6 Differentiability of vector fields

6.1 Differentiable maps $\mathbb{R}^n \to \mathbb{R}^n$

We will now generalise the idea of differentiability of scalar fields from section 5. Recall that for a scalar field $f(\underline{x}): U \to \mathbb{R}$, with U open in \mathbb{R}^n , f is differentiable at $\underline{a} \in U$ if

$$f(\underline{a} + \underline{h}) - f(\underline{a}) = \underline{h} \cdot \underline{\nabla} f(\underline{a}) + R(\underline{h})$$
 (a)

with $\lim_{\underline{h} \to \underline{0}} \frac{R(\underline{h})}{|\underline{h}|} = 0, \qquad (b)$

where you should note that the first term on RHS of (a), \underline{h} . $\underline{\nabla} f(\underline{a})$, is linear in \underline{h} .

Definition 6.1. Consider a vector field $\underline{F}(\underline{x}): U \to \mathbb{R}^n$, U open in \mathbb{R}^n . Then \underline{F} is defined to be differentiable at $\underline{a} \in U$ if there is a linear function $\underline{L}: \mathbb{R}^n \to \mathbb{R}^n$ such that

$$\underline{F}(\underline{a} + \underline{h}) - \underline{F}(\underline{a}) = \underline{L}(\underline{h}) + \underline{R}(\underline{h}) \tag{A}$$

with $\lim_{\underline{h} \to \underline{0}} \frac{\underline{R}(\underline{h})}{|\underline{h}|} = \underline{0}. \tag{B}$

Now linear functions $\mathbb{R}^n \to \mathbb{R}^n$ are matrices. To see what matrix use the standard basis,

$$\underline{F}(\underline{x}) = F_1(\underline{x})\underline{e}_1 + F_2(\underline{x})\underline{e}_2 + \dots + F_n(\underline{x})\underline{e}_n$$

$$\underline{L}(\underline{h}) = L_1(\underline{h})\underline{e}_1 + L_2(\underline{h})\underline{e}_2 + \dots + L_n(\underline{h})\underline{e}_n$$

$$\underline{R}(\underline{h}) = R_1(\underline{h})\underline{e}_1 + R_2(\underline{h})\underline{e}_2 + \dots + R_n(\underline{h})\underline{e}_n,$$

so the jth components of (A) and (B) are

$$F_j(\underline{a} + \underline{h}) - F_j(\underline{a}) = L_j(\underline{h}) + R_j(\underline{h})$$
 (A)_j

with

$$\lim_{\underline{h} \to \underline{0}} \frac{R_j(\underline{h})}{|\underline{h}|} = 0.$$
 (B)_j

These are just conditions (a) and (b) for $F_j(\underline{x})$ to be differentiable as a map $U \to \mathbb{R}$, i.e. as a scalar field. So we can use results from section 5 to see that

$$L_j(\underline{h}) = \underline{h}.\underline{\nabla}F_j(\underline{a}),$$

that is

$$L_{1} = \underline{h} \cdot \underline{\nabla} F_{1}(\underline{a}) = h_{1} \frac{\partial F_{1}}{\partial x_{1}} + h_{2} \frac{\partial F_{1}}{\partial x_{2}} + \dots + h_{n} \frac{\partial F_{1}}{\partial x_{n}}$$

$$L_{2} = \underline{h} \cdot \underline{\nabla} F_{2}(\underline{a}) = h_{1} \frac{\partial F_{2}}{\partial x_{1}} + h_{2} \frac{\partial F_{2}}{\partial x_{2}} + \dots + h_{n} \frac{\partial F_{2}}{\partial x_{n}}$$

