

Albert-Ludwigs-Universität Freiburg

Lecture notes

Probability Theory I

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1 A short repetition of Measure Theory

This chapter contains a brief summary (without proofs) of the measure theoretic foundations as already known from our Analysis III course (BSc Mathematics) or the bridging course on measure theory (MSc Data Science).

1.1 σ -Algebras and measures

Definition 1.1. Let $\Omega \neq \emptyset$. A system \mathcal{A} of subsets of Ω is called σ -algebra if

- (i) $\Omega \in \mathcal{A}$,
- (ii) $A \in \mathcal{A} \implies A^c \in \mathcal{A}$,
- (iii) $A_n \in \mathcal{A} \ \forall n \in \mathbb{N} \implies \bigcup_{n \in \mathbb{N}} A_n \in \mathcal{A}$.

\mathcal{A} is called algebra if only (iii') instead of (iii) is granted:

(iii')

$$A, B \in \mathcal{A} \implies A \cup B \in \mathcal{A}.$$

Lemma 1.2. Let $\Omega \neq \emptyset$ be a set and \mathcal{A} a σ -algebra over Ω . Then the following statements hold true:

- (i) $A_n \in \mathcal{A} \ \forall n \in \mathbb{N} \implies \bigcap_{n \in \mathbb{N}} A_n \in \mathcal{A}$
- (ii) $A, B \in \mathcal{A} \implies A \setminus B \in \mathcal{A}$.

Remark. The pair (Ω, \mathcal{A}) with a σ -algebra \mathcal{A} over Ω is called measurable space.

Arbitrary intersections of σ -algebras over Ω are σ -algebras over Ω again. For any systems \mathcal{E} of subsets of Ω , there exists a smallest σ -algebra $\sigma(\mathcal{E})$ with $\mathcal{E} \subset \sigma(\mathcal{E})$, namely the intersection of all those which contain \mathcal{E} .

Definition 1.3. If Ω is equipped with a topology, then the σ -algebra generated by the open subsets of Ω is called Borel- σ -algebra.

Definition 1.4. Let (Ω, \mathcal{A}) be a measurable space. A map $\mu : \mathcal{A} \rightarrow [0, \infty]$ with $\mu(\emptyset) = 0$ is called measure if it is σ -additive, meaning that

$$A_n \in \mathcal{A} \ \forall n \in \mathbb{N}, \ A_i \cap A_j = \emptyset \ \forall i \neq j \implies \mu \left(\bigcup_{n \in \mathbb{N}} A_n \right) = \sum_{n \in \mathbb{N}} \mu(A_n).$$

The triple $(\Omega, \mathcal{A}, \mu)$ is called measure space. If $\mu(\Omega) = 1$, then μ is named probability measure and $(\Omega, \mathcal{A}, \mu)$ correspondingly probability space. A measure μ is called finite if $\mu(\Omega) < \infty$. It is named σ -finite if there exist sets $\Omega_i \in \mathcal{A} \ \forall i \in \mathbb{N}$ such that $\mu(\Omega_i) < \infty \ \forall i \in \mathbb{N}$ and $\Omega = \bigcup_{i \in \mathbb{N}} \Omega_i$.

Notation. The following notations are occasionally used variants of " $A \cup B$ ", but also convey the information that the sets A and B are disjoint:

$$A \dot{\cup} B, \quad A + B.$$

Lemma 1.5. Let (Ω, \mathcal{A}) be a measurable space, μ a measure on (Ω, \mathcal{A}) , $A, B, A_n \in \mathcal{A} \forall n \in \mathbb{N}$. Then the following statements are satisfied:

- (i) $A \subset B \implies \mu(A) + \mu(B \setminus A) = \mu(B)$. In particular, $A \subset B \implies \mu(A) \leq \mu(B)$ (monotonicity).
- (ii) $\mu(A \cap B) + \mu(A \cup B) = \mu(A) + \mu(B)$. In particular, $\mu(A \cup B) \leq \mu(A) + \mu(B)$ (subadditivity).
- (iii) If $A_n \subset A_{n+1} \forall n \in \mathbb{N}$, then the continuity from below holds:

$$\mu\left(\bigcup_{n \in \mathbb{N}} A_n\right) = \lim_{n \rightarrow \infty} \mu(A_n).$$

- (iv) If $A_{n+1} \subset A_n \forall n \in \mathbb{N}$ and $\mu(A_1) < \infty$, then the continuity from above holds:

$$\mu\left(\bigcap_{n \in \mathbb{N}} A_n\right) = \lim_{n \rightarrow \infty} \mu(A_n).$$

We now pose the question under which conditions an additive functional (uniquely) extends to a measure.

Theorem 1.6 (Carathéodory's existence and uniqueness theorem). Let \mathcal{A} be a system of subsets of Ω with the following properties:

- (i) $\Omega \in \mathcal{A}$,
- (ii) $A, B \in \mathcal{A} \implies B \setminus A$ is a finite disjoint union of sets of \mathcal{A} ,
- (iii) $A, B \in \mathcal{A} \implies A \cap B \in \mathcal{A}$.

