteq \mathcal{F}$ a σ -algebra.

There exists a random variable

$$E_{\mu}(f | \mathcal{B}) \text{ such that}$$

(i) $E_{\mu}(f | \mathcal{B})$ is $\mathcal{B}$ -measurable and integrable

$$(ii) \forall B \in \mathcal{B} \quad \int_B E_{\mu}(f | \mathcal{B}) d\mu = \int_B f d\mu$$

Proof. Define a measure ν on $\mathcal{B}$ by

$$\nu(B) = \int_B f d\mu. \text{ This is finite, as } f \text{ is integrable}$$

and $\nu \ll \mu$. Let $h = \frac{d\nu}{d\mu}$. Then observe

$$h = E_{\mu}(f | \mathcal{B})$$

□

As usual, thus $E_{\mu}(f | \mathcal{B})$ is only defined a.e.

Any such h is called a version of the conditional

expectation

$E_{\mu}(f | \mathcal{B})$ can be interpreted as the expected

value of f for someone who knows, for each

$B \in \mathcal{B}$, whether $w \in B$ or $w \notin B$.

If $\mathcal{B} = \mathcal{F}$, then $E_{\mu}(f | \mathcal{B}) = f_{\text{a.e.}}$
 If $\mathcal{B} = \{\emptyset, \Omega\}$ then $E_{\mu}(f | \mathcal{B}) = \int_{\Omega} f d\mu$

Theorem Suppose f, g and $\{f_n\} \in L^1(X, \mu)$

Then

(i) If $f = a$ (const) a.e. then $E_N(f | \mathcal{B}) = a$

(ii) If a, b are constants then

$$E_N(af + bg | \mathcal{B}) = a E_N(f | \mathcal{B}) + b E_N(g | \mathcal{B})$$

(iii) $|E_N(f | \mathcal{B})| \leq E_N(|f| | \mathcal{B})$

(iv) If $\lim_n f_n = f$ a.e. and $|f_n| \leq g$, then

$$\lim_n (E_N f_n | \mathcal{B}) = E_N(f | \mathcal{B}) \text{ a.e.}$$

(v) If f is $\mathcal{B}$ -measurable and g and $f \cdot g \in L^1$, then

$$E_N(fg | \mathcal{B}) = f E_N(g | \mathcal{B})$$

(vi) (Jensen's Inequality)

$$|E_N(f | \mathcal{B})|^p \leq E_N(|f|^p | \mathcal{B}) \text{ for } 1 \leq p < \infty$$

— 1 —

Martingales

Let $f_1, f_2, \dots$ be a sequence of random variables on $(X, \mathcal{A}, \mu)$ and let $\mathcal{A}_1, \mathcal{A}_2, \mathcal{A}_3, \dots$ be a sequence of σ -algebras. Then $(f_n, \mathcal{A}_n)$ is called a (sub) martingale if

(i) $\mathcal{A}_n \subseteq \mathcal{A}_{n+1} \quad \forall n$

(ii) f_n is $\mathcal{A}_n$ measurable (f_n is adapted to $\mathcal{A}_n$)

(iii) $\sum_n \int |f_n| d\mu < \infty$

(iv) $E_N(f_{n+1} | \mathcal{A}_n) = f_n$ a.e. ($\leq f_n$ a.e.)

Submartingale

Example $\mathcal{N} = [0, 1]$

$A_n = \sigma$ -algebra generated by $[\frac{j}{2^n}, \frac{j+1}{2^n}]$

take $f \in L^1([0, 1])$.

$$\text{Then } f_n(x) = \frac{1}{2^n} \int_{\frac{j}{2^n}}^{\frac{j+1}{2^n}} f(t) dt \quad \text{if } x \in [\frac{j}{2^n}, \frac{j+1}{2^n}]$$

is a martingale

$$\text{eg } X = \frac{1}{2} [0, 1]$$

$$\mathcal{N}_n = \left(\frac{1}{2}, \frac{1}{2} \right) \quad \mathcal{N} = \bigotimes \mathcal{N}_n$$

