

# Jacobians with vectors and poses

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$$\mathbf{J} = \frac{\partial f(\mathbf{x})}{\partial \mathbf{x}} \triangleq \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_m} \\ \vdots & & \vdots \\ \frac{\partial f_n}{\partial x_1} & \cdots & \frac{\partial f_n}{\partial x_m} \end{bmatrix} \in \mathbb{R}^{n \times m}$$



# The Jacobian

Given a multivariate function  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$  with inputs  $\mathbf{x} \in \mathbb{R}^m$  and outputs  $f(\mathbf{x}) \in \mathbb{R}^n$  the **Jacobian matrix** stacks all the **partial derivatives** as:

$$\mathbf{J} = \frac{\partial f(\mathbf{x})}{\partial \mathbf{x}} \triangleq \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_m} \\ \vdots & & \vdots \\ \frac{\partial f_n}{\partial x_1} & \cdots & \frac{\partial f_n}{\partial x_m} \end{bmatrix} \in \mathbb{R}^{n \times m}$$

# The Jacobian

We will use the following convenient notation:

$$\mathbf{J} = \frac{\partial f(\mathbf{x})}{\partial \mathbf{x}} \triangleq \lim_{\mathbf{h} \rightarrow 0} \frac{f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x})}{\mathbf{h}} \in \mathbb{R}^{n \times m}$$

This lets us define a compact procedure for finding the Jacobian:

$$\lim_{\mathbf{h} \rightarrow 0} \frac{f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x})}{\mathbf{h}} = \dots = \lim_{\mathbf{h} \rightarrow 0} \frac{\mathbf{J}\mathbf{h}}{\mathbf{h}} \triangleq \frac{\partial(\mathbf{J}\mathbf{h})}{\partial \mathbf{h}} = \mathbf{J}$$

# The Jacobian: Example

$$\lim_{\mathbf{h} \rightarrow 0} \frac{f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x})}{\mathbf{h}} = \dots = \lim_{\mathbf{h} \rightarrow 0} \frac{\mathbf{J}\mathbf{h}}{\mathbf{h}} \triangleq \frac{\partial(\mathbf{J}\mathbf{h})}{\partial \mathbf{h}} = \mathbf{J}$$

$$\begin{aligned} \mathbf{J}_x^{\mathbf{A}\mathbf{x}} &= \frac{\partial \mathbf{A}\mathbf{x}}{\partial \mathbf{x}} = \lim_{\mathbf{h} \rightarrow 0} \frac{\mathbf{A}(\mathbf{x} + \mathbf{h}) - \mathbf{A}\mathbf{x}}{\mathbf{h}} \\ &= \lim_{\mathbf{h} \rightarrow 0} \frac{\mathbf{A}\mathbf{x} + \mathbf{A}\mathbf{h} - \mathbf{A}\mathbf{x}}{\mathbf{h}} \\ &= \lim_{\mathbf{h} \rightarrow 0} \frac{\mathbf{A}\mathbf{h}}{\mathbf{h}} \\ &= \mathbf{A}. \end{aligned}$$

# The chain rule

For  $y = f(\mathbf{x})$  and  $z = g(y)$  we have  $z = g(f(\mathbf{x}))$

We can find the derivative of  $z$  with respect to  $\mathbf{x}$  by applying the chain rule:

$$\frac{\partial z}{\partial \mathbf{x}} = \frac{\partial z}{\partial y} \frac{\partial y}{\partial \mathbf{x}}$$

We will use the corresponding notation:

$$\mathbf{J}_{\mathbf{x}}^z = \mathbf{J}_y^z \mathbf{J}_{\mathbf{x}}^y$$

# First-order Taylor approximation

We can linearise the function  $f(\mathbf{x})$  with the first-order Taylor approximation

