

Bounded Linear Operators and the Definition of Derivatives

Definition. Let V, W be normed vector spaces (both over \mathbb{R} or over \mathbb{C}). A linear transformation or linear operator $T : V \rightarrow W$ is *bounded* if there is a constant C such that

$$(1) \quad \|Tx\|_W \leq C\|x\|_V \quad \text{for all } x \in V.$$

Remark: We use the linearity of T and the homogeneity of the norm in W to see that

$$\left\| T\left(\frac{x}{\|x\|_V}\right) \right\|_W = \left\| \frac{T(x)}{\|x\|_V} \right\|_W = \frac{\|T(x)\|_W}{\|x\|_V}$$

we see that T is bounded, satisfying (1), if and only if

$$\sup_{\|x\|_V=1} \|T(x)\|_W \leq C.$$

Theorem. Let V, W be normed vector spaces and let $T : V \rightarrow W$ be a linear transformation. The following statements are equivalent.

- (i) T is a bounded linear transformation.
- (ii) T is continuous everywhere in V .
- (iii) T is continuous at 0 in V .

Proof. (i) \implies (ii). Let C as in the definition of bounded linear transformation. By linearity of T we have

$$\|T(v) - T(\tilde{v})\|_W = \|T(v - \tilde{v})\|_W \leq C\|v - \tilde{v}\|_V$$

which implies (ii).

(ii) \implies (iii) is trivial.

(iii) \implies (i): If T is continuous at 0 there exists $\delta > 0$ such that for all $v \in V$ with $\|v\| < \delta$ we have $\|Tv\| < 1$. Now let $x \in V$ and $x \neq 0$. Then

$$\left\| \delta \frac{x}{2\|x\|_V} \right\|_V = \delta/2 \text{ and thus } \left\| T\left(\delta \frac{x}{2\|x\|_V}\right) \right\|_W < 1.$$

But by the linearity of T and the homogeneity of the norm we get

$$1 \geq \left\| T\left(\delta \frac{x}{2\|x\|_V}\right) \right\|_W = \left\| \delta \frac{T(x)}{2\|x\|_V} \right\|_W = \frac{\delta}{2\|x\|_V} \|Tx\|_W$$

and therefore $\|Tx\|_W \leq C\|x\|_V$ with $C = 2/\delta$. □

Notation: If $T : V \rightarrow W$ is linear one often writes Tx for $T(x)$.

Definition. We denote by $L(V, W)$ the set of all bounded linear transformations $T : V \rightarrow W$.

$L(V, W)$ form a vector space. $S + T$ is the transformation with $(S + T)(x) = S(x) + T(x)$ and cT is the operator $x \mapsto cT(x)$. On $L(V, W)$ we define the *operator norm* (depending on the norms on V and W) by

$$\|T\|_{L(V, W)} \equiv \|T\|_{op} = \sup_{v \neq 0} \frac{\|Tv\|_W}{\|v\|_V}.$$

We can think of $\|T\|_{L(V, W)}$ as the best constant for which (1) holds. Note that

$$\|Tx\|_W \leq \|T\|_{L(V, W)} \|x\|_V.$$

Using the homogeneity of the W -norm we also can write

$$\|T\|_{L(V, W)} = \sup_{\|x\|_V=1} \|Tx\|_W.$$

We use the $\|\cdot\|_{op}$ notation if the choice of V , W and the norms are clear from the context. In the textbook, Rudin considers $V = \mathbb{R}^n$, $W = \mathbb{R}^m$ with the standard Euclidean norms and simply writes $\|T\|$ for the operator norm.

Lemma. *Let V and W be normed spaces. If V is finite dimensional then all linear transformations from V to W are bounded.*

Proof. Let v_1, \dots, v_n be a basis of V . Then for $v = \sum_{j=1}^n \alpha_j v_j$ we have

$$\|Tv\|_W = \left\| \sum_{j=1}^n \alpha_j T v_j \right\|_W \leq \sum_{j=1}^n |\alpha_j| \|T v_j\|_W \leq \sum_{j=1}^n \|T v_j\|_W \max_{k=1, \dots, n} |\alpha_k|.$$

