

Probabilities II / Truncas

Integration on a product space

Theorem: Let $(X, \mathcal{A}), (Y, \mathcal{B})$ be two measurable spaces and $f: X \times Y \rightarrow \bar{\mathbb{R}}, \mathcal{A} \otimes \mathcal{B} / \mathcal{B}(\bar{\mathbb{R}})$ measurable. Then:

i) $\forall x \in X$ the partial (fixing) function $f_x(y): Y \rightarrow \bar{\mathbb{R}}$ with $f_x(y) = f(x, y)$ is $\mathcal{B} / \mathcal{B}(\bar{\mathbb{R}})$ measurable

ii) $\forall y \in Y$ the partial function $f^y(x): X \rightarrow \bar{\mathbb{R}}$ with $f^y(x) = f(x, y)$ is $\mathcal{A} / \mathcal{B}(\bar{\mathbb{R}})$ measurable

Tonelli Theorem (for positive functions)

Let $(X, \mathcal{A}, \mu), (Y, \mathcal{B}, \nu)$ be σ -finite measure spaces, $(X \times Y, \mathcal{A} \otimes \mathcal{B}, \mu \otimes \nu)$ is the product space and $f: X \times Y \rightarrow [0, \infty]$ measurable function, then the functions:

$$x \mapsto \int f(x, y) d\nu(y), x \in X \text{ and}$$

$$y \mapsto \int f(x, y) d\mu(x), y \in Y$$

are $\mathcal{A} / \mathcal{B}([0, \infty])$ and $\mathcal{B} / \mathcal{B}([0, \infty])$ measurable respectively

and

$$\int f(x, y) d(\mu \otimes \nu)(x, y) = \int \left(\int f(x, y) d\nu(y) \right) d\mu(x) = \int \left(\int f(x, y) d\mu(x) \right) d\nu(y)$$

The integrals are well defined, since we have non negative and measurable functions.

Fubini Theorem

Let $(X, \mathcal{A}, \mu), (Y, \mathcal{B}, \nu)$ be σ -finite measure spaces, $(X \times Y, \mathcal{A} \otimes \mathcal{B}, \mu \otimes \nu)$ the product space and $f: X \times Y \rightarrow \bar{\mathbb{R}}$ a measurable function. Then, if $\int |f(x,y)| d(\mu \otimes \nu)(x,y) < \infty$, then the results of Tonelli hold.

Corollary: When the conditions (Fubini) hold, or in any case with Tonelli,

$$i) \sum_{n \geq 1} \sum_{k \geq 1} a_{n,k} = \sum_{k \geq 1} \sum_{n \geq 1} a_{n,k} \quad (\text{apud } \eta \text{ zuro } \mu \otimes \nu)$$

$$ii) \int_{\mathbb{R}} \sum_{n \geq 1} g_n(x) dx = \sum_{n \geq 1} \int_{\mathbb{R}} g_n(x) dx$$

$$iii) E\left(\sum_{n \geq 1} X_n\right) = \sum_{n \geq 1} E(X_n)$$

$$iv) E\left[\int_{\mathbb{R}} f(x, \cdot) dx\right] = \int_{\mathbb{R}} E[f(x, \cdot)] dx, \text{ where } \forall x \in \mathbb{R} \text{ } f(x, \cdot) \text{ is a random variable}$$

ex. let $X \geq 0$ be a random variable. Then

$$\int_0^{\infty} P(X > t) dt = E(X).$$

Indeed,

$$\int_0^{\infty} P(X > t) dt = \int_0^{\infty} E(\underbrace{\mathbb{1}_{\{X > t\}}}_{f(t, \cdot)}) dt \stackrel{\text{Tonelli}}{=} E \int_0^{\infty} \mathbb{1}_{\{X > t\}} dt = E \left[\int_0^X 1 dt \right] = E(X)$$

Av $X \sim \text{Geo}(p)$ where $f(x) = p q^{x-1}$, $x=1, 2, \dots$. We know $E(X) = \frac{1}{p}$.

