

2 ODE (ordinary differential equation)

Theorem 2.1.6

Let $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ be differentiable. Let $x_0 \in \mathbb{R}$ and $y_0 \in \mathbb{R}^2$. Then the ODE $y' = F(x, y)$ has a **unique solution** f defined on a "largest" open interval I containing x_0 such that $f(x_0) = y_0$.

Definition 2.2.1

Let $I \subset \mathbb{R}$ be an open interval and $k \in \mathbb{N}_0$. An **homogeneous linear ODE of order k** on I is of the form $y^{(k)} + a_{k-1}y^{(k-1)} + \dots + a_1y' + a_0y = 0$ where the coefficients a_0, \dots, a_{k-1} are complex-valued functions on I , and the unknown is a function $I \rightarrow \mathbb{C}$ that is k -times differentiable on I . An equation of the form $y^{(k)} + a_{k-1}y^{(k-1)} + \dots + a_1y' + a_0y = b$, where $b : I \rightarrow \mathbb{C}$ is another function, is called an **inhomogeneous linear ODE**.

Recognize an ODE

1. no coefficients before the highest derivative
2. all coefficients are continuous
3. no products of y or their derivatives
4. no powers of y or their derivatives
5. no functions depending on y or their derivatives

Proposition 2.3.1

Any solution of $y' + ay = 0$ is of the form $f(x) = z \exp(-A(x))$ where A is a primitive of a . The unique solution with $f(x_0) = y_0$ is $f(x) = y_0 \exp(A(x_0) - A(x))$.

Solving inhomogeneous equations

Case 1: Make a guess. For example $y' = y + x^2$ guess $f(x) = ax^2 + bx + c$, and solve the equation.

Case 2: Use the variation of the constant. Assume $f_p = z(x) \exp(-A(x))$ for $z : I \rightarrow \mathbb{C}$. Then $z'(x) = b(x) \exp(A(x)) \implies k(x) = \int b(x) \exp(A(x)) dx$.

Definition Linear differential equations with constant coefficients

Let $k \in \mathbb{N}_0$, $a_0, \dots, a_{k-1} \in \mathbb{C}$ fixed and b a general continuous function, then $y^{(k)} + a_{k-1}y^{(k-1)} + \dots + a_1y' + a_0y = b$ is such equation.

Solution of hom. diff. eq. with constant coefficients

Look for solutions of the form $f(x) = e^{\alpha x}$ for $\alpha \in \mathbb{C}$. Then we have $f^{(j)}(x) = \alpha^j e^{\alpha x}$ for all $j \geq 0$ and for all x , which means that

$$\begin{aligned} & f^{(k)}(x) + a_{k-1}f^{(k-1)}(x) + \dots + a_1f'(x) + a_0f(x) \\ &= e^{\alpha x} (\alpha^k + a_{k-1}\alpha^{k-1} + \dots + a_1\alpha + a_0). \end{aligned}$$

This translates into finding the zeros of the characteristic polynomial:

$$\begin{aligned} P(X) &= X^k + a_{k-1}X^{k-1} + \dots + a_1X + a_0 \\ &= (X - \alpha_1) \dots (X - \alpha_k) = 0 \end{aligned}$$

Imaginary roots

If a root is not real i.e. $\alpha = \beta + i\gamma$, the solution $f(x) = e^{\alpha x}$ does not take real values, but $\bar{\alpha} = \beta - i\gamma$ is also a root, hence we can write $\tilde{f}_1(x) = e^{\beta x} \cos(\gamma x)$, $\tilde{f}_2(x) = e^{\beta x} \sin(\gamma x)$ instead of $f_1(x) = e^{\alpha x}$, $f_2(x) = e^{\bar{\alpha} x}$.

Multiple roots

Case 1: no multiple roots. Any solution of the equation is of the form $f(x) = z_1 e^{a_1 x} + \dots + z_k e^{a_k x}$.

Case 2: multiple roots. Suppose that α is a multiple root of order j with $2 \leq j \leq k$. Then the k functions $f_{\alpha,0}(x) = e^{\alpha x}$, $f_{\alpha,1}(x) = x e^{\alpha x}$, \dots , $f_{\alpha,j-1}(x) = x^{j-1} e^{\alpha x}$ are linearly independent solutions. Taking the union of the functions $f_{\alpha,j}$ for all roots of P , each with its multiplicity, gives a basis of the space of solutions.

