## The implicit function theorem and convexity

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Let  $\mathbf{f} : \mathbf{R}^{n+m} \to \mathbf{R}^m$  be a continuously differentiable function.  $\mathbf{R}^{n+m}$  is the direct sum  $\mathbf{R}^n + \mathbf{R}^m$ , a point of this is  $(\mathbf{x}, \mathbf{y}) = (\mathbf{x}_1, \dots, \mathbf{x}_n, \mathbf{y}_1, \dots, \mathbf{y}_m)$ . Starting from the given function  $\mathbf{f}$ , the goal is to construct a function  $\mathbf{g}\mathbf{R}^n \to \mathbf{R}^m$  whose graph  $(\mathbf{x}, g(\mathbf{x}))$  is precisely the set of all  $(\mathbf{x}, \mathbf{y})$  such that  $\mathbf{f}(\mathbf{x}, \mathbf{y}) = \mathbf{0}$ .

Fix a point  $(\mathbf{a}, \mathbf{b}) = (a_1, \ldots, a_n, b_1, \ldots, b_m)$  with  $\mathbf{f}(\mathbf{a}, \mathbf{b}) = \mathbf{0}$ , where  $\mathbf{0}$  is the element of  $\mathbf{R}^m$ . If the matrix  $[(\partial f_i / \partial y_j)(\mathbf{a}, \mathbf{b})]$  is invertible, then there exists an open set U containing  $\mathbf{a}$ , an open set V containing  $\mathbf{b}$ , and a unique function g  $U \to V$  such that

$$\{\mathbf{x}, \mathbf{g}(\mathbf{x}) \mid \mathbf{x} \in \mathbf{U}\} = \{(\mathbf{x}, \mathbf{y}) \mid \mathbf{x} \in \mathbf{U}, \mathbf{f}(\mathbf{x}, \mathbf{y}) = \mathbf{0}\}$$

which is continuously differentiable.

In this work, it is stated that there always exists a scalar,  $C^2$  function  $F : \mathbb{R}^{n+m} \to \mathbb{R}^1$ whose partial derivative with respect to the *m* parameters of **y** is  $\mathbf{f} : \mathbb{R}^{n+m} \to \mathbb{R}^m$  and whose second partial derivative with respect to the *m* parameters of **y** is the Hessian according to the conditions of the theorem. Depending on the Hessian, the critical point where  $\mathbf{f} = \mathbf{0}$  can be a local sub-minimum, a sub-maximum and may have several other local sub-geometries according to the Morse theorem. The implicit function collects the critical points of a local projection map.

## Applications:

- 1. Any division  $(\mathbf{x}, \mathbf{y}) = (x_1, \dots, x_n, y_1, \dots, y_m)$  where the type of the critical point is a minimum for every value of  $\mathbf{x}$  may lead to a hierarchical minimisation consisting of two kinds of smaller dimensional minimisations.
- 2. The graph of the implicit function assigns a transformed merit function (a section of the original merit function) which is minimised in the second minimisation with respect to  $\mathbf{x}$  (during of which the **y**-part is eliminated in every iteration step with respect to  $\mathbf{x}$  by sub-minimisation with respect to  $\mathbf{y}$ ).
- 3. The solution of hierarchical minimisation is identical to the solution of the original minimisation if the type of the critical point is a global minimum for every value of  $\mathbf{x}$

The globally positive definite cases:

- 5. In case of strictly convex  $F : \mathbf{R}^{n+m} \to \mathbf{R}^1$ , for any division  $(\mathbf{x}, \mathbf{y}) = (x_1, \dots, x_n, y_1, \dots, y_m)$ , the Hessian is globally positive definite for every value of  $\mathbf{x}$ .
- 6. In case of linearly dependent parameters, the second partial derivative with respect to the m parameters of  $\mathbf{y}$  is globally positive definite for every value of  $\mathbf{x}$ .