

MATH 3742

We just saw that the derivative of $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ can be represented by the Jacobian Df , when it exists.

The formula for the Jacobian also reinforces something we already discovered, namely that

$$f'(\vec{c}; \vec{u}) = \vec{u}_1 \frac{\partial f}{\partial x_1}(\vec{c}) + \dots + \vec{u}_n \frac{\partial f}{\partial x_n}(\vec{c}) \quad \text{if } \vec{u} = (u_1, \dots, u_n),$$

when $f: \mathbb{R}^n \rightarrow \mathbb{R}$.

That's because when $m=1$ (so f is real-valued), then Df is a row:

$$Df(\vec{c}) = [D_1 f(\vec{c}) \ D_2 f(\vec{c}) \ \dots \ D_n f(\vec{c})]$$

and the directional derivative in the direction of \vec{u} is

$$[Df(\vec{c})] \vec{u} = [D_1 f(\vec{c}) \ \dots \ D_n f(\vec{c})] \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} = \sum_{i=1}^n u_i D_i f(\vec{c}).$$

Or, using the gradient ∇f of $f: \mathbb{R}^n \rightarrow \mathbb{R}$, this can be written as

$$Df(\vec{c}) = \nabla f(\vec{c}) \quad (\text{again, only if } m=1)$$

and for a vector-valued function $f = (f_1, \dots, f_m)$

$$(*) \quad Df(\vec{c})(\vec{v}) = f'(\vec{c}; \vec{v}) = \underbrace{\sum_{k=1}^m f'_k(\vec{c}; \vec{v}) \hat{e}_k}_{\text{this is just from the definition of}} = \underbrace{\sum_{k=1}^m (\nabla f_k(\vec{c}) \cdot \vec{v}) \hat{e}_k}_{\text{matrix multiplication, and the}}$$

this is just from the definition of matrix multiplication, and the

definition of the notation

$$\nabla f_k(\vec{c}) = (D_1 f_k(\vec{c}), \dots, D_n f_k(\vec{c})).$$

We need this equation for a lemma.

Lemma: With the setup above,

$$\|Df(\vec{c})(\vec{v})\| \leq \|\vec{v}\| \sum_{k=1}^m \|\nabla f_k(\vec{c})\|, \text{ and } \lim_{\vec{v} \rightarrow \vec{0}} Df(\vec{c})(\vec{v}) = \vec{0}.$$
$$= M \|\vec{v}\|$$

Proof: From equation (*), the inequality we want follows from the Cauchy-Schwarz inequality since

$$\left\| \sum_{k=1}^m (\nabla f_k(\vec{c}) \cdot \vec{v}) \vec{e}_k \right\| \leq \|\vec{v}\| \sum_{k=1}^m \|\nabla f_k(\vec{c})\|$$

$$\frac{\left\| \sum_{k=1}^m f'_k(\vec{c}; \vec{v}) \vec{e}_k \right\|}{\|Df(\vec{c})(\vec{v})\|}$$

This is Cauchy-Schwarz,
recall it says

$$|\vec{u} \cdot \vec{v}| \leq \|\vec{u}\| \|\vec{v}\|.$$

Second, $\lim_{\vec{v} \rightarrow \vec{0}} \|Df(\vec{c})(\vec{v})\| \leq \lim_{\vec{v} \rightarrow \vec{0}} \|\vec{v}\| \underbrace{\sum_{k=1}^m \|\nabla f_k(\vec{c})\|}_{\text{constant wrt } \vec{v}} = 0,$

and thus $Df(\vec{c})(\vec{v}) \rightarrow \vec{0}$ as $\vec{v} \rightarrow \vec{0}$.

We will use this lemma to prove:

§12.9 The chain rule.