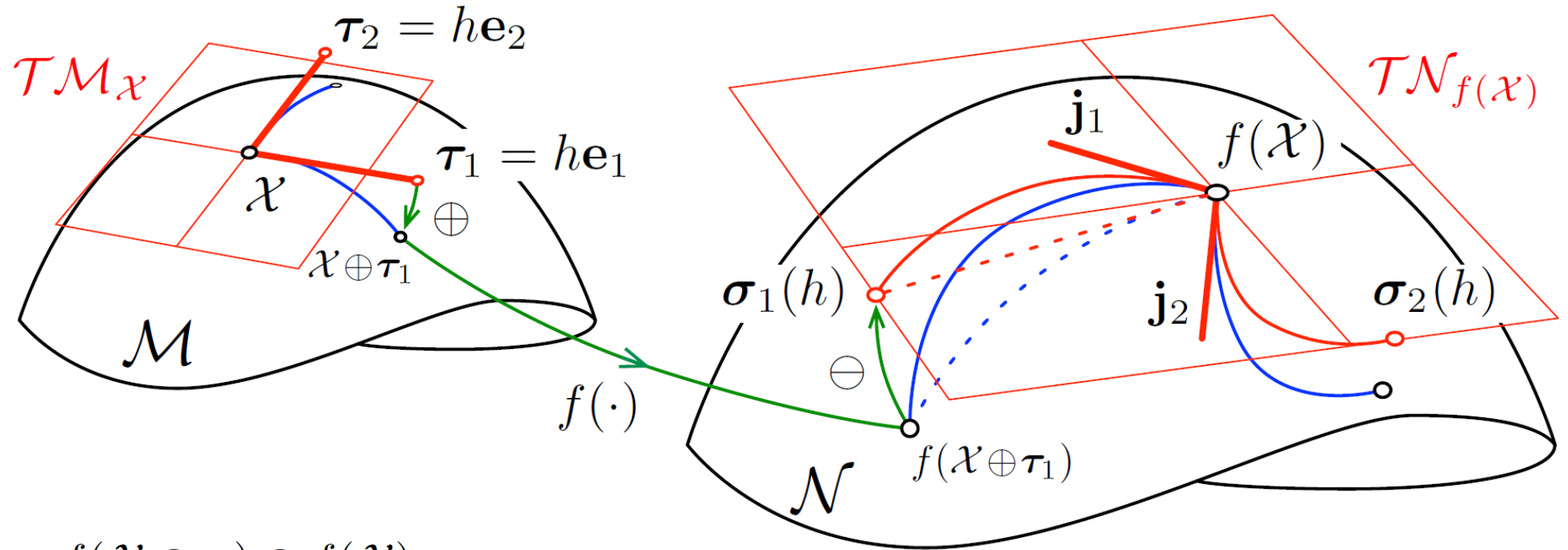
$$f(\mathbf{x} + \mathbf{h}) \xrightarrow{\mathbf{h} \rightarrow 0} f(\mathbf{x}) + \mathbf{J}_{\mathbf{x}}^{f(\mathbf{x})} \mathbf{h}$$

# Derivatives on Lie groups

We can express the derivative of functions acting on Lie groups similarly using the right plus and minus operators:

$$\begin{aligned} \mathbf{J} &= \frac{{}^x \partial f(\mathcal{X})}{\partial \mathcal{X}} \triangleq \lim_{\boldsymbol{\tau} \rightarrow 0} \frac{f(\mathcal{X} \oplus \boldsymbol{\tau}) \ominus f(\mathcal{X})}{\boldsymbol{\tau}} \in \mathbb{R}^{n \times m} \\ &= \lim_{\boldsymbol{\tau} \rightarrow 0} \frac{\text{Log}(f(\mathcal{X})^{-1} \circ f(\mathcal{X} \circ \text{Exp}(\boldsymbol{\tau})))}{\boldsymbol{\tau}} \end{aligned}$$

# Derivatives on Lie groups



$$\mathbf{J} = \frac{{}^x \partial f(x)}{\partial x} \triangleq \lim_{\tau \rightarrow 0} \frac{f(x \oplus \tau) \ominus f(x)}{\tau} \in \mathbb{R}^{n \times m}$$

$$= \lim_{\tau \rightarrow 0} \frac{\text{Log}(f(x)^{-1} \circ f(x \circ \text{Exp}(\tau)))}{\tau}$$

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# Derivatives on Lie groups: Example

$$\begin{aligned} \mathbf{J}_y^{\mathcal{X} \circ \mathcal{Y}} &= \lim_{\tau \rightarrow 0} \frac{(\mathcal{X} \circ (\mathcal{Y} \oplus \tau)) \ominus (\mathcal{X} \circ \mathcal{Y})}{\tau} \\ &= \lim_{\tau \rightarrow 0} \frac{\text{Log}((\mathcal{X} \circ \mathcal{Y})^{-1} \circ (\mathcal{X} \circ (\mathcal{Y} \circ \text{Exp}(\tau))))}{\tau} \\ &= \lim_{\tau \rightarrow 0} \frac{\text{Log}((\mathcal{X} \circ \mathcal{Y})^{-1} (\mathcal{X} \circ \mathcal{Y}) \circ \text{Exp}(\tau))}{\tau} \\ &= \lim_{\tau \rightarrow 0} \frac{\tau}{\tau} \\ &= \mathbf{I}. \end{aligned}$$

