

Lecture: Prof. Dr. Peter Pfaffelhuber

Assistance: Samuel Adeosun

https://pfaffelh.github.io/hp/2024ws_stochproc.html

<https://www.stochastik.uni-freiburg.de/>

Tutorial 2 - Definition and existence of stochastic processes

Exercise 1 (1+2+1=4 Points).

Let $\Omega = \{1,2,3,4,5\}$.

- (a) Find the smallest σ - algebra \mathcal{F}_1 containing

$$\mathcal{F}_2 := \{\{1,2,3\},\{3,4,5\}\}.$$

- (b) Is the random variable $X : \Omega \rightarrow \mathbb{R}$ defined by

$$X(1) = X(2) = 0, \quad X(3) = 10, \quad X(4) = X(5) = 1$$

measurable with respect to \mathcal{F}_1 ?

- (c) Find the σ -algebra \mathcal{F}_3 generated by $Y : \Omega \rightarrow \mathbb{R}$ and defined by

$$Y(1) = 0, \quad Y(2) = Y(3) = Y(4) = Y(5) = 1.$$

Solution.

- (a) $\mathcal{F}_1 = \{\emptyset, \Omega, \{1,2,3\}, \{3,4,5\}, \{3\}, \{1,2,4,5\}, \{1,2\}, \{4,5\}\}$.

- (b) The random variable X is measurable with respect to \mathcal{F}_1 since we have for each $A \in \mathcal{B}(\mathbb{R})$:

$$\begin{aligned} \text{if } 0 \in A, 1,10 \notin A : & \quad X^{-1}(A) = \{1,2\} \in \mathcal{F}_1, \\ \text{if } 1 \in A, 0,10 \notin A : & \quad X^{-1}(A) = \{4,5\} \in \mathcal{F}_1, \\ \text{if } 10 \in A, 1,10 \notin A : & \quad X^{-1}(A) = \{3\} \in \mathcal{F}_1, \\ \text{if } 0,1,10 \notin A : & \quad X^{-1}(A) = \emptyset \in \mathcal{F}_1, \\ \text{if } 0,1,10 \in A : & \quad X^{-1}(A) = \Omega \in \mathcal{F}_1. \end{aligned}$$

where $X^{-1}(A) = \{\omega \in \Omega : X(\omega) \in A\}$. We can reduce every other case to these, take for example, if $0,1 \in A$ but $10 \notin A$, then:

$$X^{-1}(A) = X^{-1}(\{0\}) \cup X^{-1}(\{1\}) = \{1,2\} \cup \{4,5\} = \{1,2,3,4,5\} \in \mathcal{F}_1.$$

- (c) $\mathcal{F}_3 = \sigma(Y) = \{\Omega, \emptyset, \{1\}, \{2,3,4,5\}\}$.

Exercise 2 (2+2 points).

- (a) Given an example of two stochastic processes \mathcal{X} and \mathcal{Y} which are versions of each other, but no modifications of each other.
- (b) Give an example of a real-valued stochastic process \mathcal{X} , such that $\mathbf{V}[X_t] > 0$ for all t and $\mathcal{X} = (X_t)_{t \in I}$ and $\mathcal{X}^2 := (X_t^2)_{t \in I}$ are indistinguishable.

Solution.

- (a) Let $\Omega = \{1,2,3\}$ and $\mathcal{F} = \{\{1,2,3\}, \emptyset, \{1,2\}, \{3\}\}$. Define two stochastic processes $\mathcal{X} = (X_t)_{t=1,2,3,\dots}$ and $\mathcal{Y} = (Y_t)_{t=1,2,3,\dots}$ as follows:

$$X_t(w) = 1 \quad \text{for all } t \text{ and } w \in \Omega,$$

$$Y_t(w) = 2 \quad \text{for all } t \text{ and } w \in \Omega.$$

We need to show that X_t and Y_t are versions of each other, which means we need to verify that their distributions are the same for all sets $A \in \mathcal{F}$. Observe the following for each A in \mathcal{F} :

$$\mathbf{P}(X_t \in \{1,2,3\}) = 1 = \mathbf{P}(Y_t \in \{1,2,3\}),$$

$$\mathbf{P}(X_t \in \{1,2\}) = 1 = \mathbf{P}(Y_t \in \{1,2\}),$$

$$\mathbf{P}(X_t \in \{3\}) = 0 = \mathbf{P}(Y_t \in \{3\}),$$

$$\mathbf{P}(X_t \in \emptyset) = 0 = \mathbf{P}(Y_t \in \emptyset).$$

Hence for all $A \in \mathcal{F}$:

