# Chapter 3

# Notes on Chapter 3

### Proposition 1.1

Prove two things,

- $1. \ \limsup_{r \to R} \phi(r) = \lim_{\varepsilon \to 0} \sup_{0 < |r-R| < \varepsilon} \phi(r) = \inf_{\varepsilon > 0} \sup_{0 < |r-R| < \varepsilon} \phi(r),$
- 2.  $\lim_{r\to R} \phi(r) = c \iff \limsup_{r\to R} |\phi(r) c| = 0$

### Folland Real Analysis

### Proposition 1.2

If  $U\subseteq B(1,0)=\{|x|<1\}$ , and  $U\in\mathbb{B}$ , and if m(U)>0, then the family of sets

$$E_r = \left\{ x + ry, \ y \in U \right\}$$

shrinks nicely to  $x \in \mathbb{R}^n$ .

*Proof.* Let r > 0 be fixed then  $\forall z \in E_r \hookrightarrow z = x + ry$ . Hence,

$$d(x,z) = d(x, x + ry)$$
$$= |r|d(0, y) < |r|$$

by translation invariance.

#### Definition 1.1: Signed measure

Let  $\mathcal{M}$  be a  $\sigma$ -algebra and  $v: \mathcal{M} \to [-\infty, +\infty]$  be a set function on  $\mathcal{M}$ . It is a *signed measure* on  $\mathcal{M}$  if

- $v(\varnothing) = 0$ ,
- $\nu$  assumes at most one of the values  $\pm \infty$ ,
- If  $\{E_j\}_{j\geq 1}$  is a countable, disjoint sequence of sets, the expression

$$\sum_{i\geq 1} \nu(E_i)$$
 is unambiguous, and is equal to  $\nu(\bigcup E_i)$ 

More precisely,

- if  $|\nu(\bigcup E_j)| < +\infty$ , the series  $\sum \nu(E_j)$  converges absolutely,
- if  $\nu(\bigcup E_i) = \pm \infty$ , the series  $\sum \nu(E_i)$  diverges to  $\pm \infty$  on every permutation.

#### Definition 1.2: Positive, negative, null sets

Let  $\nu$  be a signed measure on  $\mathcal{M}$ . A measurable set  $E \in \mathcal{M}$  is called *positive* (resp. negative, null) if every measurable subset  $F \subseteq E$  satisfies  $\nu(F) \ge 0$  (resp.  $\nu(F) \le 0$ ,  $\nu(F) = 0$ ).

#### Definition 1.3: Mutual singularity

Two signed measures,  $\nu$  and  $\mu$  on a common  $\sigma$ -algebra  $\mathcal{M}$  are *mutually singular*, denoted by  $\nu \perp \mu$  if there exists disjoint, measurable sets E, F whose union is  $\mathbf{X}$ .

 $\mu$  is null on E, and  $\nu$  is null on F

#### Proposition 1.3

Let v be a signed measure on  $(X, \mathcal{M})$ . If  $\{E_j\}$  is an increasing sequence in  $\mathcal{M}$ ,  $\lim_{n\to+\infty} v(E_j) = v(\bigcup E_j)$ . If  $\{E_j\}$  is a decreasing sequence in  $\mathcal{M}$ ,  $\lim_{n\to+\infty} v(E_j) = v(\bigcap E_j)$  provided  $v(E_1)$  is of finite measure.

*Proof.* Let v be a signed measure, and fix any increasing sequence  $E_j \nearrow E = \bigcup E_{j \ge 1}$  of sets. This induces a disjoint sequence in  $\{F_n\}$ . Define  $F_1 = E_1$ , and if  $n \ge 2$ ,

$$F_n = E_n \setminus \bigcup E_{j \le n-1}$$

Use  $\sigma$ -additivity of  $\nu$ , where the sum is 'defined' to be non-ambiguous.

For the second part of the proof, notice if  $A \subseteq B$  are measurable sets, if  $v(A) = \pm \infty$ , then  $v(B) = \pm \infty$ , because of the second property of v. Indeed,

$$v(B) = v(A) + v(B \setminus A) = \pm \infty + c$$

where  $c \in \mathbb{R} \cup \{\pm \infty\}$ . Therefore v(B) = v(A). By assumption  $v(E_1) \in \mathbb{R}$ , the contrapositive of the previous argument shows that the intersection  $\cap E_j$  is of finite measure as well. We can produce an increasing sequence  $G_n = E_1 \setminus E_n$  for  $n \in \mathbb{N}^+$ . Then

$$\bigcup G_n = \bigcup E_1 \setminus E_n = E_1 \cap \left[\bigcup E_n^c\right] = \left[\bigcap E_j\right]^c$$

We then write

$$E_1 = \left[\bigcup G_n\right] + \left[\bigcap E_n\right]$$

The finiteness of  $\nu(E_1)$  on the left hand side implies all the terms in the union converge absolutely. Therefore

$$v(E_1) - v(\bigcap E_n) = \lim_{n \to +\infty} v(G_n)$$
$$= \lim_{n \to +\infty} v(E_1) - v(E_n)$$
$$= v(E_1) - \lim_{n \to +\infty} v(E_n)$$

Cancelling terms finishes the proof.

