

①  $\equiv$  a very nice book "Convex Figures" Yaglom.  
used at Moscow State.

Convex functions are very much related to the convex

bodies 
$$U_f := \left\{ (x, z) \in \mathbb{R}^n \times \mathbb{R} \mid z \geq f(x) \right\}$$

is a convex body  $(1-t)(x_1, z_1) + t(x_2, z_2)$

$\in U_f$  since

$$(1-t)z_1 + tz_2 \geq (1-t)f(x_1) + tf(x_2) \geq f((1-t)x_1 + tx_2).$$

" $f$  convex"  $\Rightarrow f$  is locally Lipschitz  $\Leftrightarrow f \in W_{loc}^{1, \infty}$

Hence  $Df$  — the distributional derivative is  $L^\infty$ .

Theorem 6.9:  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex.

Then 
$$\left| f(y) - f(x) - Df(x)(y-x) - \frac{1}{2}(y-x)^T D^2 f(x)(y-x) \right|$$
$$= o(|y-x|^2) \quad \text{as } y \rightarrow x$$

a.e  $x \in \mathbb{R}^n$ .

Before we prove the thm we need to make sense of  $D^2 f(x)$ .

(A) This is NOT the distributional <sup>2nd</sup> derivative of  $f$ !

Since  $f$  is Lipschitz  $\Rightarrow f_i = \partial_i f \in L^\infty_{loc}$

$\Rightarrow f_i \in L^1_{loc}$ .

Hence  $f_{;i}(x)$  needs some deliberations.

$$\textcircled{B} \quad f_1, \dots, f_n \in BV_{loc}(\mathbb{R}^n)$$

To obtain this observe that:

$$f^\varepsilon = j_\varepsilon * f \quad \text{satisfies} \quad D^2 f^\varepsilon \geq 0.$$

$$\Rightarrow \sum_{i,j=1}^n \int f^\varepsilon \phi_{x_i x_j} V^i V^j = \int \phi \sum_{i,j=1}^n f_{x_i x_j}^\varepsilon V^i V^j$$

$$\Rightarrow L(\phi) := \int f \phi_{x_i x_j} V^i V^j \geq 0 \quad \forall \phi \geq 0, \phi \in \mathcal{D}(\mathbb{R}^n)$$

$$\text{By RM} \Rightarrow \forall V \in \mathbb{S}^{n-1} \Rightarrow$$

$$L_V(\phi) = \int \phi \, d\mu^V \quad \mu^V \text{ is a Radon measure!}$$

$$\Rightarrow \mu^{e_i} = \mu^{e_i} \quad V = \frac{e_i + e_j}{\sqrt{2}}$$

$$\Rightarrow \sum_{k,l} \phi_{x_k x_l} V^k V^l = \frac{1}{2} \sum (\phi_{x_i x_i} + 2\phi_{x_i x_j} + \phi_{x_j x_j})$$

$$\Rightarrow \int f \phi_{x_i x_j} = \int f (\phi_{VV} - \frac{1}{2}\phi_{ii} - \frac{1}{2}\phi_{jj})$$

$$\Rightarrow \mu^{j^i} = \mu^V - \frac{1}{2}\mu^{ii} - \frac{1}{2}\mu^{jj}$$

Clearly  $(\mu^{ij})$  is a  $\geq 0$  matrix of measures.

Namely  $\mu^V = \sum (\mu^{ij}) V^i V^j$  is a positive measure.

Now we may make sense of the derivation of  $f_k$

$$\begin{aligned} \int f_k \operatorname{div}(\phi) &= \int f_k \sum \phi_{x_i}^i = - \int f \sum \phi_{x_i x_k}^i \\ &= - \int \phi^i d\mu^{ik} \end{aligned}$$

Hence  $L_k(\phi) \leq C \sum_i |\mu^{ik}(K)|$  if  $\phi \in C_c^1(\mathbb{R}^n, \mathbb{R}^n)$   
 $\operatorname{spt} \phi \subset K \quad |\phi| \leq 1$

$\Rightarrow f_k$  is a BV function  
 by RM- essentially the definition!

