Empirical Processes with Applications in Statistics 2. Stochastic Convergence in Metric Spaces

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Overview

Introductory lecture: motivation why to consider weak convergence of non-measurable elements in $\ell^{\infty}(T)$ equipped with supremum distance

This lecture: weak convergence

- 1. first for general metric spaces (\mathbb{D}, d)
- 2. then specialized to $(\ell^{\infty}(T), \|\cdot\|_{\infty})$

Theory goes back to Hoffmann-Jørgensen (1984, 1991) and others. Main sources for this lecture:

- van der Vaart and Wellner (1996, Part 1)
- van der Vaart (1998, Chapter 18)

See historic notes in van der Vaart and Wellner (1996, pp. 75-78)

Stochastic Convergence in Metric Spaces

General metric spaces

Space of bounded functions

Outer expectation

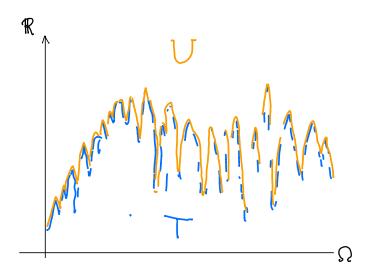
- Probability space (Ω, A, P)
- map $T: \Omega \to [-\infty, \infty]$, not necessarily measurable

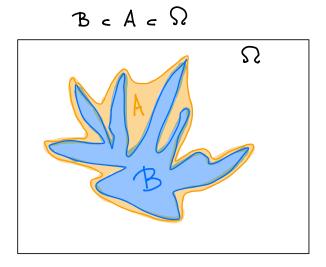
Definition: Outer expectation.

$$\mathsf{E}^*[T] = \inf \{ \mathsf{E}[U] \mid U : \Omega \to [-\infty, \infty] \text{ measurable,} \\ U \ge T \text{ and } \mathsf{E}[U] \text{ exists} \}$$

Measurable cover: Borel measurable $T^* : \Omega \to [-\infty, \infty]$ such that $T \leq T^* \leq U$ almost surely for any *U* as above

► $E^*[T] = E[T^*]$ (provided the latter expectation exists in $[-\infty, \infty]$)





Inner expectation:

$$\mathsf{E}_*[T] = -\,\mathsf{E}^*[-T]$$

Inner and outer probability:

$$\begin{aligned} \mathsf{P}^*(B) &= \mathsf{E}^*[\mathbb{1}_B] = \inf \left\{ \mathsf{P}(A) \mid B \subset A, \text{ measurable } A \subset \Omega \right\} \\ \mathsf{P}_*(B) &= \mathsf{E}_*[\mathbb{1}_B] = \mathsf{1} - \mathsf{P}^*(\Omega \setminus B) \end{aligned}$$

Some care is required:

- $(S + T)^* \leq S^* + T^*$, but no equality in general
- Fubini's theorem no longer works
- "Law" of a random variable depends on underlying probability space
 - ▶ iid samples X₁, X₂,..., X_n,... will be seen as coordinate projections on the product space (Xⁿ, Aⁿ, Pⁿ) or the sequence space (X[∞], A[∞], P[∞])

Stochastic convergence

- Metric space (D, d), Borel *σ*-field
- Random elements X_n, X : (Ω, A, P) → D X Borel measurable, but X_n not necessarily

Definition: Weak convergence. $X_n \rightsquigarrow X$ if

$$\lim_{n\to\infty}\mathsf{E}^*[f(X_n)]=\mathsf{E}[f(X)]$$

for every bounded, continuous $f : \mathbb{D} \to \mathbb{R}$

Other modes of convergence

Convergence:

- ▶ in (outer) probability: $X_n \xrightarrow{P} X$ if $P[d(X_n, X)^* > \varepsilon] \to 0$ for all $\varepsilon > 0$
- (outer) almost surely: $X_n \xrightarrow{as*} X$ if $d(X_n, X)^* \to 0$ a.s.

The usual implications hold:

- $\blacktriangleright X_n \xrightarrow{as*} X \text{ implies } X_n \xrightarrow{P} X$
- $\blacktriangleright X_n \xrightarrow{\mathsf{P}} X \text{ implies } X_n \rightsquigarrow X$
- $X_n \xrightarrow{\mathsf{P}} c$ (constant) if and only if $X_n \rightsquigarrow c$

Characterizations of weak convergene

Portmanteau lemma. Are equivalent:

(i)
$$X_n \rightsquigarrow X$$

(v) .