Compute:

$$\int_0^{\infty} P(X > t) dt \underset{\text{Geo}(p)}{=} \sum_{n=0}^{\infty} \int_n^{n+1} P(X > t) dt = \dots$$

Example: Let $X \geq 0$ be a random variable

a) If $g: [0, \infty) \rightarrow \mathbb{R}$ is differentiable with a continuous derivative on $(0, \infty)$, show that:

$$E[g(X)] = g(0) + \int_0^{\infty} g'(t) P(X > t) dt$$

if $g' \geq 0$ or if the integral converges absolutely

b) Show that

$$E(X^p) = p \int_0^{\infty} t^{p-1} P(X > t) dt, p > 0$$

Solution

$$a) g(x) = g(0) + g(x) - g(0) = g(0) + \int_0^x g'(t) dt \quad \forall x \geq 0.$$

So, $\forall \omega \in \Omega$, $g(X(\omega)) = g(0) + \int_0^{X(\omega)} g'(t) dt \Rightarrow$

$$g(X) = g(0) + \int_0^X g'(t) dt \xrightarrow[\text{conditions}]{\text{under the}}$$

$$\Rightarrow E[g(X)] = g(0) + E\left[\int_0^{+\infty} g'(t) \mathbb{1}_{\{X > t\}} dt\right]$$

$$= g(0) + \int_0^{+\infty} g'(t) E[\mathbb{1}_{\{X > t\}}] dt$$

$$= g(0) + \int_0^{+\infty} g'(t) P(X > t) dt$$

b) Set $g(x) = x^p$, $x \in [0, +\infty)$, $p > 0$. Then g is continuously differentiable on $(0, +\infty)$ and $g'(x) = p x^{p-1}$. So by (a)

$$E(X^p) = 0 + p \int_0^{+\infty} t^{p-1} P(X > t) dt.$$

Also note that; for $p=1$ and $p=2$

$$E(X) = \int_0^{+\infty} P(X > t) dt$$

$$E(X^2) = 2 \int_0^{+\infty} t P(X > t) dt$$

Product of Probability spaces (arbitrary number)

Definition: Let $\{(\Omega_i, \mathcal{A}_i, P_i)\}_{i \in I}$, $I \neq \emptyset$ be a family of probability spaces. Set

$$\Omega = \prod_{i \in I} \Omega_i = \{(\omega_i)_{i \in I} : \omega_i \in \Omega_i, \forall i \in I\}$$

$$= \{ \omega : I \rightarrow \bigcup_{i \in I} \Omega_i \mid \omega_i \in \Omega_i, \forall i \in I \}$$

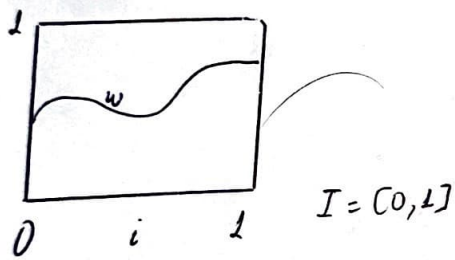
Any set $A \subset \Omega$, $A = \prod_{i \in I} A_i : A_i \in \mathcal{A}_i \forall i \in I$ and $J = \{i \in I : A_i \neq \Omega_i\}$ is finite, is called a measurable cylinder on Ω .

Set $\sigma\text{-}\mathcal{A}_i := \sigma(\{A \subset \Omega : A \text{ is a measurable cylinder}\})$ the σ -algebra generated by the measurable cylinders (the product σ -algebra)

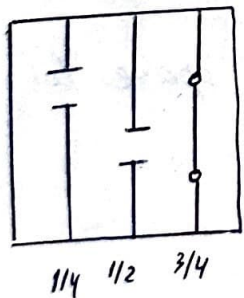
Example: $(\Omega_i, \mathcal{A}_i, P_i) = ([0, 1], \mathcal{B}([0, 1]), \mathcal{P}) \forall i \in I = [0, 1]$ (a continuous

sum of copies)

$$[0, 1]^{[0, 1]} : [0, 1] \rightarrow [0, 1]$$



An element ω of Ω is a function from $[0, 1]$ to $[0, 1]$. How to create a measurable cylinder?



$A = \prod_{i \in I} A_i$, where all A_i are $[0, 1]$ except for

$$J = \left\{ \frac{1}{4}, \frac{1}{2}, \frac{3}{4} \right\} \quad \text{and} \quad A_{1/4} = \left(\frac{2}{5}, \frac{3}{5} \right)$$

$$A_{1/2} = \left(\frac{1}{8}, \frac{1}{4} \right)$$

$$A_{3/4} = \left\{ \frac{1}{4}, \frac{1}{2} \right\}$$

ενδεχομένως

2) Take $\Omega = [0, 1]^{\mathbb{N}} = [0, 1] \times [0, 1] \times \dots \times [0, 1] \times \dots$

