

Probability and Random Processes

Lecture 7

- Conditional probability and expectation
- Decomposition of measures

Conditional Probability

- A probability space (Ω, \mathcal{A}, P)
- An event $F \in \mathcal{A}$ with $P(F) > 0$; the σ -algebra generated by F , $\mathcal{G} = \sigma(\{F\}) = \{\emptyset, F, F^c, \Omega\}$
- Elementary conditional probability of $E \in \mathcal{A}$ given F

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$

- The conditional probability of $E \in \mathcal{A}$ **conditioned on** $\mathcal{G} =$ “the probability of E knowing which events in \mathcal{G} occurred” = “probability of E knowing whether F or F^c occurred”

$$P(E|\mathcal{G}) = P(E|F)\chi_F(\omega) + P(E|F^c)\chi_{F^c}(\omega)$$

a function : $\Omega \rightarrow \mathbb{R}$

- Note that $P(E|\mathcal{G})$
 - is a random variable on (Ω, \mathcal{A}, P) ;
 - is \mathcal{G} -measurable;

and that

$$P(G \cap E) = \int_G P(E|\mathcal{G})dP, \quad G \in \mathcal{G}$$

- A basis for generalizing $P(E|\mathcal{G})$ to conditioning on arbitrary σ -algebras

- Given (Ω, \mathcal{A}, P) , $E \in \mathcal{A}$ and $\mathcal{G} \subset \mathcal{A}$, there exists a nonnegative \mathcal{G} -measurable function $P(E|\mathcal{G})$ such that

$$P(G \cap E) = \int_G P(E|\mathcal{G})dP, \quad G \in \mathcal{G}$$

Also, $P(E|\mathcal{G})$ is unique P -a.e.

- Proof: Define $\mu_E(G) = P(G \cap E)$ for any $G \in \mathcal{G}$, then $\mu_E \ll P$ and

$$P(E|\mathcal{G}) = \frac{d\mu_E}{dP}$$

- The function $P(E|\mathcal{G})$ is called the **conditional probability of E given \mathcal{G}**
 - “the probability of E knowing which events in \mathcal{G} occurred”

- Again, for fixed \mathcal{G} and E , the entity $P(E|\mathcal{G})$ is a *function* $f(\omega) = P(E|\mathcal{G})(\omega)$ on Ω
- Alternatively, by instead fixing \mathcal{G} and ω we get a set function

$$m(E) = P(E|\mathcal{G})(\omega), \quad E \in \mathcal{A}$$

- If $m(E)$ is a probability measure on (Ω, \mathcal{A}) then $P(E|\mathcal{G})$ is said to be **regular**
 - $P(E|\mathcal{G})$ is in general not necessarily regular...
- If the space (Ω, \mathcal{A}) is **standard** (more about this later in the course), then $m(E)$ is a probability measure

Conditioning on a Random Variable

- Given (Ω, \mathcal{A}, P) and a random variable X , let $\sigma(X) =$ smallest $\mathcal{F} \subset \mathcal{A}$ such that X is (still) measurable w.r.t. $\mathcal{F} =$ the **σ -algebra generated by X** ,
 - $\sigma(X)$ is exactly the class of events for which you can get to know whether they occurred or not by observing X
- The **conditional probability of $E \in \mathcal{A}$ given X** is defined as

$$P(E|X) = P(E|\sigma(X))$$

Signed Measure

- Given a measurable space (Ω, \mathcal{A}) , a **signed measure** ν on \mathcal{A} is an extended real-valued function such that
 - $\nu(\emptyset) = 0$
 - for a sequence $\{A_i\}$ of pairwise disjoint sets in \mathcal{A}

$$\nu\left(\bigcup_i A_i\right) = \sum_i \nu(A_i)$$

(i.e., simply a measure that doesn't need to be positive)

Radon–Nikodym for Signed Measures

- If μ is a σ -finite measure and ν a finite signed measure on (Ω, \mathcal{A}) , and also $\nu \ll \mu$, then there is an integrable real-valued \mathcal{A} -measurable function f on Ω such that

$$\nu(A) = \int_A f d\mu$$

for any $A \in \mathcal{A}$. Furthermore, f is unique μ -a.e.