①  $\mu_{ac}^{ij} = f_{ij} d\mu$   $\mu_s^{ij}$  — the singular part of  $\mu^{ij}$   
 w. r. t  $d\mu$ -Lebesgue.

Now  $D_{ij}^2 f = f_{ij}$ .

Namely in Alexandrov's theorem  $D^2 f(x)$  is the regular part of the distributional derivative of  $f$ .

$[D^2 f]_d$  — means  $(\mu^{ij})$  — the Radon measure

$[D^2 f] = (f_{ij})$

$[D^2 f]_s = \mu_s^{ij}$  — the singular part of  $[D^2 f]_d$ .

(2) Proof of the Thm 6.9 :

$$(i) \quad \forall a.e. x \quad \lim_{r \rightarrow 0} \int_{B(x,r)} |Df(y) - Df(x)| = 0.$$

$$(ii) \quad \lim_{r \rightarrow 0} \int_{B(x,r)} |D^2 f(y) - D^2 f(x)| = 0$$

$$(iii) \quad \lim_{r \rightarrow 0} \frac{[D^2 f]_s(B(x,r))}{|B(x,r)|} = 0$$

(ii) & (iii) follows from Lebesgue differentiation theorem.

[Thm 3.22 of Folland].

Below we shall show that if  $x$  is a point where (i), (ii), (iii) hold, then Theorem hold at  $x$ .

$$\text{Also } f^i(x) \rightarrow f(x) \text{ \& } Df^i(x) \rightarrow Df(x).$$

The tricky part is to get  $o(r^2)$  estimate on

$$f(y) - f(x) - Df(x) \cdot (y-x) - \frac{1}{2} (y-x)^T D^2 f(x) (y-x).$$

[Fix  $\varepsilon$ , it is obviously true for  $f^i$ ]

w.l.o.g assume  $x=0$ .

$$\text{Consider } H^i(y) = f^i(y) - f^i(0) - Df^i(0) \cdot y - \frac{1}{2} y^T D^2 f^i(0) y$$

We also denote

$$h(y) := f(y) - f(x_0) - Df(x_0) \cdot y - \frac{1}{2} y^T D^2 f(x_0) \cdot y$$

The goal is to show  $|h(y)| = o(|y|^2)$ .  $H^\varepsilon(y) \xrightarrow{\text{a.e.}} h(y)$

Taylor expansion of  $f^\varepsilon(y) \Rightarrow$

$$\begin{aligned} f^\varepsilon(y) &= f^\varepsilon(x_0) + \int_0^1 \frac{d}{dt} f^\varepsilon(tx) \Rightarrow f^\varepsilon(y) - f^\varepsilon(x_0) = \int_0^1 Df^\varepsilon(tx) \cdot y \, dt \\ &= \int_0^1 Df^\varepsilon((1-t)x) \cdot y \, dt = Df^\varepsilon(x_0) \cdot y + \int_0^1 y^T D^2 f^\varepsilon_{(1-t)x} y \, dt \\ &= Df^\varepsilon(x_0) \cdot y + \int_0^1 (1-s) y^T D^2 f^\varepsilon(sx) \cdot y \, ds. \end{aligned}$$

This shows

$$H^\varepsilon(y) = \int_0^1 \left[ (1-s) \left( y^T D^2 f^\varepsilon(sx) y - y^T D^2 f(x_0) y \right) \right] ds$$

$\begin{cases} D^2 f^\varepsilon(sx) \rightarrow D^2 f(x_0) \\ \text{only in average sense.} \\ \text{by Lebesgue differentiation.} \end{cases}$

Claim:  $\forall \phi \in C_c^\infty(B(r))$

$$\int h(y) \cdot \phi(y) = o(r^2) \quad \text{as } r \rightarrow 0 \quad (*)$$

Proof of (\*)

$$\begin{aligned} \int_{B(r)} \phi(y) H^\varepsilon(y) &= \int_0^1 (1-s) \left( \int_{B(r)} \phi(y) y^T \left[ \underbrace{D^2 f^\varepsilon(sx)}_z - D^2 f(x_0) \right] y \, dy \right) ds \\ &= \int_0^1 \frac{(1-s)}{s^2} \left( \int_{B(rs)} \phi\left(\frac{z}{s}\right) z^T \left[ D^2 f^\varepsilon(z) - D^2 f(x_0) \right] z \, dz \right) ds \end{aligned}$$

