

# Note: Conditional Expectation and Probability

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## 1 Conditional expectation regarding a sub-algebra

Let  $X : \Omega \rightarrow \mathbb{R}$  be an integrable random variable on a probability space  $(\Omega, \mathcal{F}, P)$ . Let  $\mathcal{G} \subset \mathcal{F}$  be a sub- $\sigma$ -algebra.

**Definition 1.** We refer to a random variable  $Y$  as a conditional expectation of  $X$  under  $\mathcal{G}$  when the following are satisfied:

- $Y$  is  $\mathcal{G}$ -measurable.
- $Y$  is integrable and satisfies for any  $A \in \mathcal{G}$ ,

$$\mathbb{E}[X; A] = \int_A X(\omega)P(d\omega) = \int_A Y(\omega)(d\omega) = \mathbb{E}[Y; A].$$

We denote by  $\mathbb{E}[X|\mathcal{G}]$  a conditional expectation defined above. The existence and almost sure uniqueness can be confirmed by Radon–Nikodým theorem. Let us define the finite measure on  $(\Omega, \mathcal{G})$  as follows:

$$\mu(A) = \int_A X(\omega)P(d\omega) \quad (A \in \mathcal{G}).$$

Then,  $\mu$  is absolutely continuous with respect to  $P|_{\mathcal{G}}$ . That is, if  $P|_{\mathcal{G}}(A) = 0$  for  $A \in \mathcal{G}$ , then  $\mu(A) = 0$ . By applying Radon–Nikodým theorem, there exists a  $\mathcal{G}$ -measurable and  $P$ -integrable function  $Y$  that satisfies  $\int_A X(\omega)P(d\omega) = \int_A Y(\omega)(d\omega)$ . Moreover,  $Y$  which satisfies these conditions is unique  $P$ -almost surely.

**Example 1.** Let  $\{\Omega_j\}_{j \in \mathbb{N}}$  be a partition of  $\Omega$ . That is,  $\Omega = \cup_j \Omega_j$ ,  $\Omega_j \cap \Omega_k = \emptyset$  ( $j \neq k$ ),  $\Omega_j \in \mathcal{F}$ , and  $\mathbb{P}[\Omega_j] > 0$  ( $j \in \mathbb{N}$ ). Let  $\mathcal{G} = \sigma(\{\Omega_j\}_{j \in \mathbb{N}})$ . Then, for an integrable random variable  $X$ ,

$$\mathbb{E}[X|\mathcal{G}] = \sum_{j \in \mathbb{N}} \frac{\mathbb{E}[X1_{\Omega_j}]}{\mathbb{P}[\Omega_j]} 1_{\Omega_j} \quad a.s.$$

*Especially, for  $A \in \mathcal{F}$  and  $X = 1_A$ ,*

$$\mathbb{E}[1_A|\mathcal{G}] = \sum_{j \in \mathbb{N}} \frac{\mathbb{P}[A \cap \Omega_j]}{\mathbb{P}[\Omega_j]} \quad \text{on } \Omega_j \text{ a.s.}$$

## 2 Conditional expectation regarding a measurable

Let  $(\Omega, \mathcal{F}, P)$  be a probability space,  $X : \Omega \rightarrow \mathbb{R}$  be an integrable  $\mathcal{F}$ -measurable function,  $(\mathcal{T}, \mathcal{B})$  be a measurable space, and  $T : \Omega \rightarrow \mathcal{T}$  be a  $\mathcal{F}/\mathcal{B}$ -measurable map. Let  $P^T$  denote the push-forward measure  $T_{\#}P$ .