$$\mathbf{P}(X_t \in A) = \mathbf{P}(Y_t \in A).$$

Thus, X_t and Y_t have the same distribution with respect to the sigma-algebra \mathcal{F} . Therefore, we conclude that X_t and Y_t are versions of each other. However, since $X_t(w) = 1$ and $Y_t(w) = 2$ for all t and for all $w \in \Omega$, it follows that $X_t(w) \neq Y_t(w)$ for all $w \in \Omega$. Thus,

$$\mathbf{P}(X_t = Y_t) = \mathbf{P}(\{\omega \in \Omega : X_t(\omega) = Y_t(\omega)\}) = \mathbf{P}(\emptyset) = 0 \neq 1 \text{ for all } t.$$

So, \mathcal{X} is not a modification of \mathcal{Y} , and vice versa.

- (b) Let $\mathcal{X} = (X_t)_{t \in I}$ be defined as:

$$X_t = \begin{cases} 1, & \text{with probability } \frac{1}{2}, \\ 0, & \text{with probability } \frac{1}{2}. \end{cases}$$

Clearly,

$$\mathbf{E}[X_t] = \frac{1}{2}, \quad \mathbf{E}[X_t^2] = \frac{1}{2} \implies \mathbf{V}[X_t] = \frac{1}{4} > 0.$$

Furthermore,

$$X_t^2 = \begin{cases} 1, & \text{with probability } \frac{1}{2}, \\ 0, & \text{with probability } \frac{1}{2}. \end{cases}$$

Thus, $\mathbf{P}(X_t = X_t^2 \text{ for all } t \in I) = 1$. Hence, \mathcal{X} and \mathcal{X}^2 are indistinguishable with $\mathbf{V}[X_t] > 0$ for all t .

Exercise 3 (4 Points).

Let I be some index set, $(E, \mathcal{B}(E))$ be Polish and $(\mathbf{P}_i)_{i \in I}$ a family of probability measures on $\mathcal{B}(E)$. Show that there exists an E -valued stochastic process $(X_t)_{t \in I}$ such that $(X_{t_1}, \dots, X_{t_n}) \sim \otimes_{i=1}^n \mathbf{P}_{t_i}$ for any $t_1, \dots, t_n \in I$. In other words, $(X_t)_{t \in I}$ is an independent family with $X_t \sim \mathbf{P}_t$.

Solution.

Recall that if (Ω, \mathcal{F}) is a measurable space, I an arbitrary index set and $(\Omega^J, \mathcal{F}^J)_{J \subseteq_f I}$ is a family of measurable product spaces, equipped with the product σ -algebra, as in Definition 5.3. A family of probability measures $(\mathbf{P}_J)_{J \subseteq_f I}$, where \mathbf{P}_J is a probability measure on \mathcal{F}^J , is called a projective family if $\mathbf{P}_H = (\pi_H^J)_* \mathbf{P}_J$ for all $H \subseteq J \subseteq_f I$. Also, if for a projective family $(\mathbf{P}_J)_{J \subseteq_f I}$ of probability measures there exists a probability measure \mathbf{P}_I on \mathcal{F}^I with $\mathbf{P}_J = (\pi_J)_* \mathbf{P}_I$ for all $J \subseteq_f I$, then \mathbf{P}_I is called the projective limit of the projective family and we write $\mathbf{P}_I = \varprojlim_{J \subseteq_f I} \mathbf{P}_J$. In the above, we have the following probability spaces: $(E, \mathcal{B}(E), \mathbf{P}_i)_{i \in I}$. Suppose $X_i : E \rightarrow E$ represent the random variables (identity maps!) which are $\mathcal{B}(E)/\mathcal{B}(E)$ measurable. As in Example 5.22 and Remark 5.23, define: $\mathbf{P}^{\otimes J} := \otimes_{j \in J} (X_j * \mathbf{P}_j)$, $J \subseteq_f I$. We claim that the family $(\mathbf{P}^{\otimes J})_{J \subseteq_f I}$ is projective. If $H \subseteq J \subseteq_f I$, then for $A_j \in \mathcal{B}(E), j \in H$,

