

36710 - 36752

ADVANCED PROBABILITY OVERVIEW

FALL 2020

LECTURE 11: WED, OCT 7, 2020

- LAST TIME: • PRODUCT σ -FIELDS, PRODUCT MEASURES AND TONELLI THEMS
FUBINI
- INDEPENDENCE

- IF X AND Y ARE RANDOM VARIABLES ON A COMMON PROBABILITY SPACE (Ω, \mathcal{F}, P) THEY ARE INDEPENDENT WHEN $\sigma(X)$ AND $\sigma(Y)$ ARE INDEPENDENT [FOR EACH $A \in \sigma(X)$ AND $B \in \sigma(Y)$, $P(A \cap B) = P(A) \cdot P(B)$]

DEF (OF MUTUAL INDEP. FOR A COLLECTION OF RV'S). LET (Ω, \mathcal{F}, P) . LET $\{(S_t, \mathcal{A}_t), t \in T\}$ BE MEASURE SPACES.
 \downarrow
ARBITRARY SET

LET $X_t: \Omega \rightarrow S_t$ BE $\mathcal{F}/\mathcal{A}_t$ -MEAS. FUNCTIONS.

(THESE ARE RV'S!) FOR EACH $t \in T$. THEN $\{X_t, t \in T\}$ ARE MUTUALLY INDEP. WHEN THEIR σ -FIELDS $\{\sigma(X_t), t \in T\}$ ARE MUTUALLY INDEPENDENT.

Thm LET (Ω, \mathcal{F}, P) BE A PROB. SPACE AND (S_1, \mathcal{A}_1) AND (S_2, \mathcal{A}_2) BE MEASURABLE SPACES. LET $X_k: \Omega \rightarrow S_k$ $k=1,2$ BE RANDOM VARIABLES AND SET $X = (X_1, X_2)$ THE DISTRIBUTION OF $X: \Omega \rightarrow (S_1, S_2)$, μ_X , IS THE PRODUCT MEASURE $\mu_{X_1} \otimes \mu_{X_2}$ IFF X_1 AND X_2 ARE INDEPENDENT.

Pf IF DIRECTION. ASSUME X_1 AND X_2 ARE INDEP. THEN FOR

ANY MEAS. RECT. $A_1 \times A_2$ $\xrightarrow{\quad} \{\omega: X_1(\omega) \in A_1\}$

$$\begin{aligned} \mu_X(A_1 \times A_2) &= P(\{X_1 \in A_1\} \cap \{X_2 \in A_2\}) \\ &= P(\{X_1 \in A_1\}) P(\{X_2 \in A_2\}) \\ &= \mu_{X_1}(A_1) \cdot \mu_{X_2}(A_2) \end{aligned}$$

FOR THE OTHER DIRECTION, ASSUME μ_X IS THE PRODUCT MEASURE, THEN

$$\begin{aligned} P(X_1 \in A_1, X_2 \in A_2) &= P(\{X_1 \in A_1\} \cap \{X_2 \in A_2\}) \\ &= \mu_X(A_1 \times A_2) \\ &= \mu_{X_1}(A_1) \cdot \mu_{X_2}(A_2) \\ &= P(\{X_1 \in A_1\}) P(\{X_2 \in A_2\}) \end{aligned}$$

EXTENSION TO MUTUAL INDEPENDENCE IS DIRECT.

INFINITE PRODUCT SPACES AND STOCHASTIC PROCESSES

Def Let (Ω, \mathcal{F}, P) be a probability space and T an arbitrary set. Suppose that for each $t \in T$ there exists a measurable space $(\mathcal{X}_t, \mathcal{F}_t)$ and a random quantity $X_t: \Omega \mapsto \mathcal{X}_t$. The collection $\{X_t, t \in T\}$ is called a stochastic process and T its index set.

REMARK $\omega \mapsto \{X_t(\omega), t \in T\}$

typically $\mathcal{X}_t = \mathbb{R}$ all t ($\mathcal{F}_t = \mathcal{B}^1$), $T = \mathbb{N}$ or $T = \mathbb{R}_{\geq 0}$

if $T = \{1, \dots, k\}$ and each X_1

is a R.V. (on a common prob. space!) the the random vector

(X_1, \dots, X_k) is a way to represent $\{X_t, t \in T\}$.

↓
DISCRETE TIME
PROCESS

↓
CONTINUOUS
TIME
PROCESS

EXAMPLE (RANDOM PROBABILITY MEASURE: MIXTURE).

