

# Measure Theory

Jesús Fernández-Villaverde  
Duke University

## Why Bother with Measure Theory?

- Kolmogorov (1933).
- Foundation of modern probability.
- Deals easily with:
  1. Continuous versus discrete probabilities. Also mixed probabilities.
  2. Univariate versus multivariate.
  3. Independence.
  4. Convergence.

## Introduction to Measure Theory

- Measure theory is an important field for economists.
- We cannot do in a lecture what it will take us (at least) a whole semester.
- Three sources:
  1. Read chapters 7 and 8 in SLP.
  2. Excellent reference: *A User's Guide to Measure Theoretic Probability*, by David Pollard.
  3. Take math classes!!!!!!!!!!!!!!!!!!!!

## $\sigma$ -Algebra

- Let  $S$  be a set and let  $\mathcal{S}$  be a family of subsets of  $S$ .  $\mathcal{S}$  is a  $\sigma$ -algebra if
  1.  $\emptyset, S \in \mathcal{S}$ .
  2.  $A \in \mathcal{S} \Rightarrow A^c = S \setminus A \in \mathcal{S}$ .
  3.  $A_n \in \mathcal{S}, n = 1, 2, \dots, \Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{S}$ .
- $(S, \mathcal{S})$ : measurable space.
- $A \in \mathcal{S}$ : measurable set.

## Borel Algebra

- Define a collection  $\mathcal{A}$  of subsets of  $S$ .
- $\sigma$ -algebra generated by  $\mathcal{A}$ : the intersection of all  $\sigma$ -algebra containing  $\mathcal{A}$  is a  $\sigma$ -algebra.
- $\sigma$ -algebra generated by  $\mathcal{A}$  is the smallest  $\sigma$ -algebra containing  $\mathcal{A}$ .
- Example: let  $\mathcal{B}$  be the collection of all open balls (or rectangles) of  $\mathbb{R}^l$  (or a restriction of).
- Borel algebra: the  $\sigma$ -algebra generated by  $\mathcal{B}$ .
- Borel set: any set in  $\mathcal{B}$ .

## Measures

- Let  $(S, \mathcal{S})$  be a measurable space.
- Measure: an extended real-valued function  $\mu : \mathcal{S} \rightarrow \mathbb{R}_\infty$  such that:
  1.  $\mu(\emptyset) = 0$ .
  2.  $\mu(A) \geq 0, \forall A \in \mathcal{S}$ .
  3. If  $\{A_n\}_{n=1}^\infty$  is a countable, disjoint sequence of subsets in  $\mathcal{S}$ , then
$$\mu\left(\bigcup_{n=1}^\infty A_n\right) = \sum_{n=1}^\infty \mu(A_n).$$
- If  $\mu(S) < \infty$ , then  $\mu$  is finite.
- $(S, \mathcal{S}, \mu)$ : measurable space.

## Probability Measures

- Probability measure:  $\mu$  such that  $\mu(S) = 1$ .
- Probability space:  $(S, \mathcal{S}, \mu)$  where  $\mu$  is a probability measure.
- Event: each  $A \in \mathcal{S}$ .
- Probability of an event:  $\mu(A)$ .
- $B(S, \mathcal{S})$ : space all probability measures on  $(S, \mathcal{S})$ .

## Almost Everywhere

- Given  $(S, \mathcal{S}, \mu)$ , a proposition holds almost  $\mu$ -everywhere ( $\mu$ -a.e.), if  $\exists$  a set  $A \in \mathcal{S}$  with  $\mu(A) = 0$ , such that the proposition holds on  $A^c$ .
- If  $\mu$  is a probability measure, we often use the phrase almost surely (a.s.) instead of almost everywhere.

## Completion

- Let  $(S, \mathcal{S}, \mu)$  be a measure space.

- Define the family of subsets of any set with measure zero:

$$\mathcal{C} = \{C \subset S : C \subseteq A \text{ for some } A \in \mathcal{S} \text{ with } \mu(A) = 0\}$$

- Completion of  $\mathcal{S}$  is the family  $\mathcal{S}'$ :

$$\mathcal{S}' = \{B' \subseteq S : B' = (B \cup C_1) \setminus C_2, B \in \mathcal{S}, C_1, C_2 \in \mathcal{C}\}$$

- $\mathcal{S}'(\mu)$ : completion of  $\mathcal{S}$  with respect to measure  $\mu$ .