Let 
$$g_\varepsilon(s) := \int_{B(rs)} \phi\left(\frac{z}{s}\right) z^T D^2 f^\varepsilon(z) \cdot z \, dz$$

$$\leftarrow = \int_0^1 \frac{(1-s)}{s^2} \frac{1}{r^n s^{2n}} \left[ g_\xi(s) - \int_{B(rs)} \phi\left(\frac{z}{s}\right) z^T D^2 f_\omega(z) z \right] ds$$

$$g_\xi(s) = \int_{B(rs)} f^\xi(z) \left( \sum_{z_i, z_j} \left[ \phi\left(\frac{z}{s}\right) z_i z_j \right]_{z_i, z_j} \right) dz$$

$$\downarrow \text{as } \xi \rightarrow 0$$

$$\int_{B(rs)} f(z) \sum_{z_i, z_j} \left[ \phi\left(\frac{z}{s}\right) z_i z_j \right]_{z_i, z_j} dz$$

$$= \int_{B(rs)} \phi\left(\frac{z}{s}\right) z_i z_j \cdot du_s^j$$

$$= \int_{B(rs)} \phi\left(\frac{z}{s}\right) z_i z_j \cdot D_j^2 f + \phi\left(\frac{z}{s}\right) z_i z_j \cdot \left( \left[ D_j^2 f \right]_s \right)_j$$

To apply DCT on  $g_\xi(s)$  we need an estimate.

Sub-claim:  $\frac{|g_\xi(s)|}{s^{n+2}} \leq C$

If holds,  $\Rightarrow$

$$\int_{B(r)} \phi(\omega) H^\xi(\omega) \rightarrow \int_0^1 \frac{(1-s)}{s^2} \frac{1}{r^n s^{2n}} \int_{B(rs)} \phi\left(\frac{z}{s}\right) z_i z_j \left[ D_j^2 f(z) - D_j^2 f(\omega) \right] + \phi\left(\frac{z}{s}\right) z_i z_j \cdot du_s^j$$

$$\downarrow \int_{B(r)} \phi(\omega) h(\omega)$$

Hence 
$$\int_{B(r)} \phi(y) |h(y)| \leq C r^2 \int_0^1 \left( \underbrace{\int_{B(rs)} |D^2 f(x) - D^2 f^{(0)}|}_{= o(1)} + \underbrace{\frac{|D^2 f|_s(B(rs))}{(rs)^n}}_{\text{as } r \rightarrow 0} \right) ds$$

$= o(r^2)$

This proves (\*).

③ Proof of subclaim & finishing the argument.

Ⓐ 
$$\int_{B(r)} |h(y)| dy = o(r^n)$$

Namely  $|h(y)| = o(|y|^n)$  in average sense.

It suffices to get estimate for  $\phi$   $\phi \in C_c^\infty(B(r))$   
 $|\phi| \leq 1$

or 
$$\int_{B(r)} \phi h$$

since one may approximate  $\text{sgn}(h)$  by  $\phi$

Ⓑ  $\exists C$

$$\sup_{B(r)} |Dh| \leq \frac{C}{r} \int_{B(r)} |h| dy + Cr$$

Recall. 
$$h = f(y) - \underbrace{f^{(0)}}_{f^{(0)}} - Df^{(0)} \cdot y - \frac{1}{2} y^T D^2 f^{(0)} y$$

Let  $\Lambda = |D^2 f^{(0)}| \Rightarrow h + \frac{\Lambda}{2} |y|^2$  is convex.

Since  $-f^{(0)} - Df^{(0)} \cdot y - \frac{1}{2} y^T D^2 f^{(0)} y + \frac{\Lambda}{2} |y|^2$  is convex.

