

# Homework 1

Kejia Huang

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## Exercise 1.5

The function  $\mathbb{I}_{[0, X(\omega)]}(x)$  is a indicator function, so  $\mathbb{I}_{[0, X(\omega)]}(x) = 1$  when  $x \in [0, X(\omega)]$

$$\int_{\Omega} \int_0^{\infty} \mathbb{I}_{[0, X(\omega)]}(x) dx d\mathbb{P}(w) = \int_{\Omega} \int_0^{X(\omega)} 1 dx d\mathbb{P}(w) = \int_{\Omega} X(\omega) d\mathbb{P}(w) = \mathbb{E}X$$

The function  $\mathbb{I}_{[0, X(\omega)]}(x) = 1$ , so it is integrable, so the order of the integration can be reversed

$$\begin{aligned} \int_{\Omega} \int_0^{\infty} \mathbb{I}_{[0, X(\omega)]}(x) dx d\mathbb{P}(w) &= \int_0^{\infty} \int_{\Omega} \mathbb{I}_{[0, X(\omega)]}(x) dx d\mathbb{P}(w) = \int_0^{\infty} \int_{\Omega} 1_{(x < X(\omega))} d\mathbb{P}(w) dx \\ &= \int_0^{\infty} P(x < X) dx = \int_0^{\infty} 1 - P(X \leq x) dx = \int_0^{\infty} 1 - F(x) dx \end{aligned}$$

so

$$\mathbb{E}(X) = \int_0^{\infty} 1 - F(x) dx$$

## Exercise 1.6

i

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Verify that

$$\mathbb{E}(e^{uX}) = e^{u\mu + \frac{1}{2}u^2\sigma^2}$$

$$\begin{aligned} \mathbb{E}(e^{uX}) &= \int_{-\infty}^{\infty} e^{ux} f(x) dx = \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{\frac{2\sigma^2 ux - (x^2 + \mu^2 - 2\mu x)}{2\sigma^2}} dx \\ &= \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{[x - (u\sigma^2 + \mu)]^2 - u^2\sigma^4 - 2u\mu\sigma^2}{2\sigma^2}} dx \\ &= e^{\frac{u^2\sigma^2 + 2u\mu}{2}} \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{[x - (u\sigma^2 + \mu)]^2}{2\sigma^2}} dx \end{aligned}$$

because

$$\int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{[x-(u\sigma^2+\mu)]^2}{2\sigma^2}} dx$$

is cdf of random variable with expectation  $u\sigma^2 + \mu$  and deviation  $\sigma^2$

so

$$\int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{[x-(u\sigma^2+\mu)]^2}{2\sigma^2}} dx = F(\infty) - F(-\infty) = 1$$

So

$$\mathbb{E}(e^{uX}) = e^{\frac{u^2\sigma^2+2u\mu}{2}}$$

### Exercise 1.7

i

$$f(x) = \lim_{n \rightarrow +\infty} f_n(x) = \lim_{n \rightarrow +\infty} \frac{1}{\sqrt{2n\pi}} e^{-\frac{x^2}{2n}}$$

$$\lim_{n \rightarrow +\infty} f_n(x) = \lim_{n \rightarrow +\infty} \frac{1}{\sqrt{2n\pi}} = 0 \quad \text{and} \quad \lim_{n \rightarrow +\infty} f_n(x) = \lim_{n \rightarrow +\infty} e^{-\frac{x^2}{2n}} = 1$$

so

$$f(x) = \lim_{n \rightarrow +\infty} f_n(x) = \lim_{n \rightarrow +\infty} \frac{1}{\sqrt{2n\pi}} e^{-\frac{x^2}{2n}} = 0$$

ii

$$\int_{\infty}^{-\infty} \frac{1}{\sqrt{2nx}} e^{-\frac{x^2}{2n}} dx$$

it is the cdf of the normal random variable with expectation 0 and deviation  $\sqrt{n}$

so it is equal to  $F(\infty) - F(-\infty) = 1$

so

$$\lim_{n \rightarrow \infty} \int_{\infty}^{-\infty} \frac{1}{\sqrt{2nx}} e^{-\frac{x^2}{2n}} dx = \lim_{n \rightarrow \infty} 1 = 1$$

iii

the condition of the Monotone Convergence Theorem is  $0 \leq f_1 \leq f_2 \leq f_3 \leq \dots$  almost everywhere to a function  $f$

set  $t = \frac{1}{\sqrt{2n}}$  and  $n \in \mathbb{R}$

$$\begin{aligned}\frac{df_n(x)}{dn} &= \frac{df_n(x)}{dt} \frac{dt}{dn} = e^{-x^2 t^2} (1 - 2t^2) \left(-\frac{1}{2\sqrt{2}} n^{-\frac{3}{2}}\right) \\ &= e^{-x^2 t^2} \left(1 - \frac{1}{n}\right) \left(-\frac{1}{2\sqrt{2}} n^{-\frac{3}{2}}\right) < 0\end{aligned}$$

so  $f(n) \geq f(n+1)$  ( $n \in 1, 2, 3, \dots$ )

