Mathematical Analysis II

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	Fubini's theorem is also useful when computing integrals.	

Example 0.1 (Gauss integral)

Recall the Gauss integral:

$$\int_0^\infty e^{-x^2} \, \mathrm{d}x = \frac{\sqrt{\pi}}{2}.$$

Here we give a different proof:

$$\int e^{-x^2} dx \int e^{-y^2} dy = \iint_{\mathbb{R}^2} e^{-(x^2 + y^2)} dx dy$$
$$= \int_0^{+\infty} e^{-r^2} dr^2 \cdot \pi = \pi.$$

§1 Lebesgue differentiation

The most important theorem in calculus is no doubt the Fundamental theorem of Calculus (which is also called Newton-Lebniz formula). Since we generalized the integrals, there must be a generalized version of this theorem:

Theorem 1.1 (Lebesgue differentiation theorem, part 1)

If f is integrable on \mathbb{R}^d , for any ball $B \subset \mathbb{R}^d$, we have

$$\lim_{x \in B, |B| \to 0} \frac{1}{m(B)} \int_B f(y) \, \mathrm{d}y = f(x), a.e.$$

This theorem clearly holds for continuous points of f.

Our basic idea is to take a continuous g, such that $||g - f||_{\mathcal{L}^1} < \varepsilon$. and to prove

$$\left\{ x: \limsup_{x \in B, |B| \to 0} \frac{1}{m(B)} \int_{B} |f(y) - f(x)| \, \mathrm{d}y \ge \varepsilon_0 \right\}$$

is a null set.

Now we estimate

$$\frac{1}{m(B)} \int_{B} |f(y) - f(x)| \, \mathrm{d}y \le \frac{1}{m(B)} \int_{B} (|f(y) - g(y)| + |g(y) - g(x)| + |g(x) - f(x)|) \, \mathrm{d}y$$

$$= |f(x) - g(x)| + \varepsilon + \frac{1}{m(B)} \int_{B} |f(y) - g(y)| \, \mathrm{d}y$$

We find that the last term is pretty hard to deal with, so we'll introduce some new tools:

Definition 1.2 (Hardy-Littlewood maximal function). Let f be an integrable function on \mathbb{R}^d . Define

$$Mf(x) = \sup_{x \in B} \frac{1}{m(B)} \int_{B} |f(y)| \, \mathrm{d}y.$$

to be the **maximal function** of f.

Theorem 1.3 (Hardy-Littlewood)

The maximal function Mf satisfies:

- ullet Mf is measurable;
- For x almost everywhere, $|f(x)| \leq Mf(x) < +\infty$.
- \bullet There exists a constant C s.t.

$$|\{x: Mf > \alpha\}| \le \frac{C}{\alpha} ||f||_{\mathcal{L}^1}.$$

Proof. First we prove $\{Mf > \alpha\}$ is measurable. If $Mf(x_0) > \alpha$, then exists an open ball $B \ni x_0$,

$$\int_{B} |f(y)| \, \mathrm{d}y > \alpha m(B).$$

This implies that $B \subset \{Mf > \alpha\} \implies \{Mf > \alpha\}$ is an open set.

For the second part, we'll prove for $\forall \varepsilon_0 > 0, N > 0$,

$$m({x : M f(x) + \varepsilon_0 < |f(x)| < N}) = 0.$$

Otherwise denote the above set as E, for $\forall 0 < \lambda < 1$, $\exists B \text{ s.t. } |E \cap B| > \lambda |B|$. Thus for $x \in E$,

$$Mf(x) \ge \frac{1}{m(B)} \int_{B} |f(y)| \, \mathrm{d}y$$

$$\ge \frac{1}{m(B)} \int_{E \cap B} |f(y)| \, \mathrm{d}y$$

$$\ge \frac{1}{m(B)} \int_{E \cap B} \varepsilon_0 + Mf(y) \, \mathrm{d}y$$

$$= \frac{m(E \cap B)}{m(B)} \varepsilon_0 + \frac{1}{|B|} \int_{E \cap B} Mf(y) \, \mathrm{d}y.$$

Taking the integral with respect to x:

$$\left(1 - \frac{|E \cap B|}{|B|}\right) \int_{E \cap B} Mf \ge \frac{|E \cap B|^2}{|B|} \varepsilon_0.$$

This implies $(1 - \lambda)N \ge \lambda \varepsilon_0$, which is impossible as $\lambda \to 1$.

