

The Łojasiewicz exponent

We begin by stating definitions. ~~Throughout~~ In the following we refer to Feschan (2019) (who at times is stating results from other work).

Let $k = \mathbb{R}$ or \mathbb{C} , & consider k^d .

Theorem (Łojasiewicz gradient inequality): Let $U \subseteq k^d$ be open and $f: U \rightarrow k$ be analytic. Let $x^* \in U$ be a critical point: $\nabla f(x^*) = 0$. For x^* , there exist constants $C > 0$, $\sigma \in (0, 1]$ and $\theta \in [\frac{1}{2}, 1)$ such that

$$\|\nabla f(x)\| \geq C |f(x) - f(x^*)|^{\theta} \text{ for all } x \in B_\sigma(x^*) \quad (1)$$

where $B_\sigma(x^*) \subseteq U$ is a σ -ball centred at x^* .

Definition The Łojasiewicz exponent of a differential function $f: U \rightarrow k$ is the smallest $\theta \geq 0$ such that there exist C, σ so that (1) holds, or more precisely:

$$\hat{\theta} = \inf \{ \theta \in [0, \infty) \mid \exists C > 0, \sigma \in (0, 1] \text{ such that (1) holds} \}$$

We focus on the case of analytic functions, when the Łojasiewicz inequality ~~guarantees~~ always takes $\hat{\theta}$ is greater than $\frac{1}{2}$.

- Feschan (2019) "Resolution of singularities & geometric proofs of the Łojasiewicz inequalities"

The Lojasевич gradient inequality has a corollary, sometimes also called the Lojasевич distance inequality. Various versions of this statement are possible, e.g. if the function is C^1 & the gradient inequality holds then so does the distance inequality. See Furukawa Corollary 4, 5, 6. We follow the proof in Bierstone 1997.

Corollary: Let $U \subseteq \mathbb{R}^d$ be open and $f: U \rightarrow \mathbb{R}$ analytic. Suppose that $\nabla f(x) = 0 \Rightarrow f(x) = 0$ for all $x \in B_r(x^*)$, and that $f(x) \geq 0$. Then there exist constants such that (1) is satisfied. (Lojasевич gradient inequality).

Then:

$$f(x) \geq C' d(x, \overline{x})^\alpha \quad \text{for all } x \in B_r(x^*)$$

where $t^* = (\theta C)^{\frac{1}{1-\theta}}$, $C' = ((1-\theta)C)^{\frac{1}{1-\theta}}$ and $\alpha = \frac{1}{1-\theta}$,

$$Z = f^{-1}(0) \cap B_{r^*}(x^*)$$

proof: Let $a \in B_r(x^*)$, s.t. $f(a) \neq 0$. Suppose that $x(t)$ is a solution to

$$\dot{x}(t) = -\frac{\nabla f(x)}{\|\nabla f(x)\|}, \quad x(0) = a.$$

We define $Q(t) = f(x(t))$. Then

$$Q'(t) = \nabla f(x(t)) \cdot \dot{x}(t) = -\|\nabla f(x(t))\|^2.$$

Hence for all t , $Q'(t) < 0$, i.e. $Q(t)$ is decreasing.

$$\frac{f(a)}{1-\theta} \geq \frac{f(x(t))^{1-\theta} - Q(t)^{1-\theta}}{1-\theta} = \frac{-1}{1-\theta} \int_0^t Q(s)^{1-\theta} ds.$$

$$= - \int_0^t Q(s)^{-\theta} Q'(s) ds$$

$$= \int_0^t f(x(s)) \cdot \|\nabla f(x(s))\| ds$$

$$\geq \frac{c}{\theta} \int_0^t ds = ct.$$

- Bierstone (1997) "Canonical desingularisation in characteristic zero by blowing up the maximum strata of a local invariant."

where in the last step we use Logisian inequality. Then $x(t)$ converges to $f^{-1}(0)$ in finite time t_0 . Since $\|x'(t)\| = 1$ we have to $\geq d(a, z)$. Rearranging gives

$$f(a) \geq ((1-\theta)c)^{1/(1-\theta)} d(a, z)^{\frac{1}{1-\theta}}$$

□.

