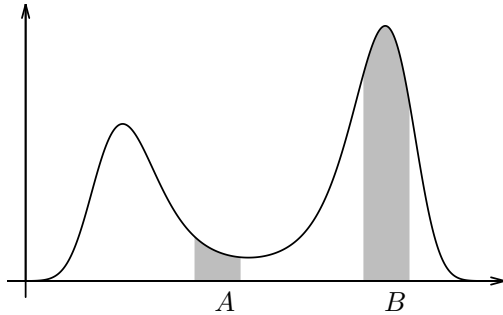


Measures with a density



Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a measure space and $f \in \mathcal{M}^+$ a function. Define

$$\nu(A) = \int_A f d\mu \quad \text{for all } A \in \mathbb{E}$$



.. p.1/22

Measures with a density



Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a measure space and $f \in \mathcal{M}^+$ a function. Define

$$\nu(A) = \int_A f d\mu \quad \text{for all } A \in \mathbb{E}$$

Lemma: ν is a measure on $(\mathcal{X}, \mathbb{E})$.

Proof: Check that $\nu(A) \in [0, \infty]$ and $\nu(\emptyset) = 0$ by the definition.

If A_1, A_2, \dots are pairwise disjoint \mathbb{E} -sets, then

$$\begin{aligned} \nu\left(\bigcup_{n=1}^{\infty} A_n\right) &= \int 1_{\bigcup_{n=1}^{\infty} A_n} f d\mu = \int \sum_{n=1}^{\infty} 1_{A_n} f d\mu \\ &\stackrel{\text{Cor. 6.20}}{=} \sum_{n=1}^{\infty} \int 1_{A_n} f d\mu = \sum_{n=1}^{\infty} \nu(A_n) \end{aligned}$$

We write $\nu = f \cdot \mu$ and say that ν has **density** (tæthed) f w.r.t. μ .

.. p.2/22

Restrictions



Example: Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a measure space and $A \in \mathbb{E}$. Let $\nu = 1_A \cdot \mu$.

We find that

$$\nu(B) = \int_B 1_A d\mu = \int 1_B 1_A d\mu = \mu(B \cap A)$$

That is, ν is the **restriction** of μ to A .

Point: The restriction (Exercise 2.3) of a measure μ to a measurable set $A \in \mathbb{E}$ is the measure with density 1_A w.r.t. μ .

.. p.3/22

Integration w.r.t. measures with a density



Theorem: Let $\nu = f \cdot \mu$ on $(\mathcal{X}, \mathbb{E})$. For all $g \in \mathcal{M}^+(\mathcal{X}, \mathbb{E})$ it holds that

$$\int g d\nu = \int g f d\mu$$

Proof: The standard scheme for these kinds of proof: Show that the result holds for

- 1) indicator functions
- 2) simple functions
- 3) \mathcal{M}^+ -functions

Point 3) is shown from 2) using monotone convergence.

.. p.4/22

Integration w.r.t. measures with a density



Theorem: Let $\nu = f \cdot \mu$ on $(\mathcal{X}, \mathbb{E})$. For a function $g \in \mathcal{M}(\mathcal{X}, \mathbb{E})$ it holds that g is ν -integrable if and only if

$$\int |g f| d\mu < \infty,$$

in which case

$$\int g d\nu = \int g f d\mu$$

Proof: First check the integrability condition using \mathcal{M}^+ -integration of $|g|$ w.r.t. $f \cdot \mu$.

Second apply the \mathcal{M}^+ -formula for g^+ - and g^- -integration.

.. p.5/22



Restrictions

Example: Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a measure space and $A \in \mathbb{E}$. The restriction, $\mu|_A$, of μ to A has density 1_A w.r.t. μ .

For $f \in \mathcal{M}^+$ we have that

$$\int f d\mu|_A = \int f 1_A d\mu = \int_A f d\mu$$

and if $f \in \mathcal{M}$ then f is integrable w.r.t. $\mu|_A$ if and only if

$$\int_A |f| d\mu < \infty$$

in which case

$$\int f d\mu|_A = \int_A f d\mu.$$

.. p.6/22

Familiar example



Consider $\nu = e^x \cdot m$ on (\mathbb{R}, \mathbb{B}) .