(ii) $\liminf_{n\to\infty} P_*(X_n \in G) \ge P(X \in G)$ for every open $G \subset \mathbb{D}$

- (iii) $\limsup_{n\to\infty} P^*(X_n \in F) \leq P(X \in F)$ for every closed $F \subset \mathbb{D}$
- (iv) For Borel sets $B \subset \mathbb{D}$ such that $P(X \in \partial B) = 0$,

$$\lim_{n\to\infty} \mathsf{P}_*(X_n \in B) = \lim_{n\to\infty} \mathsf{P}^*(X_n \in B) = \mathsf{P}(X \in B)$$

Characterizations mostly useful in proofs

Choice of state space not so important

Corollary

Subset $\mathbb{D}_0 \subset \mathbb{D}$ equipped with the same metric *d*

• Maps
$$X_n, X : \Omega \to \mathbb{D}_0$$

Then

$$X_n \rightsquigarrow X \text{ in } \mathbb{D} \iff X_n \rightsquigarrow X \text{ in } \mathbb{D}_0$$

Example:

$$\blacktriangleright \mathbb{D} = \ell^{\infty}([0,1])$$

•
$$\mathbb{D}_0 = C([0, 1]) \text{ or } \mathbb{D}_0 = \mathcal{D}([0, 1])$$

Extracting new convergence relations

Continuous Mapping Theorem

- Metric spaces D, E
- $g: \mathbb{D} \to \mathbb{E}$ is continuous at every $x \in \mathbb{D}_0 \subset \mathbb{D}$

 $\blacktriangleright X_n \rightsquigarrow X \text{ in } \mathbb{D}$

If X takes values in \mathbb{D}_0 , then $g(X_n) \rightsquigarrow g(X)$

- ► Extended continuous mapping theorem: mappings depend on *n* but $g_n(x_n) \rightarrow g(x)$ for sufficiently many sequences $x_n \rightarrow x$
- Will serve to prove the functional delta method

Extracting weakly convergent subsequences

Prohorov's Theorem. If the sequence of maps $X_n : \Omega \to \mathbb{D}$ is

asymptotically measurable

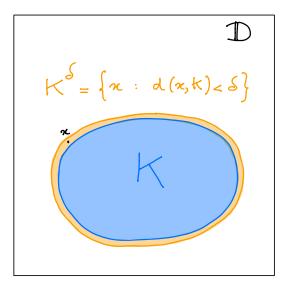
 $\lim_{n\to\infty} E^*[f(X_n)] - E_*[f(X_n)] = 0$ for every bounded, continuous $f : \mathbb{D} \to \mathbb{R}$

asymptotically tight

For every $\varepsilon > 0$ there exists a compact $K \subset \mathbb{D}$ such that, for every $\delta > 0$, we have $\liminf_{n \to \infty} P_*[\exists y \in K : d(X_n, y) < \delta] \ge 1 - \varepsilon$

then it has a *subsequence* that converges weakly to a *tight X*.

For every $\varepsilon > 0$ there exists a compact $K \subset \mathbb{D}$ such that $P(X \in K) \ge 1 - \varepsilon$.



Strategy to prove weak convergence $X_n \rightsquigarrow X$:

- 1. Show that every subsequence of X_n has a further subsequence that convergence weakly
 - Via Prohorov's theorem
- 2. Show that all weak limits that can thus arise are the identical
 - For stochastic process: via finite-dimensional distributions

Little-oh calculus

Suppose D is actually a *Banach space* Vector space over \mathbb{R} equipped with a norm $\|\cdot\|$ such that the metric $d(x, y) = \|x - y\|$ is complete

Slutsky's lemma If $X_n \rightsquigarrow X$ and $Y_n \rightsquigarrow c \in \mathbb{D}$ and if X is tight, then

$$X_n + Y_n \rightsquigarrow X + c$$

Common case: $c = 0 \in \mathbb{D}$. Then we write $Y_n = o_P(1)$ and thus

$$X_n \rightsquigarrow X \text{ tight } \implies X_n + o_P(1) \rightsquigarrow X$$

Stochastic Convergence in Metric Spaces

General metric spaces

Space of bounded functions

Space of bounded functions

Let \mathbb{D} be the space of bounded real functions on some set T:

$$\ell^{\infty}(T) = \left\{ z: T \to \mathbb{R} \mid \sup_{t \in T} |z(t)| < \infty \right\}$$