Now by Th. 6.7

$$\sup_{B(\xi)} \left| \mathcal{R}\left(h + \frac{\Delta}{z} |g|^2\right) \right| \leq \frac{C}{r} \left( \int_{B(r)} |h| + \frac{\Delta}{z} |g|^2 \right)$$

(c)  $\sup_{B(\xi)} |h| = o(r^4)$  as  $r \rightarrow 0$

$$\leq C r$$

this C

This proves the theorem.  $\forall \varepsilon > 0$

pick  $\eta \in (0, 1)$ ,  $\eta^{\frac{1}{4}} < \frac{1}{2}$ ,  $C(2\eta)^{\frac{1}{4}} = \varepsilon$ .

We shall show

$$|h(g)| \leq 2\varepsilon r^2 \text{ if } r \leq r_0 := r_0(\varepsilon, \eta).$$

Step 1.  $\frac{1}{\varepsilon} \int_{B(r)} |h| < r^2 \eta r^{n-2} \alpha(h)$  if  $r \leq r_0$

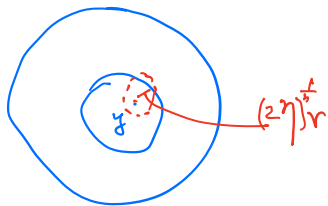
by (B).

Then  $|\{z \in B(r) \mid |h(z)| \geq \varepsilon r^2\}| \leq \eta |B(r)|$

Since

$$\underbrace{|\{z \in B(r) \mid |h(z)| \geq \varepsilon r^2\}|}_{S} \cdot \cancel{\varepsilon r^2} \leq \int_{B(r)} |h| \leq \cancel{\varepsilon \eta r^2} |B(r)|$$

Now  $\forall y \in B(\xi)$





$$\text{If } B(y, (\varepsilon\eta)^{\frac{1}{n}}r) \cap S^c = \emptyset \Rightarrow B(y, (\varepsilon\eta)^{\frac{1}{n}}r) \subset S$$

$$\Rightarrow |S| \geq 2\alpha(n)\eta r^n$$

$$\text{Hence } \Rightarrow \exists z, \quad |h(z)| < \varepsilon r^2 \quad z \in B(y, (\varepsilon\eta)^{\frac{1}{n}}r)$$

$$\text{But } |h(y)| \leq |h(z)| + \underbrace{|h(z) - h(y)|}_{\leq C(2\eta)^{\frac{1}{n}}r^2} \leq 2\varepsilon r^2$$

① proof of the sub-claim.

We only need to establish it for  $\varepsilon \ll 1$ .  $\triangleright$   $r$  small.

$$\frac{|j_\varepsilon(s)|}{s^{n+2}} \leq \frac{r^2}{s^n} \int_{B(rs)} |D^2 f^\varepsilon(z)| dz$$

$$= \frac{r^2}{s^n} \int_{B(rs)} \left| \int_{B(rs)} D^2 j_\varepsilon(z-y) f(y) dy \right| dz$$

$$= \frac{r^2}{s^n} \int_{B(rs)} \left| \int_{B(rs)} j_\varepsilon(z-y) \underbrace{d_y D_d^2 f}_{\text{measure}} \right| dz$$

$$\leq \frac{Cr^2}{\varepsilon^n s^n} \int_{B(rs+\varepsilon)} \left( \int_{B(rs) \cap B(y, \varepsilon)} dz \right) d_y D_d^2 f$$

$$\leq Cr^2 \frac{\min((rs)^\alpha, \varepsilon^n)}{\varepsilon^n s^n} \cdot \underbrace{(D_d^2 f)(B(rs+\varepsilon))}_{(\leq C(rs+\varepsilon)^\alpha)}$$

$$\leq C \cdot r^{n+2}$$

using  $r$  &  $\varepsilon$  small.

$$\begin{cases} \text{if } rs \leq \varepsilon \\ \leq C \frac{(rs)^\alpha \varepsilon^n}{\varepsilon^n s^n} \\ \text{if } rs \geq \varepsilon \\ \leq C \frac{\varepsilon^n (rs)^\alpha}{\varepsilon^n s^n} \text{ as well.} \end{cases}$$

$$D_d^2 f = D^2 f + [D^2 f]_s$$