

$$\tilde{X}: \Omega \rightarrow \mathbb{R}^k \quad \text{rvec}$$

$$(k=1 \rightarrow \text{rv})$$

$I(A)$ — indicator rv defined as

$$I(A)(\omega) = 1, \quad \omega \in A \\ = 0, \quad \omega \notin A$$

Useful results

$$I(A \cup B) = I(A) + I(B) - I(AB)$$

$$I(\bigcap_k A_k) = \prod_k I(A_k)$$

$$I(\emptyset) = 0, \quad I(\Omega) = 1$$

$$I(A) + I(A^c) = 1$$

$$A_1, A_2, \dots \text{ disjoint} \Rightarrow I(\bigcup_k A_k) = \sum_k I(A_k)$$

$$A \subset B \Leftrightarrow I(A) \leq I(B) \quad \text{— see def'n}$$

Def'n $X \leq Y$ if $X(\omega) \leq Y(\omega), \quad \forall \omega \in \Omega$

Sequence of rv's

X_1, X_2, \dots

Def'n $X_n \rightarrow X$ if $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega), \forall \omega$

Notation $X_n \uparrow X$ if

$X_1 \leq X_2 \leq \dots$ and $X_n \rightarrow X$

$X_n \downarrow X$ if

$X_1 \geq X_2 \geq \dots$ and $X_n \rightarrow X$

Def'n $A_n \rightarrow A$ if $I(A_n) \rightarrow I(A)$

\uparrow
events

Notation $A_n \uparrow A$ if $A_1 \subset A_2 \subset \dots$ and $A_n \rightarrow A$

$A_n \downarrow A$ if $A_1 \supset A_2 \supset \dots$ and $A_n \rightarrow A$

E-Axioms

$$(1) E(1) = 1$$

$$(2) X \geq 0 \Rightarrow E(X) \geq 0$$

$$(3) E(cX + dY) = cE(X) + dE(Y)$$

(4) $X_k \geq 0, \forall k$ then

$$E\left(\sum_{k=1}^{\infty} X_k\right) = \sum_{k=1}^{\infty} E(X_k)$$

$$(4)' \quad 0 \leq X_m \uparrow X \Rightarrow E(X_m) \rightarrow E(X)$$

Proposition $\{(1), (2), (3), (4)\} \Leftrightarrow \{(1), (2), (3), (4)'\}$

Remark 4' is the Monotone Convergence Theorem (MCT).

Def'n $P(A) = E[I(A)]$

Proposition (Kolmogorov Axioms)

1 $P(\Omega) = 1$

2 $P(A) \geq 0$

3 A_1, A_2 disjoint $\Rightarrow P(A_1 \cup A_2) = P(A_1) + P(A_2)$

4 A_1, A_2, \dots disjoint \Rightarrow

$$P\left(\bigcup_{k=1}^{\infty} A_k\right) = \sum_{k=1}^{\infty} P(A_k)$$

Note 4 is the σ -additive property of P

P is a f'm from events to \mathbb{R}

E is a f'm from rv's to \mathbb{R}

$(\Omega, \text{collection of events}, P)$ is a probability space

Proposition $X \leq Y \Rightarrow E(X) \leq E(Y)$

Def'n $g: \mathbb{R} \rightarrow \mathbb{R}$ is convex if
 $\forall x_0 \in \mathbb{R} \exists c$ such that
 $g(x) \geq g(x_0) + c(x - x_0), \forall x$

Jensen's Inequality

$$E[g(X)] \leq g[E(X)]$$

Markov's Inequality

$$P(|X| \geq c) \leq \frac{E(|X|)}{c}$$

($c > 0$ is a constant)