

# 1 Measure Theory Continued

**Notation:**  $\mathcal{B} := \sigma\{\text{open sets of } S^{\text{topo}}\}$ .

For  $f : S_1 \rightarrow S_2$ , have pullback  $f^{-1} : S_2 \rightarrow S_1$

(a)  $f^{-1}$  commutes with finite Boolean operations and monotone limits, i.e.

$$f^{-1}(B_1 \cup B_2) = f^{-1}(B_1) \cup f^{-1}(B_2) \quad (1.1)$$

$$f^{-1}(B_1 \cap B_2) = f^{-1}(B_1) \cap f^{-1}(B_2) \quad (1.2)$$

$$B_n \uparrow B \implies f^{-1}(B_n) \uparrow f^{-1}(B) \quad (1.3)$$

(b) Given  $S_2$ ,  $\{f^{-1}(B) : B \in S_2\}$  is a  $\sigma$ -field: "the pullback of a  $\sigma$ -field is a  $\sigma$ -field."

**Definition 1.1.** A function  $f : S_1 \rightarrow S_2$  between two measurable spaces is *measurable* if  $f^{-1}(B) \in \mathcal{S}_1$  for all  $B \in S_2$ .

**Lemma 1.2.**  $f$  is measurable if  $f^{-1}(B) \in \mathcal{S}_1$  for all  $B \in \mathcal{B}$  such that  $S_2 = \sigma(\mathcal{B})$ .

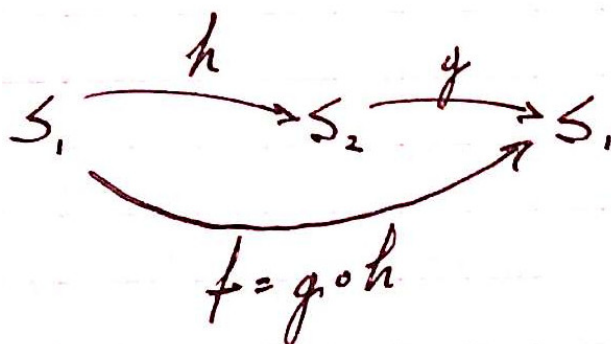
*Proof.*  $\{B \in S_2 : f^{-1}(B) \in \mathcal{S}_1\} \supset \sigma(\mathcal{B})$  is a  $\sigma$ -field by commutativity of  $f^{-1}$  wrt Boolean operations. It also  $\supset \mathcal{B}$ .  $\square$

**Lemma 1.3.**  $f^{\text{cts}} : S_1^{\text{topo}} \rightarrow S_2^{\text{topo}}$  is measurable (i.e.  $\text{cts} \implies \text{meas}$ )

*Proof.*  $\text{cts} \implies f^{-1}(G_2^{\text{open}}) \in \mathcal{S}_1^{\text{open}} \supset \mathcal{S}_1$ , where  $\mathcal{S}_1$  is the Borel  $\sigma$ -algebra on  $S_1$ . The previous lemma implies  $f$  is measurable wrt  $\sigma\{\mathcal{S}_1^{\text{open}}\} = \mathcal{S}_1$ .  $\square$

**Lemma 1.4** ( $\pi$ -system sufficiency). If  $S_2 = \mathbb{R}$ , it suffices to check  $f^{-1}((-\infty, x]) \in \mathcal{S}_1$  for all  $x \in \mathbb{R}$ .

*Proof.*  $\sigma\{(-\infty, x] : x \in \mathbb{R}\} = \sigma(\mathbb{R}) = S_2$   $\square$



**Lemma 1.5** (Composition). If  $h$  and  $g$  are measurable, then  $f = g \circ h$  is measurable.

**Lemma 1.6** (Multi-input composition). Suppose  $\{f_i : (S, \mathcal{S}) \rightarrow \mathbb{R}\}_{i=1}^d$  are measurable and  $g : \mathbb{R}^d \rightarrow \mathbb{R}$  is measurable. Then  $g(f_1(s), f_2(s), \dots, f_d(s))$  is a measurable function  $\mathcal{S}_1 \rightarrow \mathbb{R}$ .