[Such a system of subsets of Ω is called half ring.] Let $\mu : \mathcal{A} \rightarrow [0, \infty]$ be a function with the following properties:

- (iv) $\mu(\emptyset) = 0$,
- (v) $A, B \in \mathcal{A}$ with $A \cup B \in \mathcal{A}$ and $A \cap B = \emptyset \implies \mu(A \cup B) = \mu(A) + \mu(B)$ (additivity),
- (vi) There exist $\Omega_i \in \mathcal{A}$ with $\mu(\Omega_i) < \infty$ and $\Omega_i \subset \Omega_{i+1} \forall i \in \mathbb{N}$ such that $\bigcup_{i \in \mathbb{N}} \Omega_i = \Omega$.

(vii) $A, A_n \in \mathcal{A} \ \forall n \in \mathbb{N}$ with $A \subset \bigcup_{n \in \mathbb{N}} A_n$

$$\implies \mu(A) \leq \sum_{n \in \mathbb{N}} \mu(A_n).$$

Then there exists a unique extension of μ to a measure on $\sigma(\mathcal{A})$.

Remark. The system of sets

$$\{(a, b] \cap \mathbb{R} \mid a, b \in \mathbb{R} \cup \{-\infty, +\infty\}, a \leq b\}$$

satisfies the conditions (i)-(iii).

Remark. A probability measure μ is uniquely described by its values on a \cap -stable generator \mathcal{E} of the σ -algebra.

Proof. As μ is normed (i.e. $\mu(\Omega) = 1$) and additive, these values provide all the values on

$$\mathcal{E}' = \left\{ \bigcap_{i=1}^K A_i \mid A_i \text{ or } A_i^c \in \mathcal{E} \ \forall i \leq K, K \in \mathbb{N} \right\}.$$

The set \mathcal{E}' satisfies (i)-(iii):

(i) We can assume that $\emptyset \in \mathcal{E}$, as the goal is to describe μ by its values on \mathcal{E} , and we already know that $\mu(\emptyset) = 0$. Using $K = 1$ and $A_1 = \Omega$ (which is valid, as $\Omega^c = \emptyset \in \mathcal{E}$), $\Omega \in \mathcal{E}'$ follows from the definition of \mathcal{E}' .

(ii) Let $A = \bigcap_{i=1}^K A_i, B = \bigcap_{j=1}^L B_j \in \mathcal{E}'$.

$$\implies B \setminus A = B \cap A^c = \left(\bigcap_{j=1}^L B_j \right) \cap \left(\bigcup_{i=1}^K A_i^c \right) = \bigcup_{i=1}^K \underbrace{\left(A_i^c \cap \bigcap_{j=1}^L B_j \right)}_{\in \mathcal{E}'}$$

finite intersection of sets for which
the set or its complement are in \mathcal{E}

(iii) Let $A = \bigcap_{i=1}^K A_i, B = \bigcap_{j=1}^L B_j \in \mathcal{E}'$. Then,

$$A \cap B = \bigcap_{i=1}^K \bigcap_{j=1}^L (A_i \cap B_j)$$

is a finite intersection of subsets of Ω such that for every set, the set itself or its complement is in \mathcal{E} ; hence, $A \cap B \in \mathcal{E}'$.

Therefore, the claim follows from Theorem 1.6. □

1.2 Dynkin systems and the $\pi - \lambda$ theorem

Definition 1.7. A system of sets $\mathcal{D} \subset \mathcal{P}(\Omega)$ is called Dynkin system over Ω if

- (i) $\Omega \in \mathcal{D}$
- (ii) $A \in \mathcal{D} \implies A^c \in \mathcal{D}$
- (iii) If $A_n \in \mathcal{D}$ are **pairwise disjoint** sets $\forall n \in \mathbb{N}$,

$$\bigcup_{n \in \mathbb{N}} A_n \in \mathcal{D}.$$

Remark. Let \mathcal{D} be a Dynkin system over Ω . (i) and (ii) imply that $\emptyset \in \mathcal{D}$ and therefore, (iii) holds true in particular for finite unions of disjoint sets. If $E \subset D \subset \Omega$ with $E, D \in \mathcal{D}$, then $D^c \in \mathcal{D}$ by (ii), D^c and E are disjoint and therefore,

$$D \setminus E = (D^c \cup E)^c = (D^c + E)^c \in \mathcal{D}.$$

As for σ -algebras, arbitrary intersections of Dynkin systems over Ω are Dynkin systems over Ω again. For any systems \mathcal{E} of subsets of Ω , there exists a smallest Dynkin system $\mathcal{D}(\mathcal{E})$ with $\mathcal{E} \subset \mathcal{D}(\mathcal{E})$, that is

$$\mathcal{D}(\mathcal{E}) = \bigcap_{\substack{\mathcal{D} \text{ Dynkin system} \\ \mathcal{E} \subset \mathcal{D}}} \mathcal{D}.$$

Proposition 1.8. Let $\mathcal{E} \subset \mathcal{P}(\Omega)$ be stable under intersection. Then $\mathcal{D}(\mathcal{E}) = \sigma(\mathcal{E})$.

