

Lecture 04: Asymptotic Properties I

Statistical Modelling I

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Outline

1. Introduction

2. Modes of Stochastic Convergence

3. Consistency

Introduction

Evaluating estimators

Evaluating an estimator $T = T(X_1, \dots, X_n)$ of θ depends on its *sampling distribution*

This problem often simplifies by considering $T_n \equiv T_n(X_1, \dots, X_n)$ to be a sequence of random variables indexed by $n = 1, 2, 3, \dots$

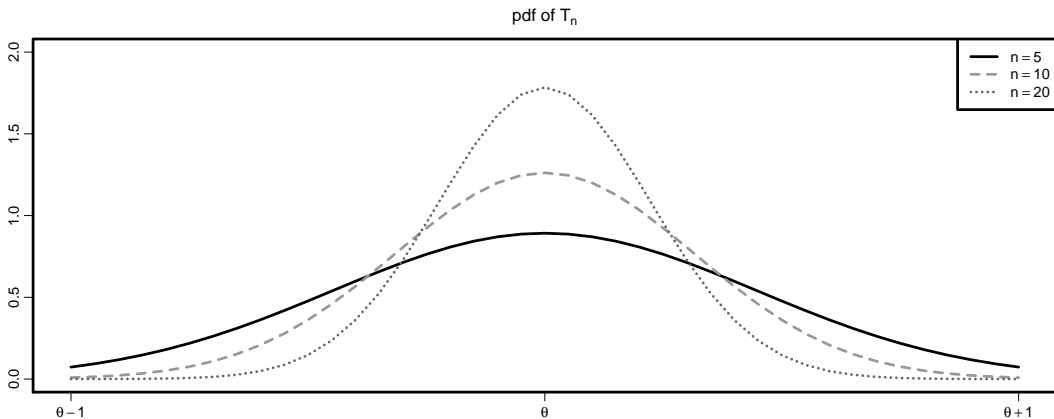
Example: $Y_1, Y_2, \dots \sim N(\theta, 1)$ iid

$$T_n = \frac{1}{n} \sum_{i=1}^n Y_i$$

$$E(T_n) = \theta$$

$$\text{Var}(T_n) = \frac{1}{n}$$

$$T_n \sim N\left(\theta, \frac{1}{n}\right)$$



Modes of Stochastic Convergence

Stochastic Convergence

Just as sequences of numbers x_1, x_2, \dots can converge, sequences of random variables X_1, X_2, \dots exhibit patterns of convergence

- ▶ Convergence almost surely
- ▶ Convergence in probability
- ▶ Convergence in distribution

Convergence almost surely and convergence in probability

X_n converges almost surely to X if

$$P\left(\lim_{n \rightarrow \infty} X_n = X\right) = 1$$

X_n converges in probability to X if, $\forall \epsilon > 0$,

$$\lim_{n \rightarrow \infty} P(|X_n - X| > \epsilon) = 0$$

Convergence in distribution

X_n converges in distribution to X with cdf F_X if

$$\lim_{n \rightarrow \infty} P(X_n \leq x) = F_X(x)$$

at all continuity points x of $F_X(x)$

Portmanteau lemma: The following statements are equivalent:

- $X_n \xrightarrow{d} X$
- $E[f(X_n)] \rightarrow E[f(X)]$ for all bounded and continuous functions $f : \mathbb{R} \rightarrow \mathbb{R}$.

Relationship between modes of stochastic convergence

We have

$$X_n \rightarrow_{as} X \quad \Rightarrow \quad X_n \rightarrow_p X \quad \Rightarrow \quad X_n \rightarrow_d X$$

