

18.024 SPRING OF 2008
DER. DERIVATIVES OF VECTOR FIELDS

Derivatives of vector fields. Derivative theory for vector fields is a straightforward extension of that for scalar fields. Given $\mathbf{f} : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ a vector field, in components, \mathbf{f} consists of m scalar fields of n variables. That is,

$$\mathbf{f}(\mathbf{x}) = (f_1(\mathbf{x}), \dots, f_m(\mathbf{x})),$$

and each $f_i : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is a scalar field, where $i = 1, \dots, m$. It seems natural to define the differentiability of \mathbf{f} , thus, componentwise.

We recall that if $f : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is a scalar field and $\mathbf{a} \in \text{int}(D)$ then the *derivative* of f at \mathbf{a} (if exists) is given by the gradient

$$\nabla f(\mathbf{a}) = (D_1 f(\mathbf{a}), D_2 f(\mathbf{a}), \dots, D_n f(\mathbf{a})).$$

For our purposes, it is convenient to understand the derivative of f as a *row matrix* rather than as a vector. We will write as

$$Df(\mathbf{a}) = [D_1 f(\mathbf{a}), D_2 f(\mathbf{a}), \dots, D_n f(\mathbf{a})].$$

Then, f is differentiable at \mathbf{a} (with the total derivative $Df(\mathbf{a})$) if

$$(*) \quad f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) = Df(\mathbf{a})\mathbf{h} + \|\mathbf{h}\|E(\mathbf{h}; \mathbf{a}),$$

and $|E(\mathbf{h}; \mathbf{a})| \rightarrow 0$ as $\|\mathbf{h}\| \rightarrow 0$. Here the first term on the right side is matrix multiplication of $Df(\mathbf{a})$ and the column matrix

$$\mathbf{h} = \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix}.$$

Definition 1. Let $\mathbf{f} : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $D \subset \mathbb{R}^n$ be open. In scalar form,

$$\mathbf{f}(\mathbf{x}) = (f_1(x_1, \dots, x_n), \dots, f_m(x_1, \dots, x_n)).$$

We say that f is *differentiable* at $\mathbf{a} \in D$ if each of the component functions f_1, \dots, f_m is differentiable at \mathbf{a} (in the sense discussed in [Apo, Section 8.11] or above).

Furthermore, by the *derivative* of \mathbf{f} at \mathbf{a} we mean the matrix

$$D\mathbf{f}(\mathbf{a}) = \begin{bmatrix} D_1 f_1(\mathbf{a}) & D_2 f_1(\mathbf{a}) & \cdots & D_n f_1(\mathbf{a}) \\ D_1 f_2(\mathbf{a}) & D_2 f_2(\mathbf{a}) & \cdots & D_n f_2(\mathbf{a}) \\ \vdots & \vdots & \cdots & \vdots \\ D_1 f_m(\mathbf{a}) & D_2 f_m(\mathbf{a}) & \cdots & D_n f_m(\mathbf{a}) \end{bmatrix}.$$

In other words, $D\mathbf{f}(\mathbf{a})$ is the matrix whose i th row is given as the derivative $Df_i(\mathbf{a})$ of the i th coordinate function of \mathbf{f} . It is often called the *Jacobian matrix** of \mathbf{f} . Another notation for this matrix is

$$\frac{\partial(f_1, \dots, f_m)}{\partial(x_1, \dots, x_n)}.$$

*It is named after the Prussian mathematician Carl Gustav Jacob Jacobi (1804-1851).

In [Apo, Section 8.18], the differentiability of a vector field \mathbf{f} is given, alternatively, by directly extending the Taylor's formula (*) in the vector-field setting. Our definition then is equivalent to that given in [Apo], which is stated below.

Theorem 2. *The vector field \mathbf{f} is differentiable at \mathbf{a} if and only if*

$$\mathbf{f}(\mathbf{a} + \mathbf{h}) - \mathbf{f}(\mathbf{a}) = D\mathbf{f}(\mathbf{a})\mathbf{h} + \|\mathbf{h}\|\mathbf{E}(\mathbf{h}; \mathbf{a})$$

and $\|\mathbf{E}(\mathbf{h}; \mathbf{a})\| \rightarrow 0$ as $\|\mathbf{h}\| \rightarrow 0$. Here, \mathbf{f} , \mathbf{h} and \mathbf{E} are written as column matrices.