Let $\Theta: \Omega \rightarrow \mathbb{R}^k$ be a random vector, with distribution μ_Θ

Let $f: \mathbb{R} \times \mathbb{R}^k \rightarrow \mathbb{R}_{\geq 0}$ be meas s.t. $\int_{\mathbb{R}} f(x, \underline{\theta}) dx = 1$

ALL $\underline{\theta} \in \mathbb{R}^k$.

[TAKE $k=1$ let μ_Θ be $N(0,1)$ AND

$$f(x, \theta) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left\{ -\frac{(x-\theta)^2}{2\sigma^2} \right\} \quad \text{some } \sigma > 0.]$$

↓
pdf of $N(\theta, \sigma^2)$

Let $B \in T = \mathcal{B}^1$ AND DEFINE

$$X_B(\omega) = \int_B f(x, \Theta(\omega)) dx$$

↳ LEBOGUE MEASURE.

ANOTHER EXAMPLE OF A RANDOM PROBABILITY MEASURE IS THE

EMPIRICAL MEASURE: X_1, X_2, \dots, X_n ARE INDEPENDENT RV'S

WITH COMMON DISTRIBUTION. LET P_n BE THE RANDOM PROB. MEASURE:

$$B \in \mathcal{B}^1 \mapsto P_n(B) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \in B\}}$$

↓

THEY ARE BOTH STOCHASTIC PROCESSES WITH $T = \mathcal{B}^1$ AND

$\mathcal{X}_t = [0, 1]$ ALL t .

Def (FUNCTIONAL REPRESENTATION OF PRODUCT SETS) LET $\mathcal{X} = \prod_{t \in T} \mathcal{X}_t$ → CARTESIAN PRODUCT

BE THE SET OF ALL FUNCTIONS $f: T \rightarrow \bigcup_{t \in T} \mathcal{X}_t$ SUCH THAT

FOR EVERY t , $f(t) \in \mathcal{X}_t$. WHEN $\mathcal{X}_t = \mathcal{Y}$ ALL t , WE

WRITE THIS AS \mathcal{Y}^T

EXAMPLE - THINK OF \mathbb{R}^k AS $\mathbb{R}^{\{1, \dots, k\}}$ AND OF A VECTOR

$(x_1, \dots, x_k) \in \mathbb{R}^k$ AS A FUNCTION S.T. $f(i) = x_i$ $i=1, \dots, k$.

• FOR THE RANDOM PROB. EXAMPLE $\mathcal{X} = [0, 1]^{\mathcal{B}^1}$ AND WE

CAN THINK OF EACH PROB. MEASURE ON $(\mathbb{R}, \mathcal{B}^1)$ AS A FUNCTION

FROM \mathcal{B}^1 INTO $[0, 1]$. WARNING THE PRODUCT SET \mathcal{X}

CONTAINS ALSO FUNCTIONS THAT ARE NOT PROBABILITIES. FOR

EXAMPLE THE FUNCTION $f(B) = 1$ ALL $B \in \mathcal{B}^1$.

KEY INTUITION: WE CAN NOW THINK OF A STOCHASTIC PROCESS AS

A RANDOM FUNCTION!

Def (COORDINATE PROJECTION AND CYLINDER SETS). LET $\mathcal{X} = \prod_{t \in T} \mathcal{X}_t$. FOR EACH $t \in T$ LET $\pi_t: \mathcal{X} \rightarrow \mathcal{X}_t$ BE DEFINED AS

$$\pi_t(f) = f(t)$$

THIS IS THE t -COORDINATE PROJECTION FUNCTION.

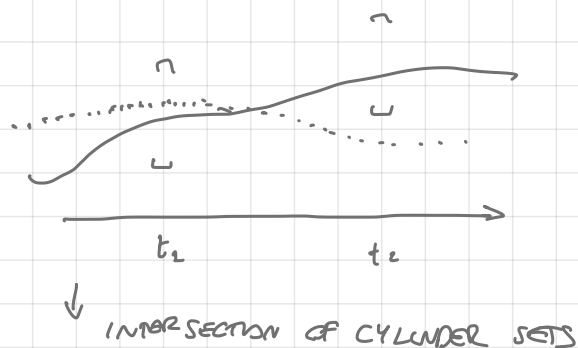
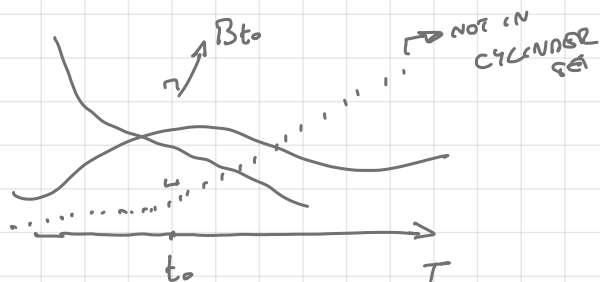
A ONE-DIMENSIONAL CYLINDER SET IS A SET OF THE FORM