### Folland Real Analysis

### Proposition 1.4

Any measurable subset of a positive set is again positive, and any countable union of positive sets is again positive. Similarly for negative, and null sets.

Proof. Trivial.

#### Proposition 1.5: Hahn Decomposition Theorem

Let  $\nu$  be a signed measure on the measurable space  $(\mathbf{X}, \mathcal{M})$ , then there exists positive and negative sets  $P, N \in \mathcal{M}$  where  $P \cup N = \mathbf{X}$ , and  $P \cap N = \emptyset$ . If P' and N' are another such decomposition,

$$P\Delta P' = N\Delta N'$$
 is  $\nu$ -null.

*Proof.* There are multiple steps to this proof. Suppose  $\nu$  does not attain  $+\infty$ . Define

$$m = \sup \{ v(P), P \text{ is a positive set} \}$$

By assumption  $m < +\infty$ , let  $\{P_j\}$  be a sequence of positive sets with  $v(P_j) \nearrow m$ . We claim the supremum is attained. Indeed, if  $P \stackrel{\triangle}{=} \cup P_j$ , then P is a positive set as well, by monotonicity  $v(P) \ge v(P_j)$ , taking the supremum on both sides reads v(P) = m.

Wanting to prove  $N \stackrel{\Delta}{=} \mathbf{X} \setminus P$  is a  $\nu$ -negative set,

• Clearly N cannot contain any positive sets  $A \subseteq N$  with a non-null measure, since

$$v(A) > 0 \implies v(A) + v(P) = v(A+P) > m$$

contradicting the supremum.

• Let us examine the properties of subsets of N with positive measure. Call this set  $A \subseteq N$ , where  $\nu(A) > 0$ .

The previous bullet point tells us A cannot be a v-positive set. There exists a  $B \subseteq A$  of strictly negative measure,

$$v(A \setminus B) + v(B) = v(A) \implies v(A \setminus B) > v(A)$$

Notice the assumption  $\nu$  does not attain  $+\infty$  allows us to subtract B over. Summarizing,

existence of subset of positive measure  $\implies$  subset with even greater positive measure

We will use the above inductively to construct a measurable subset of N, that is 'small' but has 'large' positive measure at the same time.

• Suppose N is not v-negative, so it admits a set of positive measure in  $A_1 \subseteq N$ .

Let  $n_1 = \text{least}\left\{n \in \mathbb{N}^+, \exists B \subseteq A_1, \ v(B) > v(A) + n^{-1}\right\}$ , since  $n_1$  is attained, it corresponds to some  $A_2 \subseteq A_1$  with  $v(A_2) > v(A_1) + n_1^{-1}$ .

Repeating this process inductively, we see

$$v(A_k) > v(A_{k-1}) + n_k^{-1}$$

Let  $A = \bigcap A_k$ , this should be a set of large positive measure. A simple induction will show

$$v(A_k) > v(A_1) + \sum_{j=1}^k n_j^{-1} > \sum_{j=1}^k n_j^{-1}$$

However,  $v(A) < +\infty$  by assumption. Upon taking limits and using the estimate above,

$$\sum_{j\geq 1} n_j^{-1} = \lim_{n\to\infty} \nu(A_n) = \nu(A) < +\infty$$

The sum on the left is finite, so its terms must converge to 0. Notice v(A) is a subset of N of positive measure, it admits a subset  $B \subseteq A$  with  $v(B) > v(A) + n^{-1}$  for  $n \ge 1$ .

 $n_j^{-1} \to 0$  implies  $n_j \to \infty$ . So  $n < n_j$  for large j. Notice  $B \subseteq A \subseteq A_j$ , and  $v(B) > v(A_j) + n^{-1}$ . This contradicts our definition of  $n_j$ , stated below for convenience

$$n_j = \operatorname{least}\left\{n \in \mathbb{N}^+, \exists B \subseteq A_j, v(B) > v(A_j) + n^{-1}\right\}$$

This proves N is  $\nu$ -negative.

