

Proposition 9.34 (Chain Rule)

Let D be open in \mathbb{R}^n and E be open in \mathbb{R}^m . Let $f: D \rightarrow E$ and $g: E \rightarrow \mathbb{R}^p$ be two functions, $f = \begin{pmatrix} f_1 \\ \vdots \\ f_m \end{pmatrix}$, $g = \begin{pmatrix} g_1 \\ \vdots \\ g_p \end{pmatrix}$. Assume that f is differentiable on D and g is differentiable on E . Then the function $h := g \circ f: D \rightarrow \mathbb{R}^p$, $h = \begin{pmatrix} h_1 \\ \vdots \\ h_p \end{pmatrix}$ is differentiable on D and for all $x = (x_1 \dots x_n) \in D$,

$$J_h(x) = J_g(f(x)) \cdot J_f(x).$$

Furthermore, for all $i=1, \dots, n$ and $k=1, \dots, p$

$$\begin{aligned} \frac{\partial h_k}{\partial x_i}(x) &= \sum_{j=1}^m \frac{\partial g_k}{\partial y_j}(f(x)) \cdot \frac{\partial f_j}{\partial x_i}(x) \\ &= \sum_{j=1}^m \frac{\partial g_k}{\partial y_j}(f_1(x), \dots, f_m(x)) \cdot \frac{\partial f_j}{\partial x_i}(x). \end{aligned}$$

Proof

The fact that for all $x \in D$, $J_h(x) = J_g(f(x)) \cdot J_f(x)$ follows immediately from Proposition 9.10 and the fact that the differential of a function at a point is represented by the Jacobi matrix of the function at that point. The equality $J_h(x) = J_g(f(x)) \cdot J_f(x)$ is equivalent with

$$\begin{pmatrix} \frac{\partial h_1}{\partial x_1}(x) & \dots & \frac{\partial h_1}{\partial x_n}(x) \\ \frac{\partial h_2}{\partial x_1}(x) & \dots & \frac{\partial h_2}{\partial x_n}(x) \\ \dots & \dots & \dots \\ \frac{\partial h_p}{\partial x_1}(x) & \dots & \frac{\partial h_p}{\partial x_n}(x) \end{pmatrix} = \begin{pmatrix} \frac{\partial g_1}{\partial y_1}(f(x)) & \dots & \frac{\partial g_1}{\partial y_m}(f(x)) \\ \frac{\partial g_2}{\partial y_1}(f(x)) & \dots & \frac{\partial g_2}{\partial y_m}(f(x)) \\ \dots & \dots & \dots \\ \frac{\partial g_p}{\partial y_1}(f(x)) & \dots & \frac{\partial g_p}{\partial y_m}(f(x)) \end{pmatrix} \cdot \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(x) & \dots & \frac{\partial f_1}{\partial x_n}(x) \\ \frac{\partial f_2}{\partial x_1}(x) & \dots & \frac{\partial f_2}{\partial x_n}(x) \\ \dots & \dots & \dots \\ \frac{\partial f_m}{\partial x_1}(x) & \dots & \frac{\partial f_m}{\partial x_n}(x) \end{pmatrix}$$

Then for all $i=1, \dots, n$ and $k=1, \dots, p$

$\frac{\partial h_k}{\partial x_i}(x)$ is obtained by multiplying the k th row of $J_f(f(x))$

with the i th column of $J_g(x)$. Thus

$$\frac{\partial h_k}{\partial x_i}(x) = \sum_{j=1}^m \frac{\partial h_k}{\partial y_j}(f(x)) \cdot \frac{\partial f_j}{\partial x_i}(x).$$

□

Corollary 3.35

Let D be open in \mathbb{R}^n and E be open in \mathbb{R}^m . Let $f: D \rightarrow E$ and $g: E \rightarrow \mathbb{R}$ be two functions s.t. f is differentiable on D and g is differentiable on E . Then the function $h := g \circ f: D \rightarrow \mathbb{R}$ is differentiable on D and for all $i=1, \dots, n$ and all $x \in D$

$$\frac{\partial h}{\partial x_i}(x) = \sum_{j=1}^m \frac{\partial g}{\partial y_j}(f(x)) \cdot \frac{\partial f_j}{\partial x_i}(x)$$

Proof

Apply Proposition 3.34 with $p=1$. □

Corollary 3.36

Let D be open in \mathbb{R}^n and E be open in \mathbb{R}^m . Let $f: D \rightarrow E$ and $g: E \rightarrow \mathbb{R}$ be two functions s.t. f is differentiable on D and g is differentiable on E . Then the real function $h := g \circ f: D \rightarrow \mathbb{R}$ is differentiable on D and for all $x \in D$

$$h'(x) = \sum_{j=1}^m \frac{\partial g}{\partial y_j}(f(x)) \cdot f_j'(x).$$

Proof Apply Corollary 3.35 with $n=1$. □

Example 8.37

(i) Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, $f(x) = \begin{pmatrix} x \\ x \end{pmatrix}$ and $g: \mathbb{R}^2 \rightarrow \mathbb{R}$, $g(y_1, y_2) = y_1^{y_2}$.

