

Notes on Probability

1 Radon-Nikodym Theorem

1.1 Statement

Definition 1.1.

- ***Indefinite Integrals.***

Let $(\Omega, \mathcal{F}, \mu)$ be a probability space. $p \in \mathcal{F}_+$. Define

$$\nu(A) = \int_A \mu(dx) p(x).$$

ν is a measure on (Ω, \mathcal{F}) , and is called the indefinite integral of p w.r.t μ .

- ***Absolutely continuous.***

Let μ and ν be measures on (Ω, \mathcal{F}) . ν is said to be absolutely continuous w.r.t μ if $\forall A \in \mathcal{F}$,

$$\mu(A) = 0 \implies \nu(A) = 0.$$

Remark.

- The indefinite integral gives us a convenient tool to construct new measures from the old.
- If ν is the indefinite integral of p w.r.t μ , for every $f \in \mathcal{F}_+$, we have

$$\int_{\Omega} \nu(dx) f(x) = \int_{\Omega} \mu(dx) p(x) f(x);$$

sometimes we write

$$\nu(dx) = \mu(dx) p(x).$$

- If ν is the indefinite integral of p w.r.t μ , ν is absolutely continuous w.r.t μ . (The Radon-Nikodym Theorem says the converse is true as well.)

Theorem 1.2 (Radon-Nikodym Theorem). Let μ be a σ -finite measure and ν be absolutely continuous w.r.t μ . Then there exists $p \in \mathcal{F}_+$ s.t.

$$\int_{\Omega} \nu(dx) f(x) = \int_{\Omega} \mu(dx) p(x) f(x), \quad f \in \mathcal{F}_+.$$

Moreover, p is unique for μ -almost every $\omega \in \Omega$.

Remark. p can be denoted by $d\nu/d\mu$, and be called the Radon-Nikodym derivative of ν w.r.t μ .

1.2 Application of Radon-Nikodym Theorem

Theorem 1.3 (Existence and Uniqueness of Conditional Expectation). Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, $X \in \mathcal{F}_+$, and $\mathcal{F}_1 \subset \mathcal{F}$ be a sub- σ -algebra.

$\implies \mathbb{E}(X|\mathcal{F}_1)$ exists and is unique.

Proof. For each $A \in \mathcal{F}_1$, define $P(A) = \mathbb{P}(A)$, $Q(A) = \int_A \mathbb{P}(d\omega)X(\omega)$. (P is the restriction of \mathbb{P} to \mathcal{F}_1 ; Q is the indefinite integral of X w.r.t. P .)

Check conditions: P is a finite measure; Q is absolutely continuous w.r.t P .

By the Radon-Nikodym theorem, $\exists \bar{X} \in \mathcal{F}_+$ s.t.

$$\int_{\Omega} Q(d\omega)V(\omega) = \int_{\Omega} P(d\omega)\bar{X}(\omega)V(\omega)$$

for every $V \in \mathcal{F}_+$; i.e.

$$\mathbb{E}XV = \mathbb{E}\bar{X}V$$

for every $V \in \mathcal{F}_+$.

$\implies \bar{X} = \mathbb{E}(X|\mathcal{F}_1)$, by the definition of conditional expectations. \square

2 Riesz Representation Theorem (RRT)

2.1 Setting the Stage

Theorem 2.1 (Riesz Representation Theorem). *Every continuous linear functional $\phi : C(X) \rightarrow \mathbb{R}$ can be represented by a unique signed Baire measure μ on X via*

$$\phi(f) = \int_X f \, d\mu.$$

Remark.

- Fix a compact Hausdorff space X . For each μ , we can define a bounded linear functional $\phi_{\mu} : f \mapsto \int_X f \, d\mu$.
- $\mu \neq \mu' \implies \phi_{\mu} \neq \phi_{\mu'}$. That is, $\mu \mapsto \phi_{\mu}$ is injective.
- The Riesz representation theorem says that $\mu \mapsto \phi_{\mu}$ is also surjective.

Definition 2.2. We will use the following notations to re-state the RRT:

- *Comp* - Compact Hausdorff spaces + continuous maps
- *Ban* - Banach spaces + continuous linear operators
- $C(X)$ - all continuous real-valued functions on X
- $C(X)^*$ - dual space of $C(X)$
- $M(X)$ - all Baire measures on X

Lemma 2.3. It is easy to check:

- C is a contravariant functor from *Comp* to *Ban*.

$$C : X \longmapsto C(X)$$

$$\alpha \longmapsto \alpha^{\sharp}$$

where $\alpha^{\sharp} : g \mapsto g \circ \alpha$.

- M is a covariant functor from *Comp* to *Ban*.

$$M : X \longmapsto M(X)$$

$$\alpha \longmapsto \alpha^*$$

where $\alpha^* : \mu \mapsto \mu \circ \alpha^{-1}$.

- Define $\iota_X : M(X) \rightarrow C(X)^*$ by $\mu \mapsto \phi_{\mu}$ where $\phi_{\mu}(f) = \int_X f \, d\mu$.
Then ι is a natural transformation from M to C^* .

