

ECON550: Problem Set 2

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HMC Exercises (7th edition)

1.4.1

$$\begin{aligned}P(C_2 \cup C_3 \cup \dots | C_1) &= \frac{P(C_1 \cap (C_2 \cup C_3 \cup \dots))}{P(C_1)} \\&= \frac{P((C_1 \cap C_2) \cup (C_1 \cap C_3) \cup \dots)}{P(C_1)}\end{aligned}$$

Since all the C_2, C_3, \dots are disjoint,

$$\begin{aligned}&= \frac{P(C_1 \cap C_2) + P(C_1 \cap C_3) + P(C_1 \cap C_4) + \dots}{P(C_1)} \\&= P(C_2 | C_1) + P(C_3 | C_1) + P(C_4 | C_1) + \dots\end{aligned}$$

1.4.2

$$\begin{aligned}P(C_1 \cap C_2 \cap C_3 \cap C_4) &= \frac{P(C_1)}{P(C_1)} \frac{P(C_1 \cap C_2)}{P(C_1 \cap C_2)} \frac{P(C_1 \cap C_2 \cap C_3)}{P(C_1 \cap C_2 \cap C_3)} P(C_1 \cap C_2 \cap C_3 \cap C_4) \\&= P(C_1) \frac{P(C_1 \cap C_2)}{P(C_1)} \frac{P(C_1 \cap C_2 \cap C_3)}{P(C_1 \cap C_2)} \frac{P(C_1 \cap C_2 \cap C_3 \cap C_4)}{P(C_1 \cap C_2 \cap C_3)} \\&= P(C_1) P(C_2 | C_1) P(C_3 | C_1 \cap C_2) P(C_4 | C_1 \cap C_2 \cap C_3)\end{aligned}$$

Only using Bayes' rule.

1.4.9 Let $P(a)$ denote the probability of drawing exactly a blue chips from bowl 1. The probability of drawing a blue chip from the second bowl is then given by

$$\begin{aligned}P(0) \frac{0}{5} + P(1) \frac{1}{5} + P(2) \frac{2}{5} + P(3) \frac{3}{5} + P(4) \frac{4}{5} \\= \frac{6}{252} \frac{0}{5} + \frac{60}{252} \frac{1}{5} + \frac{120}{252} \frac{2}{5} + \frac{60}{252} \frac{3}{5} + \frac{6}{252} \frac{4}{5}\end{aligned}$$

The probability of drawing exactly 3 blue chips from bowl I and then drawing a blue chip from bowl II is

$$\frac{60}{252} \frac{3}{5}$$

Thus, the conditional probability desired is

$$\frac{180}{60 + 240 + 180 + 24} = \frac{5}{14}$$

1.4.12(c) C_1 and C_2^c are independent, so

$$P(C_1 \cap C_2^c) = 0.42$$

Hence

$$P(C_1 \cup C_2^c) = P(C_1) + P(C_2^c) - P(C_1 \cap C_2^c) = 0.88$$

1.4.23

$$\begin{aligned} P(C_1 \cup C_2 \cup C_3) &= P(C_1) + P(C_2) + P(C_3) - P(C_1 \cap C_2) - P(C_2 \cap C_3) - P(C_1 \cap C_3) + P(C_1 \cap C_2 \cap C_3) \\ &= \frac{1}{2} + \frac{1}{3} + \frac{1}{4} - \frac{1}{6} - \frac{1}{12} - \frac{1}{8} + \frac{1}{24} = \frac{3}{4} \end{aligned}$$

