

# The Convergence Theorem for Discrete-time Martingales

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## Abstract

*This paper gives an insight to the mathematical theory of the discrete-time martingale, including basic definitions and results. Still, the paper is focused on proving a variant of the Convergence Theorem based on Doob's Inequality for martingales. The second part of this paper presents a proof of The Law of Large Numbers for discrete-times martingales, as an application of the Convergence Theorem.*

**Key words:** martingale, submartingale, stopping time

## Introduction

Let  $(f_n)_{n \geq 0}$  be a real valued family of random variables defined on the probability space  $(\Omega, \mathcal{F}, P)$  and let  $(\mathcal{F}_n)_{n=1,2,\dots}$  be a filtration (an increasing family of borelian fields contained in  $\mathcal{F}$ ). The stochastic process  $(f_n)_{n \geq 0}$  is said to be adapted to the family  $(\mathcal{F}_n)_{n=1,2,\dots}$  if, for all  $n$ ,  $f_n$  is  $\mathcal{F}_n$  - measurable.

**Definition 1.** The stochastic process  $(f_n)_{n \geq 0}$ , adapted to the filtration  $(\mathcal{F}_n)_{n=1,2,\dots}$ , is a martingale (respectively submartingale) with respect to the filtration, if the following conditions hold:

1.  $f_n$  is  $P$  -integrable for all  $n$ ;
2. For all  $n > m$

$$E[f_n / \mathcal{F}_m] = f_m \text{ almost surely (a.s.),} \quad (1)$$

respectively

- 2'. For all  $n > m$

$$E[f_n / \mathcal{F}_m] \geq f_m \text{ a.s. - submartingale case.} \quad (2)$$

**Remark 1.** Due to the conditional expectation properties the above conditions may be written:

2. For all  $A \in \mathcal{F}_m$ , and for all  $n > m$

$$\int_A f_n dP = \int_A f_m dP \Leftrightarrow E[f_n] = E[f_m]. \quad (3)$$

respectively

2'. For all  $A \in \mathcal{F}_n$ , and for all  $n > m$

$$\int_A f_n dP \geq \int_A f_m dP \Leftrightarrow E[f_n] \geq E[f_m] \text{ - submartingale case.} \quad (4)$$

## Stopping Times and Stopping Theorem

**Definition 2.** A random variable  $T$  defined on  $(\Omega, \mathcal{F}, P)$  and valued in  $\{0, 1, \dots, \infty\}$  is called a stopping time (or Markov time) with respect to the filtration  $(\mathcal{F}_n)_{n=1, 2, \dots}$ , or simply  $\mathcal{F}_n$  - stopping time if:

$$\text{for all } n \geq 0, \{T \leq n\} \in \mathcal{F}_n. \quad (5)$$

$$\text{or equivalent, for all } n \geq 0, \{T = n\} \in \mathcal{F}_n. \quad (6)$$

To any stopping time  $T$  we may associate a borelian field  $\mathcal{F}_T$  defined as following:

$$\mathcal{F}_T = \left\{ A \in \mathcal{B} \left( \bigcup_{n \geq 0} \mathcal{F}_n \right) : A \cap \{T \leq n\} \in \mathcal{F}_n, \forall n \geq 0 \right\}, \quad (7)$$

equivalent to:

$$\mathcal{F}_T = \left\{ A \in \mathcal{B} \left( \bigcup_{n \geq 0} \mathcal{F}_n \right) : A \cap \{T = n\} \in \mathcal{F}_n, \forall n \geq 0 \right\}, \quad (8)$$

where  $\mathcal{B} \left( \bigcup_{n \geq 0} \mathcal{F}_n \right)$  is the borelian field generated by  $\bigcup_{n \geq 0} \mathcal{F}_n$ . We consider that  $\mathcal{F}_T$  represents the events known until the moment  $T$ .

Let  $(f_n)_{n \geq 0}$  be a  $\mathcal{F}_n$  - adapted family of random variables and let  $T : \Omega \rightarrow \{0, 1, \dots, \infty\}$  be a stopping time. We define the random variable

$$f_T : \Omega \rightarrow R, f_T(\omega) = \begin{cases} 0 & , T(\omega) = \infty \\ f_{T(\omega)}(\omega) & , T(\omega) < \infty \end{cases}. \quad (9)$$

We may easily observe that  $f_T$  is  $\mathcal{F}_T$  - measurable.