3 Differential calculus in \mathbb{R}^n

Definition 3.2.1.

Let $(x_k)_{k \in \mathbb{N}}$ where $x_k \in \mathbb{R}^n$. Write $x_k = (x_{k,1}, \dots, x_{k,n})$. Let $y = (y_1, \dots, y_n) \in \mathbb{R}^n$. The sequence (x_k) **converges to** $(\rightarrow) y$ as $k \rightarrow +\infty$ if $\forall \varepsilon > 0$, if $\exists N \geq 1$ such that $\forall n \geq N$, we have $\|x_k - y\| < \varepsilon$.

Lemma 3.2.2.

$(x_k) \rightarrow y$ as $k \rightarrow +\infty \iff$ either: (1) $\forall i, 1 \leq i \leq n$, the sequence of real numbers $(x_{k,i}) \rightarrow y_i$. (2) The sequence of real numbers $\|x_k - y\| \rightarrow 0$ as $k \rightarrow +\infty$.

Definition 3.2.3.

Let $X \subset \mathbb{R}^n$ and $f : X \rightarrow \mathbb{R}^m$. (1) Let $x_0 \in X$. f is **continuous at x_0** if $\forall \varepsilon > 0 \exists \delta > 0$ s.t. $\|x - x_0\| < \delta \implies \|f(x) - f(x_0)\| < \varepsilon, \forall x \in X$. (2) f is **continuous** on X if it is continuous at $x_0 \forall x_0 \in X$.

Proposition 3.2.4.

Let $X \subset \mathbb{R}^n$ and $f : X \rightarrow \mathbb{R}^m$. Let $x_0 \in X$. The function f is continuous at $x_0 \iff \forall (x_k)_k \geq 1$ in X s.t. $x_k \rightarrow x_0$ as $k \rightarrow +\infty$, the sequence $(f(x_k))_k \geq 1$ in \mathbb{R}^m converges to $f(x_0)$.

Definition 3.2.5.

Let $X \subset \mathbb{R}^n$ and $f : X \rightarrow \mathbb{R}^m$. Let $x_0 \in X$ and $y \in \mathbb{R}^m$. We say that f has the **limit** y as $x \rightarrow x_0$ with $x \neq x_0$ if for every $\varepsilon > 0$, there exists $\delta > 0$, s.t. $\forall x \in X, x \neq x_0$, s.t. $\|x - x_0\| < \delta$, we have $\|f(x) - y\| < \varepsilon$. We then write $\lim_{x \rightarrow x_0, x \neq x_0} f(x) = y$.

Proposition 3.2.7.

Let $X \subset \mathbb{R}^n$ and $f : X \rightarrow \mathbb{R}^m$. Let $x_0 \in X$ and $y \in \mathbb{R}^m$. We have $\lim_{x \rightarrow x_0, x \neq x_0} f(x) = y \iff \forall (x_k) \in X$ s.t. $x_k \rightarrow x_0$ as $k \rightarrow +\infty$, and $x_k \neq x_0$, the sequence $(f(x_k))$ in \mathbb{R}^m converges to y .

Proposition 3.2.9.

Let $X \subset \mathbb{R}^n, Y \subset \mathbb{R}^m$ and $p \geq 1$ an integer. Let $f : X \rightarrow Y$ and $g : Y \rightarrow \mathbb{R}^p$ be continuous functions. Then the composite $g \circ f$ is continuous.

Definition 3.2.11.

(1) A subset $X \subset \mathbb{R}^n$ is **bounded** if the set of $\|x\|$ for $x \in X$ is bounded in \mathbb{R} . (2) A subset $X \subset \mathbb{R}^n$ is **closed** if for every sequence (x_k) in X that converges in \mathbb{R}^n to some vector $y \in \mathbb{R}^n$, we have $y \in X$. (3) A subset $X \subset \mathbb{R}^n$ is **compact** if it is bounded and closed.