Define $h = g \circ f: \mathbb{R} \rightarrow \mathbb{R}$, $h(x) = g(f(x)) = g(x, x) = x^x$ for all $x \in \mathbb{R}$.

We have that $f = \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}$, where $f_1(x) = f_2(x) = x$ and $f'_1(x) = f'_2(x) = 1$

for all $x \in \mathbb{R}$. Furthermore, $\frac{\partial g}{\partial y_1}(y_1, y_2) = y_2 \cdot y_1^{y_2-1}$ and

$\frac{\partial g}{\partial y_2}(y_1, y_2) = y_1^{y_2} \cdot \log y_1$. It follows that for all $x \in \mathbb{R}$,

$$h'(x) = \sum_{j=1}^2 \frac{\partial g}{\partial y_j}(f(x)) \cdot f'_j(x) = \frac{\partial g}{\partial y_1}(x, x) \cdot f'_1(x) + \frac{\partial g}{\partial y_2}(x, x) \cdot f'_2(x) = x \cdot x^{x-1} + x^x \cdot \log x = x^x (\log x + 1).$$

(ii) Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}^2$, $f(x_1, x_2) = \begin{pmatrix} x_1^2 + x_2^2 \\ x_1 x_2 \end{pmatrix}$, $g: \mathbb{R}^2 \rightarrow \mathbb{R}$,

$$g(y_1, y_2) = y_1 y_2.$$

Define $h = g \circ f: \mathbb{R}^2 \rightarrow \mathbb{R}$, $h(x_1, x_2) = g(f(x_1, x_2)) = g(x_1^2 + x_2^2, x_1 x_2) = (x_1^2 + x_2^2) \cdot x_1 x_2 = x_1^3 x_2 + x_2^3 x_1$.

We have that $f = \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}$, where $f_1(x_1, x_2) = x_1^2 + x_2^2$, $f_2(x_1, x_2) = x_1 x_2$

Then for all $(a_1, a_2) \in \mathbb{R}^2$,

$$\begin{aligned} \frac{\partial h}{\partial x_1}(a_1, a_2) &= \frac{\partial g}{\partial y_1}(f(a_1, a_2)) \cdot \frac{\partial f_1}{\partial x_1}(a_1, a_2) + \frac{\partial g}{\partial y_2}(f(a_1, a_2)) \cdot \frac{\partial f_2}{\partial x_1}(a_1, a_2) \\ &= f_2(a_1, a_2) \cdot 2a_1 + f_1(a_1, a_2) \cdot a_2 = \\ &= 2a_1^2 a_2 + a_2(a_1^2 + a_2^2) = 3a_1^2 a_2 + a_2^3. \end{aligned}$$

$$\begin{aligned} \frac{\partial h}{\partial x_2}(a_1, a_2) &= \frac{\partial f}{\partial y_1}(f(a_1, a_2)) \cdot \frac{\partial f_1}{\partial x_2}(a_1, a_2) + \frac{\partial f}{\partial y_2}(f(a_1, a_2)) \cdot \frac{\partial f_2}{\partial x_2}(a_1, a_2) \\ &= f_2(a_1, a_2) \cdot 2a_2 + f_1(a_1, a_2) \cdot a_1 = \\ &= 2a_1 a_2^2 + a_1(a_1^2 + a_2^2) = 3a_1 a_2^2 + a_1^3. \end{aligned}$$

(iii) Let $f: \mathbb{R}^3 \rightarrow \mathbb{R}^3$, $f(x, y, t) = \begin{pmatrix} x^2 \\ x^2 y \\ e^z \end{pmatrix}$ and $g: \mathbb{R}^3 \rightarrow \mathbb{R}$,

$$g(u, v, w) = u^2 + v^2 - w^2.$$

Define $h = g \circ f: \mathbb{R}^3 \rightarrow \mathbb{R}$, $h(x, y, t) = x^4(1+y^2) - e^{2t}$.

We have that $f = \begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix}$, where $f_1(x, y, t) = x^2$, $f_2(x, y, t) = x^2 y$,
 $f_3(x, y, t) = e^z$.

Then for all $(x, y, t) \in \mathbb{R}^3$,

$$\begin{aligned} \frac{\partial h}{\partial x}(x, y, t) &= \frac{\partial g}{\partial u}(f(x, y, t)) \cdot \frac{\partial f_1}{\partial x}(x, y, t) + \frac{\partial g}{\partial v}(f(x, y, t)) \cdot \frac{\partial f_2}{\partial x}(x, y, t) \\ &\quad + \frac{\partial g}{\partial w}(f(x, y, t)) \cdot \frac{\partial f_3}{\partial x}(x, y, t) \\ &= 2f_1(x, y, t) \cdot 2x + 2f_2(x, y, t) \cdot 2xy + (-2f_3(x, y, t)) \cdot 0 \\ &= 4x^3 + 4x^3 y^2 = 4x^3(1+y^2). \end{aligned}$$