1.5.1 The induced probability is:

$$P_X(0) = \frac{9}{13}$$

$$P_X(1) = \frac{1}{13}$$

$$P_X(2) = \frac{1}{13}$$

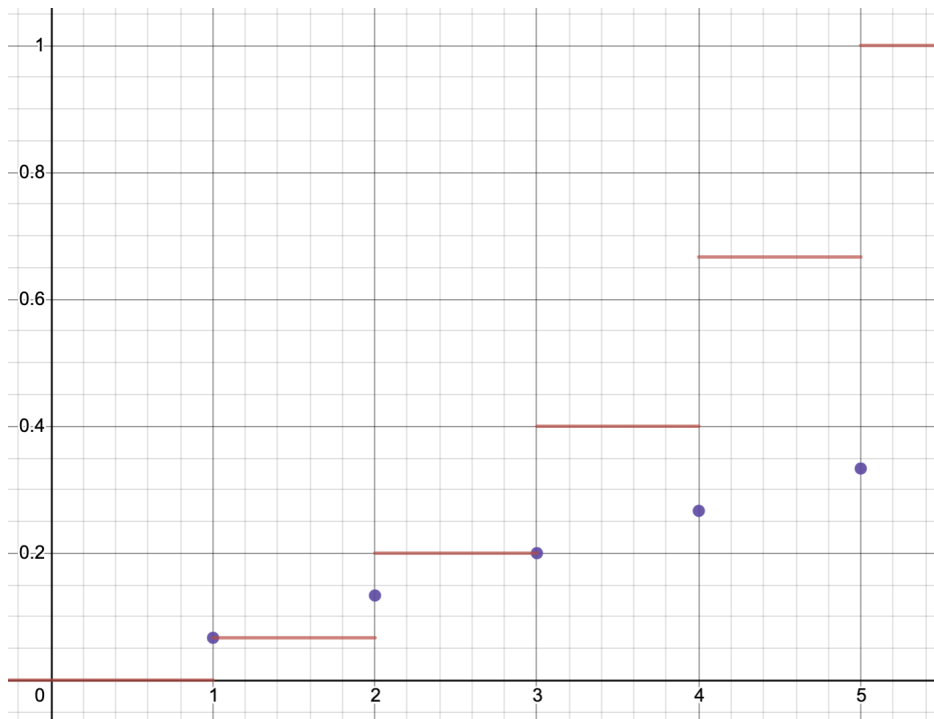
$$P_X(3) = \frac{1}{13}$$

$$P_X(4) = \frac{1}{13}$$

1.5.4(c) The cdf is given by:

$$\left\{ \begin{array}{ll} 0 & x < 1 \\ 1/15 & 1 \leq x < 2 \\ 3/15 & 2 \leq x < 3 \\ 6/15 & 3 \leq x < 4 \\ 10/15 & 4 \leq x < 5 \\ 1 & 5 \leq x \end{array} \right.$$

We have sketched the CDF in red and the PDF in purple in the next page.



1.5.6

$$P_X(D_1) = \int_1^2 \frac{2x}{9} dx = 1/9$$

1.5.8(d) This probability is 0; there is only a nonzero probability for set with a nonempty intersection with $[-1, 1]$.

1.6.3(a) The pmf is

$$p(x) = \frac{1}{6} \left(\frac{5}{6} \right)^{x-1}$$

1.6.3(b)

$$\begin{aligned} \sum p(x) &= \sum \frac{1}{6} \left(\frac{5}{6} \right)^{x-1} \\ &= \frac{1/6}{1 - (5/6)} = 1 \end{aligned}$$

1.6.3(d)

$$F(x) = \sum_{k=1}^x \frac{1}{6} \left(\frac{5}{6} \right)^{k-1} = 1 - \left(\frac{5}{6} \right)^x$$

1.7.1 CDF:

$$F_X(x) = \frac{\sqrt{x}}{10}$$

PMF:

$$f_X(x) = \frac{1}{5\sqrt{x}}$$

1.7.6(a)

$$P(|X| < 1) = \int_{-1}^1 x^2/18 \, dx = 1/3$$
$$P(X^2 < 9) = 1$$

1.7.7

$$P_X(C_1 \cup C_2) = 1/2 + (1/20) = 11/20$$
$$P_X(C_1 \cap C_2) = 0$$

1.7.9(b) Median is

$$x^3 = \frac{1}{2} \implies x = \frac{1}{\sqrt[3]{2}}$$

1.7.15 Negating the expression, we get

$$-P(X > z) \leq -P(Y > z)$$
$$1 - P(X > z) \leq 1 - P(Y > z)$$
$$P(X \leq z) \leq P(Y \leq z)$$
$$F_X(z) \leq F_Y(z)$$

Strict inequality holding for at least one z .

Problem 4

We know that $\lim_{x \rightarrow -\infty} F_X(x) = \lim_{x \rightarrow -\infty} Pr(X \leq x)$

$$= \lim_{x \rightarrow -\infty} 1 - Pr(X > x)$$
$$= 1 - \lim_{x \rightarrow -\infty} Pr(X > x)$$

But as $x \rightarrow -\infty$ the interval $(x, \infty) \rightarrow \mathbb{R}$, and hence $\lim_{x \rightarrow -\infty} Pr(X > x) = Pr(X \in \mathbb{R}) = 1$. Therefore

$$\lim_{x \rightarrow -\infty} F_X(x) = 1 - \lim_{x \rightarrow -\infty} Pr(X > x)$$
$$= 1 - 1 = 0$$

as desired.

Problem 5

To show this is a σ -field, we show this is closed under complements and finite unions. To show closure under complements, consider some $S \in \mathcal{B}(\Omega_X)$. Then we know $\exists B \in \mathcal{B}$ such that

$$S = B \cap \Omega_X$$

$$\begin{aligned}\Omega_X \setminus S &= \Omega_X \setminus (B \cap \Omega_X) = \\ &= \Omega_X \setminus B = B^c \cap \Omega_X\end{aligned}$$

Since \mathcal{B} is a σ -field, $B^c \in \mathcal{B}$, and hence $S = B^c \cap \Omega_X$ is in $\mathcal{B}(\Omega_X)$.

