

# Orthogonality and Proportional Norms

By

**P. Schöpf**

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## Abstract

Two norms on a real vectorspace define the same orthogonality relation iff they are proportional. The aim of this note is to give a proof of this statement with a minimum of results on convex sets, convex functions and real analysis. Needed is only the right derivative of a convex function and the théorème des accroissements finis as it is called by H. Cartan.

G. D. Birkhoff, R. C. James and others (see [1]) used in several important papers the following concept of orthogonality in real normed linear spaces  $(X, \|\cdot\|)$ .

**Definition 1.** *Let  $x, y \in X$ . We say  $x \perp y$  ( $x$  is orthogonal to  $y$ ) iff  $\|x + \lambda y\| \geq \|x\|$  for all  $\lambda \in \mathbb{R}$ .*

Let us now think of two norms  $\|\cdot\|_1, \|\cdot\|_2$  on  $X$ , then we can ask, when they determine the same orthogonality relation on  $X$ . If one analyzes the geometrical meaning of this orthogonality relation, then it seems that the following has to be true.

**Theorem.** *Two norms  $\|\cdot\|_1, \|\cdot\|_2$  on  $X$  determine the same orthogonality relation (i.e.  $\perp_1 = \perp_2$ ) iff they are proportional (i.e. There exists a number  $\sigma \in \mathbb{R}_{>0}$  with  $\|x\|_1 = \sigma \|x\|_2$  for all  $x \in X$ ).*

This theorem for example is useful in the proof of Theorem 4.17 in [2] as Prof. R. Ger pointed out. Every proof of this theorem will use several basic facts on convex sets and convex functions. Our aim is to give

a proof relying on a minimum of these facts and therefore we will use only, that a convex function has a right derivative at every inner point of its domain of definition. For this purpose let us fix some notations.

**Definition 2.** Let  $(X, \|\cdot\|)$  be a real normed linear space and  $x, y \in X$  linearly independent. With  $x, y$  we always can define the following functions

$$g(x, y; \bullet): \mathbb{R} \rightarrow \mathbb{R}, \quad g(x, y; \mu) := \|x + \mu y\| \quad \text{for all } \mu \in \mathbb{R},$$

$$c(x, y; \bullet): \mathbb{R} \rightarrow X, \quad c(x, y; \mu) := \frac{1}{g(x, y; \mu)}(x + \mu y) \quad \text{for all } \mu \in \mathbb{R}.$$

$g^+(x, y; \bullet), c^+(x, y; \bullet)$  are the right derivatives of these functions.

**Lemma 1.** Let  $(X, \|\cdot\|)$  be any real normed linear space,  $x, y \in X$  linearly independent, then the following statements are true.

1.  $c^+(x, y; 0) = -g^+(x, y; 0) \frac{x}{\|x\|^2} + \frac{y}{\|x\|}$
2.  $c^+(x, y; 0)$  and  $c(x, y; 0)$  are linearly independent.
3. If  $x \perp y$ , then  $-g^+(x, y; 0) \leq 0$ .
4. If  $\tilde{y} := \eta(y + \kappa x)$  with  $\kappa, \eta \in \mathbb{R}, \eta > 0$ , then  
 $c^+(x, \tilde{y}; 0) = \eta c^+(x, y; 0)$
5.  $c(x, y - x; \lambda + \mu) = c(x + \lambda(y - x), y - x; \mu)$  and  
 $c^+(x, y - x; \lambda) = c^+(x + \lambda(y - x), y - x; 0)$

*Proof:*

Ad 1.  $g^+(x, y; \mu)$  exists for every  $\mu \in \mathbb{R}$  because  $g(x, y; \bullet): \mathbb{R} \rightarrow \mathbb{R}$  is a convex function. Differentiation therefore yields

$$c^+(x, y; \mu) = \frac{-g^+(x, y; \mu)}{g^2(x, y; \mu)}(x + \mu y) + \frac{y}{g(x, y; \mu)}$$

and with  $\mu = 0$  we get

$$c^+(x, y; 0) = \frac{-g^+(x, y; 0)}{g^2(x, y; 0)}x + \frac{y}{g(x, y; 0)} = -g^+(x, y; 0) \frac{x}{\|x\|^2} + \frac{y}{\|x\|}.$$

Ad 2. 1. shows that  $c^+(x, y; 0), x$  are linearly independent, because  $x, y$  are linearly independent. But  $x = \|x\| c(x, y; 0)$ , hence  $c^+(x, y; 0), c(x, y; 0)$  are linearly independent.