$$\begin{aligned} \frac{\partial h}{\partial y}(x, y, t) &= \frac{\partial g}{\partial u}(f(x, y, t)) \cdot \frac{\partial f_1}{\partial y}(x, y, t) + \frac{\partial g}{\partial v}(f(x, y, t)) \cdot \frac{\partial f_2}{\partial y}(x, y, t) \\ &\quad + \frac{\partial g}{\partial w}(f(x, y, t)) \cdot \frac{\partial f_3}{\partial y}(x, y, t) = 2f_1(x, y, t) \cdot 0 + \\ &\quad + 2f_2(x, y, t) \cdot x^2 + (-2f_3(x, y, t)) \cdot 0 = 2x^4 y. \end{aligned}$$

$$\frac{\partial h}{\partial t}(x, y, t) = \frac{\partial g}{\partial u}(f(x, y, t)) \cdot \frac{\partial f_1}{\partial t}(x, y, t) + \frac{\partial g}{\partial v}(f(x, y, t)) \cdot \frac{\partial f_2}{\partial t}(x, y, t)$$

$$+ \frac{\partial f}{\partial x} (f(x, y, t)) \cdot \frac{\partial f_3}{\partial z} (x, y, t) = 2f_1(x, y, t) \cdot 0 + 2f_2(x, y, t) \cdot 0 +$$

$$+ (-2f_3(x, y, t)) \cdot e^z = -2e^{2z}.$$

Hence, $\nabla h(x, y, t) = \begin{pmatrix} 4x^3(1+y^2) & 2x^4y & -2e^{2t} \end{pmatrix}.$

Let us compute the derivative of h at $(1, 2, 0)$. For all $v \in \mathbb{R}^3$ we have that

$$h'(1, 2, 0)(v) = \nabla h(1, 2, 0) \cdot v = \begin{pmatrix} 20 & 4 & -2 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = 20v_1 + 4v_2 - 2v_3.$$

□

As an application of the chain rule, we shall prove the Mean Value Theorem for real-valued functions.

Definition 9.38

For any $a, b \in \mathbb{R}^n$, the line segment joining a and b , denoted by $[a, b]$, is defined as

$$[a, b] = \{ (1-t)a + tb : t \in [0, 1] \}.$$

Thus, $[a, b]$ is the image of the affine path joining a and b (see Definition 8.1).

Theorem 9.38 (Mean Value Theorem)

Let D be open in \mathbb{R}^n , $f: D \rightarrow \mathbb{R}$ be differentiable and $a, b \in D$ be s.t. $[a, b] \subseteq D$. Then there exists $c \in [a, b]$ s.t.

$$f(b) - f(a) = f'(c)(b-a).$$

Proof

Let us consider the path joining a and b : $\gamma: [0, 1] \rightarrow \mathbb{R}^n$, $\gamma(t) = (1-t)a + tb = t(b-a) + a$. Then γ is an affine function, so γ is differentiable on $[0, 1]$ and for all $t \in [0, 1]$,

$$\gamma'(t)(v) = (b-a)v \quad \text{for all } v \in \mathbb{R}.$$

Define $F := f \circ \gamma: [0, 1] \rightarrow \mathbb{R}$. Since f and γ are differentiable, by the chain rule (Proposition 9.10) we get that F is differentiable and for all $t \in [0, 1]$,

$$F'(t) = f'(\gamma(t)) \circ \gamma'(t) = f'(\gamma(t))(b-a).$$

Since F is differentiable on $[0,1]$ we can apply the Mean Value Theorem from Analysis I to get a $\xi \in]0,1[$ s.t. $F(1) - F(0) = F'(\xi)$.

Since $F(1) = f(x(1)) = f(b)$ and $F(0) = f(x(0)) = f(a)$, we get that

$$f(b) - f(a) = F'(\xi) = f'(x(\xi)) (b-a) = f'(c) (b-a), \text{ where } c = x(\xi) \in]a,b[.$$

Remark 3.40

If D is a (closed) ball, then for all $a, b \in D$ we have that $[a,b] \subseteq D$.

Proof

Let $D = U_r(x_0)$ be an open ball and $a, b \in D$. For all $t \in]0,1[$,

$$\begin{aligned} \|(1-t)a + ts - x_0\| &= \|(1-t)a + ts - (1-t)x_0 - tx_0\| = \|(1-t)(a-x_0) + t(b-x_0)\| \\ &\leq \|(1-t)(a-x_0)\| + \|t(b-x_0)\| = (1-t)\|a-x_0\| + t\|b-x_0\| < (1-t)r + tr = r, \end{aligned}$$

so $(1-t)a + ts \in U_r(x_0)$.

Similarly for $D = \overline{U_r(x_0)}$.

□

Proposition 3.41

Let $D \subseteq \mathbb{R}^n$ be open and let $f: D \rightarrow \mathbb{R}^m$. Then the following statements are equivalent:

- (i) f is continuously differentiable on D ;
- (ii) f is partial differentiable on D and all the partial derivatives

$$\frac{\partial f}{\partial x_i} : D \rightarrow \mathbb{R}^m, \quad i=1, \dots, n$$

are continuous on D .

