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Document classes

book	Default is two-sided.
report	No <code>\part</code> divisions.
article	No <code>\part</code> or <code>\chapter</code> divisions.
letter	Letter (?).
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Used at the very beginning of a document:
`\documentclass{class}`. Use `\begin{document}` to start contents and `\end{document}` to end the document.

Calculus

Fundamentals

$$\frac{d}{dx} \int_a^x f(t) dt = f(x)$$

Differentiation

In the case of a vector field, the directional derivative is also a vector each of whose components gives the rate of change of the corresponding component of \mathbf{v} in the direction of \mathbf{h} . The gradient in this case will be a tensor field (that when applied to \mathbf{h} gives the directional derivative of \mathbf{v} in the direction of \mathbf{h}).

$$\mathbf{e}_i = \frac{\partial}{\partial x^i} = \partial_i, \quad i = 1, 2, \dots, n$$

define what is referred to as the local [[basis of a vector space|basis]] of the tangent space to $\{\{\mathbf{e}^{\mathbf{M}}\}\}$ at each point of its domain. These can be used to define the [[metric tensor]]:

$$g_{ij} = \mathbf{e}_i \cdot \mathbf{e}_j$$

and its inverse:

$$g^{ij} = (g^{-1})_{ij}$$

which can in turn be used to define the dual basis:

$$\mathbf{e}^i = \mathbf{e}_j g^{ji}, \quad i = 1, 2, \dots, n$$

Some texts write \mathbf{g}_i for \mathbf{e}_i , so that the metric tensor takes the particularly beguiling form $g_{ij} = \mathbf{g}_i \cdot \mathbf{g}_j$. This convention also leaves use of the symbol e_i unambiguously for the [[vierbein]].

Gradient

$$\begin{aligned} \mathbf{e}_i \cdot \text{grad } \varphi(\mathbf{x}) &= [\text{grad } \varphi(\mathbf{x})]_i = \frac{\partial \varphi(\mathbf{x})}{\partial x_i(\mathbf{x})} \\ \mathbf{e}_i \cdot \text{grad } \mathbf{v}(\mathbf{x}) \mathbf{e}_j &= [\text{grad } \mathbf{v}(\mathbf{x})]_{ij} = \frac{\partial v_i(\mathbf{x})}{\partial x_j} \end{aligned}$$

Jacobian In general, the *derivative* of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ at a point $p \in \mathbb{R}^n$, if it exists, is the unique linear transformation $Df(p) \in L(\mathbb{R}^n, \mathbb{R}^m)$ such that

$$\lim_{h \rightarrow 0} \frac{\|f(p+h) - f(p) - Df(p)h\|}{\|h\|} = 0;$$

the matrix of $Df(p)$ with respect to the standard orthonormal bases of \mathbb{R}^n and \mathbb{R}^m , called the *Jacobian matrix* of f at p , therefore lies in $M_{m \times n}(\mathbb{R})$.

Now, suppose that $m = 1$, so that $f : \mathbb{R}^n \rightarrow \mathbb{R}$. Then if f is differentiable at p , $Df(p) \in L(\mathbb{R}^n, \mathbb{R}) = (\mathbb{R}^n)^*$ is a functional, and hence the Jacobian matrix, as you point out, lies in $M_{1 \times n}(\mathbb{R})$, i.e., is a row vector. However, by the Riesz representation theorem, $\mathbb{R}^n \cong (\mathbb{R}^n)^*$ via the map that sends a vector $x \in \mathbb{R}^n$ to the functional $y \mapsto \langle y, x \rangle$. Hence, if f is differentiable at p , then the *gradient* of f at p is the unique (column!) vector $\nabla f(p) \in \mathbb{R}^n$ such that

$$\forall h \in \mathbb{R}^n, \quad Df(p)h = \langle \nabla f(p), h \rangle;$$

in particular, if you unpack definitions, you'll find that the Jacobian matrix of f at p is precisely $\nabla f(p)^T$.

The Jacobian determinant can be viewed as the ratio of an infinitesimal change in the variables of one coordinate system to another. This requires that the function $f(x)$ maps $\mathbb{R}^n \rightarrow \mathbb{R}^n$, which produces an $n \times n$ square matrix for the Jacobian. For example:

$$\iiint_R f(x,y,z) dx dy dz = \iiint_S f(x(u,v,w), y(u,v,w), z(u,v,w)) \left| \frac{\partial(x,y,z)}{\partial(u,v,w)} \right| du dv dw \quad (3.8.4)$$

Directional Derivative

$$\text{grad } \varphi(\mathbf{x})[\mathbf{h}] = \left. \frac{d}{d\alpha} \varphi(\mathbf{x} + \alpha \mathbf{h}) \right|_{\alpha=0}$$

Divergence - Curl- Laplacian

$$\text{div } \mathbf{v} = \text{tr}[\text{grad } \mathbf{v}] = \frac{\partial v_i}{\partial x_i}$$

$$(\text{div } \mathbf{T})_i = \frac{\partial T_{ij}}{\partial x_j}$$

$$(\text{curl } \mathbf{v})_i = e_{ijk} \frac{\partial v_k}{\partial x_j}$$

$$(\text{curl } \mathbf{T})_{ij} = e_{ipq} \frac{\partial T_{jq}}{\partial x_p}$$

$$\Delta \mathbf{v} = \text{div grad } \mathbf{v}, \quad \Delta v_i = \frac{\partial^2 v_i}{\partial x_j \partial x_j}$$

$$\Delta T_{ij} = \frac{\partial T_{ij}}{\partial x_k \partial x_k}$$

Integration

Integration by parts If $u = u(x)$ and $du = u'(x)dx$, while $v = v(x)$ and $dv = v'(x)dx$, then integration by parts states that:

$$\begin{aligned} \int_a^b u(x)v'(x)dx &= [u(x)v(x)]_a^b - \int_a^b u'(x)v(x)dx \\ &= u(b)v(b) - u(a)v(a) - \int_a^b u'(x)v(x)dx \end{aligned}$$

