# Breiman's Law of Large Numbers

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## 1 Introduction

This seminar will follow the proof of Breiman's Law of Large Numbers from [Bre60] and [BQ16a]. We'll begin with introducing the conceptually similar Birkhoff Ergodic Theorem (Compact Metric, Uniquely Ergodic case) with proof from [EW11].

### 2 Birkhoff - Uniquely Ergodic Case

Recall the Birkhoff Ergodic Theorem, and the Ergodic case:

**Theorem 1** (Birkhoff Ergodic Theorem). Let  $(X, \mathcal{B}, \mu, T)$  be a p.p.s.,  $f \in L^1(X, \mathcal{B}, \mu)$ . Then for  $\mu$ -a.e.  $x \in X$ :

$$f^{*}(x) = \lim_{N \to \infty} \frac{1}{N} \sum_{n=0}^{N-1} f(T^{n}x)$$
(1)

The limit exists,  $f^* \in L^1$ ,  $f^* = f^* \circ T$ . Denote by  $\mathcal{E} \subseteq \mathcal{B}$  the set of T invariant subsets. Then,  $\forall A \in \mathcal{E}$ :

$$\int_{A} f^* \,\mathrm{d}\mu = \int_{A} f \,\mathrm{d}\mu \tag{2}$$

Note:  $f^* = E(f \mid \mathcal{E})$ 

**Corollary 1** (Pointwise Ergodic Theorem). If T is ergodic, then  $\mathcal{E}$  is trivial, and then  $\forall f \in L^1$  and for  $\mu$ -a.e.  $x \in X$ :

$$\lim_{N \to \infty} \frac{1}{N} \sum_{n=0}^{N-1} f(T^n x) = \int f \, \mathrm{d}\mu$$
(3)

Now we shall continue to an even more specific case, when the space X is a compact metric space, and T is uniquely ergodic (recall, a uniquely ergodic transformation is one that admits a single invariant probability measure).

**Theorem 2** (Birkhoff - Uniquely Ergodic Case). Let  $(X, \mathcal{B}, \mu, T)$  be a p.p.s., where X is a compact metric space and T is uniquely ergodic. The for every  $x \in X$ :

$$\lim_{N \to \infty} \frac{1}{N} \sum_{n=0}^{N-1} f(T^n x) = \int f \, \mathrm{d}\mu$$
(4)

And this convergence is **uniform** across X.

We will need the following lemma (see appendix - 4.1 for proof):

**Lemma 1.** Let  $(X, \mathcal{B}, \mu, T)$  be a p.p.s., where X is a compact metric space. Then the space of probability measures: Prob(X) is weak-\* sequentially compact. It is characterized by:

$$\mu_n \to \mu \Leftrightarrow \forall f \in C(X) \quad \int_X f \, \mathrm{d}\mu_n \to \int_X f \, \mathrm{d}\mu \tag{5}$$

We'll start by proving another lemma:

**Lemma 2.** Let  $T: X \to X$  be a continuous map of a compact metric space X, and  $(\nu_n)$  be a sequence in Prob (X). Then any weak-\* limit point of the sequence  $(\mu_n)$  defined by  $\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} T_*^j \nu_n$  is in Prob<sup>T</sup> (X) (the space of T invariant probability measures).

Note: since by Lemma 1,  $\operatorname{Prob}^{T}(X)$  is weak-\* compact, the sequence must have some limit point.

*Proof.* Notice:  $||f||_{\infty} = \sup_{x \in X} |f| < \infty$  by compactness of X and continuity of f.