Proof

(i) \Rightarrow (ii) Assume that f is continuously differentiable on D .

That is, f is differentiable on D and the map

$$f': D \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m), \quad a \mapsto f'(a)$$

is continuous.

By Proposition 3.28, we have that f is partial differentiable on D .

Let us prove that for all $i=1, \dots, n$ the i -th partial derivative

$$\frac{\partial f}{\partial x_i} : D \rightarrow \mathbb{R}^m \text{ is continuous.}$$

Let $x, a \in D$. Then

$$\begin{aligned} \left\| \frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right\| &= \left\| D_{e_i} f(x) - D_{e_i} f(a) \right\| \stackrel{P.3.21}{=} \| f'(x)(e_i) - f'(a)(e_i) \| \\ &\leq \| f'(x) - f'(a) \| \cdot \| e_i \| = \\ &= \| f'(x) - f'(a) \|. \end{aligned}$$

Since f' is continuous, $\lim_{x \rightarrow a} f'(x) = f'(a)$, so $\lim_{x \rightarrow a} \| f'(x) - f'(a) \| = 0$.

It follows that $\lim_{x \rightarrow a} \left\| \frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right\| = 0$, that is $\lim_{x \rightarrow a} \frac{\partial f}{\partial x_i}(x) = \frac{\partial f}{\partial x_i}(a)$.

That is, $\frac{\partial f}{\partial x_i}$ is continuous at a . Since $a \in D$ was arbitrary, we get

that $\frac{\partial f}{\partial x_i}$ is continuous on D .

(ii) \Rightarrow (i) Assume that f is partial differentiable and all the partial derivatives $\frac{\partial f}{\partial x_i}, i=1, \dots, n$ are continuous. By Proposition 3.32, it follows

that f is differentiable on D and for all $a \in D$, $f'(a)$ has as associated

matrix the Jacobi matrix $Jf(a)$ of f at a .

It remains to prove that $f': D \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ is continuous.

Let $x, a \in D$. Then

$$\|f'(x) - f'(a)\| = \sup_{\|h\| \leq 1} \|f'(x)h - f'(a)h\| =$$

$$= \sup_{\|h\| \leq 1} \|\partial f(x) \cdot h - \partial f(a) \cdot h\|$$

$$= \sup_{\|h\| \leq 1} \left\| \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x) h_i - \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a) h_i \right\|$$

$$= \sup_{\|h\| \leq 1} \|h\| \cdot \left\| \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right) \right\|$$

$$\leq \left\| \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right) \right\| \text{ if we consider}$$

the $\|\cdot\|_2$ on \mathbb{R}^n , so $\|h\| \leq \|h\| \leq 1$.

$$\leq \sum_{i=1}^n \left\| \frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right\|.$$

Since $\lim_{x \rightarrow a} \left\| \frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right\| = 0$ for all $i=1, \dots, n$, we

get that $\lim_{x \rightarrow a} \|f'(x) - f'(a)\| = 0$, that is $\lim_{x \rightarrow a} f'(x) = f'(a)$.

Thus, f' is continuous on D . □

The Contraction Mapping Principle

Let (X, d) be a metric space. A function $T: X \rightarrow X$ is called Lipschitz or Lipschitzian if there is a constant $k > 0$ s.t. for all $x, y \in X$

$$(i) \quad d(T(x), T(y)) \leq k \cdot d(x, y).$$

The smallest number k for which (i) holds is called the Lipschitz constant of T .

Definition 9.42

A Lipschitzian function $T: X \rightarrow X$ with Lipschitz constant $k < 1$ is said to be a contraction.

Theorem 9.43 (Banach's Contraction Mapping Principle)

Let (X, d) be a complete metric space and let $T: X \rightarrow X$ be a contraction. Then T has a unique fixed point x_0 . Moreover, for each $x \in X$,

$$\lim_{n \rightarrow \infty} T^n(x) = x_0$$

and in fact for each $x \in X$

$$d(T^n(x), x_0) \leq \frac{k^n}{1-k} d(x, T(x)), \quad n=1, 2, \dots$$

Proof

Since T is a contraction, we know that for each $x \in X$,

$$d(T(x), T^2(x)) \leq K \cdot d(x, T(x)).$$

Adding $d(x, T(x))$ to both sides of the above yields

$$d(x, T(x)) + d(T(x), T^2(x)) \leq d(x, T(x)) + K \cdot d(x, T(x)),$$

which can be rewritten

$$d(x, T(x)) - K d(x, T(x)) \leq d(x, T(x)) - d(T(x), T^2(x)).$$

This in turn is equivalent to

$$d(x, T(x)) \leq \frac{1}{1-K} (d(x, T(x)) - d(T(x), T^2(x))).$$

Now define the functions $\varphi: X \rightarrow \mathbb{R}^+$ by setting $\varphi(x) = \frac{1}{1-K} d(x, T(x))$,

for $x \in X$. This gives us the basic inequality

$$d(x, T(x)) \leq \varphi(x) - \varphi(T(x)), \quad x \in M.$$

Therefore if we fix $x \in X$ and take $m, n \in \mathbb{N}$ with $n < m$, we obtain

$$d(T^n(x), T^{m+1}(x)) \leq \sum_{i=n}^m d(T^i(x), T^{i+1}(x)) \leq \varphi(T^n(x)) - \varphi(T^{m+1}(x)).$$

