

Tail Fields

(ref: § 3.5 Rosenthal, § 4 Billingsley)

in a prob. space
(Ω, \mathcal{F}, P)

def. Given a sequence of events $A_1, A_2, \dots \in \mathcal{F}$,
we define their tail field by

$$\mathcal{T} = \bigcap_{n=1}^{\infty} \sigma(A_n, A_{n+1}, \dots)$$

In words, an event $A \in \mathcal{T}$ must have the property that for any $n \in \mathbb{N}$, it depends only on the events A_n, A_{n+1}, \dots . In particular, $A \in \mathcal{T}$ iff changing/omitting a finite number of values does not affect the occurrence of A .
(does not depend on any finite set $\{A_1, A_2, \dots, A_n\}$)

* Tail event depends only on limiting behavior

Examples

- $\limsup_n A_n \in \mathcal{T}$
- $\liminf_n A_n \in \mathcal{T}$

Pf.: See Billingsley or Rosenthal

Kolmogorov's 0-1 Law: If events A_1, A_2, \dots are independent with tail field \mathcal{T} , then for $A \in \mathcal{T}$, $P(A) = 0$ or 1 .

Expectation

(ref: § 5, Rosenthal § 4)

Simple Random Variables

Let (Ω, \mathcal{F}, P) be a probability space and let X be a random variable defined on this space.

def: A random variable X is simple if $\text{range}(X)$ is finite, where $\text{range}(X) \equiv \{X(\omega) : \omega \in \Omega\}$.

↑
assumes only
finitely many
values

We can list the distinct elements in the range of X as x_1, x_2, \dots, x_n

and write
$$X = \sum_{i=1}^n x_i \mathbb{1}_{A_i}$$

where $A_i = \{\omega \in \Omega : X(\omega) = x_i\} = X^{-1}(\{x_i\}) \in \mathcal{F}$

and where $\mathbb{1}_{A_i}$ are indicator functions.

$$\mathbb{1}_{A_i} = \begin{cases} 1 & \text{if } \omega \in A_i \\ 0 & \text{if } \omega \notin A_i \end{cases}$$

Note that the A_i 's form a finite partition of Ω .

def. For a simple RV $X = \sum_{i=1}^n x_i \mathbb{1}_{A_i}$, we define its expected value (or expectation or mean) by

$$E[X] = E\left[\sum_{i=1}^n x_i \mathbb{1}_{A_i}\right] = \sum_{i=1}^n x_i P(A_i)$$

where $\{A_i\}$ is a finite partition of Ω .

Notation: often write $\mu_X = E[X]$.

Example: Let (Ω, \mathcal{F}, P) be Lebesgue measure on $[0,1]$, and define simple RVs X & Y by

$$X(\omega) = \begin{cases} 5, & \text{if } \omega > \frac{1}{3} \\ 3, & \text{if } \omega \leq \frac{1}{3} \end{cases}$$

$$Y(\omega) = \begin{cases} 2, & \text{if } \omega \text{ is rational } (\mathbb{Q} \cap [0,1]) \\ 4, & \text{if } \omega = \frac{1}{\sqrt{2}} (\approx 0.707) \\ 6, & \text{if other } \omega \leq \frac{1}{4} \text{ i.e. } [0, \frac{1}{4}] \setminus ((\mathbb{Q} \cap [0,1]) \cup \{\frac{1}{\sqrt{2}}\}) \\ 8, & \text{otherwise} \end{cases}$$

Compute $E[X]$ and $E[Y]$.

- Partition of $[0,1]$: $\underbrace{[0, \frac{1}{3})}_{A_1}, \underbrace{[\frac{1}{3}, 1]}_{A_2}$

$$E[X] = 3 P([0, \frac{1}{3})) + 5 P([\frac{1}{3}, 1]) \\ = 3 \cdot \frac{1}{3} + 5 \cdot \frac{2}{3} = \frac{3}{3} + \frac{10}{3} = \frac{13}{3}$$

- Partition of $[0,1]$: $\underbrace{\mathbb{Q} \cap [0,1]}_{A_1}, \underbrace{\{\frac{1}{\sqrt{2}}\}}_{A_2}, \underbrace{[0, \frac{1}{4}] \setminus (\mathbb{Q} \cap [0,1])}_{A_3}, A_4$

where $A_4 = [0,1] \setminus (A_1 \cup A_2 \cup A_3) = (\frac{1}{4}, 1] \setminus (\mathbb{Q} \cap [0,1] \cup \{\frac{1}{\sqrt{2}}\})$

$$E[Y] = 2 \underbrace{P(A_1)}_{\substack{0 \\ \text{countable}}} + 4 \underbrace{P(A_2)}_{\substack{0 \\ \text{singleton}}} + 6 \underbrace{P(A_3)}_{\frac{1}{4}} + 8 \underbrace{P(A_4)}_{\frac{3}{4}} \\ = \frac{6}{4} + \frac{24}{4} = \frac{30}{4} = \frac{15}{2}$$

