

The background of the slide features a large, faint watermark of the University of Bonn seal. The seal is circular and contains a central figure of a seated woman, likely a personification of Wisdom or Philosophy, holding a book. Above her are three portraits of men. The seal is surrounded by Latin text: 'SIGILLUM UNIVERSITATIS BONNENSIS' at the top and 'MDCCCXXXIII' at the bottom.

Probability Theory

10. Introduction to weak convergence

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Definition

- ▶ Preliminary remark: For the expected value with respect to P we now write

$$P[X] := \int X dP.$$

- ▶ $\mathcal{P}(E)$: probability measure on $\mathcal{B}(E)$;
 $\mathcal{P}_{\leq 1}(E)$: finite measure μ on $\mathcal{B}(E)$ with $\mu(E) \leq 1$.
 $\mathcal{C}_b(E)$: continuous, bounded functions $E \rightarrow \mathbb{R}$
 $\mathcal{C}_c(E)$: continuous functions $E \rightarrow \mathbb{R}$ with compact support

Definition 9.1

- ▶ Weak convergence of $P_1, P_2, \dots \in \mathcal{P}(E)$ to $P \in \mathcal{P}(E)$:

$$(P_n \xrightarrow{n \rightarrow \infty} P) \quad : \iff \quad P_n[f] \xrightarrow{n \rightarrow \infty} P[f], \quad f \in \mathcal{C}_b(E)$$

- ▶ Vague convergence of $\mu_1, \mu_2, \dots \in \mathcal{P}_{\leq 1}$ to μ :

$$(\mu_n \xrightarrow{n \rightarrow \infty} \nu \mu) \quad : \iff \quad \mu_n[f] \xrightarrow{n \rightarrow \infty} \mu[f], \quad f \in \mathcal{C}_c(E)$$

- ▶ Convergence in distribution of X_1, X_2, \dots to X :

$$\begin{aligned} X_n \xrightarrow{n \rightarrow \infty} X \quad : \iff \quad (X_n)_* P_n \xrightarrow{n \rightarrow \infty} X_* P \\ \iff \quad P[f(X_n)] \xrightarrow{n \rightarrow \infty} P[f(X)], \quad f \in \mathcal{C}_b(E). \end{aligned}$$

Examples

- ▶ $x, x_1, x_2, \dots \in \mathbb{R}$ with $x_n \xrightarrow{n \rightarrow \infty} x$ and

$P = \delta_x, P_1 = \delta_{x_1}, P_2 = \delta_{x_2}, \dots$. Then, $P_n \xrightarrow{n \rightarrow \infty} P$, since

$$P_n[f] = f(x_n) \xrightarrow{n \rightarrow \infty} f(x) = P[f], \quad f \in C_b(\mathbb{R}).$$

With $x_n = n$, we have $P_n \xrightarrow{n \rightarrow \infty} \nu_0$, since

$$P_n[f] = f(x_n) \xrightarrow{n \rightarrow \infty} 0 = \nu_0[f], \quad f \in C_c(\mathbb{R}).$$

- ▶ Let X, X_1, X_2, \dots identically distributed. Then $X_n \xrightarrow{n \rightarrow \infty} X$.
- ▶ Central Limit Theorem, for example as the theorem of

deMoivre-Laplace: For $p \in (0, 1)$ let

$X_n \sim B(n, p), n = 1, 2, \dots$ and $X \sim N(0, 1)$. Then,

$$\frac{X_n - np}{\sqrt{np(1-p)}} \xrightarrow{n \rightarrow \infty} X.$$

Uniqueness of the limit

- Lemma 9.4: Let $P, Q, P_1, P_2, \dots \in \mathcal{P}(E)$ with $P_n \xrightarrow{n \rightarrow \infty} P$ and $P_n \xrightarrow{n \rightarrow \infty} Q$. Then $P = Q$.

Proof: to show: $P(A) = Q(A)$ for A closed. We set

$$r(x, A) := \inf_{y \in A} r(x, y)$$

and

$$f_m(x) \mapsto (1 - m \cdot r(x, A))^+$$

for $m = 1, 2, \dots$. Then $f_m \xrightarrow{m \rightarrow \infty} 1_A$. Further

$$P(A) = \lim_{m \rightarrow \infty} P[f_m] = \lim_{m \rightarrow \infty} \lim_{n \rightarrow \infty} P_n[f_m] = \lim_{m \rightarrow \infty} Q[f_m] = Q(A)$$

Convergence in probability and weak convergence

- Proposition 9.5: X, X_1, X_2, \dots rvs. If $X_n \xrightarrow{n \rightarrow \infty}_p X$, then $X_n \xrightarrow{n \rightarrow \infty} X$. If X is constant, then the inverse is also true.