Proposition 3.2.13.

Let $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a continuous map. For any closed set $Y \subset \mathbb{R}^m$, $f^{-1}(Y) = \{x \in \mathbb{R}^n : f(x) \in Y\} \subset \mathbb{R}^n$ is closed.

Theorem 3.2.15.

Let $X \subset \mathbb{R}^n$ be a non-empty compact set and $f : X \rightarrow \mathbb{R}$ a continuous function. Then f is bounded and achieves its max and min. I.e. $\exists x_+, x_- \in X$ s.t. $f(x_+) = \sup_{x \in X} f(x), f(x_-) = \inf_{x \in X} f(x)$.

Definition 3.3.1.

A subset $X \subset \mathbb{R}^n$ is **open** if, for any $x = (x_1, \dots, x_n) \in X$, there exists $\delta > 0$ such that the set

$$\{y = (y_1, \dots, y_n) \in \mathbb{R}^n : |x_i - y_i| < \delta \text{ for all } i\}$$

is contained in X . In other words: any point of \mathbb{R}^n obtained by changing any coordinate of x by at most δ is still in X .

Proposition 3.3.2.

A set $X \subset \mathbb{R}^n$ is open if and only if the complement $Y = \{x \in \mathbb{R}^n : x \notin X\}$ is closed.

Corollary 3.3.3.

If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is continuous and $Y \subset \mathbb{R}^m$ is open, then $f^{-1}(Y)$ is open in \mathbb{R}^n .

Definition 3.3.5

Let $X \subset \mathbb{R}^n$ be an open set. Let $f : X \rightarrow \mathbb{R}^m$ be a function. Let $1 \leq i \leq n$. We say that f has a **partial derivative** on X with respect to the i -th variable, or coordinate, if for all $x_0 = (x_{0,1}, \dots, x_{0,n}) \in X$, the function defined by

$$g(t) = f(x_{0,1}, \dots, x_{0,i-1}, t, x_{0,i+1}, \dots, x_{0,n})$$

on the set $I = \{t \in \mathbb{R} : (x_{0,1}, \dots, x_{0,i-1}, t, x_{0,i+1}, \dots, x_{0,n}) \in X\}$ is differentiable at $t = x_{0,i}$. Its **derivative** $g'(x_{0,i})$ at $x_{0,i}$ is denoted

$$\frac{\partial f}{\partial x_i}(x_0), \quad \partial_{x_i} f(x_0), \quad \partial_i f(x_0)$$

Proposition 3.3.7.

Consider $X \subset \mathbb{R}^n$ open and f, g functions from X to \mathbb{R}^m . Let $1 \leq i \leq n$.

(1) If f and g have partial derivatives with respect to the i -th coordinate on X , then $f + g$ also does, and

$$\partial_{x_i}(f + g) = \partial_{x_i}(f) + \partial_{x_i}(g).$$

(2) If $m = 1$, and if f and g have partial derivatives with respect to the i -th coordinate on X , then fg also does and

$$\partial_{x_i}(fg) = \partial_{x_i}(f)g + f\partial_{x_i}(g).$$

Furthermore, if $g(x) \neq 0$ for all $x \in X$, then f/g has a partial derivative with respect to the i -th coordinate on X , with

$$\partial_{x_i}(f/g) = (\partial_{x_i}(f)g - f\partial_{x_i}(g))/g^2.$$

Definition 3.3.9. (Jacobi matrix)

Let $X \subset \mathbb{R}^n$ open and $f : X \rightarrow \mathbb{R}^m$ a function with partial derivatives on X . Write

$$f(x) = (f_1(x), \dots, f_m(x)).$$

For any $x \in X$, the matrix

$$J_f(x) = (\partial_{x_j} f_i(x))_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}}$$

with m rows and n columns is called the **Jacobi matrix** of f at x .