*Proof.* Apply lemma 1.5 to  $(S, \mathbb{R}^d, \mathbb{R})$  and  $h(s_1) = [f_1(s_1) \ f_2(s_1) \ \dots \ f_d(s_1)]$ . Suffices to show  $h : S \rightarrow \mathbb{R}^d$  measurable.

Use fact that  $\mathcal{B}^d = \text{Borel } \sigma\text{-field on } \mathbb{R}^d = \sigma\text{-field generated by } \left\{ \prod_{i=1}^d (-\infty, \hat{x}_i] : \hat{x} \in \mathbb{R}^d \right\}$ .

Then

$$h^{-1} \left( \prod_{i=1}^d (-\infty, x_i] \right) = \bigcap_{i=1}^d \{s_1 : f_i(s_1) = x_i\} \in \mathcal{S}_1 \quad (1.4)$$

and by lemma 1.4 we are done.  $\square$

**Corollary 1.7.**  $\{f_i : S \rightarrow \mathbb{R}\}$  measurable, then  $f_1 + f_2$ ,  $f_1 \cdot f_2$ , and  $\max\{f_1, f_2\}$  are measurable.

*Proof.*  $g(x_1, x_2) = x_1 + x_2$ ,  $x_1 \cdot x_2$ , and  $\max\{x_1, x_2\}$  are all continuous hence measurable. Applying lemma 1.6 with  $\{f_i\}$  and  $g$  shows that the composition is measurable.  $\square$

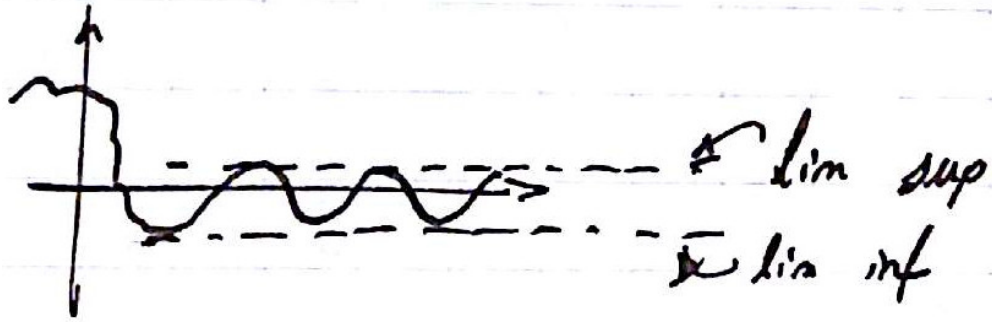
This is *very important*, make sure to grok the following definition:

**Definition 1.8.** For arbitrary  $x_n \in \bar{\mathbb{R}}$ ,  $n \in \mathbb{N}$ , define

$$\limsup_n x_n := \lim_{N \uparrow \infty} \sup_{n \geq N} x_n = \inf_{N \geq 1} \sup_{n \geq N} x_n \in \bar{\mathbb{R}} \quad (1.5)$$

$$\liminf_n x_n := \lim_{N \uparrow \infty} \inf_{n \geq N} x_n = \sup_{N \geq 1} \inf_{n \geq N} x_n \in \bar{\mathbb{R}} \quad (1.6)$$

$$(1.7)$$



Note that both  $\limsup$  and  $\liminf$  exist  $\in \bar{\mathbb{R}}$ , regardless of whether  $\lim_n x_n$  does, and  $\limsup \geq \lim \geq \liminf$ .

These definitions may be generalized to ascending and descending sequences of sets, where  $\sup$  is taken to be  $\cup$  and  $\inf$  as  $\cap$ .

