

# 18.175: Lecture 9

## Borel-Cantelli and strong law

Scott Sheffield

MIT

Laws of large numbers: Borel-Cantelli applications

Strong law of large numbers

Laws of large numbers: Borel-Cantelli applications

Strong law of large numbers

- ▶ **First Borel-Cantelli lemma:** If  $\sum_{n=1}^{\infty} P(A_n) < \infty$  then  $P(A_n \text{ i.o.}) = 0$ .

- ▶ **First Borel-Cantelli lemma:** If  $\sum_{n=1}^{\infty} P(A_n) < \infty$  then  $P(A_n \text{ i.o.}) = 0$ .
- ▶ **Second Borel-Cantelli lemma:** If  $A_n$  are independent, then  $\sum_{n=1}^{\infty} P(A_n) = \infty$  implies  $P(A_n \text{ i.o.}) = 1$ .

- ▶ **Theorem:**  $X_n \rightarrow X$  in probability if and only if for every subsequence of the  $X_n$  there is a further subsequence converging a.s. to  $X$ .

# Convergence in probability $\Rightarrow$ subsequential a.s. convergence

- ▶ **Theorem:**  $X_n \rightarrow X$  in probability if and only if for every subsequence of the  $X_n$  there is a further subsequence converging a.s. to  $X$ .
- ▶ **Main idea of proof:** Consider event  $E_n$  that  $X_n$  and  $X$  differ by  $\epsilon$ . Do the  $E_n$  occur i.o.? Use Borel-Cantelli.

## Pairwise independence example

- ▶ **Theorem:** Suppose  $A_1, A_2, \dots$  are pairwise independent and  $\sum P(A_n) = \infty$ , and write  $S_n = \sum_{i=1}^n 1_{A_i}$ . Then the ratio  $S_n/ES_n$  tends a.s. to 1.



## Pairwise independence example

- ▶ **Theorem:** Suppose  $A_1, A_2, \dots$  are pairwise independent and  $\sum P(A_n) = \infty$ , and write  $S_n = \sum_{i=1}^n 1_{A_i}$ . Then the ratio  $S_n/ES_n$  tends a.s. to 1.
- ▶ **Main idea of proof:** First, pairwise independence implies that variances add. Conclude (by checking term by term) that  $\text{Var}S_n \leq ES_n$ . Then Chebyshev implies

$$P(|S_n - ES_n| > \delta ES_n) \leq \text{Var}(S_n)/(\delta ES_n)^2 \rightarrow 0,$$

which gives us convergence in probability.

## Pairwise independence example

- ▶ **Theorem:** Suppose  $A_1, A_2, \dots$  are pairwise independent and  $\sum P(A_n) = \infty$ , and write  $S_n = \sum_{i=1}^n 1_{A_i}$ . Then the ratio  $S_n/ES_n$  tends a.s. to 1.
- ▶ **Main idea of proof:** First, pairwise independence implies that variances add. Conclude (by checking term by term) that  $\text{Var}S_n \leq ES_n$ . Then Chebyshev implies

$$P(|S_n - ES_n| > \delta ES_n) \leq \text{Var}(S_n)/(\delta ES_n)^2 \rightarrow 0,$$

which gives us convergence in probability.

- ▶ Second, take a smart subsequence. Let  $n_k = \inf\{n : ES_n \geq k^2\}$ . Use Borel Cantelli to get a.s. convergence along this subsequence. Check that convergence along this subsequence deterministically implies the non-subsequential convergence.

Laws of large numbers: Borel-Cantelli applications

Strong law of large numbers

Laws of large numbers: Borel-Cantelli applications

Strong law of large numbers

- ▶ **Theorem (strong law):** If  $X_1, X_2, \dots$  are i.i.d. real-valued random variables with expectation  $m$  and  $A_n := n^{-1} \sum_{i=1}^n X_i$  are the *empirical means* then  $\lim_{n \rightarrow \infty} A_n = m$  almost surely.

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.



## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .
- ▶  $E[A_n^4] = n^{-4} E[S_n^4] = n^{-4} E[(X_1 + X_2 + \dots + X_n)^4]$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .
- ▶  $E[A_n^4] = n^{-4} E[S_n^4] = n^{-4} E[(X_1 + X_2 + \dots + X_n)^4]$ .
- ▶ Expand  $(X_1 + \dots + X_n)^4$ . Five kinds of terms:  $X_i X_j X_k X_l$  and  $X_i X_j X_k^2$  and  $X_i X_j^3$  and  $X_i^2 X_j^2$  and  $X_i^4$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .
- ▶  $E[A_n^4] = n^{-4} E[S_n^4] = n^{-4} E[(X_1 + X_2 + \dots + X_n)^4]$ .
- ▶ Expand  $(X_1 + \dots + X_n)^4$ . Five kinds of terms:  $X_i X_j X_k X_l$  and  $X_i X_j X_k^2$  and  $X_i X_j^3$  and  $X_i^2 X_j^2$  and  $X_i^4$ .
- ▶ The first three terms all have expectation zero. There are  $\binom{n}{2}$  of the fourth type and  $n$  of the last type, each equal to at most  $K$ . So  $E[A_n^4] \leq n^{-4} \left( 6 \binom{n}{2} + n \right) K$ .