(Notice that the last inequality comes from cancellation in the telescoping sum.) In particular by taking $n=1$ and letting $m \rightarrow \infty$ we conclude that

$$\sum_{i=1}^{\infty} d(T^i(x), T^{i+1}(x)) \leq \varphi(T(x)) < \infty.$$

This implies that $(T^n(x))_{n \in \mathbb{N}}$ is a Cauchy sequence. Since X is complete there exists $x_0 \in X$ s.t.

$$\lim_{n \rightarrow \infty} T^n(x) = x_0$$

and since T is continuous

$$x_0 = \lim_{n \rightarrow \infty} T^n(x) = \lim_{n \rightarrow \infty} T^{n+1}(x) = T(x_0).$$

Thus x_0 is a fixed point of T .

In order to see that x_0 is the only fixed point of T , suppose $T(y) = y$.

Then by what we have just shown

$$x_0 = \lim_{n \rightarrow \infty} T^n(y) = y.$$

Returning to the inequality

$$d(T^n(x), T^{n+1}(x)) \leq \varphi(T^n(x)) - \varphi(T^{n+1}(x)),$$

upon letting $n \rightarrow \infty$ we see that

$$d(T^n(x), x_0) \leq \varphi(T^n(x)) = \frac{1}{1-k} d(T^n(x), T^{n+1}(x)).$$

Since $\frac{1}{1-k} d(T^n(x), T^{n+1}(x)) \leq \frac{k^n}{1-k} d(x, T(x))$ we obtain

$$d(T^n(x), x_0) \leq \frac{k^n}{1-k} d(x, T(x)).$$

This provides an estimate on the rate of convergence for the sequence $(T^n(x))$ which depends only on $d(x, T(x))$.

□

Inverse Function Theorem

Motivation

1. Let $f: D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^n$ be differentiable. Note that the dimension of the domain and the range are the same. Suppose $f(a) = b$. Then an approximation to $f(x)$ for x near a is given by the map

$$(*) \quad x \mapsto f(a) + f'(a)(x-a).$$

We expect that if $f'(a)$ is an invertible linear function, which implies the map in $(*)$ is bijective, then f should be also bijective near a . We shall see from the Inverse Function Theorem that this is true if f is continuously differentiable.

2. Consider the system of equations $f(x) = y$, where $f: D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^n$,

$$f = \begin{pmatrix} f_1 \\ \vdots \\ f_n \end{pmatrix}, \quad x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}. \quad \text{Thus,}$$

$$f_1(x_1, \dots, x_n) = y_1$$

$$f_2(x_1, \dots, x_n) = y_2$$

$$\vdots$$

$$f_n(x_1, \dots, x_n) = y_n.$$

Suppose that these equations are satisfied if $x = a = (a_1, \dots, a_n)$ and $y = b = (b_1, \dots, b_n)$ and that f is of class C^1 with $f'(a)$ invertible. Then it follows from the Inverse Function Theorem that for all y in some neighborhood of b the equations have a unique solution

x in some neighborhood of a .

Before we proceed with the Inverse Function Theorem, we give without proof the following result.

Proposition 1.44

Let V be a finite dimensional normed space and consider the set

$$M = \{ T \in \mathcal{L}(V, V) : T \text{ invertible} \}.$$

Then

- (i) M is open in $\mathcal{L}(V, V)$. More precisely, if $T_0 \in M$ and $\|T - T_0\| < \frac{1}{2\|T_0^{-1}\|}$, then $T \in M$ and $\|T^{-1}\| \leq 2\|T_0^{-1}\|$.
- (ii) the function $\varphi: M \rightarrow M$, $\varphi(T) = T^{-1}$ is continuous.

Theorem 1.45 (Inverse Function Theorem)

Let D be open in \mathbb{R}^n and $f: D \rightarrow \mathbb{R}^n$ be continuously differentiable.

Assume that $f'(a)$ is invertible for some $a \in D$.

Then there exist an open neighborhood U of a and an open neighborhood V of $b = f(a)$ s.t.

- (i) $f'(x)$ is invertible at every $x \in U$
- (ii) $f|_U: U \rightarrow V$ is bijective, and hence has an inverse $g: V \rightarrow U$
- (iii) g is continuously differentiable and

$$g'(f(x)) = (f'(x))^{-1} \quad \text{for every } x \in U.$$

Proof

We consider $\|\cdot\|_\infty$ on \mathbb{R}^n .

Step 1

We prove that there exists an open neighborhood E of a s.t. $f'(x)$ is invertible for all $x \in E$.