Let  $\mu_{n_k} \to \mu$  be a converging subsequence. We'd like to show that  $\mu$  is T invariant, i.e.  $\forall f \in C(X)$ :

$$\int_{X} f \, \mathrm{d}\mu = \int_{X} f \circ T \, \mathrm{d}\mu \tag{6}$$

Recall, that by convergence we have:

$$\mu_{n_k} \to \mu \Rightarrow \forall g \in C(X) \ \int_X g \, \mathrm{d}\mu_{n_k} \to \int_X g \, \mathrm{d}\mu \tag{7}$$

So we observe:

$$0 \le \left| \int_X f \, \mathrm{d}\mu_{n_k} - \int_X f \circ T \, \mathrm{d}\mu_{n_k} \right| = \left| \frac{1}{n_k} \sum_{j=0}^{n_k-1} \int_X f \circ T^{j-1} \, \mathrm{d}\nu_{n_k} - \frac{1}{n_k} \sum_{j=0}^{n_k-1} \int_X f \circ T^j \, \mathrm{d}\nu_{n_k} \right| = \left| \frac{1}{n_k} \left( \int_X f \, \mathrm{d}\nu_{n_k} - \int_X f \circ T^{n_k} \, \mathrm{d}\nu_{n_k} \right) \right| \le \frac{1}{n_k} 2 \|f\|_{\infty} \to 0 \quad (8)$$

And so we must have that  $\int_X f \, d\mu = \int_X f \circ T \, d\mu$  as required, so  $\mu \in \operatorname{Prob}^T(X)$ .  $\Box$ 

Proof of Theorem 2. Denote by  $\mu$  the unique invariant measure. Take the sequence  $(\delta_x)$  and apply Lemma 2 (i.e. in the notation of the lemma, we have:  $\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} \delta_{T^j x}$ ).

As the limit of every converging subsequence must be invariant (by the lemma) then by uniqueness it must be  $\mu$ . By compactness there are converging subsequences, and as the limit is unique we must have that:

$$\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} \delta_{T^j x} \to \mu \tag{9}$$

This is convergence in weak-\* topology, so we have that for every  $f \in C(X)$ :

$$\frac{1}{n}\sum_{j=0}^{n-1}f\left(T^{j}x\right)\to\int_{X}f\,\mathrm{d}\mu\tag{10}$$

And so we have everywhere convergence.

Now assume this convergence is not uniform. Denote:  $E[f] = \int_X f \, d\mu$ . So there is some  $f \in C(X)$ ,  $\epsilon > 0$  and  $(N_k)$ , such that  $N_k \to \infty$  and for every  $N_k$  exists  $x_{N_k} \in X$  such that:

$$\left|\frac{1}{N_k}\sum_{j=0}^{N_k-1} f\left(T^j x_{N_k}\right) - E\left[F\right]\right| > \epsilon \tag{11}$$

Taking the sequence  $\left(\delta_{x_{N_k}}\right)$  and in a similar manner to that of Lemma 2, we conclude that:

$$\mu_k = \frac{1}{N_k} \sum_{j=0}^{N_k - 1} \delta_{T^j x_{N_k}} \to \mu$$
 (12)

And so:

$$\int_{X} f \, \mathrm{d}\mu_k \to \int_{X} f \, \mathrm{d}\mu = E\left[f\right] \tag{13}$$

But by (11), notice that for all  $N_k$ :

$$\left| \int_{X} f \, \mathrm{d}\mu_{k} - E\left[F\right] \right| = \left| \frac{1}{N_{k}} \sum_{j=0}^{N_{k}-1} f\left(T^{j} x_{N_{k}}\right) - E\left[F\right] \right| > \epsilon \tag{14}$$

And so the limit must be  $\geq \epsilon$  - a contradiction. Therefore we must have uniform convergence.

#### **3** Breiman's Law of Large Numbers

We will observe a Markov random walk on a compact metric space  $(\Omega, \mathcal{B})$ , determined by the transition probabilities,  $P(A \mid x)$  from x into  $A \in \mathcal{B}$ , or alternatively defined by a Borel map from  $\Omega$  to to  $\operatorname{Prob}(\Omega)$ ,  $x \mapsto P_x$ .

