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A Glivenko-Cantelli Bootstrap Theorem for the Foster-Greer-Thorbecke Poverty Index

Un Teorema Glivenko-Cantelli Bootstrap para la Medida de Pobreza de Foster-Greer-Thorbecke

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ABSTRACT. We assume the Foster-Greer-Thorbecke (FGT) poverty index as an empirical process indexed by a particular Glivenko-Cantelli class or collection of functions and define this poverty index as a functional empirical process of the bootstrap type, to show that the outer almost sure convergence of the FGT empirical process is a necessary and sufficient condition for the outer almost sure convergence of the FGT bootstrap empirical process; that is: both processes are asymptotically equivalent respect to this type of convergence.

 $Key\ words\ and\ phrases.$ Foster-Greer-Thorbecke poverty index, convergence of empirical processes, Glivenko-Cantelli classes, bootstrap empirical processes.

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Resumen. Asumimos el indicador de pobreza de Foster-Greer-Thorbecke (FGT) como un proceso empírico indexado por una particular clase o colección de funciones Glivenko-Cantelli y definimos este indicador de pobreza como un proceso empírico funcional del tipo bootstrap, para probar que la convergencia casi segura exterior del proceso empírico FGT es una condición necesaria y suficiente para la convergencia casi segura exterior del proceso empírico bootstrap FGT; esto es: ambos procesos son asintóticamente equivalentes respecto de este tipo de convergencia.

Palabras y frases clave. Indicador de pobreza de Foster-Greer-Thorbecke, convergencia de procesos empíricos, clases Glivenko-Cantelli, procesos empíricos bootstrap.

1. Introduction

The problem of estimating one-dimensional poverty measures is theoretically addressed in this paper, developing a characterization in law of large numbers, in the framework of bootstrap empirical processes. To achieve this goal, first we introduce some basics elements: let N be a statistical universe of individuals (let us say households), such that for each one of them it is possible to determine its level of income following e.g. [12, 17], for a random sample of n individuals withdrawn from this population, a measure or classic index of poverty is a function $\mathcal{P}: \mathbb{R}^{n+1}_+ \to [0,1]$, where the value of $\mathcal{P}(y,z)$ indicates the degree or level of poverty associated with the vector of incomes $y = (y_1, y_2, \dots, y_n) \in \mathbb{R}^n_+$ and the fixed poverty line $z \in \mathbb{R}_+$, such that any j-th individual of the random sample is considered poor if $y_i < z$.

This type of measures is commonly denominated one-dimensional poverty indices because in their construction only one economic dimension is considered. With the research published by Sen in 1976 about the first axioms or properties of the axiomatic method of poverty (see [18]), the idea of studying this problem as a phenomenon that depends only on the income acquires greater mathematical rigor within the economic theory, and various measures of poverty begin to be proposed, all of which are supported in the Sen's axiomatic definition. In this approach, one of the most important measures is the Foster-Greer-Thorbecke (FGT) poverty index (1984, [7]):

$$FGT(y, z, \alpha) = \frac{1}{n} \sum_{j=1}^{q} \left(\frac{z - y_j}{z}\right)^{\alpha}, \tag{1}$$

that emphasizes the degree of aversion to poverty by including the parameter $\alpha \geq 0$, where according with [12], [17] and [22] among others, the sum in (1) is only over q: the number of poor individuals for the random sample¹.

On the other hand, the bootstrap technique was introduced by Efron in 1979 and 1982 [5, 6], as a method to estimate the sample distribution of a statistics. In general, let Y_1, Y_2, \ldots, Y_n be a finite collection of i.i.d. random variables with law of probability \mathbb{P} , if $\theta := \theta(\mathbb{P})$ is a parameter of interest, $\theta_n := \theta_n(Y_1, Y_2, \ldots, Y_n; \mathbb{P})$ an estimator of θ , and $\hat{Y}_1, \hat{Y}_2, \ldots, \hat{Y}_n$ an i.i.d. random sample with replacement of the empirical probability measure \mathbb{P}_n . Then, the "bootstrap principle" consists in estimating the unknown distribution of θ_n through $\hat{\theta}_n := \theta_n(\hat{Y}_1, \hat{Y}_2, \ldots, \hat{Y}_n; \mathbb{P}_n)$.

In 2009, Lo and Seck found that the FGT poverty index defined in (1) understood as an empirical process satisfies a very particular law of large numbers (see [14]). Now, we found that it is possible to establish an important

¹For example, if $\alpha = 0$ the index is interpreted as the *incidence of poverty*, while for $\alpha = 1$ and $\alpha = 2$ is interpreted as the *intensity or severity of poverty* and the *depth or inequality among the poor*, respectively. For a detailed discussion about the axiomatic method and all the one-dimensional poverty indices proposed in the literature, see *e.g.* [22].

convergence relationship between the FGT empirical process of Lo and Seck and another one of the bootstrap type defined below. The statements of our main result presented in the paper are inspired (among others) in the theorems 3.3 of [21] and 10.15 of [13], that succinctly tell the reader: if one wants to obtain a uniform bootstrap approximation one should check if a certain class is Glivenko-Cantelli. Indeed, the theoretical proposal presented here is a particular contribution over the literature: it formally states that under certain conditions, the FGT empirical process considered as an estimator of the average poverty level (statistics) converges almost surely to the real and unknown average poverty level (parameter) reflected in the mean function of the corresponding process, if and only if the FGT bootstrap empirical process considered as a bootstrap estimator of the average poverty level (bootstrap statistics) converges almost surely to the correspondent estimator, for a random sample of incomes statistically large and representative of a statistical universe of households.

The article is structured in four Sections, including this introduction. In Section 2, we present the problem statement. Consequently, Section 3 presents the main result, and finally, Section 4 contains all the tools required for its development.

2. The problem

Consider the product probability space $(\Omega, \Sigma, P) := (\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}})$. In this framework $\mathcal{X}^{\mathbb{N}}$ is the sample space of all infinite-numerable sequences of incomes, such that for any infinite-numerable sequence $\omega := (y_1, y_2, \ldots) \in \mathcal{X}^{\mathbb{N}}$, we can define a function or coordinate projection $Y : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}$ such that $Y(\omega) = y \in \mathcal{X}$, with probability distribution function $\mathbb{F}(z) = \mathbb{P}(Y \leq z)$ for $z \in \mathbb{R}_+$ fixed. Moreover, according with [3] and [9] among others, we can define a finite collection of functions Y_1, Y_2, \ldots, Y_n i.i.d. $\sim \mathbb{P}$, so that for each $j \in \mathbb{N}$, $Y_j : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}$ is the j-th coordinate projection on $(\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}})$, such that for all $\omega := (y_1, y_2, \ldots) \in \mathcal{X}^{\mathbb{N}}$, $Y_j(\omega) = y_j \in \mathcal{X}$, with empirical distribution function:

$$\mathbb{F}_n(z) = \frac{1}{n} \sum_{j=1}^n \mathbb{I}\{Y_j < z\} = \frac{\sharp\{Y_j < z : 1 \le j \le n\}}{n},\tag{2}$$

for $z \in \mathbb{R}_+$ fixed, where $q = n\mathbb{F}_n(z)$. Particularly, the i.i.d. collection $\{Y_j\}_{j=1}^n$ is an *empirical process of n random variables*, where each projection Y_j on the product probability space $(\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}})$ represents the *observed level of income for the j-th statistical individual of the random sample of size n* in the probability space $(\mathcal{X}, \mathcal{A}, \mathbb{P})$.

