#### Notation

We sometimes write  $A \in \aleph_0$  to mean "A is countable". and  $A \in \aleph_1$  to mean "A is uncountable".  $\chi \sim N(0,1)$  means  $\chi$  has a normal distribution with mean 0 and std 1.

(TBD) means to be done (NPR) means "needs peer review". If you are not the original author and agree the problem is correct, you can remove this tag.

#### 1.1.1 on midterm

We show that  $(\mathbb{R}, \mathcal{F}, P)$  is a probability space, where  $\mathcal{F} = \{A \in \mathbb{R} : A \in \aleph_0 \text{ or } A^c \in \aleph_0\}$  and P(A) = 0 if  $A \in \aleph_0$  and 1 if  $A^c \in \aleph_0$ . We must first show that  $\mathcal{F}$  is a  $\sigma$ -algebra.

- Let  $A \in \mathcal{F}$ . If  $A \in \aleph_0$ , then  $A^c \in \mathcal{F}$  since  $(A^c)^c = A \in \aleph_0$ . If  $A^c \in \aleph_0$ , then  $(A^c)^c = A \in \mathcal{F}$  since  $A^c \in \aleph_0$ .
- Let  $(A_i) \subset \mathcal{F}$  be a countable sequence of sets. Then either  $A = \bigcup A_i$  is countable or not. If A is countable, then  $A \in \mathcal{F}$ . If A is uncountable, then at least one  $A_k$  must be uncountable. But  $A_k \in F$ , so  $A_k^c \in \aleph_0$ . But then  $A^c = \bigcap A_i^c \subset A_k^c \in \aleph_0$ . Hence  $A \in F$ .

It remains to check that P is a probability measure.

- Let  $A \in \mathcal{F}$ . Then P(A) = 0 or 1, so  $P(A) \ge P(\emptyset) = 0$  since  $\emptyset \in \aleph_0$ .
- $P(\mathbb{R}) = 1$  since  $\mathbb{R}^c = \emptyset \in \aleph_0$ .
- Let  $(A_i) \subset \mathcal{F}$  be a countable sequence of disjoint sets. We must show  $P(\bigcup A_i) = \sum_i P(A_i)$ . Observe that if each  $A_i$  is countable, then  $\bigcup A_i$  is countable and so both sides of the equality are zero. We claim that in fact at most one  $A_i$  can be uncountable. In that case,  $\bigcup A_i$  is uncountable and  $\sum_i P(A_i) = 1$ , so the equality holds. Suppose  $A_i$  and  $A_j$  are two uncountable disjoint sets in  $\mathcal{F}$ . Then,

$$\mathbb{R} = \varnothing^c = (A_i \cap A_j)^c = A_i^c \cup A_j^c.$$

This is a contradiction, since  $A_i^c \cup A_j^c$  is countable, but  $\mathbb{R}$  is not.

### 1.1.2 on midterm

Let  $\mathcal{S}_d$  be the collection of sets of the form

$$(a_1, b_1] \times ... \times (a_d, b_d] \subset \mathbb{R}^d, -\infty \le a_i < b_i \le \infty$$

and in addition the empty set. We claim that  $\sigma(S_d) = \mathcal{R}^d$ , the Borel subsets of  $\mathbb{R}$ . First observe that if  $\mathcal{T}_d$  is the collection of sets of the form

$$(a_1, b_1) \times ... \times (a_d, b_d) \subset \mathbb{R}^d, -\infty \le a_i < b_i \le \infty$$

then  $\sigma(S_d) = \sigma(\mathcal{T}_d)$ , because if  $U = (a_1, b_1) \times ... \times (a_d, b_d) \in \mathcal{T}_d$  then there is a countable sequence of sets in  $S_d$  whose union is U: specifically sets of the form  $V_n = (a_1, b_1 - \frac{1}{n}] \times ... \times (a_d, b_d - \frac{1}{n}]$ . Similarly, since  $\sigma$ -algebras are also closed under countable intersection via de Morgan's laws, we can write any set of the form  $(a_1, b_1] \times ... \times (a_d, b_d]$  as the intersection of sets of the form  $(a_1, b_1 + \frac{1}{n}) \times ... \times (a_d, b_d + \frac{1}{n})$ .