Now, we show closure under countable union. Consider $S_1, S_2, \dots \in \mathcal{B}(\Omega_X)$. We know $\exists B_1, B_2, \dots \in \mathcal{B}$ such that $S_i = B_i \cap \Omega_X$. Let

$$\begin{aligned}S &= \cup_i S_i \\ &= \cup_i (B_i \cap \Omega_X) \\ &= (\cup_i B_i) \cap (\cup_i \Omega_X) \\ &= (\cup_i B_i) \cap \Omega_X\end{aligned}$$

But $\cup_i B_i \in \mathcal{B}$ since \mathcal{B} is a σ -field, and so we have that $S \in \mathcal{B}(\Omega_X)$, and hence we have the closure properties. So $\mathcal{B}(\Omega_X)$ is a σ -field.

Problem 6

Let s_1, s_2 be simple functions. Then there exists some disjoint sets A_j such that

$$s_1(x) = \sum_{j=1}^M c_j 1_{A_j}$$

$$s_2(x) = \sum_{j=1}^M d_j 1_{A_j}$$

Then

$$\begin{aligned}\int (s_1 + s_2) d\mu &= \sum_{j=1}^M (c_j + d_j) \mu(A_j) \\ &= \sum_{j=1}^M c_j \mu(A_j) + \sum_{j=1}^M d_j \mu(A_j) \\ &= \sum_{j=1}^M c_j \mu(A_j) + \sum_{j=1}^M d_j \mu(A_j)\end{aligned}$$

$$= \int s_1 d\mu + \int s_2 d\mu$$

Further, we have that

$$\begin{aligned} \int a s_1 d\mu &= \sum_{j=1}^M a c_j \mu(A_j) \\ &= a \sum_{j=1}^M c_j \mu(A_j) \\ &= a \int s_1 d\mu \end{aligned}$$

So we have linearity of the Lebesgue integral on simple functions.

Problem 7

We first show linearity for nonnegative, measurable functions. Suppose we have two such functions f, g . Then, we note that we can construct two increasing sequences of simple functions $\{s_n\} \rightarrow f, \{s'_n\} \rightarrow g$, such that by the monotone convergence

$$\begin{aligned} \int f d\mu &= \lim_{n \rightarrow \infty} \int s_n d\mu \\ \int g d\mu &= \lim_{n \rightarrow \infty} \int s'_n d\mu \end{aligned}$$

So

$$\int f d\mu + \int g d\mu = \lim_{n \rightarrow \infty} \left(\int s_n d\mu + \int s'_n d\mu \right)$$

By linearity of simple functions as we showed in the previous problem,

$$\begin{aligned} \int f d\mu + \int g d\mu &= \lim_{n \rightarrow \infty} \left(\int s_n d\mu + \int s'_n d\mu \right) \\ &= \lim_{n \rightarrow \infty} \left(\int (s_n + s'_n) d\mu \right) \end{aligned}$$

We note then that the sequence $s_n^* = s_n + s'_n$ must also be increasing, since both s_n and s'_n are increasing. Further, $\{s_n^*\} \rightarrow f + g$. Hence

$$\begin{aligned} \int f d\mu + \int g d\mu &= \lim_{n \rightarrow \infty} \left(\int (s_n + s'_n) d\mu \right) \\ &= \int (f + g) d\mu \end{aligned}$$

We also know that for scalar a , $\{a s_n\} \rightarrow a f$, and hence

$$\int a f d\mu = \lim_{n \rightarrow \infty} \int a s_n d\mu$$

$$\begin{aligned}
&= \lim_{n \rightarrow \infty} a \int s_n d\mu \\
&= a \lim_{n \rightarrow \infty} \int s_n d\mu \\
&= a \int f d\mu
\end{aligned}$$

So Lebesgue integration is linear for nonnegative, measurable functions.

Now more generally, suppose f, g are integrable. Then we have for $f = f^+ - f^-$, $g = g^+ - g^-$. f^+, g^+, f^-, g^- all nonnegative. Then

$$\begin{aligned}
\int f d\mu &= \int f^+ d\mu - \int f^- d\mu \\
\int g d\mu &= \int g^+ d\mu - \int g^- d\mu \\
\int f d\mu + \int g d\mu &= \int f^+ d\mu - \int f^- d\mu + \int g^+ d\mu - \int g^- d\mu \\
&= \int (f^+ + g^+) d\mu - \int (f^- + g^-) d\mu \\
&= \int ((f^+ + g^+) - (f^- + g^-)) d\mu \\
&= \int (f + g) d\mu
\end{aligned}$$

using the linearity of nonnegative functions we proved earlier. Further,

$$\begin{aligned}
\int a f d\mu &= \int a f^+ d\mu - \int a f^- d\mu \\
&= a \left(\int f^+ d\mu - \int f^- d\mu \right) \\
&= a \int f d\mu
\end{aligned}$$

So we have shown linearity of the Lebesgue integral on integrable functions.