Let $\mathbb{P}_n: \mathcal{X}^{\mathbb{N}} \times \mathcal{A} \to [0,1]$ be the *empirical measure* associated with this sequence of random variables, where:

$$\mathbb{P}_n := \frac{1}{n} \sum_{j=1}^n \delta_{Y_j}. \tag{3}$$

Lo and Seck (2009, [14]) define the class or collection of functions $\mathcal{F}_{\Gamma} := \{f_{\alpha}, \alpha \geq 0\}$:

$$f_{\alpha}(x) = \left| \frac{z - x}{z} \right|^{\alpha} \cdot \mathbb{I}\{x < z\},\tag{4}$$

where the second term of (4) is an indicator function. In this setting, following the criterion frequently used in *probability theory*, if we "omit" the dependence on ω in the next notation, then (with some abuse of notation) for each $\omega := (y_1, y_2, \ldots) \in \mathcal{X}^{\mathbb{N}}$ fixed as infinite-numerable sequence of sample points, are obtained the *trajectories or realizations*:

$$f_{\alpha} \mapsto \mathbb{P}_{n}(f_{\alpha}) = \frac{1}{n} \sum_{j=1}^{n} f_{\alpha}(Y_{j}) = \frac{1}{n} \sum_{j=1}^{n} \left| \frac{z - Y_{j}}{z} \right|^{\alpha} \cdot \mathbb{I}\{Y_{j} < z\}$$

$$= \int_{\mathcal{X}} f_{\alpha}(y_{j}) d\mathbb{P}_{n}(y_{j}) = \int_{0}^{z} \left| \frac{z - y_{j}}{z} \right|^{\alpha} d\mathbb{F}_{n}(y_{j}) = \mathbb{E}_{\mathbb{F}_{n}}(f_{\alpha}(Y_{j})), \quad (5)$$

which allows to describe the FGT poverty index defined in (1) like the functional or \mathcal{F}_{Γ} -indexed empirical process $\{\mathbb{P}_n(f_{\alpha}): f_{\alpha} \in \mathcal{F}_{\Gamma}, \alpha \geq 0\}$, with $\mathcal{F}_{\Gamma} \subset L_1(\mathcal{X}, \mathcal{A}, \mathbb{P})$ and the compositions $f_{\alpha}(Y_j) \equiv f_{\alpha} \circ Y_j: \mathcal{X}^{\mathbb{N}} \to \mathcal{X} \to \mathbb{R}$, for all $\alpha \geq 0, j = 1, 2, \ldots, n$.

Let

$$\hat{\mathbb{P}}_n := \frac{1}{n} \sum_{j=1}^n \delta_{\hat{Y}_j} \tag{6}$$

be the classical Efron nonparametric bootstrap empirical measure, where it is possible to consider n bootstrap samples $\hat{Y}_1, \hat{Y}_2, \dots, \hat{Y}_n$ of a determined collection of i.i.d.~ \mathbb{P} functions or coordinate projections Y_1, Y_2, \dots, Y_n . Following e.g. [2, 15, 21], we can consider a triangular array of exchangeable random variables $\mathbf{W} := \{W_{nj} : j = 1, 2, \dots, n, n = 1, 2, \dots\}$ defined on $(\mathcal{W}, \mathcal{D}, \mathbb{P}_W)$, such that these random variables can be interpreted as random weights, in the sense that each component W_{nj} reflects the number of times that Y_j is selected for the n trials of the bootstrap sample with replacement, where:

$$\hat{\mathbb{P}}_n^W := \frac{1}{n} \sum_{j=1}^n W_{nj} \delta_{Y_j},\tag{7}$$

is just the exchangeably weighted bootstrap empirical measure, such that the classical measure $\hat{\mathbb{P}}_n$ defined above is a special case of $\hat{\mathbb{P}}_n^W$ obtained by taking $(W_{n1}, W_{n2}, \ldots, W_{nn})' = \underline{W}_n = \underline{M}_n$, with $\underline{M}_n = (M_{n1}, M_{n2}, \ldots, M_{nn})' \sim Mult_n(n, (1/n, 1/n, \ldots, 1/n))$. Consequently, we can define the functional or \mathcal{F}_{Γ} -indexed FGT bootstrap empirical process $\{\hat{\mathbb{P}}_n^W(f_{\alpha}) : f_{\alpha} \in \mathcal{F}_{\Gamma}, \alpha \geq 0\}^2$.

In [14], Lo and Seck show that the class \mathcal{F}_{Γ} is strong \mathbb{P} -Glivenko-Cantelli; that is:

$$\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} = \sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} |\mathbb{P}_n(f_{\alpha}) - \mathbb{P}(f_{\alpha})| \xrightarrow{a.s.^*} 0, \tag{8}$$

as $n \to \infty$, where

$$\mathbb{P}(f_{\alpha}) = \int_{\mathcal{X}} f_{\alpha}(y) d\mathbb{P}_{Y}(y) = \int_{0}^{z} \left| \frac{z - y}{z} \right|^{\alpha} d\mathbb{F}(y) = \mathbb{E}_{\mathbb{F}}(f_{\alpha}(Y)), \tag{9}$$

is the correspondent mean function of the FGT empirical process $\mathbb{P}_n(f_\alpha)$ defined above. Now, we find that under the G.-C. hypothesis, the trajectories or realizations of \mathbb{P}_n^W get uniformly closer to \mathbb{P}_n as $n \to \infty$, and that the reciprocal is also true.

3. The Main Result

Specifically, we suppose in this paper that $\underline{W}_n = (W_{n1}, W_{n2}, \dots, W_{nn})'$ satisfies the following conditions:

A1. $\underline{W}_n = (W_{n1}, W_{n2}, \dots, W_{nn})'$ is exchangeable for all $n = 1, 2, \dots$, that is, for any permutation $\pi = (\pi(1), \pi(2), \dots, \pi(n))$ of $(1, 2, \dots, n)$, the joint distribution of $\pi(\underline{W}_n) = (W_{n\pi(1)}, W_{n\pi(2)}, \dots, W_{n\pi(n)})'$ is the same as that of \underline{W}_n .

A2.
$$W_{nj} \ge 0$$
 for all n, j , and $\sum_{j=1}^{n} W_{nj} = n$, for all n .

Using the last condition like in [21], page 598:

$$\hat{\mathbb{P}}_{n}^{W} - \mathbb{P}_{n} = \frac{1}{n} \sum_{j=1}^{n} (W_{nj} - 1)(\delta_{Y_{j}} - \mathbb{P}) =: \frac{1}{n} \sum_{j=1}^{n} \xi_{nj} Z_{j}.$$
 (10)

That is, we have a multiplier process with $\xi_{nj} := W_{nj} - 1$ and $Z_j := \delta_{Y_j} - \mathbb{P}$, respectively, such that for any $f_{\alpha} \in \mathcal{F}_{\Gamma} : f_{\alpha} \mapsto (\hat{\mathbb{P}}_n^W - \mathbb{P}_n)(f_{\alpha}) = \frac{1}{n} \sum_{j=1}^n \xi_{nj} Z_j(f_{\alpha}),$

with $Z_j(f_\alpha) = f_\alpha(Y_j) - \mathbb{P}(f_\alpha)$. Particularly, we redefine **A1** and consider a couple of additional conditions for the weights ξ_{nj} :

B1.
$$\underline{\xi}_n = (\xi_{n1}, \xi_{n2}, \dots, \xi_{nn})'$$
 is exchangeable for all n .

the correspondent bootstrap processes in our main result, respect to the Σ_n -measurability and the backward submartingale property required in the Strobl's result, with Σ_n being the filtration defined in remark 4.7 of this Section. Additionally, following e.g. [21], our main result developed in Section 3 will allow to present in future papers, at least one Glivenko-Cantelli theorem for the classical nonparametric bootstrap empirical measure $\hat{\mathbb{P}}_n := \frac{1}{n} \sum_{j=1}^n M_{nj} \delta_{Y_j}$ as a direct consequence of this main result (see e.g. Theorem 3.3 for the exchangeable bootstrap and Theorem 3.2 for Efron's bootstrap, that follows as a corollary of the first mentioned here in [21]).

