

THE PLENARY HULL OF THE GENERALIZED JACOBIAN MATRIX  
AND THE INVERSE FUNCTION THEOREM  
IN SUBDIFFERENTIAL CALCULUS

BY

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**Introduction.** Some of the most important recent advances in optimization have come about as a result of systematic replacement of smoothness assumptions by convexity. This is exemplified by the work of Rockafellar [6]. It is natural to ask whether analogous results can be proven without either smoothness or convexity. A general theory of necessary conditions for such problems has been obtained [1]. The conditions are expressed, in part, by means of generalized gradients.

The classical inverse function theorem gives conditions under which a  $C^r$  function admits (locally) a  $C^r$  inverse. The purpose of this paper is to give conditions under which a Lipschitzian (not necessarily differentiable) function admits (locally) a Lipschitzian inverse by means of the characterization of the plenary hull of the generalized Jacobian matrix.

**1. Locally Lipschitz functions.** Let  $f : B \rightarrow \mathbf{R}$  be locally Lipschitz on a bounded subset  $B$  of  $\mathbf{R}^n$ . It is known [7] that such a function has at almost all points  $x$  a derivative (gradient), which we denote by  $\nabla f(x)$ . It is easily verified that the function  $\nabla f$  is bounded on bounded subsets of its domain of definition.

Let now  $F : O \rightarrow \mathbf{R}^m$  be locally Lipschitz,  $O$  a nonempty open subset of  $\mathbf{R}^n$ . One is tempted to define the generalized derivative of  $F = (f_1, \dots, f_m)^t$  at  $x_0 \in O$  by simply considering  $[\partial f_1(x_0), \dots, \partial f_m(x_0)]^t$  (for undefined concepts the reader is referred to [1]).

The usual  $m \times n$  Jacobian matrix of partial derivatives, when it exists, is denoted by  $JF(x)$ . We topologize the vector space of  $m \times n$  matrices with the norm

$$\|M\| = \max |m_{ij}| \quad \text{where } M = (m_{ij}), \quad 1 \leq i \leq m, \quad 1 \leq j \leq n.$$

A mathematical tool is what F. H. Clarke called the generalized Jacobian matrix, defined in the following way:

DEFINITION 1.1. The generalized Jacobian matrix of  $F$  at  $x_0 \in O$ , denoted by  $\tilde{J}F(x_0)$ , is the *convex hull* of all matrices  $M$  of the form  $M = \lim_{n \rightarrow \infty} JF(x_n)$  where  $x_n$  converges to  $x_0$  in  $\text{dom } F'$ .

In this definition,  $\text{dom } F'$  denotes the subset of full measure of  $O$  where  $F$  is differentiable.

$\tilde{J}F(x_0)$  is a nonempty compact convex subset of the vector space of  $m \times n$  matrices, which reduces to  $\{JF(x_0)\}$  whenever  $F$  is  $C^1$  in some neighborhood of  $x_0$ .

DEFINITION 1.2.  $\tilde{J}F(x_0)$  is said to be of *maximal rank* if every  $M$  in  $\tilde{J}F(x_0)$  is of maximal rank.

**2. Plenary hull of the generalized Jacobian matrix.** Let us denote by  $\langle\langle \cdot, \cdot \rangle\rangle$  the inner product on the vector space of  $m \times n$  matrices defined by  $\langle\langle M, U \rangle\rangle = \text{Trace of } M \circ U^t$ ; it follows from Definition 1.1 that for all  $U \in \mathbb{R}^{m \times n}$

$$(2.1) \quad \max_{M \in \tilde{J}F(x_0)} \langle\langle M, U \rangle\rangle = \limsup_{\substack{x \rightarrow x_0 \\ x \in \text{dom } F'}} \langle\langle JF(x), U \rangle\rangle .$$

Consider  $U \in \mathbb{R}^{m \times n}$  of the form  $u \otimes v : x \rightarrow \langle u, x \rangle v$ , where  $u \in \mathbb{R}^n$  and  $v \in \mathbb{R}^m$ . Then  $\langle\langle M, u \otimes v \rangle\rangle$  reduces to  $\langle Mu, v \rangle$  and (2.1) can be rephrased as

$$(2.2) \quad \max_{M \in \tilde{J}F(x_0)} \langle Mu, v \rangle = \limsup_{\substack{x \rightarrow x_0 \\ x \in \text{dom } F'}} \langle JF(x)u, v \rangle .$$