Proof. We observe that both sides of the equation represent column matrices. Considering the i th entries of these matrices, we obtain the following equation

$$f_i(\mathbf{a} + \mathbf{h}) - f_i(\mathbf{a}) = Df_i(\mathbf{a}) \cdot \mathbf{h} + \|\mathbf{h}\|\mathbf{E}_i(\mathbf{h}; \mathbf{a}).$$

Now if \mathbf{f} is differentiable at \mathbf{a} if and only if each f_i is, and f_i is differentiable at \mathbf{a} if and only if $|\mathbf{E}_i(\mathbf{h}; \mathbf{a})| \rightarrow 0$ as $\|\mathbf{h}\| \rightarrow 0$. But, $|\mathbf{E}_i(\mathbf{h})| \rightarrow 0$ as $\|\mathbf{h}\| \rightarrow 0$ for each i if and only if $\|\mathbf{E}(\mathbf{h})\| \rightarrow 0$ as $\|\mathbf{h}\| \rightarrow 0$, and we are done. \square

The following result is immediate from our definition of differentiability.

Theorem 3. *If \mathbf{f} is differentiable at \mathbf{a} then \mathbf{f} is continuous at \mathbf{a} .*

The chain rule for derivatives of vector fields. Before considering the chain rule for vector fields, let us take the chain rule for scalar fields we have already proved [Apo, Section 8.15] and reformulate it in terms of matrices.

Let $f : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$ be a scalar field defined in an open ball about \mathbf{a} , and let $\mathbf{x} : I \subset \mathbb{R} \rightarrow \mathbb{R}^n$ be a vector-valued function defined in an open interval about t_0 . Let $\mathbf{x}(t_0) = \mathbf{a}$ and $\mathbf{x}(I) \subset D$. If f is differentiable at \mathbf{a} and if \mathbf{x} is differentiable at t_0 , then we have shown in Theorem 8.8 of [Apo] that $f \circ \mathbf{x}$ is differentiable at t_0 and its derivative is given by

$$\frac{d}{dt}f(\mathbf{x}(t_0)) = \nabla f(\mathbf{x}(t_0)) \cdot \mathbf{x}'(t_0).$$

We now write this formula in scalar form as

$$\frac{d}{dt}(f \circ \mathbf{x}) = \frac{\partial f}{\partial x_1} \frac{dx_1}{dt} + \cdots + \frac{\partial f}{\partial x_n} \frac{dx_n}{dt};$$

or, we write it in matrix form as

$$\frac{d}{dt}(f \circ \mathbf{x}) = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix} \begin{bmatrix} \frac{dx_1}{dt} \\ \vdots \\ \frac{dx_n}{dt} \end{bmatrix}.$$

Recalling the definition of the Jacobian matrix Df the latter formula is recognized as

$$\frac{d}{dt}(f(\mathbf{x}(t))) = Df(\mathbf{x}(t))D\mathbf{x}(t).$$

(Note that Df is a row matrix while $D\mathbf{x}$ is by definition a column matrix.) This is the form of the chain rule which is useful to extend to vector fields.

Let $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^p$ and $\mathbf{g} : \mathbb{R}^p \rightarrow \mathbb{R}^m$ be vector fields such that the composition field $\mathbf{h} = \mathbf{g} \circ \mathbf{f}$ is defined in an open ball about $\mathbf{a} \in \mathbb{R}^n$. Let $\mathbf{f}(\mathbf{a}) = \mathbf{b} \in \mathbb{R}^p$. We write these fields as

$$\mathbf{f}(\mathbf{x}) = \mathbf{f}(x_1, \dots, x_n), \quad \mathbf{g}(\mathbf{y}) = \mathbf{g}(y_1, \dots, y_p).$$