Then for $x \in \mathbb{R}$

$$\nu((-\infty, x]) = \int_{-\infty}^x e^y dm(y) = e^x.$$

We have seen this measure before – via transformations – now we have another explicit characterization of the measure; for all $A \in \mathbb{B}$

$$\nu(A) = \int_A e^y dm(y).$$

.. p.7/22



Successive densities

Theorem: Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a measure space, and let $f, g \in \mathcal{M}^+(\mathcal{X})$. Then

$$g \cdot (f \cdot \mu) = (fg) \cdot \mu$$

Proof: For $A \in \mathbb{E}$ it holds that

$$(g \cdot (f \cdot \mu))(A) = \int 1_A g d(f \cdot \mu) = \int (1_A g) f d\mu = \int_A (fg) d\mu$$

.. p.8/22

Uniqueness



Theorem: Let $(\mathcal{X}, \mathbb{E}, \mu)$ be a σ -finite measure spaces and let $f, g \in \mathcal{M}^+$. If $f \cdot \mu = g \cdot \mu$ then $f = g$ μ -almost everywhere.

Proof: Define

$$A = \{x \mid f(x) < g(x) < \infty\}, \quad B = \{x \mid g(x) < f(x) < \infty\},$$

$$C = \{x \mid f(x) < g(x) = \infty\}, \quad D = \{x \mid g(x) < f(x) = \infty\}.$$

and we show that they all have μ -measure 0.

□

.. p.9/22

Product measures



Corollary: If $\lambda = h \cdot (\mu \otimes \nu)$ then $\lambda = (f \cdot \mu) \otimes (g \cdot \nu)$ if and only if $h = f \otimes g$ $\mu \otimes \nu$ -a.e., that is there is a set $A \subset \mathcal{X} \times \mathcal{Y}$ such that for $(x, y) \in A$

$$h(x, y) = f(x)g(y)$$

and $A \subset B \in \mathbb{E}$ with $\mu \otimes \nu(B^c) = 0$, cf. also Exercise 11.6.

Important observation: If $C \in \mathbb{E} \otimes \mathbb{K}$ then the measure

$$(f \otimes g)1_C \cdot \mu \otimes \nu = (f \otimes g) \cdot \mu \otimes \nu|_C$$

is a product measure if $C = A \times B$ is a **product set** (up to a nullset).

.. p.11/22

Product measures



Let $(\mathcal{X}, \mathbb{E}, \mu)$ and $(\mathcal{Y}, \mathbb{K}, \nu)$ be two measure spaces and $f \in \mathcal{M}^+(\mathcal{X}, \mathbb{E})$ and $g \in \mathcal{M}^+(\mathcal{Y}, \mathbb{K})$.

Theorem: If $f \cdot \mu$ and $g \cdot \nu$ are σ -finite then

$$(f \cdot \mu) \otimes (g \cdot \nu) = (f \otimes g) \cdot (\mu \otimes \nu)$$

with $f \otimes g(x, y) = f(x)g(y)$.

Lemma: If $(\mathcal{X}, \mathbb{E}, \mu)$ is σ -finite and $f \in \mathcal{M}^+(\mathcal{X}, \mathbb{E})$ then $f \cdot \mu$ is σ -finite if and only if $f < \infty$ μ -a.e.

.. p.10/22

Discrete integration



Let \mathcal{X} be countable and τ the counting measure.

Let $p : \mathcal{X} \rightarrow [0, \infty]$ denote a map and consider the measure $\nu = p \cdot \tau$ on $(\mathcal{X}, \mathbb{P}(\mathcal{X}))$. Then

$$\nu(A) = \int_A p \, d\tau = \sum_{x \in A} p(x).$$

If \mathcal{X} is infinite and x_1, x_2, \dots is a bijective counting of the elements in \mathcal{X} we can write the sum as

$$\nu(A) = \lim_{n \rightarrow \infty} \sum_{i=1}^n 1_A(x_i) p(x_i).$$

.. p.12/22



Let \mathcal{X} be countable and τ the counting measure.