B2. The norm $L_{2,1}$ of ξ_{n1} is *finite*; that is,

$$\|\xi_{n1}\|_{2,1} = \int_0^\infty \sqrt{\mathbb{P}_W(|\xi_{n1}| \ge t)} dt \le k < \infty.$$

B3. ξ_{n1} satisfies the weak second-moment condition:

$$\lim_{t \to \infty} \limsup_{n \to \infty} t^2 \mathbb{P}_W(|\xi_{n1}| \ge t) = 0.$$

Theorem 3.1. Let $\{Z_j\}_{j=1}^n$ be an i.i.d. empirical process, where $Z_j := \delta_{Y_j} - \mathbb{P}$, with $Y_j : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}$ i.i.d. $\sim \mathbb{P}$. Let $\{\xi_{nj}\}_{j=1}^n$ be an i.i.d. collection of random weights independent of the collection $\{Z_j\}_{j=1}^n$, with mean $\mathbb{E}(\xi_{nj}) = \mu$ and that satisfies **B1-B3**, such that $\xi_{nj} := W_{nj} - 1$ with $\{W_{nj}\}_{j=1}^n$ that satisfies **A1-A2**. Let the class or collection $\mathcal{F}_{\Gamma} := \{f_{\alpha}, \alpha \geq 0\}$ with measurable cover function $F^* : \mathcal{X} \to \mathbb{R}$ defined by $F^* := (\|f_{\alpha}\|_{\mathcal{F}_{\Gamma}})^* \in L_1(\mathcal{X}, \mathcal{A}, \mathbb{P})$ and

$$f_{\alpha}(y_j) = \left| \frac{z - y_j}{z} \right|^{\alpha} \cdot \mathbb{I}\{y_j < z\},$$

for all $y_j \in \mathcal{X}$ and the fixed poverty line $z \in \mathbb{R}_+$. Then, the following are equivalent:

- (i) \mathcal{F}_{Γ} is strong \mathbb{P} -Glivenko-Cantelli.
- (ii) $\|\hat{\mathbb{P}}_n^W \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}} \xrightarrow{a.s.^*} 0$, as $n \to \infty$.

Proof. (i) \Rightarrow (ii): We know that $\hat{\mathbb{P}}_n^W - \mathbb{P}_n = \frac{1}{n} \sum_{i=1}^n \xi_{nj} Z_j$, by (10). It follows:

$$\mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \xi_{nj} Z_j \right\|_{\mathcal{F}_{\Gamma}} \le 2n_0 \mathbb{E}^* \| Z_1 \|_{\mathcal{F}_{\Gamma}} \cdot \mathbb{E} \left(\max_{1 \le j \le n} \frac{|\xi_{nj}|}{n} \right) + 4 \frac{\|\xi_{n1}\|_{2,1}}{\sqrt{n}} \cdot \max_{n_0 < k \le n} \left\{ \mathbb{E}^* \left\| \frac{1}{\sqrt{k}} \sum_{j=n_0+1}^k \epsilon_j Z_j \right\|_{\mathcal{F}_{\Gamma}} \right\},$$

for any $1 \leq n_0 < n$. This is the right side in the lemma of inequalities for the bootstrap process (Lemma 4.12). We point out a few properties related to this upper bound: (a) Since \mathcal{F}_{Γ} is strong \mathbb{P} -Glivenko-Cantelli by hypothesis; i.e., $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} \stackrel{a.s.^*}{\longrightarrow} 0$, as $n \to \infty$, then this implies $\mathbb{E}^* \|Z_1\|_{\mathcal{F}_{\Gamma}} = \mathbb{P}^* \|f_{\alpha}(Y_1) - \mathbb{P}(f_{\alpha})\|_{\mathcal{F}_{\Gamma}} < \infty$, by the lemma of the necessity of integrability of the envelope function (Lemma 4.6). (b) $\mathbb{E}\left(\max_{1\leq j\leq n} \frac{|\xi_{nj}|}{n}\right) \to 0$, follows from the lemma of convergence in mean for the maximum of the exchangeable weights (lemma 4.13), under the conditions **B2** and **B3**. (c) $\|\xi_{n1}\|_{2,1} < \infty$, by **B2**. (d) Let F^*

the measurable cover function of \mathcal{F}_{Γ} defined above, by the inequality (19) of the Strobl's theorem for backward submartingales (Lemma 4.9), it follows that

$$\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* = \left(\sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} \left| \frac{1}{n} \sum_{j=1}^n f_{\alpha}(Y_j) \right| \right)^*$$

$$\leq \frac{1}{n} \sum_{j=1}^n F^*(Y_j)$$

$$= \mathbb{P}_n(F^*) < \infty.$$

Emulating the Strobl's theorem for $\{\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^*\}_{n\in\mathbb{N}}$, we can conclude that this process is a backward submartingale respect to $\{\Sigma_n\}_{n\in\mathbb{N}}$, the σ -algebra defined in Remark 4.7. Since $\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* \cdot \mathbb{I}\{\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* > M\} \leq \mathbb{P}_n(F^*) \cdot \mathbb{I}\{\mathbb{P}_n(F^*) > M\}$ for M > 0, then $\lim_{M \to \infty} \mathbb{E}(\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* \cdot \mathbb{I}\{\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* > M\}) = 0$; that is, $\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^*$ is uniformly integrable, and consequently L_1 -bounded. By the Doob's theorem of convergence (Lemma 4.2), it follows that $\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^*$ converges almost surely to a finite limit, and by the G.-C. hypothesis and the continuity of the uniform norm, $\|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* \xrightarrow{a.s.^*} \|\mathbb{P}\|_{\mathcal{F}_{\Gamma}}$. Then, applying the theorem of convergence in mean for backward submartingales (Lemma 4.3), $\mathbb{E}^*\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} \to 0$, as $n \to \infty$, and by the lower bound of (24) in the lemma of Rademacher symmetrization (Lemma 4.10), we have:

$$\lim_{n \to \infty} \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \epsilon_j Z_j \right\|_{\mathcal{F}_{\Gamma}} = 0,$$

and this concludes (d). Therefore, by (a)-(b)-(c)-(d):

$$\lim_{n \to \infty} \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \xi_{nj} Z_j \right\|_{\mathcal{F}_{\Gamma}} = 0, \tag{11}$$

that is, $\mathbb{E}^* \|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}} \to 0$, as $n \to \infty$. This implies that $\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* \xrightarrow{\mathbb{P}} 0$, because the *convergence in outer mean* implies *convergence in outer probability*.