We observe the space  $(\Omega^{\mathbb{N}}, \mathcal{B}^{\mathbb{N}})$  the space of infinite sequences with coordinates in  $\Omega$ , describing the random walk, and denote by  $X_n$  the random variable which describes the position at the *n*-th step, where  $X_0$  is the starting point of the random walk. For each  $x \in \Omega$ , we define a probability  $\nu_x$  on  $(\Omega^{\mathbb{N}}, \mathcal{B}^{\mathbb{N}})$  induced by P when  $X_0 = x$ .

We shall consider only Markov-Feller processes: If T is the action defined on functions as:

$$T\phi(x) = \int_{\Omega} \phi \, \mathrm{d}P_x = E\left(\phi(X_1) \mid X_0 = x\right) \tag{15}$$

Notice that T is a bounded operator w.r.t. the sup norm (as an operator on bounded functions): for every point, the value is an average of the original function over some set, and therefore not greater than the sup of the original function. Therefore,  $||T(f)||_{\infty} \leq ||f||_{\infty}$ , i.e. T is bounded. Note that as our space is compact metric, all continuous functions are bounded.

A Markov process is a Markov-Feller process if T acts on continuous functions (i.e. T(f) is continuous if f is). Note: This is equivalent to the map  $x \mapsto P_x$  being continuous w.r.t. the weak-\* topology. Denote this map by  $M(x) = P_x$ : In one direction, assume T acts on continuous functions. Notice that if  $x_n \to x$ , then for every  $f \in C(\omega)$ , as Tf is continuous:

$$M(x_{n})(f) = P_{x_{n}}(f) = \int_{\Omega} f \, \mathrm{d}P_{x_{n}} = T\phi(x_{n}) \to T\phi(x) = \int_{\Omega} f \, \mathrm{d}P_{x} = P_{x}(f) = M(x)(f)$$
(16)

And so M is continuous w.r.t. to weak-\*, as the integrals converge for every continuous f.

Now assume that M is continuous. Then for every  $f \in C(\omega), x_n \to x$ :

$$T\phi(x_n) = \int_{\Omega} f \, dP_{x_n} = P_{x_n}(f) = M(x_n)(f) \to M(x)(f) = P_x(f) = \int_{\Omega} f \, dP_x = T\phi(x)$$
(17)

And so Tf is continuous as required.

**Example 1.** (Group Action) Consider the case where we have a random walk on a compact metric space  $(\Omega, \mathcal{B})$ , determined by a (continuous) group action of G on  $\Omega$ , where choice of G action is determined by a compactly supported probability measure on G,  $\mu_G$ .

This is a Markov-Feller operator, as we can see that for  $\phi \in C(\Omega)$ :

$$T\phi(x) = \int_{\Omega} \phi \, \mathrm{d}P_x = \int_G \phi(gx) \,\mu_G(g) \tag{18}$$

Here we have that  $P_x$  is the push forward of  $\mu_G$  under the map  $g \mapsto gx$ .

Notice that because  $\Omega$  is compact,  $\phi$  is equicontinuous. Choose some  $\epsilon > 0$ , and  $\epsilon_1 = \epsilon(\epsilon) > 0$  that exists by equicontinuity of  $\phi$  s.t. if  $|x - y| < \epsilon_1$ , then  $|\phi(x) - \phi(y)| < \epsilon$ . Each g is also equicontinuous, with a corresponding  $\epsilon_g$  s.t. if  $|x - y| < \epsilon_g$  then  $|gx - gy| < \frac{\epsilon_1}{2}$ . For each g we can take the neighborhood where the difference is  $< \frac{\epsilon_1}{4}$  by the sup norm, and this gives an open cover, so by compactness we have a finite subcover. Note that for g we have that if h is in it's neighborhood, then if  $|x - y| < \epsilon_g$  then  $|gx - gy| < \epsilon_1$ . Taking the minimum over the  $\epsilon_g$  for the g's chosen for the subcover, we get some  $\epsilon_2 > 0$  such that for all g, if  $|x - y| < \epsilon_2$  then  $|gx - gy| < \epsilon_1$ . And so  $|\phi(gx) - \phi(gy)| < \epsilon$ , so integrating (against the probability measure  $\mu_G$ ) gives us that  $|T\phi(y) - T\phi(x)| < \epsilon$  - i.e. we have continuity.