**Definition 2.** We refer to a function  $g : \mathcal{T} \rightarrow \mathbb{R}$  as a conditional expectation of  $X$  under  $T = t$  when the following are satisfied:

- $g$  is  $\mathcal{B}$ -measurable.
- $g$  is  $P^T$ -integrable and satisfies for any  $B \in \mathcal{B}$ ,

$$\int_{T^{-1}(B)} X(\omega) P(d\omega) = \int_B g(t) P^T(dt).$$

We denote by  $\mathbb{E}[X|T = t]$  a conditional expectation  $g$ . The existence and uniqueness up to 0-measurable set regarding  $P^T$  can be proven in the same way as in the previous section. That is, we define the finite measure on  $(\mathcal{T}, \mathcal{B})$  by

$$\mu(B) = \int_{T^{-1}(B)} X(\omega) P(d\omega) \quad (B \in \mathcal{B}).$$

Then,  $\mu$  is absolutely continuous in  $P^T$ . By Radon–Nikodým derivative  $d\mu/dP^T(t)$  satisfies the condition.

**Example 2.** Let  $Y : \Omega \rightarrow \mathbb{R}^{d_1}$  and  $T : \Omega \rightarrow \mathbb{R}^{d_2}$  be a measurable maps defined on the probability space  $(\Omega, \mathcal{F}, P)$ . Suppose the joint distribution  $P^{(Y,T)}$  of  $(Y, T)$  has a probability density function  $p^{(Y,T)}(y, t)$ . We denote by  $p^T(t)$  the density of the marginal distribution in  $T$ . We set

$$p(y|t) = \begin{cases} \frac{p^{(Y,T)}(y,t)}{p^T(t)} & (p^T(t) \neq 0), \\ 0 & (p^T(t) = 0). \end{cases}$$

Suppose for a measurable function  $h : \mathbb{R}^{d_1} \times \mathbb{R}^{d_2} \rightarrow \mathbb{R}$ ,  $h(Y, T) : \Omega \rightarrow \mathbb{R}$  is integrable. Then, it follows that for  $B \in \mathcal{B}(\mathbb{R}^{d_2})$ ,

$$\begin{aligned} \int_{T^{-1}(B)} h(Y, T)(\omega) P(d\omega) &= \int_{\mathbb{R}^{d_1} \times B} h(y, t) P^{(Y,T)}(dy, dt) \\ &= \int_{\mathbb{R}^{d_1} \times B} h(y, t) p^{(Y,T)}(y, t) dy dt \\ &= \int_B \left\{ \int_{\mathbb{R}^{d_1}} h(y, t) p(y|t) dx \right\} p^T(t) dt. \end{aligned}$$

This means

$$\mathbb{E}[h(Y, T)|T = t] = \int_{\mathbb{R}^{d_1}} h(y, t) p(y|t) dx \quad P^T\text{-a.s.}$$

The next proposition shows that the conditional expectations under a map  $T$  and  $\sigma(T)$  are equivalent. Here,  $\sigma(T)$  is the minimum  $\sigma$ -algebra so that  $T$  is measurable, which is  $\sigma(T) = \{T^{-1}(B) | B \in \mathcal{B}\}$ .

**Proposition 1.** *Let  $(\Omega, \mathcal{F}, P)$  be a probability space,  $(\mathcal{T}, \mathcal{B})$  be a measurable space,  $T : \Omega \rightarrow \mathcal{T}$  be a random variable, and  $X : \Omega \rightarrow \mathbb{R}$  be a  $\mathcal{F}$ -measurable integrable random variable. Then,*

$$\mathbb{E}[X|\sigma(T)](\omega) = \mathbb{E}[X|T](\omega) \quad P\text{-a.s. } \omega,$$

where we define  $\mathbb{E}[X|T](\omega) = \mathbb{E}[X|T = t]|_{t=T(\omega)}$ .