It remains to show that  $\sigma(\mathcal{T}_d) = \mathcal{R}^d$ . Let A be any open subset of  $\mathbb{R}^d$ . Then let  $\mathcal{D}$  be the subcollection of  $\mathcal{T}_d$  consisting of sets contained in A. Then  $\bigcup_{B \in \mathcal{D}} B = A$ , since if  $x \in A$ , then there exists an open rectangle R containing x contained in A, since A is open, and R is a member of  $\mathcal{D}$ .

#### 1.1.3 on midterm

We show that the  $\sigma$ -alegbra  $\mathcal{R}^d$  of Borel sets in  $\mathbb{R}^d$  is countably generated. It suffices to provide a countable collection of sets in  $\mathcal{R}^d$  that generate any open subset of  $\mathbb{R}^d$ . Let Q be the following countable collection:

$$Q = \{B(x, 1/n) : x \in \mathbb{Q}^d, n \in \mathbb{N}\}\$$

where  $B(x, 1/n) := \{ y \in \mathbb{R}^d : |x - y| < 1/n \}$ . Let  $U \subset \mathbb{R}^d$  be any open set. Let  $Q' = Q \cap 2^U$  (i.e. the subcollection of Q consisting of subsets of U). We claim that

$$U = \bigcup_{V \in Q'} V$$

Clearly  $\bigcup_{V \in Q'} V \subset U$ . For the other inclusion, let  $x \in U$ . Then, since U is open, there exists a number  $N \in \mathbb{N}$  such that  $B(x, 1/N) \subset U$ . By the density of  $\mathbb{Q}^d \subset \mathbb{R}^d$ , there exists a point  $p \in \mathbb{Q}^d$  such that  $d(x, p) < \frac{1}{2N}$ . But then  $x \in B(p, \frac{1}{2N})$  and this ball is a member of Q' since if  $z \in U^c$  is arbitrary then

$$\frac{1}{N} < d(x,z) \le d(x,p) + d(p,z) < \frac{1}{2N} + d(p,z)$$

and hence  $d(p,z) > \frac{1}{2N}$ , implying  $d(p,U^c) = \inf_{z \in U^c} d(p,z) > \frac{1}{2N}$ . Thus  $x \in \bigcup_{V \in Q'} V$ .

#### 1.1.4 on midterm

- (i) Suppose  $F_1 \subset F_1 \subset ...$  are  $\sigma$ -algebras. We show  $F \coloneqq \bigcup F_i$  is a  $\sigma$ -algebra. Let  $A \in F$ . Then  $A \in F_k$  for some k. Hence  $A^c \in F_k \subset F$ , since  $F_k$  is a  $\sigma$ -algebra. If also  $B \in F$  then  $B \in F_j$  for some j. and we have  $A \in F_N$  for all  $N \ge k$  and  $B \in F_M$  for all  $M \ge j$ . In particular  $A \in F_s$  and  $B \in F_s$ , where  $s = \max j, k$  so  $A \cup B \in F_s$  since  $F_s$  is a  $\sigma$ -algebra. Hence  $A \cup B \in F$ , and we are done.
- (ii) If  $\Omega = \mathbb{N}$  and  $F_i := \sigma(\{1\}, \{2\}, ..., \{i\})$  then each  $F_i$  is a  $\sigma$ -algebra and we have  $F_i \uparrow F = \bigcup F_i$ , but F is not a  $\sigma$ -algebra. If  $U \in F$ , then  $U \in F_k$  for some k. But every set in  $F_k$  is either finite or has finite complement in  $\mathbb{N}$  (because union preserves finiteness or finite-complementness and because complement switches them, so you can't get anything else stating from a collection of finite sets). So every set in F is finite or has finite complement. But then F doesn't contain some countable unions of its own sets, such as  $\{2,4,6,\ldots\}$ .