# First-order Taylor approximation

We can linearise the function  $f(\mathcal{X})$  with the first-order Taylor approximation

$$f(\mathcal{X} \oplus \boldsymbol{\tau}) \xrightarrow{\boldsymbol{\tau} \rightarrow 0} f(\mathcal{X}) \oplus \mathbf{J}_{\mathcal{X}}^{f(\mathcal{X})} \boldsymbol{\tau}$$

# Elementary Lie group Jacobians

Jacobians for  $SO(3)$  and  $SE(3)$  are given in the compendium!

- Try computing a few of them by hand!
- They are also implemented in [pylie](#)

## 5.4.1 Jacobian of the inverse operation

$$\mathbf{J}_R^{\mathbf{R}^{-1}} = -\text{Ad}_R = -R. \quad (5.50)$$

## 5.4.2 Jacobians of the composition operation

$$\mathbf{J}_{R_a}^{\mathbf{R}_a \mathbf{R}_b} = \text{Ad}_{R_b}^{-1} = R_b^\top \quad (5.51)$$

$$\mathbf{J}_{R_b}^{\mathbf{R}_a \mathbf{R}_b} = \mathbf{I}. \quad (5.52)$$

## 5.4.3 Jacobians of the group action

From Example 5.14 we have

$$\mathbf{J}_R^{\mathbf{R}^x} = -R[x]_\times \quad (5.53)$$

$$\mathbf{J}_x^{\mathbf{R}^x} = R. \quad (5.54)$$

## 5.4.4 Jacobians of the plus and minus operators

$$\mathbf{J}_R^{\mathbf{R} \oplus \theta} = \text{Ad}_{\text{Exp}(\theta)}^{-1} = R(\theta)^\top \quad (5.55)$$

$$\mathbf{J}_\theta^{\mathbf{R} \oplus \theta} = \mathbf{J}_r(\theta). \quad (5.56)$$

For  $\theta = R_a \oplus R_b$ , we have

$$\mathbf{J}_{R_a}^{\mathbf{R}_a \oplus \mathbf{R}_b} = -\mathbf{J}_l^{-1}(\theta) \quad (5.57)$$

$$\mathbf{J}_{R_b}^{\mathbf{R}_a \oplus \mathbf{R}_b} = \mathbf{J}_r^{-1}(\theta). \quad (5.58)$$

## 5.5 Jacobian blocks for $SE(3)$

The left Jacobian and its inverse have the following closed form expressions:

$$\mathbf{J}_l(\xi) = \begin{bmatrix} \mathbf{J}_l(\theta) & \mathbf{Q}(\xi) \\ 0 & \mathbf{J}_l(\theta) \end{bmatrix} \quad (5.59)$$

$$\mathbf{J}_l^{-1}(\xi) = \begin{bmatrix} \mathbf{J}_l^{-1}(\theta) & -\mathbf{J}_l^{-1}(\theta)\mathbf{Q}(\xi)\mathbf{J}_l^{-1}(\theta) \\ 0 & \mathbf{J}_l^{-1}(\theta) \end{bmatrix}. \quad (5.60)$$

# Vectors are also Lie groups!

The group of vectors under addition  $(\mathbb{R}^n, +)$  is a trivial Lie group where the group elements, the Lie algebra and the tangent spaces are all the same:

$$\mathbf{t} = \mathbf{t}^\wedge = \text{Exp}(\mathbf{t})$$

This means that

$$\mathbf{t}_1 \oplus \mathbf{t}_2 = \mathbf{t}_1 + \mathbf{t}_2$$

$$\mathbf{t}_2 \ominus \mathbf{t}_1 = \mathbf{t}_2 - \mathbf{t}_1$$

and that everything we have developed for Lie groups also applies for vectors, including:

$$\mathbf{J} = \frac{{}^x \partial f(\mathcal{X})}{\partial \mathcal{X}} \triangleq \lim_{\boldsymbol{\tau} \rightarrow 0} \frac{f(\mathcal{X} \oplus \boldsymbol{\tau}) \ominus f(\mathcal{X})}{\boldsymbol{\tau}} \in \mathbb{R}^{n \times m}$$