Remark: $E[\underbrace{\mathbb{1}_A}_{\text{simple RV}}] = P(A)$ for all $A \in \mathcal{F}$

$$E[\underbrace{c}_{\text{simple RV}}] = c \quad \forall c \in \mathbb{R} \text{ (constants)} \\ X(\omega) = c$$

Expectation is Linear

Let $X = \sum_{i=1}^n x_i \mathbb{1}_{A_i}$ and $Y = \sum_{j=1}^m y_j \mathbb{1}_{B_j}$ where

$\{A_i\}$ and $\{B_j\}$ are finite partitions of Ω .

If $a, b \in \mathbb{R}$, then $\{A_i \cap B_j\}$ is again a finite partition of Ω and we have

$$E[aX + bY] = E\left[\sum_{i,j} (ax_i + by_j) \mathbb{1}_{(A_i \cap B_j)}\right]$$

$$= \sum_{i,j} (ax_i + by_j) P(A_i \cap B_j)$$

$$= a \sum_i x_i P(A_i) + b \sum_j y_j P(B_j)$$

$$= a E[X] + b E[Y].$$

Q. why?
indep?
partition?

Note: It follows that $E\left[\sum_{i=1}^n x_i \mathbb{1}_{A_i}\right] = \sum_{i=1}^n x_i P(A_i)$

for ANY finite collection of subsets $A_i \subseteq \Omega$, even if they do not form a partition.

[since $P(A) = \sum_i P(A \cap B_i)$ for any partition $\{B_i\}$]

Expectation is Order Preserving

If simple RVs $X \leq Y$ are s.t. $X \leq Y$

(meaning that $X(\omega) \leq Y(\omega) \forall \omega \in \Omega$), then

$$E[X] \leq E[Y].$$

Why is this true?

$$X \leq Y \Rightarrow Y - X \geq 0$$

$$\Rightarrow E[Y - X] \geq 0$$

$$\Rightarrow E[Y] - E[X] \geq 0 \text{ by linearity}$$

$$\Rightarrow E[X] \leq E[Y].$$

In particular, since

$$-|X| \leq X \leq |X|, \text{ we have that } |E[X]| \leq E[|X|].$$

aka Generalized Triangle Inequality

Suppose X & Y are independent simple RVs. Then

$$\boxed{E[XY] = E[X] \cdot E[Y]}$$

Why?

$$X = \sum_{i=1}^n x_i \mathbb{1}_{A_i} \quad \text{and} \quad Y = \sum_{j=1}^m y_j \mathbb{1}_{B_j},$$

$\{A_i\}$ and $\{B_j\}$ finite partitions of Ω ,

$\{x_i\}$ and $\{y_j\}$ are distinct, then

X & Y are independent iff

$$P(A_i \cap B_j) = P(A_i) P(B_j) \quad \forall i, j.$$

In that case,

$$\begin{aligned} E[XY] &= \sum_{i,j} x_i y_j P(A_i \cap B_j) \\ &= \sum_{i,j} x_i y_j P(A_i) P(B_j) \\ &= E[X] E[Y]. \end{aligned}$$

Note: This may be false if X & Y not indep!

Example: $X = \begin{cases} +1 & \text{with prob. } \frac{1}{2} \\ -1 & \text{with prob. } \frac{1}{2} \end{cases}$

e.g. fair coin flip
where heads = 1 $\frac{1}{2}$
tails = -1

If $Y = X$, then

(or modified Bernoulli trial)

$$E[X] = E[Y] = 1\left(\frac{1}{2}\right) + (-1)\left(\frac{1}{2}\right) = 0$$

BUT $E[XY] = 1$. $\left[(1)(1)\left(\frac{1}{2}\right) + (-1)(-1)\left(\frac{1}{2}\right) = 1 \right]$

Example:

We may have $E[XY] = E[X]E[Y]$ even if $X \not\perp Y$
are NOT independent.

$$X = \begin{cases} 0 & \text{w/prob } \frac{1}{3} \\ 1 & \text{"} \\ 2 & \text{"} \end{cases}$$

$$Y(\omega) = 1 \quad \text{if } X(\omega) = 0 \text{ or } 2$$

$$Y(\omega) = 5 \quad \text{if } X(\omega) = 1$$

Compute $E[XY]$, $E[X]$, $E[Y]$ to verify claim above