$A_n = \mathcal{C}_n = \sigma$ -alg generated by rectangles $B_1 \times \dots \times B_n$ for n fixed

$f \in L^1(X)$

$$f_n(x) = \int \dots \int f(x_1, x_2, \dots, x_n, \sum_{n+1}^{\infty} x_i, \sum_{n+2}^{\infty} x_i, \dots) d\mu_{n+1}(\sum_{n+1}^{\infty} x_i) \dots$$

is a martingale

—

Indeed, let f be any integrable f_n in $L^1(X, \mathcal{N})$

or let $A_1 \subset A_2 \subset A_3 \subset \dots \subset A_n \subset \dots \subset A$ be a sequence

of σ -algebras. Then $f_n = E_n(f | \mathcal{A}_n)$ is a martingale

Theorem (The martingale convergence theorem)

Let $f_1, f_2, \dots, f_n, \dots$ be a (sub) martingale s.t.

$$K = \sup_n \int |f_n| d\mu < \infty$$

Then $\exists$ random vble f on X s.t. $f_n \rightarrow f$ a.s.

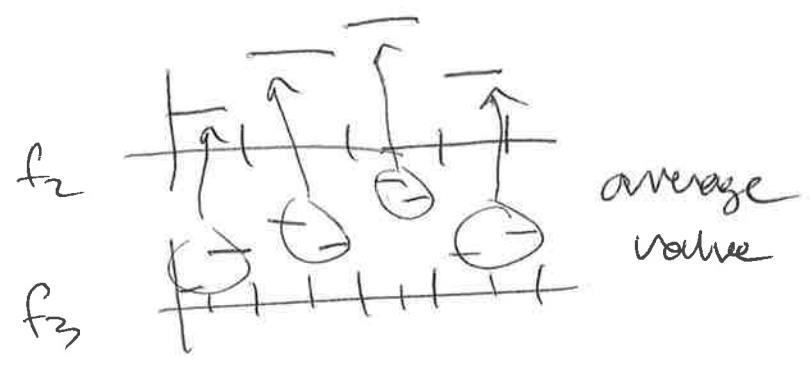
and $\int |f| d\mu \leq K$. $\Leftrightarrow$ (If unif. integr. $f_n \rightarrow f$ in L^1)

Proof. In fact, we want to define f by

$$E_\mu(f_n | A_n) = f_n$$

To do this, let $f = \limsup f_n$ and then show that $\limsup f_n = \liminf f_n$ by analysing the number of upcrossings in the martingale.

Examples On $[0,1]$



Then $f_i \rightarrow f$ a.s.

On $\mathbb{R}$ $f_n = \alpha$ -abs gen by $B \cap [0, n]$



$f_n \rightarrow f$ in L^1 .

To prove the martingale convergence theorem requires two basic inequalities

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Lemma 1 (Kolmogorov's Inequality)

If $X_1, X_2, \dots, X_n$ is a submartingale, then for $\alpha > 0$

$$\mu \{ \max_{i \leq n} X_i \geq \alpha \} \leq \frac{1}{\alpha} \int |X_n| d\mu$$

Proof Let $A_1 = \{ \omega : X_1(\omega) \geq \alpha \}$

$$A_k = \{ \omega : \max_{i \leq k} X_i(\omega) < \alpha \leq X_k(\omega) \}$$

$$A = \bigcup_{k=1}^n A_k = \{ \omega : \max_{i \leq n} X_i(\omega) \geq \alpha \}$$

Since $A_k \in \mathcal{F}_k = \sigma(X_1, \dots, X_k)$

$$\begin{aligned} \int_A X_n d\mu &= \sum_{k=1}^n \int_{A_k} X_n d\mu = \sum_{k=1}^n \int_{A_k} E\{X_n | \mathcal{F}_k\} d\mu \\ &\geq \sum_{k=1}^n \int_{A_k} X_k d\mu \geq \alpha \sum \mu(A_k) = \alpha \mu(A) \end{aligned}$$

$$\text{Thus } \alpha \mu \{ \omega : \max_{i \leq n} X_i \geq \alpha \} \leq \int_{\{ \omega : \max_{i \leq n} X_i(\omega) \geq \alpha \}} X_n d\mu \leq \int X_n^+ d\mu \leq \int |X_n| d\mu \quad \square$$