I feel like there are some missing details with this proof. What if $x(t)$ leaves $B_r(x^*)$? This seems to make the application of the gradient inequality invalid. Likewise the dependence on a ~~constant point is interesting~~. Let's put this issue aside (it will come up again).

Taylor's theorem inequalities provide results about convergence of gradient flow:

Corollary. Let $f: U \rightarrow \mathbb{R}$ be an analytic function which is bounded below and let $f^* \in \mathbb{R}$ be the minimum (which we suppose is actually attained on U). Suppose $\nabla f(x) = 0 \Rightarrow f(x) = f^*$. Let θ be a value such that the Logisian inequality holds. Then: let $x(t)$ be a gradient flow of f starting sufficiently close to z .

(1) If $\theta = \frac{1}{2}$ then the distance $d(x(t), z)$ of $x(t)$ to z decays to zero at least as quickly as $K e^{-ct}$ for some $K, c > 0$.

Continuing Lojasiewicz exponents

Recall+

Let $f: U \rightarrow \mathbb{R}$, $U \subseteq \mathbb{R}^d$ be an analytic function. Recall

Theorem (Lojasiewicz): Let $x^* \in U$ be a critical point of f , and suppose $\nabla f(x) = 0 \Rightarrow f(x) = f(x^*)$, and that $f(x) \geq f(x^*)$. Then there exist constants $C, C > 0$, $\theta \in [\frac{1}{2}, 1)$ and a nbhd V of x^* such that

$$(1) \quad \|\nabla f(x)\| \geq C |f(x) - f(x^*)|^{\theta} \quad \text{for all } x \in V$$

$$(2) \quad |f(x) - f(x^*)| \geq C^l d(x, z)^{\alpha} \quad -" -$$

where $z = f^{-1}(x^*)$ and $\alpha = \frac{1-\theta}{l}$.

The smallest value of θ satisfying (1) is the Lojasiewicz exponent of f at x^* .

One reason to care about the Lojasiewicz exponent is that it controls the convergence rate of gradient flow, with qualitatively different behavior at $\theta = \frac{1}{2}$ vs $\theta > \frac{1}{2}$.

Corollary Let f be as in the theorem and θ any value such that (1) & (2) hold. Let $x(t)$ be a gradient flow of f starting sufficiently close to x^* .

$$(1) \quad \text{If } \theta = \frac{1}{2} \quad d(x(t), z) \leq A e^{-\alpha t}, \quad A, \alpha > 0.$$

$$(2) \quad \text{If } \theta > \frac{1}{2} \quad d(x(t), z) \leq A(t+B)^{-\frac{1-\theta}{2\theta-1}} \quad A, B > 0.$$

The exponent $\frac{1-\theta}{2\theta-1} \in (0, \infty)$ and is a decreasing function of θ . That is, when θ is larger convergence is slower.

proof: For simplicity suppose $x^* = 0 \Rightarrow f(x^*) = 0$. Then we consider the transformed function $f(x + x^*) - f(x^*)$.

When $\theta = \frac{1}{2}$ the log

We consider $Q(t) = \log f(x(t))$. Note Note that.

$$Q'(t) = \nabla f(x(t)) \cdot x'(t) = -\|\nabla f(x(t))\|^2.$$

When $\theta = \frac{1}{2}$ the Logasimic inequality says that.

$$-Q'(t) \geq C Q(t)$$

Rearranging then integrating gives

$$-\int_0^t \frac{Q'(t)}{Q(t)} dt \geq Ct$$

$$\Rightarrow -\log Q(t) \geq Ct + D$$

$$\Rightarrow Q(t) \leq A e^{-at}, A, a > 0.$$

The Logasimic distance inequality gives

$$d(x(t), z)^2 \leq Q(t)$$

and so

$$d(x(t), z) \leq A e^{-at}$$

proving (1).