Let $\dot{F}^* := (\|f_{\alpha} - \mathbb{P}(f_{\alpha})\|_{\mathcal{F}_{\Gamma}})^* \in L_1(\mathcal{X}, \mathcal{A}, \mathbb{P})$ the measurable cover function of the class $\dot{\mathcal{F}}_{\Gamma} := \{f_{\alpha} - \mathbb{P}(f_{\alpha}) : f_{\alpha} \in \mathcal{F}_{\Gamma}\}$, then for $f_{\alpha} \in \mathcal{F}_{\Gamma}$ fixed:

$$\left| \frac{1}{n} \sum_{j=1}^{n} \xi_{nj} [f_{\alpha} - \mathbb{P}(f_{\alpha})] \right| \leq \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| f_{\alpha} - \mathbb{P}(f_{\alpha})|$$

$$= \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| \cdot |f_{\alpha} - \mathbb{P}(f_{\alpha})|$$

$$\leq \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| \cdot ||g_{\alpha} - \mathbb{P}(g_{\alpha})||_{\mathcal{F}_{\Gamma}}^{*},$$

$$\Rightarrow \left| \frac{1}{n} \sum_{j=1}^{n} \xi_{nj} [f_{\alpha} - \mathbb{P}(f_{\alpha})] \right| \leq \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| \cdot ||g_{\alpha} - \mathbb{P}(g_{\alpha})||_{\mathcal{F}_{\Gamma}}^{*}, \quad \text{for all } f_{\alpha} \in \mathcal{F}_{\Gamma}$$

$$= \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| \cdot ||f_{\alpha} - \mathbb{P}(f_{\alpha})||_{\mathcal{F}_{\Gamma}}^{*}, \quad \text{for all } f_{\alpha} \in \mathcal{F}_{\Gamma}$$

$$\Rightarrow \left| \left| \frac{1}{n} \sum_{j=1}^{n} \xi_{nj} [f_{\alpha} - \mathbb{P}(f_{\alpha})] \right| \right|_{\mathcal{F}_{\Gamma}}^{*} \leq \frac{1}{n} \sum_{j=1}^{n} |\xi_{nj}| \cdot \dot{F}^{*}, \quad (12)$$

where the right side of (12) is integrable by remark 4.5. Using this fact and the condition of exchangeability **B1** for the random weights ξ_{nj} , we can emulate again the Strobl's theorem to see that $\{\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_\Gamma}^*, \Sigma_n\}_{n\in\mathbb{N}}$ is a backward submartingale. By inequality (12), it is clear that $\sup_{n\in\mathbb{N}} \mathbb{E}(\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_\Gamma}^*) < \infty$,

and from Doob's theorem applied above, it follows that $\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^*$ converges almost surely to a finite limit, but $\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}^* \xrightarrow{\mathbb{P}} 0$, thus this limit must be equal to zero and then $\|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}} \xrightarrow{a.s.^*} 0$.

(ii)
$$\Rightarrow$$
(i): Let $F^* := (\|f_{\alpha}\|_{\mathcal{F}_{\Gamma}})^* \in L_1(\mathcal{X}, \mathcal{A}, \mathbb{P}),$

$$\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^* = (\sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} |\hat{\mathbb{P}}_n^W(f_{\alpha})|)^* \le \hat{\mathbb{P}}_n^W(F^*) < \infty.$$

Under the condition of exchangeability **A1**, it is clear that $\{\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^*, \Sigma_n\}_{n \in \mathbb{N}}$ is a *backward submartingale* too by Strobl's result. Since

$$\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_\Gamma}^* \cdot \mathbb{I}\{\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_\Gamma}^* > M\} \leq \hat{\mathbb{P}}_n^W(F^*) \cdot \mathbb{I}\{\hat{\mathbb{P}}_n^W(F^*) > M\}$$

for M > 0, then $\lim_{M \to \infty} \mathbb{E}(\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^* \cdot \mathbb{I}\{\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^* > M\}) = 0$; that is, $\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^*$ is uniformly integrable, and in fact L_1 -bounded. Applying again the Doob's theorem 4.2, $\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}}^*$ converges almost surely to a finite limit, and by the hypothesis (ii) and the continuity of $\|\cdot\|_{\mathcal{F}_{\Gamma}}$, we have $\|\hat{\mathbb{P}}_n^W\|_{\mathcal{F}_{\Gamma}} \stackrel{a.s.^*}{\longrightarrow} \|\mathbb{P}_n\|_{\mathcal{F}_{\Gamma}}$, and therefore, by the Theorem 4.3 (theorem of convergence in mean 4.3), it

follows $\mathbb{E}^* \|\hat{\mathbb{P}}_n^W - \mathbb{P}_n\|_{\mathcal{F}_{\Gamma}} \to 0$, when $n \to \infty$. Since

$$\frac{1}{2} \|\xi_{n1} - \mu\|_1 \cdot \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \epsilon_j Z_j \right\|_{\mathcal{F}_{\Gamma}} \le \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \xi_{nj} Z_j \right\|_{\mathcal{F}_{\Gamma}},$$

is the left side in (26) of the Lemma 4.12 of inequalities for the bootstrap process, where $\|\xi_{n1} - \mu\|_1 < \infty$ under **B2** by Remark 4.11, it is clear that

$$\lim_{n \to \infty} \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \epsilon_j Z_j \right\|_{\mathcal{F}_{\Gamma}} = 0.$$

From the upper bound of (24) in the Lemma 4.10 of Rademacher symmetrization, it follows:

$$\lim_{n \to \infty} \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n Z_j \right\|_{\mathcal{F}_{\Gamma}} = 0, \tag{13}$$

that is, $\mathbb{E}^* \| \mathbb{P}_n - \mathbb{P} \|_{\mathcal{F}_{\Gamma}} \to 0$, when $n \to \infty$. This implies $\| \mathbb{P}_n - \mathbb{P} \|_{\mathcal{F}_{\Gamma}}^* \xrightarrow{\mathbb{P}} 0$ using the convergence argument discussed above, we can conclude that \mathcal{F}_{Γ} is weak \mathbb{P} -Glivenko-Cantelli.

To finish the proof, the process $\{\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*, \Sigma_n\}_{n \in \mathbb{N}}$ is a backward submartingale by Theorem 4.9 (Strobl's theorem) and $\mathbb{E}(\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*) \leq 2\mathbb{P}(F^*) < \infty$, by inequality (20), for each $n \in \mathbb{N}$. Consequently, $\sup_{n \in \mathbb{N}} \mathbb{E}(\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*) < \infty$, and from Doob's theorem, it follows that $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*$ converges almost surely to a finite limit, but $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* \xrightarrow{\mathbb{P}} 0$, thus this limit must be equal to zero, and then $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}$ a.s.* 0.

Note 1. Following page 15 in [8], if we assume measurability for the collection $\{Y_j\}_{j=1}^n$, then the index "*" can be removed in the notation, and all the results presented here also hold (see e.g. the definition 4.4 in the next Section). Lemma 4.6 is similar to Lemma 8.13, page 141 in [13], applied now to the classes of functions \mathcal{F}_{Γ} and $\dot{\mathcal{F}}_{\Gamma}$, respectively. Lemma 4.9 is basically the original Theorem 1.1, pages 826–829 in [19], for the class \mathcal{F}_{Γ} . Lemma 4.10 corresponds to Lemma 11.2.12, pages 343–344 in [4], applied to the class \mathcal{F}_{Γ} . The Lemma 4.12 developed for this collection, is similar to Lemma 2.9.1, pages 177–179 in [20]; or Lemma 2.2, pages 595–596 in [21]. Lemma 4.13 is similar to Lemma 4.7, page 2071 in [15], considering the random weights ξ_{nj} . For details about all the proofs see [11], and for a detailed discussion about the bootstrap see e.g. [10].