Note if $\{X_i\}$ is a mgale, $|X_i|$ is a submartingale

$$\text{so } \mu \{ \max_{i \leq n} |X_i| \geq \alpha \} \leq \frac{1}{\alpha} \int |X_n| d\mu$$

Definition If $[\alpha, \beta]$, $\alpha < \beta$, is an interval and

$X_1, \dots, X_n$ are random variables, the number of

upcrossings of $[\alpha, \beta]$ by $X_1(\omega), \dots, X_n(\omega)$ is the number

of times the sequence $X_1(\omega), \dots, X_n(\omega)$ passes from below α to above β .

Specifically, it is the number of integers u, v (109)

$$1 \leq u < v \leq n \quad \text{st.} \quad X_u \leq \alpha$$

$$\alpha < X_i < \beta \quad \text{for } u < i < v \quad \left(\begin{array}{l} \text{vacuous} \\ \text{condition if} \\ u+1=v \end{array} \right)$$

$$\text{or } X_v \geq \beta$$

Theorem (The upcrossings lemma)

If $X_1, X_2, \dots, X_n$ is a submartingale and

$$U(w) = \# \text{ upcrossings of } [\alpha, \beta] \text{ by } X_1(w), \dots, X_n(w)$$

then

$$\int U(w) d\mu(w) \leq \frac{1}{\beta - \alpha} \left(\int |X_n| d\mu + \alpha \right)$$

Proof. We require a lemma

Lemma B. If $X_1, \dots, X_n$ is a submartingale and

τ_1, τ_2 are stopping times (functions $X \rightarrow \mathbb{N}$ st.

$\{\tau_i(w) \leq k\} \in \mathcal{F}_k$) such that $1 \leq \tau_1(w) \leq \tau_2(w) \leq n$ a.p. w

$$\text{then } \int X_{\tau_1(w)}(w) d\mu(w) \leq \int X_{\tau_2(w)}(w) d\mu(w)$$

$$\text{L.V.B. Since } \int |X_{\tau(w)}(w)| d\mu(w) \leq \sum_{k=1}^n \int |X_k| d\mu$$

the functions are integrable]

Proof Let $\Delta_k = X_k - X_{k-1}$. Then

$$\begin{aligned} \int (X_{\tau_1} - X_{\tau_2}) d\mu &= \int \left(\sum_{k=1}^n \Delta_k X_{\tau_1 < k < \tau_2} \right) d\mu \\ &= \sum_{k=1}^n \int_{\tau_1 < k < \tau_2} \Delta_k d\mu \end{aligned}$$

Since $\{w : r_1(w) < k < r_2(w)\} = \{w : r_1(w) \leq k-1\}$

$\bigcap \{w : r_2(w) \leq k-1\} \in \mathcal{F}_{k-1}$ the integrals are non-negative $\square$

Proof of the upcrossings theorem

Let $Y_k = \max\{0, Y_k - \alpha\}$

$$\theta = \beta - \alpha$$

Then U is also the # of upcrossings of $[0, \theta]$ by

the submartingale $X_1, \dots, X_n$.

Define inductively $\tau_0, \tau_1, \dots$

$$\tau_0 = 1$$

τ_k the even k is the smallest j s.t. $\tau_{k-1} < j \leq n$ and $Y_j \geq \theta$, and n if there is no such j

τ_k the odd k is the smallest j s.t. $\tau_{k-1} < j \leq n$

and $Y_j = 0$ and is n if there is no such j

Note that $\tau_k \leq n$ and $\tau_{k+1} = n \Rightarrow \tau_k = n$

Also $\tau_{k+1} < n \Rightarrow \tau_{k+1} < \tau_k$ and hence $\tau_k = \tau_{k+1} = \dots = n$

if k is odd and τ_k is a stopping time, then for $i < n$

$$\{w : \tau_k(w) = j\} = \bigcup_{i=1}^{j-1} \{w : \tau_{k-1}(w) = i\}, \quad Y_i(w) \geq \theta \text{ and } Y_i(w) = 0$$

