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# 1 Chapter 1

## 1.1 Section 1

**Exercise 1.1:** Let  $P$  be a probability measure on  $(\Omega, \mathcal{F})$

i) **Monotonicity:** If  $A \subset B$  then  $P(B) - P(A) = P(B - A) \geq 0$ .

Let  $C = B \cap A^c$ . Then  $A \cup C = B$ . Because  $A$  and  $C$  are disjoint, we have that  $P(B) = P(A \cup C) = P(A) + P(C)$ . Then  $P(B) - P(A) = P(C) = P(B \setminus A)$  which is greater or equal to zero since  $P(C) \geq 0$ . □

ii) **Subadditivity:** For  $\langle A_m \rangle \in \mathcal{F}$  and  $A \subset \cup^\infty A_m$  it follows that  $P(A) \leq \sum^\infty P(A_m)$

Since  $\langle A_m \rangle \in \mathcal{F}$  and  $\mathcal{F}$  is a sigma field, there exists  $\langle B_m \rangle$  such that  $B_i \cap B_j = \emptyset$  for  $i \neq j$  and  $\cup^\infty A_m = \cup^\infty B_m$  [2, pp. 17–18]. From (i) we know that  $A \subset \cup B_m \implies P(A) \leq P(\cup B_m)$ . Then

$$P(A) \leq P(\cup^\infty B_m) = \sum^\infty P(B_m) \leq \sum_*^\infty P(A_m)$$

The  $(*)$  inequality follows from how we define each element of  $\langle B_m \rangle$ :

$$\begin{aligned} B_i &= A_i \setminus [A_1 \cup \dots \cup A_{i-1}] \implies P(B_i) = P(A_i) - \sum_{j=1}^{i-1} P(A_j) \\ &\implies P(B_i) \leq P(A_i), \forall i \end{aligned}$$

□

iii) **Continuity from below:** If  $A_i \uparrow A$  then  $P(A_i) \uparrow P(A)$

By supposition, we know that for every  $i$  we have  $A_i = \cup_{j=1}^i A_j$

$$\lim_{i \rightarrow \infty} P(A_i) = \lim_{i \rightarrow \infty} P(\cup_{j=1}^i A_j) = P(A)$$

(I'm not sure if this is rigorous enough but ...)

□

iv) **Continuity from above:** If  $A_i \downarrow A$  then  $P(A_i) \downarrow P(A)$

For every  $i$  we have that  $A_i = \cap_{j=1}^i A_j$

$$\lim_{i \rightarrow \infty} P(A_j) = \lim_{i \rightarrow \infty} P(\cap^i A_j) = P(A)$$

□

### Exercise 1.2:

i) If  $\mathcal{F}_i$ ,  $i \in I$  are  $\sigma$ -fields then  $\cap_{i \in I} \mathcal{F}_i$  is.

Suppose  $\mathcal{F}_i$  are  $\sigma$ -fields.

- Consider  $A \in \cap_{i \in I} \mathcal{F}_i$ . Then  $\exists i$  such that  $A \in \mathcal{F}_i$ . So  $A^c \in \mathcal{F}_i$  because  $\mathcal{F}_i$  is a  $\sigma$ -field. Then  $A^c \in \cap_{i \in I} \mathcal{F}_i$ .
- Consider  $A_j \in \cap_{i \in I} \mathcal{F}_i$  a countable sequence of sets. Then  $\forall j, \forall i$ ,  $A_j \in \mathcal{F}_i$ . Then  $\cup A_j \in \mathcal{F}_i$ ,  $\forall i$  because  $\mathcal{F}_i$  is a  $\sigma$ -field. Then  $\cup A_j \in \cap_{i \in I} \mathcal{F}_i$

□

ii) Use the result in (i) to show if we are given a set  $\Omega$  and a collection  $\mathcal{A}$  of subsets of  $\Omega$  then there is a smallest  $\sigma$ -field containing  $\mathcal{A}$ .

Let  $\mathcal{A}$  be a collection of subsets of  $\Omega$ . let  $\mathcal{F}_{\mathcal{A}}$  be the set of sigma fields that contain  $\mathcal{A}$ . Define  $\mathcal{F} = \cap_{\mathcal{A}} \mathcal{F}_{\mathcal{A}}$ . From (i) we know that  $\mathcal{F}$  is a sigma field. By definition  $\mathcal{F}$  is the smallest sigma field containing  $\mathcal{A}$  since for sigma field  $\mathcal{C}$  such that  $\mathcal{A} \subset \mathcal{C}$ , we have  $\mathcal{F} = \cap_{\mathcal{A}} \mathcal{F}_{\mathcal{A}} \subset \mathcal{C}$

□

With  $(\mathbb{R}, \mathcal{F}, P)$  and  $\mathcal{B}$  the borel sets, define a random variable as a real valued function such that  $X : \Omega \rightarrow \mathbb{R}$  is  $\mathcal{F}$  measurable for every borel set

$$X^{-1}(B) \in \mathcal{F}, \quad B \in \mathcal{B}$$

Then  $X$  induces a probability measure on  $\mathbb{R}$  called its distribution

$$\mu(A) = P(X \in A) = P(X^{-1}(A)), \quad A \in \mathcal{B}$$

The **distribution function** is defined as

$$F(x) = P(X \leq x)$$

When the distribution function  $F(x) = P(X \leq x)$  has the form

$$F(x) = \int_{-\infty}^x f(y)dy$$

we say that  $X$  has **density function**  $f$ .

