

Recap.

Thm 3.8 of R & W (2018)

If $f_n \xrightarrow{e} f$, $\varepsilon_n \downarrow 0$.

$\limsup_n \varepsilon_n - \argmin f_n \subseteq \argmin f$.

$$f_n \xrightarrow{e} f \iff \text{epi-} f_n \xrightarrow{PK} \text{epi-} f$$

$$\text{epi-} f = \{(\alpha, \alpha) \in \mathbb{R} \times \mathbb{R} : f(\alpha) \leq \alpha\}.$$

We will define 'epi-distance' btw fcs by
'set-distance' btw epi-graphs.

Def. ρ -epidistance. $\forall \rho > 0$ ρ -AW distance.

$$d_\rho(C, D) = \sup_{x \in B(0, \rho)} |d(x, C) - d(x, D)|$$

$$d_\rho(f, g) = d_\rho(\text{epi-}f, \text{epi-}g)$$

Rmk. d_ρ is a pseudo metric: $d_\rho(C, D) = 0 \nRightarrow C = D$.

d_ρ is non-decreasing in ρ .

Def. Epi-distance.

$$d(C, D) = \int_0^\infty e^{-\rho} d_\rho(C, D) d\rho.$$

$$d(f, g) = d(\text{epi-}f, \text{epi-}g)$$

Thm 4.36 of RW (1999)

$$C_n \xrightarrow{PK} C \iff d_p(C_n, C) \rightarrow 0 \quad \forall p > 0$$

$$\iff d(C_n, C) \rightarrow 0$$

Thm 7.58 of RW (1998)

$$f_n \xrightarrow{e} f \iff d(f_n, f) \rightarrow 0$$

Moreover, $(\text{lsc-fn}(\mathbb{R}^d), d)$ is a proper, complete, separable metric space.

Remark 1. On reference point 0.

$$\theta_0 = \operatorname{argmin} f(\theta) \quad 0 = \operatorname{argmin} [f(u + \theta_0) - f(\theta_0)]$$

$\Rightarrow$

$$\hat{\theta}_n \in \operatorname{argmin} f_n(\theta) \quad \hat{\theta}_n - \theta_0 = \operatorname{argmin} [f_n(u + \theta_0) - f_n(\theta_0)]$$

If we change the reference point, induced topology on

Remark 2. (Pompeiu-Hausdorff distance) $(\text{lsc-fn}(\mathbb{R}^d))$ doesn't change

$$d_{\infty}(C, D) = \sup_x |d(x, C) - d(x, D)|$$

$$d_p \leq d_{\infty} \quad \forall p > 0 \quad d \leq d_{\infty}$$

Therefore,

$$d_{\infty}(C_n, C) \rightarrow 0 \Rightarrow C_n \xrightarrow{PK} C$$

In fact, convergence in d_{∞} is strictly stronger notion than PK-convergence.

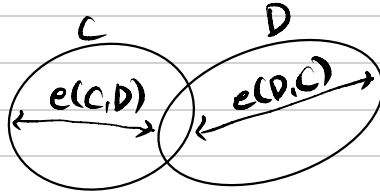
$$a_n \rightarrow a \quad |b_n| \rightarrow \infty \quad C_n = \{a_n, b_n\} \quad PK\text{-}\lim_n C_n = \{a\}$$

$$d_p(C_n, C) \rightarrow 0 \quad d_{\infty}(C_n, C) \rightarrow \infty \quad \text{If } C_n, C \subseteq B(0, R) \text{ then } \dots$$

Def. Pseudo p -epi distance.

Def. p -truncated Hausdorff distance.

$$e(C, D) = \sup_{x \in C} d(x, D) = \sup_{x \in C} \inf_{y \in D} d(x, y)$$



$$d_{\text{Haus}}(C, D) = \max \{ e(C, D), e(D, C) \}$$

$$\hat{d}_p(C, D) = \max \{ e(C \cap B(o, p), D), e(D \cap B(o, p), C) \}$$

Remark. $\hat{d}_p$ is non-decreasing in p .

$\hat{d}_p(C, D)$ is not a pseudo-metric. Triangle inequality does not hold.

Thm 4.36. $C_n \xrightarrow{pk} C \iff \hat{d}_p(C_n, C) \rightarrow 0 \quad \forall p > 0$.

Prop 4.37 of RW(1998) If C, D are closed,

$$\hat{d}_p(C, D) \leq d_p(C, D) \leq \hat{d}_{p'}(C, D) \quad p' = p + \frac{d(o, C)}{v d(o, D)}$$

If C, D are convex, $p' = p + d(o, C) \vee d(o, D)$

If C, D are convex, $0 \in C \cap D$ then $d|_p = \hat{d}|_p$

Thm 3.1 of Attouch and Wets (1983), Thm 4.1 of Royset (2020)

For any l.s.c. fns f, g . Suppose that

$$\inf f \in (-p, p), \varepsilon\text{-argmin } f \cap B(0, p) \neq \emptyset$$

Then,

$$\inf g - \inf f \leq \hat{d}|_p(f, g).$$

To be able to consider any perturbation of f and still obtain a proximity of minimizers, we need to know some geometrical shape at minimizer.