If $\nu = p \cdot \tau$ and $t : \mathcal{X} \mapsto \mathcal{Y}$ is a map we find for $y \in \mathcal{Y}$ that

$$t(\nu)(\{y\}) = \nu(t^{-1}(\{y\})) = \sum_{x \in t^{-1}(\{y\})} p(x).$$

Introducing

$$g(y) = \sum_{x \in t^{-1}(\{y\})} p(x)$$

we see that

$$t(\nu) = g \cdot \tau.$$

– p.13/22

Example



Consider $t : \{0, 1\}^n \rightarrow \{0, 1, \dots, n\}$ defined by

$$t(x_1, \dots, x_n) = \sum_{i=1}^n x_i.$$

With τ the counting measure on $\{0, 1\}^n$ we find for $m = 0, 1, \dots, n$ that

$$t(\tau)(\{m\}) = |t^{-1}(\{m\})| = \binom{n}{m} = \frac{n!}{m!(n-m)!}$$

Thus $t(\tau)$ has density

$$g(m) = \binom{n}{m}$$

w.r.t. the counting measure on $\{0, 1, \dots, n\}$.

– p.14/22



Let $(\mathcal{X}, \mathbb{E}, \mu)$ and $(\mathcal{Y}, \mathbb{K}, \xi)$ be two measure spaces. We think of μ and ξ as natural **base measures**.

Let $t : \mathcal{X} \rightarrow \mathcal{Y}$ be \mathbb{E} - \mathbb{K} -measurable.

The central problem: With $\nu = f \cdot \mu$ being a measure with density f on $(\mathcal{X}, \mathbb{E})$.

- 1) Does $t(\nu)$ have a density w.r.t. ξ ?
- 2) If so, how do we find the density?

Central question in **probability theory** as well as **statistics**.

– p.15/22

Special transformations



Theorem: If $\nu = (g \circ t) \cdot \mu$ on $(\mathcal{X}, \mathbb{E})$, and if $t : (\mathcal{X}, \mathbb{E}) \rightarrow (\mathcal{Y}, \mathbb{K})$ is measurable then

$$t(\nu) = g \cdot t(\mu).$$

Corollary: If $t : (\mathcal{X}, \mathbb{E}) \rightarrow (\mathcal{Y}, \mathbb{K})$ is bijective and **bi-measurable** and if $\nu = f \cdot \mu$ is a measure on $(\mathcal{X}, \mathbb{E})$, then

$$t(\nu) = f \circ t^{-1} \cdot t(\mu).$$

– p.16/22

Example



Recall that with $t: \{0, 1\}^n \rightarrow \{0, 1, \dots, n\}$ defined by $t(x_1, \dots, x_n) = \sum_{i=1}^n x_i$ then $t(\tau)$ has density $g(m) = \binom{n}{m}$ w.r.t. the counting measure on $\{0, 1, \dots, n\}$. If $\nu = (p \circ t) \cdot \tau$ on $\{0, 1\}^n$ then $t(\nu)$ has density

$$h(m) = \binom{n}{m} p(m)$$

w.r.t. τ on $\{0, 1, \dots, n\}$.