**Lemma 1.9.** Given measurable functions  $\{f_i : S \rightarrow \bar{\mathbb{R}}\}_{i=1}^\infty$ , define  $f^*(s) = \limsup_n f_n(s)$  and  $f_*(s) = \liminf_n f_n(s)$ . Then  $f^*$  and  $f_*$  are measurable functions  $S \rightarrow \bar{\mathbb{R}}$ .

*Proof.*

$$\{s : \limsup_n f_n(s) \leq x\} = \{s : f_n(s) \leq x + 1/i \text{ ult. } \forall i \in \mathbb{N}\} \quad (1.8)$$

$$= \bigcap_{i=1}^{\infty} \{s : f_n(s) \leq x + 1/i \text{ ult.}\} \quad (1.9)$$

$$= \bigcap_{i=1}^{\infty} \bigcup_{N=1}^{\infty} \{s : f_n(s) \leq x + 1/i \text{ } \forall n \geq N\} \quad (1.10)$$

$$= \bigcap_{i=1}^{\infty} \bigcup_{N=1}^{\infty} \underbrace{\bigcap_{n=N}^{\infty} \{s : f_n(s) \leq x + 1/i\}}_{\in \mathcal{S}} \quad (1.11)$$

so  $f^*$  measurable. □

## 2 On $\mathbb{R}$ -valued measurable functions $(S, \mathcal{S}) \rightarrow \mathbb{R}$

**Definition 2.1.** For  $A \in \mathcal{S}$ , the *indicator function*  $1_A(s) = \begin{cases} 1, & \text{if } s \in A \\ 0, & \text{otherwise} \end{cases}$

Let  $\vec{c} \in \mathbb{R}^n$  and  $\{A_i\}_1^n$  be a partition of  $S$  into measurable sets.  $f : (S, \mathcal{S}) \rightarrow \mathbb{R}$  is a *simple function* if  $f(s) = \sum_i \underbrace{c_i 1_{A_i}}_{\text{step function on } A_i} (= c_i \text{ for } s \in A_i)$ .

**Lemma 2.2.** Let  $h^{meas} : S \rightarrow [0, L]$ . For  $i \geq 1$ , define

$$h_i(s) = \max_{j \geq 0} \left\{ \frac{j}{2^i} : \frac{j}{2^i} \leq h(s) \right\} = 2^{-i} \lfloor 2^i h(s) \rfloor \leq h(s) \quad (2.1)$$

Then  $h_i(s) \uparrow h(s)$  and each  $h_i$  is a simple function.

**Exercise 2.3.** Prove this.

## 3 Measures

$(S, \mathcal{S})$  a measurable space.

**Definition 3.1.** A *measure* is a function  $\mu : \mathcal{S} \rightarrow [0, \infty]$  such that

(a)  $\mu(\emptyset) = 0$

(b) (Countable additivity) For countable disjoint  $A_i \in \mathcal{S}$ ,  $\mu(\cup_i A_i) = \sum_i \mu(A_i) \leq \infty$

**Definition 3.2.**  $\mu$  is a *probability measure* if in addition  $\mu(S) = 1$ .

$\mu(S) < \infty$  is a *finite measure*.

If  $\exists S_n \uparrow S$  s.t.  $\mu(S_n) < \infty$  for all  $n$ , then  $\mu$  is a  $\sigma$ -finite measure

### 3.1 Elementary Properties

- If  $A \subset B$ , then  $\mu(A) \leq \mu(B)$
- If  $\mu(A \cup B) \leq \mu(A) + \mu(B)$ , with equality if  $A \cap B = \emptyset$
- For probability measures,  $\mu(A^c) = 1 - \mu(A)$
- (Monotonicity)  $A_n \uparrow A \implies \mu(A_n) \uparrow \mu(A)$ .  
 $A_n \downarrow A$  and some  $\mu(A_n) < \infty \implies \mu(A_n) \downarrow \mu(A)$
- (Continuity)  $A_n \downarrow \emptyset$ ,  $\exists n : \mu(A_n) < \infty$ , then  $\mu(A_n) \downarrow 0$ .