$$\prod_{t \in T} B_t \quad \text{s.t.} \quad B_{t_0} \in \mathcal{F}_{t_0} \quad \text{SOME } t_0 \in T$$

$$B_t = \mathcal{X}_t \quad \text{ALL } t \neq t_0.$$

DEFINE ON \mathcal{X} THE PRODUCT σ -FIELD $\bigotimes_{t \in T} \mathcal{F}_t$ AS THE

σ -FIELD GENERATED BY ALL CYLINDERS.



Lemma 34 THE PRODUCT σ -FIELD IS THE SMALLEST σ -FIELD S.T. ALL COORDINATE PROJECTIONS ARE MEAS. FUNCTIONS.

Thm 35 LET (Ω, \mathcal{F}, P) BE PROB. SPACE, AND T BE A SET INDEXING MEASURABLE SPACE $\{(\mathcal{X}_t, \mathcal{F}_t), t \in T\}$ S.T. $X_t: \Omega \rightarrow \mathcal{X}_t$ IS A FUNCTION FOR ALL t . DEFINE $X: \Omega \rightarrow \mathcal{X} = \prod_t \mathcal{X}_t$ BY SETTING $X(\omega)$ TO BE THE FUNCTION $f \in \mathcal{X}$ DEFINED BY $f_t = X_t(\omega)$, ALL t . THEN X IS $\mathcal{F} / \bigotimes_t \mathcal{F}_t$ IFF EACH X_t IS $\mathcal{F} / \mathcal{F}_t$ MEAS.

$\hookrightarrow X$ IS A RANDOM FUNCTION DESCRIBING THE STOCHASTIC PROCESS

$\cdot X(\omega)$ IS A REALIZATION OR PATH OF THE PROCESS

HOW DO WE CONSTRUCT PROB. DISTRIBUTIONS OVER $(\mathcal{X}, \bigotimes_{t \in T} \mathcal{F}_t)$ TO

\downarrow HANDLE STOCHASTIC PROCESSES?

■ KOLMOGOROV EXTENSION THEOREM.

Def (FINITE DIMENSIONAL PROJECTIONS). T . LET $V \subset T$ FINITE, SO

$$V = \{t_1, t_2, \dots, t_n\}. \quad \text{LET } \mathcal{X}_V = \prod_{t \in V} \mathcal{X}_t \quad \text{AND, SIMILARLY,} \\ \hookrightarrow \text{FINITE PRODUCT}$$

LET \mathcal{F}_V BE THE PRODUCT σ -FIELD OVER \mathcal{X}_V . FOR ANY $U \subset V$

OF THE FORM $U = \{t_{i_1}, \dots, t_{i_m}\}$ $m < n$ AND FOR ANY

$$x = (x(t_1), \dots, x(t_n)) \in \mathcal{X}_V \quad \text{LET } x_U = (x(t_{i_1}), \dots, x(t_{i_m}))$$

BE THE CORRESPONDING SUB-VECTOR.

LET P_V IS A PROB. MEASURE ON $(\mathcal{X}_V, \mathcal{F}_V)$. THE PROJECTION

OF P_V ON $(\mathcal{X}_U, \mathcal{F}_U)$ IS THE PROB. MEASURE $\pi_U(P_V)$

GIVEN BY

$$\pi_U(P_V)(B) = P_V(\{x \in \mathcal{X}_V : x_U \in B\}) \quad B \in \mathcal{F}_U$$

SIMILARLY IF Q IS A PROB. DIST. OVER $(\prod_{t \in T} \mathcal{X}_t, \bigotimes_{t \in T} \mathcal{F}_t)$

THE PROJECTION OF Q ON $(\mathcal{X}_V, \mathcal{F}_V)$ IS

$$\pi_V(Q)(B) = Q(\{x \in \prod_{t \in T} \mathcal{X}_t : x_V \in B\}) \quad B \in \mathcal{F}_V$$