Also in $\mathcal{F}_j$, and $(\tau_k = n) = [\tau_k \leq n-1] \in \text{Mem } \mathcal{F}_n$

so τ_k is a stopping time

A similar argument for k even shows that (111)
 the τ_k are stopping time, as τ_0 is.

$$\text{Since } Y_n = Y_{\tau_n} \geq Y_{\tau_n} - Y_{\tau_0}$$

$$Y_n \geq \sum_{\substack{\uparrow \\ \text{even}}} (Y_{\tau_k} - Y_{\tau_{k-1}}) + \sum_{\substack{\uparrow \\ \text{odd}}} (Y_{\tau_k} - Y_{\tau_{k-1}})$$

By Lemma B, the last sum has non-negative integral

$$\text{so } \int Y_n d\mu \geq \int \sum_{\substack{\uparrow \\ \text{even}}} (Y_{\tau_k} - Y_{\tau_{k-1}}) d\mu$$

Suppose now k is even

$$\text{If } \tau_{k-1} = n \text{ then } Y_{\tau_{k-1}} = Y_n = Y_{\tau_k}$$

while if $\tau_{k-1} < n$ then $Y_{\tau_{k-1}} = 0$. In either case

$$Y_{\tau_k} - Y_{\tau_{k-1}} \geq 0. \text{ On the other hand, if } 1 \leq u < v \leq n$$

$$Y_u = 0 \text{ and } 0 < Y_i < \theta \text{ for } u < i < v, Y_v = 0$$

(thanks ~~to~~ u, v contributes to an upcrossing of $(0, \theta]$ by $Y_1, \dots, Y_n$). Then $\exists$ even k $1 \leq k \leq n$ st.

$$\tau_{k-1} \leq u < v = \tau_k$$

$$\text{Thus } Y_{\tau_{k-1}} = 0 \text{ and } Y_{\tau_k} \geq \theta \text{ and so } Y_{\tau_k} - Y_{\tau_{k-1}} \geq \theta$$

Since different upcrossings give different values of

$$\sum_{\substack{\uparrow \\ \text{even}}} Y_{\tau_k} - Y_{\tau_{k-1}} \geq U \theta \text{ so } \int Y_n d\mu \geq \theta \int U d\mu$$

In terms of the original random variables

$$(\beta - \alpha) \int U d\mu \leq \int_{X_n > \alpha} (X_n - \alpha) d\mu = \int |X_n| d\mu + |\alpha| \quad \square$$

Proof of the monotone convergence theorem

(112)

Fix α, β and let $V_n = \text{number of upcrossings of } [\alpha, \beta] \text{ by } X_1, \dots, X_n$. Then

$$\int V_n d\mu \leq \int \frac{\beta - \alpha}{|X_n| + |\alpha|} \leq \beta - \alpha$$

Since V_n is non-decreasing and $\int V_n d\mu$ is

bounded, the monotone convergence theorem implies $\sup V_n$ is integrable, hence finite valued a.e.

Let $X^* = \limsup X_i$ $\} \text{ possibly } \infty$
 $X_* = \liminf X_i$

If $X_* < \alpha < \beta < X^*$, then V_n must go to ∞ .
 Since V_n is a.e. bounded

$$\mu\{w: X_*(w) < \alpha < \beta < X^*(w)\} = 0$$

New $\{w: X_* < X^*\} = \cup_{\alpha, \beta \text{ rational}} \{X_* < \alpha < \beta < X^*\}$
 Therefore $X_* = X^*$ a.e. and X_k converges to this common value - which may be ∞ .
 By Fatou's lemma

$$\int |X| \leq \liminf \int |X_n| \leq K$$

Since X is integrable, it is finite a.e.

□