Divergence Theorem

$$\begin{aligned} \int \varphi n_i da &= \int_R \frac{\partial \varphi}{\partial x_i} dv \\ \int_{\partial R} v_i n_i da &= \int_R \frac{\partial v_i}{\partial x_i} dv \\ \int_{\partial R} T_{ij} n_j da &= \int_R \frac{\partial T_{ij}}{\partial x_j} dv \end{aligned}$$

II Kinematics

Displacement:

$$\mathbf{u}(\mathbf{X}, t) = \chi(\mathbf{X}, t) - \mathbf{X}, \quad u_i(X_1, X_2, X_3) = \chi_i(X_1, X_2, X_3) - X_i$$

Velocity/Acceleration

$$\dot{\mathbf{u}}(\mathbf{X}, t) = \frac{\partial \chi(\mathbf{X}, t)}{\partial t} \quad (3.8.3)$$

$$\dot{\mathbf{u}}(\mathbf{X}, t) = \frac{\partial \chi(\mathbf{X}, t)}{\partial t} \quad (3.8.4)$$

3.2 Deformation/Displacement Gradient

$$\mathbf{F}(\mathbf{X}, t) = \frac{\partial}{\partial \mathbf{X}} \chi(\mathbf{X}, t), \quad F_{ij} = \frac{\partial}{\partial X_j} \chi_i(X_1, X_2, X_3, t), \quad \det \mathbf{F}(\mathbf{X}, t) > 0$$

$$\mathbf{H}(\mathbf{X}, t) = \frac{\partial}{\partial \mathbf{X}} \mathbf{u}(\mathbf{X}, t), \quad H_{ij} = \frac{\partial}{\partial X_j} u_i(X_1, X_2, X_3, t)$$

$$\mathbf{H}(\mathbf{X}, t) = \mathbf{F}(\mathbf{X}, t) - \mathbf{1}, \quad H_{ij} = F_{ij} - \delta_{ij}$$

$$J \equiv \det \left(\frac{\partial \chi}{\partial \mathbf{X}} \right) = \det \mathbf{F} = \frac{dv}{dv_R} \neq 0$$

3.3 Stretch & Rotation

Polar Decomposition: $\mathbf{F} = \mathbf{R}\mathbf{U} = \mathbf{V}\mathbf{R}$

$$\begin{aligned} \mathbf{C} &= \mathbf{U}^2 = \mathbf{F}^T \mathbf{F}, \quad C_{ij} = F_{ki} F_{kj} = \frac{\partial \chi_k}{\partial X_i} \frac{\partial \chi_k}{\partial X_j} \\ \mathbf{B} &= \mathbf{V}^2 = \mathbf{F} \mathbf{F}^T, \quad B_{ij} = F_{ik} F_{jk} = \frac{\partial \chi_i}{\partial X_k} \frac{\partial \chi_j}{\partial X_k} \end{aligned}$$

$$\lambda \stackrel{\text{def}}{=} \frac{ds}{dS} = |\mathbf{U}\mathbf{e}| = \sqrt{\mathbf{e} \cdot \mathbf{C}(\mathbf{X})\mathbf{e}}$$

$$\text{where } dS = |d\mathbf{X}|, ds = |d\mathbf{x}|, \mathbf{e} = \frac{d\mathbf{X}}{|d\mathbf{X}|}$$

$$\textbf{Engineering shear: } \gamma = \sin^{-1} \left[\frac{\mathbf{e}^{(1)} \cdot \mathbf{C} \mathbf{e}^{(2)}}{\lambda(\mathbf{e}^{(1)}) \lambda(\mathbf{e}^{(2)})} \right]$$

3.4 Strain

$$\textbf{Green strain: } \mathbf{E} \stackrel{\text{def}}{=} \frac{1}{2} \left(\mathbf{F}^T \mathbf{F} - \mathbf{1} \right) = \frac{1}{2} \left(\mathbf{H} + \mathbf{H}^T + \mathbf{H}^T \mathbf{H} \right)$$

$$\textbf{Hencky's Log strain: } \ln \mathbf{U} \stackrel{\text{def}}{=} \sum_{i=1}^3 (\ln \lambda_i) \mathbf{r}_i \otimes \mathbf{r}_i$$

3.5.2 Infinitesimal Strain

ϵ' : distortion $\epsilon_M \delta_{ij}$: dilation

$$\begin{aligned}\epsilon &= \frac{1}{2} \left[\mathbf{H} + \mathbf{H}^\top \right], & \epsilon &= \epsilon^\top, \quad |\mathbf{H}| \ll 1 \\ \epsilon_{ij} &= \frac{1}{2} \left[\frac{\partial u_i}{\partial X_j} + \frac{\partial u_j}{\partial X_i} \right], & \epsilon_{ji} &= \epsilon_{ij}, \quad \left| \frac{\partial u_i}{\partial X_j} \right| \ll 1\end{aligned}$$

3.A Linearization

$$\begin{aligned}\lim_{\alpha \rightarrow 0} \mathbf{Y}_o f(\mathbf{Y}) &= f(\mathbf{Y}_o) + \left. \frac{d}{d\alpha} f(\mathbf{Y}_o + \alpha(\mathbf{Y} - \mathbf{Y}_o)) \right|_{\alpha=0} \\ \lim_0 f(\mathbf{H}) &= f(0) + \left. \frac{d}{d\alpha} f(\alpha \mathbf{H}) \right|_{\alpha=0}\end{aligned}$$

3.B Compatibility

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