

## Product measure

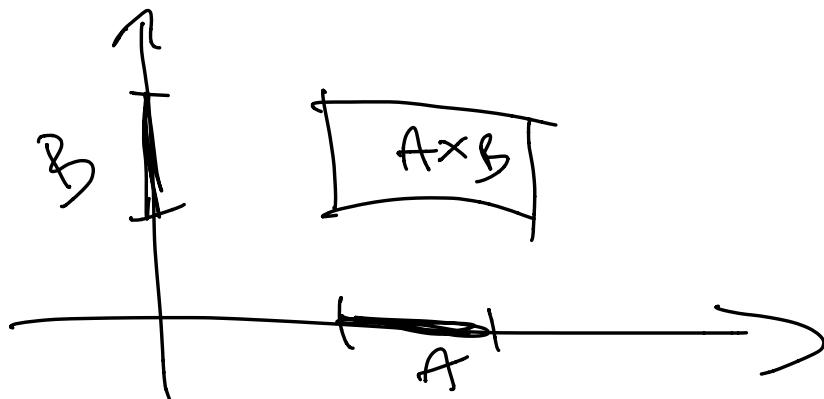
Suppose that  $(\Omega_1, \mathcal{F}_1, P_1)$ ,  $(\Omega_2, \mathcal{F}_2, P_2)$  are probability spaces.

Construct a new one with

$$\Omega = \Omega_1 \times \Omega_2$$

$\bar{\mathcal{F}}$  should contain all sets of the form  $A \times B$  with  $A \in \bar{\mathcal{F}}_1$ ,  $B \in \bar{\mathcal{F}}_2$

$$P(A \times B) = P_1(A) \times P_2(B)$$

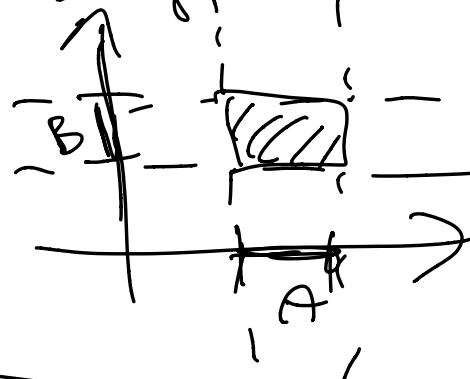


Th: With  $\bar{\mathcal{F}} = \sigma(\mathcal{F}_1 \times \mathcal{F}_2)$

(this is the  $\sigma$ -field generated by sets of the form  $A \times B$  with  $A \in \bar{\mathcal{F}}_1$ ,  $B \in \bar{\mathcal{F}}_2$ )

there is a unique  $R$  on  $(\Omega, \mathcal{F})$   
 with  $R(A \times B) = R_1(A) R_2(B)$ .  $\Omega_1 \times \Omega_2$

Proof: The sets  $A \times B$  form a  
 semi-algebra so  
 we can use the  
 previous sum.



You can iterate this to define  
 the product of finitely many  
 probability spaces.

Soon: infinitely many probability spaces

Theorem (Fubini)  $(\Omega_1, \mathcal{F}_1, R_1), (\Omega_2, \mathcal{F}_2, R_2)$

$f: \Omega_1 \times \Omega_2 \rightarrow \mathbb{R}$  measurable

If  $f \geq 0$  or  $\int |f| d(R_1 \times R_2) < \infty$

$$\int_{\Omega_1 \times \Omega_2} f d(R_1 \times R_2) = \int_{\Omega_2} \left( \int_{\Omega_1} f dR_1 \right) dR_2 = \int_{\Omega_1} \left( \int_{\Omega_2} f dR_2 \right) dR_1 = \int_{\Omega_1} f dR_1$$

# Independence

Def:  $(\Omega, \mathcal{F}, P)$  is a prob space

1)  $A, B \in \mathcal{F}$  are independent if

$$P(A \cap B) = P(A) P(B)$$

2, Two random variables  $X, Y : \Omega \rightarrow \mathbb{R}$   
are independent if  $\{X \in A\}, \{Y \in B\}$   
are independent for any  $A, B$  Borel set.

$$P(X \in A, Y \in B) = P(X \in A) P(Y \in B)$$

3, Two  $\sigma$ -fields  $\mathcal{A}, \mathcal{B} \subseteq \mathcal{F}$  are independent  
if all pairs  $A \in \mathcal{A}, B \in \mathcal{B}$  are independent.