$$\begin{aligned} (\pi_H^J)_* \mathbf{P}^{\otimes J} \left(\prod_{i \in H} A_j \right) &= \mathbf{P}^{\otimes J} \left((\pi_H^J)^{-1} \left(\prod_{j \in H} A_j \right) \right) \\ &= \mathbf{P}^{\otimes J} \left(\prod_{j \in H} A_j \times \prod_{j \in J \setminus H} E \right) \\ &= \prod_{j \in J} \mathbf{P}_j(X_j \in A_j) \cdot \prod_{j \in J \setminus H} \mathbf{P}_j(X_j \in E) \\ &= \prod_{j \in H} \mathbf{P}_j(X_j \in A_j) \\ &= \mathbf{P}^{\otimes H} \left(\prod_{j \in H} A_j \right). \end{aligned}$$

Thus, we claim that the projective limit exists! (See Theorem 5.24 [Existence of processes, Kolmogorov])

$$\mathbf{P}^{\otimes I} = \otimes_{i \in I} (X_i * \mathbf{P}_i) = \varprojlim_{J \subseteq_f I} \otimes_{j \in J} (X_j * \mathbf{P}_j) = \varprojlim_{J \subseteq_f I} \mathbf{P}^{\otimes J}.$$

Thus, for any $j = \{t_1, t_2, \dots, t_n\} \subseteq I$ and for all $A := (A_j)_{j \in J} \in \mathcal{B}(E)$,

$$\begin{aligned} \mathbf{P}^{\otimes J}(A) &= \otimes_{j \in J} (X_j * \mathbf{P}_j(A_j)) = \otimes_{j \in J} \left(\mathbf{P}_j(X_j^{-1}(A_j)) \right) \\ &= \otimes_{j \in J} (\mathbf{P}_j(A_j)) = \otimes_{j \in J} \mathbf{P}_j(A). \end{aligned}$$

That is, such E -valued stochastic process $\mathcal{X} = (X_t)_{t \in I}$ exists!

Exercise 4 (1+1+2= 4 Points).

Let $(X_k)_{k \in \mathbb{N}_0}$ be a **symmetric** simple random walk, i.e. $X_k = \sum_{i=0}^{k-1} Z_i$ where Z_1, Z_2, \dots are iid with $\mathbf{P}(Z_1 = \pm 1) = \frac{1}{2}$ ¹. For $n \in \mathbb{N}$ define

$$S_n = \sum_{k=1}^n X_k.$$

¹Recall the convention that $\sum_{i=0}^{-1} a_i = 0$

- (a) Show whether or not $(S_n)_{n \in \mathbb{N}_0}$ is a simple random walk (not necessarily symmetric).
- (b) Compute the covariance $\mathbf{COV}[X_k, X_l]$ for $k \leq l \in \mathbb{N}$.
- (c) Compute the variance of S_n for $n \in \mathbb{N}$.

Note: You may need to recall that

$$\sum_{k=1}^n k = \frac{n(n+1)}{2} \quad \text{and} \quad \sum_{k=1}^n k^2 = \frac{n(n+1)(2n+1)}{6}, \quad \text{for } n \in \mathbb{N}.$$

Solution.

- (a) We start by writing

$$S_n = \sum_{k=1}^n \sum_{i=0}^{k-1} Z_i = \sum_{i=0}^{n-1} \sum_{k=i+1}^n Z_i = \sum_{i=0}^{n-1} (n-i)Z_i$$

This expression shows that S_n is a linear combination of the Z_i values, where each Z_i is multiplied by the number of times it contributes to the sum S_n . Specifically, Z_i appears in S_n a total of $n - i$ times. To verify whether S_n constitutes a simple random walk, we analyze its increment:

$$S_{n+1} - S_n = X_{n+1}.$$

Since $X_{n+1} = Z_0 + Z_1 + \dots + Z_n$, we can express:

$$S_{n+1} = S_n + X_{n+1} = S_n + (Z_0 + Z_1 + \dots + Z_n).$$

Each increment $S_{n+1} - S_n$ depends on the Z values, but it does not result in independent increments because the contribution of each Z_i to S_n depends on their indices. Thus, although S_n is a sum of random variables, it is not a simple random walk because the increments are not independent and identically distributed. Therefore, we conclude that:

The process $(S_n)_{n \in \mathbb{N}_0}$ is not a simple random walk.