To show this composition is  $\nu$ -unique, let P' and N' be disjoint, measurable positive and negative sets of X. Then

$$P \setminus P' \subseteq P$$
 and  $P \setminus P' \setminus \mathbf{X} \setminus P' \subseteq N'$ 

So  $P \setminus P'$  is at the same time a  $\nu$ -positive and a  $\nu$ -negative set, hence it is  $\nu$ -null by Lemma 3.2.

Finally, the case for when  $\nu$  attains  $+\infty$  can be handled if we consider  $-\nu$ . P is positive for  $-\nu$  iff it is negative for  $\nu$ , and similarly for N. Relabelling P and N finishes the proof.

Proposition 1.6

### Proposition 1.7

Proposition 1.8

Proposition 1.9

Proposition 1.10

### Proposition 1.11

Proposition 1.12

Proposition 1.13

Proposition 1.14

Proposition 1.15

Proposition 1.16

Proposition 1.17

Proposition 1.18

#### Proposition 1.19

Let the maximal function of any measurable  $f \in \mathbb{B}_{\mathbb{R}^n}$  be denoted by Hf(x), more precisely,

$$Hf(x) = \sup_{r>0} A_r |f|(x) = \sup_{r>0} \frac{1}{m(B(r,x))} \int_{B(r,x)} f(y) dy$$

where  $A_r|f|$  is the average of |f| on a ball with radius r>0 centered at  $x\in\mathbb{R}^n$ . In symbols,

$$A_r|f| = \frac{1}{m(B(r,x))} \int_{B(r,x)} f(y) dy$$

The maximal theorem makes two claims:

- 1.  $(Hf)^{-1}((\alpha,+\infty)) = \bigcup_{r>0} (A_r|f|)^{-1}((\alpha,+\infty))$ , and Hf is measurable for every  $f \in L^1_{loc}$ .
- 2. There exists a C > 0, for every  $f \in L^1$

$$m(\{Hf(x) > \alpha\}) \le \frac{C}{\alpha} \|f\|_1$$

for every  $\alpha > 0$ .

*Proof.* Let  $\alpha > 0$  and fix  $z \in (Hf)^{-1}((\alpha, +\infty))$ , so  $Hf(z) > \alpha$  and

$$\sup_{r>0} A_r |f|(z) > \alpha$$

and with  $Hf(z) - \alpha > 0$ , we get some  $r_0 > 0$ 

$$Hf(z)-(Hf(z)-\alpha)=\alpha < A_{r_0}|f|(z) \implies z \in \bigcup_{r>0} (A_r|f|)^{-1}((\alpha,+\infty))$$

Next, let  $z \in \bigcup_{r>0} (A_r|f|)^{-1}((\alpha,+\infty))$ , it is clear that

$$Hf(z) \geq A_{r_0}|f|(z) > \alpha$$

for some  $r_0 > 0$ . Since  $A_r|f|$  (a function indexed by r > 0) is continuous in  $x \in \mathbb{R}^n$ ,  $(A_r|f|)^{-1}((\alpha, +\infty))$  is open, and Hf is measurable.

The second claim is slightly more intricate than the first. Define

$$E_{\alpha} = \left\{ Hf > \alpha \right\} = \bigcup_{r > 0} \left\{ A_r |f| > \alpha \right\}$$

Let  $x \in E_{\alpha}$ , this induces a  $r_x > 0$  where  $x \in \left\{A_{r_x} | f | > \alpha \right\}$ . Rearranging gives

$$\left(\frac{1}{\alpha} \int\limits_{B(r,x)} |f| dz\right) < m(B(r,x))$$

We wish to apply Theorem 3.15 to this family of open balls. Notice

- Each  $x \in E_{\alpha} \hookrightarrow r_x > 0 \hookrightarrow A_{r_x}|f|$ ,
- If  $U = \bigcup_{x \in E_{\alpha}} B(r_x, x)$ , then  $E_{\alpha} \subseteq U$ ,
- Choose  $c < m(E_{\alpha}) \le m(U)$  (by monotonicity) arbitrarily,
- By Theorem 3.15, there exists a finite disjoint subcollection of points indexed by

$$x_1,\ldots,x_N\in E_\alpha$$

so that  $\bigsqcup_{j\leq N} B(r_{x_i},x_j) = U \supseteq E_{\alpha}$ , and  $c < 3^n \sum_{j\leq k} m(B_j)$ 