Proof. Apparently, every σ -algebra is a Dynkin system, so $\mathcal{D}(\mathcal{E}) \subset \sigma(\mathcal{E})$. For the other direction, $\sigma(\mathcal{E}) \subset \mathcal{D}(\mathcal{E})$, it is sufficient to prove that $\mathcal{D}(\mathcal{E})$ is a σ -algebra, that is, it must be stable under intersection (because then, for $A_1, A_2 \in \mathcal{D}$, we have $A_1 \cup A_2 = A_1 + (A_2 \setminus (A_1 \cap A_2)) \in \mathcal{D}$). Let $D \in \mathcal{D}(\mathcal{E})$ be fixed and consider

$$\tilde{D} := \{A \subset \Omega \mid A \cap D \in \mathcal{D}(\mathcal{E})\}.$$

We want to show that $\mathcal{D}(\mathcal{E})$ is contained in \tilde{D} . We observe that \tilde{D} is itself a Dynkin system:

- (a) $\Omega \in \tilde{D}$ because $\Omega \cap D = D \in \mathcal{D}(\mathcal{E})$.
- (b) $A \in \tilde{D} \implies A^c \cap D = D \setminus (A \cap D) \in \mathcal{D}(\mathcal{E})$ by the previous remark, as $A \cap D \subset D$, so $A^c \in \tilde{D}$.
- (c) Let $A_n \in \tilde{D}$ be pairwise disjoint sets. $A_n \cap D \in \mathcal{D}(\mathcal{E})$ are pairwise disjoint sets for $n \in \mathbb{N}$, so

$$\left(\bigcup_{n \in \mathbb{N}} A_n \right) \cap D = \bigcup_{n \in \mathbb{N}} (A_n \cap D) \in \mathcal{D}(\mathcal{E}),$$

i.e. $\bigcup_{n \in \mathbb{N}} A_n \in \tilde{D}$.

For all $E \in \mathcal{E}$, we have $\mathcal{E} \subset \tilde{E}$, because by requirement, \mathcal{E} is stable under intersection. Hence, $\mathcal{D}(\mathcal{E}) \subset \tilde{E}$, as $\mathcal{D}(\mathcal{E})$ is the *smallest* Dynkin system which contains \mathcal{E} , and \tilde{E} is *some* Dynkin system which contains \mathcal{E} . In particular, $D \cap E \in \tilde{E}$ for the fixed D from above, so $E \in \tilde{D}$. As $E \in \mathcal{E}$ was arbitrary, this implies $\mathcal{E} \subset \tilde{D}$, so $\mathcal{D}(\mathcal{E}) \subset \tilde{D}$ (because \tilde{D} is a Dynkin system) for every $D \in \mathcal{D}(\mathcal{E})$. Therefore, $\mathcal{D}(\mathcal{E})$ is stable under intersection. \square

Often, an \cap -stable system of sets is called π -system, whereas a Dynkin system is also referred to as λ -system.

Corollary 1.9 (Dynkin's $\pi - \lambda$ theorem). *If \mathcal{E} is a π -system and \mathcal{L} is a λ -system that contains \mathcal{E} , then $\sigma(\mathcal{E}) \subset \mathcal{L}$.*

1.3 Measurable maps and measure integral

Definition 1.10. *Let $(\Omega, \mathcal{A}), (\Omega', \mathcal{A}')$ be measurable spaces. A map $f : \Omega \rightarrow \Omega'$ is called measurable (or \mathcal{A} - \mathcal{A}' -measurable) if*

$$f^{-1}(A') \in \mathcal{A} \quad \forall A' \in \mathcal{A}'$$

where $f^{-1}(A') := \{\omega \in \Omega \mid f(\omega) \in A'\}$.

Lemma 1.11. *Let $(\Omega, \mathcal{A}), (\Omega', \mathcal{A}')$ be measurable spaces and $\mathcal{E} \subset \mathcal{A}'$ with $\sigma(\mathcal{E}) = \mathcal{A}'$ (meaning \mathcal{E} is a generator of \mathcal{A}'). Then:*

$$f^{-1}(E') \in \mathcal{A} \quad \forall E' \in \mathcal{E}' \iff f^{-1}(A') \in \mathcal{A} \quad \forall A' \in \mathcal{A}'$$

This means that it is sufficient to check measurability on a generator of the σ -algebra.

Notation. *For the sake of readability, we define $\overline{\mathbb{R}} := \mathbb{R} \cup \{-\infty, +\infty\}$ and*

$$\mathcal{B}(\overline{\mathbb{R}}) = \{B \subset \mathbb{R} \mid B \cap \mathbb{R} \in \mathcal{B}(\mathbb{R})\} = \{B \cup E \mid B \in \mathcal{B}(\mathbb{R}), E \subset \{-\infty, +\infty\}\}.$$

Proposition 1.12. *Let (Ω, \mathcal{A}) be a measurable space and $f_n : \Omega \rightarrow \overline{\mathbb{R}}$ an \mathcal{A} - $\mathcal{B}(\overline{\mathbb{R}})$ -measurable function $\forall n \in \mathbb{N}$. Then, the functions*

$$\inf_{n \in \mathbb{N}} f_n, \sup_{n \in \mathbb{N}} f_n, \liminf_{n \rightarrow \infty} f_n, \limsup_{n \rightarrow \infty} f_n$$

are also \mathcal{A} - $\mathcal{B}(\overline{\mathbb{R}})$ -measurable.

Monotone limits play a crucial role for the construction of the measure integral.

Definition 1.13. Let $A \subset \Omega$. The function

$$\mathbb{1}_A : \Omega \rightarrow \{0, 1\}, \omega \mapsto \begin{cases} 1 & \omega \in A \\ 0 & \omega \notin A \end{cases}$$

is called indicator function. If $A_k \in \mathcal{A}$ and $c_k \in \mathbb{R} \forall k \in \{1, \dots, n\}$ for some $n \in \mathbb{N}$, then $\sum_{k=1}^n c_k \mathbb{1}_{A_k}$ is called \mathcal{A} -elementary function.

