## Homework 3

## Question 1

(Change-of-variable formula, Klenke Theorem 4.10) Let  $(\Omega, \mathcal{A})$  and  $(\Omega', \mathcal{A}')$  be measurable spaces, let  $\mu$  be a measure on  $(\Omega, \mathcal{A})$  and let  $X : \Omega \to \Omega'$  be measurable. Let  $\mu' = \mu \circ X^{-1}$  be the image measure of  $\mu$  under the map X, i.e., for every  $E \in \mathcal{A}'$ , we define

$$\mu'(E) \triangleq \mu(X^{-1}(E)).$$

Assume that  $f: \Omega' \to \mathbb{R}$  is  $\mu'$ -integrable. Show that  $f \circ X$  is  $\mu$ -integrable and  $\int (f \circ X) d\mu = \int f d\mu'$ .

(Hint: You can follow the proof of Theorem 1.6.9 in Durrett, but you need to write the answer in your own words.)

#### **Solution:**

We will prove this result by verifying it in four increasingly general special cases.

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Case 1: Indicator functions. If E \in \mathcal{A}' and f = 1_E then \int (f \circ X) d\mu = \int 1_E(X(\omega)) d\mu(\omega) = \mu(X^{-1}(E)) = \mu'(E) = \int 1_E d\mu' = \int f d\mu'. Case 2: Simple functions. If f(x) = \sum_{i=1}^n c_i 1_{E_i} where c_i \in \mathbf{R}, E_i \in \mathcal{A}' then \int (f \circ X) d\mu = \int \sum_{i=1}^n c_i 1_{E_i}(X(\omega)) d\mu = \sum_{i=1}^n c_i \int 1_{E_i}(X(\omega)) d\mu = \sum_{i=1}^n c_i \int 1_{E_i} d\mu' = \int \sum_{i=1}^n c_i 1_{E_i} d\mu' = \int f d\mu'. Case 3: Nonnegative functions. Now, if f > 0 and we let
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Case 3: Nonnegative functions. Now if  $f \ge 0$  and we let

$$f_n(x) = ([2^n f(x)]/2^n) \wedge n$$

where [x] = the largest integer  $\leq x$  and  $a \wedge b = min\{a, b\}$ , then the  $f_n$  are simple and  $f_n \uparrow f$ , so using the result for simple functions and the monotone convergence theorem:

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\int (f \circ X) d\mu = \lim_n \int (f_n \circ X) d\mu = \lim_n \int f_n d\mu' = \int f d\mu'. Case 4: Integrable functions. The general case now follows by writing f(x) = f(x)^+ - f(x)^- where f(x)^+ and f(x)^- are nonnegative functions. Then \int (f \circ X) d\mu = \int (f^+ \circ X) d\mu - \int (f^- \circ X) d\mu = \int f^+ d\mu' - \int f^- d\mu' = \int f d\mu'. In addition, \int |f \circ X| d\mu = \int (f^+ \circ X) d\mu + \int (f^- \circ X) d\mu = \int f^+ d\mu' + \int f^- d\mu' = \int |f| d\mu' < \infty. So f \circ X is \mu-integrable.
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# Question 2

(Klenke exercise 4.2.1) Let  $(\Omega, \mathcal{A}, \mu)$  be a measure space and let  $f \in L^1(\mu)$ . Show that for any  $\epsilon > 0$ , there is an  $A \in \mathcal{A}$  with  $\mu(A) < \infty$  and

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|\int_A f \ d\mu - \int f \ d\mu| < \epsilon. Solution:
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// We can answer this question without using Markov inequality, or MCT, or DCT.

// We note that this question is trivial if  $(\Omega, \mathcal{A}, \mu)$  is a probability space, because in this case we can simply take  $A = \Omega$ . Sketch of proof.

We can first assume that the function f is non-negative and  $\int f d\mu$  is finite. Let c denote the value of  $\int f d\mu$ . By the definition of Lebesgue integral for nonnegative function, c is the supremum of  $\int h \ d\mu$ , with h ranging over all simple functions such that  $0 \le h \le f$ . For any  $\epsilon > 0$ , we can find a simple function h in the form

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h(\omega)=\sum_{i=1}^m\ a_i\mathbf{1}_{A_i}(\omega), so that 0\leq h\leq f , a_i>0 for all i, and c-\epsilon\leq\int h\ d\mu\leq c.
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With this choice of h, we get

(\*) 
$$\int f \ d\mu - \int h \ d\mu \le \epsilon \ .$$

The measure of  $A_i$  could not be infinite, otherwise the integral of h would be infinite, violating the assumption that  $\int f d\mu$  is finite. We take A to be the union of  $A_i$ , for i = 1, 2, ..., m. As each  $A_i$  has finite measure, the union A also has finite measure.

