

STATS219: Stochastic Processes

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1 Measure theory

1.1 Definitions of measure theory

1.1.1 Powerset (2^Ω)

Set of all possible subsets of Ω

1.1.2 σ -algebra (\mathcal{F})

(a) $\Omega \in \mathcal{F}$

(b) $A \in \mathcal{F} \Rightarrow A^C \in \mathcal{F}$

(c) $A_1, \dots, A_\infty \in \mathcal{F} \Rightarrow \cup_i A_i \in \mathcal{F}$

1.1.3 Probability space $((\Omega, \mathcal{F}, P))$

$P : \mathcal{F} \rightarrow [0, 1]$ such that

(a) $0 \leq P(A) \leq 1 \forall A \in \mathcal{F}$

(b) $P(\Omega) = 1$

(c) $P(A) = \sum_i P(A_i)$ whenever $A = \cup_i A_i$ and $A_n \cap A_m = \emptyset$ for $n \neq m$

2 Random variables

2.1 Definitions of random variables

2.1.1 Random variable (X , "mathcal{F}-measurable function")

$X : \Omega \rightarrow \mathbb{R}$ such that

$$\{\omega : X(\omega) \leq \alpha\} \in \mathcal{F}, \forall \alpha \in \mathbb{R}$$

2.1.2 Indicator function (\mathbb{I}_A)

A rv $\forall A \in \mathcal{F}$ such that

$$I_A(\omega) = \begin{cases} 1 & \text{if } \omega \in A \\ 0 & \text{else} \end{cases}$$

2.1.3 Borel function

$g : \mathbb{R} \rightarrow \mathbb{R} \Rightarrow g$ is an rv on $(\mathbb{R}, \mathcal{B})$

2.1.4 σ -algebra generator ($\sigma(\{A_\alpha\})$)

$$\sigma(\{A_\alpha\}) = \cap \{ \mathcal{G} : \mathcal{G} \subseteq 2^\Omega \text{ } \sigma\text{-field, } A_\alpha \in \mathcal{G}, \forall \alpha \in \Gamma, \text{ a countable or uncountable index} \}$$
$$\sigma(X) = \mathcal{F}_X = \sigma(\{\omega : X(\omega) \leq \alpha\} \forall \alpha)$$

2.1.5 Expectation of an rv ($E[X]$)

$$E[X] = \lim_{n \rightarrow \infty} \left[\sum_{k=0}^{\infty} x_{k,n} * P(\{\omega : X(\omega) \in I_{k,n}\}) \right] \text{ for } x_{k,n} = k2^{-n}, I_{k,n} = (x_{k,n}, x_{k+1,n}]$$

2.1.6 Independence

events A, B independent $\implies P(A \cap B) = P(A)P(B)$ for $A, B \in \mathcal{F}$
 σ -fields $\mathcal{H}, \mathcal{G} \subseteq \mathcal{F}$ independent $\implies P(G \cap H) = P(G)P(H), \forall G \in \mathcal{G}, \forall H \in \mathcal{H}$

2.1.7 Uncorrelated

$E(XY) = E(X)E(Y)$, for $X, Y \in L^2$

2.1.8 L^q spaces ($L^q(\Omega, \mathcal{F}, P)$)

The collection of all rv X on (Ω, \mathcal{F}) where $E(|X|^q) < \infty$

2.1.9 Law of an rv (\mathcal{P}_x)

Probability measure on $(\mathbb{R}, \mathcal{B})$ such that $\mathcal{P}_x(B) = P(\{\omega : X(\omega) \in B\})$ for all $B \in \mathcal{B}$

2.1.10 Distribution function of an rv (F_X)

$F_X(\alpha) = P(\{\omega : X(\omega) \leq \alpha\}) = \mathcal{P}_X((-\infty, \alpha]) \quad \forall \alpha \in \mathbb{R}$