: :

$$L_n = \underline{h}.\underline{\nabla}F_n(\underline{a}) = h_1\frac{\partial F_n}{\partial x_1} + h_2\frac{\partial F_n}{\partial x_2} + \dots + h_n\frac{\partial F_n}{\partial x_n},$$

or

$$\begin{pmatrix} L_1 \\ L_2 \\ \vdots \\ L_n \end{pmatrix} = \begin{pmatrix} h_1 \frac{\partial F_1}{\partial x_1} + h_2 \frac{\partial F_1}{\partial x_2} + \dots + h_n \frac{\partial F_1}{\partial x_n} \\ h_1 \frac{\partial F_2}{\partial x_1} + h_2 \frac{\partial F_2}{\partial x_2} + \dots + h_n \frac{\partial F_2}{\partial x_n} \\ & \vdots \\ h_1 \frac{\partial F_n}{\partial x_1} + h_2 \frac{\partial F_n}{\partial x_2} + \dots + h_n \frac{\partial F_n}{\partial x_n} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{\partial F_1}{\partial x_1} & \frac{\partial F_1}{\partial x_2} & \dots & \frac{\partial F_1}{\partial x_n} \\ \frac{\partial F_2}{\partial x_1} & \frac{\partial F_2}{\partial x_2} & \dots & \frac{\partial F_2}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial F_n}{\partial x_1} & \frac{\partial F_n}{\partial x_2} & \dots & \frac{\partial F_n}{\partial x_n} \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \\ \vdots \\ h_n \end{pmatrix}.$$

The $n \times n$ matrix on the RHS of the last equation is called the <u>Jacobian matrix</u>, or <u>differential</u>, of $\underline{F}(\underline{x})$ at $\underline{x} = \underline{a}$; it is written as $D\underline{F}(\underline{a})$, or $d\underline{F}(\underline{a})$ (or $D\underline{F}_{\underline{a}}$ or $d\underline{F}_{\underline{a}}$ or even J_{ij}).

Definition 6.2. The determinant of the differential,

$$\det(D\underline{v}) \equiv |D\underline{v}|$$

is called the <u>Jacobian</u>, $J(\underline{v})$.

Example 33. If

$$\underline{v}(\underline{x}) = \begin{pmatrix} x^2 - y^2 \\ 2xy \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix},$$

then

$$D\underline{v}(\underline{x}) = \begin{pmatrix} \frac{\partial v_1}{\partial x} & \frac{\partial v_1}{\partial y} \\ \frac{\partial v_2}{\partial x} & \frac{\partial v_2}{\partial y} \end{pmatrix} = \begin{pmatrix} 2x & -2y \\ 2y & 2x \end{pmatrix}$$

The Jacobian is then given by

$$J(\underline{v}) = \begin{vmatrix} 2x & -2y \\ 2y & 2x \end{vmatrix}$$
$$= 4(x^2 + y^2) = 4|\underline{x}|^2$$

Example 34. If $\underline{x} \in \mathbb{R}^n$ and $\underline{v}(\underline{x}) = \underline{x}$ then

$$D\underline{v}(\underline{x}) = \begin{pmatrix} \frac{\partial x_1}{\partial x_1} & \frac{\partial x_1}{\partial x_2} & \cdots & \frac{\partial x_1}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial x_n}{\partial x_1} & \frac{\partial x_n}{\partial x_2} & \cdots & \frac{\partial x_n}{\partial x_n} \end{pmatrix} = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{pmatrix} = \mathbb{I}_n,$$

the $n \times n$ identity matrix, and

$$J(\underline{v}) = |\mathbb{I}_n| = 1$$

6.2 Diffeomorphisms and the inverse function theorem

We can think of a vector field $\underline{v}(\underline{x})$ as a mapping $\mathbb{R}^n \to \mathbb{R}^n$, or equivalently as a <u>coordinate transformation</u> on \mathbb{R}^n . If we think of the components of \underline{h} as the coordinates of a point $\underline{x} = \underline{a} + \underline{h}$ relative to an "origin" at \underline{a} , then the components of $\underline{v}(\underline{a} + \underline{h}) - \underline{v}(\underline{a})$ are the transformed coordinates relative to the transformed origin $\underline{v}(\underline{a})$. Then for differentiable \underline{v} (and small \underline{h})

$$\underline{v}(\underline{a} + \underline{h}) - \underline{v}(\underline{a}) \simeq D\underline{v}(\underline{a}) \cdot \underline{h}$$

new coordinates \simeq matrix \times old coordinates,

which is a <u>linear</u> transformation, invertible if the determinant of $D\underline{v}(\underline{a})$ (i.e. the Jacobian) is non-zero.