Since $f'(a) \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n)$ and $f'(a)$ is represented by the Jacobian matrix $\mathcal{J}f(a)$, we have that $f'(a)$ is invertible iff $\det \mathcal{J}f(a) \neq 0$. Let us consider the function

$$G: D \rightarrow \mathbb{R}, \quad G(x) = \det \mathcal{J}f(x) = \begin{vmatrix} \frac{\partial f_1}{\partial x_1}(x) & \dots & \frac{\partial f_1}{\partial x_n}(x) \\ \vdots & & \vdots \\ \frac{\partial f_n}{\partial x_1}(x) & \dots & \frac{\partial f_n}{\partial x_n}(x) \end{vmatrix}.$$

Since f is continuously differentiable, we get from Proposition 2.41 that all the partial derivatives, $\frac{\partial f_j}{\partial x_i}: D \rightarrow \mathbb{R}$, $i, j=1, \dots, n$ are continuous, hence all the coefficients of the matrix $\mathcal{J}f(x)$ are continuous. Since $G(x) = \det \mathcal{J}f(x)$ consists of sums and products of these coefficients, it follows that G is a continuous function. We have that $G(a) = \det \mathcal{J}f(a) \neq 0$, so by continuity, there exists an open neighborhood E of a s.t. $G(x) \neq 0$ for all $x \in E$. That is, $f'(x)$ is invertible at every $x \in E$.

Step 2

Since we can consider in the sequel the restriction $f|_E: E \rightarrow \mathbb{R}^n$ of f to the open neighborhood E of a , we can assume without loss of generality that $D=E$. Thus, $f'(x)$ is invertible at every $x \in D$.

Step 3

Let $y \in B_\delta(f(a))$,

where $\delta > 0$ will be chosen later. (we will take the set V in the theorem to be $B_\delta(f(a))$.)

For each such y , we want to prove the existence of $x \in D$ s.t.

(1) $f(x) = y.$

Since f is differentiable at a with derivative $f'(a)$, there exists

$r: D \rightarrow \mathbb{R}^n$ s.t.

$f(x) = f(a) + f'(a)(x-a) + r(x)$, and $\lim_{x \rightarrow a} \frac{r(x)}{\|x-a\|} = 0.$

Then we obtain the following chain of equivalences:

$$\begin{aligned}
 & f(x) = y \\
 & \iff \\
 & f(x) + f'(a)(x-a) + r(x) = y \\
 & \iff \\
 & f'(a)(x-a) = y - f(a) - r(x) \\
 & \iff \\
 & x-a = (f'(a))^{-1}(f'(a)(x-a)) = (f'(a))^{-1}(y - f(a) - r(x)) = \\
 & \quad = (f'(a))^{-1}(y - f(a)) - (f'(a))^{-1}(r(x)) \\
 & \iff \\
 & x = a + (f'(a))^{-1}(y - f(a)) - (f'(a))^{-1}(r(x)).
 \end{aligned}$$

For brevity, let us denote $T := (f'(a))^{-1}$. Then (1) \iff

(2) $x = a + T(y - f(a)) - T(r(x)).$

Thus, we want to find $x \in D$ s.t. (2) is satisfied.

Step 4

We define $\phi_y: D \rightarrow \mathbb{R}^n$,

$$\phi_y(x) = a + T(y - f(a)) - T(r(x)).$$

Note the following:

$$\begin{aligned} x \text{ is a fixed point of } \phi_y &\Leftrightarrow \phi_y(x) = x \Leftrightarrow (2) \text{ holds} \Leftrightarrow \\ &\Leftrightarrow x \text{ solves the equation (1).} \end{aligned}$$

We claim that

$$\phi_y(\overline{U_\varepsilon(a)}) \subseteq U_\varepsilon(a)$$

and that ϕ_y is a contraction, provided $\varepsilon > 0$ is sufficiently small (ε will depend only on f and a) and provided $y \in B_\delta(f(a))$, where $\delta > 0$ also depends only on f and a .

To prove the claim, let $\varepsilon > 0$ and $x_1, x_2 \in \overline{U_\varepsilon(a)}$. Then

$$\begin{aligned} \phi_y(x_1) - \phi_y(x_2) &= (a + T(y - f(a)) - T(r(x_1))) - (a + T(y - f(a)) - T(r(x_2))) \\ &= T(r(x_2)) - T(r(x_1)) = T(r(x_2) - r(x_1)). \end{aligned}$$

and so

$$\begin{aligned} (3) \quad \|\phi_y(x_1) - \phi_y(x_2)\| &\leq \|T(r(x_2) - r(x_1))\| \leq \|T\| \cdot \|r(x_2) - r(x_1)\| \\ &= K \cdot \|r(x_2) - r(x_1)\|, \text{ where } K := \|T\|. \end{aligned}$$

We have that

$$\begin{aligned} r(x_2) - r(x_1) &= \left(f(x_2) - f(a) - f'(a)(x_2 - a) \right) - \left(f(x_1) - f(a) - f'(a)(x_1 - a) \right) \\ &= f(x_2) - f(x_1) - f'(a)(x_2 - x_1). \end{aligned}$$

Since $r, f: D \rightarrow \mathbb{R}^n$, $r = \begin{pmatrix} r_1 \\ \vdots \\ r_n \end{pmatrix}$, $f = \begin{pmatrix} f_1 \\ \vdots \\ f_n \end{pmatrix}$, where $r_i, f_i: D \rightarrow \mathbb{R}$ for all $i=1, \dots, n$. Moreover, by Proposition 3.15, $f'(a) = \begin{pmatrix} f_1'(a) \\ \vdots \\ f_n'(a) \end{pmatrix}$.