## Universal $\sigma$ -Algebra

- $\mathcal{U} = \bigcap_{\mu \in B(S, \mathcal{S})} \mathcal{S}'(\mu)$ .
- Note:
  1.  $\mathcal{U}$  is a  $\sigma$ -algebra.
  2.  $\mathcal{B} \subset \mathcal{U}$ .
- Universally measurable space is a measurable space with its universal  $\sigma$ -algebra.
- Universal  $\sigma$ -algebras avoid a problem of Borel  $\sigma$ -algebras: projection of Borel sets are not necessarily measurable with respect to  $\mathcal{B}$ .

## Measurable Function

- Measurable function into  $\mathbb{R}$ : given a measurable space  $(S, \mathcal{S})$ , a real-valued function  $f : S \rightarrow \mathbb{R}$  is measurable with respect to  $\mathcal{S}$  (or  $\mathcal{S}$ -measurable) if

$$\{s \in \mathcal{S} : f(s) \leq a\} \in \mathcal{S}, \forall a \in \mathbb{R}$$

- Measurable function into a measurable space: given two measurable spaces  $(S, \mathcal{S})$  and  $(T, \mathcal{T})$ , the function  $f : S \rightarrow T$  is measurable if:

$$\{s \in \mathcal{S} : f(s) \in A\} \in \mathcal{S}, \forall A \in \mathcal{T}$$

- If we set  $(T, \mathcal{T}) = (\mathbb{R}, \mathcal{B})$ , the second definition nests the first.
- Random variable: a measurable function in a probability space.

## Measurable Selection

- Measurable selection: given two measurable spaces  $(S, \mathcal{S})$  and  $(T, \mathcal{T})$  and a correspondence  $\Gamma$  of  $S$  into  $T$ , the function  $h : S \rightarrow T$  is a measurable selection from  $\Gamma$  if  $h$  is measurable and:

$$h(s) \in \Gamma(s), \forall s \in S$$

- Measurable Selection Theorem: Let  $S \subseteq \mathbb{R}^l$  and  $T \subseteq \mathbb{R}^m$  and  $\mathcal{S}$  and  $\mathcal{T}$  be their universal  $\sigma$ -algebras. Let  $\Gamma: S \rightarrow T$  be a (nonempty) compact-valued and u.h.c. correspondence. Then,  $\exists$  a measurable selection from  $\Gamma$ .

## Measurable Simple Functions

- $M(S, \mathcal{S})$ : space of measurable, extended real-valued functions on  $S$ .
- $M^+(S, \mathcal{S})$ : subset of nonnegative functions.
- Measurable simple function:

$$\phi(s) = \sum_{i=1}^n a_i \chi_{A_i}(s)$$

- Importance: for any measurable function  $f$ ,  $\exists \{\phi_n\}$  such that  $\phi_n(s) \rightarrow f$  pointwise.

## Integrals

- Integral of  $\phi$  with respect to  $\mu$ :

$$\int_S \phi(s) \mu(ds) = \sum_{i=1}^n a_i \mu(A_i)$$

- Integral of  $f \in M^+(S, \mathcal{S})$  with respect to  $\mu$ :

$$\int_S f(s) \mu(ds) = \sup_{\phi(s) \in M^+(S, \mathcal{S})} \int_S \phi(s) \mu(ds)$$

such that  $0 \leq \phi \leq f$ .

- Integral of  $f \in M^+(S, \mathcal{S})$  over  $A$  with respect to  $\mu$ :

$$\int_A f(s) \mu(ds) = \int_S f(s) \chi_A(s) \mu(ds)$$

## Positive and Negative Parts

- We define the previous results with positive functions.
- How do we extend to the general case?
- $f^+$ : positive part of a function

$$f^+(s) = \begin{cases} f(s) & \text{if } f(s) \geq 0 \\ 0 & \text{if } f(s) < 0 \end{cases}$$

- $f^-$ : negative part of a function

$$f^-(s) = \begin{cases} -f(s) & \text{if } f(s) \leq 0 \\ 0 & \text{if } f(s) > 0 \end{cases}$$