Proof: Suppose $X_n \xrightarrow{n \rightarrow \infty}_p X$ and

$\lim_{n \rightarrow \infty} P[f(X_n)] \neq P[f(X)]$ for some $f \in \mathcal{C}_b(E)$. Choose a subsequence $(n_k)_{k=1,2,\dots}$ and $\varepsilon > 0$ with

$\lim_{k \rightarrow \infty} |P[f(X_{n_k})] - P[f(X)]| > \varepsilon$. Select a further

subsequence $(n_{k_\ell})_{\ell=1,2,\dots}$ with $X_{n_{k_\ell}} \xrightarrow{\ell \rightarrow \infty}_{as} X$. Then,

$$\lim_{\ell \rightarrow \infty} P[f(X_{n_{k_\ell}})] = P[f(X)]$$

Convergence in probability and weak convergence

- ▶ Proposition 9.5: X, X_1, X_2, \dots rvs. If $X_n \xrightarrow{n \rightarrow \infty} \rightarrow_p X$, then $X_n \xrightarrow{n \rightarrow \infty} \rightarrow X$. If X is constant, then the inverse is also true.

Proof: weak convergence \Rightarrow conv. in probability with $X = c$:

Select $x \mapsto r(x, c) \wedge 1 \in \mathcal{C}_b(E)$, such that

$$P[r(X_n, c) \wedge 1] \xrightarrow{n \rightarrow \infty} P[r(X, c) \wedge 1] = 0.$$

From this, $X_n \xrightarrow{n \rightarrow \infty} \rightarrow_p X$ follows.

The Portmanteau Theorem

► Theorem 10.6: X, X_1, X_2, \dots rvs. Equivalent are:

- (i) $X_n \xrightarrow{n \rightarrow \infty} X$;
- (ii) $P[f(X_n)] \xrightarrow{n \rightarrow \infty} P[f(X)]$ for $f \in \mathcal{C}_b(E)$ Lipschitz;
- (iii) $\liminf_{n \rightarrow \infty} P(X_n \in G) \geq P(X \in G)$ for all open $G \subseteq E$.
- (iv) $\limsup_{n \rightarrow \infty} P(X_n \in F) \leq P(X \in F)$ for all closed $F \subseteq E$.
- (v) $\lim_{n \rightarrow \infty} P(X_n \in B) = P(X \in B)$ for all $B \in \mathcal{B}(E)$ with $P(\partial B) = 0$.

Proof: (i) \Rightarrow (ii): clear; (iii) \iff (iv) clear;

(ii) \Rightarrow (iv) $F \subseteq E$ closed, $\varepsilon_k \downarrow 0$ and

$$f_k(x) = \left(1 - \frac{1}{\varepsilon_k} r(x, F)\right)^+.$$

$$\limsup_{n \rightarrow \infty} P(X_n \in F) \leq \limsup_{n \rightarrow \infty} P[f_k(X_n)] = P[f_k(X)] \downarrow P(X \in F).$$

The Portmanteau Theorem

► Theorem 10.6: X, X_1, X_2, \dots rvs. Equivalent are:

- (i) $X_n \xrightarrow{n \rightarrow \infty} X$;
- (ii) $P[f(X_n)] \xrightarrow{n \rightarrow \infty} P[f(X)]$ for $f \in \mathcal{C}_b(E)$ Lipschitz;
- (iii) $\liminf_{n \rightarrow \infty} P(X_n \in G) \geq P(X \in G)$ for all open $G \subseteq E$.
- (iv) $\limsup_{n \rightarrow \infty} P(X_n \in F) \leq P(X \in F)$ for all closed $F \subseteq E$.
- (v) $\lim_{n \rightarrow \infty} P(X_n \in B) = P(X \in B)$ for all $B \in \mathcal{B}(E)$ with $P(\partial B) = 0$.

Proof: (iii) \Rightarrow (i) $f \in \mathcal{C}_b, 0 \leq f \leq c$:

$$\begin{aligned} P[f(X)] &= \int_0^\infty P(f(X) > t) dt \leq \int_0^\infty \liminf_{n \rightarrow \infty} P(f(X_n) > t) dt \\ &\leq \liminf_{n \rightarrow \infty} \int_0^\infty P(f(X_n) > t) dt = \liminf_{n \rightarrow \infty} P[f(X_n)], \end{aligned}$$

$$\limsup_{n \rightarrow \infty} P[f(X_n)] = c - \liminf_{n \rightarrow \infty} P[-f(X_n) + c] \leq P[f(X)]$$

The Portmanteau Theorem

► 10.6: X, X_1, X_2, \dots ZV. Equivalent are:

(i) $X_n \xrightarrow{n \rightarrow \infty} X$;

(ii) $P[f(X_n)] \xrightarrow{n \rightarrow \infty} P[f(X)]$ for $f \in \mathcal{C}_b(E)$ Lipschitz;

(iii) $\liminf_{n \rightarrow \infty} P(X_n \in G) \geq P(X \in G)$ for all open $G \subseteq E$.

(iv) $\limsup_{n \rightarrow \infty} P(X_n \in F) \leq P(X \in F)$ for all closed $F \subseteq E$.

(v) $\lim_{n \rightarrow \infty} P(X_n \in B) = P(X \in B)$ for all $B \in \mathcal{B}(E)$ with $P(\partial B) = 0$.