The expression $\max_{k=1, \dots, n} |\alpha_k|$ defines a norm on V . Since all norms on V are equivalent, there is a constant C_1 such that

$$\max_{j=1, \dots, n} |\alpha_j| \leq C \left\| \sum_{j=1}^n \alpha_j v_j \right\|_V$$

for all choices of $\alpha_1, \dots, \alpha_n$. Thus we get $\|Tv\|_W \leq C \|v\|_V$ for all $v \in V$, where the constant C is given by $C = C_1 \sum_{j=1}^n \|T v_j\|_W$. \square

Lemma. *On \mathbb{R}^n , \mathbb{R}^m use the Euclidean norms $\|x\|_2 = (\sum_{j=1}^n |x_j|^2)^{1/2}$, $\|y\|_2 = (\sum_{i=1}^m |y_i|^2)^{1/2}$. Let A be an $m \times n$ matrix and consider the linear operator $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ defined by $T(x) = Ax$. Let*

$$\|A\|_{HS} := \left(\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2 \right)^{1/2}.$$

Then

$$\|T\|_{op} \leq \|A\|_{HS}.$$

Proof. $\|Ax\|_2^2 = \sum_{i=1}^m \left| \sum_{j=1}^n a_{ij} x_j \right|^2$. By the Cauchy-Schwarz inequality $\left| \sum_{j=1}^n a_{ij} x_j \right|^2 \leq \sum_{j=1}^n |a_{ij}|^2 \|x\|_2^2$ and hence $\|Ax\|_2 \leq \|A\|_{HS} \|x\|_2$ for all $x \in \mathbb{R}^n$. Thus $\|T\|_{op} \leq \|A\|_{HS}$. \square

Remark. We often write A for both the matrix and the linear operator $x \rightarrow Ax$. The expression $\|A\|_{HS}$ is a norm on the space of $m \times n$ matrices called the Hilbert-Schmidt norm of A . In many cases it is substantially larger than the operator norm (and so the estimate in the Lemma is rather inefficient). To illustrate this let D be an $n \times n$ diagonal matrix (with entries $\lambda_1, \dots, \lambda_n$ on the diagonal and zeroes elsewhere). Then it is easy to verify the inequality

$$\|Dx\|_2 \leq \max_{j=1, \dots, n} |\lambda_j| \|x\|_2$$

and this is seen to be sharp by testing on the unit vectors $x = e_j$. Thus $\|D\|_{op} = \max_{j=1, \dots, n} |\lambda_j|$ while $\|D\|_{HS} = (\sum_{j=1}^n |\lambda_j|^2)^{1/2}$. In particular for the identity operator I we have $\|I\|_{op} = 1$ but $\|I\|_{HS} = \sqrt{n}$.

Normed vector spaces and completeness. Let V be a normed space. Then V is a metric space with respect to the metric $d(x, y) = \|x - y\|$. Recall that a sequence $\{x_n\}_{n=1}^\infty$ is a Cauchy-sequence in V if for every $\varepsilon > 0$ there is K such that $\|x_n - x_m\|_V < \varepsilon$ for $m, n \geq K$.

Definition. A *Banach space* is a normed space which is complete (i.e. every Cauchy sequence converges).

Lemma: A finite dimensional normed space over \mathbb{R} or \mathbb{C} is complete.

Proof. Let v_1, \dots, v_N be a basis of V and for $x \in V$ let $\beta_1(x), \dots, \beta_N(x)$ be the coordinates of x with respect to the basis v_1, \dots, v_N .

We need to show that $\{x_n\}$ has a limit in V . For every n we can write

$$x_n = \sum_{i=1}^N \beta_i(x_n) v_i.$$

The expression

$$\|x\|_\infty := \max_{i=1, \dots, N} |\beta_i(x)|$$

defines a norm on V which is equivalent to the given norm. In particular we have

$$|\beta_i(x_n) - \beta_i(x_m)| \leq C \|x_n - x_m\|_V, \quad i = 1, \dots, N.$$

This shows that for each $i = 1, \dots, N$ the numbers $\beta_i(x_n)$ form a Cauchy-sequence of scalars and thus converge to a scalar β_i . Define $x = \sum_{i=1}^N \beta_i v_i$ (so that $\beta_i = \beta_i(x)$). Then

$$\|x_n - x\|_V = \left\| \sum_{i=1}^N (\beta_i(x_n) - \beta_i) v_i \right\|_V \leq \sum_{i=1}^N |\beta_i(x_n) - \beta_i| \|v_i\|_V$$

which converges to 0 as $n \rightarrow \infty$. Hence $x_n \rightarrow x$ in V . □

Theorem: Let V and W be normed spaces and assume that W is complete. Then the space $L(V, W)$ of bounded linear operators from V to W is a Banach space.