## Simple observations

- $X$  and  $Y$  are independent if and only if  $\sigma(X)$  and  $\sigma(Y)$  are independent.
- if  $\mathcal{A}$  and  $\mathcal{B}$  are independent and  $X \in \mathcal{A}, Y \in \mathcal{B}$  then  $X$  and  $Y$  are independent
- $A, B$  are independent  $\Leftrightarrow 1_A$  and  $1_B$  are independent

II

$\mathcal{G}(\{A\})$  and  $\mathcal{G}(\{B\})$  are independent  
 " "  
 $\{A, A^c, \phi, S\}$

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Ex:  $(\Omega_1, \mathcal{F}_1, P_1)$ ,  $(\Omega_2, \mathcal{F}_2, P_2)$  are prob spaces  
 Consider the product prob. space on  $\Omega_1 \times \Omega_2$ .

↳ If  $A \in \mathcal{F}_1$ ,  $B \in \mathcal{F}_2$  then  $A \times \Omega_2$ ,  $\Omega_1 \times B$   
 are independent

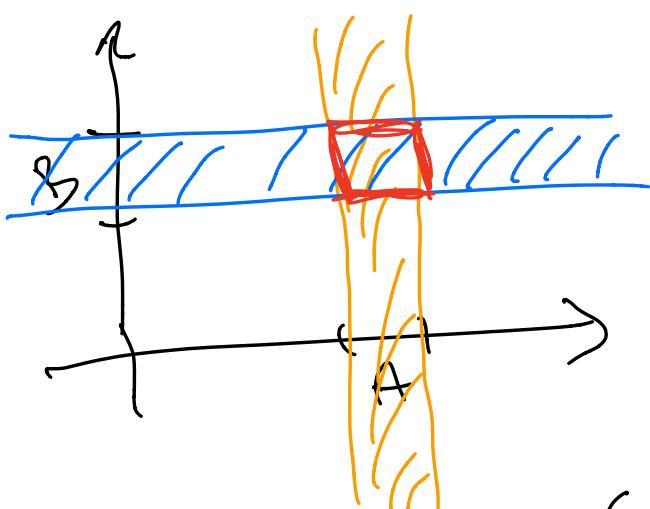
$$P(A \times \Omega_2) = P_1(A) P_2(\Omega_2) = P_1(A)$$

$$P(\Omega_1 \times B) = P_2(B)$$

$$P(A \times \Omega_2 \cap \Omega_1 \times B) = P(A \times B)$$

$$= P_1(A) P_2(B)$$

$$= P(A \times \Omega_2) P(\Omega_1 \times B)$$



2, The  $\mathcal{G}$ -fields  $\{A \times \Omega_2 : A \in \mathcal{F}_1\}$ ,

$\{\Omega_1 \times B : b \in \mathcal{F}_2\}$  are independent

3, If  $X, Y : \Omega_1 \times \Omega_2 \rightarrow \mathbb{R}$  are random variables so that  $X$  only depends on the first coordinate and  $Y$  only on the 2nd then  $X$  and  $Y$  are independent.

Definition: We say that  $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n \subset \mathcal{F}$  are independent if for every  $A_i \in \mathcal{F}_i$  we have  $P(\bigcap_{i=1}^n A_i) = \prod_{i=1}^n P(A_i)$

Random variables  $X_1, X_2, \dots, X_n$  are independent if the generated  $\sigma$ -fields are independent.

Events  $A_1, A_2, \dots, A_n$  are independent if the generated  $\sigma$ -fields are independent.

Claim:  $A_1, A_2, \dots, A_n$  are independent if and only if 1, for any  $I \subset \{1, \dots, n\}$  we have

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

$2^{n-1}$  equations

$$2) P\left(\bigcap_{i=1}^n A_i^*\right) = \prod_{i=1}^n P(A_i^*)$$

$$A_i^* : A_i \text{ or } A_i^c \quad \text{2 equations}$$


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Q: How can we construct a prob space with independent random variables with given distributions?