**Proposition 1.** If  $(f_n)_{n \geq 0}$  is  $\mathcal{F}_n$  - adapted, then for all  $A \in \mathcal{B}_R$  the random variable

$$D_A(\omega) = \begin{cases} \min\{n : f_n(\omega) \in A\} & , \{n : f_n(\omega) \in A\} \neq \emptyset \\ \infty & , \text{otherwise} \end{cases} \quad (10)$$

is a  $\mathcal{F}_n$  - stopping time (it is called the moment of the first entrance in  $A$ ).

The following result will be given without proof. The ones who are interested in the proof may consult [2].

**Theorem 1.** (The Stopping Theorem for bounded stopping times) Let  $(f_n)_{n \geq 0}$  be a  $\mathcal{F}_n$  - submartingale (respectively  $\mathcal{F}_n$  - martingale) and let  $T_1 \leq T_2$  be two stopping times almost surely finite. Let us suppose that  $T_1 \leq T_2 \leq m \in (0, \infty)$ , then:

$$E[f_{T_2} / \mathcal{F}_{T_1}] \geq f_{T_1}, \quad (11)$$

respectively

$$E[f_{T_2} / \mathcal{F}_{T_1}] = f_{T_1}. \quad (12)$$

In particular,

$$E[f_{T_2} / \mathcal{F}_{T_1}] \geq E[f_{T_1}], \quad (13)$$

respectively

$$E[f_{T_2} / \mathcal{F}_{T_1}] = E[f_{T_1}]. \quad (14)$$

### Doob's Maximal Lemma

**Theorem 2.** (Doob's Maximal Lemma) Let  $(f_n)_{n \geq 0}$  be a nonnegative  $\mathcal{F}_n$  - submartingale. Then for all  $\varepsilon > 0$  and  $n = 1, 2, \dots$  :

$$P\left(\max_{0 \leq k \leq n} f_k \geq \varepsilon\right) \leq \frac{1}{\varepsilon} \int_{\{\max_{0 \leq k \leq n} f_k \geq \varepsilon\}} f_n dP \leq \frac{E[f_n]}{\varepsilon}. \quad (15)$$

**Proof.** The second inequality in (15) is obvious if considering the integral properties. In order to prove the first inequality, we define the random variable  $f_n^* = \max_{0 \leq k \leq n} f_k$  and the stopping time  $T = \min(k \geq 0, f_k \geq \varepsilon)$ . According to Proposition 1 we have  $T = D_{[\varepsilon, \infty)}$ . We use Theorem 1 for the bounded stopping times  $T \wedge n \leq n$ , and also the equality  $\{f_n^* \geq \varepsilon\} = \{f_{T \wedge n} \geq \varepsilon\}$ . Considering that on the set  $\{f_n^* \geq \varepsilon\}$  the inequalities  $T \leq n$  and  $f_T \geq \varepsilon$  hold, we obtain:

$$\varepsilon P(f_n^* \geq \varepsilon) = \varepsilon P(f_{T \wedge n} \geq \varepsilon) = \int_{\{f_{T \wedge n} \geq \varepsilon\}} \varepsilon dP \leq \int_{\{f_{T \wedge n} \geq \varepsilon\}} f_{T \wedge n} dP.$$

Due to Theorem 1 (13) we have:

$$\int_{\{f_{T \wedge n} \geq \varepsilon\}} f_{T \wedge n} dP \leq \int_{\{f_{T \wedge n} \geq \varepsilon\}} f_n dP = \int_{\{f_n^* \geq \varepsilon\}} f_n dP.$$

Finally, the relation (15) is provided as:

$$P(\{f_n^* \geq \varepsilon\}) \leq \frac{1}{\varepsilon} \int_{\{f_n^* \geq \varepsilon\}} f_n dP.$$

**Proposition 2.** If  $(f_n)_{n \geq 0}$  is a  $\mathcal{F}_n$  -martingale so that  $E(|f_n|^p) < \infty$  for all  $n$  and for a  $p \geq 1$ , then

$$\text{for all } \varepsilon > 0, P\left(\max_{0 \leq k \leq n} |f_k| \geq \varepsilon\right) \leq \frac{E[|f_n|^p]}{\varepsilon^p}. \quad (16)$$