Then we get that for all $i=1, \dots, n$

$$r_i(x_2) - r_i(x_1) = f_i(x_2) - f_i(x_1) - f_i'(a)(x_2 - x_1).$$

Since f_i is differentiable and $x_1, x_2 \in \overline{U_\varepsilon(a)}$ implies $[x_1, x_2] \in \overline{U_\varepsilon(a)}$ (by Remark 3.40), we can apply the Mean Value Theorem 3.3) to get $c_i \in [x_1, x_2]$ s.t.

$$f_i(x_2) - f_i(x_1) = f_i'(c_i)(x_2 - x_1).$$

It follows that

$$\begin{aligned} r_i(x_2) - r_i(x_1) &= f_i'(c_i)(x_2 - x_1) - f_i'(a)(x_2 - x_1) = \\ &= (f_i'(c_i) - f_i'(a))(x_2 - x_1). \end{aligned}$$

Hence,

$$(4) \quad \frac{|r_i(x_2) - r_i(x_1)|}{\|x_2 - x_1\|} = |(f_i'(c_i) - f_i'(a))(x_2 - x_1)| \leq \|f_i'(c_i) - f_i'(a)\|.$$

Since f' is continuous by the hypothesis, we get that f_i' is also continuous at a , so there exists $\varepsilon_i > 0$ (depending only on f_i and a) s.t.

$$\left(\forall x \in \overline{U_{\varepsilon_i}(a)} \Rightarrow \|f_i'(x) - f_i'(a)\| \leq \frac{\lambda}{2\kappa} \right).$$

Let $\varepsilon := \min_{i=1, \dots, n} \varepsilon_i$. Then $\overline{U_\varepsilon(a)} \subseteq \overline{U_{\varepsilon_i}(a)}$, so for any $i=1, \dots, n$

$$(5) \left(\forall x \in \overline{U_\varepsilon(a)} \implies \|f_i'(x) - f_i'(a)\| \leq \frac{\varepsilon}{2k} \right).$$

Using (4) and (5) we get that for all $x_1, x_2 \in \overline{U_\varepsilon(a)}$ and for all $i=1, \dots, n$,

$$|r_i(x_2) - r_i(x_1)| \leq \|f_i'(c_i) - f_i'(a)\| \cdot \|x_1 - x_2\| \leq \frac{\varepsilon}{2k} \cdot \|x_1 - x_2\|,$$

since $c_i \in [x_1, x_2] \subseteq \overline{U_\varepsilon(a)}$.

Thus, $\|r(x_2) - r(x_1)\| \leq \frac{\varepsilon}{2k} \cdot \|x_1 - x_2\|$

By (3), it follows that

$$\begin{aligned} \|\phi_y(x_1) - \phi_y(x_2)\| &\leq k \cdot \|r(x_2) - r(x_1)\| \leq k \cdot \frac{\varepsilon}{2k} \cdot \|x_1 - x_2\| \\ &= \frac{\varepsilon}{2} \|x_1 - x_2\|. \end{aligned}$$

This proves that $\phi_y|_{\overline{U_\varepsilon(a)}} : \overline{U_\varepsilon(a)} \rightarrow \mathbb{R}^n$ is a contraction.

It remains to prove that $\phi_y(\overline{U_\varepsilon(a)}) \subseteq U_\varepsilon(a)$.

For this, let $x \in \overline{U_\varepsilon(a)}$. Then

$$\begin{aligned} \|\phi_y(x) - a\| &= \|a + T(y - f(a)) - T(r(x)) - a\| = \|T(y - f(a)) - T(r(x))\| \\ &\leq \|T(y - f(a))\| + \|T(r(x))\| \leq \|T\| \cdot \|y - f(a)\| + \\ &\quad + \|T\| \cdot \|r(x)\| = k \cdot \|y - f(a)\| + k \cdot \|r(x) - r(a)\|, \\ &\quad \text{since } r(a) = 0. \\ &\leq k \cdot \|y - f(a)\| + k \cdot \frac{\varepsilon}{2k} \|x - a\| \leq k \cdot \delta + \frac{\varepsilon}{2}. \end{aligned}$$

Let us choose $\delta > 0$ s.t. $k\delta < \frac{\varepsilon}{2}$, so $\delta > \frac{\varepsilon}{2k}$. Then for all

$y \in B_\delta(f(a))$, we have that $\|\phi_y(x) - a\| < \varepsilon$ for all $x \in \overline{U_\varepsilon(a)}$, that is $\phi_y(x) \in U_\varepsilon(a)$ for all $x \in \overline{U_\varepsilon(a)}$.