**Definition 1.** (Stationary Measure) A measure  $\pi$  on  $\Omega$  shall be called stationary if  $\forall A \in \mathcal{B}$ :

$$\pi(A) = \int_{\Omega} P(A \mid x) \, \mathrm{d}\pi(x) \tag{19}$$

Note that in Example 1, we have that this translates to:  $\forall A \in \mathcal{B}$ :

$$\pi(A) = \int_{\Omega} P(A \mid x) \, \mathrm{d}\pi(x) = \int_{G} \int_{\Omega} \mathbb{1}_{A}(gx) \, \mathrm{d}\pi(x) \, \mathrm{d}\mu_{G}(g) = \mu_{G} * \pi(A) \tag{20}$$

I.e.,  $\pi$  is stationary if  $\pi = \mu_G * \pi$ , where  $\mu_G * \pi$  is convolution of the two measures, which is defined by:

$$\mu_G * \pi (A) = \int_G \int_\Omega \mathbb{1}_A (gx) \, \mathrm{d}\pi (x) \, \mathrm{d}\mu_G(g) \tag{21}$$

We shall see that in our setting, a stationary measure always exists. Recall we have the operator  $T: C \to C$  defined by:

$$T\phi(x) = \int_{\Omega} \phi \, \mathrm{d}P_x = E\left(\phi(X_1) \mid X_0 = x\right) \tag{22}$$

We have:  $(T^k \phi)(x) = E(\phi(X_k) | X_0 = x)$ . We've shown that T is bounded, with  $||Tf||_{\infty} \leq ||f||_{\infty}$ , so  $||T^k f||_{\infty} \leq ||f||_{\infty}$  too.

In addition denote:  $\bar{T}_N \phi = \frac{1}{N} \sum_{n=1}^N T^n \phi.$ 

**Proposition 1.** Observe a Markov-Feller random walk on  $\Omega$ . Let  $(\nu_n)$  be a sequence in Prob  $(\Omega)$ . Then any weak-\* limit point of the sequence  $(\mu_n)$  defined by  $\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} T_*^j \nu_n$  is stationary.

*Proof.* Proof is somewhat similar to that of Lemma 2.

Denote a limit point of a subsequence by  $\mu$ . We would like to show that is is a stationary measure, i.e. that for all measurable A:

$$\mu(A) = \int_{\Omega} P(A \mid x) \, d\mu = \int_{\Omega} E(\mathbb{1}_A(X_1) \mid X_0 = x) \, d\mu = (T_*\mu)(A)$$
(23)

So we would like to show that  $\mu = T_*\mu$ . It would suffice to prove for each  $f \in C(\Omega)$  that:

$$\int_{\Omega} f \, \mathrm{d}\mu = \int_{\Omega} f \, \mathrm{d}\left(T_*\mu\right) = \int_{\Omega} Tf \, \mathrm{d}\mu \tag{24}$$

But as  $f \in C(\Omega)$  we have that  $Tf \in C(\Omega)$ , so by weak-\* convergence of the subsequence  $(\mu_{n_k})$ , we have:

$$\int_{\Omega} f \, \mathrm{d}\mu_{n_k} \to \int_{\Omega} f \, \mathrm{d}\mu \qquad \text{and} \qquad \int_{\Omega} Tf \, \mathrm{d}\mu_{n_k} \to \int_{\Omega} Tf \, \mathrm{d}\mu \tag{25}$$