Proof. We need to show that $L(V, W)$ is complete. Let T_n be a Cauchy sequence in $L(V, W)$ (with respect to the norm $\|T\|_{op} = \sup_{\|x\|_V=1} \|Tx\|_W$). Then for each $x \in V$, $T_n x$ is Cauchy in W and by assumption $T_n x$ converge to some vector Tx . Check, using limiting arguments, that $T : V \rightarrow W$ is linear. Since T_n is Cauchy the sequence T_n is bounded in $L(V, W)$ and thus there exists M with $\|T_n\|_{op} \leq M$.

We need to show that T is a bounded operator. We have for all n and all $x \in V$

$$\|Tx\|_W \leq \|Tx - T_n x\|_W + \|T_n x\|_W \leq \|Tx - T_n x\|_W + \|T_n\|_{op} \|x\|_V.$$

Letting $n \rightarrow \infty$ we see that $\|Tx\|_W \leq M\|x\|_V$. Hence $T \in L(V, W)$ with $\|T\|_{op} \leq M$.

Finally we need to show that $T_n \rightarrow T$ in $L(V, W)$. Given $\varepsilon > 0$ we choose N such that $\|T_n - T_m\|_{op} < \varepsilon/2$ for $m, n \geq N$.

We prove the following *Claim*:

For each $x \in V$, $x \neq 0$, and $n \geq N$ we have $\|T_n x - Tx\|_W < \varepsilon \|x\|_V$.

Let $n \geq N$. Since $T_m x \rightarrow Tx$ in W we can choose $m(x) \geq N$ such that $\|T_{m(x)} x - Tx\|_W \leq \varepsilon/4 \|x\|_V$. We estimate we estimate (with $m \geq N$)

$$\begin{aligned} \|Tx - T_n x\|_W &\leq \|Tx - T_{m(x)} x + T_{m(x)} x - T_n x\|_W \\ &\leq \|Tx - T_{m(x)} x\|_W + \|T_{m(x)} x - T_n x\|_W \\ &\leq \frac{\varepsilon}{4} \|x\|_V + \|T_{m(x)} - T_n\|_{op} \|x\|_V \\ &\leq \frac{\varepsilon}{4} \|x\|_V + \frac{\varepsilon}{2} \|x\|_V < \varepsilon \|x\|_V. \end{aligned}$$

Hence the claim is established and thus $\|T_n - T\|_{op} < \varepsilon$ for $n \geq N$. This shows $T_n \rightarrow T$ with respect to the operator norm. \square

Definition. (i) Let V be a normed vector space over \mathbb{R} . Bounded linear transformations from V to \mathbb{R} are called bounded linear functionals. The space $L(V, \mathbb{R})$ with the operator norm is called the dual space to V , or V^* .

(ii) Similarly if V be a normed vector space over \mathbb{C} we call the bounded linear transformations from V to \mathbb{C} bounded linear functionals and refer to the space $L(V, \mathbb{C})$ with the operator norm as the dual space V^* .

Since \mathbb{R} and \mathbb{C} are complete the above theorem about completeness of $L(V, W)$ immediately yields the

Corollary. Dual spaces of normed vector spaces are Banach spaces.

Exercise: The space $L(\mathbb{R}^n, \mathbb{R})$ of linear functionals $\ell : \mathbb{R}^n \rightarrow \mathbb{R}$ can be identified with \mathbb{R} , since for any such ℓ there is a unique vector $y \in \mathbb{R}^n$ such that $\ell(x) = \sum_{j=1}^n x_j y_j$ for all $x \in \mathbb{R}^n$. We consider the p -norm on \mathbb{R}^n , $\|x\|_p = (\sum_{j=1}^n |x_j|^p)^{1/p}$, $1 \leq p < \infty$. Also set $\|x\|_\infty = \max_{j=1, \dots, n} |x_j|$. The norm on \mathbb{R} is just the absolute value $|\cdot|$.