(iii), (iv) \Rightarrow (v) For $B \in \mathcal{B}(E)$ is

$$P(X \in B^\circ) \leq \liminf_{n \rightarrow \infty} P(X_n \in B^\circ) \leq \limsup_{n \rightarrow \infty} P(X_n \in \bar{B}) \leq P(X_n \in \bar{B}).$$

(v) \Rightarrow (iv) For $F \subseteq E$ closed let $F^\varepsilon := \{x \in E : r(x, F) \leq \varepsilon\}$.

Then $P(X \in \partial F^\varepsilon) = 0$ for Lebesgue-almost every ε . So,

$$\limsup_{n \rightarrow \infty} P(X_n \in F) \leq \limsup_{n \rightarrow \infty} P(X_n \in F^{\varepsilon_k}) = P(X \in F^{\varepsilon_k}) \downarrow P(X \in F).$$

Convergence of distribution functions

- Corollary 9.7: $P, P_1, P_2, \dots \in \mathcal{P}(\mathbb{R})$ with distribution functions F, F_1, F_2, \dots . Then,

$$P_n \xrightarrow{n \rightarrow \infty} P \iff \left(F_n(x) \xrightarrow{n \rightarrow \infty} F(x) \text{ for all continuity points } x \text{ of } F. \right)$$

' \Rightarrow ': If x is the continuity point of F , then

$P(\partial(-\infty; x]) = P(\{x\}) = 0$. Also

$$F_n(x) = P_n((-\infty; x]) \xrightarrow{n \rightarrow \infty} P((-\infty; x]) = F(x).$$

' \Leftarrow ': See manuscript;

The Theorem of deMoivre-Laplace

- ▶ For $X_n \sim B(n, p)$,

$$P\left(\frac{X_n - np}{\sqrt{np(1-p)}} \leq x\right) \xrightarrow{n \rightarrow \infty} \Phi(x),$$

where Φ is the distribution function of the standard normal distribution is.

This also means

$$\frac{X_n - np}{\sqrt{np(1-p)}} \xrightarrow{N \rightarrow \infty} Z \sim N(0, 1).$$

Slutzky's Theorem

- ▶ Corollary 9.9: $X, X_1, X_2, \dots, Y_1, Y_2, \dots$ rvs. If $X_n \xrightarrow{n \rightarrow \infty} X$ and $r(X_n, Y_n) \xrightarrow{n \rightarrow \infty}_p 0$, then $Y_n \xrightarrow{n \rightarrow \infty} X$.
- ▶ Proof: $f \in \mathcal{C}_b(E)$ Lipschitz. Then

$$|f(x) - f(y)| \leq L \cdot r(x, y) \wedge (2\|f\|_\infty), \quad x, y \in E$$

$$\Rightarrow \limsup_{n \rightarrow \infty} |E[f(X_n) - f(Y_n)]| \leq \limsup_{n \rightarrow \infty} E[L \cdot r(X_n, Y_n) \wedge (2\|f\|_\infty)] = 0.$$

Also,

$$\begin{aligned} \limsup_{n \rightarrow \infty} |E[f(Y_n)] - E[f(X)]| \\ \leq \limsup_{n \rightarrow \infty} |E[f(Y_n)] - E[f(X_n)]| + |E[f(X_n)] - E[f(X)]| = 0. \end{aligned}$$

The Continuous Mapping Theorem

- ▶ Remark: X, X_1, X_2, \dots ZV,

$\varphi : E \rightarrow E'$ continuous. Then

$$X_n \xrightarrow{n \rightarrow \infty} X \quad \Rightarrow \quad \varphi(X_n) \xrightarrow{n \rightarrow \infty} \varphi(X).$$

Indeed: For $g \in \mathcal{C}_b(E')$ we have $g \circ \varphi \in \mathcal{C}_b(E)$, therefore

$$P[g(\varphi(X_n))] \xrightarrow{n \rightarrow \infty} P[g(\varphi(X))].$$

- ▶ Theorem 9.10: X, X_1, X_2, \dots ZV,

$\varphi : E \rightarrow E'$ measurable,

$U_\varphi := \{x : \varphi \text{ discontinuous in } x\} \subseteq E$.

$$P(X \in U_\varphi) = 0, \quad X_n \xrightarrow{n \rightarrow \infty} X \quad \Rightarrow \quad \varphi(X_n) \xrightarrow{n \rightarrow \infty} \varphi(X).$$

Weak and almost sure convergence, Skorohod

- ▶ Theorem 9.11: X, X_1, X_2, \dots ZV. Then, $X_n \xrightarrow{n \rightarrow \infty} X$ holds if and only if there is a probability space exists on which random variables Y, Y_1, Y_2, \dots are defined with $Y_n \xrightarrow{n \rightarrow \infty}_{as} Y$ and $Y \stackrel{d}{=} X, Y_1 \stackrel{d}{=} X_1, Y_2 \stackrel{d}{=} X_2, \dots$
- ▶ Example: If X, X_1, X_2, \dots are iid, then $X_n \xrightarrow{n \rightarrow \infty} x$. With $X = Y_1, Y_2, \dots$ we find $X_n \sim Y_n$ and $Y_n = X$, in particular $Y_n \xrightarrow{n \rightarrow \infty}_{as} X$.