Theorem 12.7. Let $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $g: \mathbb{R}^p \rightarrow \mathbb{R}^m$ be functions such that their composition $h = f \circ g$

is defined in a neighbourhood of $\vec{a} \in \mathbb{R}^p$. Suppose g is differentiable at \vec{a} with total derivative $g'(\vec{a})$, and f is differentiable at $\vec{b} = g(\vec{a})$ with total derivative $f'(\vec{b})$. Then $h = f \circ g$ is differentiable at $\vec{a} \in \mathbb{R}^p$ and

$$h'(\vec{a}) = \underbrace{f'(\vec{b}) \cdot g'(\vec{a})}_{\text{matrix multiplication, or composition of linear functions, depending on how you are thinking of total derivatives.}}$$

Proof: We consider $h(\vec{a} + \vec{y}) - h(\vec{a})$ for small $\|\vec{y}\|$, and show $h(\vec{a} + \vec{y}) - h(\vec{a}) = [f'(\vec{b}) \cdot g'(\vec{a})] \vec{y} + \|\vec{y}\| E_a(\vec{y})$.

for an appropriate error function $E_a(\vec{y})$.

First

$$h(\vec{a} + \vec{y}) - h(\vec{a}) = (f \circ g)(\vec{a} + \vec{y}) - (f \circ g)(\vec{a}) = f(\vec{b} + \vec{v}) - f(\vec{b}),$$

if we take $\vec{b} = g(\vec{a})$ and $\vec{v} = g(\vec{a} + \vec{y}) - \vec{b}$.

Then since g obeys a Taylor formula we get

$$\vec{v} = g'(\vec{a})(\vec{y}) + \|\vec{y}\| E_a(\vec{y}), \quad E_a(\vec{y}) \rightarrow \vec{0} \text{ as } \vec{y} \rightarrow \vec{0},$$

and since f obeys a Taylor formula we get

$$f(\vec{b} + \vec{v}) - f(\vec{b}) = f'(\vec{b})(\vec{v}) + \|\vec{v}\| E_b(\vec{v}), \quad \text{where}$$

$$E_b(\vec{v}) \rightarrow \vec{0} \text{ as } \vec{v} \rightarrow \vec{0}.$$

Combining these, we get:

$$f(\vec{b} + \vec{v}) - f(\vec{b}) = f'(\vec{b})(g(\vec{a})(\vec{y}) + \|\vec{y}\| E_a(\vec{y})) + \|\vec{v}\| E_b(\vec{v})$$

$$= f'(\vec{b}) \cdot g'(\vec{a})(\vec{y}) + E(\vec{y}) \|\vec{y}\|, \text{ where}$$

$$\|\vec{y}\| E(\vec{y}) = f'(\vec{b}) \|\vec{y}\| E_a(\vec{y}) + \|\vec{v}\| E_b(\vec{v})$$

$$\Rightarrow E(\vec{y}) = f'(\vec{b}) E_a(\vec{y}) + \frac{\|\vec{v}\|}{\|\vec{y}\|} E_b(\vec{v}) \text{ if } \vec{y} \neq 0.$$

Thus the proof is done if we can show $E(\vec{y}) \rightarrow 0$ as $\vec{y} \rightarrow \vec{0}$.

Now as $\vec{y} \rightarrow \vec{0}$ the formula for \vec{v}

$$\vec{v} = g'(\vec{a})(\vec{y}) + \|\vec{y}\| E_a(\vec{y})$$

shows that $\vec{v} \rightarrow \vec{0}$ as well. Thus the term $f'(\vec{b}) E_a(\vec{y}) \rightarrow \vec{0}$ as $\vec{y} \rightarrow \vec{0}$, but $\frac{\|\vec{v}\|}{\|\vec{y}\|} E_b(\vec{v})$ may or may not. It depends on "how fast" $\vec{v} \rightarrow \vec{0}$ as $\vec{y} \rightarrow \vec{0}$. We'll show $\frac{\|\vec{v}\|}{\|\vec{y}\|}$ is bounded as $\|\vec{y}\| \rightarrow 0$, completing the proof.