4. Tools Required for the Main Result

Definition 4.1. Let $Y_n : \Omega \to \mathbb{R}$ be a sequence of random variables and $\{\Sigma_n\}_{n\in\mathbb{N}}$ be a *filtration* on (Ω, Σ, P) ; that is, a decrescent sequence of sub- σ -algebras of Σ ,

$$\Sigma \supset \Sigma_n \supset \Sigma_{n+1}$$

for each $n \in \mathbb{N}$ such that:

- (1) $\{Y_n\}_{n\in\mathbb{N}}$ is adapted to the filtration $\{\Sigma_n\}_{n\in\mathbb{N}}$ (or more generally, adapted to the P-completion of $\{\Sigma_n\}_{n\in\mathbb{N}}$). That is, Y_n is Σ_n -measurable for each $n\in\mathbb{N}$ (or more generally, measurable for the P-completion of $\{\Sigma_n\}_{n\in\mathbb{N}}$);
- (2) $\mathbb{E}(|Y_n|) < \infty$, for each $n \in \mathbb{N}$;
- (3) $\mathbb{E}(Y_{n+1}|\Sigma_n) \geq Y_{n+1}$ a.s., for each $n \in \mathbb{N}$.

Then $\{Y_n, \Sigma_n\}_{n\in\mathbb{N}}$ is said to be a reversed or backward submartingale.

Lemma 4.2 (Doob's theorem of convergence). Let $\{Y_n\}_{n\in\mathbb{N}}$ be a backward submartingale on (Ω, Σ, P) such that $\sup_{n\in\mathbb{N}} \mathbb{E}(|Y_n|) < \infty$. Then there is an integrable random variable Y such that:

$$\lim_{n \to \infty} Y_n = Y \ a.s. \tag{14}$$

Lemma 4.3 (Theorem of convergence in mean). Let $\{Y_n\}_{n\in\mathbb{N}}$ be a backward submartingale on (Ω, Σ, P) uniformly integrable; that is,

$$\lim_{M \to \infty} \mathbb{E}(|Y_n| \cdot \mathbb{I}\{|Y_n| > M\}) = \lim_{M \to \infty} \int_{(|Y_n| > M)} |Y_n| dP = 0.$$

Then this process is convergent in mean; that is:

$$\lim_{n \to \infty} \mathbb{E}(|Y_n - Y|) = 0. \tag{15}$$

Definition 4.4. Let $f: \Omega \to \overline{\mathbb{R}}$ be an arbitrary function not necessarily measurable on a probability space (Ω, Σ, P) , where $\overline{\mathbb{R}} \equiv [-\infty, \infty]$ is the set of extended real numbers. The *outer expectation or outer integral of* f *with respect to* P, is defined as:

$$\mathbb{E}_{P}^{*}(f) := \int_{-\infty}^{\infty} f dP = \inf\{\mathbb{E}_{P}(g) : g \geq f, g : \Omega \to \overline{\mathbb{R}} \text{ measurable and } \mathbb{E}_{P}(g) \text{ exists}\},$$
(16)

where $\mathbb{E}_P(g) = \int g dP = \int g^+ dP + \int g^- dP = \mathbb{E}_P(g^+) + \mathbb{E}_P(g^-)$, such that $\mathbb{E}_P(g)$ exist, if at least $\mathbb{E}_P(g^+)$ or $\mathbb{E}_P(g^-)$ is finite.

If f is a measurable function quasi-integrable, then $\mathbb{E}_P^*(f) = \mathbb{E}_P(f)$. Moreover, if $\mathbb{E}_P^*(f) < \infty$, then $\mathbb{E}_P^*(f) = \mathbb{E}_P(f^*)$, where $f^* : \Omega \to \overline{\mathbb{R}}$ is the minimal measurable majorant or smallest measurable function above f, that satisfies:

- (1) $f^*(\omega) \ge f(\omega)$ for each $\omega \in \Omega$;
- (2) For any measurable function $g: \Omega \to \overline{\mathbb{R}}$ with $g \geq f$ a.s., $f^* \leq g$ a.s.

Remark 4.5. Consider the class $\mathcal{F}_{\Gamma} := \{f_{\alpha}, \alpha \geq 0\}$, and let $F : \mathcal{X} \to \mathbb{R}$ be the *envelope function* of this collection, with $F := \|f_{\alpha}\|_{\mathcal{F}_{\Gamma}}$ such that $|f_{\alpha}(y)| \leq F(y) = \sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} |f_{\alpha}(y)|$ for each $y \in \mathcal{X}$ and $f_{\alpha} \in \mathcal{F}_{\Gamma}$. Now we can define the class

 $\dot{\mathcal{F}}_{\Gamma} := \{f_{\alpha} - \mathbb{P}(f_{\alpha}) : f_{\alpha} \in \mathcal{F}_{\Gamma}\}$, with $\dot{F} : \mathcal{X} \to \mathbb{R}$ defined by $\dot{F} := \|f_{\alpha} - \mathbb{P}(f_{\alpha})\|_{\mathcal{F}_{\Gamma}}$, such that $|f_{\alpha} - \mathbb{P}(f_{\alpha})| \leq \dot{F}$. According to the above definition, under integrability, the outer expectation of an envelope function for a determined class or collection is equal to the expected value of the measurable cover function respect to this envelope. In other terms, if $\mathbb{P}^*(F) < \infty$ and $\mathbb{P}^*(\dot{F}) < \infty$, then is true that $\mathbb{P}^*(F) = \mathbb{P}(F^*)$ and $\mathbb{P}^*(\dot{F}) = \mathbb{P}(\dot{F}^*)$, where $F^* := (\|f_{\alpha}\|_{\mathcal{F}_{\Gamma}})^*$ and $\dot{F}^* := (\|f_{\alpha} - \mathbb{P}(f_{\alpha})\|_{\mathcal{F}_{\Gamma}})^*$ are the measurable cover functions for F and F of the classes \mathcal{F}_{Γ} and $\dot{\mathcal{F}}_{\Gamma}$, respectively. In this setting, $F \leq F^*$, F^* is measurable, and $F^* \leq h$ P-a.s. for all measurable function $h \geq F$. The same analysis follows for \dot{F} and \dot{F}^* .

Lemma 4.6 (Necessity of integrability of the envelope function). If the class of functions \mathcal{F}_{Γ} is strong \mathbb{P} -Glivenko-Cantelli; i.e., $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} \xrightarrow{a.s.^*} 0$, as $n \to \infty$. Then this implies that $\mathbb{P}^*\|f_{\alpha} - \mathbb{P}(f_{\alpha})\|_{\mathcal{F}_{\Gamma}} < \infty$. Consequently, if \mathcal{F}_{Γ} is $L_1(\mathbb{P})$ -bounded; i.e., $\|\mathbb{P}\|_{\mathcal{F}_{\Gamma}} = \sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} |\mathbb{P}(f_{\alpha})| < \infty$, then $\mathbb{P}^*(F) < \infty$ for an envelope function F.

Proof. See [11], Lemma 3.1, pages 83–84.