- ► Banach space with norm $||z||_{\infty} = \sup_{t \in T} |z(t)|$
- Supremum distance $d(z_1, z_2) = ||z_1 z_2||_{\infty} = \sup_{t \in T} |z_1(t) z_2(t)|$
- Non-separable when T is uncountable: Borel σ -field is very large
- Natural space in which to study empirical processes such as

$$f \mapsto \mathbb{P}_n f = \frac{1}{n} \sum_{i=1}^n f(X_i)$$

indexed by $f \in \mathcal{F} = T$, for i.i.d. X_1, X_2, \ldots taking values in some measurable space (X, \mathcal{A}, P) and \mathcal{F} some subset of $L_1(P)$ or $L_2(P)$

Theorem: Weak convergence in $\ell^{\infty}(T)$ **.** Let $X_n : \Omega \to \ell^{\infty}(T)$. Then

 X_n converges weakly to some tight limit

if and only if the following two properties hold:

- 1. X_n is asymptotically tight
- 2. for every $(t_1, \ldots, t_k) \in T^k$, the random vectors $(X_n(t_1), \ldots, X_n(t_k))$ converge weakly in \mathbb{R}^k

Special case of proof strategy on slide 15

- 1. Asymptotic tightness: by controlling the oscillations of the trajectories (Arzèla–Ascoli)
- 2. Convergence of finite-dimensional distributions: via classical limit theorems

Theorem: Asymptotic tightness in $\ell^{\infty}(T)$.

Let $X_n : \Omega \to \ell^{\infty}(T)$ be such that $X_n(t) : \Omega \to \mathbb{R}$ is asymptotically tight for every $t \in T$. Are equivalent:

(i) X_n is asymptotically tight

(ii) For every $\varepsilon, \eta > 0$ there exists a finite partition $T = \bigcup_{i=1}^{k} T_i$ with

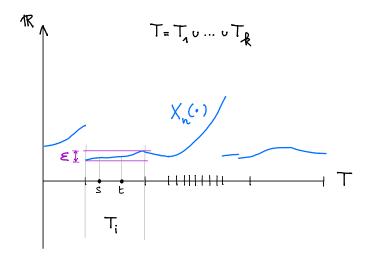
$$\limsup_{n\to\infty}\mathsf{P}^*\left[\max_{i=1,\dots,k}\sup_{s,t\in T_i}|X_n(s)-X_n(t)|>\varepsilon\right]<\eta$$

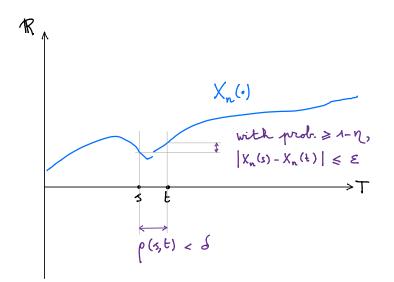
(iii) there exists a semimetric ρ on T such that (T, ρ) is totally bounded and such that for every $\varepsilon, \eta > 0$ there exists $\delta > 0$ with

$$\limsup_{n\to\infty}\mathsf{P}^*\left[\sup_{\rho(s,t)<\delta}|X_n(s)-X_n(t)|>\varepsilon\right]<\eta$$

i.e., X_n is asymptotically uniformly ρ -equicontinuous in probability

 (T, ρ) totally bounded: for every $\delta > 0$, there exists finitely many $t_1, \ldots, t_\ell \in T$ such that for every $t \in T$, there is t_j with $\rho(t, t_j) < \delta$





- If, moreover, X_n → X, then almost all trajectories t → X(t) are uniformly ρ-continuous
- If, moreover, X is a Gaussian process, then the following semimetric always works:

$$\rho(\boldsymbol{s},t) = \left(\mathsf{E}\left[\{\boldsymbol{X}(\boldsymbol{s}) - \boldsymbol{X}(t)\}^2\right]\right)^{1/2}, \qquad \boldsymbol{s},t \in \mathcal{T}$$

- Techniques to control the probabilities in (ii) and (iii):
 - maximal inequalities
 - symmetrization
 - entropy: bracketing and covering numbers
 - ▶ ...

Summary

Classical theory of weak convergence in metric spaces as in Billingsley (1968) works well in *separable metric spaces*

- C([0, 1]) with supremum distance
- D([0, 1]) with Skorohod topology

The space $\ell^{\infty}(T)$ in empirical process theory is non-separable and requires handling non-measurable mappings: *Hoffmann-Jørgensen theory*

Classical results can be mostly recovered:

- Portmanteau lemma
- Continuous mapping theorem
- Prohorov's theorem
- Slutsky's lemma
- Tightness criteria for sequences of stochastic processes

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