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Pairwise independence of events:

$A_1, A_2, \dots, A_n$  are pairwise independent if for any  $i \neq j$   $A_i$  and  $A_j$  are independent.

This is weaker than the "full" independence

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Example: Suppose that  $X_1, X_2, X_3$  are independent Bernlli( $\frac{1}{2}$ ) random variables (independent coin flips)

Then the events  $A_{ij} = \{X_i = X_j\}$  ( $i \neq j$ ) are pairwise independent, but not independent.

$A_{12}, A_{13}, A_{23}$

Ex: Suppose that  $A_1, A_2, \dots, A_\varepsilon$  are finite subsets of  $\mathbb{R}$ . Consider the probability space of choosing a uniformly chosen element of  $A_1 \times A_2 \times \dots \times A_\varepsilon$ , and let  $X_j$  be the  $j^{\text{th}}$  chosen number. Then  $X_1, X_2, \dots, X_\varepsilon$  are independent.

If we consider the prob space of choosing uniformly from  $A_j$ , then the product of these probability spaces is choosing uniformly from  $A_1 \times \dots \times A_\varepsilon$ .

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Q: How can we construct a prob space with independent random variables with given distributions?

For any distribution we realize that on  $(\Omega, \mathcal{B}, \mathbb{P})$  as a random variable.

## Sufficient conditions for independence

Def: Collection of sets  $A_1, \dots, A_n \subseteq \mathcal{F}$  are independent if for any subset  $I \subset \{1, \dots, n\}$

$$A_i \in \mathcal{A}_i \text{ then } P\left(\bigcap_{i \in I} A_i\right) = \prod_{i \in I} P(A_i).$$

(if  $\Omega \in \mathcal{A}_i$  for each  $i \leq n$  then this is equivalent to  $P\left(\bigcap_{i=1}^n A_i\right) = \prod_{i=1}^n P(A_i)$ .)

Then: Suppose that a collection of sets  $A_1, \dots, A_n$  are independent, and each one is closed for intersection. Then the generated  $\sigma$ -fields  $\mathcal{G}(A_1), \mathcal{G}(A_2), \dots, \mathcal{G}(A_n)$  are also independent.

Def:  $\pi$ -system: closed for intersection  
 $\mathcal{B}$  is a  $\lambda$ -system:  $\Omega \in \mathcal{B}$ , if  $A \in \mathcal{B} \in \mathcal{B}$   
if  $B_n \in \mathcal{B}$  with  $B_n \uparrow B$  then  $B \cap A \in \mathcal{B}$   
then  $B \in \mathcal{B}$ .

Lemma: If  $\mathcal{A}$  is a  $\pi$ -system and  $\mathcal{A} \subset \mathcal{B}$  is

a  $\lambda$ -system then  $\sigma(\mathcal{A}) \subset \mathcal{B}$ .

Outline of the proof of the theorem using the lemma. Assume that  $\mathcal{S} \in \mathcal{A}_n$ .

Fix  $A_i \in \mathcal{A}_i$ ,  $i \geq 2$ . Let  $F = \bigcap_{i=2}^n A_i$ .

$$\mathcal{B} \stackrel{\text{def}}{=} \left\{ A \in \mathcal{F} : P(AF) = P(A)P(F) \right\}$$

Check:  $\mathcal{B}$  is a  $\lambda$ -system

$\sigma(\mathcal{A}_n) \subseteq \mathcal{B}$  which means that

$\sigma(\mathcal{A}_1), \mathcal{A}_2, \mathcal{A}_3, \dots, \mathcal{A}_n$  are

independent. Now iterate this to

get that  $\sigma(\mathcal{A}_1), \sigma(\mathcal{A}_2), \dots, \sigma(\mathcal{A}_n)$

are independent.

Lemma:  $X_1, X_2, \dots, X_n$  random variables  
are independent if and only if

$$P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n) = \prod_{i=1}^n P(X_i \leq x_i)$$

for all choices of  $x_i \in (-\infty, \infty]$

Joint CDF

Product of the  
individual CDFs