If \mathbf{f} and \mathbf{g} are differentiable at \mathbf{a} and \mathbf{b} , respectively, then partial derivatives of \mathbf{h} exist at \mathbf{a} by applying the chain rule for each $h_i(\mathbf{x}) = g_i(\mathbf{f}(\mathbf{x}))$ as

$$\frac{\partial h_i}{\partial x_j} = \frac{\partial g_i}{\partial y_1} \frac{\partial f_1}{\partial x_j} + \cdots + \frac{\partial g_i}{\partial y_p} \frac{\partial f_p}{\partial x_j}.$$

This is recognized as matrix multiplication

$$[D_1 g_i \ D_2 g_i \ \cdots \ D_p g_i] \begin{bmatrix} D_j f_1 \\ \vdots \\ D_j f_p \end{bmatrix}.$$

In other words, its multiplication of the i th row of $D\mathbf{g}$ and the j th column of $D\mathbf{f}$. Thus, the Jacobian matrix of \mathbf{h} is expected to satisfy the matrix equation

$$D\mathbf{h}(\mathbf{a}) = D\mathbf{g}(\mathbf{b})D\mathbf{f}(\mathbf{a}).$$

Not exactly. If \mathbf{f} and \mathbf{g} are differentiable then we know that the partial derivatives of the composite function \mathbf{h} exist. But, the mere *existence* of the partial derivatives of the function h_i does not imply the differentiability of h_i , and thus, one needs to provide a separate argument that \mathbf{h} is differentiable.

One may avoid giving a separate proof of the differentiability of \mathbf{h} by assuming a stronger hypothesis, namely, that both \mathbf{f} and \mathbf{g} are *continuously differentiable*. Then, the formula,

$$D_j h_i(\mathbf{x}) = \sum_{l=1}^p D_l g_i(\mathbf{f}(\mathbf{x})) D_j f_l(\mathbf{x}),$$

which we have proved, shows that $D_j h_i$ is continuous, that is, \mathbf{h} is continuously differentiable. Therefore, \mathbf{h} is differentiable.

We summarize these.

Theorem 4. *Let $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^p$ and $\mathbf{g} : \mathbb{R}^p \rightarrow \mathbb{R}^m$ be vector fields defined in open balls about $\mathbf{a} \in \mathbb{R}^n$ and $\mathbf{b} \in \mathbb{R}^p$, respectively. Let $\mathbf{f}(\mathbf{a}) = \mathbf{b}$. If \mathbf{f} and \mathbf{g} are continuously differentiable on their respective domains, then $\mathbf{h}(\mathbf{x}) = \mathbf{g} \circ \mathbf{f}(\mathbf{x})$ is continuously differentiable on its domain and*

$$D\mathbf{h}(\mathbf{x}) = D\mathbf{g}(\mathbf{f}(\mathbf{x})) \cdot D\mathbf{f}(\mathbf{x}).$$

Most applications of the chain rule we will use in future are in the C^1 -class setting, and thus this theorem will suffice. But, one may remove the C^1 condition. Please read the statement and its proof of [Apo, Theorem 8.11].

Example 5 (Polar coordinates). Given a function $f(x, y)$ defined on the (x, y) plane, we introduce polar coordinates

$$x = r \cos \theta \quad \text{and} \quad y = r \sin \theta$$

and f becomes a function of r and θ as $\phi(r, \theta) = f(r \cos \theta, r \sin \theta)$. Then,

$$\frac{\partial \phi}{\partial r} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial r} = \frac{\partial f}{\partial x} \cos \theta + \frac{\partial f}{\partial y} \sin \theta$$

and

$$\frac{\partial \phi}{\partial \theta} = -r \frac{\partial f}{\partial x} \sin \theta + r \frac{\partial f}{\partial y} \cos \theta.$$

The second derivative is computed as

$$\begin{aligned}
\frac{\partial^2 \phi}{\partial r^2} &= \frac{\partial}{\partial r} \left(\frac{\partial \phi}{\partial r} \right) = \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial x} \cos \theta + \frac{\partial f}{\partial y} \sin \theta \right) \\
&= \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial x} \right) \cos \theta + \frac{\partial f}{\partial x} \frac{\partial(\cos \theta)}{\partial r} + \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial y} \right) \sin \theta + \frac{\partial f}{\partial y} \frac{\partial(\sin \theta)}{\partial r} \\
&= \left(\frac{\partial}{\partial x} \left(\frac{\partial f}{\partial x} \right) \cos \theta + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right) \sin \theta \right) \cos \theta + \left(\frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right) \cos \theta + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial y} \right) \sin \theta \right) \sin \theta \\
&= \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial x \partial y} \sin \theta \cos \theta.
\end{aligned}$$

An analogous formula for $\partial^2 \phi / \partial \theta^2$ is worked in [Apo, Example 3] and $\partial^2 \phi / \partial r \partial \theta$ is in [Apo, Section 8.22, Exercise 5].