For $\theta > \frac{1}{2}$ we consider

$$Q(t)^{1-2\theta} = \int_0^t \frac{d}{dt} Q(t)^{1-2\theta} dt \rightarrow Q(0)^{1-2\theta}$$

$$= \int_0^t (1-2\theta) Q'(t) Q(t)^{-2\theta} dt + 1$$

$$Q(t)^{1-2\theta} = - \int_0^t (1-2\theta) \| \nabla f(x(s)) \|^2 f(x(s))^{-2\theta} ds + D$$

$$Q(t)^{1-2\theta} = (2\theta-1) \int_0^t \| \nabla f(x(s)) \|^2 f(x(s))^{-2\theta} ds + D$$

Note that $(2\theta-1) > 0$. Then:

$$\begin{aligned} Q(t)^{1-2\theta} &\geq (2\theta-1) \int_0^t C f(x(s))^{2\theta} f(x(s))^{-2\theta} ds + D \\ &\geq A t + D \end{aligned}$$

Rearranging (noting $1-2\theta < 0$) gives

$$Q(t) \leq A'(t+B)^{\frac{-1}{2\theta-1}}$$

Applying the distance inequality gives

$$d(x(t), z) \leq A'(t+B)^{\frac{1-\theta}{2\theta-1}}$$

as required. \square

For functions in normal crossing form

Now suppose that f is in normal crossing form.

$$f(x) = a(x) x_1^{n_1} \cdot \dots \cdot x_d^{n_d}, \quad n_i \in \{0, 1, 2, \dots\}$$

where $a(0) \neq 0$. Also assume $d \geq 2$.

Theorem (Theorem 3 Fuchs'): If f is as above and $\nabla f(0) = 0$ (i.e. at least one $n_i > 0$) then there exists a neighborhood V of 0 , $C > 0$ such that

$$\|\nabla f(x)\| \geq C |f(x)|^\theta$$

$$\text{where } \theta = 1 - \frac{1}{N}, \quad N = \sum_{i=1}^d n_i.$$

Note that if the hypothesis $\nabla f(0) = 0$ is satisfied we must have $N \geq 2$, and so $\theta \in [\frac{1}{2}, 1)$. Furthermore, an immediate corollary is that $\theta = \frac{1}{2}$ if and only if $f(x) = a(x)x_i^2$ or $f(x) = a(x)x_i x_j$ for some $i, j \in [d]$.

proof (Fuchs' p12): We have

$$\|\nabla f(x)\| =$$

~~Without loss~~

By relabelling coordinates suppose $f(x) = a(x) \cdot \prod_{i=1}^c x_i^{n_i}$ for $c \leq d$. Then we have

$$\nabla f(x) = \sum_{i=1}^c \frac{\partial f}{\partial x_i}(x) e_i^*$$

$$\leftarrow \sum_{i=1}^c \left(x_i^{n_i} \frac{\partial a}{\partial x_i} + n_i x_i^{n_i-1} a(x) \right) \prod_{j=1, j \neq i}^c x_j^{n_j} e_i^*$$

$$= \sum_{i=1}^c \left(x_i^{n_i} \frac{\partial a}{\partial x_i} + n_i x_i^{n_i-1} a(x) \right) \cdot x_1^{n_1} \cdot \dots \cdot \overset{\wedge}{x_i^{n_i}} \cdot \dots \cdot x_c^{n_c} e_i^*$$

$$+ \left(\prod_{i=1}^c x_i^{n_i} \right) \sum_{j=c+1}^d \frac{\partial a}{\partial x_j} e_j^*$$

* Fuchs (2019) "Resolution of singularities and geometric proofs of the Kojasowicz inequalities".

By slight abuse of notation:

$$\nabla f(x) = \prod_{i=1}^c x_i^{n_i} \cdot \left(\left(\sum_{i=1}^c \frac{\partial a}{\partial x_i} + n_i a(x) x_i^{-1} \right) e_i + \sum_{j=c+1}^k \frac{\partial a}{\partial x_j} e_j \right)$$

Then, by just dropping " $\sum_{j=c+1}^k \frac{\partial a}{\partial x_j} e_j$ " (note this is zero if $c=k$) the following is an equality we get

$$\|\nabla f(x)\|^2 \geq \left(\prod_{i=1}^c x_i^{2n_i} \right) \cdot \left(\sum_{i=1}^c \left(x_i \frac{\partial a}{\partial x_i} + n_i a(x) \right)^2 x_i^{-2} \right) \quad (1)$$