We estimate:

$$\begin{aligned} \|\vec{v}\| &\leq \|g'(\vec{a})(\vec{y})\| + \|\vec{y}\| \|E_a(\vec{y})\| \\ &\leq \|\vec{y}\| (M + \|E_a(\vec{y})\|) \end{aligned}$$

This is the M
provided by our previous
lemma

Thus

$$\frac{\|\vec{v}\|}{\|\vec{y}\|} \leq M + \|E_a(\vec{y})\|, \text{ so as } \vec{y} \rightarrow \vec{0} \text{ this term remains bounded.}$$

It follows that $\lim_{\vec{y} \rightarrow 0} \frac{\|\vec{v}\|}{\|\vec{y}\|} E_b(\vec{v}) = \vec{0}$,

so $\lim_{\vec{y} \rightarrow 0} E(\vec{y}) = 0$ and we have the Taylor formula

$$h(\vec{a} + \vec{y}) - h(\vec{a}) = f'(b) \cdot g'(\vec{a})(\vec{y}) + \|\vec{y}\| E(\vec{y})$$

as desired. So the total derivative of $h = f \circ g$ at \vec{a} is $f'(b) \cdot g'(\vec{a})$.

Start here

This means that the Jacobians Dh , Df , Dg obey the law

$$Dh(\vec{a}) = Df(b) Dg(\vec{a}),$$

which our book calls "the matrix form of the chain rule". This reduces to more familiar-looking forms in low dimensions. E.g.

Example: If $m=1$, so that $h = f \circ g$ is a real-valued function. Then we have, if ~~$h = (h_1, \dots, h_m)$~~ $h = h(y_1, \dots, y_n)$

$$\left[\frac{\partial h}{\partial y_1}(\vec{a}) \quad \frac{\partial h}{\partial y_2}(\vec{a}) \quad \dots \quad \frac{\partial h}{\partial y_n}(\vec{a}) \right] = \left[\frac{\partial f}{\partial x_1}(b) \quad \dots \quad \frac{\partial f}{\partial x_p}(b) \right]$$

$$\cdot \begin{bmatrix} \frac{\partial g_1}{\partial y_1}(\vec{a}) & \frac{\partial g_1}{\partial y_2}(\vec{a}) & \dots & \frac{\partial g_1}{\partial y_n}(\vec{a}) \\ \vdots & & & \\ \frac{\partial g_p}{\partial y_1}(\vec{a}) & \frac{\partial g_p}{\partial y_2}(\vec{a}) & \dots & \frac{\partial g_p}{\partial y_n}(\vec{a}) \end{bmatrix}$$

1947. 11. 22.

So that

$$\frac{\partial h}{\partial y_i}(\bar{a}) = \frac{\partial f \circ g}{\partial y_i}(a) = \sum_{k=1}^p \frac{\partial f}{\partial x_k} \frac{\partial g_k}{\partial y_i}, \text{ which we normally}$$

would write as

$$= \sum_{k=1}^p \frac{\partial f}{\partial x_k} \frac{\partial x_k}{\partial y_i}, \text{ thinking of the}$$

output of the k^{th} coordinate function of
 $g: \mathbb{R}^n \rightarrow \mathbb{R}^p$ as the k^{th} argument of
 $f: \mathbb{R}^p \rightarrow \mathbb{R}$.

This gives the familiar chain rule for calculating derivatives of functions $\mathbb{R}^n \rightarrow \mathbb{R}$.

MATH 3472

Example (Chain rule matrix version vs. case of functions $\mathbb{R}^n \rightarrow \mathbb{R}$)

Define $g: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ by

$$g(s,t) = (s^2t, st+2t^2, st)$$

(from Dartmouth College notes)

and $f: \mathbb{R}^3 \rightarrow \mathbb{R}$ by

$$f(x,y,z) = e^{2x-y+z}.$$

We will calculate the derivative of $h = f \circ g$ in two ways at the point $(1,1)$.