*Proof.* For any  $A \in \sigma(T)$ , there exists  $B \in \mathcal{B}$  such that  $A = T^{-1}(B)$ . Then,

$$\begin{aligned} \int_A \mathbb{E}[X|\sigma(T)](\omega) P(d\omega) &= \int_A X(\omega) P(d\omega) \\ &= \int_{T^{-1}(B)} X(\omega) P(d\omega) \\ &= \int_B \mathbb{E}[X|T = t] P^T(dt) \\ &= \int_{T^{-1}(B)} \mathbb{E}[X|T = t]|_{t=T(\omega)} P(d\omega) \\ &= \int_A \mathbb{E}[X|T = t]|_{t=T(\omega)} P(d\omega). \end{aligned}$$

Therefore, we can conclude  $\mathbb{E}[X|\sigma(T)] = \mathbb{E}[X|T = t]$   $P$ -a.s. because of the uniqueness of the conditional expectation (i.e., Radon–Nikodým derivative).  $\square$

Hence,  $\mathbb{E}[X|\sigma(T)]$  is basically a function induced by the composition of functions  $T : \Omega \rightarrow \mathcal{T}$  and  $\mathbb{E}[X|T = \cdot] : \mathcal{T} \rightarrow \mathbb{R}$ .

### 3 Conditional Probability

**Definition 3.** *Let  $(\Omega, \mathcal{F}, P)$  be a probability space.*

1. *Let  $\mathcal{G} \subset \mathcal{F}$  be a sub  $\sigma$ -algebra. The conditional probability of an event  $A \in \mathcal{F}$  under  $\mathcal{G}$  is defined by  $\mathbb{P}[A|\mathcal{G}] = \mathbb{E}[1_A|\mathcal{G}]$ .*
2. *Let  $(\mathcal{T}, \mathcal{B})$  be a measurable space and  $T : \Omega \rightarrow \mathcal{T}$  be a  $\mathcal{F}/\mathcal{B}$ -random variable. The conditional probability of an event  $A \in \mathcal{F}$  under the condition  $T = t$  is defined by  $\mathbb{P}[A|T = t] = \mathbb{E}[1_A|T = t]$ .*

For each case, we see the following relationship. (i) For any  $B \in \mathcal{G}$ , it follows that

$$\mathbb{P}[A \cap B] = \mathbb{E}[1_A; B] = \mathbb{E}[\mathbb{E}[1_A|\mathcal{G}]; B] = \mathbb{E}[\mathbb{P}[A|\mathcal{G}]; B].$$

An example under  $\sigma(\{\Omega_j\}_{j \in \mathbb{N}})$  where  $\{\Omega_j\}_{j \in \mathbb{N}}$  is a partition of  $\Omega$  is given in Example 1. (ii) For any  $B \in \mathcal{B}$ , it follows that

$$\mathbb{P}[A \cap \{T \in B\}] = \mathbb{P}[A \cap T^{-1}(B)] = \int_{T^{-1}(B)} 1_A(\omega) P(d\omega) = \int_B \mathbb{P}[A|T = t] P^T(dt).$$

For  $\mathcal{G} = \sigma(T)$ , it follows that by Proposition 1 for any  $A \in \mathcal{F}$ ,

$$\mathbb{P}[A|\sigma(T)](\omega) = \mathbb{P}[A|T = t]|_{t=T(\omega)} \quad P\text{-a.s. } \omega,$$

**Example 3.** Let  $Y : \Omega \rightarrow \mathbb{R}^{d_1}$  and  $T : \Omega \rightarrow \mathbb{R}^{d_2}$  be measurable maps defined on the probability space  $(\Omega, \mathcal{F}, P)$ . Suppose  $P^{(Y,T)}$  has a probability density function  $p^{(Y,T)}(y, t)$ . Then for any  $A \in \mathcal{B}(\mathbb{R}^{d_1})$  and  $B \in \mathcal{B}(\mathbb{R}^{d_2})$ , it follows that

$$\begin{aligned}
\mathbb{P}[Y \in A, T \in B] &= \mathbb{P}[Y^{-1}(A) \cap T^{-1}(B)] \\
&= \int_{T^{-1}(B)} 1_{Y(\omega) \in A} P(d\omega) \\
&= \int_{\mathbb{R}^{d_1} \times B} 1_A(y) P^{(Y,T)}(dy, dt) \\
&= \int_{\mathbb{R}^{d_1} \times B} 1_A(y) p^{(Y,T)}(y, t) dy dt \\
&= \int_{A \times B} p^{(Y,T)}(y, t) dy dt \\
&= \int_B \left\{ \int_A p(y|t) dy \right\} p^T(t) dt.
\end{aligned}$$