We can use results on chain rules so that the left-hand of (2.2) appears as the generalized gradient of a particular real-valued function. Given  $v \in \mathbb{R}^m$ , the *generalized gradient* of  $F_v : x \rightarrow \langle F(x), v \rangle$  at  $x_0$  can be exactly described as  $\partial F_v(x_0) = \tilde{J}^t F(x_0)v$  ([3]). Therefore, for all  $u \in \mathbb{R}^n$ , we have

$$(2.3) \quad \max_{M \in \tilde{J}F(x_0)} \langle u, M^t v \rangle = F_v^\circ(x_0; u) \quad (\text{see [1, Definition 1.3]}).$$

Although  $\tilde{J}F(x_0)$  is convex and compact, one generally cannot separate an  $M_0$  from  $\tilde{J}F(x_0)$  by using only linear mappings (in  $\mathbb{R}^{m \times n}$ ) of the form  $u \otimes v$ ,  $u \in \mathbb{R}^n$ ,  $v \in \mathbb{R}^m$ . This led Sweetser [8] to introduce the following definition:

DEFINITION 2.1. A subset  $A \subset \mathbb{R}^{m \times n}$  is *plenary* if it includes every  $M$  in  $\mathbb{R}^{m \times n}$  satisfying  $Mu \in Au$  for all  $u \in \mathbb{R}^n$ .

Since the intersection of plenary sets is plenary, Sweetser defined the *plenary hull* of  $A$ , denoted  $\text{plen } A$ , as the smallest plenary set containing  $A$ . When  $\min(m, n) > 1$ ,  $\text{plen } \tilde{J}F(x_0)$  is a convex compact (plenary) set of matrices containing  $\tilde{J}F(x_0)$ .

Since  $\tilde{J}F(x_0)u = [\text{plen } \tilde{J}F(x_0)]u$  for all  $u \in \mathbf{R}^n$ , Hiriart-Urruty and Thibault [4] formulated the following theorem:

**THEOREM 2.1.** *Let  $u \in \mathbf{R}^n$  and  $v \in \mathbf{R}^m$ . Then*

$$(2.4) \quad \max_{M \in \text{plen } \tilde{J}F(x_0)} \langle Mu, v \rangle = F^\circ(x_0; u, v).$$

In other words,  $M \in \text{plen } \tilde{J}F(x_0)$  if and only if

$$\langle Mu, v \rangle \leq F^\circ(x_0; u, v) \quad \text{for all } (u, v) \in \mathbf{R}^n \times \mathbf{R}^m.$$

To summarize, let us say that  $\text{plen } \tilde{J}F(x_0)$  is the convex compact (plenary) set of matrices satisfying  $[\text{plen } \tilde{J}F(x_0)]u = \tilde{J}F(x_0)u$  for all  $u \in \mathbf{R}^n$ . When  $F = (f_1, \dots, f_m)^t$  we have

$$\tilde{J}F(x_0) \subset \text{plen } \tilde{J}F(x_0) \subset [\partial f_1(x_0), \dots, \partial f_m(x_0)]^t.$$

The set  $[\partial f_1(x_0), \dots, \partial f_m(x_0)]^t$  is obviously convex, compact and plenary. It actually yields the same image set as  $\tilde{J}F(x_0)$  does when the considered vectors  $u$  are the elements  $e_i$  of the canonical basis in  $\mathbf{R}^n$ . In other words,

$$\{x_i^*, [x_1^*, \dots, x_i^*, \dots, x_m^*]^t \in \tilde{J}F(x_0)\} = \partial f_i(x_0) \quad ([3]).$$

### 3. The plenary hull of $\tilde{J}F(x_0)$ and the inverse function theorem.

**THEOREM 3.1.** *Let  $F : O \rightarrow \mathbf{R}^n$ ,  $O \subset \mathbf{R}^n$ . If every matrix  $M$  in  $\text{plen } \tilde{J}F(x_0)$  is of maximal rank, then there exist neighborhoods  $U$  and  $V$  of  $x_0$  and  $F(x_0)$  respectively, and a Lipschitzian function  $G : V \rightarrow \mathbf{R}^n$  such that for all  $(u, v) \in U \times V$ ,  $F(u) = v$  if and only if  $G(v) = u$ .*

When  $F$  is  $C^1$ ,  $\tilde{J}F(x_0)$  reduces to  $JF(x_0)$  and the function  $G$  above is necessarily  $C^1$  as well. Thus we recover the classical theorem.