Since $a(x)$ is C^1 & $a(0) \neq 0$ there is a ball $B=B_\sigma(0)$ such that

$$|x_j \frac{\partial a}{\partial x_j}| \leq \frac{n_j}{2} |a(x)|, \quad \forall x \in B_\sigma, j=1, \dots, c$$

(this just follows because $g(x)=|x_j \frac{\partial a}{\partial x_j}|$ and $h(x)=\frac{n_j}{2} |a(x)|$ are continuous and $g(0) < h(0)$). Thus,

$$n_j |a(x)| = |n_j a(x) + x_j \frac{\partial a}{\partial x_j} - x_j \frac{\partial a}{\partial x_j}|$$

$$\leq |n_j a(x) + x_j \frac{\partial a}{\partial x_j}| + \alpha |x_j \frac{\partial a}{\partial x_j}|$$

$$\leq |n_j a(x)| + |x_j \frac{\partial a}{\partial x_j}| + \frac{n_j}{2} |x_j \frac{\partial a}{\partial x_j}|$$

$$\Rightarrow \frac{n_j}{2} |a(x)| \leq |n_j a(x) + x_j \frac{\partial a}{\partial x_j}|$$

Since $n_j \geq 1$ for $j=1, \dots, c$, from (1) we obtain

$$\|\nabla f(x)\|^2 \geq \frac{a(x)^2}{4} \prod_{i=1}^c x_i^{2n_i} \sum_{j=1}^c x_j^{-2} \quad \forall x \in B_\sigma. \quad (A)$$

Separately
Define

$$m = \inf_{x \in \overline{B_\sigma}} |a(x)|, \quad M = \sup_{x \in \overline{B_\sigma}} |a(x)|$$

Note that since $a(x) \neq 0$ on $x \in \overline{B_\sigma}$ $a(\overline{B_\sigma})$ is compact $a(x) > 0$. Likewise $M < \infty$.

We first handle the case that $c=1$. In this case (A) becomes

$$\|\nabla f(x)\|^2 \geq \frac{a(x)^2}{4} x_1^{2(n_1-1)}$$

Noting $f(x) = a(x) x_1^{n_1}$ Taking square roots & using we give

$$\|\nabla f(x)\| \geq \frac{1}{2} |a(x)| x_1^{n_1-1}$$

Since $|f(x)| = |a(x)| x_1^{n_1}$

Then

$$\|\nabla f(x)\| \geq \frac{1}{2} |a(x)| |x_1|^{n_1-1}$$

$$\geq \frac{1}{2} |a(x)|^{\frac{1}{n_1}} |a(x) x_1^{n_1}|^{1-\frac{1}{n_1}}$$

$$\geq \frac{1}{2} m^{\frac{1}{n_1}} |f(x)|^{1-\frac{1}{n_1}}$$

which proves the result when $c=1$.

Now suppose $c>1$ and set $N = \sum_{i=1}^c n_i$. Recall:

Lemma (Generalized Young Inequality): For $c \geq 2$, $a_j > 0$, $p_j > 0$ and $r > 0$ st $\frac{1}{r} = \sum_{i=1}^c \frac{1}{p_i}$ we have:

$$\left(\prod_{i=1}^c a_i \right)^r \leq r \sum_{i=1}^c \frac{a_i p_i}{p_i} \quad (2)$$

Then we have that

$$\left(\prod_{i=1}^c x_i \right)^{-\frac{2n_i}{N}} \leq \frac{1}{N} \sum_{j=1}^c n_j x_j^{-2} \quad (2)$$

is true as an instance of (2) with $a_j = x_j^{-\frac{2n_i}{N}}$, $p_j = \frac{N}{n_j}$ and $r=1$. Setting $n = \max\{n_j\}$, we immediately get

$$\left(\prod_{i=1}^c x_i \right)^{-\frac{2n_i}{N}} \leq \frac{n}{N} \sum_{j=1}^c x_j^{-2} \quad (3)$$

