556: MATHEMATICAL STATISTICS I

THE BOREL-CANTELLI LEMMA

DEFINITION Limsup and liminf events

Let $\{E_n\}$ be a sequence of events in sample space Ω . Then

$$E^{(S)} = \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m$$

is the **limsup** event of the infinite sequence; event $E^{(S)}$ occurs if and only if

- for all $n \ge 1$, there exists an $m \ge n$ such that E_m occurs.
- infinitely many of the *E_n* occur.

Similarly, let

$$E^{(I)} = \bigcup_{n=1}^{\infty} \bigcap_{m=n}^{\infty} E_m$$

is the **liminf** event of the infinite sequence; event $E^{(I)}$ occurs if and only if

- there exists $n \ge 1$, such that for all $m \ge n$, E_m occurs.
- only finitely many of the *E_n* do not occur.

THEOREM The Borel-Cantelli Lemma

Let $\{E_n\}$ be a sequence of events in sample space Ω . Then

(a) If

$$\sum_{n=1}^{\infty} P(E_n) < \infty, \qquad \Longrightarrow \qquad P\left(E^{(S)}\right) = 0,$$

that is,

 $P[E_n \text{ occurs infinitely often }] = 0.$

(b) If the events $\{E_n\}$ are **independent**

$$\sum_{n=1}^{\infty} P(E_n) = \infty \qquad \Longrightarrow \qquad P\left(E^{(S)}\right) = 1.$$

that is,

 $P[E_n \text{ occurs infinitely often }] = 1.$

Note: This result is useful for assessing almost sure convergence. For a sequence of random variables $\{X_n\}$ and limit random variable X, suppose, for $\epsilon > 0$, that $A_n(\epsilon)$ is the event

$$A_n(\epsilon) \equiv \{\omega : |X_n(\omega) - X(\omega)| > \epsilon\}$$

The BC Lemma says

(a) if
$$\sum_{n=1}^{\infty} P(A_n(\epsilon)) = \sum_{n=1}^{\infty} P[|X_n - X| > \epsilon] < \infty$$
 then $X_n \xrightarrow{a.s.} X$

(b) if
$$\sum_{n=1}^{\infty} P(A_n(\epsilon)) = \sum_{n=1}^{\infty} P[|X_n - X| > \epsilon] = \infty$$
 with the X_n independent then $X_n \xrightarrow{a.s.} X$

Proof. NOT EXAMINABLE

(a) Note first that

$$\sum_{n=1}^{\infty} P(E_n) < \infty \quad \Longrightarrow \quad \lim_{n \to \infty} \sum_{m=n}^{\infty} P(E_m) = 0.$$

because if the sum on the left-hand side is finite, then the tail-sums on the right-hand side tend to zero as $n \longrightarrow \infty$. But for every $n \ge 1$,

$$E^{(S)} = \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \subseteq \bigcup_{m=n}^{\infty} E_m \qquad \therefore \qquad P\left(E^{(S)}\right) \leq P\left(\bigcup_{m=n}^{\infty} E_m\right) \leq \sum_{m=n}^{\infty} P(E_m). \tag{1}$$

Thus, taking limits as $n \longrightarrow \infty$, we have that

$$P\left(E^{(S)}\right) \leq \lim_{n \to \infty} \sum_{m=n}^{\infty} P(E_m) = 0.$$