**Remark 1.** This theorem remains true (without modifications in the proof) if we impose the maximality of rank for all  $M \in \tilde{J}_\Lambda F(x_0)$  where  $\Lambda \subset \text{dom } F'$  has complement in  $O$  of null measure and  $\tilde{J}_\Lambda F(x_0)$  is defined as in (1.1) except that the points  $x_n$  belong to  $\Lambda$  only.

**Remark 2.** By the very definitions,

$$\partial_\Lambda F_v(x_0) = \tilde{J}_\Lambda^t F(x_0)v \quad ([3]).$$

It is known that the generalized gradient of a real-valued function does not change when we alter the values of the function on a set of null measure [1, Proposition 1.11]. The desire to make the generalized derivative insensitive to sets of null measure led B. H. Pourciau [5] to alter Clarke's original definition by considering the Lebesgue set  $\text{Leb } F'$  of  $F'$  instead of  $\text{dom } F'$  in the definition of  $\tilde{J}F(x_0)$ , but, since  $F'$  is locally in  $L^\infty(O, \mathbf{R}^m)$ , almost every  $x$  in  $\text{dom } F'$  belongs to  $\text{Leb } F'$ .

**Remark 3.** Let  $\Lambda$  and  $\tilde{J}_\Lambda F(x_0)$  be as in Remark 1. Then  $\text{plen } \tilde{J}_\Lambda F(x_0) = \text{plen } \tilde{J}F(x_0)$ . So,  $\text{plen } \tilde{J}F(x_0)$  is insensitive to sets of measure zero.

**Proof of Theorem 3.1.**

**LEMMA 1** (An exact chain rule in the finite-dimensional case [5].) *Let  $F : \mathbf{R}^n \rightarrow \mathbf{R}^m$  be a locally Lipschitz function and let  $g : \mathbf{R}^m \rightarrow \mathbf{R}$  be continuously differentiable. Then*

$$\partial(g \circ F) = \tilde{J}^t F(x_0) \nabla g(F(x_0)).$$

**LEMMA 2.** *Let  $\beta$  be a positive number. Then for all  $x$  sufficiently near  $x_0$ ,  $\text{plen } \tilde{J}F(x) \subset [\text{plen } \tilde{J}F(x_0) + \beta M(0, 1)]$  where  $M(0, 1)$  denotes the unit ball in the vector space of  $m \times n$  matrices.*

This is a direct consequence of the definition of the plenary hull of the generalized Jacobian matrix.

**LEMMA 3.** *There are positive numbers  $r$  and  $\lambda$  with the following property: given any unit vector  $v$  in  $\mathbf{R}^n$ , there is a unit vector  $u$  in  $\mathbf{R}^n$  such that, whenever  $x$  lies in  $x_0 + rB$  and  $M \in \text{plen } \tilde{J}F(x_0)$ , then  $\langle Mv, u \rangle \geq \lambda$  for all  $M$ , where  $B$  denotes the open unit ball in  $\mathbf{R}^n$ .*

**Proof.** Let  $\Sigma_1$  denote the unit sphere in  $\mathbf{R}^n$ . Then the subset  $(\text{plen } \tilde{J}F(x_0))\Sigma_1$  of  $\mathbf{R}^n$  is compact and does not contain 0 since  $\text{plen } \tilde{J}F(x_0)$  is of maximal rank.

Hence for some  $\lambda > 0$ ,  $(\text{plen } \tilde{J}F(x_0))\Sigma_1$  is distant at least  $2\lambda$  from 0. For positive  $\beta$  sufficiently small,  $[\text{plen } \tilde{J}F(x_0) + \beta M(0, 1)]\Sigma_1$  is distant at least  $\lambda$  from 0.

By Lemma 2, it follows that for some positive  $r$ ,

$$x \in x_0 + rB \Rightarrow \text{plen } \tilde{J}F(x) \subset \text{plen } \tilde{J}F(x_0) + \beta M(0, 1).$$

We may suppose  $r$  chosen so that  $F$  satisfies the Lipschitz condition on  $x_0 + r\bar{B}$ .