Now for the last part, since $\{Mf > \alpha\}$ is open, $\forall x \in \{Mf > \alpha\}$, $\exists B \text{ s.t.}$

$$\int_{B} |f(y)| \, \mathrm{d}y > \alpha m(B).$$

Hence for disjoint balls B_{i_k} ,

$$||f||_{\mathcal{L}^1} \ge \sum_{l=1}^k \int_{B_{i_l}} |f(y)| \, \mathrm{d}y > \alpha \sum_{l=1}^k |B_{i_l}|.$$

If we could select B_{i_l} 's such that their measure achieves say 1% of E, then we're done.

Lemma 1.4

Let B_1, \ldots, B_n be open balls in \mathbb{R}^d . There exists i_1, \ldots, i_k such that B_{i_j} 's are pairwise disjoint, and

$$\bigcup_{i=1}^{n} B_i \subset \bigcup_{j=1}^{k} 3B_{i_j}.$$

Here 3B means to multiply the radius of the ball by 3.

Proof of lemma. Trivial, just take the largest ball first and using greedy algorithm. \Box

Remark 1.5 — For countable many balls, the conculsion holds with 3 replaced by 5.

In particular, for all compact sets $K \subset \{Mf > \alpha\}$, there exists a finite open cover B_1, B_2, \ldots, B_n of K. By lemma we can select B_{i_j} 's satisfying

$$\sum_{i=1}^{k} m(B_{i_j}) \ge \frac{1}{3^d} m\left(\bigcup_{i=1}^{n} B_i\right) \ge \frac{1}{3^d} m(K).$$

Combining with our previous discussion we get $||f||_{\mathcal{L}^1} \geq \frac{\alpha}{3^d} m(K)$.

Returning to the proof of Theorem 1.1, we can assume g is continuous with compact support,

$$\frac{1}{m(B)} \int_{B} |f(y) - g(y)| \, \mathrm{d}y \le M(f - g)(x)$$

Hence by taking B sufficiently small s.t. $|g(y) - g(x)| \le \varepsilon_0$ for all $x, y \in B$,

$$\frac{1}{m(B)} \int_{B} f(y) \, \mathrm{d}y \ge 3\varepsilon_0$$

$$\iff |f(x) - g(x)| + M(f - g)(x) \ge 2\varepsilon_0.$$

But

$$m\{|f(x) - g(x)| \ge \varepsilon_0\} + m\{M(f - g) > \varepsilon_0\} \le \frac{\|f - g\|_{\mathcal{L}^1}}{\varepsilon_0} + \frac{3^d}{\varepsilon_0}\|f - g\|_{\mathcal{L}^1} \le \frac{3^d + 1}{\varepsilon_0}\varepsilon.$$

This completes the proof.

Definition 1.6 (Lebesgue points). Let $|f(x)| < \infty$, f is locally integrable. If x satisfies

$$\lim_{|B| \to 0, B \ni x} \frac{1}{|B|} \int_{B} |f(y) - f(x)| \, \mathrm{d}y = 0,$$

we say x is a **Lebesgue point** of f.

Remark 1.7 — Here "locally integrable" means for all bounded measurable sets $E, f\chi_E \in \mathcal{L}^1$. This is denoted by $f \in \mathcal{L}^1_{loc}$.

Let E be a measurable set, χ_E locally integrable, If point x is called a **density point** of E if it's a Lebesgue point of χ_E .

Theorem 1.8

Let E be a measurable set, then almost all the points in E are density points of E, almost all the points outside of E are not density points of E.

Proof. This is a direct corollary of Theorem 1.1.

The differentiation theorem has some applications in convolution:

$$\begin{split} \frac{1}{|B|} \int_B f(y) \, \mathrm{d}y &= c_d^{-1} \varepsilon^{-d} \int_{B(x,\varepsilon)} f(y) \, \mathrm{d}y \\ &= \int f(x-y) \cdot c_d^{-1} \varepsilon^{-d} \chi_{B(0,\varepsilon)}(y) \, \mathrm{d}y \\ &= f * K_{\varepsilon}. \end{split}$$

where $K_{\varepsilon} = c_d^{-1} \varepsilon^{-d} \chi_{B(0,\varepsilon)}$, c_d is the measure of a unit sphere in \mathbb{R}^d .