**Exercise 1.5:** A  $\sigma$ -field  $\mathcal{F}$  is said to be **countably generated** if there is a countable collection  $\mathcal{C} \subset \mathcal{F}$  so that  $\sigma(\mathcal{C})$ . Show that  $\mathcal{R}^d$  is countably generated.

$\mathcal{R}^d$  are the Borel subsets of  $\mathbb{R}^n$ . First let's look at  $\mathcal{R}$ . We'll show that  $\mathcal{G} = \{[q, \infty) : q \in \mathbb{Q}\}$  generates  $\mathcal{R}$ . Consider an arbitrary open interval  $(a, b) \subset \mathbb{R}$ ,  $a < b$ . See that

$$[b, \infty)^c \cap \bigcup_n [a + 1/n, \infty) = (-\infty, b) \cap (a, \infty) = (a, b) \quad (1)$$

Remembering that every open set of real numbers is the countable union of disjoint open intervals [2, p. 42], we observe that using 1 as a way to generate open intervals, we can also generate any open set. Therefore  $\mathcal{R} \subset \sigma(\mathcal{G})$ . To see that  $\sigma(\mathcal{B}) \subset \mathcal{R}$ , observe that any interval  $[q, \infty)$  can be generated by unions, intersections, and complements of open sets (very easy to show). Therefore  $\mathcal{R} = \sigma(\mathcal{G})$ . Since  $\mathcal{G}$  is countable,  $\mathcal{G} \times \cdots \times \mathcal{G}$  is countable and  $\sigma(\mathcal{G} \times \cdots \times \mathcal{G}) = \mathcal{R} \times \cdots \times \mathcal{R} = \mathcal{R}^d$ .  $\square$

**Exercise 1.6:** Suppose  $X$  and  $Y$  are random variables on  $(\Omega, \mathcal{F}, P)$  and let  $A \in \mathcal{F}$ . Show that if we let  $Z(\omega) = X(\omega)$  for  $\omega \in A$  and  $Z(\omega) = Y(\omega)$  for  $\omega \in A^c$ , then  $Z$  is a random variable.

We want to show that  $Z^{-1}(B) \in \mathcal{F}$  given  $B \in \mathcal{R}$  (borel set). By supposition we have  $Z^{-1}(B) \cap A = X^{-1}(B) \cap A$  and  $Z^{-1}(B) \cap A^c = Y^{-1}(B) \cap A^c$ . Therefore,

$$\begin{aligned} Z^{-1}(B) &= (Z^{-1}(B) \cap A) \cup (Z^{-1}(B) \cap A^c) \\ &= (X^{-1}(B) \cap A) \cup (Y^{-1}(B) \cap A^c) \stackrel{*}{\in} \mathcal{F} \end{aligned}$$

To show (\*), observe that since  $X$  and  $Y$  are random variables, we know that  $X^{-1}(B), Y^{-1}(B) \in \mathcal{F}$ . Then since  $A, A^c \in \mathcal{F}$  it's clear that both  $X^{-1}(B) \cap A$  and  $Y^{-1}(B) \cap A^c \in \mathcal{F}$ . So their intersection must also be an element of  $\mathcal{F}$ .  $\square$

**Exercise 1.8:** Show that a distribution function has at most countably many discontinuities.

Let  $D = \{x \in \mathbb{R} : F(x-) \neq F(x+)\}$ . Since  $F$  is increasing,  $x \in D \implies F(x-) < F(x+) \implies F(x+) - F(x-) > 0$ . However, it's also clear from the fact that  $F$  is increasing and  $F(-\infty) = 0, F(\infty) = 1$  that  $\sum_{x \in D} F(x+) - F(x-) \leq 1 < \infty$ . Therefore  $D$  must be countable infinite since the sum over the terms is finite [1, p. 11].  $\square$

*Alternate solution partially inspired from a clever answer on stack exchange*

Let  $D$  be the set of points of discontinuity. Since  $F(x-) < F(x+)$  for all  $x \in D$ , choose a rational  $r_x$  such that  $F(x-) < r_x < F(x+)$ . Because  $F$  is increasing, the intervals  $(F(x-), F(x+))$  are mutually disjoint. So  $x \rightarrow r_x$  is an injective function from  $D \rightarrow \mathbb{Q}$ . Therefore  $D$  must be countably infinite.  $\square$

