

Epi-Convergence of M-estimators When Objective Functions are Convex

Kosaku Takanashi
Graduate School of Economics
Keio University

The main purpose of this paper is to present an approach, based on Mosco-convergence, for proving the almost sure convergence of the estimation problem defined by convex minimization. This is done under weaker hypotheses than those usually assumed. Mosco-convergence that our approach in this study is based on is weaker topology than uniform convergence. Mosco-convergence ensures the convergence of empirical minimizer to the exact minimizer. Unlike bracketing condition, maximal entropy and sieve, epi-convergence dose not require compactness assumptions on the parameter spaces.

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be probability triple and $\omega \in \Omega$. Let $\Theta \subseteq \mathcal{X}$ be a parameter set in Hilbert space and $\theta \in \Theta$. To show the consistency, we need the convergence of empirical minimizer

$$\arg \min_{\theta} \frac{1}{n} \sum_{i=1}^n \rho(\omega_i, \theta) \rightarrow \arg \min_{\theta} \mathbb{E} [\rho(\omega, \theta)].$$

Consider the uniform convergence

$$\sup_{\theta} \left| \frac{1}{n} \sum_{i=1}^n \rho(\omega_i, \theta) - \mathbb{E} [\rho(\omega, \theta)] \right| \xrightarrow{P} 0$$

that implies consistency and model selection optimality. For uniform convergence, we need impose one of either assumptions (van der Vaart and Wellner, 96, Springer); 1:Totally boundedness of parameter space, 2:Random entropy condition, 3: ρ is finite convex.

These assumptions are little bit restrictive for a functional estimation. Instead of uniform convergence, we shall consider a weaker topology than uniform convergence. We focus on the Mosco topology. Mosco convergence ensures the convergence of empirical minimizer.

To obtain Mosco-convergence of the objective function, we use the theorem of the equivalences between Mosco-convergence, Graph convergence(G-convergence) of subdifferential operators, pointwise convergence of Moreau-Yosida approximation and pointwise convergence of resolvent. Using these equivalences, we can establishe the consistency and weak convergence of a estimator defined on infinite dimensionl parameter space.

Our main results are as follows. Let f_1, f_2, \dots and f_0 be a sequence of proper lsc convex function defined on (Ω, \mathcal{X}) such that

$$f_n(\omega, \theta) \triangleq \frac{1}{n} \sum_{i=1}^n \rho_i(\omega, \theta)$$

$$f_0(\omega, \theta) \triangleq \mathbb{E} [\rho(\omega, \theta)].$$

We can obtain the following results of the consistency and achivement of infimum. There exists a \mathbb{P} - negligible subset \mathcal{N} of Ω . For almost all $\omega \in \Omega \setminus \mathcal{N}$, we have

$$\limsup_{i \rightarrow \infty} \left(\arg \min_{\theta \in \Theta} f_i(\omega, \theta) \right) \subset \arg \min_{\theta \in \Theta} f_0(\omega, \theta)$$

where the lim sup is in the weak topology. For all $\omega \in \Omega \setminus \mathcal{N}$ if there is a weakly compact set $K \subset \mathcal{X}$ such that $\arg \min_{\theta} f_i(\omega, \theta) \subset K$ and $\arg \min_{\theta} f_i \neq \emptyset$ for all i , then

$$\lim_{i \rightarrow \infty} (\inf f_i(\omega, \theta)) = \inf f_0(\omega, \theta).$$

This can be interapted as consistency and risk optimality in infinite dimensional parameter spaces. And We establish the existence of minimum defined on Hilbert spaces.

If the set of $\arg \min f$ is sigleton, we can obtain a convergence in distribution of the sequences $\arg \min f_i$ and $\inf f_i$

$$\arg \min_{\theta \in \Theta} f_i(\omega, \theta) \rightsquigarrow \arg \min_{\theta \in \Theta} f_0(\omega, \theta)$$

$$\lim_{i \rightarrow \infty} (\inf f_i(\omega, \theta)) \rightsquigarrow \inf f_0(\omega, \theta).$$