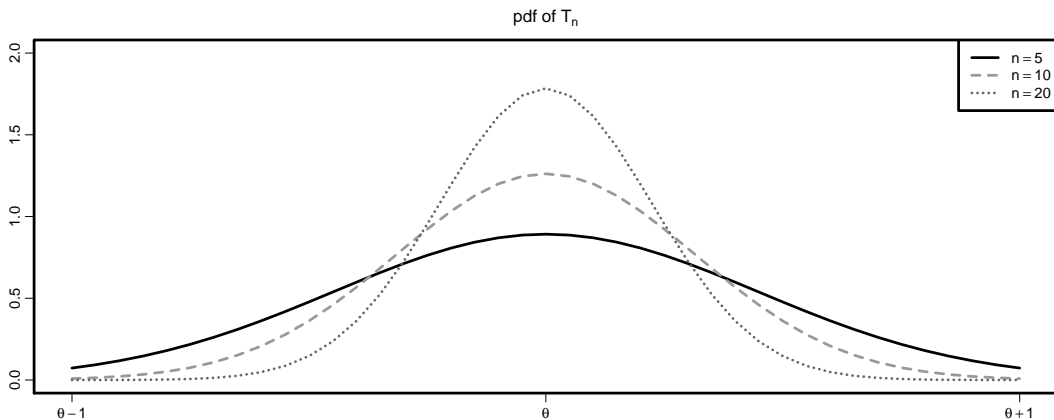
and if $X = c$ is a constant

$$X_n \rightarrow_p X \quad \Leftrightarrow \quad X_n \rightarrow_d X$$

Consistency

Concentration

Estimating θ from $N(\theta, 1)$, we see distribution of T_n concentrates around θ_0



Definition: Consistency

A sequence of estimators $(T_n)_{n \in \mathbb{N}}$ for $g(\theta)$ is called **(weakly) consistent** if for all $\theta \in \Theta$:

$$T_n \xrightarrow{P_\theta} g(\theta) \quad (n \rightarrow \infty)$$

Definition: Asymptotically Unbiased Estimator

A sequence of estimators $(T_n)_{n \in \mathbb{N}}$ for $g(\theta)$ is called **asymptotically unbiased** if for all $\theta \in \Theta$:

$$E_{\theta}(T_n) \rightarrow g(\theta) \quad (n \rightarrow \infty)$$

Markov's Inequality

$$P(|X| \geq a) \leq \frac{E|X|}{a} \quad \forall a > 0$$

Proof

$$a\mathcal{I}(|X| \geq a) \leq |X| \quad \Rightarrow \quad P(|X| \geq a) = E\{\mathcal{I}(|X| \geq a)\} \leq \frac{1}{a}E|X|$$

Lemma: MSE consistency

If T_n is asymptotically unbiased for $g(\theta)$ and $\forall \theta \in \Theta$

$$\text{Var}_{\theta}(T_n) \rightarrow 0 \quad (n \rightarrow \infty)$$

then T_n is consistent for $g(\theta)$

Proof

Applying Markov's inequality, we have

$$\begin{aligned} \forall \epsilon > 0 : \quad P_{\theta}(|T_n - g(\theta)| \geq \epsilon) &= P_{\theta}((T_n - g(\theta))^2 \geq \epsilon^2) \leq \frac{E_{\theta}(T_n - g(\theta))^2}{\epsilon^2} = \frac{MSE_{\theta}(T_n)}{\epsilon^2} \\ &= \frac{1}{\epsilon^2} \underbrace{(Var_{\theta} T_n)}_{\rightarrow 0} + \underbrace{(E_{\theta} T_n - g(\theta))^2}_{\rightarrow 0} \end{aligned}$$

Example: $X_i \sim \text{Bernoulli}(\theta)$, $\theta \in \Theta = [0, 1]$

$$T_n(x_1, \dots, x_n) = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\left. \begin{array}{l} E_{\theta} T_n(X_1, \dots, X_n) = \theta \quad \forall \theta \in \Theta \\ \text{Var}_{\theta}(T_n) = \frac{1}{n} \theta(1 - \theta) \rightarrow 0 \quad (n \rightarrow \infty) \end{array} \right\} \implies T_n \text{ consistent}$$

Summary

- ▶ Consistency is an important property, but does not tell us all about the sampling distribution of T_n
- ▶ To establish approximations to the sampling distribution of T_n , we will appeal to convergence in distribution of sequences of the form $a_n(T_n - g(\theta))$ for deterministic a_n
- ▶ We typically find that a_n is proportional to $SE(T_n)$
- ▶ Many estimators “look like” sample means, so $a_n = \sqrt{n}$ and $N(0, \sigma^2)$ limits are extremely common!