Measure integral for elementary functions

Let f be an elementary function, i.e. $f = \sum_{k=1}^n c_k \mathbb{1}_{A_k}$ with measurable sets $A_k \subset \Omega$. Then the elementary integral is defined as

$$\int f \, d\mu := \sum_{k=1}^n c_k \mu(A_k).$$

If there is an alternative representation $f = \sum_{k=1}^{n'} c'_k \mathbb{1}_{A'_k}$, one can easily prove that

$$\sum_{k=1}^n c_k \mu(A_k) = \sum_{k=1}^{n'} c'_k \mu(A'_k).$$

Hence, this integral is well-defined.

Lemma 1.14. Let (Ω, \mathcal{A}) be a measurable space and $f : \Omega \rightarrow \mathbb{R}$ an \mathcal{A} - $\mathcal{B}(\mathbb{R})$ -measurable function with $f \geq 0$. Then the following statements hold true.

- (i) There exists a sequence $(f_n)_{n \in \mathbb{N}}$ of elementary functions with $f_n \leq f_{n+1} \forall n \in \mathbb{N}$ and $\lim_{n \rightarrow \infty} f_n = f$.
- (ii) There exist sets $A_k \in \mathcal{A} \forall k \in \mathbb{N}$ and a sequence $(c_k)_{k \in \mathbb{N}}$ of non-negative real numbers such that $f = \sum_{k \in \mathbb{N}} c_k \mathbb{1}_{A_k}$.

That is: Any measurable function $f \geq 0$ is the monotone limit of elementary functions $f_n \nearrow f$.

The crucial idea is to define the integral for general non-negative, measurable functions f by monotone approximation: For non-negative elementary functions $f_n \nearrow f$,

$$\int f \, d\mu := \lim_{\substack{\uparrow \\ = \lim_{n \rightarrow \infty} f_n}} \int f_n \, d\mu = \sup_{n \in \mathbb{N}} \int f_n \, d\mu$$

To ensure that this integral is well-defined, we have to verify that

$$\lim_{n \rightarrow \infty} \int f_n \, d\mu = \lim_{n \rightarrow \infty} \int g_n \, d\mu$$

if g_n is another sequence of elementary functions with $g_n \nearrow f$. This is the content of the next lemma.

Lemma 1.15. *Let $(\Omega, \mathcal{A}, \mu)$ be a measure space and $f : \Omega \rightarrow \mathbb{R}$ a measurable, bounded function with $f \geq 0$. If*

$$f = \sum_{n=1}^{\infty} \alpha_n \mathbb{1}_{A_n} = \sum_{n=1}^{\infty} \beta_n \mathbb{1}_{B_n}$$

with constants $\alpha_n, \beta_n > 0 \forall n \in \mathbb{N}$, then

$$\sum_{n \in \mathbb{N}} \alpha_n \mu(A_n) = \sum_{n \in \mathbb{N}} \beta_n \mu(B_n).$$

We summarize the important findings and state formally the definition.

Definition 1.16. *Let $f \geq 0$ be a non-negative measurable function. By Lemma 1.14, f is the monotone limit of non-negative elementary functions $f_n = \sum_{k=1}^n c_k \mathbb{1}_{A_k}$, that is $f = \sum_{n \in \mathbb{N}} c_n \mathbb{1}_{A_n}$, and we define*

$$\int f \, d\mu := \lim_{n \rightarrow \infty} \int f_n \, d\mu = \sum_{n \in \mathbb{N}} c_n \mu(A_n).$$

By Lemma 1.15, this integral is well-defined.

A measurable (not necessarily non-negative) function $f : \Omega \rightarrow \mathbb{R}$ is called (finitely) integrable if $f^+ := f \cdot \mathbb{1}_{\{f \geq 0\}}$ and $f^- := -f \cdot \mathbb{1}_{\{f < 0\}}$ are integrable, i.e.

$$\int f^+ \, d\mu < \infty \text{ and } \int f^- \, d\mu < \infty.$$

This is the case if and only if $\int |f| \, d\mu < \infty$ (which we call absolute integrability). In this case, we define

$$\int f \, d\mu := \int f^+ \, d\mu - \int f^- \, d\mu.$$

Lemma 1.17 (Properties of the integral). *(i) The integral is linear on the finitely integrable functions, i.e.*

$$\int (\alpha f + \beta g) \, d\mu = \alpha \int f \, d\mu + \beta \int g \, d\mu$$

for all finitely integrable functions f, g and $\alpha, \beta \in \mathbb{R}$.

(ii) The integral is monotone, i.e.

$$g \leq f \implies \int g \, d\mu \leq \int f \, d\mu$$

for all integrable functions f, g .