## Integrability

- Let  $(S, \mathcal{S}, \mu)$  be a measure space and let  $f$  be measurable, real-valued function on  $S$ . If  $f^+$  and  $f^-$  both have finite integrals with respect to  $\mu$ , then  $f$  is integrable and the integral is given by

$$\int f d\mu = \int f^+ d\mu - \int f^- d\mu$$

- If  $A \in \mathcal{S}$ , the integral of  $f$  over  $A$  with respect to  $\mu$ :

$$\int_A f d\mu = \int_A f^+ d\mu - \int_A f^- d\mu$$

## Transition Functions

- Transition function: given a measurable space  $(Z, \mathcal{Z})$ , a function  $Q : Z \times \mathcal{Z} \rightarrow [0, 1]$  such that:

1. For  $\forall z \in Z$ ,  $Q(z, \cdot)$  is a probability measure on  $(Z, \mathcal{Z})$ .

2. For  $\forall A \in \mathcal{Z}$ ,  $Q(\cdot, A)$  is  $\mathcal{Z}$ -measurable.

- $(Z^t, \mathcal{Z}^t) = (Z \times \dots \times Z, \mathcal{Z} \times \dots \times \mathcal{Z})$  ( $t$  times).

- Then, for any rectangle  $B = A_1 \times \dots \times A_t \in \mathcal{Z}^t$ , define:

$$\mu^t(z_0, B) = \int_{A_1} \dots \int_{A_{t-1}} \int_{A_t} Q(z_{t-1}, dz_t) Q(z_{t-2}, dz_{t-1}) \dots Q(z_0, dz_1)$$

## Two Operators

- For any  $\mathcal{Z}$ -measurable function  $f$ , define:

$$(Tf)(z) = \int f(z') Q(z, dz'), \quad \forall z \in \mathcal{Z}$$

Interpretation: expected value of  $f$  next period.

- For any probability measure  $\lambda$  on  $(\mathcal{Z}, \mathcal{Z})$ , define:

$$(T^*\lambda)(A) = \int Q(z, A) \lambda(dz), \quad \forall A \in \mathcal{Z}$$

Interpretation: probability that the state will be in  $A$  next period.

## Basic Properties

- $T$  maps the space of bounded  $\mathcal{Z}$ -measurable functions,  $B(Z, \mathcal{Z})$ , into itself.
- $T^*$  maps the space of probability measures on  $(Z, \mathcal{Z})$ ,  $\Lambda(Z, \mathcal{Z})$ , into itself.
- $T$  and  $T^*$  are adjoint operators:

$$\int (Tf)(z) \lambda(dz) = \int f(z') (T^*\lambda)(dz'), \quad \forall \lambda \in \Lambda(Z, \mathcal{Z})$$

for any function  $f \in B(Z, \mathcal{Z})$ .

## Two Properties

- A transition function  $Q$  on  $(Z, \mathcal{Z})$  has the Feller property if the associated operator  $T$  maps the space of bounded continuous function on  $Z$  into itself.
- A transition function  $Q$  on  $(Z, \mathcal{Z})$  is monotone if for every nondecreasing function  $f$ ,  $Tf$  is also non-decreasing.

## Consequences of our Two Properties

- If  $Z \subset \mathbb{R}^l$  is compact and  $Q$  has the Feller property, then  $\exists$  a probability measure  $\lambda^*$  that is invariant under  $Q$ :

$$\lambda^* = (T^* \lambda^*)(A) = \int Q(z, A) \lambda^*(dz)$$

- Weak convergence: a sequence  $\{\lambda_n\}$  converges weakly to  $\lambda$  ( $\lambda_n \Rightarrow \lambda$ ) if

$$\lim_{n \rightarrow \infty} \int f d\lambda_n = \int f d\lambda, \forall f \in C(S)$$

- If  $Q$  is monotone, has the Feller property, and there is enough “mixing” in the distribution, there is a unique invariant probability measure  $\lambda^*$ , and  $T^{*n} \lambda_0 \Rightarrow \lambda^*$  for  $\forall \lambda_0 \in \Lambda(Z, \mathcal{Z})$ .