## Probability and Random Processes Lecture 4

General integration theory

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## Measurable Extended Real-valued Functions

- R\* = the extended real numbers; a subset O ⊂ R\* is open if it can be expressed as a countable union of intervals of the form (a, b), [-∞, b), (a, ∞]
- A measurable space (Ω, A); an extended real-valued function
  f : Ω → ℝ\* is measurable if f<sup>-1</sup>(O) ⊂ A for all open O ⊂ ℝ\*
- A sequence  $\{f_n\}$  of measurable extended real-valued functions: for any x,  $\limsup f_n(x)$  and  $\liminf f_n(x)$  are measurable  $\Rightarrow$  if  $f_n \rightarrow g$  pointwise, then g is measurable
  - Hence, with the definition above, e.g.

$$f_n(x) = \frac{n}{\sqrt{2\pi}} \exp\left(-\frac{(nx)^2}{2}\right)$$

converges to a measurable function on  $(\mathbb{R}, \mathcal{B})$  or  $(\mathbb{R}, \mathcal{L})$ 

#### Measurable Simple Function

• An  $\mathcal{A}$ -measurable function s is a simple function if its range is a finite set  $\{a_1, \ldots, a_n\}$ . With  $A_k = \{x : s(x) = a_k\}$ , we get

$$s(x) = \sum_{k=1}^{n} a_k \chi_{A_k}(x)$$

(since s is measurable,  $A_k \in \mathcal{A}$ )

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# Integral of a Nonnegative Simple Function

• A measure space  $(\Omega, \mathcal{A}, \mu)$  and  $s : \Omega \to \mathbb{R}$  a nonnegative simple function which is  $\mathcal{A}$ -measurable, represented as

$$s(x) = \sum_{k=1}^{n} a_k \chi_{A_k}(x)$$

The integral of s over  $\Omega$  with respect to  $\mu$  is defined as

$$\int s(x)d\mu(x) = \sum_{k=1}^{n} a_k \mu(A_k)$$

#### Approximation by a Simple Function

• For any nonnegative extended real-valued and  $\mathcal{A}$ -measurable function f, there is a nondecreasing sequence of nonnegative  $\mathcal{A}$ -measurable simple functions that converges pointwise to f,

$$0 \le s_1(x) \le s_2(x) \le \dots \le f(x)$$
$$f(x) = \lim_{n \to \infty} s_n(x)$$

• If f is the pointwise limit of an increasing sequence of nonnegative  $\mathcal{A}$ -measurable simple functions, then f is an extended real-valued A-measurable function



 $\iff$  The nonnegative extended real-valued  $\mathcal{A}$ -measurable functions are exactly the ones that can be approximated using sequences of A-measurable simple functions

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## Integral of a Nonnegative Function

• A measure space  $(\Omega, \mathcal{A}, \mu)$  and  $f : \Omega \to \mathbb{R}^*$  a nonnegative extended real-valued function which is A-measurable. The integral of f over  $\Omega$  is defined as

$$\int_{\Omega} f d\mu = \sup_{s} \int_{\Omega} s d\mu$$

where the supremum is over all nonnegative A-measurable simple functions dominated by f.

• Integral over an arbitrary set  $E \in \mathcal{A}$ ,

$$\int_E f d\mu = \int_\Omega f \chi_E d\mu$$

#### **Convergence** Results

• MCT: if {*f<sub>n</sub>*} is a monotone nondecreasing sequence of nonnegative extended real-valued *A*-measurable functions, then

$$\int_E \lim f_n d\mu = \lim \int_E f_n d\mu$$

for any  $E \in \mathcal{A}$ 

• Fatou: if  $\{f_n\}$  is a sequence of nonnegative extended real-valued  $\mathcal{A}$ -measurable functions, then

$$\int_E \liminf f_n d\mu \le \liminf \int_E f_n d\mu$$

for any  $E \in \mathcal{A}$ 

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Integral of a General Function

• Let f be an extended real-valued A-measurable function, and let  $f^+ = \max\{f, 0\}, f^- = -\min\{f, 0\}$ , then the integral of fover E is defined as

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

for any  $E \in \mathcal{A}$ 

• f is integrable over E if

$$\int_E |f|d\mu = \int_E f^+ d\mu + \int_E f^- d\mu < \infty$$

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#### Integral of a Function Defined A.E.