# 1.1.5 (TBD)

#### 1.2.1 on midterm

Suppose X and Y are random variables on  $(\Omega, F, P)$  and let  $A \in F$ . We show that if  $Z(\omega) := X(\omega)$  for  $\omega \in A$  and  $Z(\omega) := Y(\omega)$  for  $\omega \in A^c$ , then Z is a random variable. Clearly Z maps

 $\Omega$  to  $\mathbb{R}$ . It remains to show that Z is F-measurable. Let B be a Borel subset of  $\mathbb{R}$ . Then:

$$Z^{-1}(B) = \{\omega \in \Omega : Z(\omega) \in B\}$$

$$= \{\omega \in A : Z(\omega) \in B\} \cup \{\omega \in A^c : Z(\omega) \in B\}$$

$$= \{\omega \in A : X(\omega) \in B\} \cup \{\omega \in A^c : Y(\omega) \in B\}$$

$$\subset \{\omega \in \Omega : X(\omega) \in B\} \cup \{\omega \in \Omega : Y(\omega) \in B\}$$

$$= X^{-1}(B) \cup Y^{-1}(B)$$

$$\in F,$$

as desired.

#### 1.2.2 on midterm

Suppose  $\chi \sim N(0,1)$ . Using Theorem 1.2.6, we have the estimate:

$$\left(\frac{1}{4} - \frac{1}{4^3}\right)e^{\frac{-4^2}{2}} \le \int_4^\infty e^{\frac{-y^2}{2}} \, dy \le \frac{1}{4}e^{\frac{-4^2}{2}}.$$

Hence:

$$\frac{1}{\sqrt{2\pi}} \left( \frac{1}{4} - \frac{1}{4^3} \right) e^{\frac{-4^2}{2}} \le P(\chi \ge 4) \le \frac{1}{4\sqrt{2\pi}} e^{\frac{-4^2}{2}}.$$

i.e.

$$3.13664 \cdot 10^{-5} \le P(\chi \ge 4) \le 3.34575 \cdot 10^{-5}$$

Note that the true value is  $P(\chi \ge 4) \approx 3.16712 \cdot 10^{-5}$ .

### 1.2.3 on midterm

We show that a distribution function has only countably many discontinuities. Suppose  $J \subset \mathbb{R}$  is the set of discontinuities of a distribution function F. Observe that, for each  $a \in J$ , we have  $F(a) - F(a-) = F(a) - \lim_{y \uparrow a} F(y) > 0$  since F is discontinuous at a and nondecreasing. Hence the open intervals  $(F(a-), F(a)) \subset \mathbb{R}$  have nonzero measure. The intervals are also disjoint, since F is nondecreasing. But then,

$$\{(F(a-), F(a)) : a \in J\}$$

is a collection of nonempty disjoint intervals in  $\mathbb{R}$ . We may (by the Axiom of Choice) define a function  $\phi: J \to \mathbb{Q}$  via  $\phi(a) = q_a$  where  $q_a$  is an arbitrary rational number in the interval (F(a-), F(a)). This function is injective because the intervals are disjoint. Hence the cardinality of J must be at most the cardinality of  $\mathbb{Q}$ . In particular J cannot be uncountable.

### 1.2.4 on midterm

We show that if  $F(x) = P(X \le x)$  is continuous, then Y := F(X) has a uniform distribution on (0,1). Since F is continuous, for each  $y \in (0,1)$  there exists by IVT an  $x \in \mathbb{R}$  such that F(x) = y. Define  $G: (0,1) \to \mathbb{R}$  via  $G(y) = \sup\{x \in \mathbb{R} : F(x) = y\}$ . Observe that:

- (a) G is strictly increasing,
- (b)  $G(F(x)) \ge x$  for all  $x \in \mathbb{R}$ , and

(c) 
$$F(G(y)) = y$$
 for all  $y \in (0, 1)$ .