Theorem 2.4 (Riesz Representation Theorem Revisited). ι is a natural equivalent from M to C^* .

2.2 Proof of RRT

Definition 2.5.

- A topological space is called extremally disconnected if each open set has an open closure.
- \mathcal{C} - all clopen sets

Lemma 2.6. Let $X \in \text{Comp}$ be extremally disconnected. Then

- a) the Baire sets in X are generated by \mathcal{C} ;
- b) the simple functions based on \mathcal{C} are uniformly dense in $C(X)$.

Lemma 2.7. The Stone-Čech compactification of a discrete space is extremally disconnected.

Lemma 2.8. For each object $Y \in \text{Comp}$, there is a morphism $\alpha : X \rightarrow Y$ such that X is extremally disconnected.

Proof of RRT. Let Y be an object in Comp , and $\alpha : X \rightarrow Y$ be the morphism starting at an extremally disconnected object X .

- $\alpha^\#$ is a norm isomorphism.
- By the Hahn-Banach Theorem, $\alpha^{\#\#}$ is also surjective.
- ι_X is surjective $\implies \iota_Y$ is surjective.

□

2.3 Application of RRT in the Moment Problem

Exsercise 2.3.1. Show that the moments completely determine the probability distribution if it is concentrated on a finite interval (that if, $\mathbb{P}(X_n \in [a, b]) = 1$, for all n).

Proof. The first proof is based on the Riesz representation theorem.

- Assume F, G are concentrated on $[a, b]$ and

$$\int_a^b x^k \, dF = \int_a^b x^k \, dG \quad \forall k \in \mathbb{R}.$$

We want to prove $F = G$ on $[a, b]$.

- Let $C[a, b]$ be the space of all continuous functions on $[a, b]$ with the uniform norm. Define

$$\begin{aligned} T_F : C[a, b] &\longrightarrow \mathbb{R} \\ h &\longmapsto \int_{[a, b]} h \, dF \\ T_G : C[a, b] &\longrightarrow \mathbb{R} \\ h &\longmapsto \int_{[a, b]} h \, dG \end{aligned}$$

It is easy to check that T_F and T_G are both continuous linear functionals.

- Obviously, for each polynomial $h(x) = \sum_{i=0}^n a_i x^i$, $T_F(h) = T_G(h)$. Thus, $T_F = T_G$ on $C[a, b]$, since polynomials are a dense set of $C[a, b]$ by the Stone-Weierstrass theorem.

- (Riesz representation theorem) Every continuous linear functional ψ over $C[a, b]$ can be uniquely represented as

$$\psi(f) = \int_{[a,b]} f \, d\mu$$

where μ is .

It suffices to check F and G are bounded variation functions.

□

3 Poisson Random Measures

3.1 Characterization via Laplace functionals

Definition 3.1. A random measure N on (E, \mathcal{E}) is called Poisson with mean ν if

- a) for every $A \in \mathcal{E}$, $N(A) \sim \text{Poisson}(\nu(A))$.
- b) for disjoint $A_1, \dots, A_n \in \mathcal{E}$, $N(A_1), \dots, N(A_n)$ are independent.

Remark.

- For every $f \in \mathcal{E}_+$, define

$$Nf(\omega) = \int_E N(\omega, dx) f(x).$$

- $f \mapsto \mathbb{E}e^{-Nf}$ is called the Laplace functional of N .
- We say a sequence of RVs X taking values in (E, \mathcal{E}) forms a Poisson random measure on (E, \mathcal{E}) , if

$$Nf = \sum_{i \in I} f \circ X_i, \quad f \in \mathcal{E}_+$$

is Poisson with some mean measure ν .

Theorem 3.2. Let N be a random measure on (E, \mathcal{E}) . It is Poisson with mean ν if and only if

$$\mathbb{E}e^{-Nf} = e^{-\nu(1-e^f)}, \quad f \in \mathcal{E}_+,$$

where $\nu f = \int_E \nu(dx) f(x) = \int_E f d\nu$.

Proof. \Rightarrow : It suffices to prove this result for all simple functions.

Assume $f = \sum a_i \mathbf{1}_{A_i}$. Then $Nf = \sum a_i N(A_i)$. By the definition, all A_i are independent, and the distribution of $N(A_i)$ is Poisson(νA_i), so

$$\begin{aligned} \mathbb{E}e^{-Nf} &= \mathbb{E}e^{-\sum a_i N(A_i)} \\ &= \prod \mathbb{E}e^{-a_i N(A_i)} \\ &= \prod \exp[\nu A_i \cdot (e^{-a_i} - 1)] \\ &= \exp\left[-\sum \nu A_i \cdot (1 - e^{-a_i})\right] = e^{-\nu(1-e^f)} \end{aligned}$$

If $f \in \mathcal{E}_+$, we can use a sequence of simple functions in \mathcal{E}_+ increasing to f . Then use the continuity of Laplace functionals and the monotone convergence theorem.

\Leftarrow : Note that $\{\mathbb{E}e^{-Nf}\}$ uniquely determines the probability law of N . And by the necessity part, it must be Poisson. □