Each space is $(\Omega, \mathcal{B}(\Omega), \mathcal{P})$

Theorem: If for any measurable cylinder, we define

$$P(A) = \prod_{i \in J} P_i(A_i) = \prod_{i \in I} P_i(A_i)$$

because for $i \notin J$, $A_i = \Omega \Rightarrow P_i(\Omega_i) = 1$.

then P extends uniquely to a probability measure on the measure space $(\Omega = \prod_{i \in I} \Omega_i, \otimes_{i \in I} \mathcal{A}_i)$ and we denote P by $\otimes_{i \in I} P_i$

Example: Let us consider the random experiment of tossing a coin an infinite number of times where

$$P_p(\text{"tails"}) = p \quad \wedge \quad P_p(\text{"heads"}) = 1-p$$

\downarrow \downarrow
 πάτηρα \downarrow \downarrow
 κεφαλή

We seek to construct an appropriate probability space. Here:

$$(\Omega_i, \mathcal{A}_i, P) = (\{H, T\}, \mathcal{P}(\{H, T\}), P_p), \quad 0 < p < 1.$$

By using the above theorem, there exists a probability space for this experiment which is:

$$(\Omega, \mathcal{A}, P) = \left(\{H, T\}^{\mathbb{N}^*}, \otimes_{i \in \mathbb{N}^*} \mathcal{P}(\{H, T\}), \otimes_{i \in \mathbb{N}^*} P_p \right)$$

$\underbrace{\hspace{10em}}$
 $(\mathcal{A}, \mathcal{P}(\{H, T\}^{\mathbb{N}^*}))$
 HW

είναι η περίπτωση κεντρικού μωρι
 να άλλα προει να είναι ότι δέδο

example: $P(\text{"5th toss H"}, \text{"10th toss T"})$

$$= P_p("H") P_p("T") = (1-p)p$$

SOS (H, T, H, T, H, T, ...)

έχω αποφάσει μέχρι η δέση και άλλα
ευδαιμόνια

$A_n = "H, T, H, T, \dots, H, T"$

φθίνουσα αλφά ενδεχομένων

Αυτή έχω αποφάσει μέχρι $n+1$.

$$P(A_n) = p^{\frac{n}{2}} (1-p)^{\frac{n}{2}} \xrightarrow{n \rightarrow \infty} 0$$

Independence of Events

Independence of a family of events (ανεξαρτησία μιας οικογένειας ενδεχομένων)

Definition: Let (Ω, \mathcal{A}, P) be a probability space and $(A_i)_{i \in I}$ in \mathcal{A} a family of events. The events $(A_i)_{i \in I}$ are called independent if for every finite $J \subset I$,

$$P\left(\bigcap_{i \in J} A_i\right) = \prod_{i \in J} P(A_i)$$

Independence of a family of collections of events (οικογένεια συλλογών ενδεχομένων)

Let $(\mathcal{A}_i)_{i \in I}$ be a family of collections \mathcal{A}_i of events. The family $(\mathcal{A}_i)_{i \in I}$ is called independent, if for every finite $J \subset I$ and $A_i \in \mathcal{A}_i, i \in J$:

$$P\left(\bigcap_{i \in J} A_i\right) = \prod_{i \in J} P(A_i)$$

Example: Take A, B be two independent events ($A \perp B$). A independent of $B \Leftrightarrow P(A \cap B) = P(A) \cdot P(B)$

(Exercise in Prob I: $A \perp B \Leftrightarrow A \perp B^c \Leftrightarrow A^c \perp B \Leftrightarrow A^c \perp B^c$)

So if $\mathcal{A}_1 = \{A, A^c\}$, $\mathcal{A}_2 = \{B, B^c\}$, then \mathcal{A}_1 and \mathcal{A}_2 are independent collections of events.

Μπορεί να επεκταίνουμε σε \emptyset και Ω και να έχουμε ανεξ. σ -alg.

$\mathcal{A}_1' = \{\emptyset, A, A^c, \Omega\}$, $\mathcal{A}_2' = \{\emptyset, B, B^c, \Omega\}$ are independent

collections bcs each subcollection ανεξ από \emptyset και Ω και η προσαρτημένη τους δεν αλλάζει κάτι

$$P(A \cap \emptyset) = P(\emptyset) = 0 = P(\emptyset) P(A)$$

$$P(A \cap \Omega) = P(A) = 1 \cdot P(A) = P(\Omega) \cdot P(A)$$

So every event A is independent of \emptyset and Ω .