so  $f_n(x)$  is not in accordance with the condition of the Monotone Convergence Theorem

so this does not violate the Monotone Convergence Theorem

### Exercise 1.10

i

$\mathbb{B}[0, 1]$  is Borel  $\sigma$ -algebra and  $A \in \mathbb{B}[0, 1]$

$P(\Omega) = 1$  and  $A$  satisfied with countable additivity

so  $S_0(\Omega, \mathbb{A}, \mathbb{P})$  is a probability space

because  $Z$  is a nonnegative random variable and  $\mathbb{E}(Z) = 1$

defined  $\tilde{\mathbb{P}}(A) = \int^A Z(\omega) d\mathbb{P}(\omega)$

so  $\tilde{\mathbb{P}}$  is a probability measure

ii

if  $\tilde{\mathbb{P}}(A) = 0$  then  $A \notin \mathbb{B}[0, 1]$

$$\tilde{\mathbb{P}}(A) = \int_{A \cup [0, \frac{1}{2})} 0 d\mathbb{P}(\omega) + \int_{A \cup [\frac{1}{2}, 1]} 2 d\mathbb{P}(\omega) = 2P(A \cup [\frac{1}{2}, 1]) + 0P(A \cup [0, \frac{1}{2})) = 0$$

iii

$$\tilde{\mathbb{P}}(A) = 2P(A \cup [\frac{1}{2}, 1]) + 0P(A \cup [0, \frac{1}{2}))$$

$$\tilde{\mathbb{P}}(A) = 0, P(A) > 0 \text{ so } A \text{ is the subset of } [0, \frac{1}{2})$$

### Exercise 1.13

i

$$\frac{1}{\epsilon} \mathbb{P}\{X \in B(x, \epsilon)\} = \frac{1}{\epsilon} \int_{x+\frac{\epsilon}{2}}^{x-\frac{\epsilon}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du$$

according to the Mean Value Theorem

$$\frac{\int_{x+\frac{\epsilon}{2}}^{x-\frac{\epsilon}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du}{(x+\frac{\epsilon}{2}) - (x-\frac{\epsilon}{2})} = \frac{1}{\sqrt{2\pi}} e^{-\frac{\xi^2}{2}} \quad \xi \in [x - \frac{\epsilon}{2}, x + \frac{\epsilon}{2}]$$

because  $x \in [x - \frac{\epsilon}{2}, x + \frac{\epsilon}{2}]$  so

$$\frac{\int_{x+\frac{\epsilon}{2}}^{x-\frac{\epsilon}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du}{(x+\frac{\epsilon}{2}) - (x-\frac{\epsilon}{2})} = \frac{\int_{x+\frac{1}{\epsilon}}^{x-\frac{1}{\epsilon}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du}{\epsilon} = \frac{1}{\epsilon} \int_{x+\frac{1}{\epsilon}}^{x-\frac{1}{\epsilon}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du \doteq \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

ii

just use the Mean Value Theorem too

$$\frac{1}{\epsilon} \tilde{\mathbb{P}}\{Y \in B(y, \epsilon)\} = \frac{1}{\epsilon} \int_{y+\frac{\epsilon}{2}}^{y-\frac{\epsilon}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du \doteq \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}$$

iii

$$\{X \in B(x, \epsilon)\} \rightarrow \{Y - \theta \in B(x, \epsilon)\} \rightarrow \{Y \in B(x + \theta, \epsilon)\} \rightarrow \{Y \in B(y, \epsilon)\}$$

iv

$$\frac{\tilde{\mathbb{P}}(A)}{\mathbb{P}(A)} \doteq \exp \frac{X^2(\bar{\omega}) - Y^2(\bar{\omega})}{2} = \exp \frac{X^2(\bar{\omega}) - (X(\bar{\omega}) + \theta)^2}{2} = \exp \left\{ \frac{-\theta^2 - 2\theta X(\bar{\omega})}{2} \right\}$$

**Exercise 2.2**

i

$$\sigma(X) = \{\emptyset, \Omega, X = 1, X = 0\} = \{\emptyset, \Omega, \{HT, TH\}, \{HH, TT\}\}$$

ii

$$\sigma(S_1) = \{\emptyset, \Omega, S_1 = 8, S_1 = 2\} = \{\emptyset, \Omega, \{HH, HT\}, \{TH, TT\}\}$$

iii the condition for A and B is independent under probability measure  $\tilde{\mathbb{P}}$  is

$$\tilde{\mathbb{P}}(A \cap B) = \tilde{\mathbb{P}}(A) * \tilde{\mathbb{P}}(B)$$

$$\tilde{\mathbb{P}}(\{HT, TH\} \cap \{HH, HT\}) = \tilde{\mathbb{P}}(\{HT\}) = \frac{1}{4}, \quad \tilde{\mathbb{P}}(\{HT, TH\}) = \frac{1}{2}, \quad \tilde{\mathbb{P}}(\{HH, HT\}) = \frac{1}{2}$$

so

$$\tilde{\mathbb{P}}(\{HT, TH\} \cap \{HH, HT\}) = \tilde{\mathbb{P}}(\{HT, TH\}) * \tilde{\mathbb{P}}(\{HH, HT\})$$