Let  $y \in (0,1)$  and consider the sets:

$$A = \{\omega \in \Omega : G(F(X(\omega))) \le G(y)\}$$
$$B = \{\omega \in \Omega : X(\omega) \le G(y)\}.$$

We claim A = B as follows. Let  $\omega \in A$ . Then  $X(\omega) \leq G(F(X(\omega))) \leq G(y)$  by (b). Hence  $\omega \in B$ . On the other hand, if  $\omega \in B$ , then  $G(F(X(\omega))) \leq G(F(G(y)))$  since both F and G are nondecreasing. But G(F(G(y))) = G(y) by (c). Hence  $\omega \in A$ . Therefore

$$(*) \qquad P(G(F(X)) \le G(y)) = P(X \le G(y)).$$

Hence, we have for any  $y \in (0,1)$ :

$$P(Y \le y) = P(F(X) \le y)$$

$$= P(G(F(X)) \le G(y)) \quad \text{by (a)}$$

$$= P(X \le G(y)) \quad \text{by (*)}$$

$$= F(G(y))$$

$$= y \quad \text{by (c)}$$

as desired.

#### 1.2.5

Let  $X : \Omega \to \mathbb{R}$  be continuous with density f. Let  $g : \mathbb{R} \to \mathbb{R}$  be strictly increasing and differentiable on  $(a,b) \subset \mathbb{R}$ . We compute the density of g(X).

Let  $a, b \in \text{range}(g(X(\Omega))) \subset \mathbb{R}$ . Since g is strictly increasing, there are unique values  $g^{-1}(a)$  and  $g^{-1}(b)$  such that  $g(g^{-1}(a)) = a$  and  $g(g^{-1}(b)) = b$ . Thus consider the sets:

$$A = \{ \omega \in \Omega : a \le g(X(\omega)) \le b \}$$

$$A = \{\omega \in \Omega : g^{-1}(a) \le X(\omega) \le g^{-1}(b)\}$$

Since X has densite f, we have

$$P(a \le g(X) \le b) = P(A) = \int_{g^{-1}(a)}^{g^{-1}(b)} f(s) ds.$$

Note that  $s \in (g^{-1}(a), g^{-1}(b))$  and hence we may make the substitution  $s = g^{-1}(y)$ , since g is strictly increasing. Hence g(s) = y and we can rewrite the integral as:

$$P(a \le g(X) \le b) = \int_a^b \frac{f(g^{-1}(y))}{g'(g^{-1}(y))} \, dy$$

as desired.

#### 1.2.6 on midterm

Suppose X has a normal distribution with density f. We compute the density of  $Y := e^X$ . By 1.2.5, Y has distribution function

$$P(a < Y < b) = \int_a^b \frac{f(g^{-1}(y))}{g'(g^{-1}(y))} dy = \int_a^b \frac{f(\ln(y))}{e^{\ln(y)}} dy = \int_a^b \frac{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(\ln(y))^2}}{y} dy$$

Hence the density of Y is  $g(t) = \frac{1}{t\sqrt{2\pi}}e^{-\frac{1}{2}(\ln(t))^2}$ .

### 1.2.7 on midterm

Suppose X has the density function f. We compute the density of  $Y = X^2$ . If F is the CDF of Y then:

$$F(y) = P(Y \le y) = P(X^2 \le y) = P(|X| \le \sqrt{y}) = P(-\sqrt{y} < X < \sqrt{y}) = \int_{-\sqrt{y}}^{\sqrt{y}} f(s) \, ds$$

Observe that:

$$F'(y) = \frac{d}{dy} \left( \int_0^{\sqrt{y}} f(s) \, ds - \int_0^{-\sqrt{y}} f(s) \, ds \right) = f(\sqrt{y}) \frac{1}{2\sqrt{y}} - f(-\sqrt{y}) \frac{-1}{2\sqrt{y}} = \frac{f(\sqrt{y}) + f(-\sqrt{y})}{2\sqrt{y}}$$

and hence  $F(y) = \int_{-\infty}^{y} \frac{f(\sqrt{y}) + f(-\sqrt{y})}{2\sqrt{y}} dy$ .