(i) Prove

$$\sup_{\|x\|_1 \neq 0} \frac{\left| \sum_{j=1}^n x_j y_j \right|}{\|x\|_1} = \|y\|_\infty$$

and

$$\sup_{\|x\|_\infty \neq 0} \frac{\left| \sum_{j=1}^n x_j y_j \right|}{\|x\|_\infty} = \|y\|_1.$$

(ii) Prove that if $1 < p < \infty$ then

$$\|\ell\|_{op} \equiv \sup_{\|x\|_p \neq 0} \frac{\left| \sum_{j=1}^n x_j y_j \right|}{\|x\|_p} = \|y\|_{p'} \text{ where } p' = \frac{p}{p-1}.$$

Hint: Hölder's inequality should be used here.

Finally we consider compositions of linear transformations. The operator norms are submultiplicative in the sense of the following lemma.

Lemma. Let V_1, V_2, V_3 be normed vector spaces and let $T \in L(V_1, V_2)$ and $S \in L(V_2, V_3)$. Define the composition $ST : V_1 \rightarrow V_3$ by $ST(x) = S(T(x))$. Then $ST \in L(V_1, V_3)$ and we have

$$\|ST\|_{L(V_1, V_3)} \leq \|S\|_{L(V_2, V_3)} \|T\|_{L(V_1, V_2)}.$$

Proof: For $x \in V_1$,

$$\|ST(x)\|_{V_3} \leq \|S\|_{L(V_2, V_3)} \|T(x)\|_{V_2} \leq \|S\|_{L(V_2, V_3)} \|T\|_{L(V_1, V_2)} \|x\|_{V_1}.$$

This implies $ST \in L(V_1, V_3)$ and the asserted submultiplicativity property. \square

Differentiation

Definition. Let V and W be normed vector spaces. Let $U \subset V$ be an open subset of V and let $F : U \rightarrow W$ be a function. Let $a \in U$. We say that F is differentiable at a if there exists a bounded linear transformation $T : V \rightarrow W$ (usually denoted by Df_a or by $f'(a)$) such that

$$\lim_{h \rightarrow 0} \frac{\|f(a+h) - f(a) - Th\|_W}{\|h\|_V} = 0$$

Uniqueness: For the derivative of f at a to be well defined we must prove uniqueness of T . Suppose there are two linear bounded transformations T and \tilde{T} with $\frac{\|f(a+h) - f(a) - Th\|_W}{\|h\|_V} \rightarrow 0$ and $\frac{\|f(a+h) - f(a) - \tilde{T}h\|_W}{\|h\|_V} \rightarrow 0$ as $\|h\|_V \rightarrow 0$.

0. The triangle inequality implies that $\lim_{h \rightarrow 0} \frac{\|Th - \tilde{T}h\|_W}{\|h\|_V} = 0$. I.e. for every $\varepsilon > 0$ there is a $\delta > 0$ such that $\|Th - \tilde{T}h\|_W \leq \varepsilon \|h\|_V$ for all $\|h\| \leq \delta$. However since T, \tilde{T} are linear this inequality also holds with h replaced by ch for all scalars c , and thus we have

$$\|Th - \tilde{T}h\|_W \leq \varepsilon \|h\|_V \text{ for all } h \in V.$$

Hence $\|T - \tilde{T}\|_{op} \leq \varepsilon$ for all $\varepsilon > 0$ which implies $T = \tilde{T}$. Thus if the derivative of f at a exists it is uniquely defined and we shall henceforth denote it by Df_a or by $f'(a)$.

Remark: Other terms for this derivative Df_a are “total derivative of f at a ” or “Fréchet derivative of f at a ”.

Exercise: Prove that if f is differentiable at a then f is continuous at a .

Example: Let $M(n, n)$ be the space of $n \times n$ matrices, with any norm. Define $F(A) = A^2$. Then F is differentiable at any A and its derivative DF_A is given by

$$DF_A(H) = AH + HA.$$

For the proof observe that $F(A + H) - F(A) = (A + H)^2 - A^2 = AH + HA + H^2$ and check that $\lim_{H \rightarrow 0} \|H^2\|/\|H\| = 0$. This is accomplished by showing that $\|H^2\| \leq C\|H\|_2$ and if one uses a submultiplicative norm on $M(n, n)$ one gets this even with the constant 1.