Now let a unit vector  $v$  be given. It follows from the above that the convex set  $[\text{plen } \tilde{J}F(x_0) + \beta M(0, 1)]v = [\tilde{J}F(x_0) + \beta M(0, 1)]v$ , for all  $v$  in  $\mathbf{R}^n$ , is distant at least  $\lambda$  from 0. By the usual separation theorem for convex sets, there is a unit vector  $u$  such that  $\langle u, Mv \rangle \geq \lambda$  for all  $M \in \text{plen } \tilde{J}F(x)$ .

**LEMMA 4.** *If  $x_1$  and  $x_2$  lie in  $x_0 + r\bar{B}$ , then*

$$\|F(x_1) - F(x_2)\| \geq \lambda \|x_1 - x_2\|.$$

**Proof.** We may suppose  $x_1 \neq x_2$  and by the continuity of  $F$  that  $x_1, x_2 \in x_0 + rB$ .

Set  $v = (x_2 - x_1)/\|x_2 - x_1\|$ ,  $\alpha = \|x_2 - x_1\|$  so that  $x_2 = x_1 + \alpha v$ . Let  $\pi$  be the plane perpendicular to  $v$  and passing through  $x_1$ . The set  $P$  of

points  $x$  in  $x_0 + rB$  where  $F'$  fails to exist is of measure zero, and hence by Fubini's theorem, for almost every  $x$  in  $\pi$  the ray  $x + tv$ ,  $t \geq 0$ , meets  $P$  in a set of null one-dimensional measure. Choose an  $x$  with the above property and sufficiently close to  $x_1$  so that  $x + tv$  lies in  $x_0 + rB$  for every  $t$  in  $[0, \alpha]$ . Then the function  $t \rightarrow F(x + tv)$  is Lipschitzian for  $t$  in  $[0, \alpha]$  and has a.e. on this interval the derivative  $JF(x + tv)v$ . Thus

$$F(x + \alpha v) - F(x) = \int_0^\alpha JF(x + tv)v dt.$$

Let  $u$  be as in Lemma 3. We deduce that

$$\langle u, F(x + \alpha v) - F(x) \rangle = \left\langle u, \int_0^\alpha JF(x + tv)v dt \right\rangle \geq \int_0^\alpha \lambda dt = \lambda \alpha.$$

Recalling the definition of  $\alpha$ , we arrive at

$$\|F(x + \alpha v) - F(x)\| \geq \lambda \|x_2 - x_1\|.$$

This may be done for  $x$  arbitrarily close to  $x_1$ . Since  $F$  is continuous, the lemma follows.

**LEMMA 5.**  $F(x_0 + rB)$  contains  $F(x_0) + (r\lambda/2)B$ .

**PROOF.** Let  $y$  be any point in  $F(x_0) + (r\lambda/2)B$ , and let the minimum of  $\|y - F(x)\|^2$  over  $x_0 + r\bar{B}$  be attained at  $x$ . We claim that  $x$  belongs to  $x_0 + rB$ . Indeed, otherwise

$$\begin{aligned} r\lambda/2 &> \|y - F(x_0)\| \geq \|F(x) - F(x_0)\| - \|y - F(x)\| \\ &\geq \lambda \|x - x_0\| - \|y - F(x)\| \\ &\geq \lambda r - \|y - F(x_0)\| > \lambda r - r\lambda/2 = r\lambda/2, \end{aligned}$$

which is a contradiction. Thus  $x$  yields a local minimum for the function  $\|y - F(x)\|^2$ , and consequently [1, Corollary 1.10],

$$0 \in \partial \|y - F(x)\|^2.$$

We now use Lemma 1 to conclude that 0 belongs to the set

$$\tilde{J}^t F(x)(y - F(x)) \quad ([3]),$$

which coincides with  $[\text{plen } \tilde{J}^t F(x)](y - F(x))$  for all vectors in  $\mathbf{R}^n$  by Theorem 2.1. But Lemma 3 implies that every matrix in  $\text{plen } \tilde{J} F(x)$  is nonsingular, hence the above is possible only if  $F(x) = y$ .

We now set  $V = F(x_0) + (r\lambda/2)B$ , and we define  $G$  on  $V$  as follows:  $G(v)$  is the unique  $x$  in  $x_0 + rB$  such that  $F(x) = v$ . For  $U$  we can choose any neighborhood of  $x_0$  satisfying  $F(U) \supset V$ . The theorem is now seen to follow, since Lemma 4 implies that  $G$  is Lipschitz with constant  $\lambda^{-1}$ .

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