By differentiation theorem, $\lim_{\varepsilon\to 0} f * K_{\varepsilon} = f(x)$, a.e.. In the homework we proved that there doesn't exist a function I s.t. f * I = f for all $f \in \mathcal{L}^1$, but the functions K_{ε} is approximating this "convolution identity".

Definition 1.9. In general, if $\int K_{\varepsilon} = 1$, $|K_{\varepsilon}| \leq A \min\{\varepsilon^{-d}, \varepsilon |x|^{-d-1}\}$ for some constant A, we say K_{ε} is an **approximation to the identity**.

"convolution kernel"

Let φ be a smooth function whose support is in $\{|x| \leq 1\}$, and $\int \varphi = 1$. The function $K_{\varepsilon} := \varepsilon^{-d} \varphi(\varepsilon^{-1} x)$ is called the Friedrichs smoothing kernel.

Theorem 1.10

If K_{ε} is an approximation to the identity, f integrable,

$$\lim_{\varepsilon \to 0} \|f * K_{\varepsilon} - f\|_{\mathcal{L}^{1}} = 0.$$

Proof.

$$|(f * K_{\varepsilon})(x) - f(x)| = \left| \int f(x - y) K_{\varepsilon}(y) \, \mathrm{d}y - f(x) \right|$$

$$\leq \int |f(x - y) - f(x)| |K_{\varepsilon}(y)| \, \mathrm{d}y$$

$$\leq \int_{|y| \leq R} |f(x - y) - f(x)| A\varepsilon^{-d} \, \mathrm{d}y + \int_{|y| > R} |f(x - y) - f(x)| A\varepsilon|y|^{-d-1} \, \mathrm{d}y.$$

Taking the integral over \mathbb{R}^d :

$$\begin{split} &\|K_{\varepsilon}*f-f\|_{\mathcal{L}^{1}}\\ &\leq A\varepsilon^{-d}\int\int_{|y|\leq R}|f(x-y)-f(x)|\,\mathrm{d}y\,\mathrm{d}x + A\varepsilon\int\int_{|y|>R}|f(x-y)-f(x)||y|^{-d-1}\,\mathrm{d}y\,\mathrm{d}x\\ &\leq A\varepsilon^{-d}\int\int_{|y|\leq R}|\tau_{-y}f(x)-f(x)|\,\mathrm{d}y\,\mathrm{d}x + A\varepsilon\int_{|y|>R}|y|^{-d-1}\int|\tau_{-y}f(x)|+|f(x)|\,\mathrm{d}x\,\mathrm{d}y\\ &\leq A\varepsilon^{-d}\int_{|y|\leq R}\|\tau_{-y}f-f\|_{\mathcal{L}^{1}}\,\mathrm{d}y + A\varepsilon\int_{|y|>R}|y|^{-d-1}2\|f\|_{\mathcal{L}^{1}}\,\mathrm{d}y. \end{split}$$

By the continuity of translation, $\forall \varepsilon_0$, let R be sufficiently small we have

$$\|\tau_{-y}f - f\|_{\mathcal{L}^1} < \varepsilon_0, \forall |y| < R.$$

$$||K_{\varepsilon} * f - f||_{C^1} < A\varepsilon^{-d}R^dc_d\varepsilon_0 + \varepsilon R^{-1}C$$

where C is a constant.

Take suitable $\varepsilon, \varepsilon_0$ s.t. $R = \varepsilon \varepsilon_0^{-\frac{1}{2d}}$, then $LHS \leq C_1 \varepsilon_0^{\frac{1}{2}} + C_2 \varepsilon_0^{\frac{1}{2d}} \to 0$.

Theorem 1.11

Let K_{ε} be an approximation to the identity, f integrable,

$$\lim_{\varepsilon \to 0} f * K_{\varepsilon} = f(x)$$

holds for Lebesgue points x of f.

Proof. WLOG x = 0, let

$$\omega(r) = \frac{1}{r^d} \int_{B(0,r)} |f(y) - f(0)| \, \mathrm{d}y,$$

we have $\lim_{r\to 0} \omega(r) = 0$, and ω is continuous.

$$\omega(r) \le \frac{\|f\|_{\mathcal{L}^1}}{r^d} + |f(0)|,$$

Thus ω is bounded.