Now $g(1,1) = (1, 3, 1)$ and so

$$(f \circ g)'(1,1) = f'(1, 3, 1) \circ g'(1, 1)$$

Compute:

$$g'(s,t) = \begin{bmatrix} D_1g_1(s,t) & D_2g_1(s,t) \\ D_1g_2(s,t) & D_2g_2(s,t) \\ D_1g_3(s,t) & D_2g_3(s,t) \end{bmatrix} = \begin{bmatrix} 2st & s^2 \\ 1 & 4t \\ t & s \end{bmatrix}$$

$$\Rightarrow g'(1,1) = \begin{bmatrix} 2 & 1 \\ 1 & 4 \\ 1 & 1 \end{bmatrix}$$

and similarly

$$f'(x,y,z) = [D_1f(x,y,z) \ D_2f(x,y,z) \ D_3f(x,y,z)]$$

$$= [2e^{2x-y+z} \ -e^{2x-y+z} \ e^{-2x-y+z}]$$

Therefore

$$f'(1,3,1) = \begin{bmatrix} 2 & -1 & 1 \end{bmatrix}. \text{ Then we compute}$$

$$(f \circ g)'(1,1) = \begin{bmatrix} 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 4 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 4 & -1 \end{bmatrix}.$$

Another way of seeing this is to use the notation just discussed, ie the equation $\frac{\partial y}{\partial t_j} = \sum_{k=1}^n \frac{\partial y}{\partial x_k} \frac{\partial x_k}{\partial t_j}$. In our case,

set $w = f(x,y,z)$ and $(x,y,z) = g(s,t)$ (so x = outputs of coordinate function g_1 , etc).

Then the above formula in our present setting gives

$$\frac{\partial w}{\partial s} = \frac{\partial w}{\partial x} \cdot \frac{\partial x}{\partial s} + \frac{\partial w}{\partial y} \cdot \frac{\partial y}{\partial s} + \frac{\partial w}{\partial z} \cdot \frac{\partial z}{\partial s}$$

then compute

$$\frac{\partial w}{\partial s} = 2(2) + (-1)(1) + (1)(1) = 4, \text{ which}$$

agrees with the first entry in our matrix. Similarly

$$\frac{\partial w}{\partial t} = \frac{\partial w}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial w}{\partial y} \cdot \frac{\partial y}{\partial t} + \frac{\partial w}{\partial z} \cdot \frac{\partial z}{\partial t} = -1.$$

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Sample application of Chain Rule.

Theorem: Suppose f and $D_2 f$ are continuous on $[a,b] \times [c,d]$. Let $p(y), q(y)$ be functions $p,q: [c,d] \rightarrow [a,b]$. Define

$$F(y) = \int_{p(y)}^{q(y)} f(x,y) dx \text{ if } y \in [c,d]$$

Then $F'(y)$ exists and for each $y \in [c, d]$ we have

$$F'(y) = \int_{p(y)}^{q(y)} D_2 f(x, y) dx + f(q(y), y) q'(y) - f(p(y), y) p'(y).$$

Remark: This is viewed as an improvement of a theorem proved in Chapter 7 (Riemann-Stieltjes integration).

Namely if α is of bounded variation on $[a, b]$, $f \in R(\alpha)$ and

$$F(x) = \int_a^x f d\alpha \quad (x \in [a, b])$$

f Riemann int. on $[a, b]$ wrt α .

then

$F'(x) = f(x) \alpha'(x)$ (provided α is increasing and $\alpha'(x)$ exists. When $\alpha(x) = x$ is the identity, this reduces to the expected formula:

$$\left(\int_a^x f dx \right)' = f(x).$$

Proof of theorem: Set

$$G(x_1, x_2, x_3) = \int_{x_1}^{x_2} f(t, x_3) dt \text{ whenever } x_1, x_2 \in [a, b]$$

and $x_3 \in [c, d]$. Then we can rewrite F as

$$F = G(p(y), q(y), y)$$

i.e. the composition

$$\mathbb{R} \xrightarrow{(p(y), q(y), y)} \mathbb{R}^3 \xrightarrow{G} \mathbb{R}$$