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 \geq 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .
- ▶  $E[A_n^4] = n^{-4} E[S_n^4] = n^{-4} E[(X_1 + X_2 + \dots + X_n)^4]$ .
- ▶ Expand  $(X_1 + \dots + X_n)^4$ . Five kinds of terms:  $X_i X_j X_k X_l$  and  $X_i X_j X_k^2$  and  $X_i X_j^3$  and  $X_i^2 X_j^2$  and  $X_i^4$ .
- ▶ The first three terms all have expectation zero. There are  $\binom{n}{2}$  of the fourth type and  $n$  of the last type, each equal to at most  $K$ . So  $E[A_n^4] \leq n^{-4} \left( 6 \binom{n}{2} + n \right) K$ .
- ▶ Thus  $E[\sum_{n=1}^{\infty} A_n^4] = \sum_{n=1}^{\infty} E[A_n^4] < \infty$ . So  $\sum_{n=1}^{\infty} A_n^4 < \infty$  (and hence  $A_n \rightarrow 0$ ) with probability 1.

## General proof of strong law

- ▶ Suppose  $X_k$  are i.i.d. with finite mean. Let  $Y_k = X_k 1_{|X_k| \leq k}$ . Write  $T_n = Y_1 + \dots + Y_n$ . **Claim:**  $X_k = Y_k$  all but finitely often a.s. so suffices to show  $T_n/n \rightarrow \mu$ . (Borel Cantelli, expectation of positive r.v. is area between cdf and line  $y = 1$ )

## General proof of strong law

- ▶ Suppose  $X_k$  are i.i.d. with finite mean. Let  $Y_k = X_k 1_{|X_k| \leq k}$ . Write  $T_n = Y_1 + \dots + Y_n$ . **Claim:**  $X_k = Y_k$  all but finitely often a.s. so suffices to show  $T_n/n \rightarrow \mu$ . (Borel Cantelli, expectation of positive r.v. is area between cdf and line  $y = 1$ )
- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to prove it?



## General proof of strong law

- ▶ Suppose  $X_k$  are i.i.d. with finite mean. Let  $Y_k = X_k 1_{|X_k| \leq k}$ . Write  $T_n = Y_1 + \dots + Y_n$ . **Claim:**  $X_k = Y_k$  all but finitely often a.s. so suffices to show  $T_n/n \rightarrow \mu$ . (Borel Cantelli, expectation of positive r.v. is area between cdf and line  $y = 1$ )
- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to prove it?
- ▶ **Observe:**  $\text{Var}(Y_k) \leq E(Y_k^2) = \int_0^{\infty} 2yP(|Y_k| > y)dy \leq \int_0^k 2yP(|X_1| > y)dy$ . Use Fubini (interchange sum/integral, since everything positive)

$$\begin{aligned} \sum_{k=1}^{\infty} E(Y_k^2)/k^2 &\leq \sum_{k=1}^{\infty} k^{-2} \int_0^{\infty} 1_{(y < k)} 2yP(|X_1| > y)dy = \\ &\int_0^{\infty} \left( \sum_{k=1}^{\infty} k^{-2} 1_{(y < k)} \right) 2yP(|X_1| > y)dy. \end{aligned}$$

Since  $E|X_1| = \int_0^{\infty} P(|X_1| > y)dy$ , complete proof of claim by showing that if  $y \geq 0$  then  $2y \sum_{k > y} k^{-2} \leq 4$ .

## General proof of strong law

- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to use it?