We next use the fact that the simple function h is less than or equal to f on the set A. By monotone property of Lebesgue integral,

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\int h \ d\mu = \int_A h \ d\mu \le \int_A f \ d\mu.
Combining this with (*), we obtain
0 \le \int f \ d\mu - \int_A f \ d\mu \le \epsilon.
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This proves the statement in the question for non-negative function f.

For general real-valued f, we apply the above argument for the positive and negative parts of f. Because f is assumed to be  $\mu$ -integrable,  $\int f^+ d\mu$  and  $\int f^- d\mu$  are both finite. Take A to be the union of all the sets appearing in the indicator functions that approximate the positive and negative part.

# Question 3

(Minkowski's inequality, Durrett Exercise 1.5.3) Let  $(\Omega, \mathcal{A}, \mu)$  be a measure space and f, g be real-valued  $\mu$ -measurable functions. Suppose that  $p \in (1, \infty)$ . The p-norm of function f, denoted by  $||f||_p$  is defined as  $(\int |f|^p d\mu)^{1/p}$ . The inequality  $|f+g|^p \leq 2^p (|f|^p + |g|^p)$  shows that if  $||f||_p$  and  $||g||_p$  are finite, then so is  $||f+g||_p$ . Apply Holder's inequality to  $|f| ||f+g||^{p-1}$  and  $|g| ||f+g||^{p-1}$  to show

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||f + g||_p \le ||f||_p + ||g||_p.
(Hint: See Theorem 7.17 in Klenke)
Solution:
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$$||f+g||_p^p = \int (f+g)^p d\mu = \int f(f+g)^{p-1} d\mu + \int g(f+g)^{p-1} d\mu$$

$$\leq ||f||_p \cdot ||(f+g)^{p-1}||_q + ||g||_p \cdot ||(f+g)^{p-1}||_q$$

$$= (||f||_p + ||g||_p) \cdot ||f+g||_p^{p-1}.$$
(1)

Note that in the last step, we used the fact that p - p/q = 1. Dividing both sides by  $||f + g||_p^{p-1}$  yields the result.

## Question 4

(Durrett Exercise 1.6.11) Prove that if  $E[|X|^k] < \infty$ , then for  $0 < j < k, E[|X|^j] < \infty$ , and furthermore  $E[|X|^j] \le E[|X|^k]^{j/k}$ .

(The purpose of this question is to show that the j-norm of a random variable is less than or equal to the corresponding k-norm for j < k.)

#### **Solution:**

// The asserted inequality is false in general when the measure of the whole space is not  $1\,$ 

Apply Jensen's inequality with the function  $\phi(x) = x^{k/j}$ , which is a convex function on the nonnegative real numbers when  $k \geq j$ . Suppose that  $E|X^k|$  is finite.

$$E[|X|^k] = E[\phi(|X|^j)]$$

$$\geq \phi(E[|X|^j])$$

$$= E[|X|^j]^{k/j}.$$

This is the same as  $E[\,|X|^j\,] \le E[\,|X|^k\,]^{j/k}$  .

Because  $E[|X|^k]$  is finite, the right-hand side of the above inequality is finite. The left-hand side is finite as well.

// For finite measure space, we can still prove that finite  $L_k$  norm implies finite  $L_j$  norm. See this wikipedia page. This uses the Holder inequality.

# Question 5

(Lebesgue integral with counting measure) Let  $\mathbb Z$  denote the set of integers and  $\mu$  be the counting measure, i.e., for subset  $S\subseteq \mathbb Z$ ,

$$\mu(S) = \begin{cases} |S| & \text{if } |S| \text{ is finite} \\ \infty & \text{otherwise.} \end{cases}$$

For any function  $f: \mathbb{Z} \to \mathbb{R}$ , let  $S_+$  be the set  $\{x \in \mathbb{Z} : f(x) > 0\}$  and  $S_-$  be the set  $\{x \in \mathbb{Z} : f(x) < 0\}$ .