2.1.11 Convergence and equality almost surely (a.s.)

$$\begin{aligned} X &\stackrel{a.s.}{=} Y \iff P(\{\omega : X(\omega) \neq Y(\omega)\}) = 0 \\ X &\stackrel{a.s.}{\leq} 0 \iff P(\{\omega : X(\omega) > 0\}) = 0 \\ X_n &\stackrel{a.s.}{\rightarrow} X \iff X_n(\omega) \rightarrow X(\omega) \text{ as } n \rightarrow \infty \quad \forall \omega \in A \in \mathcal{F} \text{ with } P(A) = 1 \end{aligned}$$

2.1.12 Convergence in probability (p)

$$X_n \xrightarrow{p} X \iff P(\{\omega : |X_n(\omega) - X(\omega)| > \epsilon\}) \longrightarrow 0 \text{ as } n \rightarrow \infty \quad \forall \epsilon \in \mathbb{R}$$

2.1.13 Convergence in L^q (or in q-mean)

$$X_n \xrightarrow{q.m.} X \iff \|X_n - X\|_q = [E(|X_n - X|^q)]^{\frac{1}{q}} \longrightarrow 0 \text{ as } n \rightarrow \infty, \text{ for } X_n, X \in L^q$$

2.1.14 Convergence in law (or weak convergence or in distribution)

$$X_n \xrightarrow{d} X \iff F_{X_n}(\alpha) \longrightarrow F_X(\alpha) \text{ as } n \rightarrow \infty \quad \forall \alpha \text{ continuity points of } F_X$$

3 Conditional expectation

3.1 Definitions of conditional expectation

3.1.1 In discrete space

$$f(y) := E(X|Y = y) = \frac{E(X * I\{Y = y\})}{P(Y = y)} \quad \forall y \text{ requiring } P(Y = y) > 0$$

3.1.2 In L^2 space (Hilbert space)

$Z = E(X|Y)$ unique random variable satisfying :

$$Z \in H_Y = L^2(\Omega, \sigma(Y), P)$$

$$E[(X - Z)V] = 0 \quad \forall V \in H_Y$$

$$\min_Z \{E[(X - Z)^2] : Z \in H_Y\} \text{ (interchangeable with line above it b/c of } L^2 \text{ properties)}$$

3.1.3 In L^1 space

$Z = E(X|Y)$ such that :

$$E[(X - Z)I\{Y = y\}] = 0, \forall y \in \mathbb{R}$$

$$E[(X - Z)I\{Y \in B\}] = 0, \forall B \in \mathcal{B}$$

3.2 Properties of conditional expectation

For $X, Y \in L^1$, σ -fields $\mathcal{H} \subseteq \mathcal{G} \subseteq \mathcal{F}$

- $X \geq 0 \implies E(X|\mathcal{G}) \stackrel{a.s.}{\geq} 0$
- If $\mathcal{G}, \sigma(X)$ independent $\implies E(X|\mathcal{G}) = E(X)$
- If X is \mathcal{G} -measurable $\implies E(X|\mathcal{G}) = X$
- $E[\alpha X + \beta Y|\mathcal{G}] = \alpha E(X|\mathcal{G}) + \beta E(Y|\mathcal{G}) \quad \forall \alpha, \beta \in \mathbb{R}$
- $E[E(X|\mathcal{G})|\mathcal{H}] = E(X|\mathcal{H})$
- $E[E(X|\mathcal{G})] = E(X)$
- If Y is \mathcal{G} -measurable and $X, XY \in L^1 \implies E[XY|\mathcal{G}] = Y E[X|\mathcal{G}]$

For $Var(Y|\mathcal{G}) := E(Y^2|\mathcal{G}) - E(Y|\mathcal{G})^2$, $Y \in L^2$

- If $Y \in L^2(\Omega, \mathcal{G}, P) \implies Var(Y|\mathcal{G}) = 0$
- $Var(Y) = E[Var(Y|\mathcal{G})] + Var[E(Y|\mathcal{G})]$
- If $E(Y|\mathcal{G}) = X$ and $E(X^2) = E(Y^2) < \infty \implies X \stackrel{a.s.}{=} Y$
- If $\mathcal{H} \subseteq \mathcal{G}$ and $X \in L^2(\Omega, \mathcal{F}, P) \implies E[(X - E(X|\mathcal{G}))^2] \leq E[(X - E(X|\mathcal{H}))^2]$