 A measure space (Ω, A, μ), and a function f defined μ-a.e. on Ω (if D is the domain of f then μ(D<sup>c</sup>) = 0). If there is an extended real-valued A-measurable function g such that g = f μ-a.e., then define the integral of f as

$$\int_E f d\mu = \int_E g d\mu$$

for any  $E \in \mathcal{A}$ .

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DCT, General Version

A measure space (Ω, A, μ), and a sequence {f<sub>n</sub>} of extended real-valued A-measurable functions that converges pointwise μ-a.e. Assume that there is a nonnegative integrable function g such that |f<sub>n</sub>| ≤ g μ-a.e. for each n. Then

$$\int_E \lim f_n d\mu = \lim \int_E f_n d\mu$$

for any  $E \in \mathcal{A}$ 

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### DCT: Proof

• Let  $f(x) = \lim f_n(x)$  if  $\lim f_n(x)$  exists, and f(x) = 0 o.w., then f is measurable and  $f_n \to f$   $\mu$ -a.e. Hence

$$\int_E \lim f_n d\mu = \int_E f d\mu$$

• Fatou  $\Rightarrow$ 

$$\int (g-f)d\mu \le \liminf_{n \to \infty} \int (g-f_n)d\mu = \int gd\mu - \limsup_{n \to \infty} \int f_n d\mu$$
$$\Rightarrow \limsup_{n \to \infty} \int f_n d\mu \le \int f d\mu$$
Fatou 
$$\Rightarrow$$

$$\int (g+f)d\mu \le \liminf_{n\to\infty} \int (g+f_n)d\mu \Rightarrow \int fd\mu \le \liminf_{n\to\infty} \int f_n d\mu$$

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## DCT for Convergence in Measure

A measure space (Ω, A, μ), and a sequence {f<sub>n</sub>} of extended real-valued A-measurable functions that converges in measure to the A-measurable function f. Assume that there is a nonnegative integrable function g such that |f<sub>n</sub>| ≤ g μ-a.e. for each n. Then

$$\int_E f d\mu = \lim \int_E f_n d\mu$$

for any  $E \in \mathcal{A}$ 

## **Distribution Functions**

• Let  $\mu$  be a finite measure on  $(\mathbb{R}, \mathcal{B})$ , then the distribution function of  $\mu$  is defined as

$$F_{\mu}(x) = \mu((-\infty, x])$$

- A (general) real-valued function F on  $\mathbb{R}$  is called a distribution function if the following holds
  - 1 F is monotone nondecreasing
  - **2** F is right continuous
  - $\mathbf{3} \ F$  is bounded
  - $4 \lim_{x \to -\infty} F(x) = 0$
- Each distribution function is the distribution function corresponding to a unique finite measure on (R, B)
- The finite measure µ corresponding to F is called the Lebesgue–Stieltjes measure corresponding to F

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## The Lebesgue–Stieltjes Integral

 Let F be a distribution function with corresponding Lebesgue–Stieltjes measure µ. Let f be a Borel measurable function, then the Lebesgue–Stieltjes integral of f w.r.t. F is defined as

$$\int f(x)dF(x) = \int f(x)d\mu(x)$$

## The Lebesgue–Stieltjes Integral: Example

• Take the Dirac measure

$$\delta_b(E) = \begin{cases} 1, & b \in E \\ 0, & \text{o.w.} \end{cases}$$

and restrict it to  $\mathcal{B}$ , then the corresponding distribution function is

$$F(x) = \begin{cases} 1, & x \ge b \\ 0, & \text{o.w.} \end{cases}$$

• Let f be finite and Borel measurable, then

$$\int f(x)dF(x) = f(b)$$

• A way of handling discrete (random) variables and expectation, without having to resort to 'Dirac  $\delta$ -functions'

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