**Exercise 1.9:** Show that if  $F(x) = P(X \leq x)$  is continuous then  $Y = F_X(X)$  has a uniform distribution on  $(0, 1)$ . That is, if  $y \in [0, 1]$ ,  $P(Y \leq y) = y$

Suppose  $F_X(x) = P(X \leq x)$  is continuous. Let  $Y = F_X(X)$ . Then for  $y \in [0, 1]$

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(F_X(X) \leq y) \\ &= P(X \leq F_X^{-1}(y)) && \text{(using } F^{-1} \text{ on page 6)} \\ &= F_X(F_X^{-1}(y)) = y \end{aligned}$$

$\square$

**Exercise 1.10:** Suppose that  $X$  has density  $f$ ,  $P(\alpha \leq X \leq \beta) = 1$  and  $g$  is a function that is increasing and differentiable on  $(\alpha, \beta)$ . Then  $g(X)$  has density  $f(g^{-1}(x))/g'(g^{-1}(x))$  for  $x \in (g(\alpha), g(\beta))$  and 0 otherwise. When  $g(x) = ax + b$  with  $a \geq 0$  the answer is  $f((y - b)/a)/a$ .

Let  $F_{g(X)}(x)$  be the distribution function for  $g(x)$ . Then the density function is  $h = \frac{d}{dx} F_{g(X)}(x)$ . Then

$$\begin{aligned} F_{g(X)}(x) &= P(g(X) \leq x) \\ &= P(X \leq g^{-1}(x)) && \text{(this exists because } g \text{ is increasing)} \\ &= F_X(g^{-1}(x)) \end{aligned}$$

So by definition

$$\begin{aligned} h &= F_{g(X)}(x) = \frac{d}{dx} F_X(g^{-1}(x)) \\ &= f(g^{-1}(x)) \left( \frac{d}{dx} g^{-1}(x) \right) \\ &= f(g^{-1}(x)) \frac{1}{g'(g^{-1}(x))} = \frac{f(g^{-1}(x))}{g'(g^{-1}(x))} \end{aligned}$$

Also notice that

$$P(\alpha \leq X \leq \beta) = 1 \implies P(g(\alpha) \leq g(X) \leq g(\beta)) = 1$$

So then  $h(x)$  is the density function for  $x \in (g(\alpha), g(\beta))$ .

**Exercise 1.11:** Suppose  $X$  has a normal distribution. Use the previous exercise to compute the density of  $\exp(X)$ . This is called the **lognormal distribution**.

I'm assuming  $X$  has a standard normal distribution ( $\mu = 0, \sigma = 1$ ). Then using the previous problem we have that  $g(X) = \exp(X)$  has density

$$\begin{aligned} \frac{f(g^{-1}(x))}{g'(g^{-1}(x))} &= \frac{f(\ln(x))}{\exp(\ln(x))} \\ &= \frac{f(\ln(x))}{x} \\ &= \frac{1}{x\sqrt{2\pi}} \exp\left(-\frac{\ln(x)^2}{2}\right), \quad \text{for } x \in (\exp(-\infty), \exp(\infty)) = (0, \infty) \end{aligned}$$

**Exercise 1.12:**

- i) Suppose  $X$  has density function  $f$ . Compute the distribution function of  $X^2$  and then differentiate to find its density function.

$$\begin{aligned} F_{X^2}(x) &= P(X^2 \leq x) \\ &= P(X \leq \sqrt{x}) + P(X \leq -\sqrt{x}) \\ &= F(\sqrt{x}) + F(-\sqrt{x}) \end{aligned}$$

Now differentiating

$$\begin{aligned}\frac{d}{dx}F(\sqrt{x}) - F(-\sqrt{x}) &= \frac{d}{dx} \left( \int_0^{\sqrt{x}} f dx + \int_0^{-\sqrt{x}} f dx \right) \\ &= \frac{1}{2\sqrt{x}} (f(\sqrt{x}) + f(-\sqrt{x}))\end{aligned}$$

□

- ii) Work out the answer when  $X$  has a standard normal distribution to find the density of the **chi-square distribution**.

With  $f = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$

$$\begin{aligned}\frac{1}{2\sqrt{x}} (f(\sqrt{x}) + f(-\sqrt{x})) &= \frac{1}{2\sqrt{2\pi x}} \left( \exp\left(-\frac{x}{2}\right) + \exp\left(-\frac{x}{2}\right) \right) \\ &= \frac{1}{\sqrt{2\pi x}} \exp\left(-\frac{x}{2}\right)\end{aligned}$$

□

## References

- [1] G.B. Folland. *Real Analysis: Modern Techniques and Their Applications*. Pure and Applied Mathematics: A Wiley Series of Texts, Monographs and Tracts. Wiley, 2013. ISBN: 9781118626399.
- [2] H. L. Royden. *Real analysis / H.L. Royden*. eng. Third edition. New York: Macmillan, 1988 - 1988. ISBN: 0024041513.