Therefore we can compute

$$|K_{\varepsilon} * f(x) - f(x)| \leq \int |K_{\varepsilon}(y)| |f(x - y) - f(x)| \, \mathrm{d}y$$

$$\leq \int_{B(0,r)} |K_{\varepsilon}(y)| |f(x - y) - f(x)| \, \mathrm{d}y + \sum_{k \geq 0} \int_{2^k r \leq |y| < 2^{k+1}r} |K_{\varepsilon}(y)| |f(x - y) - f(x)| \, \mathrm{d}y$$

$$\leq A\varepsilon^{-d} \cdot r^d \omega(r) + \sum_{k \geq 0} \int_{2^k r \leq |y| < 2^{k+1}r} A\varepsilon |y|^{-d-1} |f(x - y) - f(x)| \, \mathrm{d}y$$

$$\leq A\varepsilon^{-d} r^d \omega(r) + \sum_{k \geq 0} A\varepsilon 2^{-k(d+1)} r^{-d-1} \cdot 2^{(k+1)d} r^d \omega(2^{k+1}r)$$

$$= A\varepsilon^{-d} r^d \omega(r) + A2^d \varepsilon r^{-1} \sum_{k \geq 0} 2^{-k} \omega(2^{k+1}r).$$

Let $r = \varepsilon$, since $\omega(r)$ is continuous and bounded, we're done.

§1.1 Lebesgue Differentiation theorem part 2

Recall that Fundamental theorem of Calculus has a second part, Given any differentiable function F(x), if F'(x) Riemann integrable, then

$$F(b) - F(a) = \int_a^b F'(x) dx.$$

When F has nice properties (smooth), then it has a generalization to higher dimensions (Stokes' theorem).

But in Lebesgue integration theory, the derivative of a function is hard to define. Hence we'll find an alternative for F'(x).

Example 1.12

Consider Heaviside function $H(t) = \begin{cases} 1, & t > 0 \\ 0, & t \le 0 \end{cases}$

Then H is differentiable almost everywhere, but $\int_{-1}^{1} H'(t) dt = 0 \neq H(1) - H(-1)$.

Example 1.13

Consider Cantor-Lebesgue function F, similarly we have F'(x) = 0, a.e., but $\int_0^1 F'(x) dx = 0 \neq F(1) - F(0)$.

Definition 1.14 (Dini derivatives). Let f(x) be a measurable function, define

$$D^{+}(f)(x) = \limsup_{h > 0, h \to 0} \frac{f(x+h) - f(x)}{h}, \quad D^{-}(f)(x) = \limsup_{h < 0, h \to 0} \frac{f(x+h) - f(x)}{h}.$$

$$D_{+}(f)(x) = \liminf_{h>0, h\to 0} \frac{f(x+h) - f(x)}{h}, \quad D_{-}(f)(x) = \liminf_{h<0, h\to 0} \frac{f(x+h) - f(x)}{h}.$$

Theorem 1.15 (Lebesgue Differentiation theorem for increasing functions)

Let f be an increasing function on [a, b], then F'(x) exists almost everywhere, and

$$\int_{a}^{b} F'(x) \, \mathrm{d}x \le F(b) - F(a).$$

Sketch of the proof. The outline of the proof is as follows:

Step 1: Decompose $F = F_c + J$, where F_c is continuous, J is a jump function.

Step 2: Prove F_c increasing and J' = 0, a.e..

Step 3: Prove
$$D^+(F) < +\infty$$
, a.e., $D^+(F) \le D_-(F)$, a.e., and $D^-(F) \le D_+(F)$, a.e..

We proceed step by step.

Step 1 Denote $F(x+0) = \lim_{h\to 0^+} F(x+h)$, $F(x-0) = \lim_{h\to 0^-} F(x+h)$. Since F increasing, let $\{x_n\}$ be all the discontinuous points of F. Define:

$$j_n(x) = \begin{cases} 0, & x < x_n \\ \beta_n, & x = x_n \\ \alpha_n, & x > x_n \end{cases}$$

where $\alpha_n = F(x_n + 0) - F(x_n - 0), \beta_n = F(x_n) - F(x_n - 0)$

Hence the jump function

$$J_F(x) = \sum_{n=1}^{+\infty} j_n(x) \le \sum_{n=1}^{+\infty} \alpha_n = \sum_{n=1}^{+\infty} (F(x_n + 0) - F(x_n - 0)) \le F(b) - F(a)$$

is well-defined and increasing.