Definition 3.3.11 (Gradient, Divergence)

Let $X \subset \mathbb{R}^n$ be open. (1) Let $f : X \rightarrow \mathbb{R}$ be a function. If all partial derivatives of f exist at $x_0 \in X$, then the column vector

$$\begin{pmatrix} \partial_{x_1} f(x_0) \\ \vdots \\ \partial_{x_n} f(x_0) \end{pmatrix}$$

is called the **gradient** at x_0 , and is denoted $\nabla f(x_0)$. (2) Let $f = (f_1, \dots, f_n) : X \rightarrow \mathbb{R}^n$ be a function with values in \mathbb{R}^n such that all partial derivatives of all coordinates f_i of f exist at $x_0 \in X$. Then the real number

$$\text{Tr}(J_f(x_0)) = \sum_{i=1}^n \partial_{x_i} f_i(x_0),$$

the trace of the Jacobi matrix, is called the **divergence** of f at x_0 , and is denoted $\text{div}(f)(x_0)$.

Definition 3.4.2.

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^m$ be a function. Let u be a linear map $\mathbb{R}^n \rightarrow \mathbb{R}^m$ and $x_0 \in X$. We say that f is **differentiable at x_0 with differential u** if

$$\lim_{x \rightarrow x_0} \frac{1}{\|x - x_0\|} (f(x) - f(x_0) - u(x - x_0)) = 0$$

where the limit is in \mathbb{R}^m . We then denote $df(x_0) = u$. If f is differentiable at every $x_0 \in X$, then we say that f is differentiable on X .

Proposition 3.4.4.

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^m$ be a function that is differentiable on X . (1) The function f is continuous on X . (2) The function f admits partial derivatives on X with respect to each variable. (3) Assume that $m = 1$. Let $x_0 \in X$, and let $u(x_1, \dots, x_n) = a_1 x_1 + \dots + a_n x_n$ be the

differential of f at x_0 . We then have $\partial_{x_i} f(x_0) = a_i$ for $1 \leq i \leq n$.

Proposition 3.4.6.

Let $X \subset \mathbb{R}^n$ be open, $f : X \rightarrow \mathbb{R}^m$ and $g : X \rightarrow \mathbb{R}^m$ differentiable functions on X . (1) The function $f + g$ is differentiable with differential $d(f + g) = df + dg$, and if $m = 1$, then fg is differentiable. (2) If $m = 1$ and if $g(x) \neq 0$ for all $x \in X$, then f/g is differentiable.

Proposition 3.4.7.

Let $X \subset \mathbb{R}^n$ be open, $f : X \rightarrow \mathbb{R}^m$ a function on X . If f has all partial derivatives on X , and if the partial derivatives of f are continuous on X , then f is differentiable on X , with differential determined by its partial derivatives, in the sense that the matrix of the differential $df(x_0)$, with respect to the canonical basis of \mathbb{R}^n and \mathbb{R}^m , is the Jacobi matrix of f at x_0 .

Proposition 3.4.9 (Chain rule)

Let $X \subset \mathbb{R}^n$ be open, $Y \subset \mathbb{R}^m$ be open, and let $f : X \rightarrow Y$ and $g : Y \rightarrow \mathbb{R}^p$ be differentiable functions. Then $g \circ f : X \rightarrow \mathbb{R}^p$ is differentiable on X , and for any $x \in X$, its differential is given by the composition

$$d(g \circ f)(x_0) = dg(f(x_0)) \circ df(x_0).$$

In particular, the Jacobi matrix satisfies

$$J_{g \circ f}(x_0) = J_g(f(x_0)) J_f(x_0)$$

where the right-hand side is a matrix product.

Definition 3.4.11.

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^m$ a function that is differentiable. Let $x_0 \in X$ and $u = df(x_0)$ be the differential of f at x_0 . The graph of the affine linear approximation

$$g(x) = f(x_0) + u(x - x_0)$$

from \mathbb{R}^n to \mathbb{R}^m , or in other words the set

$$\{(x, y) \in \mathbb{R}^n \times \mathbb{R}^m : y = f(x_0) + u(x - x_0)\}$$

is called the **tangent space** at x_0 to the graph of f .

Definition 3.4.13.