In the case that X has a normal distribution, the density is

$$\frac{\frac{1}{\sqrt{2\pi}}e^{-\frac{1}{2}(\sqrt{y})^2} + \frac{1}{\sqrt{2\pi}}e^{-\frac{1}{2}(-\sqrt{y})^2}}{2\sqrt{y}} = \frac{\frac{1}{\sqrt{2\pi}}e^{-\frac{y}{2}} + \frac{1}{\sqrt{2\pi}}e^{-\frac{y}{2}}}{2\sqrt{y}} = \frac{\frac{2e^{-\frac{y}{2}}}{\sqrt{2\pi}}}{2\sqrt{y}} = \frac{e^{-\frac{y}{2}}}{\sqrt{2\pi y}}$$

#### 1.3.1 (TBD)

# 1.3.2 on midterm

We show that  $X_1 + X_2$  is a random variable if  $X_1$  and  $X_2$  are. Observe that

$$\{X_1 + X_2 < a\} = \{X_2 < a - X_1\} = \bigcup_{q \in \mathbb{Q}} \{X_2 < q < a - X_1\} = \bigcup_{q \in \mathbb{Q}} \{X_2 < q\} \cap \{q < a - X_1\}$$

but the latter is in  $\mathcal{F}$ , since the union is countable and since  $X_1$  and  $X_2$  are random variables.

### 1.3.3 on midterm

We show that if  $X_n \to X : \Omega \to \mathbb{R}$  almost surely and f is continuous on  $\mathbb{R}$ , then  $f(X_n) \to f(X)$  almost surely. Assume that Recall that  $X_n \to X$  a.s means  $P(\Omega_0) = 1$ , where

$$\Omega_0 = \{ \omega \in \Omega : \lim_{n \to \infty} X_n(\omega) \text{ exists} \}.$$

Let

$$\Omega_1 = \{ \omega \in \Omega : \lim_{n \to \infty} f(X_n(\omega)) \text{ exists} \}.$$

If  $\omega \in \Omega_0$ , then the limit  $X(\omega) := \lim_{n \to \infty} X_n(\omega)$  exists and so, by continuity, the limit

$$\lim_{x\to\infty} f(X_n(\omega)) = f(\lim_{n\to\infty} X_n(\omega)) = f(X(\omega))$$

exists. Hence  $\Omega_0 \subset \Omega_1$ . But then  $1 \geq P(\Omega_1) \geq P(\Omega_0) = 1$ , hence  $f(X_n) \to f(X)$  a.s.

## 1.3.4 (TBD) on midterm

in notebook

#### 1.3.5 on midterm

Let  $f: \mathbb{R} \to \mathbb{R}$  and let  $S_a = \{x \in \mathbb{R} : f(x) \leq a\}$ . We show that f is lower semicontinuous (that is:  $\liminf_{y\to x} f(y) \geq f(x)$  for each  $x \in \mathbb{R}$ ) iff  $S_a$  is closed for each  $a \in \mathbb{R}$ . Suppose first that f is l.s.c. and let  $a \in \mathbb{R}$ . Let  $(x_n) \subset S_a$  be a sequence such that  $x_n \to y$ . Hence for all n, we have  $f(x_n) \leq a$ . Then:

$$f(y) \le \liminf_{y \to x} f(y) = \liminf_{n \to \infty} f(x_n) \le a.$$

Hence  $a \in S_a$ , so  $S_a$  is closed.

Conversely, suppose  $S_a$  is closed for each  $a \in \mathbb{R}$ . In particular, fixing  $x \in \mathbb{R}$  and  $\epsilon > 0$ , we have that

$$S = x \in \mathbb{R} : f(x) \le f(x) - \epsilon$$

is closed. Now, observe that  $x \in S^c$  since  $f(x) > f(x) - \epsilon$ . Since  $S^c$  is open, there exists an  $N \in \mathbb{N}$  and a sequence  $y_k \to x$  such that  $y_k \in S^c$  for all  $k \ge N$ . Then if  $k \ge N$ , we have

$$f(y_k) > f(x) - \epsilon$$
 since  $y_k \in S^c$   
 $\inf_{k \ge N} f(y_k) > f(x) - \epsilon$  since  $f(x) - \epsilon$  is a lower bound  
 $\sup_{N} \inf_{k \ge N} f(y_k) \ge f(x)$  since  $\epsilon$  was arbitrary

as desired. Since x was arbitrary, f is l.s.c. on  $\mathbb{R}$ .