- The function f is the **Radon–Nikodym derivative** of ν w.r.t. μ , notation $f = \frac{d\nu}{d\mu}$

Conditional Expectation

- Given (Ω, \mathcal{A}, P) , a random variable Y (with $E[|Y|] < \infty$) and $\mathcal{G} \subset \mathcal{A}$, there exists a \mathcal{G} -measurable function $E[Y|\mathcal{G}]$ such that

$$\int_G Y dP = \int_G E[Y|\mathcal{G}] dP, \quad G \in \mathcal{G}$$

Also, the function $E[Y|\mathcal{G}]$ is unique P -a.e.

- Proof: Define $\mu_Y(G) = \int_G Y dP$ for any $G \in \mathcal{G}$, then $\mu_Y \ll P$ and

$$E[Y|\mathcal{G}] = \frac{d\mu_Y}{dP}$$

- The function $E[Y|\mathcal{G}]$ is called the **conditional expectation of Y given \mathcal{G}**
 - “the expectation of Y knowing which events in \mathcal{G} occurred”

- Note that with $\mathcal{G} = \{\emptyset, \Omega\}$ and $G = \Omega$, we get

$$E[Y] = \int_{\Omega} Y dP = \int_{\Omega} E[Y|\mathcal{G}] dP \Rightarrow E[Y|\{\emptyset, \Omega\}] = E[Y] \quad P\text{-a.e.}$$

(noting that the definition is verified also for $G = \emptyset$)

- If $\mathcal{G}_1 \subset \mathcal{G}_2 \subset \mathcal{A}$ then

$$E[E[Y|\mathcal{G}_1]|\mathcal{G}_2] = E[E[Y|\mathcal{G}_2]|\mathcal{G}_1] = E[Y|\mathcal{G}_1] \quad P\text{-a.e.}$$

so in particular, for any $\mathcal{G} \subset \mathcal{A}$,

$$E[E[Y|\mathcal{G}]] = E[E[Y|\mathcal{G}|\{\emptyset, \Omega\}]] = E[Y|\{\emptyset, \Omega\}] = E[Y] \quad P\text{-a.e.}$$

- If Z is (already) \mathcal{G} -measurable, then

$$E[ZY|\mathcal{G}] = ZE[Y|\mathcal{G}] \quad P\text{-a.e.}$$

Conditional Expectation vs. Probability

- The entity $E[Y|\mathcal{G}]$ is a function $g(\omega) = E[Y|\mathcal{G}](\omega)$
- If (Ω, \mathcal{A}) is standard, then $P(E|\mathcal{G})$ is regular
 $\Rightarrow m(E) = P(E|\mathcal{G})(\omega)$ is a probability measure on (Ω, \mathcal{A}) for fixed ω and \mathcal{G} . Furthermore, in this case

$$E[Y|\mathcal{G}] = \int Y(u)dm(u) = \int Y(u)dP(u|\mathcal{G})$$

- This interpretation for conditional expectation does not hold in general (for non-standard (Ω, \mathcal{A}))

Projections and Atoms

Conditional expectation as a projection

- Given (Ω, \mathcal{A}, P) assume $\mathcal{G} \subset \mathcal{A}$ and let $\mathcal{M} = \{ \mathcal{G}\text{-measurable functions} \}$
- For an \mathcal{A} -measurable Y , let $g(\omega) = E[Y|\mathcal{G}](\omega)$, then

$$E[(Y - g)^2] \leq E[(Y - g')^2] \text{ for all } g' \in \mathcal{M}$$

- If Y is already in \mathcal{M} , then $g(\omega) = Y(\omega)$ P -a.e.