Therefore, we see  $\mathbb{P}[Y^{-1}(A)|T = t] = \mathbb{P}[Y \in A|T = t] = \int_A p(y|t) dy$ . In other words,  $p(y|t)$  is a density function of  $\mathbb{P}[Y \in \cdot | T = t]$ .

**Proposition 2.** Let  $(\Omega, \mathcal{F}, P)$  be a probability space and  $X : \Omega \rightarrow \mathbb{R}$  an integrable function. Let  $\mathcal{G} \subset \mathcal{F}$  be a sub  $\sigma$ -algebra. Then,  $\mathbb{E}[X|\mathcal{G}] = \int X(\omega) \mathbb{P}[d\omega|\mathcal{G}]$  a.s.

*Proof.* Let  $\{X_n\}_{n \in \mathbb{N}}$  be a sequence of the step-function  $X_n = \sum_{i=1}^{m_n} a_{n,i} 1_{A_{n,i}}$  ( $A_{n,i} \in \mathcal{F}$ ) such that for any  $\omega \in \Omega$ ,  $X_n(\omega) \rightarrow X(\omega)$ . Then, for any  $A \in \mathcal{G}$ ,

$$\begin{aligned}
\mathbb{E}[X; A] &= \lim_{n \rightarrow \infty} \mathbb{E}[X_n; A] \\
&= \lim_{n \rightarrow \infty} \sum_{i=1}^{m_n} a_{n,i} \mathbb{E}[1_{A_{n,i}}; A] \\
&= \lim_{n \rightarrow \infty} \sum_{i=1}^{m_n} a_{n,i} \mathbb{E}[\mathbb{P}[A_{n,i}|\mathcal{G}]; A] \\
&= \mathbb{E} \left[ \lim_{n \rightarrow \infty} \sum_{i=1}^{m_n} a_{n,i} \mathbb{P}[A_{n,i}|\mathcal{G}]; A \right] \\
&= \mathbb{E} \left[ \int X(\omega) \mathbb{P}[d\omega|\mathcal{G}]; A \right].
\end{aligned}$$

Therefore,  $\mathbb{E}[X|\mathcal{G}] = \int X(\omega) \mathbb{P}[d\omega|\mathcal{G}]$  a.s. □

Applying this proposition for the composition  $X = f(Y)$  of a measurable map  $Y : \Omega \rightarrow \mathbb{R}^{d_1}$  and measurable function  $f : \mathbb{R}^{d_1} \rightarrow \mathbb{R}$ , we have

$$\begin{aligned}
\mathbb{E}[f(Y)|\mathcal{G}] &= \int f(Y(\omega)) \mathbb{P}[d\omega|\mathcal{G}] \\
&= \int f(y) \mathbb{P}[Y \in dy|\mathcal{G}] \quad \text{a.s.},
\end{aligned}$$

where  $\int \cdot \mathbb{P}[Y \in dy|\mathcal{G}]$  represents an integral by the push-forward measure  $Y_{\#} \mathbb{P}[\cdot|\mathcal{G}]$ .

In the same way, we can show the following. For a measurable map  $T : \Omega \rightarrow \mathbb{R}^{d_2}$ , we have

$$\mathbb{E}[f(Y)|T = t] = \int f(y)\mathbb{P}[Y \in dy|T = t].$$

Especially, when  $P^{(Y,T)}$  has the density function  $p^{(Y,T)}(y, t)$  as in Example 3,

$$\mathbb{E}[f(Y)|T = t] = \int f(y)p(y|t)dy.$$