- (iii) Theorem of monotone convergence: If $(f_n)_{n \in \mathbb{N}}$ is a sequence of finitely integrable or non-negative measurable functions with $f_n \leq f_{n+1} \forall n \in \mathbb{N}$, then

$$\int \lim_{n \rightarrow \infty} f_n \, d\mu = \lim_{n \rightarrow \infty} \int f_n \, d\mu.$$

- (iv) Theorem of dominated convergence: If $(f_n)_{n \in \mathbb{N}}$ is a sequence of measurable functions with $|f_n| \leq f \forall n \in \mathbb{N}$ for some function f with $\int f \, d\mu < \infty$, such that the pointwise limit $g := \lim_{n \rightarrow \infty} f_n$ exists, then

$$\int g \, d\mu = \lim_{n \rightarrow \infty} \int f_n \, d\mu.$$

- (v) Fatou's Lemma: If f, f_n are finitely integrable and $f \leq f_n \forall n \in \mathbb{N}$, then the function $\liminf_{n \rightarrow \infty} f_n$ is finitely integrable, and

$$\liminf_{n \rightarrow \infty} \int f_n \, d\mu \geq \int \liminf_{n \rightarrow \infty} f_n \, d\mu.$$

1.4 Image measure and transformation formula

Definition 1.18. Let $(\Omega, \mathcal{A}, \mu)$ be a measure space, (Ω', \mathcal{A}') a measurable space and $X : \Omega \rightarrow \Omega'$ a measurable map. The image measure μ^X on (Ω', \mathcal{A}') is defined as

$$\mu^X(A') := \mu(X^{-1}(A')) \quad \forall A' \in \mathcal{A}'.$$

We have seen already that μ^X is actually a measure on (Ω', \mathcal{A}') .

Theorem 1.19 (Transformation formula). Let $(\Omega, \mathcal{A}, \mu)$ be a measure space, (Ω', \mathcal{A}') a measurable space and $X : \Omega \rightarrow \Omega'$ a measurable map. Then:

- (i) If $f' : \Omega' \rightarrow \mathbb{R}$ is \mathcal{A}' - $\mathcal{B}(\mathbb{R})$ -measurable and non-negative, then

$$\int f' \, d\mu^X = \int f' \circ X \, d\mu.$$

- (ii) If $f' : \Omega' \rightarrow \mathbb{R}$ is finitely absolutely integrable w.r.t. μ^X , then so is $f' \circ X$ and vice-versa. In this case, we also have:

$$\int f' \, d\mu^X = \int f' \circ X \, d\mu.$$

1.5 Product spaces and Kolmogorov's consistency theorem

Definition 1.20. Let I be an arbitrary index set and $(\Omega_i)_{i \in I}$ a family of sets. We define

$$\Omega = \prod_{i \in I} \Omega_i := \left\{ \omega : I \rightarrow \bigcup_{i \in I} \Omega_i \mid \omega(i) \in \Omega_i \ \forall i \in I \right\}.$$

Its elements ω are sometimes written as $\omega = (\omega_i)_{i \in I}$. Let $\pi_i : \Omega \rightarrow \Omega_i$, $\omega \mapsto \omega_i$ be the coordinate map (i.e. the projection onto the i -th coordinate).

Definition 1.21. Let $(\Omega_i, \mathcal{A}_i)$ be a measurable space for every $i \in I$. The smallest σ -algebra over Ω with respect to which the function π_i is measurable $\forall i \in I$ is called the product- σ -algebra. This means that the product- σ -algebra is

$$\bigotimes_{i \in I} \mathcal{A}_i := \sigma \left(\pi_i^{-1}(A_i) \mid A_i \in \mathcal{A}_i \ \forall i \in I \right).$$

Theorem 1.22. Let $(\Omega_i, \mathcal{A}_i, \mu_i)$ be σ -finite measure spaces $\forall i \in \{1, \dots, n\}$ for some $n \in \mathbb{N}$. Then there exists a unique σ -finite measure $\bigotimes_{i=1}^n \mu_i$ on the product space

$$\left(\prod_{i=1}^n \Omega_i, \bigotimes_{i=1}^n \mathcal{A}_i \right)$$

with the property

$$\left(\bigotimes_{i=1}^n \mu_i \right) (A_1 \times \dots \times A_n) = \prod_{i=1}^n \mu_i(A_i)$$

for all sets $A_i \in \mathcal{A}_i \ \forall i \in \{1, \dots, n\}$. We call this measure the product measure.

Example. Let $\Omega_1 = \mathbb{R} = \Omega_2$, $\mathcal{A}_1 = \mathcal{B}(\mathbb{R}) = \mathcal{A}_2$, $\mu_1 = \lambda = \mu_2$. Then the product measure λ^2 satisfies $\lambda^2([a_1, b_1] \times [a_2, b_2]) = \lambda([a_1, b_1]) \cdot \lambda([a_2, b_2])$.

One nice property of the integral with respect to product measures is that it can be iteratively boiled down to integrals over the marginals.

Theorem 1.23 (Fubini). Let $(\Omega_1, \mathcal{A}_1, \mu_1)$ and $(\Omega_2, \mathcal{A}_2, \mu_2)$ be σ -finite measure spaces and $f : \Omega_1 \times \Omega_2 \rightarrow \mathbb{R}$ a $(\mathcal{A}_1 \otimes \mathcal{A}_2)$ - $\mathcal{B}(\mathbb{R})$ -measurable function. If f is non-negative or absolutely integrable, then the maps

$$\Omega_1 \rightarrow \overline{\mathbb{R}}, \omega_1 \mapsto \int_{\Omega_2} f(\omega_1, \omega_2) \, d\mu_2(\omega_2) \quad \text{and} \quad \Omega_2 \rightarrow \overline{\mathbb{R}}, \omega_2 \mapsto \int_{\Omega_1} f(\omega_1, \omega_2) \, d\mu_1(\omega_1)$$

are measurable, and

$$\int_{\Omega} f \, d\mu = \int_{\Omega_1} \left(\int_{\Omega_2} f(\omega_1, \omega_2) \, d\mu_2(\omega_2) \right) d\mu_1(\omega_1)$$

$$= \int_{\Omega_2} \left(\int_{\Omega_1} f(\omega_1, \omega_2) \, d\mu_1(\omega_1) \right) \, d\mu_2(\omega_2)$$

Definition 1.24. A topological space is called Polish, if there exists a complete metric inducing its topology and a countable base of the topology.