As illustration: Consider for $k = 2$:

$$X_2 = Z_0 + Z_1.$$

The possible values of X_2 are:

$$\begin{cases} 2 & \text{if } Z_0 = 1, Z_1 = 1, \\ 0 & \text{if } Z_0 = 1, Z_1 = -1 \quad \text{or} \quad Z_0 = -1, Z_1 = 1, \\ -2 & \text{if } Z_0 = -1, Z_1 = -1. \end{cases}$$

The increment $S_{n+1} - S_n = X_{n+1}$ is not independent of previous Z_i . The distribution of X_k changes with k (e.g., X_2 can take values $-2, 0, 2$). Note that, $\mathbf{P}(X_2 = \pm 1) = 0$.

(b) We compute as follows:

$$\begin{aligned}\mathbf{COV}[X_k, X_l] &= \mathbf{E}[(X_k - \mathbf{E}[X_k])(X_l - \mathbf{E}[X_l])] \\ &= \mathbf{E}[X_k X_l] = \mathbf{E} \left[\left(\sum_{i=0}^{k-1} Z_i \right) \left(\sum_{j=0}^{l-1} Z_j \right) \right]\end{aligned}$$

When we expand the sum above, we observe that the terms of the form $\mathbf{E}[Z_i Z_j]$, for $i \neq j$ will vanish because Z_i and Z_j are independent for $i \neq j$. The only terms left are the those of the form $\mathbf{E}[Z_i Z_i]$ or $\mathbf{E}[Z_j Z_j]$. Their value is 1, and since there are k of them ($k \leq l$), then

$$\mathbf{COV}[X_k, X_l] = k.$$

(c) We already know that $\mathbf{E}[S_n] = 0$. So, it suffices to compute

$$\mathbf{V}[S_n] = \mathbf{E}[S_n^2] = \mathbf{E} \left[\left(\sum_{k=1}^n X_k \right) \left(\sum_{l=1}^n X_l \right) \right]$$

Recall the identity of Bienamyé (see Proposition 6.9!)

$$\begin{aligned}\mathbf{V} \left[\sum_{k=1}^n X_k \right] &= \mathbf{E} \left[\left(\sum_{k=1}^n X_k \right)^2 \right] = \sum_{k=1}^n \sum_{l=1}^n \mathbf{E}[X_k X_l] = \sum_{k=1}^n \mathbf{E}[X_k^2] + 2 \sum_{1 \leq k < l \leq n} \mathbf{E}[X_k X_l] \\ &= \sum_{k=1}^n \mathbf{V}[X_k] + 2 \sum_{1 \leq k < l \leq n} \mathbf{COV}[X_k X_l].\end{aligned}$$

For each $k = 1, 2, \dots, n$, there are exactly $n - k$ values of l such that $k < l \leq n$. Therefore, $\sum_{1 \leq k < l \leq n} k = \sum_{k=1}^n k(n - k)$, and so

$$\begin{aligned}\mathbf{V}[S_n] &= \sum_{k=1}^n k + 2 \sum_{k=1}^n k(n - k) = (2n + 1) \sum_{k=1}^n k - 2 \sum_{k=1}^n k^2 \\ &= (2n + 1) \frac{n(n + 1)}{2} - 2 \frac{n(n + 1)(2n + 1)}{6} = \frac{n(n + 1)(2n + 1)}{6}.\end{aligned}$$

Alternatively: We can use the representation $X_k = \sum_{i=1}^k Y_i$ so that

$$\begin{aligned}\mathbf{V}[S_n] &= \mathbf{E} \left[\left(\sum_{k=1}^n X_k \right)^2 \right] = \mathbf{E} \left[\left(\sum_{k=1}^n \sum_{i=1}^k Y_i \right)^2 \right] \\ &= \mathbf{E} \left[\left(\sum_{i=1}^n \sum_{k=1}^n Y_i \right)^2 \right] = \mathbf{E} \left[\left(\sum_{i=1}^n (n - i + 1) Y_i \right)^2 \right] \\ &= \sum_{i=1}^n (n - i + 1)^2 = \sum_{i=1}^n i^2 = \frac{n(n + 1)(2n + 1)}{6}\end{aligned}$$

where we have used the fact that $\mathbf{E}[Y_i Y_j] = 0$ for $i \neq j$.