

Advanced Calculus (II)

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11.5: Mean Value Theorem and Taylor's Formula

Theorem (11.31 Mean Value Theorem on \mathbf{R}^n)

Let V be open in \mathbf{R}^n and suppose that $f : V \rightarrow \mathbf{R}^m$ is differentiable on V . If $\mathbf{x}, \mathbf{a} \in V$ and $L(\mathbf{x}; \mathbf{a}) \subseteq V$, then for each $\mathbf{u} \in \mathbf{R}^m$, there is a $\mathbf{c} \in L(\mathbf{x}; \mathbf{a})$ such that

$$\mathbf{u} \cdot (f(\mathbf{x}) - f(\mathbf{a})) = \mathbf{u} \cdot (Df(\mathbf{c})(\mathbf{x} - \mathbf{a})).$$

Proof.

Let

$$g(t) = \mathbf{a} + t(\mathbf{x} - \mathbf{a}), \quad t \in \mathbf{R},$$

and notice by Exercise 8, p.338, that $g : \mathbf{R} \rightarrow \mathbf{R}^n$ is differentiable with $Dg(t) = \mathbf{x} - \mathbf{a}$ for all $t \in \mathbf{R}$. Since $L(\mathbf{x}; \mathbf{a}) \subseteq V$ and V is open, choose $\delta > 0$ such that $g(t) \in V$ for all $t \in I_\delta := (-\delta, 1 + \delta)$. By the Chain Rule,

$$(24) \quad D(f \circ g)(t) = Df(g(t))(\mathbf{x} - \mathbf{a}), \quad t \in I_\delta.$$

Fix $\mathbf{u} \in \mathbf{R}^m$, and consider the function

$$F(t) = \mathbf{u} \cdot (f \circ g)(t), \quad t \in I_\delta.$$



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The function F is a real-valued function on I_δ . By the Dot Product Rule: $D(f \cdot g)(\mathbf{a}) = g(\mathbf{a})Df(\mathbf{a}) + f(\mathbf{a})Dg(\mathbf{a})$ and (24), F is differentiable on I_δ with

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Definition (11.32)

A subset E of \mathbf{R}^n is said to be *convex* if and only if $L(\mathbf{x}; \mathbf{a}) \subseteq E$ for all $\mathbf{x}, \mathbf{a} \in E$.

Corollary (11.33)

Let V be convex and open in \mathbf{R}^n and suppose that $f : V \rightarrow \mathbf{R}$. If f is differentiable on V and $\mathbf{a} + \mathbf{h}$, \mathbf{a} both belong to V , then there is a $0 < t < 1$ such that

$$(25) \quad f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) = Df(\mathbf{a} + t\mathbf{h}) \cdot \mathbf{h} := \sum_{j=1}^n \frac{\partial f}{\partial x_j}(\mathbf{a} + t\mathbf{h}) h_j.$$

Proof.

Let u be a nonzero scalar, and suppose that $\mathbf{a} + \mathbf{h}$, \mathbf{a} both belong to V . Since V is convex, $L(\mathbf{a} + \mathbf{h}; \mathbf{a}) \subseteq V$. Hence, by Theorem 11.31,

$$u(f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a})) = u(\nabla f(\mathbf{c}) \cdot \mathbf{h})$$

for some $\mathbf{c} \in L(\mathbf{a} + \mathbf{h}; \mathbf{a})$. Dividing this inequality by u and choosing $t \in (0, 1)$ such that $\mathbf{c} = \mathbf{a} + t\mathbf{h}$, we conclude that (25) holds. □

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Corollary (11.34)

Let V be an open set in \mathbf{R}^n , let H be a compact subset of V , and suppose that $f : V \rightarrow \mathbf{R}^m$ is \mathcal{C}^1 on V . If E is a convex subset of H , then there is a constant M (which depends on H and f but not on E) such that

$$\|f(\mathbf{x}) - f(\mathbf{a})\| \leq M\|\mathbf{x} - \mathbf{a}\|$$

for all $\mathbf{x}, \mathbf{a} \in E$.

Proof.