Classical example: n independent flips of a (skewed) coin are modelled by the probability measure on $\{0, 1\}^n$ with density

$$(x_1, \dots, x_n) \mapsto \gamma^{\sum_{i=1}^n x_i} (1 - \gamma)^{n - \sum_{i=1}^n x_i} = \gamma^{t(x_1, \dots, x_n)} (1 - \gamma)^{n - t(x_1, \dots, x_n)}$$

for $\gamma \in [0, 1]$. The model for the sum of heads, say, is the probability measure with density

$$\binom{n}{m} \gamma^m (1 - \gamma)^{n - m} \quad (\text{the binomial distribution}).$$

... p.17/22

Strategy for \mathbb{R}^k



Let $\nu = f \cdot m_k$ be a measure on $(\mathbb{R}^k, \mathbb{B}_k)$ with density f w.r.t. the Lebesgue measure m_k . Consider $t: \mathbb{R}^k \mapsto \mathbb{R}^k$. The strategy for finding $t(\nu)$ is as follows:

- Find disjoint $A_1, \dots, A_n \in \mathbb{B}_k$ with $\nu((\cup_i A_i)^c) = 0$ such that t is bijective and bi-measurable on each A_i .
- Observe that on A_i

$$t(\nu|_{A_i}) = (f \circ t^{-1})t(m_k|_{A_i}).$$

- Find $t(m_k|_{A_i})$.

Moral: If the strategy works, the transformation problem boils down to transformations of the Lebesgue measure.

... p.19/22

Marginalization



Lemma: Let $(\mathcal{X}, \mathbb{E}, \mu)$ and $(\mathcal{Y}, \mathbb{K}, \lambda)$ be two σ -finite measure spaces.

Let $f \in \mathcal{M}^+(\mathcal{X} \times \mathcal{Y})$ and consider $\nu = f \cdot \mu \otimes \lambda$ on $\mathcal{X} \times \mathcal{Y}$.

Then $\hat{X}(\nu) = g \cdot \mu$ with

$$g(x) = \int f(x, y) d\lambda(y) \quad \text{for } x \in \mathcal{X}$$

... p.18/22

Example



Consider the map $t(x) = a + Bx$ for B and $k \times k$, invertible matrix. The (global) inverse map is

$$t^{-1}(y) = B^{-1}(y - a) \quad \text{for } y \in \mathbb{R}^k$$

Due to the transformation results for the Lebesgue measure under affine transformations we have that $t(m_k) = |\det B^{-1}| m_k$, hence for $\nu = f \cdot m_k$ we have that $t(\nu) = g \cdot m_k$ where

$$g(y) = f(t^{-1}(y)) |\det B^{-1}| = f(B^{-1}(y - a)) |\det B^{-1}|.$$

For $k = 1$ this is

$$g(y) = f\left(\frac{y - a}{B}\right) \frac{1}{|B|} = f\left(\frac{y - a}{B}\right) |(t^{-1})'(y)|$$

and we say $t(\nu)$ has **location parameter** a and **scale parameter** b w.r.t. μ .

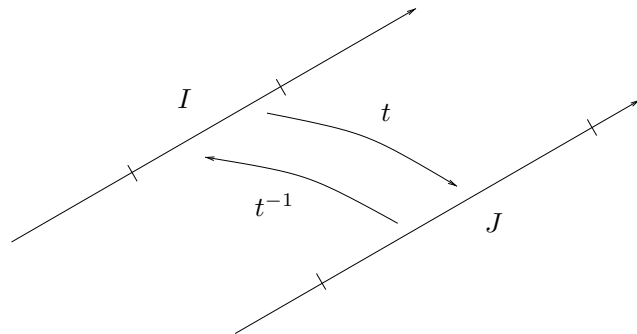
... p.20/22

Nice transformations



Consider a map $t : \mathbb{R} \rightarrow \mathbb{R}$ and open intervals I and J with

- 1) t maps I bijectively on J ,
- 2) t^{-1} is a C^1 -map on J .



... p.21/22

Transformation of the Lebesgue measure



Theorem: For $t : \mathbb{R} \rightarrow \mathbb{R}$ a Borel measurable map that fulfills that

- 1) t maps I bijectively on J ,
- 2) t^{-1} is a C^1 -map on J .

it holds that $t(m|_I) = g \cdot m$ where g is

$$g(y) = \begin{cases} |(t^{-1})'(y)| & \text{for } y \in J, \\ 0 & \text{for } y \notin J. \end{cases}$$

... p.22/22