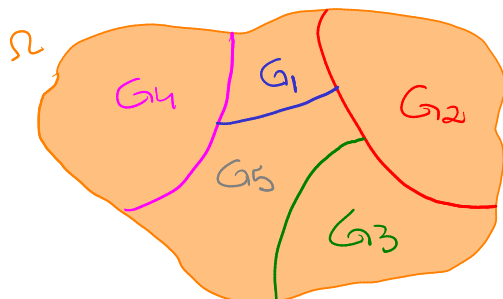


- 3.17.** Let  $(\Omega, \mathcal{F}, P)$  be a probability space and let  $X: \Omega \rightarrow \mathbf{R}$  be a random variable with  $E[|X|] < \infty$ . If  $\mathcal{G} \subset \mathcal{F}$  is a *finite*  $\sigma$ -algebra, then by Exercise 2.7 there exists a partition  $\Omega = \bigcup_{i=1}^n G_i$  such that  $\mathcal{G}$  consists of  $\emptyset$  and unions of some (or all) of  $G_1, \dots, G_n$ .
- a) Explain why  $E[X|\mathcal{G}](\omega)$  is constant on each  $G_i$ . (See Exercise 2.7 c).)



**2.7.** a) Suppose  $G_1, G_2, \dots, G_n$  are disjoint subsets of  $\Omega$  such that

$$\Omega = \bigcup_{i=1}^n G_i.$$

Prove that the family  $\mathcal{G}$  consisting of  $\emptyset$  and all unions of some (or all) of  $G_1, \dots, G_n$  constitutes a  $\sigma$ -algebra on  $\Omega$ .

b) Prove that any *finite*  $\sigma$ -algebra  $\mathcal{F}$  on  $\Omega$  is of the type described in a).

c) Let  $\mathcal{F}$  be a *finite*  $\sigma$ -algebra on  $\Omega$  and let  $X: \Omega \rightarrow \mathbf{R}$  be  $\mathcal{F}$ -measurable. Prove that  $X$  assumes only finitely many possible values. More precisely, there exists a disjoint family of subsets  $F_1, \dots, F_m \in \mathcal{F}$  and real numbers  $c_1, \dots, c_m$  such that

$$X(\omega) = \sum_{i=1}^m c_i \chi_{F_i}(\omega).$$

By the exercise 2.7, for  $c_i \in \mathbf{R}$ ,  $i=1, \dots, n$ ,

$$E[X|\mathcal{G}](\omega) = E\left[\underbrace{\sum_{i=1}^n c_i \chi_{G_i}(\omega)}_{\text{constant } C}\right] = E[C] = C$$

b) Assume that  $P[G_i] > 0$ . Show that

$$E[X|\mathcal{G}](\omega) = \frac{\int_{G_i} X dP}{P(G_i)} \quad \text{for } \omega \in G_i.$$

$$\int_H E[X|\mathcal{H}] dP = \int_H X dP, \text{ for all } H \in \mathcal{H}.$$

gives

$$\int_{G_i} E[X|\mathcal{G}] dP = \int_{G_i} X dP$$

however,

$$\int_{G_i} E[X|\mathcal{G}] dP = E[X|\mathcal{G}] \int_{G_i} dP = E[X|\mathcal{G}] P(G_i)$$

therefore,

$$E[X|\mathcal{G}] = \frac{\int_{G_i} X dP}{P(G_i)}$$

- c) Suppose  $X$  assumes only finitely many values  $a_1, \dots, a_m$ . Then from elementary probability theory we know that (see Exercise 2.1)

$$E[X|G_i] = \sum_{k=1}^m a_k P[X = a_k | G_i] .$$

Compare with b) and verify that

$$E[X|G_i] = E[X|\mathcal{G}](\omega) \quad \text{for } \omega \in G_i .$$

Thus we may regard the conditional expectation as defined in Appendix B as a (substantial) generalization of the conditional expectation in elementary probability theory.

Using the previous item,

$$\begin{aligned} E[X|\mathcal{G}] &= \frac{\int_{G_i} X dP}{P(G_i)} = \frac{\int_{G_i} \sum_{k=1}^m a_k \chi_{a_k} dP}{P(G_i)} = \sum_{k=1}^m \frac{\int_{G_i} a_k \chi_{a_k} dP}{P(G_i)} \\ &= \sum_{k=1}^m a_k \frac{P(G_i \cap G_k)}{P(G_i)} = \sum_{k=1}^m a_k P[X = a_k | G_i] \end{aligned}$$

Alternative answer

b) Since

$$\mathbb{E}[f|A] = \int f d\mu_A, \quad \mu_A(B) = \frac{\mu(B \cap A)}{\mu(A)}$$

We have

$$\mathbb{E}\left[\sum_{i=1}^n c_i \chi_{G_i}(\omega) \mid G_i\right] = \int_{G_i} \frac{X dP(\omega)}{P(G_i)}$$

c) Given the definition of  $a_k$ ,  $k=1, \dots, m$ ,

$$\mathbb{E}[X|G_i] = \sum_{k=1}^m a_k P[X=a_k|G_i] = \int X dP_{G_i}$$

By the fact that

$$\mu_A(B) = \frac{\mu(B \cap A)}{\mu(A)}$$

we obtain

$$\mathbb{E}[X|G_i] = \int X dP_{G_i} = \int_{G_i} \frac{X dP}{P(G_i)} = \mathbb{E}[X|g]$$