From part of the RHS of (A), applying (3) gives us

$$\prod_{i=1}^c x_i^{2n_i} \cdot \sum_{j=1}^c x_j^{-2} \geq \prod_{i=1}^c x_i^{2n_i} - \prod_{i=1}^c x_i^{-\frac{2n_i}{N}} \cdot \frac{N}{n}$$

$$\geq \frac{N}{n} \prod_{i=1}^c x_i^{2n_i(1-\frac{1}{N})}$$

Thus, from (A) above we get

$$\|\nabla f(x)\| \geq \frac{1}{2} |a(x)| \cdot \sqrt{\frac{N}{n}} \left(\prod_{i=1}^c |x_i|^{n_i} \right)^{1-\frac{1}{N}}$$

$$\geq \frac{1}{2} |a(x)| \cdot \sqrt{\frac{N}{n}} \left(|a(x)| \cdot \prod_{i=1}^c |x_i|^{n_i} \right)^{1-\frac{1}{N}}$$

$$\geq \frac{1}{2} m^{1/N} \cdot \sqrt{\frac{N}{n}} \|f(x)\|^{1-\frac{1}{N}}$$

which completes the proof. \square

We now consider a general analytic map f . For simplifying suppose $f(0)=0$ and $\nabla f(0)=0$. Let $g: \tilde{U} \rightarrow \mathbb{C}^d$ be a ~~continuous~~ \mathcal{C}^1 function of singularities of f in a neighbourhood of the origin:

$$f \circ g(u) = a(u) \cdot \prod_{i=1}^c u_i^{n_i}$$

where $c \leq d$.

of (logarithmic gradient inequality, Fulton p17): let $N = \sum_{i=1}^c n_i$. Applying the previous theorem gives

$$\|(\nabla f \circ g)(u)\| \geq C \|f \circ g(u)\|^{1-\frac{1}{N}}, \quad u \in \tilde{V} \subset \tilde{U}$$

at $B_\sigma \subset \tilde{U}$.

Let $x \in U$, and $u \in \tilde{U}$ s.t. $x = g(u)$

Now g is an open map (?) so $g(B_\sigma) = V$ is open. Let $x \in V$ and $u \in B_\sigma$ be such that $x = g(u)$.

Using the chain rule

$$\|(\nabla f \circ g)(u)\| = \|\nabla f(g(u))\| \cdot Jg(u) \leq \underbrace{\|Jg(u)\|}_{\leq M} \cdot \|\nabla f(x)\|$$

Combining these results gives

$$\|\nabla f(\bar{x})\| \leq \frac{L}{N}$$

$$\|\nabla f(x)\| \geq L^{-\frac{1}{N}}$$

so we have (A) now note

$$\left(\frac{\|\nabla f(\bar{x})\|}{N} \right)^{\frac{1}{N}} \leq \frac{1}{N} \leq \|\nabla f(\bar{x})\|$$

$$\left(\frac{\|\nabla f(\bar{x})\|}{N} \right)^{\frac{1}{N}} \leq \|\nabla f(\bar{x})\| \leq \frac{1}{N}$$

$$\|\nabla f(\bar{x})\| \leq \frac{1}{N^{\frac{1}{N}}} = \frac{1}{e}$$

□

long and somewhat bulky

minimize over \bar{x} of given objective function & subject to the
constraints $\nabla f(\bar{x}) = 0$ and $\bar{x} \in \mathbb{R}^d$. This is called a
constrained optimization problem.

$$\text{Now suppose } \bar{x} \text{ and } \nabla f(\bar{x}) = 0 \text{ and } \nabla^2 f(\bar{x}) = I_d$$

then $\nabla f(\bar{x}) = 0$ and $\nabla^2 f(\bar{x})$ is positive definite (assumption)

definition

$$\|\nabla f(\bar{x})\| \leq \|\nabla f(\bar{x})\|$$

and so

$\|\nabla f(\bar{x})\| = 0$ since $\nabla f(\bar{x}) = 0$

now if $V = (\nabla f(\bar{x}))^\top$ and suppose V is a matrix
then $V^\top V = 0$ since $V^\top V = V^\top (\nabla f(\bar{x}))^\top = 0$

so V is zero

$$\|\nabla f(\bar{x})\| \cdot \|\nabla f(\bar{x})\| \geq \|\nabla f(\bar{x})\| \cdot \|\nabla f(\bar{x})\| = \|\nabla f(\bar{x})\|^2$$