Example. A well-known example of a Polish space is the set of all continuous functions on the unit interval, $\mathcal{C}([0, 1])$, equipped with the topology of uniform convergence:

$$d_{\text{sup}} : \mathcal{C}([0, 1]) \times \mathcal{C}([0, 1]) \rightarrow \mathbb{R}, \quad (f, g) \mapsto \|f - g\|_{\infty} = \sup_{x \in [0, 1]} |f(x) - g(x)|.$$

Proposition 1.25 (Ulam's lemma). Let (Ω, τ) be a Polish space and $\mathcal{B}(\Omega)$ the Borel- σ -algebra on this topological space. Let μ be a probability measure on $(\Omega, \mathcal{B}(\Omega))$. Let $E \in \mathcal{B}(\Omega)$ and $\varepsilon > 0$. Then there exists a compact set $K \subset E$ with $\mu(E \setminus K) < \varepsilon$.

Definition 1.26. Let I be an index set, $(\Omega_i, \mathcal{A}_i)$ a measurable space for every $i \in I$ and

$$(\Omega, \mathcal{A}) := \left(\prod_{i \in I} \Omega_i, \bigotimes_{i \in I} \mathcal{A}_i \right).$$

A family of probability measures $\{\mu_J \mid J \subset I \text{ is finite}\}$ on

$$(\Omega_J, \mathcal{A}_J) := \left(\prod_{i \in J} \Omega_i, \bigotimes_{i \in J} \mathcal{A}_i \right)$$

is called projective family if for every finite set $J \subset I$ and every $K \subset J$, we have

$$\mu_K = \mu_J^{\pi_K^J} \quad (\text{image measure of } \mu_J \text{ under } \pi_K^J),$$

where $\pi_K^J : \Omega_J \rightarrow \Omega_K$, $(\omega_i)_{i \in J} \mapsto (\omega_i)_{i \in K}$ is the coordinate projection from Ω_J to Ω_K .

Theorem 1.27 (Consistency theorem of Kolmogorov). Suppose that we have Polish spaces (Ω_n, τ_n) for every $n \in \mathbb{N}$ and a projective family $\{\mu_J \mid J \subset \mathbb{N} \text{ is finite}\}$ where μ_J is a probability measure on

$$\left(\prod_{k \in J} \Omega_k, \bigotimes_{k \in J} \mathcal{B}(\tau_k) \right)$$

for all finite $J \subset \mathbb{N}$. Then, there exists a unique probability measure μ on

$$\left(\prod_{n \in \mathbb{N}} \Omega_n, \bigotimes_{n \in \mathbb{N}} \mathcal{B}(\tau_n) \right)$$

such that $\mu_J = \mu^{\pi_J}$ for the projections $\pi_J : \prod_{n \in \mathbb{N}} \Omega_n \rightarrow \prod_{n \in J} \Omega_n$, $(\omega_n)_{n \in \mathbb{N}} \mapsto (\omega_i)_{i \in J}$.

Definition 1.28. Let $(\Omega_1, \mathcal{A}_1)$ and $(\Omega_2, \mathcal{A}_2)$ be measurable spaces. A map $K : \Omega_1 \times \mathcal{A}_2 \rightarrow [0, 1]$ is called probability kernel if the following properties hold:

- (i) $\forall \omega_1 \in \Omega_1$, the map $K(\omega_1, \cdot)$ is a probability measure on $(\Omega_2, \mathcal{A}_2)$.
- (ii) $\forall A_2 \in \mathcal{A}_2$, the map $K(\cdot, A_2)$ is \mathcal{A}_1 - $\mathcal{B}([0, 1])$ -measurable (where $\mathcal{B}([0, 1])$ is the Borel- σ -algebra on $[0, 1]$).

Theorem 1.29. Let $(\Omega_1, \mathcal{A}_1, \mathbb{P}_1)$ be a probability space, $(\Omega_2, \mathcal{A}_2)$ a measurable space and $K : \Omega_1 \times \mathcal{A}_2 \rightarrow [0, 1]$ a probability kernel. Then, there exists a unique probability measure \mathbb{P} on the product space $(\Omega_1 \times \Omega_2, \mathcal{A}_1 \otimes \mathcal{A}_2)$ with

$$\mathbb{P}(A_1 \times A_2) = \int_{A_1} K(\omega_1, A_2) \, d\mathbb{P}_1(\omega_1) \quad \forall A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2.$$

Moreover, for every bounded, measurable function $f : \Omega_1 \times \Omega_2 \rightarrow \mathbb{R}$, we have:

$$\int f \, d\mathbb{P} = \int \underbrace{\int f(\omega_1, \omega_2) K(\omega_1, d\omega_2)}_{\text{bounded, measurable function from } \Omega_1 \text{ to } \mathbb{R}} \, d\mathbb{P}_1(\omega_1)$$

This is a generalization of Fubini's theorem.

1.6 Lebesgue decomposition and densities

Definition 1.30. Let (Ω, \mathcal{A}) be a measurable space and μ, ν measures on it.