(We may call  $S_+$  the positive support of f and  $S_-$  the negative support of f.) If  $f \in L^1(\mu)$ , show that

 $\int f \, d\mu = \sup\{ \sum_{x \in A} f(x) : |A| < \infty, A \subseteq S_+ \} - \sup\{ -\sum_{x \in B} f(x) : |B| < \infty, B \subseteq S_- \}.$ 

The first supremum is taken over all finite subset  $A \in \mathbb{Z}$  such that f(x) > 0 for all  $x \in A$ . Likewise, the second supremum is taken over all finite subset B such that f(x) < 0 for all  $x \in B$ .

### Solution:

Consider the positive part  $f^+$  of the function f first. By definition, the Lebesgue integral of  $f^+$  with counting measure on  $\mathbb Z$  is the supremum of the set  $S_1 = \{ \int_{\mathbb Z} \ g \ d\mu : 0 \le g \le f^+, g \ \text{simple} \}.$ 

For simple function g,  $\int_{\mathbb{Z}} g \ d\mu$  is just a finite summation.

This question is suggesting that it suffices to consider the set

 $S_2 = \{ \sum_{x \in S_1} f(x) : |S_1| < \infty, \ f(x) > 0 \ \forall x \in S_1 \},$  which may look simpler than  $S_1$ .

Let  $s_1 = \sup S_1$  and  $s_2 = \sup S_2$ . Because  $S_2$  is a subset of  $S_1$ . We have  $s_2 \leq s_1$ . Our goal is to prove the reverse inequality. Because it is assumed that f is in  $L^1(\mu)$ ,  $s_1$  is a finite number.

We claim that  $s_2 \geq s_1 - \epsilon$  for every arbitrarily small positive real numbers  $\epsilon$ . Since  $s_1 - \epsilon$  is strictly less than  $s_1$ ,  $s_1 - \epsilon$  is no longer an upper bound of  $S_1$ . This means that we can find a simple function g(x) in the form

 $g(x) = \sum_{i=1}^{k} c_i \mathbf{1}_{A_i}(x)$  for some positive constant  $c_i$  and subset  $A_i$  of  $\mathbb{Z}$ , such that  $0 \le g(x) \le f^+(x)$  for all x, and  $\int g(x) d\mu \ge s_1 - \epsilon$ . Because  $s_1$  is assumed to be finite,  $A_i$ 's are all finite sets.

From g, we can construct a larger function h by taking A to be the union of  $A_1, \ldots, A_k$  and let

 $h(x) = f(x)\mathbf{1}_A(x).$ 

Because the sample space  $\mathbb Z$  is discrete and A is finite, the function h(x) is a simple function. Thus

 $\int g(x) d\mu \leq \int h(x)d\mu = \sum_{x\in A} f^+(x)$ . We observe that  $\sum_{x\in A} f^+(x)$  is an element in  $S_2$ , the supremum  $s_2$  of  $S_2$  must be larger than or equal to  $\sum_{x\in A} f^+(x)$ , and hence larger than or equal to  $\int g(x) d\mu$ . This proves that  $s_2 \geq s_1 - \epsilon$ .

Since  $s_2 \ge s_1 - \epsilon$  holds for arbitrarily small positive real number  $\epsilon$ , we must have  $s_2 \ge s_1$ . This proves that  $s_2 = s_1$ .

The proof for the negative part is the same as above. This completes the proof.

- // There are two messages from this exercise.
- // The first message is that there is no "conditional convergence" in Lebesgue measure.
- // Summation like  $\sum_{n=1}^{\infty} (-1)^n/n$  does not exist in the Lebesgue sense.
- // See https://math.stackexchange.com/questions/1472173/no-conditional-convergence-in-lebesgue-integration
- // and https://math.stackexchange.com/questions/1095666/lebesgue-integral-and-absolute-value
- // The second one is the proof technique of "Give yourself an epsilon of room".
- // This is a common trick in real analysis

// See https://terrytao.wordpress.com/2009/02/28/tricks-wiki-give-yourself-anepsilon-of-room/

## Question 6

(Improper Riemann integral) In Lecture 10 we demonstrated the relationship between Riemann integral  $\int_a^b f(x) dx$  and Lebesgue integral  $\int_{[a,b]}^a f(x) d\lambda(x)$ . The purpose of this question is to consider the the case of improper Riemann integral.