**Proof.** We may prove that if  $(f_n)_{n \geq 0}$  is a  $\mathcal{F}_n$ -martingale,  $p \geq 1$ , and  $E(|f_n|^p) < \infty$  for all  $n$ , then  $(|f_n|^p)_{n \geq 0}$  is a  $\mathcal{F}_n$ -martingale. Then, by (15) we have:

$$P\left(\max_{0 \leq k \leq n} |f_k| \geq \varepsilon\right) \leq P\left(\max_{0 \leq k \leq n} |f_k|^p \geq \varepsilon^p\right) \leq \frac{E(|f_n|^p)}{\varepsilon^p}.$$

## The Convergence Theorem and The Law of Large Numbers

**Theorem 3** (The Convergence Theorem). Let  $(f_n)_{n \geq 0}$  be a  $\mathcal{F}_n$ -martingale so that  $\sup_{n \geq 0} E(f_n^2) < \infty$ . Then there exists a random variable  $f$  such that  $f_n \xrightarrow{a.s.} f$  and  $f_n \xrightarrow{L^2} f$  and, moreover:

$$\text{for all } n \geq 0, f_n = E[f / \mathcal{F}_n]. \quad (17)$$

**Proof.** We first notice that it is enough to prove the convergence in  $L^2$  in order to obtain the a.s. convergence. Therefore, let us suppose that  $f_n \xrightarrow{L^2} f$ . Then, by applying Proposition 2 (16) for the martingale  $(f_{n+k} - f_n)_{k \geq 1}$  and for  $p = 2$ , we have: for all  $\varepsilon > 0$ ,

$$P\left(\max_{0 \leq k \leq m} |f_{n+k} - f_n| \geq \varepsilon\right) \leq \frac{E|f_{n+m} - f_n|^2}{\varepsilon^2}.$$

As a consequence, we have:

$$P\left(\sup_k |f_{n+k} - f_n| \geq \varepsilon\right) = \lim_{m \rightarrow \infty} P\left(\max_{0 \leq k \leq m} |f_{n+k} - f_n| \geq \varepsilon\right) \leq \frac{1}{\varepsilon^2} \lim_{m \rightarrow \infty} E|f_{n+m} - f_n|^2.$$

Furthermore,  $(f_n)_{n \geq 0}$  is Cauchy in  $L^2$ , so  $\frac{1}{\varepsilon^2} \lim_{m \rightarrow \infty} E|f_{n+m} - f_n|^2 \xrightarrow{n \rightarrow \infty} 0$ .

We have obtained that  $(f_n)_{n \geq 0}$  is Cauchy a.s., implying that  $(f_n)_{n \geq 0}$  is convergent a.s.

We prove now the convergence in  $L^2$ . We may easily observe that for  $(f_n)_{n \geq 0}$  a  $\mathcal{F}_n$ -martingale such that  $E(|f_n|^2) < \infty$  for all  $n$ , then the following relations hold:

$$E\left[(f_{n+m} - f_n)^2 / \mathcal{F}_n\right] = E\left[(f_{n+m}^2 - f_n^2) / \mathcal{F}_n\right], \quad (18)$$

$$0 \leq \int (f_{n+m} - f_n)^2 dP = \int f_{n+m}^2 dP - \int f_n^2 dP. \quad (19)$$

Relation (19) implies that the sequence  $(\int f_n^2)_{n \geq 0}$  is increasing. The sequence is also bounded (as  $\sup_{n \geq 0} E(f_n^2) < \infty$ ), and thus convergent. From (19) we have  $\lim_{m, n \rightarrow \infty} \int (f_{n+m} - f_n)^2 dP = 0$ ,

meaning that  $(f_n)_{n \geq 0}$  converges in  $L^2$  to a limit  $f$ . This implies that  $(f_n)_{n \geq 0}$  converges in  $L^1$  to the limit  $f$ , meaning that  $\int_A f_n dP \xrightarrow{n \rightarrow \infty} \int_A f dP$ .