The inverse function theorem says that this invertibility can be extended beyond the linear behaviour:

Theorem 6.3. Let $\underline{v}:U\to\mathbb{R}^n$ (with U open in \mathbb{R}^n) be a differentiable vector field with continuous partial derivatives, and let $\underline{a}\in U$. Then if $J(\underline{v}(\underline{a}))\neq 0$, \exists an open set $\tilde{U}\subseteq U$ containing \underline{a} such that

- (i) $v(\tilde{U})$ is open
- (ii) The mapping \underline{v} from \tilde{U} to $\underline{v}(\tilde{U})$ has a differentiable inverse i.e. there exists a differentiable vector field $\underline{w}:\underline{v}(\tilde{U})\to\mathbb{R}^n$ such that $\underline{w}(\underline{v}(\underline{x}))=\underline{x}$ and $\underline{v}(\underline{w}(y))=y$

Definition 6.4.

- A mapping $\underline{v}: \tilde{U} \to V \subset \mathbb{R}^n$ satisfying (i) and (ii) above is called a <u>diffeomorphism</u> of \tilde{U} onto $\tilde{V} = v(\tilde{U})$, and \tilde{U} and \tilde{V} are said to be diffeomorphic.
- More generally, a mapping $\underline{v}:U\to V$ is called a <u>local diffeomorphism</u> if for every point $\underline{a}\in U$ there is an open set $\tilde{U}\subset U$ containing \underline{a} such that $\underline{v}:\tilde{U}\to\underline{v}(\tilde{U})$ is a diffeomorphism.

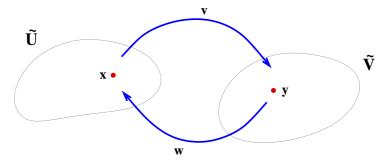


Figure 29: A diffeomorphism from \tilde{U} to \tilde{V}

Remarks In general, suppose that

$$\underline{v}: U \to V \subset \mathbb{R}^n$$
$$\underline{w}: V \to W \subset \mathbb{R}^n$$

(with U and V both open in \mathbb{R}^n) are both continuously differentiable vector fields (not necessarily diffeomorphisms).

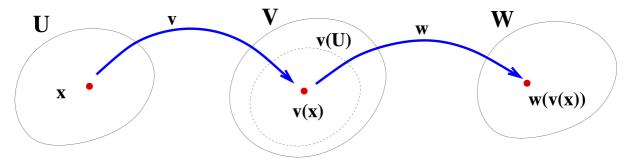


Figure 30: Composition of maps \underline{v} and \underline{w}

Then $\underline{w}(\underline{v}(\underline{x}))$ is a mapping $U \to W \subset \mathbb{R}^n$ and its differential can be calculated using the chain rule:

$$D\underline{w}(\underline{v}(\underline{x})) = \begin{pmatrix} \frac{\partial w_1(\underline{v}(\underline{x}))}{\partial x_1} & \frac{\partial w_1(\underline{v}(\underline{x}))}{\partial x_2} & \dots & \frac{\partial w_1(\underline{v}(\underline{x}))}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial w_n(\underline{v}(\underline{x}))}{\partial x_1} & \frac{\partial w_n(\underline{v}(\underline{x}))}{\partial x_2} & \dots & \frac{\partial w_n(\underline{v}(\underline{x}))}{\partial x_n} \end{pmatrix}$$

$$= \begin{pmatrix} \sum_j \frac{\partial w_1}{\partial v_j} \frac{\partial v_j}{\partial x_1} & \sum_j \frac{\partial w_1}{\partial v_j} \frac{\partial v_j}{\partial x_2} & \dots & \sum_j \frac{\partial w_1}{\partial v_j} \frac{\partial v_j}{\partial x_n} \\ \vdots & & \vdots & & \vdots \\ \sum_j \frac{\partial w_n}{\partial v_j} \frac{\partial v_j}{\partial x_1} & \sum_j \frac{\partial w_n}{\partial v_j} \frac{\partial v_j}{\partial x_2} & \dots & \sum_j \frac{\partial w_n}{\partial v_j} \frac{\partial v_j}{\partial x_n} \end{pmatrix}$$