Observe that that:

$$0 \leq \left| \int_{\Omega} f \, \mathrm{d}\mu_{n_{k}} - \int_{\Omega} Tf \, \mathrm{d}\mu_{n_{k}} \right| = \left| \frac{1}{n_{k}} \sum_{j=0}^{n_{k}-1} \int_{\Omega} T^{j-1} f \, \mathrm{d}\nu_{n_{k}} - \frac{1}{n_{k}} \sum_{j=0}^{n_{k}-1} \int_{\Omega} T^{j} f \, \mathrm{d}\nu_{n_{k}} \right| = \left| \frac{1}{n_{k}} \left( \int_{\Omega} f \, \mathrm{d}\nu_{n_{k}} - \int_{\Omega} T^{n_{k}} f \, \mathrm{d}\nu_{n_{k}} \right) \right| \leq \frac{1}{n_{k}} \left( \|f\|_{\infty} + \|T^{k}f\|_{\infty} \right) \leq \frac{2}{n_{k}} \|f\|_{\infty} \to 0 \quad (26)$$

And therefore we must have that the limits are equal as required, and  $\mu$  is stationary.

**Corollary 2.** In the setting of a Markov-Feller random walk on a compact metric base space  $(\Omega, \mathcal{B})$ , a stationary measure always exists.

*Proof.* Take an arbitrary sequence of measures  $(\nu_n)$  in Prob  $(\Omega)$ , and observe the sequence  $(\mu_n)$  as defined in Proposition 1. By Lemma 1, Prob  $(\Omega)$  is sequentially compact, so  $(\mu_n)$  has a converging subsequence to some  $\mu$ . By Proposition 1 this limit is stationary, and so a stationary probability measure exists, as required.  $\Box$ 

From here on, we shall assume that there exists a **unique** stationary measure.

**Theorem 3** (Breiman's Law of Large Numbers). In the setting of a Markov-Feller random walk on a compact metric base space  $(\Omega, \mathcal{B})$  with a unique stationary measure  $\pi$ , we have that  $\forall \phi \in C(\Omega)$ , then for any  $x \in \Omega$  for  $\nu_x$  a.e. sequence  $(X_N)$ :

$$\frac{1}{N}\sum_{n=1}^{N}\phi\left(X_{n}\right)\rightarrow\int_{\Omega}\phi\left(x\right) \,\mathrm{d}\pi\left(x\right) \tag{27}$$

Note the similarities to the Birkhoff theorem: the time averages converge to the space average, for any starting point, and almost every path.

**Proposition 2.** For any  $\phi \in C(\Omega)$ ,  $\overline{T}_N(\phi)$  converges uniformly to  $\pi(\phi) = \int_{\Omega} \phi \, d\pi$ .

*Proof.* Proof is somewhat similar to that of Theorem 2. Let  $(\nu_n)$  be a sequence in Prob $(\Omega)$ . Then any weak-\* limit point of the sequence  $(\mu_n)$  defined by  $\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} T_*^j \nu_n$  is stationary by Proposition 1, and therefore  $\pi$ . By compactness, this is the limit of the entire sequence  $(\mu_n)$ .

As  $\phi$  is continuous, we get that taking  $\nu_n^x = \delta_x$  gives us everywhere convergence, i.e.:

$$\bar{T}_{n}(\phi)(x) = \int_{\Omega} \phi \, \mathrm{d}\mu_{n}^{x} \to \int_{\Omega} \phi \, \mathrm{d}\pi$$
(28)

Now assume this convergence is not uniform. Denote:  $E[\phi] = \int_{\omega} \phi \, d\mu$ . So there is some  $f \in C(\Omega)$ ,  $\epsilon > 0$ and  $(N_k)$ , such that  $N_k \to \infty$  and for every  $N_k$  exists  $x_{N_k} \in \Omega$  such that:

$$\left|\frac{1}{N_k}\sum_{j=0}^{N_k-1}T^jf(x_{N_k}) - E\left[F\right]\right| > \epsilon$$
(29)