Theorem 1.16

 $F - J_F$ is continuous and increasing.

Proof. First note that

$$\lim_{h \to 0^+} (F(x+h) - J_F(x+h)) = F(x+0) - \lim_{h \to 0^+} J_F(x+h) = F(x-0) - \lim_{h \to 0^+} J_F(x-h)$$

This can be derived from the definition of J_F : If F is continuous at x, the equality is obvious; If $x = x_n$ for some n,

$$\lim_{h \to 0^+} J_F(x+h) = \sum_{x_k \le x_n} \alpha_k + \lim_{h \to 0^+} \sum_{x_n < x_k \le x_n + h} j_k(x+h) = \sum_{x_k \le x_n} \alpha_k$$

$$\lim_{h \to 0^+} J_F(x-h) = \lim_{h \to 0^+} \sum_{x_k < x_n - h} \alpha_k + \lim_{j \to 0^+} \sum_{x_k = x_n - h} \beta_k = \sum_{x_k < x_n} \alpha_k$$

Note that $\alpha_n = F(x_n + 0) - F(x_n - 0)$, thus $F - J_F$ is continuous. Secondly,

$$F(x) - J_F(x) \le F(y) - J_F(y), \quad \forall a \le x \le y \le b.$$

Because

$$J_F(y) - J_F(x) = \sum_{x < x_j < y} \alpha_j + \sum_{x_k = y} \beta_k - \sum_{x_k = x} \beta_k \le \sum_{x < x_j < y} \alpha_j + F(y) - F(y - 0) \le F(y) - F(x).$$

which means $F - J_F$ is increasing.

Step 2

Proposition 1.17

The jump function J(x) is differentiable almost everywhere, and J'(x) = 0, a.e..

Proof. The Dini derivatives of J(x) exist and are non-negative (since J is increasing).

$$\overline{D}(J)(x) = \max\{D^+(J)(x), D^-(J)(x)\}.$$

Let $E_{\varepsilon} = \{\overline{D}(J)(x) > \varepsilon > 0\}$. We'll prove E_{ε} is null for all ε . If $x \in E_{\varepsilon}$, $\exists h$ s.t.

$$\frac{J(x+h)-J(x)}{h}>\varepsilon\implies J(x+h)-J(x-h)>\varepsilon h.$$

Let $N \in \mathbb{N}$ s.t. $\sum_{n>N} \alpha_n < \frac{\varepsilon \delta}{10}$. Define $J_N(x) = \sum_{n>N} j_n(x)$.

$$E_{\varepsilon,N} = \{\overline{D}(J_N)(x) > \varepsilon\}, \quad E_{\varepsilon} \subset E_{\varepsilon,N} \cup \{x_1, \dots, x_N\},$$

Since for $x \neq x_i$,

$$\overline{D}(J)(x) = \limsup_{h \to 0} \frac{J(x+h) - J(x)}{h} = \limsup_{h \to 0} \left(\frac{J_N(x+h) - J_N(x)}{h} + \frac{1}{h} \sum_{n=1}^{N} (j_n(x+h) - j_n(x)) \right).$$

Lemma 1.18

Let \mathcal{B} be a collection of balls with bounded radius in \mathbb{R}^d . There exists countably many disjoint balls B_i s.t.

$$\bigcup_{B\in\mathcal{B}}B\subset\bigcup_{i=1}^\infty 5B_i.$$

Proof. Let r(B) denote the radius of B. Take B_1 s.t. $r(B_1) > \frac{1}{2} \sup_{b \in \mathcal{B}} r(B)$. The rest is the same as before.

By lemma, there exists countably many disjoint intervals $(x_i + h_i, x_i - h_i)$ s.t.

$$m^*(E_{\varepsilon,N}) \le 5 \sum_{i=1}^{\infty} 2h_i$$

$$\le 10 \sum_{i=1}^{\infty} \varepsilon^{-1} (J_N(x_i + h_i) - J_N(x_i - h_i))$$

$$\le 10 \varepsilon^{-1} (J_N(b) - J_N(a)) < \delta.$$

Hence $E_{\varepsilon,N}$ is a null set $\implies E_{\varepsilon}$ null, and at last $\overline{D}(J) = 0, a.e.$.