• Define  $B_j = B(r_{x_i}, x_j)$  for all  $j \le k$ , and

$$m(B_j) < \frac{1}{\alpha} \cdot \int_{B_j} |f| dz$$

by finite additivity,

$$c3^{-n} < \sum_{j \le k} m(B_j) < \frac{1}{\alpha} \cdot \sum_{j \le k} \int_{B_j} |f| dz$$

and finally

$$c < \frac{3^n}{\alpha} \sum_{j \le k} \int_{B_j} |f| dz \le \frac{3^n}{\alpha} \|f\|_1$$

• By inner regularity, of m on  $\mathbb{B}$ , since

$$m(E_{\alpha}) = \sup \left\{ m(K), K \subseteq \mathbb{R}^{n}, \text{ compact. } K \subseteq E_{\alpha} \right\}$$

for any compact  $K, K \subseteq E_{\alpha}$ , we have  $m(K) < +\infty, m(K) \le m(E_{\alpha})$  and

$$m(K) = c < \frac{3^n}{\alpha} \|f\|_1 \implies m(E_\alpha) \le \frac{3^n}{\alpha} \|f\|_1$$

#### Remark 1.1

We used the properties of a Radon Measure here, without relying on the phrase 'sending  $c \to E_{\alpha}$ ', which would require us to deal with two cases  $m(E_{\alpha}) < +\infty$  and  $m(E_{\alpha}) = +\infty$ .

Proposition 1.20

Proposition 1.21

Proposition 1.22

#### Proposition 1.23

The Lebesgue Differentiation Theorem. Suppose  $f \in L^1_{loc}$ , and for every  $x \in \mathcal{L}_f$ , (so that  $x \in \mathbb{R}^n$  a.e). We have

1. 
$$\lim_{r\to 0} \frac{1}{m(E_r)} \int_{E_r} |f(y) - f(x)| dy = 0$$
,

2. 
$$\lim_{r\to 0} \frac{1}{m(E_r)} \int_{E_r} f(y) dy = f(x),$$

For every family  $\{E_r\}_{r>0}$  that shrinks nicely to  $x \in \mathbb{R}^{n'}$ .

*Proof.* Since the family  $\{E_r\}_{r>0}$  shrinks nicely, we have

$$m(E_r) \ge m(B(r, x)) \implies m(E_r) > \alpha \cdot m(B(r, x))$$

for some  $\alpha > 0$ , independent on r. Rearranging gives

$$m^{-1}(E_r) < \alpha^{-1} m^{-1}(B(r, x))$$

And monotonicity of the integral

$$\int_{E_r} |f(y) - f(x)| \, dy \le \int_{B(r,x)} |f(y) - f(x)| \, dy$$

Combining the last two results, for every  $\varepsilon > 0$ , if  $0 < r < \varepsilon$ , then

$$m^{-1}(E_r) \int_{E_r} |f(y) - f(x)| dy \le m^{-1} B(r, x) \int_{B(r, x)} |f(y) - f(x)| dy$$

Taking the supremum on both sides,

$$\sup_{0 < r < \varepsilon} m^{-1}(E_r) \int_{E_r} |f(y) - f(x)| dy \le \sup_{0 < r < \varepsilon} m^{-1} B(r, x) \int_{B(r, x)} |f(y) - f(x)| dy$$

and sending  $\varepsilon \to 0$ , proves the first claim. The second claim is immediate upon applying the  $L^1$  inequality.

Fix any  $\varepsilon > 0$ , and

$$\lim_{r \to 0} m^{-1}(E_r) \int_{E_r} f(y) dy = f(x) \iff \lim_{\varepsilon \to 0} \sup_{0 < r < \varepsilon} \left| m^{-1}(E_r) \int_{E_r} f(y) dy - f(x) \right|$$

$$\iff \lim_{\varepsilon \to 0} \sup_{0 < r < \varepsilon} \left| m^{-1}(E_r) \int_{E_r} [f(y) - f(x)] dy \right|$$

$$\leq \lim_{\varepsilon \to 0} \sup_{0 < r < \varepsilon} m^{-1}(E_r) \int_{E_r} |f(y) - f(x)| dy$$

$$= \lim_{r \to 0} m^{-1}(E_r) \int_{E_r} |f(y) - f(x)| dy$$

$$= 0$$

Updated 2023-12-02 at 00:25

Proposition 1.24

Proposition 1.25

Proposition 1.26

Proposition 1.27

Proposition 1.28

Proposition 1.29

Proposition 1.30

Proposition 1.31

Proposition 1.32

Proposition 1.33

Proposition 1.34

Proposition 1.35

Proposition 1.36

Proposition 1.37

Proposition 1.38