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Remark 4.7. Let \mathbb{P}_n be the *n*-th empirical measure on $(\Omega, \Sigma, P) := (\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}})$, it follows that the class \mathcal{C} of all sets invariant under permutations of the *n* first coordinates $Y_j : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}$ is a σ -algebra³. Specifically, we can define Σ_n as the smallest σ -algebra that contains all the sets

$${A \in \Sigma : \mathbb{I}_A(y) = \mathbb{I}_A(\pi y)};$$

that is, invariants under any permutation $\pi \in S(n)$ of the first n coordinates $Y_j : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}$, such that $\Sigma_n \supset \Sigma_{n+1}$, for each $n \in \mathbb{N}$.

$$S(n) := \{\pi : \pi(j) = j' \text{ if and only if } \pi(j') = j, \text{ for all } j, j' \leq n \in \mathbb{N}\}.$$

³See e.g. [3], Lemma A.2.8, pages 127–128. In general, for a non empty set X, a permutation of X is a bijection $f_{\pi}: X \to X$, such that if $(x_1, x_2, \ldots, x_n, x_{n+1}, x_{n+2}, \ldots) \in X$, then any permutation π of the first n terms of this sequence; that is, $(x_{\pi(1)}, x_{\pi(2)}, \ldots, x_{\pi(n)}, x_{n+1}, x_{n+2}, \ldots)$ also belongs to X, and it is part of the symmetric group of n denoted by S(n), where:

Remark 4.8. Let $A \in \Sigma_n$. For $1 \leq j < j' \leq n$,

$$\mathbb{E}[f_{\alpha}(Y_{j})|\Sigma_{n}] = \mathbb{E}[f_{\alpha}(Y_{j}^{'})|\Sigma_{n}]. \tag{17}$$

To see this, consider the two sides of (17) are Σ_n -measurable by properties of the conditional expectation. Now, for any $A \in \Sigma_n$, $\pi \in S(n)$ the symmetric group of order n:

$$(y_1, y_2, \dots, y_n, y_{n+1}, y_{n+2}, \dots) \in A \Leftrightarrow (y_{\pi(1)}, y_{\pi(2)}, \dots, y_{\pi(n)}, y_{n+1}, y_{n+2}, \dots) \in A.$$

Hence, if $y \in Y_j(A)$, then for some $\{z_m\} \in A : z_j = y$, but now we have $(z_{\pi(1)}, z_{\pi(2)}, \dots, z_{\pi(n)}, z_{n+1}, z_{n+2}, \dots) \in A$, where $\pi \in S(n) : \pi(j') = j$, so

$$Y_{j'}((z_{\pi(1)}, z_{\pi(2)}, \dots, z_{\pi(n)}, z_{n+1}, z_{n+2}, \dots)) = z_{\pi(j')} = z_j = y \Rightarrow y \in Y_{j'}(A).$$

Applying the same argument as above, it is clear that if $y \in Y_{j'}(A)$, then $y \in Y_j(A)$. Therefore $Y_j(A) = Y_{j'}(A)$, and it follows that:

$$\begin{split} \mathbb{E}[f_{\alpha}(Y_{j})\cdot\mathbb{I}_{C}] &= \int_{\Omega}f_{\alpha}(Y_{j}(\{y_{m}\}_{m\in\mathbb{N}}))\cdot\mathbb{I}_{A}(\{y_{m}\}_{m\in\mathbb{N}})d\mathbb{P}^{\mathbb{N}}(\{y_{m}\}_{m\in\mathbb{N}})\\ &= \int_{A}f_{\alpha}(Y_{j}(\{y_{m}\}_{m\in\mathbb{N}}))d\mathbb{P}^{\mathbb{N}}(\{y_{m}\}_{m\in\mathbb{N}})\\ &= \int_{Y_{j}(A)}f_{\alpha}(y_{j})d(\mathbb{P}^{\mathbb{N}}\circ Y_{j}^{-1})(y_{j}) = \int_{Y_{j}(A)}f_{\alpha}(y)d\mathbb{P}(y)\\ &= \int_{Y_{j'}(A)}f_{\alpha}(y)d\mathbb{P}(y) = \int_{Y_{j'}(A)}f_{\alpha}(y_{j'})d(\mathbb{P}^{\mathbb{N}}\circ Y_{j'}^{-1})(y_{j'})\\ &= \int_{A}f_{\alpha}(Y_{j'}(\{y_{m}\}_{m\in\mathbb{N}}))d\mathbb{P}^{\mathbb{N}}(\{y_{m}\}_{m\in\mathbb{N}})\\ &= \int_{\Omega}f_{\alpha}(Y_{j'}(\{y_{m}\}_{m\in\mathbb{N}}))\cdot\mathbb{I}_{A}(\{y_{m}\}_{m\in\mathbb{N}})d\mathbb{P}^{\mathbb{N}}(\{y_{m}\}_{m\in\mathbb{N}})\\ &= \mathbb{E}[f_{\alpha}(Y_{j'})\cdot\mathbb{I}_{C}]. \end{split}$$

Lemma 4.9 (Strobl's theorem). Let $(\Omega, \Sigma, P) := (\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}})$, and consider the collection of functions $\mathcal{F}_{\Gamma} \subset L_1(\mathcal{X}, \mathcal{A}, \mathbb{P})$. If \mathcal{F}_{Γ} has a measurable cover function $F^* \in L_1(\mathcal{X}, \mathcal{A}, \mathbb{P})$, then $\{\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*, \Sigma_n\}_{n \in \mathbb{N}}$ is a backward submartingale; that is, $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*$ is Σ_n -measurable, P-integrable, and

$$\|\mathbb{P}_{n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* \le \mathbb{E}(\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* | \Sigma_{n+1}) \ P\text{-}a.s.$$
 (18)

for each $n \in \mathbb{N}$.

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Proof. First, we show the integrability of $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*$. Let

$$\|\mathbb{P}_{n} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^{*} = \left(\sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} \left| \frac{1}{n} \sum_{j=1}^{n} f_{\alpha}(Y_{j}) - \int_{\mathcal{X}} f_{\alpha}(y) d\mathbb{P}_{Y}(y) \right| \right)^{*}$$

$$\leq \frac{1}{n} \sum_{j=1}^{n} F^{*}(Y_{j}) + \int_{\mathcal{X}} F^{*}(y) d\mathbb{P}_{Y}(y)$$

$$= \mathbb{P}_{n}(F^{*}) + \mathbb{P}(F^{*}). \tag{19}$$

The right side of (19) is Σ -measurable, P-integrable and real-valued. Hence we have that $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* \in L_1(\Omega, \Sigma, P)$ for each $n \in \mathbb{N}$. In fact, respect to the integrability of the outer empirical discrepancy, by (19),

$$\mathbb{E}(\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_n}^*) \le \mathbb{E}[\mathbb{P}_n(F^*) + \mathbb{P}(F^*)] < 2\mathbb{P}(F^*) < \infty, \tag{20}$$

for each $n \in \mathbb{N}$.

Now, let $g := \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* : \mathcal{X}^{\mathbb{N}} \to \mathbb{R}$. We define $f_{\pi} : \mathcal{X}^{\mathbb{N}} \to \mathcal{X}^{\mathbb{N}}$ for a permutation $\pi \in S(n)$ by

$$f_{\pi}(y_1, y_2, \dots) := (y_{\pi(1)}, y_{\pi(2)}, \dots, y_{\pi(n)}, y_{n+1}, y_{n+2}, \dots).$$

Since $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}$ is invariant under all permutations of the first n coordinates,

$$g \circ f_{\pi} \geq \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} \circ f_{\pi} = \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}},$$

for all $\pi \in S(n)$ and then,

$$\min_{\pi \in S(n)} g \circ f_{\pi} \ge \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}},$$

where the left side of the inequality is a Σ -measurable function. Therefore:

$$\min_{\pi \in S(n)} g \circ f_{\pi} \ge \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* = g \text{ } P\text{-a.s.},$$

by the definition of measurable cover functions. Thus,

$$\min_{\pi \in S(n)} g \circ f_{\pi} = g \ \text{P-a.s.}$$

For each rational $q \geq 0$, let $A_q := A_{q,n}$ the sets where $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_r}^* \geq q$; i.e.

$$A_q := A_{q,n} := \{ \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* \ge q \},$$

we have that

$$\{g \ge q\} = \{\min_{\pi \in S(\pi)} g \circ f_{\pi} \ge q\},$$
 (21)

with $\{\min_{\pi \in S(n)} g \circ f_{\pi} \geq q\} \in \Sigma_n$. Consequently, $\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^*$ is Σ_n -measurable.