Step 5

Thus, we have found $\epsilon > 0$ and $\delta > 0$ depending only on f and a s.t. for all $y \in B_\delta(f(a))$

$$\phi_y: \overline{U_\epsilon(a)} \rightarrow \overline{U_\epsilon(a)}, \quad \phi_y(x) = a + T(y - f(a)) - T(f(x))$$

is a contraction. Since $\overline{U_\epsilon(a)}$ is closed in the complete metric space \mathbb{R}^n with the distance induced by $\|\cdot\|_\infty$, it follows by Proposition 6.15 that $\overline{U_\epsilon(a)}$ is complete. So, we can apply Banach's Contraction Mapping Principle (Theorem 2.43) to get that there exists a unique $x \in \overline{U_\epsilon(a)}$ s.t. $\phi_y(x) = x$. Moreover, since $\phi_y(x) \in U_\epsilon(a)$ we get that $x \in U_\epsilon(a)$. Thus, for any $y \in B_\delta(f(a))$ there is a unique $x \in B_\epsilon(a)$ s.t. $f(x) = y$.

Step 6

Let $U := B_\epsilon(a) \cap f^{-1}(V)$ and $V := B_\delta(f(a))$. Then U, V are open, $b \in V$ and $a \in U$, since $a \in B_\epsilon(a)$ and $f(a) = b \in V$, so $a \in f^{-1}(V)$.

From the definition of U , we have that for all $x \in U$, $f(x) \in V$, so $f(U) \subseteq V$. Thus, $f|_U: U \rightarrow V$ is well-defined. Since we have proved in Step 5 that for all $y \in V$ there exists a unique $x \in U$ s.t. $f(x) = y$, it follows that $f|_U: U \rightarrow V$ is bijective.

Step 7

Let $g: V \rightarrow U$ be the inverse of f . Thus for all $y \in V$, we have that $g(y) = x$, where x is unique s.t. $f(x) = y$ or, equivalently, $\phi_y(x) = x$.

We prove that g is Lipschitz, hence continuous. For all $y_1, y_2 \in V$, if we denote $x_1 = g(y_1)$ and $x_2 = g(y_2)$, we get that

$$\begin{aligned} \|g(y_1) - g(y_2)\| &= \|x_1 - x_2\| = \|\Phi_{y_1}(x_1) - \Phi_{y_2}(x_2)\| = \\ &= \|(a + T(y_1 - f(a)) - T(r(x_1))) - (a + T(y_2 - f(a)) - T(r(x_2)))\| \\ &= \|T(y_1 - y_2) - T(r(x_1) - r(x_2))\| \leq \|T\| \cdot \|y_1 - y_2\| + \\ &\quad + \|T\| \cdot \|r(x_1) - r(x_2)\| \leq K \cdot \|y_1 - y_2\| + K \cdot \frac{1}{2K} \|x_1 - x_2\| \\ &= K \cdot \|y_1 - y_2\| + \frac{1}{2} \|g(y_1) - g(y_2)\|. \end{aligned}$$

Thus,

$$\frac{1}{2} \|g(y_1) - g(y_2)\| \leq K \cdot \|y_1 - y_2\|$$

and so

$$\|g(y_1) - g(y_2)\| \leq 2K \cdot \|y_1 - y_2\|.$$

Step 8

Since g is continuous on V and $f'(x)$ is invertible for all $x \in U$, we can apply Proposition 1.11 to get that g is differentiable on V with derivative

$$g'(y) = [f'(g(y))]^{-1} \quad \text{for all } y \in V,$$

$$\text{so } g'(f(x)) = (f'(x))^{-1} \quad \text{for all } x \in U.$$

Step 9.

It remains to prove that g' is ~~continuous~~ continuous.

This follows immediately by using Proposition 1.44 and reusing

(51)

that $g' = \varphi \circ f' \circ \sigma$, where φ is the continuous function defined in Proposition 9.44. Since φ is continuous, g is continuous as we proved in Step 7 and f' is continuous by the hypothesis, it follows that g' is continuous.

□

Remark 1.46

Local invertibility does not imply global invertibility.

Proof

Let us consider the following example.

$$f: \underbrace{\{(x,y) \in \mathbb{R}^2 : y > 0\}}_D \rightarrow \mathbb{R}^2, \quad f(x,y) = \begin{pmatrix} y \cos x \\ y \sin x \end{pmatrix}.$$

Then f is continuously differentiable with $\partial f(x,y) = \begin{pmatrix} -y \sin x & \cos x \\ y \cos x & \sin x \end{pmatrix}$

for all $(x,y) \in D$. Then

$$\det \partial f(x,y) = -y \sin^2 x - y \cos^2 x = -y \neq 0$$

for all $(x,y) \in D$. Consequently, f is locally invertible at every point in D .

Yet, f is not globally invertible, since f is 2π -periodic with respect to x , so f is not injective.

□