Let $X \subset \mathbb{R}^n$ be an open set and let $f : X \rightarrow \mathbb{R}^m$ be a function. Let $v \in \mathbb{R}^n$ be a non-zero vector and $x_0 \in X$. We say that f has **directional derivative $w \in \mathbb{R}^m$ in the direction v** , if the function g defined on

the set

$$I = \{t \in \mathbb{R} : x_0 + tv \in X\}$$

by

$$g(t) = f(x_0 + tv)$$

has a derivative at $t = 0$, and this is equal to w . In other words, this means that the limit

$$\lim_{\substack{t \rightarrow 0 \\ t \neq 0}} \frac{f(x_0 + tv) - f(x_0)}{t}$$

exists and is equal to w .

Proposition 3.4.15.

Let $X \subset \mathbb{R}^n$ be an open set and let $f : X \rightarrow \mathbb{R}^m$ be a differentiable function. Then for any $x \in X$ and non-zero $v \in \mathbb{R}^n$, the function f has a directional derivative at x_0 in the direction v , equal to $\text{df}(x_0)(v)$.

Definition 3.5.1.

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^m$. We say that f is of class C^1 if f is differentiable on X and all its partial derivatives are continuous. The set of functions of class C^1 from X to \mathbb{R}^m is denoted $C^1(X; \mathbb{R}^m)$. Let $k \geq 2$. We say, by induction, that f is of class C^k if it is differentiable and each partial derivative $\partial_{x_i} f : X \rightarrow \mathbb{R}^m$ is of class C^{k-1} . The set of functions of class C^k from X to \mathbb{R}^m is denoted $C^k(X; \mathbb{R}^m)$. If $f \in C^k(X; \mathbb{R}^m)$ for all $k \geq 1$, then we say that f is of class C^∞ . The set of such functions is denoted $C^\infty(X; \mathbb{R}^m)$.

Proposition 3.5.4 (Mixed derivatives commute)

$k \geq 2$. Let $X \subset \mathbb{R}^n$ be open and let $f : X \rightarrow \mathbb{R}^m$ be a function of class C^k . Then the partial derivatives of order k are independent of the order in which the partial derivatives are taken: for any variables x and y , we have $\partial_{x,y} f = \partial_{y,x} f$ and for any variables x, y, z , we have

$$\partial_{x,y,z} f = \partial_{x,z,y} f = \partial_{y,z,x} f = \partial_{z,x,y} f = \dots$$

Definition 3.5.9 (Hessian).

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ a C^2 function. For $x \in X$, the **Hessian matrix** of f at x is the symmetric square matrix

$$\text{Hess}_f(x) = (\partial_{x_i, x_j} f)_{1 \leq i, j \leq n}.$$

We also sometimes write simply $H_f(x)$.

Definition 3.7.1 (Taylor polynomials).

Let $k \geq 1$ be an integer. Let $f : X \rightarrow \mathbb{R}$ be a function of class C^k on X , and fix $x_0 \in X$. The k -th Taylor polynomial of f at the point x_0 is the polynomial in n variables of degree $\leq k$ given by

$$T_k f(y; x_0) = f(x_0) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x_0) y_i + \dots + \sum_{m_1 + \dots + m_n = k} \frac{1}{m_1! \dots m_n!} \frac{\partial^k f}{\partial x_1^{m_1} \dots \partial x_n^{m_n}}(x_0) y_1^{m_1} \dots y_n^{m_n}$$

where the last sum ranges over the tuples of n non-negative integers such that the sum is k .

Proposition 3.7.3 (Taylor approximation)

Let $k \geq 1$ be an integer. Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ be a function of class C^k . For x_0 in X , if we define $E_k f(x; x_0)$ by

$$f(x) = T_k f(x - x_0; x_0) + E_k f(x; x_0)$$

then we have

$$\lim_{\substack{x \rightarrow x_0 \\ x \neq x_0}} \frac{E_k f(x; x_0)}{\|x - x_0\|^k} = 0$$

Proposition 3.8.1.

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ a differentiable function. If $x_0 \in X$ is such that $f(y) \leq f(x_0)$ for all y close enough to x_0 (local maximum at x_0) or $f(y) \geq f(x_0)$ for all y close enough to x_0 (local minimum at x_0). Then we have $df(x_0) = 0$, or in other words $\nabla f(x_0) = 0$, or equivalently $\frac{\partial f}{\partial x_i}(x_0) = 0$ for $1 \leq i \leq n$.