Since H is compact and the entries of Df are continuous on H , we have by the Extreme Value Theorem (Theorem 9.32 or 10.63) and the proof of Theorem 8.17 that the operator norm of Df is bounded on H , i.e., that

$$M := \sup_{c \in H} \|Df(c)\|$$

is finite. Notice that M depends only on H and f .

Let $\mathbf{x}, \mathbf{a} \in E$ and $\mathbf{u} = f(\mathbf{x}) - f(\mathbf{a})$. Since E is convex, $L(\mathbf{x}; \mathbf{a}) \subseteq E$. Hence, by Theorem 11.31, there is a $\mathbf{c} \in L(\mathbf{x}; \mathbf{a})$ such that

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Let $\mathbf{x}, \mathbf{a} \in E$ and $\mathbf{u} = f(\mathbf{x}) - f(\mathbf{a})$. Since E is convex, $L(\mathbf{x}; \mathbf{a}) \subseteq E$. Hence, by Theorem 11.31, there is a $\mathbf{c} \in L(\mathbf{x}; \mathbf{a})$ such that

$$\begin{aligned} \|f(\mathbf{x}) - f(\mathbf{a})\|^2 &= \mathbf{u} \cdot (f(\mathbf{x}) - f(\mathbf{a})) = \mathbf{u} \cdot (Df(\mathbf{c})(\mathbf{x} - \mathbf{a})) \\ &= (f(\mathbf{x}) - f(\mathbf{a})) \cdot (Df(\mathbf{c})(\mathbf{x} - \mathbf{a})). \end{aligned}$$

Proof.

It follows from the Cauchy-Schwarz Inequality and the definition of the operator norm that

$$\|f(\mathbf{x}) - f(\mathbf{a})\|^2 \leq \|f(\mathbf{x}) - f(\mathbf{a})\| \|Df(\mathbf{c})\| \|\mathbf{x} - \mathbf{a}\|.$$

If $\|f(\mathbf{x}) - f(\mathbf{a})\| = 0$, there is nothing to prove. Otherwise, we can divide the inequality above by $\|f(\mathbf{x}) - f(\mathbf{a})\|$ to obtain

$$\|f(\mathbf{x}) - f(\mathbf{a})\| \leq \|Df(\mathbf{c})\| \|\mathbf{x} - \mathbf{a}\| \leq M \|\mathbf{x} - \mathbf{a}\|.$$



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Corollary (11.35)

Suppose that V is open and connected in \mathbf{R}^n and that $f : V \rightarrow \mathbf{R}^m$ is differentiable on V . If $Df(\mathbf{c}) = \mathbf{0}$ for all $\mathbf{c} \in V$, then f is constant on V .

Proof.

Fix $\mathbf{a} \in V$, and let $\mathbf{x} \in V$. Since V is open and connected, V is polygonally connected (see Exercise 10, p.277).

Thus, there exist points $\mathbf{x}_0 = \mathbf{a}, \mathbf{x}_1, \dots, \mathbf{x}_k = \mathbf{x}$ such that $L(\mathbf{x}_{j-1}; \mathbf{x}_j)$ for $j = 1, 2, \dots, k$ (see Figure 11.4).

Let $\mathbf{u} = f(\mathbf{x}) - f(\mathbf{a})$ and choose by Theorem 11.31 points $\mathbf{c}_j \in L(\mathbf{x}_{j-1}; \mathbf{x}_j)$ such that

$$\mathbf{u} \cdot (f(\mathbf{x}_j) - f(\mathbf{x}_{j-1})) = \mathbf{u} \cdot (Df(\mathbf{c}_j)(\mathbf{x}_j - \mathbf{x}_{j-1})) = 0$$

for $j = 1, 2, \dots, k$. Summing over j and telescoping, we see by the choice of \mathbf{u} that

$$0 = \sum_{j=1}^k \mathbf{u} \cdot (f(\mathbf{x}_j) - f(\mathbf{x}_{j-1})) = \mathbf{u} \cdot (f(\mathbf{x}) - f(\mathbf{a})) = \|f(\mathbf{x}) - f(\mathbf{a})\|^2.$$

Therefore, $f(\mathbf{x}) = f(\mathbf{a})$.

Proof.