## General proof of strong law

- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to use it?
- ▶ Consider subsequence  $k(n) = [\alpha^n]$  for arbitrary  $\alpha > 1$ . Using Chebyshev, if  $\epsilon > 0$  then

$$\begin{aligned} \sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) &\leq \epsilon^{-1} \sum_{n=1}^{\infty} \text{Var}(T_{k(n)})/k(n)^2 \\ &= \epsilon^{-2} \sum_{n=1}^{\infty} k(n)^{-2} \sum_{m=1}^{k(n)} \text{Var}(Y_m) = \epsilon^{-2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n:k(n) \geq m} k(n)^{-2}. \end{aligned}$$

# General proof of strong law

- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to use it?
- ▶ Consider subsequence  $k(n) = [\alpha^n]$  for arbitrary  $\alpha > 1$ . Using Chebyshev, if  $\epsilon > 0$  then

$$\begin{aligned} \sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) &\leq \epsilon^{-1} \sum_{n=1}^{\infty} \text{Var}(T_{k(n)})/k(n)^2 \\ &= \epsilon^{-2} \sum_{n=1}^{\infty} k(n)^{-2} \sum_{m=1}^{k(n)} \text{Var}(Y_m) = \epsilon^{-2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n:k(n) \geq m} k(n)^{-2}. \end{aligned}$$

- ▶ **Sum series:**

$$\sum_{n:\alpha^n \geq m} [\alpha^n]^{-2} \leq 4 \sum_{n:\alpha^n \geq m} \alpha^{-2n} \leq 4(1 - \alpha^{-2})^{-1} m^{-2}.$$

# General proof of strong law

- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to use it?
- ▶ Consider subsequence  $k(n) = [\alpha^n]$  for arbitrary  $\alpha > 1$ . Using Chebyshev, if  $\epsilon > 0$  then

$$\begin{aligned} \sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) &\leq \epsilon^{-1} \sum_{n=1}^{\infty} \text{Var}(T_{k(n)})/k(n)^2 \\ &= \epsilon^{-2} \sum_{n=1}^{\infty} k(n)^{-2} \sum_{m=1}^{k(n)} \text{Var}(Y_m) = \epsilon^{-2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n:k(n) \geq m} k(n)^{-2}. \end{aligned}$$

- ▶ **Sum series:**

$$\sum_{n:\alpha^n \geq m} [\alpha^n]^{-2} \leq 4 \sum_{n:\alpha^n \geq m} \alpha^{-2n} \leq 4(1 - \alpha^{-2})^{-1} m^{-2}.$$

- ▶ **Combine computations (observe RHS below is finite):**

$$\sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) \leq 4(1 - \alpha^{-2})^{-1} \epsilon^{-2} \sum_{m=1}^{\infty} E(Y_m^2) m^{-2}.$$

# General proof of strong law

- ▶ **Claim:**  $\sum_{k=1}^{\infty} \text{Var}(Y_k)/k^2 \leq 4E|X_1| < \infty$ . How to use it?
- ▶ Consider subsequence  $k(n) = [\alpha^n]$  for arbitrary  $\alpha > 1$ . Using Chebyshev, if  $\epsilon > 0$  then

$$\begin{aligned} \sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) &\leq \epsilon^{-2} \sum_{n=1}^{\infty} \text{Var}(T_{k(n)})/k(n)^2 \\ &= \epsilon^{-2} \sum_{n=1}^{\infty} k(n)^{-2} \sum_{m=1}^{k(n)} \text{Var}(Y_m) = \epsilon^{-2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n:k(n) \geq m} k(n)^{-2}. \end{aligned}$$

- ▶ **Sum series:**

$$\sum_{n:\alpha^n \geq m} [\alpha^n]^{-2} \leq 4 \sum_{n:\alpha^n \geq m} \alpha^{-2n} \leq 4(1 - \alpha^{-2})^{-1} m^{-2}.$$

- ▶ **Combine computations (observe RHS below is finite):**

$$\sum_{n=1}^{\infty} P(|T_{k(n)} - ET_{k(n)}| > \epsilon k(n)) \leq 4(1 - \alpha^{-2})^{-1} \epsilon^{-2} \sum_{m=1}^{\infty} E(Y_m^2) m^{-2}.$$

- ▶ Since  $\epsilon$  is arbitrary, get  $(T_{k(n)} - ET_{k(n)})/k(n) \rightarrow 0$  a.s.

## General proof of strong law

- ▶ Conclude by taking  $\alpha \rightarrow 1$ . This finishes the case that the  $X_1$  are a.s. positive.

## General proof of strong law

- ▶ Conclude by taking  $\alpha \rightarrow 1$ . This finishes the case that the  $X_1$  are a.s. positive.
- ▶ Can extend to the case that  $X_1$  is a.s. positive within infinite mean.



## General proof of strong law

- ▶ Conclude by taking  $\alpha \rightarrow 1$ . This finishes the case that the  $X_1$  are a.s. positive.
- ▶ Can extend to the case that  $X_1$  is a.s. positive within infinite mean.
- ▶ Generally, can consider  $X_1^+$  and  $X_1^-$ , and it is enough if one of them has a finite mean.