Definition 3.8.2 (Critical point)

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ a differentiable function. A point $x_0 \in X$ such that $\nabla f(x_0) = 0$ is called a **critical point** of the function f .

Definition 3.8.6 (Non-degenerate critical point)

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ a function of class C^2 . A critical point $x_0 \in X$ of f is called **non-degenerate** if the Hessian matrix has non-zero determinant.

Corollary 3.8

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}$ a function of class C^2 . Let x_0 be a non-degenerate critical point of f . Let p and q be the number of positive and negative eigenvalues of $\text{Hess}_f(x_0)$. (1) If $p = n$, equivalently if $q = 0$, the function f has a local minimum at x_0 . (2) If $q = n$, equivalently if $p = 0$, the function f has a local maximum at x_0 . (3) Otherwise, equivalently if $pq \neq 0$, the function f does not have a local extremum at x_0 . One then says that f has a saddle point at x_0 .

Proposition 3.9.2

Let $X \subset \mathbb{R}^n$ be open and let $f : X \rightarrow \mathbb{R}$ and $g : X \rightarrow \mathbb{R}$ be functions of class C^1 . If $x_0 \in X$ is a local extremum of the function f restricted to the set $Y = \{x \in X : g(x) = 0\}$ then either $\nabla g(x_0) = 0$, or there exists $\lambda_0 \in \mathbb{R}$ such that

$$\begin{cases} \nabla f(x_0) = \lambda \nabla g(x_0) \\ g(x_0) = 0 \end{cases}$$

or in other words, there exists λ such that (x_0, λ) is a critical point of the differentiable function $h : X \times \mathbb{R} \rightarrow \mathbb{R}$ defined by $h(x, \lambda) = f(x) - \lambda g(x)$. Such a value λ is called a Lagrange multiplier at x_0 .

Definition 3.10.1 (Change of variable)

Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^n$ be differentiable. Let $x_0 \in X$. We say that f is a **change of variable** around x_0 if there is a radius $r > 0$ such that the restriction of f to the ball

$$B = \{x \in \mathbb{R}^n : \|x - x_0\| < r\}$$

of radius r around x_0 has the property that the image $Y = f(B)$ is open in \mathbb{R}^n , and if there is a differentiable map $g : Y \rightarrow B$ such that $f \circ g = \text{Id}_Y$ and $g \circ f = \text{Id}_B$.

Theorem 3.10.2 (Inverse function theorem)

(Inverse function theorem). Let $X \subset \mathbb{R}^n$ be open and $f : X \rightarrow \mathbb{R}^n$ differentiable. If $x_0 \in X$ is such that $\det(J_f(x_0)) \neq 0$, i.e., such that the Jacobian trix of f at x_0 is invertible, then f is a change of variable around x_0 . Moreover, the Jacobian of g at x_0 is determined by

$$J_g(f(x_0)) = J_f(x_0)^{-1}$$

In addition, if f is of class C^k , then g is of class C^k .

Theorem 3.10.4 (Implicit Function Theorem).

Let $X \subset \mathbf{R}^{n+1}$ be open and let $g : X \rightarrow \mathbf{R}$ be of class C^k with $k \geq 1$. Let $(x_0, y_0) \in \mathbf{R}^n \times \mathbf{R}$ be such that $g(x_0, y_0) = 0$. Assume that

$$\partial_y g(x_0, y_0) \neq 0$$

Then there exists an open set $U \subset \mathbf{R}^n$ containing x_0 , an open interval $I \subset \mathbf{R}$ containing y_0 , and a function $f : U \rightarrow \mathbf{R}$ of class C^k such that the system of equations

$$\begin{cases} g(x, y) = 0 \\ x \in U, \quad y \in I \end{cases}$$

is equivalent with $y = f(x)$. In particular, $f(x_0) = y_0$. Moreover, the gradient of f at x_0 is given by

$$\nabla f(x_0) = -\frac{1}{(\partial_y g)(x_0, y_0)} \nabla_x g(x_0, y_0)$$

where $\nabla_x g = (\partial_{x_1} g, \dots, \partial_{x_n} g)$