Example: Let $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by $F(x_1, x_2) = x_1^2 + \sin(\pi x_2)$. You are being asked to check from the definition whether F differentiable at $(2, 1)$ and to determine the derivative $DF_{(2,1)}$, as a linear transformation $\mathbb{R}^2 \rightarrow \mathbb{R}$. Observe that

$$\begin{aligned} F(2 + h_1, 1 + h_2) - F(2, 1) &= (2 + h_1)^2 - 2^2 + (\sin(\pi(1 + h_2)) - \sin \pi) \\ &= 4h_1 + \pi \cos(\pi)h_2 + O(h_1^2) + O(h_2^2). \end{aligned}$$

Hence $DF_{(2,1)} : \mathbb{R}^2 \rightarrow \mathbb{R}$ is the linear transformation given by

$$DF_{(2,1)}(h) = 4h_1 - \pi h_2.$$

Example: Let $C([0, 1])$ be the space of continuous functions with the usual max-norm $\|f\|_\infty = \max_{x \in [0, 1]} |f(x)|$. We define a function

$$\Gamma : C([0, 1]) \rightarrow C([0, 1])$$

by saying that for $f \in C([0, 1])$, $\Gamma[f]$ is the function whose values at $x \in [0, 1]$ are given by

$$\Gamma[f](x) = \int_0^x f(t)^2 \cos t \, dt.$$

Show that Γ is differentiable at every $g \in C([0, 1])$ and determine $D\Gamma_g$ for all $g \in C([0, 1])$.

For $h \in C([0, 1])$ we compute

$$\begin{aligned}\Gamma[g + h](x) - \Gamma[g](x) &= \int_0^x (g(t) + h(t))^2 \cos t \, dt - \int_0^x (g(t))^2 \cos t \, dt \\ &= \int_0^x h(t) 2g(t) \cos t \, dt + \int_0^x (h(t))^2 \cos t \, dt\end{aligned}$$

We claim that $D\Gamma_g : C([0, 1]) \rightarrow C([0, 1])$ is the linear transformation which to each $h \in C([0, 1])$ assigns the function $T[h]$ whose values at x are given by

$$T[h](x) = \int_0^x h(t) 2g(t) \cos t \, dt.$$

We must show that T is a bounded linear transformation and that it is really the derivative of Γ at g . Clearly T is linear and maps $C([0, 1])$ to itself.

It is bounded since

$$\max_{0 \leq x \leq 1} \left| \int_0^x h(t) 2g(t) \cos t \, dt \right| \leq 2 \max_{[0, 1]} |g(x)| \max_{x \in [0, 1]} |h(x)|;$$

this inequality shows that the operator norm of T is at most $2\|g\|_\infty$.

To verify that T is really the derivative of Γ at g we must analyze the error term and show that

$$\frac{\max_{x \in [0, 1]} \left| \int_0^x h(t)^2 \cos t \, dt \right|}{\|h\|_\infty} \rightarrow 0 \text{ as } \|h\|_\infty \rightarrow 0.$$

The numerator is bounded by $\max_{t \in [0, 1]} |h(t)^2| = \|h\|_\infty^2$ and hence the displayed expression tends to 0 as $\|h\|_\infty \rightarrow 0$. Thus we have shown that Γ is differentiable at g and $D\Gamma_g = T$.

Example: Let $C([0, 1])$ be the space of continuous functions with the usual max-norm $\|f\|_\infty = \max_{x \in [0, 1]} |f(x)|$. Define $\Lambda : C([0, 1]) \rightarrow \mathbb{R}$ by

$$\Lambda[f] = \int_0^1 f(t)^2 \cos t \, dt.$$

Show that Λ is differentiable at every $g \in C([0, 1])$ and that

$$D\Lambda_g : C([0, 1]) \rightarrow \mathbb{R}$$

is the bounded linear functional defined by

$$D\Lambda_g[h] = \int_0^1 h(t) 2g(t) \cos t \, dt.$$