Step 3 First we prove $D^+(F) < \infty, a.e.$. Let $E_{\gamma} = \{x : D^+(F)(x) > \gamma\}.$ When $h \in \left[\frac{1}{n+1}, \frac{1}{n}\right]$:

$$\frac{F(x+h) - F(x)}{h} \le \frac{n+1}{n} \frac{F(x+\frac{1}{n}) - F(x)}{\frac{1}{n}},$$
$$\ge \frac{n}{n+1} \frac{F(x+\frac{1}{n+1}) - F(x)}{\frac{1}{n+1}}.$$

Thus

$$D^{+}(F)(x) = \limsup_{h \to 0} \frac{F(x+h) - F(x)}{h} = \limsup_{n \to \infty} \frac{F(x + \frac{1}{n}) - F(x)}{\frac{1}{n}}$$

is measurable.

Lemma 1.19 (Sunrise lemma)

Let G(x) be a continuous function on \mathbb{R} . Define

$$E = \{x : \exists h > 0, G(x+h) > G(x)\}.$$

Then E is open and $E = \bigcup_{i=1}^{\infty} (a_i, b_i)$, where (a_i, b_i) are disjoint finite intervals s.t. $G(a_i) = G(b_i)$.

When G is defined on finite interval [a, b], we also have $G(a) \leq G(b_1)$.

Proof. E is open since G is continuous.

Take an interval (a, b), by definition $a, b \notin E$, so $G(a) \ge G(b)$.

Since $b \notin E, G(x) \leq G(b), \forall x > b$. If G(a) > G(b), Let $G(a + \varepsilon) > G(b)$, as $a + \varepsilon \in E$, exists h > 0 s.t. $G(a + \varepsilon + h) > G(a + \varepsilon)$.

But G has a maximum on $[a + \varepsilon, b]$, say G(c), we must have $c \neq a + \varepsilon, b$. This leads to a contradiction.

For $x \in E_{\gamma}$, $\exists h > 0$ s.t. $F(x+h) - F(x) > \gamma h$, by Sunrise Lemma on $F(x) - \gamma x$,

$$m(E_{\gamma}) \leq \sum_{k=1}^{\infty} (b_k - a_k) \leq \gamma^{-1} \sum_{k=1}^{\infty} (F(b_k) - F(a_k)) \leq \gamma^{-1} (F(b) - F(a)).$$

Therefore when $\gamma \to \infty$, $m(E_{\gamma}) \to 0$.

The last part is $D^+(F) \leq D_-(F)$, a.e..

Similarly let

$$E_{r,R} = \{D^+(F)(x) > R, D_-(F)(x) < r\}.$$

WLOG $E_{r,R} \subset [c,d]$, and $d-c < \frac{R}{r}m(E_{r,R})$.

Let G(x) = F(-x) + rx, by Sunrise Lemma on [-d, -c],

$${s: \exists h > 0, G(x+h) > G(x)} = \bigcup_{k} (-b_k, -a_k).$$

Note that $-E_{r,R}$ is contained in the above set, and $G(-b_k) \leq G(-a_k) \iff F(b_k) - F(a_k) \leq r(b_k - a_k)$,

We use Sunrise Lemma again on each (a_k, b_k) and F(x) - Rx,

$$E_{r,R} \cap (a_k, b_k) \subset \{x : \exists h > 0, F(x+h) - F(x) \ge Rh\} = \bigcup_{l=1}^{\infty} (a_{k,l}, b_{k,l}).$$

Hence

$$m(E_{r,R}) \le \sum_{k,l=1}^{\infty} (b_{k,l} - a_{k,l})$$

$$\le R^{-1} \sum_{k,l=1}^{\infty} (F(b_{k,l}) - F(a_{k,l})) \le R^{-1} \sum_{k=1}^{\infty} (F(b_k) - F(a_k))$$

$$\le R^{-1} r \sum_{k=1}^{\infty} (b_k - a_k) \le R^{-1} r (d - c),$$

which gives a contradiction!