using chain rule on each matrix element,

$$=\begin{pmatrix} \frac{\partial w_1}{\partial v_1} & \frac{\partial w_1}{\partial v_2} & \cdots & \frac{\partial w_1}{\partial v_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial w_n}{\partial v_1} & \frac{\partial w_n}{\partial v_2} & \cdots & \frac{\partial w_n}{\partial v_n} \end{pmatrix} \begin{pmatrix} \frac{\partial v_1}{\partial x_1} & \frac{\partial v_1}{\partial x_2} & \cdots & \frac{\partial v_1}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial v_n}{\partial x_1} & \frac{\partial v_n}{\partial x_2} & \cdots & \frac{\partial v_n}{\partial x_n} \end{pmatrix}$$

$$= D\underline{w}(\underline{v}) D\underline{v}(\underline{x}) \qquad \text{i.e. matrix multiplication}$$

For the special case when \underline{v} is a local diffeomorphism and \underline{w} is its inverse map,

$$\underline{w}(\underline{v}(\underline{x})) = \underline{x}
\Rightarrow D\underline{w} D\underline{v} = D\underline{w}(\underline{v}(\underline{x})) = D\underline{x}(\underline{x}) = \mathbb{I}_n,$$

using Example 34 above. Likewise

$$\underline{v}(\underline{w}(\underline{y})) = \underline{y}$$

$$\Rightarrow D\underline{v} \ D\underline{w} = D\underline{v}(\underline{w}(y)) = Dy(y) = \mathbb{I}_n.$$

So $D\underline{v}$ is an invertible matrix, with inverse $(D\underline{v})^{-1} = D\underline{w}$. Taking determinants,

$$J(\underline{w}) = \frac{1}{J(v)},$$

and in particular, $J(\underline{v}) \neq 0$, which was the main condition of the inverse function theorem.

Definition 6.5. Such a \underline{v} is called <u>orientation preserving</u> if $J(\underline{v}) > 0$, and <u>orientation reversing</u> if $J(\underline{v}) < 0$.

Example 35. Continuing Example 33 we had

$$\underline{v}(\underline{x}) \; = \; \begin{pmatrix} x^2 - y^2 \\ 2xy \end{pmatrix} \qquad \Rightarrow \qquad J(\underline{v}) \; = \; 4(x^2 + y^2).$$

So for $(x, y) \neq (0, 0), J(\underline{v}) > 0$.

Hence if $U = \mathbb{R}^2 - \{0\}$, $\underline{v}: U \to U$ is an orientation preserving local diffeomorphism. However it is not a global diffeomorphism since $\underline{v}(-\underline{x}) = \underline{v}(\underline{x})$ (so no inverse can exist globally). But \underline{v} does map $\{(x,y): x>0\}$ onto $\mathbb{R}^2 - \{(x,0), x\leq 0\}$ diffeomorphically.

Example 36. Consider the transformation from polar coordinates (r, θ) back to cartesians (x, y). We have

$$\underline{v}(r,\theta) = \begin{pmatrix} x(r,\theta) \\ y(r,\theta) \end{pmatrix} = \begin{pmatrix} r\cos\theta \\ r\sin\theta \end{pmatrix}$$

Differential:

$$D\underline{v} = \begin{pmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{pmatrix} = \begin{pmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{pmatrix}$$
(6.1)

Jacobian:

$$J(\underline{v}(r,\theta)) = \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix}$$
$$= r \cos^2 \theta + r \sin^2 \theta = r$$

For r > 0, $J(\underline{v}) > 0$, and the transformation is therefore orientation preserving.

The inverse mapping is

$$\underline{w}(x,y) = \begin{pmatrix} r(x,y) \\ \theta(x,y) \end{pmatrix} = \begin{pmatrix} \sqrt{x^2 + y^2} \\ \tan^{-1}(\frac{y}{x}) \end{pmatrix}$$

Exercise: check this!