Derivatives of inverses. Recall that if $f(x)$ is a differentiable real-valued function of a single variable x , and if $f'(x) > 0$ for $x \in [a, b]$ (or $f'(x) < 0$), then f is strictly increasing (or decreasing), and it has an inverse, say g . Furthermore, g is differentiable and its derivative satisfies

$$g'(f(x)) = \frac{1}{f'(x)}.$$

Part of this theory extends to vector fields.

Theorem 6. Let $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be defined in an open ball of \mathbf{a} and $\mathbf{f}(\mathbf{a}) = \mathbf{b}$. Suppose that \mathbf{f} has an inverse, say \mathbf{g} .

If \mathbf{f} is differentiable at \mathbf{a} and if \mathbf{g} is differentiable at \mathbf{b} then

$$D\mathbf{g}(\mathbf{b}) = (D\mathbf{f}(\mathbf{a}))^{-1}.$$

In interpretation, the jacobian matrix of the inverse function is just the inverse of the jacobian matrix.

Proof. We observe that $\mathbf{g} \circ \mathbf{f}(\mathbf{x}) = \mathbf{x}$. Since \mathbf{f} and \mathbf{g} are differentiable and so is the composite function $\mathbf{g} \circ \mathbf{f}$, by the chain rule, it follows that

$$D\mathbf{g}(\mathbf{b})D\mathbf{f}(\mathbf{a}) = I_n,$$

which implies the assertion. \square

The theorem shows that in order for the differentiable function \mathbf{f} to have a differentiable inverse, it is *necessary* that the Jacobian matrix $D\mathbf{f}$ have rank n . Roughly speaking, it is also *sufficient* for \mathbf{f} to have an inverse. More precisely, one has the following *Inverse Function Theorem*, which is one of the fundamental results in mathematical analysis.

Theorem 7 (Inverse function Theorem). If $f : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ is of C^1 in a neighborhood of \mathbf{a} and $D\mathbf{f}(\mathbf{a})$ has rank n , then there is an open ball $B \subset \mathbb{R}^n$ about \mathbf{a} such that $\mathbf{f} : B \rightarrow C = \mathbf{f}(B)$ is one-to-one and onto. Furthermore, the inverse function $\mathbf{g} : C \rightarrow B$ of \mathbf{f} is continuously differentiable and $D\mathbf{g}(\mathbf{f}(\mathbf{x})) = (D\mathbf{f}(\mathbf{x}))^{-1}$.

The proof and further application of the inverse function theorem is learned in 18.100A.

Example 8 (Invertibility of the polar coordinate transformation). Let

$$\mathbf{f} : \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad \mathbf{f}(r, \theta) = (x, y)$$

be the differentiable transformation from the (x, y) -plane to (r, θ) plane by polar coordinates:

$$x = r \cos \theta \quad \text{and} \quad y = r \sin \theta.$$

Its Jacobian matrix is computed as

$$\frac{\partial(x, y)}{\partial(r, \theta)} = \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}.$$

Note that its determinant is r and thus, it is of rank 2 provided that $r \neq 0$. According to the inverse function theorem, thus, away from the origin ($r = 0$), the polar coordinate transformation \mathbf{f} is invertible locally, and its inverse is given by $\mathbf{g}(x, y) = (r, \theta)$. Moreover,

$$\frac{\partial(r, \theta)}{\partial(x, y)} = \frac{\partial(x, y)}{\partial(r, \theta)}^{-1} = \frac{1}{r} \begin{bmatrix} r \cos \theta & r \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}.$$

REFERENCES

[Apo] T. Apostol, *Calculus*, vol. II, Second edition, Wiley, 1967.

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