Taking the sequence  $\left(\delta_{x_{N_k}}\right)$  we get:

$$\mu_k = \frac{1}{N_k} \sum_{j=0}^{N_k - 1} T^j_* \delta_{x_{N_k}} \to \pi$$
(30)

And so:

$$\int_{\Omega} f \, \mathrm{d}\mu_k \to \int_{\Omega} f \, \mathrm{d}\pi = E\left[f\right] \tag{31}$$

But by (29), notice that for all  $N_k$ :

$$\left| \int_{\Omega} f \, \mathrm{d}\mu_k - E\left[F\right] \right| = \left| \frac{1}{N_k} \sum_{j=0}^{N_k - 1} T^j f\left(x_{N_k}\right) - E\left[F\right] \right| > \epsilon \tag{32}$$

And so the limit must be  $\geq \epsilon$  - a contradiction. Therefore we must have uniform convergence.

The following lemma is a derivative of Doob's martingale convergence theorem, and a proof can be found in 4.2.

**Lemma 3.** Let  $(Y_n)_{n\geq 1}$  be a sequence of random variables which are uniformly bounded in  $L^2$  (i.e.  $\exists M < \infty$  s.t. for all  $n \geq 1$ ,  $E(Y_n^2) < M$ ), and such that:

$$E(Y_n \mid Y_1, ..., Y_{n-1}) = 0 \quad \forall n \ge 1$$
(33)

Then the sequence  $\frac{1}{n} \sum_{k=1}^{n} Y_k$  converges to 0 almost surely and in  $L^2$ .

**Proposition 3.** Let  $\phi \in C(\Omega)$ ,  $x \in \Omega$ . Let  $(X_n)_{n \geq 1}$  be distributed according to  $\nu_x$  and define:

$$Z_{n}^{1} = \begin{cases} \phi(X_{n}) - E(\phi(X_{n}) \mid X_{n-1}) & n > 1\\ 0 & n \le 1 \end{cases}$$

$$Z_{n}^{k} = \begin{cases} E(\phi(X_{n}) \mid X_{n-k+1}) - E(\phi(X_{n}) \mid X_{n-k}) & n > k\\ 0 & n \le k \end{cases}$$
(34)

Then  $\frac{1}{N}\sum_{n=1}^{N} Z_n^k \to 0$  a.s.  $\nu_x$ .

*Proof.* To prove we would like to use Lemma 3. First note that clearly as  $Z_n^k \leq 2 \|\phi\|_{\infty}$ , we have that  $E\left(\left(Z_n^k\right)^2\right) \leq 4 \|\phi\|_{\infty}^2 < \infty$  for all n, k.

Next we must check that:  $E(Z_n^k | Z_{n-1}^k, ..., Z_1^k) = 0$ . Note that  $Z_{n-1}^k, ..., Z_1^k$  are random variables in the  $\sigma$ -algebra of  $X_{n-k}, ..., X_1$ , so:

$$E\left(Z_{n}^{k} \mid Z_{n-1}^{k}, ..., Z_{1}^{k}\right) = E\left(E\left(Z_{n}^{k} \mid X_{n-k}, ..., X_{1}\right) \mid Z_{n-1}^{k}, ..., Z_{1}^{k}\right)$$
(35)

Notice that by the Markov property:

$$E(Z_{n}^{k} | X_{n-k}, ..., X_{1}) = E(Z_{n}^{k} | X_{n-k}) = E(E(\phi(X_{n}) | X_{n-k+1}) | X_{n-k}) - E(\phi(X_{n}) | X_{n-k}) = E(E(\phi(X_{n}) | X_{n-k+1}, X_{n-k}) | X_{n-k}) - E(\phi(X_{n}) | X_{n-k}) = E(\phi(X_{n}) | X_{n-k}) - E(\phi(X_{n}) | X_{n-k}) = 0 \quad (36)$$