Thm 4.2 of Royset (2020). For any l.s.c. fns f, g .

Suppose that $\inf f, \inf g \in (-p, p)$ and

$$\text{argmin } f \cap B(0, p) \neq \emptyset, \text{argmin } g \cap B(0, p) \neq \emptyset$$

$$f(x) - \inf f \geq \varphi(\text{dist}(x, \text{argmin } f))$$

for some increasing $\varphi: [0, \infty) \rightarrow [0, \infty)$ s.t. $\varphi(0) = 0$

Then,

$$e(\text{argmin } g \cap B(0, p), \text{argmin } f) \leq \hat{d}|_p(f, g) + \varphi^{-1}(\hat{d}|_p(f, g))$$

$$(\text{Convexity}) \quad f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle$$

$$(\text{Strong Convexity}) \quad \exists \mu > 0 \text{ s.t. } f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle + \frac{\mu}{2} \|y - x\|_2^2$$

S.C.

$$(\text{Weak SC}) \quad \exists \mu > 0. \quad f(x_p) \geq f(x) + \langle \nabla f(x), x_p - x \rangle + \frac{\mu}{2} \|x_p - x\|_2^2$$

$$x_p = \text{Proj}_{\text{argmin} f}(x)$$

(Polyak-Łojasiewicz, PL)

$$\frac{1}{2} \|\nabla f(x)\|_2^2 \geq \mu (f(x) - \inf f)$$

$$x_{t+1} = x_t - \eta_t \nabla f(x_t)$$

(Quadratic growth)

$$f(x) - f^* \geq \frac{\mu}{2} \|x - x_p\|_2^2$$

Thm 2 of Karimi et al. (2020)

$$(\text{SC}) \rightarrow (\text{WSC}) \rightarrow (\text{PL}) \rightarrow (\text{QG})$$

If f is diff'l, ∇f is Lipschitz, f is convex,

$$(\text{QG}) \rightarrow (\text{PL})$$

Ex. Many (linear) moment inequalities

θ^* solves

$$\begin{aligned} \mathbb{E} h_1(x, \theta) &\leq 0 \\ \vdots \\ \mathbb{E} h_J(x, \theta) &\leq 0 \end{aligned} \quad \left[\begin{aligned} -\mathbb{P}(X \geq \theta) + \frac{1}{2} &\leq 0 \\ -\mathbb{P}(X \leq \theta) + \frac{1}{2} &\leq 0 \end{aligned} \right]$$

$$h_j(x, \theta) = a_j^T \theta - h_j(x)$$

$$f(\theta) = \max_{j=1 \dots J} (a_j^T \theta - \mathbb{E} h_j(x))_+$$

$$f_n(\theta) = \max_{j=1 \dots J} (a_j^T \theta - \bar{h}_{j,n})_+$$

$$A = \{ \theta : A\theta \leq h \text{ coordinate wise} \} \quad A = \begin{bmatrix} a_1^T \\ \vdots \\ a_J^T \end{bmatrix}$$

(Hoffman's constant)

$$h = \begin{bmatrix} \mathbb{E} h_1(x) \\ \vdots \\ \mathbb{E} h_J(x) \end{bmatrix}$$

$$\exists K(A) > 0$$

$\forall u,$

$$\text{dist}_{\|\cdot\|_p}(u, A) \leq K(A)_{p,q} \| (Au - h)_+ \|_q$$

$$f(u) - f^* \leq K(A)_{p,\infty}^T \text{dist}_{\|\cdot\|_p}(u, A) \quad (\text{linear conditioning})$$

Thm 4.4 of Royset (2020)

For l.s.c fns f_n, f . Suppose $f_n \xrightarrow{e} f$.

$\inf f_n \in (-p, p-\varepsilon)$, $\inf f \in (-p, p)$.

ε -argmin $f \cap B(0, p) \neq \emptyset \quad \forall \varepsilon > 0$.

Then,

$$e(\varepsilon\text{-argmin } f_n \cap B(0, p), \text{argmin } f) \leq \hat{d}_p(f_n, f)$$

$$\forall \varepsilon > 0$$

Prop 4.2 of Royset (2018, SIAM) For any l.s.c fns.

$f, g: \mathcal{X} \rightarrow \overline{\mathbb{R}}$. $\forall p > 0$.

$$\begin{aligned} \hat{d}_p(f, g) &\leq \sup_{A_p} |f - g| & A_p &= \{x \in B(0, p) : f(x) \leq p \\ &&& \text{or } g(x) \leq p\} \\ &\leq \sup_{B(0, p)} (f - g) \end{aligned}$$

Ex. Many linear manatees ineq.

$$\hat{d}_p(f, f_n) \leq \max_{j=1, \dots, J} |E h_j(x) - h_{j,n}(x)| \leq \sqrt{\frac{\log J}{n}}$$

if $h_j(x)$'s are sub-g.