in this way, we can get  $\tilde{\mathbb{P}}(A \cap B) = \tilde{\mathbb{P}}(A) * \tilde{\mathbb{P}}(B)$  for  $A \in \sigma(X)$   $B \in \sigma(S_1)$

so  $\sigma(X)$  and  $\sigma(S_1)$  are independent under measure  $\tilde{\mathbb{P}}$

iv use the same method as in iii

$$\tilde{\mathbb{P}}(\{HT, TH\} \cap \{HH, HT\}) = \tilde{\mathbb{P}}(\{HT\}) = \frac{2}{9}, \quad \tilde{\mathbb{P}}(\{HT, TH\}) = \frac{4}{9}, \quad \tilde{\mathbb{P}}(\{HH, HT\}) = \frac{2}{3}$$

so

$$\tilde{\mathbb{P}}(\{HT, TH\} \cap \{HH, HT\}) \neq \tilde{\mathbb{P}}(\{HT, TH\}) * \tilde{\mathbb{P}}(\{HH, HT\})$$

so  $\sigma(X)$  and  $\sigma(S_1)$  are not independent under measure  $\mathbb{P}$

v because  $\sigma(X)$  and  $\sigma(S_1)$  are not independent under measure  $\mathbb{P}$ , under the condition of  $X=1$ , the distribution of  $S_1$  can be changed

**Exercise 2.6**

i  $\sigma(X) = \{\emptyset, \Omega, X = 1, X = 2\} = \{\emptyset, \Omega, \{a, b\}, \{c, d\}\}$

ii

$$\mathbb{E}[Y|X](a) = \mathbb{E}[Y|X](b) = \frac{Y(a)\mathbb{P}(a) + Y(b)\mathbb{P}(b)}{\mathbb{P}(a) + \mathbb{P}(b)} = \frac{1}{3} - \frac{2}{3} = -\frac{1}{3}$$

$$\mathbb{E}[Y|X](c) = \mathbb{E}[Y|X](d) = \frac{Y(c)\mathbb{P}(c) + Y(d)\mathbb{P}(d)}{\mathbb{P}(c) + \mathbb{P}(d)} = 0$$

$$\int_{\Omega} \mathbb{E}[Y|X](\omega) d\mathbb{P}(\omega) = -\frac{1}{3} * (\frac{1}{6} + \frac{1}{3}) = -\frac{1}{6}$$

$$\int_{\Omega} X(\omega) d\mathbb{P}(\omega) = \frac{1}{6} - \frac{1}{3} + \frac{1}{4} - \frac{1}{4} = -\frac{1}{6} = \int_{\Omega} \mathbb{E}[Y|X](\omega) d\mathbb{P}(\omega)$$

so the partial-averaging property is satisfied

iii

$$\int_{\Omega} \mathbb{E}[Z|X](\omega) d\mathbb{P}(\omega) = \mathbb{E}[Z|X] = \mathbb{E}[X+Y|X] = X + \mathbb{E}[Y|X] = X_{(-1,1)} - \frac{1}{6} = -\frac{1}{6}$$

$$\int_{\Omega} Z(\omega) d\mathbb{P}(\omega) = \mathbb{E}[Z] = \frac{1}{3} * (-2 * \frac{1}{4}) = -\frac{1}{6}$$

so the partial-averaging property is satisfied

iv

$$\mathbb{E}[Z|X] - \mathbb{E}[Y|X] = -\frac{1}{6} + X - (-\frac{1}{6}) = X$$

according to Taking out what is known in Theorem 2.3.2

### Exercise 2.8

set  $\alpha$  is one of the  $\sigma(X)$  measurable r.v.

$$\mathbb{E}[Y_2|\alpha] = \mathbb{E}[Y - \mathbb{E}[Y|X]|\alpha] = \mathbb{E}[Y|\alpha] - \mathbb{E}[\mathbb{E}[Y|X]|\alpha]$$

because  $\alpha$  is  $\sigma(X)$  measurable

so

$$\mathbb{E}[Y_2|\alpha] = \mathbb{E}[Y|\alpha] - \mathbb{E}[Y|\alpha] = 0$$

according to Iterated conditioning

so  $Y_2$  and  $X$  are uncorrelated

**Exercise 2.10**

$$\begin{aligned}\int_A g(X) d\mathbb{P} &= \int_{-\infty}^{\infty} g(x) f_X(x) dx = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{y f_{x,y}(x, y)}{f_X(x)} f_X(x) dy dx \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f_{x,y}(x, y) dy dx = \mathbb{E}[Y] = \int_A Y d\mathbb{P}\end{aligned}$$

**extra problem**

**A**

$$E[M_1] = p - q = M_0 = 0 \quad p + q = 1$$

so  $q = p = 0.5$

**B**

$$E[M_{302} | M_{300} = 60] = E[M_{300}] + E[M_2] = 60 + 2 * \frac{1}{4} + 0 * \frac{1}{2} - 2 * \frac{1}{4} = 60$$