Particularly, because  $(f_n)_{n \geq 0}$  is a  $\mathcal{F}_n$ -martingale, we have:

$$\text{For all } n \geq k, E[f_n / \mathcal{F}_k] = f_k, \quad (20)$$

$$\text{For all } n \geq k, E[f_n] = E[f_k]. \quad (21)$$

That means that:

$$\text{For all } A \in \mathcal{F}_k \text{ and } n \geq k, \int_A f_k dP = \int_A f_n dP \rightarrow \int_A f dP. \quad (22)$$

Hence:

$$\text{For all } A \in \mathcal{F}_k, \int_A f_k dP = \int_A f dP. \quad (23)$$

The last relation justifies relation (17): for all  $n \geq 0, f_n = E[f / \mathcal{F}_n]$ .

We remind the following lemma known as Kronecker's Lemma, that will be used in the further proof.

**Lemma 1.** Let  $(a_n)_n$  be an increasing sequence of positive numbers having the limit  $\infty$  and let

$(x_n)_n$  be a sequence of real numbers such that  $\sum \frac{x_n}{a_n}$  is convergent. Then:

$$\frac{1}{a_n} \sum_{k=1}^n x_k \xrightarrow{n \rightarrow \infty} 0. \quad (24)$$

**Theorem 4.** (The Law of Large Numbers for martingales) Let  $(f_n)_{n \geq 0}$  be a  $\mathcal{F}_n$ -martingale such that:

$$\sum_{n=1}^{\infty} \frac{E(|f_n - f_{n-1}|^2)}{n^2} < \infty. \quad (25)$$

Then: 
$$\frac{f_n}{n} \xrightarrow{n \rightarrow \infty} 0 \text{ a.s.} \quad (26)$$

**Proof.** Without loss of generality, we may suppose that  $f_0 = 0$ . If defining the family of random variables  $(h_n)_n$  as following:  $h_0 = 0$ , for all  $n \geq 1, h_n = \sum_{k=1}^n \frac{f_k - f_{k-1}}{k}$ . We obtain

$$E[h_{n+1} - h_n / \mathcal{F}_n] = E\left[\frac{f_{n+1} - f_n}{n+1} / \mathcal{F}_n\right] = \frac{1}{n+1} E[(f_{n+1} - f_n) / \mathcal{F}_n] = 0, \text{ implying that}$$

$(h_n)_{n \geq 0}$  is a martingale. Observe that  $\frac{f_n}{n} = \frac{1}{n} \sum_{k=1}^n k(h_k - h_{k-1})$  and, due to Lemma 1, in order to

show that  $\frac{f_n}{n} \xrightarrow{n \rightarrow \infty} 0$  a.s. is enough to prove that  $\sum_n \frac{n(h_n - h_{n-1})}{n} = \sum_n (h_n - h_{n-1})$  is a.s.

convergent, meaning that  $h_n$  is a.s. convergent. According to Theorem 3 it is enough to prove

that  $\sup_{n \geq 0} E(h_n^2) < \infty$ . Indeed, as  $E[h_n^2] = \sum_{k=1}^n \frac{E[|f_k - f_{k-1}|^2]}{k^2}$ , then using (25), we obtain:

$\sup_{n \geq 0} E(h_n^2) < \sum_{k=1}^{\infty} \frac{E[|f_k - f_{k-1}|^2]}{k^2} < \infty$  and the proof is complete.

**Conclusion.** For  $(f_n)_{n \geq 0}$  a  $\mathcal{F}_n$ -discrete martingale, under the condition of  $\sup_{n \geq 0} E(f_n^2) < \infty$  there exists a random variable  $f$  such that for all  $n \geq 0$ ,  $f_n = E[f / \mathcal{F}_n]$  and, moreover,  $(f_n)_{n \geq 0}$  converges in  $L^2$  and a.s. to  $f$ . The proof of the law of the large numbers presented before, is based on this result.

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## Teorema de convergență pentru martingale cu timp discret

### Rezumat

Articolul prezintă câteva rezultate importante ale teoriei martingalelor cu timp. În prima parte este demonstrată o variantă a Teoremei de Convergență a martingalelor cu timp discret pe baza inegalității lui Doob pentru martingale. A doua parte a acestui articol prezintă o demonstrație a Legii Numerelor Mari, ca o consecință a Teoremei de Convergență.