Ad 3.  $x \perp y$  is defined by  $\|x + \mu y\| \geq \|x\|$  for all  $\mu \in \mathbb{R}$ , but this implies that  $\mu = 0$  is an argument where the absolute minimum  $\|x\|$  of  $g(x, y; \bullet)$  is attained and therefore we must have  $-g^+(x, y; 0) \leq 0$ .

Ad 4. One easily can check that

$$c(x_2, y; \mu) = c\left(x_2, \tilde{y}; \frac{\mu}{\eta(1 - \kappa\mu)}\right)$$

for all  $\mu \in \mathbb{R}$  with  $\kappa\mu < 1$ . Differentiation yields

$$c^+(x_2, y; \mu) = c^+\left(x_2, \tilde{y}; \frac{\mu}{\eta(1 - \kappa\mu)}\right) \frac{1}{\eta(1 - \kappa\mu)^2}$$

and for  $\mu = 0$  we get the desired equation. ■

Ad 5. A trivial computation.

**Lemma 2.**  $x \perp c^+(x_2, y; 0)$  for every pair of linearly independent vectors  $x_2, y \in X$ .

*Proof:*

We only have to show that  $\|x + \lambda c^+(x_2, y; 0)\| \geq \|x\|$  for every  $\lambda \in \mathbb{R}$ . For shorter notation we will write  $c(\delta) := c(x_2, y; \delta)$ . Let  $\delta > 0$  and  $\mu \notin [0, 1]$ , we then get

$$\|(1 - \mu)c(\delta) + \mu c(0)\| \geq (1 - \mu)\|c(\delta)\| + \mu\|c(0)\| = 1 = \|c(0)\|$$

or equivalently

$$\left\|c(\delta) + (-\mu)\delta \left(\frac{c(\delta) - c(0)}{\delta}\right)\right\| \geq \|c(0)\|.$$

The last inequality says that for all  $\lambda \notin [-\delta, 0]$

$$\left\|c(\delta) + \lambda \left(\frac{c(\delta) - c(0)}{\delta}\right)\right\| \geq \|c(0)\|.$$

If we choose an  $\varepsilon > 0$ , then for every  $0 < \delta < \varepsilon$  and every  $\lambda \notin [-\varepsilon, 0]$  we get

$$\left\|c(\delta) + \lambda \left(\frac{c(\delta) - c(0)}{\delta}\right)\right\| \geq \|c(0)\|.$$

Taking the limit  $\delta \rightarrow 0$  yields

$$\|c(0) + \lambda c^+(0)\| \geq \|c(0)\|$$

for every  $\lambda \notin [-\varepsilon, 0]$ . But  $\varepsilon$  was arbitrary and therefore

$$\|x + \lambda c^+(0)\| \geq \|x\|$$

for all  $\lambda \in \mathbb{R}$ . ■

**Lemma 3.** *Let  $\|\cdot\|_1, \|\cdot\|_2$  be two norms on  $X$  which determine the same orthogonality relation  $\perp$  on  $X$  (i.e.  $\perp_1 = \perp_2 =: \perp$ ). Let  $g_i(x, y; \bullet), c_i(x, y; \bullet)$  be the functions with respect to  $\|\cdot\|_i, i = 1, 2$ , then*

$$c_2^+(x, y; 0) = \frac{\|x\|_1}{\|x\|_2} c_1^+(x, y; 0).$$

*Proof:*

By Lemma 1 we have

$$c_1^+(x, y; 0) = -g_1^+(x, y; 0) \frac{x}{\|x\|_1^2} + \frac{y}{\|x\|_1} \quad \text{and}$$

$$c_2^+(x, y; 0) = -g_2^+(x, y; 0) \frac{x}{\|x\|_2^2} + \frac{y}{\|x\|_2}$$

Substituting

$$\tilde{y} := c_1^+(x, y; 0) = \frac{1}{\|x\|_1} \left( y + \frac{-g_1^+(x, y; 0)}{\|x\|_1} x \right)$$

in these two equations yields (according to Lemma 1.4)

$$c_1^+(x, \tilde{y}; 0) = \frac{\tilde{y}}{\|x\|_1} \quad \text{and} \quad c_2^+(x, \tilde{y}; 0) = \frac{1}{\|x\|_1} c_2^+(x, y; 0)$$