- (i) μ is called (absolutely) continuous with respect to ν (notation: $\mu \ll \nu$) if for every $A \in \mathcal{A}$ with $\nu(A) = 0$, we also have $\mu(A) = 0$.
- (ii) μ and ν are called equivalent if $\nu \ll \mu$ and $\mu \ll \nu$.
- (iii) μ and ν are called singular (notation: $\mu \perp \nu$) if there exists a set $A \in \mathcal{A}$ with $\mu(A) = \nu(A^c) = 0$.
- (iv) ν has a density f with respect to μ if there exists a measurable function f with

$$\nu(A) = \int f \cdot \mathbb{1}_A \, d\mu = \int_A f \, d\mu.$$

In this case, we write:

$$f = \frac{d\nu}{d\mu}.$$

Proposition 1.31. Let ν be a σ -finite measure. If f_1 and f_2 both are densities of ν w.r.t. μ , then

$$\mu(\{f_1 \neq f_2\}) = 0.$$

Theorem 1.32 (Lebesgue-decomposition). *Let μ and ν be σ -finite measures on some measurable space (Ω, \mathcal{A}) . Then, ν possesses a unique decomposition*

$$\nu = \nu_a + \nu_s$$

where $\nu_a \ll \mu$, $\nu_s \perp \mu$, ν_a has a density w.r.t. μ and

$$\mu \left(\frac{d\nu_a}{d\mu} = \infty \right) = 0.$$

Corollary 1.33 (Theorem of Radon-Nikodym). *Let (Ω, \mathcal{A}) be a measurable space and μ, ν σ -finite measures on it. Then,*

$$\nu \ll \mu \iff \nu \text{ has a density w.r.t. } \mu.$$

2 Stochastic independence

From now on, we always assume $(\Omega, \mathcal{A}, \mathbb{P})$ is a probability space.

Definition 2.1. *A family of set systems $(\mathcal{C}_i)_{i \in I}$ with $\mathcal{C}_i \subset \mathcal{A} \ \forall i \in I$ is called stochastically independent if*

$$\mathbb{P} \left(\bigcap_{i \in S} C_i \right) = \prod_{i \in S} \mathbb{P}(C_i) \quad \forall \text{ finite } S \subset I \text{ and } C_i \in \mathcal{C}_i \ \forall i \in S.$$

Theorem 2.2. *Let $(\mathcal{C}_i)_{i \in I}$ be a stochastically independent family of set systems with $\mathcal{C}_i \subset \mathcal{A} \ \forall i \in I$, which are stable under intersection. Then, the family $(\sigma(\mathcal{C}_i))_{i \in I}$ is also stochastically independent.*

Proof. Let $S = \{i_1, \dots, i_n\} \subset I$ be finite and $l \in \{1, \dots, n\}$ fixed. We define:

$$\mathcal{D}_l = \left\{ A \in \mathcal{A} \mid \mathbb{P} \left(A \cap \bigcap_{j \neq l} C_{i_j} \right) = \mathbb{P}(A) \cdot \prod_{j \neq l} \mathbb{P}(C_{i_j}) \right. \\ \left. \text{for all } C_{i_j} \in \sigma(\mathcal{C}_{i_j}) \ \forall j < l \text{ and } C_{i_j} \in \mathcal{C}_{i_j} \ \forall j > l \right\}$$

We now want to prove that \mathcal{D}_l is a Dynkin system, which we will do by induction on l .

Initialization step: We show that \mathcal{D}_1 is a Dynkin system. Let $C_{i_j} \in \mathcal{C}_{i_j} \ \forall j > 1$.

(a) As \mathcal{C}_{i_j} are independent, we have:

$$\mathbb{P} \left(\Omega \cap \bigcap_{j \neq 1} C_{i_j} \right) = \mathbb{P} \left(\bigcap_{j \neq 1} C_{i_j} \right) = \prod_{j \neq 1} \mathbb{P}(C_{i_j})$$

Therefore, $\Omega \in \mathcal{D}_1$.

(b) Let $D \in \mathcal{D}_1$. We have:

$$D^c \cap \bigcap_{j \neq 1} C_{i_j} = \left(\Omega \cap \bigcap_{j > 1} C_{i_j} \right) \setminus \left(D \cap \bigcap_{j > 1} C_{i_j} \right)$$

As

$$\mathbb{P}\left(\Omega \cap \bigcap_{j > 1} C_{i_j}\right) = \mathbb{P}\left(\bigcap_{j > 1} C_{i_j}\right) = \prod_{j > 1} \mathbb{P}(C_{i_j}),$$

$$\mathbb{P}\left(D \cap \bigcap_{j > 1} C_{i_j}\right) = \mathbb{P}(D) \cdot \prod_{j > 1} \mathbb{P}(C_{i_j}),$$

and $D \cap \bigcap_{j > 1} C_{i_j} \subset \Omega \cap \bigcap_{j > 1} C_{i_j}$, we obtain:

$$\mathbb{P}\left(D^c \cap \bigcap_{j > 1} C_{i_j}\right) = (1 - \mathbb{P}(D)) \cdot \prod_{j > 1} \mathbb{P}(C_{i_j}) = \mathbb{P}(D^c) \cdot \prod_{j > 1} \mathbb{P}(C_{i_j}).$$

Therefore, $D^c \in \mathcal{D}_1$.