so the chain rule gives

$$F'(y) = \begin{bmatrix} D_1 G(p(y), q(y), y) & D_2 G(p(y), q(y), y) & D_3 G(p(y), q(y), y) \end{bmatrix} \begin{bmatrix} p'(y) \\ q'(y) \\ 1 \end{bmatrix}$$
$$\Rightarrow F'(y) = D_1 G(p(y), q(y), y) p'(y) + D_2 G(p(y), q(y), y) q'(y) + D_3 G(p(y), q(y), y)$$

By Theorems from Chapter 7,

$$D_1 G(x_1, x_2, x_3) = -f(x_1, x_3) \text{ and } D_2 G(x_1, x_2, x_3) = f(x_2, x_3).$$

and

$$D_3 G(x_1, x_2, x_3) = \int_{x_1}^{x_2} D_2 f(t, x_3) dt.$$

These equations result in the claimed formula for $F'(y)$.

Note : $D_1 G$ and $D_2 G$ are relatively straight forward consequences of what is called (sometimes) the fundamental theorem of calculus:

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

Application: Multivariable chain rule and change of coordinates.

Recall that cartesian (x,y) -coordinates are related to polar coordinates (r,θ) via

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

$$r = \sqrt{x^2 + y^2}, \quad \theta = \tan^{-1} \frac{x}{y}.$$

Suppose we have a function expressed in xy -coordinates, like $g(x,y) = x^2y^3$. Then its derivative (in xy coordinates)

is

$$Dg(x,y) = [2xy^3 \quad 3x^2y^2].$$

What is its derivative with respect to (r,θ) coordinates? One way: plug $x = r\cos\theta, y = r\sin\theta$ into g , and differentiate. Alternatively, set

$$f: \mathbb{R}^2 \longrightarrow \mathbb{R}^2, \quad f(r,\theta) = (r\cos\theta, r\sin\theta)$$

Then $g(r,\theta) = g \circ f(r,\theta)$, and the chain rule gives

$$D(g \circ f)(r,\theta) = Dg(f(r,\theta)) \circ Df(r,\theta)$$

$$= [2(r\cos\theta)(r\sin\theta)^3 \quad 3(r\cos\theta)^2(r\sin\theta)^2] \cdot \begin{bmatrix} \cos\theta & -r\sin\theta \\ \sin\theta & r\cos\theta \end{bmatrix}$$

$$= [5r^4\cos^2\theta\sin^3\theta \quad r^5\cos\theta\sin^2\theta(3\cos^2\theta - 2\sin^2\theta)]$$

$$\frac{\partial g}{\partial r}^{\parallel}$$

$$\frac{\partial g}{\partial \theta}^{\parallel}$$

This could have been done by substituting at the beginning $x=r\cos\theta$, $y=r\sin\theta$. But this is not always possible.

Example: Suppose $u:\mathbb{R}^2 \rightarrow \mathbb{R}$ satisfies

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = 0 \quad (\text{Laplace's equation}).$$

What form would Laplace's equation take in polar coordinates?

From the chain rule,

$$Du(r, \theta) = D(u \circ f)(r, \theta) = Du(f(r, \theta)) \circ Df(r, \theta)$$

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

But the matrix on the right is invertible, with inverse

$$\begin{bmatrix} \cos\theta & \sin\theta \\ -\frac{\sin\theta}{r} & \frac{\cos\theta}{r} \end{bmatrix}, \text{ so left-multiplying the above eqn by}$$

this inverse gives:

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

Now we compute second derivatives, and get

$$\left. \begin{aligned} \frac{\partial^2 u}{\partial x^2} &= \frac{\partial}{\partial x} \left(\frac{\partial u}{\partial x} \right) = \frac{\partial}{\partial x} (v) \\ &= \cos \theta \frac{\partial v}{\partial r} - \frac{\sin \theta}{r} \frac{\partial v}{\partial \theta} \\ &= \cos \theta \frac{\partial}{\partial r} \left(\cos \theta \frac{\partial u}{\partial r} - \frac{\sin \theta}{r} \frac{\partial u}{\partial \theta} \right) - \frac{\sin \theta}{r} \frac{\partial}{\partial \theta} \left(\sin \theta \frac{\partial u}{\partial r} + \frac{\cos \theta}{r} \frac{\partial u}{\partial \theta} \right) \\ &= \text{some equation giving } \frac{\partial^2 u}{\partial x^2} \text{ in terms of } r, \theta. \end{aligned} \right\} \begin{array}{l} \text{since } u \text{ was arbitrary, same} \\ \text{formula holds for } v \end{array}$$

Similarly we can compute $\frac{\partial^2 u}{\partial y^2}$, and substitute into

Laplace's equation. We get:

$$\frac{\partial^2 u}{\partial r^2} + \frac{1}{r^2} \frac{\partial^2 u}{\partial \theta^2} + \frac{1}{r} \frac{\partial u}{\partial r} = 0.$$

§ 12.11 The Mean Value Theorem.

If $f: \mathbb{R} \rightarrow \mathbb{R}$, then the Mean-Value theorem says that if f is differentiable on $[a, b]$ then $\exists c$ such that

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

When $m > 1$, no such formula can hold: Consider $f: \mathbb{R} \rightarrow \mathbb{R}^2$ given by

$$f(t) = (\cos t, \sin t), \text{ and then}$$

$$f'(t)(u) = u[-\sin t, \cos t]^T \text{ for all } u \in \mathbb{R}.$$

Now consider on the interval $[0, 2\pi]$. There we find that the MVT equation gives

$$f(0) - f(2\pi) = \underbrace{u[-\sin t, \cos t]^T \cdot 2\pi}_{\begin{array}{l} \parallel \\ (1, 0) - (1, 0) \\ \parallel \\ \vec{0} \end{array}} \quad \text{a vector of length } 2\pi|u|.$$

so the two sides cannot be equal. So the "obvious" generalization from functions $f: \mathbb{R} \rightarrow \mathbb{R}$ to functions $\mathbb{R} \rightarrow \mathbb{R}^m$ cannot hold.

The proper generalization is as follows: Use the notation

$$L(\vec{x}, \vec{y}) = \{t\vec{x} + (1-t)\vec{y} \mid t \in [0, 1]\}$$

to denote the line segment connecting \vec{x} and \vec{y} .

Theorem (Mean-Value Theorem) Let $S \subset \mathbb{R}^n$ be open and assume $f: S \rightarrow \mathbb{R}^m$ is differentiable at every $s \in S$. Choose two points $\vec{x}, \vec{y} \in S$ such that $L(\vec{x}, \vec{y}) \in S$. Then for every vector $\vec{a} \in \mathbb{R}^m$ there is a point $z \in L(\vec{x}, \vec{y})$ such that

$$\vec{a} \cdot (f(\vec{y}) - f(\vec{x})) = \vec{a} \cdot \underbrace{(f'(\vec{z})(\vec{y} - \vec{x}))}_{\text{matrix multiplication!}}$$

Proof: Let $\vec{u} = \vec{y} - \vec{x}$. Then S is open and $L(\vec{x}, \vec{y}) \subseteq S$, so $\exists \delta > 0$ such that $\vec{x} + t\vec{u} \in S$ for all $t \in (-\delta, 1+\delta)$.

Choose $\vec{a} \in \mathbb{R}^m$ and define $F: \mathbb{R} \rightarrow \mathbb{R}$ by

$$F(t) = \vec{a} \cdot f(\vec{x} + t\vec{u})$$

(actually, $F: (-\delta, 1+\delta) \rightarrow \mathbb{R}$).