Finally, for $j = 1, 2, \ldots, n + 1$, let

$$\mathbb{P}_{n,j} := \frac{1}{n} \sum_{j'=1,j'\neq j}^{n+1} \delta_{Y_{j'}},$$

then $\mathbb{P}_{n,n+1} \equiv \mathbb{P}_n := \frac{1}{n} \sum_{j=1}^n \delta_{Y_j}$, and for $j \neq n+1$, $\mathbb{P}_{n,j}$ has the same properties as \mathbb{P}_n . Therefore,

$$\|\mathbb{P}_{n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}} = \left\| \frac{1}{n+1} \left(\sum_{l=1}^{n+1} \delta_{Y_{l}} \right) - \mathbb{P} \right\|_{\mathcal{F}_{\Gamma}}$$

$$= \frac{1}{n+1} \left\| \frac{1}{n} \left(\sum_{l=1}^{n+1} n \delta_{Y_{l}} \right) - (n+1) \mathbb{P} \right\|_{\mathcal{F}_{\Gamma}}$$

$$= \frac{1}{n+1} \left\| \left(\sum_{j=1}^{n+1} \mathbb{P}_{n,j} \right) - (n+1) \mathbb{P} \right\|_{\mathcal{F}_{\Gamma}}$$

$$= \frac{1}{n+1} \left\| \sum_{j=1}^{n+1} \left(\frac{1}{n} \sum_{j'=1,j'\neq j}^{n+1} \delta_{Y_{j'}} - \mathbb{P} \right) \right\|_{\mathcal{F}_{\Gamma}}$$

$$\leq \frac{1}{n+1} \sum_{j=1}^{n+1} \left\| \frac{1}{n} \sum_{j'=1,j'\neq j}^{n+1} \delta_{Y_{j'}} - \mathbb{P} \right\|_{\mathcal{F}_{\Gamma}}$$

$$= \frac{1}{n+1} \sum_{j=1}^{n+1} \|\mathbb{P}_{n,j} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}$$

$$\leq \frac{1}{n+1} \sum_{j=1}^{n+1} \|\mathbb{P}_{n,j} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}.$$
(22)

The right side of (22) is Σ -measurable, so it is an upper bound of the outer empirical discrepancy $\|\mathbb{P}_{n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* P$ -a.s., too. Therefore,

$$\|\mathbb{P}_{n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* = \mathbb{E}(\|\mathbb{P}_{n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* | \Sigma_{n+1})$$

$$\leq \frac{1}{n+1} \sum_{i=1}^{n+1} \mathbb{E}(\|\mathbb{P}_{n,j} - \mathbb{P}\|^* | \Sigma_{n+1}) \quad P\text{-a.s.}$$

Thus, it is enough to prove that for each $1 \le j \le n+1$,

$$\mathbb{E}(\|\mathbb{P}_{n,j} - \mathbb{P}\|^* | \Sigma_{n+1}) = \mathbb{E}(\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* | \Sigma_{n+1}) \text{ } P\text{-a.s.}$$
(23)

For $m \in \mathbb{N}$ fix but arbitrary, let $h : \mathcal{X} \to \mathbb{R}$ such that

$$h(y_1, y_2, \dots y_m) = \left(\sup_{f_{\alpha} \in \mathcal{F}_{\Gamma}} \left| \frac{1}{m} \sum_{j=1}^{m} f_{\alpha}(y_j) - \mathbb{P}(f_{\alpha}) \right| \right)^*,$$

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if

$$\|\mathbb{P}_{n,n+1} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* = \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* = h \circ (Y_1, Y_2, \dots, Y_{j-1}, Y_j, Y_{j+1}, Y_{j+2}, \dots, Y_n),$$

we have

$$\|\mathbb{P}_{n,j} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* = h \circ (Y_1, Y_2, \dots, Y_{j-1}, Y_{j+1}, Y_{j+2}, \dots, Y_{n+1}).$$

However

$$(Y_1, \dots, Y_{j-1}, Y_j, Y_{j+1}, \dots, Y_n)(A) = (Y_1, \dots, Y_{j-1}, Y_{j+1}, \dots, Y_{n+1})(A),$$

for all $A \in \Sigma_{n+1}$. Thus, we can use the argument developed in Remark 4.8 to verify (17), and obtain:

$$\int_{\Delta} \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* d\mathbb{P}^{\mathbb{N}} = \int_{\Delta} \|\mathbb{P}_{n,j} - \mathbb{P}\|_{\mathcal{F}_{\Gamma}}^* d\mathbb{P}^{\mathbb{N}},$$

using the fact that the sets A in Σ_{n+1} are invariants under all permutations of the first n+1 coordinates. Hence, the equality (23) is true, and consequently the backward submartingale property (18) is satisfied.

Lemma 4.10 (Rademacher symmetrization). Given the basic product probability space $(\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}}) \times (\mathcal{Z}, \mathcal{C}, \mathbb{P}_{\epsilon})$. Let $\{Y_j\}_{j=1}^n$ be an $i.i.d. \sim \mathbb{P}$ empirical process, independent of the i.i.d. Rademacher collection $\{\epsilon_j\}_{j=1}^n$. Then,

$$\frac{1}{2}\mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \epsilon_j (\delta_{Y_j} - \mathbb{P}) \right\|_{\mathcal{F}_{\Gamma}} \le \mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n (\delta_{Y_j} - \mathbb{P}) \right\|_{\mathcal{F}_{\Gamma}} \le 2\mathbb{E}^* \left\| \frac{1}{n} \sum_{j=1}^n \epsilon_j (\delta_{Y_j} - \mathbb{P}) \right\|_{\mathcal{F}_{\Gamma}},$$
(24)

where (24) it is also true if $\mathbb{P}(f_{\alpha})$ is deleted.

Proof. See [11], Lemma 3.3, pages 93–94.

Remark 4.11. If $\|\xi_{n1}\|_{2,1} < \infty$, then this implies $\|\xi_{n1}\|_2 := \mathbb{E}(|\xi_{n1}|^2) < \infty$. Indeed,

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$$\mathbb{E}(|\xi_{n1}|^{2}) = \int_{0}^{\infty} \mathbb{P}_{W}(|\xi_{n1}|^{2} > t)dt$$

$$= \int_{0}^{\infty} \mathbb{P}_{W}(|\xi_{n1}| > \sqrt{t})dt$$

$$= \int_{0}^{\infty} \mathbb{P}_{W}(|\xi_{n1}| > u)2udu \quad (u = \sqrt{t} \Rightarrow u^{2} = t, \text{ y } 2udu = dt)$$

$$= \int_{0}^{\infty} 2u\sqrt{\mathbb{P}_{W}(|\xi_{n1}| > u)}\sqrt{\mathbb{P}_{W}(|\xi_{n1}| > u)}du$$

$$\leq 2\int_{0}^{\infty} u\sqrt{\frac{1}{u^{2}}\mathbb{E}(|\xi_{n1}|^{2})}\sqrt{\mathbb{P}_{W}(|\xi_{n1}| > u)}du \text{ (by Markov's inequality)}$$

$$= 2\sqrt{\mathbb{E}(|\xi_{n1}|^{2})}\int_{0}^{\infty} \sqrt{\mathbb{P}_{W}(|\xi_{n1}| > u)}du$$

$$\Rightarrow \sqrt{\mathbb{E}(|\xi_{n1}|^{2})} \leq 2\int_{0}^{\infty} \sqrt{\mathbb{P}_{W}(|\xi_{n1}| > u)}du$$

$$\Rightarrow \sqrt{\mathbb{E}(|\xi_{n1}|^{2})} \leq 2||\xi_{n1}||_{2.1}. \tag{25}$$

Consequently $\|\xi_{n1}\|_2$ is finite, and therefore $\|\xi_{n1}\|_1 := \mathbb{E}(|\xi_{n1}|) < \infty$. Furthermore, $\mathbb{E}(\xi_{n1})$ is defined and finite if and only if $\mathbb{E}(|\xi_{n1}|) < \infty$. Hence, condition **B2** for $\|\xi_{n1}\|_{2,1}$ ensures that all these moments are finite.