Fix $\mathbf{a} \in V$, and let $\mathbf{x} \in V$. Since V is open and connected, V is polygonally connected (see Exercise 10, p.277).

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Therefore, $f(\mathbf{x}) = f(\mathbf{a})$.

Theorem (11.37 Taylor's Formula on \mathbf{R}^n)

Let $p \in \mathbf{N}$, let V be open in \mathbf{R}^n , let $\mathbf{x}, \mathbf{a} \in V$, and suppose that $f : V \rightarrow \mathbf{R}$. If the p th total total differential of f exists on V and $L(\mathbf{x}; \mathbf{a}) \subseteq V$, then there is a point $\mathbf{c} \in L(\mathbf{x}; \mathbf{a})$ such that

$$f(\mathbf{x}) = f(\mathbf{a}) + \sum_{k=1}^{p-1} \frac{1}{k!} D^{(k)} f(\mathbf{a}; \mathbf{h}) + \frac{1}{p!} D^{(p)} f(\mathbf{c}; \mathbf{h})$$

for $\mathbf{h} := \mathbf{x} - \mathbf{a}$.

Note: These hypotheses are met if V is convex and f is C^p on V .

Proof.

Let $\mathbf{h} = \mathbf{x} - \mathbf{a}$. As in the proof Theorem 11.31, choose $\delta > 0$ so small that $\mathbf{a} + t\mathbf{h} \in V$ for $t \in I_\delta := (-\delta, 1 + \delta)$. The function $F(t) = f(\mathbf{a} + t\mathbf{h})$ is differentiable on I_δ and, by the Chain Rule,

$$F'(t) = Df(\mathbf{a} + t\mathbf{h})(\mathbf{h}) = \sum_{k=1}^n \frac{\partial f}{\partial x_k}(\mathbf{a} + t\mathbf{h})h_k.$$

In fact, a simple induction argument can be used to verify

$$F^{(j)}(t) = \sum_{i_1=1}^n \cdots \sum_{i_j=1}^n \frac{\partial^j f}{\partial x_{i_1} \cdots \partial x_{i_j}}(\mathbf{a} + t\mathbf{h})h_{i_1} \cdots h_{i_j}$$

for $j = 1, 2, \dots, p$. Thus



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Let $\mathbf{h} = \mathbf{x} - \mathbf{a}$. As in the proof Theorem 11.31, choose $\delta > 0$ so small that $\mathbf{a} + t\mathbf{h} \in V$ for $t \in I_\delta := (-\delta, 1 + \delta)$. The function $F(t) = f(\mathbf{a} + t\mathbf{h})$ is differentiable on I_δ and, by the Chain Rule,

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for $j = 1, 2, \dots, p$. Thus



Proof.

$$(26) \quad F^{(j)}(0) = D^{(j)}f(\mathbf{a}; \mathbf{h}) \text{ and } F^{(p)}(t) = D^{(p)}f(\mathbf{a} + t\mathbf{h}; \mathbf{h})$$

for $j = 1, \dots, p-1$, and $t \in I_\delta$.

We have proved that $F : I_\delta \rightarrow \mathbf{R}$ has a derivative of order p everywhere on $I_\delta \supset [0, 1]$. Therefore, by the one-dimensional Taylor Formula and (26),

$$\begin{aligned} f(\mathbf{x}) - f(\mathbf{a}) &= F(1) - F(0) = \sum_{j=1}^{p-1} \frac{1}{j!} F^{(j)}(0) + \frac{1}{p!} F^{(p)}(t) \\ &= \sum_{j=1}^{p-1} \frac{1}{j!} D^{(j)}f(\mathbf{a}; \mathbf{h}) + \frac{1}{p!} D^{(p)}f(\mathbf{a} + t\mathbf{h}; \mathbf{h}) \end{aligned}$$

for some $t \in (0, 1)$. Thus set $\mathbf{c} = \mathbf{a} + t\mathbf{h}$.

Proof.

$$(26) \quad F^{(j)}(0) = D^{(j)}f(\mathbf{a}; \mathbf{h}) \text{ and } F^{(p)}(t) = D^{(p)}f(\mathbf{a} + t\mathbf{h}; \mathbf{h})$$

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Thank you.