Suppose that  $f: \mathbb{R} \to \mathbb{R}$  is a measurable function satisfying

- (i)  $\int_a^b f(x) dx$  is Riemann integrable for any a and b s.t.  $-\infty < a < b < \infty$ ;

(ii) The double limit  $\lim_{a\to-\infty}\lim_{b\to\infty}\int_a^b|f(x)|\ dx$  exists. The double limit above is usually denoted as  $\int_{-\infty}^{\infty}|f(x)|\ dx$ .

In this question we denote the Lebesgue measure on  $\mathbb{R}$  by  $\lambda$ .

(a) For positive integer n, define the truncated function  $g_n(x) \triangleq |f(x)| \mathbf{1}_{[-n,n]}(x)$ , where  $\mathbf{1}_{[-n,n]}$  is the indicator function of [-n,n]. From the second theorem in lecture 10, we obtain

$$\int_{[-n,n]} |g_n(x)| \ d\lambda(x) = \int_{-n}^n |f(x)| \ dx.$$

The LHS is the Lebesgue integral of  $|g_n(x)|$  on the integral [-n, n] w.r.t. the Lebesgue measure  $\lambda$ , and the RHS is the Riemann integral of |f(x)| from -n to

Verify that  $g_n$  is an increasing sequence of non-negative functions converging pointwise to |f|. Prove that  $\int_{\mathbb{R}} |f(x)| \ d\lambda(x) = \int_{-\infty}^{\infty} |f(x)| \ dx < \infty,$  and hence show that f(x) is  $\lambda$ -integrable.

$$\int_{\mathbb{R}} |f(x)| d\lambda(x) = \int_{-\infty}^{\infty} |f(x)| dx < \infty,$$

(Hint: Use monotone convergence theorem.)

(b) For positive integer n, define  $h_n(x) \triangleq f(x) \mathbf{1}_{[-n,n]}(x)$ . Verify that the functions  $h_n(x)$ 's are dominated by |f(x)| for all n, and converge to f(x) as n increases. By applying dominated convergence theorem, prove that

$$\int_{\mathbb{R}} f(x) \ d\lambda(x) = \int_{-\infty}^{\infty} f(x) dx.$$

#### Solution:

(a) The function  $g_n$  is increasing because it is a product of two nonnegative functions |f(x)| and  $\mathbf{1}_{[n,n]}$ , and  $(\mathbf{1}_{[n,n]})_{n\geq 1}$  is an increasing sequence of functions. It is assumed that the Riemann integral of  $g_n$  from -n to n exists for all n. Using Theorem 2 in lecture 10, we have

 $\int g_n(x) \ d\lambda(x) = \int_{[n,n]} g_n(x) \ d\lambda(x) = \int_{-n}^n g_n(x) \ dx$ . If we take limits as  $n \to \infty$ , the left-hand side approaches the Lebesuge integral  $\int |f(x)| \ d\lambda(x)$  by MCT. The right-hand side approaches  $\int_{-\infty}^{\infty} |f(x)| dx$ .

Since  $\int_{-\infty}^{\infty} |f(x)| dx$  is finite by assumption,  $\int |f(x)| d\lambda$  is also finite. Therefore f(x) is  $\lambda$ -integrable.

(b) For each n, the function  $h_n(x)$  is dominated by f(x) because

$$|h_n(x)| \le |f(x)| \cdot |\mathbf{1}_{[n,n]}(x)| \le |f(x)|.$$

For a given x,  $h_n(x)$  is equal to f(x) for all  $n \geq x$ , and hence converges to f(x) as  $n \to \infty$ .

By dominated convergence theorem,

- (1)  $\lim_n \int h_n(x) \ d\lambda = \int \lim_n h_n(x) \ d\lambda = \int f(x) \ d\lambda$ . Since  $h_n$  has finite support, by Theorem 2 in lecture 10,  $\int h_n(x) \ d\lambda(x) = \int_{[n,n]} h_n(x) \ d\lambda(x) = \int_{-n}^n h_n(x) \ dx$ .
- (2)  $\lim_n \int h_n(x) = \lim_n \int_n^n h_n(x) dx = \int_{-\infty}^{\infty} f(x) dx$ . (The second equality comes from the definition of improper Riemann integral.)

Putting (1) and (2) together, we get  $\int f(x) d\lambda = \int_{-\infty}^{\infty} f(x) dx$ .