From this we get (by Lemma 1.1)

$$\begin{aligned} \frac{1}{\|x\|_1} c_2^+(x, y; 0) &= c_2^+(x, \tilde{y}; 0) = -g_2^+(x, \tilde{y}; 0) \frac{x}{\|x\|_2^2} + \frac{\tilde{y}}{\|x\|_2} \\ &= -g_2^+(x, \tilde{y}; 0) \frac{x}{\|x\|_2^2} + \frac{1}{\|x\|_2} c_1^+(x, y; 0). \end{aligned}$$

By Lemma 2 we have  $x \perp \tilde{y}$  and Lemma 1.3 yields therefore

$$-g_2^+(x, \tilde{y}; 0) \leq 0.$$

If we substitute  $\tilde{y} := c_2^+(x, y; 0)$  in our starting equations, we get the analogous equation

$$\frac{1}{\|x\|_2} c_1^+(x, y; 0) = -g_1^+(x, \tilde{y}; 0) \frac{x}{\|x\|_1^2} + \frac{1}{\|x\|_1} c_2^+(x, y; 0),$$

with the analogous statement that

$$-g_1^+(x, \tilde{y}; 0) \leq 0.$$

Adding these last two equations yields

$$-g_2^+(x, \tilde{y}; 0) \frac{x}{\|x\|_2^2} + -g_1^+(x, \tilde{y}; 0) \frac{x}{\|x\|_1^2} = 0,$$

which implies that both coefficients are zero, because they are both  $\leq 0$ . Now we are ready, because of this we have

$$\frac{1}{\|x\|_1} c_2^+(x, y; 0) = \frac{1}{\|x\|_2} c_1^+(x, y; 0),$$

what we wanted. ■

*Proof of the theorem:*

Let  $\|\cdot\|_1, \|\cdot\|_2$  be two norms on  $X$ , which define the same orthogonality relation  $\perp$  on  $X$  and  $x, y \in X$  linearly independent, then we will show that

$$\frac{\|y\|_1}{\|y\|_2} = \frac{\|x\|_1}{\|x\|_2}.$$

Our two curves are related in the following form

$$c_2(x, y - x; \lambda) = \frac{g_1(x, y - x; \lambda)}{g_2(x, y - x; \lambda)} c_1(x, y - x; \lambda)$$

and therefore we get by differentiation

$$\begin{aligned} c_2^+(x, y - x; \lambda) &= \left( \frac{g_1}{g_2} \right)^+ (x, y - x; \lambda) c_1(x, y - x; \lambda) \\ &\quad + \left( \frac{g_1}{g_2} \right) (x, y - x; \lambda) c_1^+(x, y - x; \lambda). \end{aligned}$$

Lemma 1.5, Lemma 3 and again Lemma 1.5 yield

$$\begin{aligned} c_2^+(x, y - x; \lambda) &= c_2^+(x + \lambda(y - x), y - x; 0) \\ &= \frac{\|x + \lambda(y - x)\|_1}{\|x + \lambda(y - x)\|_2} c_1^+(x + \lambda(y - x), y - x; 0) \\ &= \frac{\|x + \lambda(y - x)\|_1}{\|x + \lambda(y - x)\|_2} c_1^+(x, y - x; \lambda) \end{aligned}$$

i.e.  $c_2^+(x, y - x; \lambda), c_1^+(x, y - x; \lambda)$  are linearly dependent. On the other side we have according to Lemma 1.2, that  $c_1(x, y - x; \lambda), c_1^+(x, y - x; \lambda)$

are linearly independent and hence we conclude, that

$$\left( \begin{array}{c} g_1 \\ g_2 \end{array} \right)^+ (x, y - x; \lambda) = 0$$

for all  $\lambda \in \mathbb{R}$ . From this we get (by the théorème des accroissements finis [4])

$$\left| \frac{g_1(x, y - x; 1)}{g_2(x, y - x; 1)} - \frac{g_1(x, y - x; 0)}{g_2(x, y - x; 0)} \right| \leq 0$$

or, what is the same

$$\frac{\|y\|_1}{\|y\|_2} = \frac{\|x\|_1}{\|x\|_2}.$$

This implies the existence of a number  $\sigma \in \mathbb{R}$ , such that for all  $x \in X$

$$\|x\|_1 = \sigma \|x\|_2.$$

The implication from proportionality of the two norms on equality of there associated orthogonality relations is trivial. ■

## References

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**Author's address:** P. Schöpf, Institut für Mathematik, Universität Graz, Heinrichstraße 36/III, A-8010 Graz, Austria.