Conditioning on a random variable

- For two random variables X and Y , $E[Y|X] = E[Y|\sigma(X)]$
- If $E|Y| < \infty$ then there is a Borel-measurable $f : \mathbb{R} \rightarrow \mathbb{R}$ such that $E[Y|X] = f(X(\omega))$ P -a.e.
- Thus $E[Y|X](\omega)$ is constant on the sets $\{\omega : X(\omega) = x\}$

Atoms

- $A \in \mathcal{A}$ is an **atom** of \mathcal{A} if the only set in \mathcal{A} which is a proper subset of A is \emptyset
- If there is a countable $\{A_i\}$ such that $\mathcal{A} = \sigma(\{A_i\})$ then \mathcal{A} is **separable**
- (Ω, \mathcal{A}) standard $\Rightarrow \mathcal{A}$ separable [more about “standard” later]
- \mathcal{A} separable \Rightarrow every $A \in \mathcal{A}$ is a union of atoms
- If f is \mathcal{A} -measurable, then f is constant on the atoms of \mathcal{A}
- If \mathcal{A} is separable and $\mathcal{G} \subset \mathcal{A}$, then the atoms of \mathcal{G} are bigger
- If G is an atom of $\mathcal{G} \subset \mathcal{A}$ and $P(G) > 0$, then

$$E[Y|\mathcal{G}](\omega) = \frac{1}{P(G)} \int_G Y dP, \quad \text{for } \omega \in G$$

“smoothing over atoms”

Mutually Singular Measures

- Given (Ω, \mathcal{A}) , two measures μ_1 and μ_2 are **mutually singular**, notation $\mu_1 \perp \mu_2$, if there is a set $E \in \mathcal{A}$ such that $\mu_1(E^c) = 0$ and $\mu_2(E) = 0$.
- **Lebesgue decomposition**: Given a σ -finite measure space $(\Omega, \mathcal{A}, \mu)$ and an additional σ -finite measure ν on \mathcal{A} , there exist measures ν_1 and ν_2 on \mathcal{A} such that $\nu_1 \ll \mu$, $\nu_2 \perp \mu$ and $\nu = \nu_1 + \nu_2$. This representation is unique.

Continuous and Discrete Measures

- For a measure space $(\Omega, \mathcal{A}, \mu)$ such that $\{x\} \in \mathcal{A}$ for all $x \in \Omega$:
 - $x \in \Omega$ is an **atom** of μ if $\mu(\{x\}) > 0$
 - μ is **continuous** if it has no atoms
 - μ is **discrete** if there is a countable $K \subset \Omega$ such that $\mu(K^c) = 0$
- Let $(\Omega, \mathcal{A}, \mu)$ be a σ -finite measure space and ν an additional σ -finite measure on \mathcal{A} . Assume that $\{x\} \in \mathcal{A}$ for all $x \in \Omega$. Then there exist measures ν_{ac} , ν_{sc} and ν_d such that
 - $\nu_{ac} \ll \mu$, $\nu_{sc} \perp \mu$ and $\nu_d \perp \mu$
 - ν_{sc} is continuous and ν_d is discrete
 - $\nu = \nu_{ac} + \nu_{sc} + \nu_d$, uniquely

Decomposition on the Real Line

- Let ν be a finite measure on $(\mathbb{R}, \mathcal{B})$, then ν can be decomposed uniquely as $\nu = \nu_{ac} + \nu_{sc} + \nu_d$ where
 - ν_{ac} is absolutely continuous w.r.t. Lebesgue measure
 - ν_{sc} is continuous and singular w.r.t. Lebesgue measure
 - ν_d is discrete
- Furthermore, if F_ν is the distribution function of ν , then

$$\nu(\{x\}) = F_\nu(x) - \lim_{x' \rightarrow x^-} F_\nu(x')$$

That is, if there are atoms, they are the points of discontinuity of F_ν