(b) Consider $N \ge n$, and the union of events

$$E_{n,N} = \bigcup_{m=n}^{N} E_m.$$

 $E_{n,N}$ corresponds to the collection of sample outcomes that are in *at least one* of the collections corresponding to events $E_n, ..., E_N$. Therefore, $E'_{n,N}$ is the collection of sample outcomes in Ω that are **not** in **any** of the collections corresponding to events $E_n, ..., E_N$, and hence

$$E'_{n,N} = \bigcap_{m=n}^{N} E'_m \tag{2}$$

Now,

$$E_{n,N} \subseteq \bigcup_{m=n}^{\infty} E_m \implies P(E_{n,N}) \leq P\left(\bigcup_{m=n}^{\infty} E_m\right)$$

and hence, by assumption and independence,

$$1 - P\left(\bigcup_{m=n}^{\infty} E_{m}\right) \leq 1 - P\left(\bigcup_{m=n}^{N} E_{m}\right) = 1 - P(E_{n,N}) = P\left(E_{n,N}'\right) = P\left(\bigcap_{m=n}^{N} E_{m}'\right) = \prod_{m=n}^{N} P\left(E_{m}'\right)$$
$$= \prod_{m=n}^{N} (1 - P(E_{m})) \leq \exp\left\{-\sum_{m=n}^{N} P(E_{m})\right\},$$

as $1 - x \le \exp\{-x\}$ for 0 < x < 1. Now, taking the limit of both sides as $N \longrightarrow \infty$, for fixed n,

$$1 - P\left(\bigcup_{m=n}^{\infty} E_m\right) \leq \lim_{N \to \infty} \exp\left\{-\sum_{m=n}^{N} P\left(E_m\right)\right\} = 0$$

as, by assumption $\sum_{n=1}^{\infty} P(E_n) = \infty$. Thus, for each *n*, we have that

$$P\left(\bigcup_{m=n}^{\infty} E_m\right) = 1$$
 \therefore $\lim_{n \to \infty} P\left(\bigcup_{m=n}^{\infty} E_m\right) = 1.$

But the sequence of events $\{A_n\}$ defined for $n \ge 1$ by

$$A_n = \bigcup_{m=n}^{\infty} E_m$$

is monotone non-increasing, and hence, by continuity,

$$P\left(\lim_{n \to \infty} A_n\right) = \lim_{n \to \infty} P(A_n).$$
(3)

From (4), we have that the right hand side of equation (5) is equal to 1, and, by definition,

$$\lim_{n \to \infty} A_n = \bigcap_{n=1}^{\infty} A_n = \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m.$$
(4)

Hence, combining (2), (3) and (4) we have finally that

$$P\left(\bigcap_{n=1}^{\infty}\bigcup_{m=n}^{\infty}E_{m}\right) = 1 \implies P\left(E^{(S)}\right) = 1.$$

Interpretation and Implications

The Borel-Cantelli result is concerned with the calculation of the probability of the limsup event $E^{(S)}$ occurring for general infinite sequences of events $\{E_n\}$. From previous discussion, we have seen that $E^{(S)}$ corresponds to the collection of sample outcomes in Ω that are in **infinitely many** of the E_n collections. Alternately, $E^{(S)}$ occurs if and only if **infinitely many** $\{E_n\}$ occur. The Borel-Cantelli result tells us conditions under which $P(E^{(S)}) = 0$ or 1.

EXAMPLE : Consider the event *E* defined by

"*E* occurs" = "run of 100^{100} Heads occurs in an infinite sequence of independent coin tosses"

We wish to calculate P(E), and proceed as follows; consider the infinite sequence of events $\{E_n\}$ defined by

" E_n occurs" = "run of 100¹⁰⁰ Heads occurs in the *n*th block of 100¹⁰⁰ coin tosses"

Then $\{E_n\}$ are independent events, and

$$P(E_n) = \frac{1}{2^{100^{100}}} > 0 \implies \sum_{n=1}^{\infty} P(E_n) = \infty,$$

and hence by part (b) of the Borel-Cantelli result,

$$P\left(E^{(S)}\right) = P\left(\bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m\right) = 1$$

so that the probability that infinitely many of the $\{E_n\}$ occur is 1. But, crucially,

$$E^{(S)} \subseteq E \implies P(E) = 1.$$

Therefore the probability that *E* occurs, that is that a run of 100^{100} Heads occurs in an infinite sequence of independent coin tosses, is 1.