

This set of notes aims to provide complete proofs of a number of asymptotic results regarding the Bootstrap [2] contained in Bickel and Freedman [1].

## 1 Bootstrap asymptotics for sample mean

**Theorem 1.1 (Bootstrap Central Limit Theorem for I.I.D. sample mean, Theorem 2.1 [1])**

Let  $(\Omega, \mathcal{A}, \nu)$  be a probability space. Let  $X, X_1, X_2, \dots : \Omega \rightarrow \mathbb{R}$  be a sequence of independent and identically distributed  $\mathbb{R}$ -valued random variables defined on  $\Omega$  *with finite expectation value  $\mu_X \in \mathbb{R}$  and variance  $\sigma_X^2 < \infty$* . For each  $n \in \mathbb{N}$  be fixed, define:

$$\bar{X}_n : \Omega \rightarrow \mathbb{R} : \omega \mapsto \frac{1}{n} \sum_{i=1}^n X_i(\omega).$$

For  $n, m \in \mathbb{N}$ , define  $\mathcal{S}_m^{(n)}$  to be the set of all functions from  $\{1, 2, \dots, m\} \rightarrow \{1, 2, \dots, n\}$ . Thus, each

$$s = (s(1), s(2), \dots, s(m)) \in \mathcal{S}_m^{(n)}$$

can be regarded as a length- $m$  finite (ordered) sequence of positive integers between 1 and  $n$ , inclusive. Note that  $\mathcal{S}_m^{(n)}$  is a finite set with  $|\mathcal{S}_m^{(n)}| = n^m$ . Endow  $\mathcal{S}_m^{(n)}$  with the probability space structure induced by the uniform probability function:

$$P_{\mathcal{S}_m^{(n)}}(s) := \frac{1}{n^m}, \quad \text{for each } s \in \mathcal{S}_m^{(n)}.$$

Let  $\Omega \times \mathcal{S}_m^{(n)}$  be the product probability space of  $\Omega$  and  $\mathcal{S}_m^{(n)}$ . Define:

$$\bar{X}_m^{(n)} : \Omega \times \mathcal{S}_m^{(n)} \rightarrow \mathbb{R} : (\omega, s) \mapsto \frac{1}{m} \sum_{j=1}^m X_{s(j)}(\omega).$$

For each  $\omega \in \Omega$ , define:

$$\Phi_{m,\omega}^{(n)} : \mathcal{S}_m^{(n)} \rightarrow \mathbb{R} : s \mapsto \sqrt{m} \left( \bar{X}_m^{(n)}(\omega, s) - \bar{X}_n(\omega) \right)$$

Then,

$$P \left( \Phi_{m,\omega}^{(n)} \xrightarrow{d} N(0, \sigma_X^2), \text{ as } n, m \rightarrow \infty \right) = \nu \left( \left\{ \omega \in \Omega \mid \Phi_{m,\omega}^{(n)} \xrightarrow{d} N(0, \sigma_X^2), \text{ as } n, m \rightarrow \infty \right\} \right) = 1.$$

**Remark 1.2**

For each fixed  $\omega \in \Omega$ ,  $\left\{ \Phi_{m,\omega}^{(n)} : \mathcal{S}_m^{(n)} \rightarrow \mathbb{R} \right\}_{n,m \in \mathbb{N}}$  is a doubly indexed sequence of  $\mathbb{R}$ -valued random variables. Note that their respective domains  $\mathcal{S}_m^{(n)}$  are pairwise distinct probability spaces. The **Bootstrap Central Limit Theorem for I.I.D. sample mean** asserts that for almost every  $\omega \in \Omega$ , the doubly indexed sequence  $\left\{ \Phi_{m,\omega}^{(n)} \right\}$  of  $\mathbb{R}$ -valued random variables converges in distribution to  $N(0, \sigma_X^2)$  as  $n, m \rightarrow \infty$ .

**Remark 1.3** The following results are well known from classical asymptotic theory:

By the **Weak Law of Large Numbers**,  $\bar{X}_n$  converges in probability to  $\mu_X$ , as  $n \rightarrow \infty$ ; in other words,

$$\lim_{n \rightarrow \infty} P \left( |\bar{X}_n - \mu_X| > \varepsilon \right) = \lim_{n \rightarrow \infty} \nu \left( \left\{ \omega \in \Omega : |\bar{X}_n(\omega) - \mu_X| > \varepsilon \right\} \right) = 0, \quad \text{for each } \varepsilon > 0.$$

By the **Strong Law of Large Numbers**,  $\bar{X}_n$  converges almost surely to  $\mu_X$ , as  $n \rightarrow \infty$ ; in other words,

$$P \left( \lim_{n \rightarrow \infty} \bar{X}_n = \mu_X \right) = \nu \left( \left\{ \omega \in \Omega \mid \lim_{n \rightarrow \infty} \bar{X}_n(\omega) = \mu_X \right\} \right) = 1.$$

By the **Central Limit Theorem**,  $\sqrt{n}(\bar{X}_n - \mu_X)$  converges in distribution to  $N(0, \sigma_X^2)$ .

## References

- [1] BICKEL, P. J., AND FREEDMAN, D. A. Some asymptotic theory for the bootstrap. *The Annals of Statistics* 9, 6 (1981), 1196–1217.
- [2] EFRON, B. Bootstrap methods: another look at the jackknife. *The Annals of Statistics* 7, 1 (1979), 1–26.