Lemma 4.12 (Inequalities for the bootstrap process with exchangeable multiplier random weights). Given $(\mathcal{X}^{\mathbb{N}}, \mathcal{A}^{\mathbb{N}}, \mathbb{P}^{\mathbb{N}}) \times (\mathcal{W}, \mathcal{D}, \mathbb{P}_W) \times (\mathcal{Z}, \mathcal{C}, \mathbb{P}_{\epsilon})$, the basic product probability space. Let $\{Z_j\}_{j=1}^n$ be an i.i.d. empirical process such that $\mathbb{E}^* \| Z_j \|_{\mathcal{F}_{\Gamma}} < \infty$ for each $j \leq n$, independent of the i.i.d. Rademacher collection $\{\epsilon_j\}_{j=1}^n$. Then, for a collection of i.i.d. exchangeable random weights $\{\xi_{nj}\}_{j=1}^n$ with $\|\xi_{n1}\|_{2,1} < \infty$ and $\mathbb{E}(\xi_{nj}) = \mu$, independent of the collection $\{Z_j\}_{j=1}^n$ and any $1 \leq n_0 < n$,

$$\frac{1}{2} \|\xi_{n1} - \mu\|_{1} \cdot \mathbb{E}^{*} \left\| \frac{1}{n} \sum_{j=1}^{n} \epsilon_{j} Z_{j} \right\|_{\mathcal{F}_{\Gamma}} \leq \mathbb{E}^{*} \left\| \frac{1}{n} \sum_{j=1}^{n} \xi_{nj} Z_{j} \right\|_{\mathcal{F}_{\Gamma}} \\
\leq 2n_{0} \mathbb{E}^{*} \|Z_{1}\|_{\mathcal{F}_{\Gamma}} \cdot \mathbb{E} \left(\max_{1 \leq j \leq n} \frac{|\xi_{nj}|}{n} \right) \\
+ 4 \frac{\|\xi_{n1}\|_{2,1}}{\sqrt{n}} \cdot \max_{n_{0} < k \leq n} \left\{ \mathbb{E}^{*} \left\| \frac{1}{\sqrt{k}} \sum_{j=n_{0}+1}^{k} \epsilon_{j} Z_{j} \right\|_{\mathcal{F}_{\Gamma}} \right\}.$$
(26)

For symmetrically distributed variables ξ_{nj} around μ , the constants 1/2, 2 and 4 can all be replaced by 1, and μ in the left-hand side of (26) is equal to zero.

 \checkmark

Lemma 4.13 (Uniformity square-integrable for the second moment, and convergence in mean for the maximum of the exchangeable multiplier weights). Let $\xi := \{|\xi_{nj}| : j = 1, 2, ..., n, n = 1, 2, ...\}$ be a triangular array of non-negative and exchangeable random variables, defined on the probability space $(\mathcal{W}, \mathcal{D}, \mathbb{P}_W)$. If ξ satisfies conditions **B2** and **B3**, this implies that the sequence $\{|\xi_{n1}|\}_{n\in\mathbb{N}}$ is uniformly square-integrable; that is,

$$\lim_{t \to \infty} \limsup_{n \to \infty} \mathbb{E}(|\xi_{n1}|^2 \cdot \mathbb{I}\{|\xi_{n1}| \ge t\}) = 0.$$
 (27)

Furthermore, **B2** and **B3** also imply that

$$\mathbb{E}\left(\max_{1\leq j\leq n} \frac{|\xi_{nj}|}{n}\right) \to 0. \tag{28}$$

Proof. See [11], Lemma 3.5, pages 102–104.

References

- [1] K. B. Athreya and S. N. Lahiri, Measure theory and probability theory, Springer, New York, 2006.
- [2] G. Cheng and J. Z. Huang, Bootstrap consistency for general semiparametric m-estimation, The Annals of Statistics 38 (2010), no. 5, 2884–2915.
- [3] G. Dierickx, *Limit theorems for general empirical processes*, Master thesis in mathematics, Vrije Universiteit Brussel, may 2012.
- [4] R. M. Dudley, *Uniform central limit theorems*, Cambridge Studies in advanced mathematics, 1999.
- [5] B. Efron, Bootstrap methods: Another look at the jacknife, The Annals of Statistics 7 (1979), 1–26.
- [6] _____, The jacknife, the bootstrap and other resampling plans, SIAM, Philadelphia, 1982.
- [7] J. Foster, J. Greer, and E. Thorbecke, A class of decomposable poverty measures, Econometrica 52 (1984), no. 3, 761–766.
- [8] E. Giné, Empirical processes and applications: an overview, Bernoulli 2 (1996), no. 1, 1–28.
- [9] E. Giné and R. Nickl, Mathematical foundations of infinite-dimensional statistical models, Cambridge University Press, 2015.
- [10] E. Giné and J. Zinn, Bootstrapping general empirical measures, The Annals of Probability 18 (1990), no. 2, 851–869.

- [11] P. Harmath, Análisis y estimación de medidas de pobreza unidimensionales bajo la teoría de procesos empíricos bootstrap, Tesis doctoral en matemáticas, Universidad Centroccidental Lisandro Alvarado, enero 2018.
- [12] P. Harmath, A. Monsalve, and J. Ramoni, *Medición, método axiomático y medidas de pobreza: Una revisión*, Revista de Matemáticas de la Universidad del Atlántico 4 (2017), no. 1, 1–15.
- [13] M. R. Kosorok, Introduction to empirical processes and semipara-metric inference, Springer, New York, 2006.
- [14] G. Samb Lo and Ch. Tidiane Seck, *Uniform convergence of the non-weighted poverty measures*, Communications in Statistics-Theory and Methods **38** (2009), no. 20, 3697–3704.
- [15] J. Praestgaard and J. A. Wellner, Exchangeably weighted bootstraps of the general empirical process, The Annals of Probability 21 (1993), no. 4, 2053–2086.
- [16] L. Rincón, *Introducción a los procesos estocásticos*, Departamento de Matemáticas, Facultad de Ciencias UNAM, México DF, 2012.
- [17] Ch. Tidiane Seck, Estimation non-paramétrique et convergence faible des mesures de pauvreté, Thèse de doctorat, Université Pierre et Marie Curie, mars 2011.
- [18] A. Sen, Poverty: an ordinal approach to measurement, Econometrica 44 (1976), no. 2, 219–231.
- [19] F. Strobl, On the reversed sub-martingale property of empirical discrepancies in arbitrary sample spaces, Journal of Theoretical Probability 8 (1995), no. 4, 825–831.
- [20] A. W. van der Vaart and J. A. Wellner, Weak convergence and empirical processes with applications to statistics, Springer-Verlag, New York, 1996.
- [21] J. A. Wellner, Some converse limit theorems for exchangeable bootstraps, The Institute of Mathematical Statistics Lecture Notes-Monograph Series 36 (2001), 593–606.
- [22] B. Zheng, Aggregate poverty measures, Journal of Economic Surveys 11 (1997), no. 2, 123–162.

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