(c) Let $A_n \in \mathcal{D}_1$ be pairwise disjoint sets $\forall n \in \mathbb{N}$. We have:

$$\left(\bigcup_{n \in \mathbb{N}} A_n \right) \cap \left(\bigcap_{j > 1} C_{i_j} \right) = \bigcup_{n \in \mathbb{N}} \underbrace{\left(A_n \cap \bigcap_{j > 1} C_{i_j} \right)}_{\text{pairwise disjoint}}$$

$$\begin{aligned} \Rightarrow \mathbb{P}\left(\left(\bigcup_{n \in \mathbb{N}} A_n\right) \cap \left(\bigcap_{j > 1} C_{i_j}\right)\right) &= \sum_{n \in \mathbb{N}} \mathbb{P}\left(A_n \cap \bigcap_{j > 1} C_{i_j}\right) && \text{by } \sigma\text{-additivity} \\ &= \sum_{n \in \mathbb{N}} \mathbb{P}(A_n) \cdot \prod_{j > 1} \mathbb{P}(C_{i_j}) \\ &= \mathbb{P}\left(\bigcup_{n \in \mathbb{N}} A_n\right) \cdot \prod_{j > 1} \mathbb{P}(C_{i_j}) && \text{by } \sigma\text{-additivity} \end{aligned}$$

Therefore, $\bigcup_{n \in \mathbb{N}} A_n \in \mathcal{D}_1$.

As we have shown that \mathcal{D}_1 is a Dynkin system, and we know that it contains \mathcal{C}_{i_1} , we know that $\sigma(\mathcal{C}_{i_1}) = \mathcal{D}(\mathcal{C}_{i_1}) \subset \mathcal{D}_1$. The induction step follows analogously. \square

Definition 2.3. Let I be an arbitrary index set, $(\Omega, \mathcal{A}, \mathbb{P})$ a measure space and $(\Omega_i, \mathcal{A}_i)$ a measurable space for every $i \in I$. Let $X_i : \Omega \rightarrow \Omega_i$ be random variables for every $i \in I$. The family $(X_i)_{i \in I}$ is called stochastically independent if $(X_i^{-1}(\mathcal{A}_i))_{i \in I}$ is stochastically independent.

Theorem 2.4. *Let \mathcal{E}_i be a \cap -stable generator of \mathcal{A}_i for all $i \in I$. Then $(X_i)_{i \in I}$ is stochastically independent if $(X_i^{-1}(\mathcal{E}_i))_{i \in I}$ is stochastically independent.*

Proof. We know that $X_i^{-1}(B) \cap X_i^{-1}(C) = X_i^{-1}(B \cap C)$, so $X_i^{-1}(\mathcal{E}_i)$ is stable under intersection (for every $i \in I$). Thanks to theorem 2.2, it therefore suffices to show that

$$\sigma(X_i^{-1}(\mathcal{E}_i)) = X_i^{-1}(\underbrace{\sigma(\mathcal{E}_i)}_{=\mathcal{A}_i}).$$

The inclusion $\sigma(X_i^{-1}(\mathcal{E}_i)) \subset X_i^{-1}(\sigma(\mathcal{E}_i))$ is easy:

$$\mathcal{E}_i \subset \sigma(\mathcal{E}_i) \implies X_i^{-1}(\mathcal{E}_i) \subset \underbrace{X_i^{-1}(\sigma(\mathcal{E}_i))}_{\sigma\text{-algebra}} \implies \sigma(X_i^{-1}(\mathcal{E}_i)) \subset X_i^{-1}(\sigma(\mathcal{E}_i)).$$

To show the other inclusion, we define

$$\mathcal{D} = \{B \in \sigma(\mathcal{E}_i) \mid X_i^{-1}(B) \in \sigma(X_i^{-1}(\mathcal{E}_i))\}$$

and aim to show that \mathcal{D} is a Dynkin system over Ω_i .

(a) We have $\Omega_i \in \sigma(\mathcal{E}_i)$ and $X_i^{-1}(\Omega_i) = \Omega \in \sigma(X_i^{-1}(\mathcal{E}_i))$, so $\Omega_i \in \mathcal{D}$.

(b) Let $B \in \mathcal{D}$. We have

$$X_i^{-1}(B^c) = (X_i^{-1}(B))^c \in \sigma(X_i^{-1}(\mathcal{E}_i)),$$

so $B^c \in \mathcal{D}$.

(c) Let $B_n \in \mathcal{D}$ be pairwise disjoint sets $\forall n \in \mathbb{N}$. We have

$$X_i^{-1}\left(\bigcup_{n \in \mathbb{N}} B_n\right) = \bigcup_{n \in \mathbb{N}} X_i^{-1}(B_n) \in \sigma(X_i^{-1}(\mathcal{E}_i)),$$

so $\bigcup_{n \in \mathbb{N}} B_n \in \mathcal{D}$.

We deduce

$$\mathcal{E}_i \subset \mathcal{D} \subset \sigma(\mathcal{E}_i) \implies \mathcal{D}(\mathcal{E}_i) \subset \mathcal{D} \subset \sigma(\mathcal{E}_i) \xrightarrow{\text{Prop. 1.8}} \mathcal{D} = \sigma(\mathcal{E}_i)$$

and therefore, $X_i^{-1}(\sigma(\mathcal{E}_i)) \subset \sigma(X_i^{-1}(\mathcal{E}_i))$. □