1.3.6 (TBD)

f

- 1.3.7 (TBD)
- 1.3.8 (TBD)
- 1.3.9 (TBD)

#### 1.4.1 on midterm

We show that if  $f \ge 0$  and  $\int f d\mu = 0$ , then f = 0 a.e. Let  $n \in \mathbb{N}$  and consider the set  $A_n = \{x \in \mathbb{R} : f(x) \ge \frac{1}{n}\}$ . Observe that  $A_n \uparrow A := \{x \in \mathbb{R} : f(x) > 0\}$ . Hence, by continuity of measure from below we have  $\mu(A_n) \uparrow \mu(A)$ . We will show below that  $\mu(A_n) = 0$ , implying  $\mu(A) = 0$ . This gives the result because  $\mu(A) = 0$  (together with  $f \ge 0$ ) means precisely that f = 0 except on a set of measure zero.

Observe that for each  $x \in \mathbb{R}$  we have  $f(x) \ge \frac{1}{n} \mathbb{I}_{A_n}(x)$ , where  $\mathbb{I}_{A_n}$  is the indicator function of  $A_n$ , since if  $x \in A_n$ , then  $f(x) \ge \frac{1}{n} \cdot 1 = \frac{1}{n} \mathbb{I}_{A_n}(x)$  and if  $x \notin A_n$ , then  $f(x) \ge 0 = \frac{1}{n} \mathbb{I}_{A_n}(x)$ . We then have by monotonicity:

$$0 = \int f \, d\mu \ge \int \frac{1}{n} \mathbb{I}_{A_n}(x) \, d\mu = \frac{1}{n} \mu(A_n) \ge 0,$$

which implies  $\mu(A_n) = 0$ , as desired.

## 1.4.2 (TBD) on midterm

## 1.4.3 (TBD)

#### 1.4.4 on midterm

We prove the Riemann-Lebesgue lemma, that is: if g is integrable then

$$\lim_{n\to\infty}\int g(x)\cos(nx)\,dx=0.$$

Suppose first that  $\phi$  is a step function with finite support. Then

$$\int \phi(x)\cos(nx) dx = \int \sum_{k=0}^{n} 1_{(a_k,b_k)}\cos(nx) dx$$

$$= \sum_{k=0}^{m} \int_{a_k}^{b_k} c_k \cos(nx) dx$$

$$= -\frac{1}{n} \sum_{k=0}^{m} c_k (\sin(nb_k) - \sin(na_k)) dx$$

$$\to 0 \quad \text{as } n \to \infty, \text{ since sin is bounded}$$

Now, by the previous exercise, if g is any integrable function, then there exists a step function  $\phi$  with finite support such that  $\int |g - \phi| dx \to 0$ . Hence we have:

$$\left| \int g(x) \cos(nx) \, dx \right| = \left| \int g(x) \cos(nx) + \phi(x) \cos(nx) - \phi(x) \cos(nx) \, dx \right|$$

$$\leq \left| \int (g(x) - \phi(x)) \cos(nx) \, dx \right| + \left| \int \phi(x) \cos(nx) \, dx \right|$$

$$\leq \int \left| (g(x) - \phi(x)) \right| \left| \cos(nx) \right| \, dx + \left| \int \phi(x) \cos(nx) \, dx \right|$$

$$\leq \int \left| (g(x) - \phi(x)) \right| \, dx + \left| \int \phi(x) \cos(nx) \, dx \right|$$

$$\Rightarrow 0$$

#### 1.5.1 on midterm

Suppose  $m \in \mathbb{R}$  such that  $\mu(A_m) = 0$  where  $A_m = \{x : |g(x)| > m\}$ . Then:

$$\int |fg| \, d\mu = \int_{A_m} |fg| \, d\mu + \int_{A_m^c} |fg| \, d\mu$$

$$= \int |f||g| 1_{A_m} \, d\mu + \int |f||g| 1_{A_m^c} \, d\mu$$

$$\leq 0 + \int |f| m \, d\mu$$

$$= m \|f\|_1$$

So  $\frac{1}{\|f\|_1} \int |fg| d\mu$  is a lower bound for  $\{m \in \mathbb{R} : \mu(\{x : |g(x)| > m\}) = 0\}$ , so it is also less than  $\|f\|_{\infty} := \inf\{m \in \mathbb{R} : \mu(\{x : |g(x)| > m\}) = 0\}$ , hence  $\int |fg| d\mu \le \|f\|_1 \|f\|_{\infty}$  as desired.

## 1.5.2 (TBD)

f

## 1.5.3 (TBD) on midterm

## 1.5.4 on midterm

Let f be integrable and let  $E_m$  be disjoint sets with union E. Then we show

$$\sum_{m=0}^{\infty} \int_{E_m} f \, d\mu = \int_{E} f \, d\mu$$

Let  $f_k := \sum_{m=0}^k f \cdot 1_{E_m}$ . Then  $f_k \to f \cdot 1_E$  a.e. since  $E_m$  are disjoint. Note that  $|f_n| \le f$  and f is integrable, so by DCT and linearity:

$$\sum_{m=0}^{\infty} \int_{E_m} f \, d\mu = \lim_{k \to \infty} \int f_k \, d\mu = \int f \cdot 1_E \, d\mu = \int_E f \, d\mu$$

as desired.

### 1.5.5 (TBD) on midterm

in notebook

## 1.5.6 (TBD) on midterm

## 1.5.7 on midterm

Let  $f \ge 0$ . We show that  $\int f \wedge n \, d\mu \uparrow \int f \, d\mu$  as  $n \to \infty$ . Observe that for  $f_n := f \wedge n$ , we have  $f_n \ge 0$  and  $f_n \uparrow f$  (since for each x,  $f(x) \le N$  for some  $N \in \mathbb{N}$ ). Hence by MCT,  $\int f_n \, d\mu \uparrow \int f \, d\mu$ , as desired.

Suppose  $g \ge 0$ . Let  $\epsilon > 0$ . Then by the above, there exists n large enough that  $|\int_A g - \int_A g \wedge n \, d\mu| < \epsilon/2$ . Let  $\delta = \frac{\epsilon}{2n}$  and suppose  $\mu(A) < \delta$ . Then:

$$\int_{A} g \, d\mu < \epsilon/2 + \int_{A} g \wedge n \, d\mu$$

$$\leq \epsilon + \int_{A} n \, d\mu$$

$$= \epsilon/2 + n\mu(A)$$

$$< \epsilon/2 + n\delta$$

$$= \epsilon$$

Now, if g is integrable, we have  $|g| = g^+ + g^- = g \vee 0 + (-g) \vee 0$ , which is nonnegative, and the result holds.

### 1.5.8 on midterm

Suppose f is integrable on [a,b] and define  $g(x) \coloneqq \int_a^x f(s) \, ds$ . We show that g is continuous on (a,b). Let  $x \in (a,b)$ . Let  $\epsilon > 0$ . Then, by 1.5.7, there exists a  $\delta > 0$  such that  $\int_x^{x+\delta} |f(s)| \, ds < \epsilon$ .

Hence, if  $x < y < x + \delta$ :

$$|g(y) - g(x)| = \left| \int_{a}^{y} f(s) \, ds - \int_{a}^{x} f(s) \, ds \right|$$

$$= \left| \int_{x}^{y} f(s) \, ds \right|$$

$$\leq \int_{x}^{y} |f(s)| \, ds$$

$$\leq \int_{x}^{x+\delta} |f(s)| \, ds$$

$$< \epsilon$$

and we are done.

- 1.5.9 (TBD)
- 1.5.10 (TBD)