And so we have all that is required in order to use Lemma 3, which gives us the desired result.  $\Box$ 

Proof. (Breiman's Law of Large Numbers)

To prove the theorem, we first notice that for n > k:

$$\phi(X_n) - E(\phi(X_n) \mid X_{n-k}) = \sum_{i=1}^{k} Z_n^i$$
(37)

By Proposition 3, we have that for  $(X_n)$  a.e.- $\nu_x$  (notice that we are neglecting k bounded terms when comparing to  $\sum_{i=1}^{k} \sum_{n=1}^{N} Z_n^i$ , but in the limit these are negligible - as we take k to be constant, and we have the  $\frac{1}{N}$  normalization):

$$\left|\frac{1}{N}\sum_{n=1}^{N}\phi(X_{n}) - \frac{1}{N}\sum_{n=k+1}^{N}E(\phi(X_{n}) \mid X_{n-k})\right| \to 0$$
(38)

Again neglecting at most k bounded (by  $\|\phi\|_{\infty}$ ) terms, we get:

$$\left|\frac{1}{N}\sum_{n=1}^{N}\phi(X_{n}) - \frac{1}{N}\sum_{n=1}^{N}E(\phi(X_{n+k}) \mid X_{n})\right| \to 0$$
(39)

Summing up this limit for  $k \leq M$  for some fixed M (and dividing by M) we get get:

$$\left|\frac{1}{N}\sum_{n=1}^{N}\phi(X_{n}) - \frac{1}{N}\sum_{n=1}^{N}\frac{1}{M}\sum_{k=1}^{M}E\left(\phi(X_{n+k}) \mid X_{n}\right)\right| \to 0$$
(40)

By Proposition 2, for any  $\epsilon > 0$ , we can choose an M, s.t. for all n:

$$\left|\frac{1}{M}\sum_{k=1}^{M} E\left(\phi\left(X_{n+k}\right) \mid X_{n}\right) - \pi\left(\phi\right)\right| < \epsilon$$

$$(41)$$

As we have uniform convergence.

For such a choice of M, we get via substituting in (40) that  $\nu_x$ -a.e.:

$$\lim_{N} \left| \frac{1}{N} \sum_{n=1}^{N} \phi\left(X_{n}\right) - \pi\left(\phi\right) \right| \le \epsilon$$

$$\tag{42}$$

As  $\epsilon$  is arbitrary, taking a countable sequence going to 0 will give us the required result, that  $\nu_x$ -a.e.:

$$\lim_{N} \left| \frac{1}{N} \sum_{n=1}^{N} \phi\left(X_{n}\right) - \pi\left(\phi\right) \right| = 0$$

$$\tag{43}$$

### 4 Appendix

#### 4.1 Proof of Lemma 1

The space of signed measures (linear functionals), M is the dual of  $C_0(X)$ . As X is compact, we have  $C_0(X) = C_b(X) = C(X)$ . By Banach-Alaoglu, the unit ball B in M is sequentially compact w.r.t. the weak-\* topology. Notice that  $\operatorname{Prob}(X) \subseteq B$ , and that it is closed, as if  $\mu_n \to \mu$  for  $\mu_n \in \operatorname{Prob}(X)$ , then  $\mu$  must still be a positive functional (as  $\mu_n$  are), and as  $\mathbb{1}_X \in C_b(X) = C_0(X)$ , we have that:

$$1 = \mu_n(X) = \langle \mathbb{1}_X, \mu_n \rangle \to \langle \mathbb{1}_X, \mu \rangle = \mu(X)$$
(44)

And so  $\mu(X) = 1$  (and is a positive functional), and therefore  $\mu \in \text{Prob}(X)$ . Therefore Prob(X) is a closed subset of a sequentially compact set (the unit ball in M), and is therefore sequentially compact w.r.t. the weak-\* topology.

#### 4.2 Proof of Lemma 3

See [BQ16b], Corollary A.8.

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