Then F is differentiable with derivative

$$\begin{aligned} F'(t) &= \vec{a} \cdot f'(\vec{x} + t\vec{u}; \vec{u}) && \leftarrow \text{directional derivative} \\ &= \vec{a} \cdot f'(\vec{x} + t\vec{u})(\vec{u}) && \leftarrow \text{total derivative} \end{aligned}$$

We can apply the usual mean value theorem to F on $[0, 1]$ to arrive at $\theta \in (0, 1)$ such that

$$F(1) - F(0) = F'(\theta)(1-0) = F'(\theta),$$

and then observe

$$F'(\theta) = \vec{a} \cdot f'(\vec{x} + \theta\vec{u})(\vec{u}),$$

If we set $\vec{z} = \vec{x} + \theta \vec{u}$, this becomes

$$\vec{a} \cdot f'(\vec{z})(\vec{y} - \vec{x}) \quad (\text{recall } \vec{u} = \vec{y} - \vec{x} \text{ by def.})$$

Further, $F(1) - F(0)$

$$\begin{aligned} &= \vec{a} \cdot f(\vec{x} + \vec{u}) - \vec{a} \cdot f(\vec{x}) \\ &= \vec{a} \cdot (f(\vec{y}) - f(\vec{x})), \end{aligned}$$

so we arrive at the desired equation.

Remarks: ① If $f: \mathbb{R} \rightarrow \mathbb{R}$ note that this reduces to the original MVT, and the " \vec{a} " term becomes redundant (in the sense that it can simply be canceled from each side of the equation, it's just multiplication by a scalar)

② In general, there seems to be some disagreement between texts as to what the "correct" generalization of the MVT to higher dimensions should be.

E.g. Marsden-Hoffman states it as:

(i) If $f: S \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable on S open, then $\forall \vec{x}, \vec{y} \in S$ st. $L(\vec{x}, \vec{y}) \subset S$ $\exists c \in L(\vec{x}, \vec{y})$ such that

$$f(\vec{y}) - f(\vec{x}) = Df(\vec{c}) \cdot (\vec{y} - \vec{x})$$

and

(ii) If $f: S \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, S open, then $\exists c_1, \dots, c_m$ on $L(\vec{x}, \vec{y})$ such that

$$f_i(\vec{y}) - f_i(\vec{x}) = Df_i(\vec{c})(\vec{y} - \vec{x}), \text{ for } i=1, \dots, m$$

where $f = (f_1, \dots, f_m)$.

Certainly, wikipedia supports the notion that no generalization to $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is widely accepted as "correct".

The Apostol version of the MVT does allow for the following simple proof, however, which is analogous to the development of calculus for functions $\mathbb{R} \rightarrow \mathbb{R}$.

Theorem: Let S be an open, connected subset of \mathbb{R}^n , and $f: S \rightarrow \mathbb{R}^m$ differentiable at all points in S . If $f'(\vec{c}) = 0$ for all $\vec{c} \in S$, then f is constant on S .

Proof:

Since S is open and connected, every pair of points \vec{x} and \vec{y} in S can be connected by a polygonal arc (see Chapter 4). Denote the vertices of this arc by $\vec{x} = \vec{p}_1, \vec{p}_2, \vec{p}_3, \dots, \vec{p}_r = \vec{y}$.

Since each segment $L(\vec{p}_{i+1}, \vec{p}_i) \subseteq S$, the MVT

$$\text{gives } \vec{a} \cdot (f(\vec{p}_{i+1}) - f(\vec{p}_i)) = \vec{0}$$

for every \vec{a} , assuming the derivative is everywhere $\vec{0}$.

If we ~~take~~ add together all these equations for $i=1, \dots, r-1$, we get

$$\vec{a} \cdot (f(\vec{x}) - f(\vec{y})) = \vec{0}.$$

But \vec{a} can be anything! So in particular, when \vec{a} is $f(\vec{x}) - f(\vec{y})$ we get

$$\|f(\vec{x}) - f(\vec{y})\| = 0$$

so $f(\vec{x}) = f(\vec{y})$ and f is constant.