

# Convex sets

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## Abstract

In these notes I review some facts about convex sets. First I treat when convex combinations of elements are unique. Then I review the Carathéodory theorem. Next I consider extreme elements and minimal sets that generate the convex set. I finish with convex isomorphisms. I assume that the concepts of convex sets and convex hulls are familiar.

## Contents

1	Unique convex combinations	1
2	Theorem of Carathéodory (1907)	3
3	Minimal sets and extreme elements	4
4	Convex isomorphisms	6

## 1 Unique convex combinations

**Definition 1.** Let  $C$  be the convex hull of  $v_1, \dots, v_m$ . An element  $v \in C$  has a unique convex combination of elements  $v_1, \dots, v_m$  when  $\lambda_1, \dots, \lambda_m, \mu_1, \dots, \mu_k \geq 0$ ,  $\sum_{i=1}^k \mu_i = \sum_{i=1}^m \lambda_i = 1$  and  $v = \lambda_1 v_1 + \dots + \lambda_m v_m = \mu_1 v_1 + \dots + \mu_k v_k$  implies  $\lambda_i = \mu_i$ , for all  $i = 1, \dots, m$ .

**Lemma 2.** Let  $V$  be a real vector space. Let  $k \in \mathbb{N}$ . Let  $v_1, \dots, v_{k+1} \in V$  and let  $C$  be the convex hull of  $v_1, \dots, v_{k+1}$ . Then each element of  $C$  has a unique convex combination of elements of  $v_1, \dots, v_{k+1}$  if and only if  $v_1 - v_{k+1}, \dots, v_k - v_{k+1}$  are linearly independent.

*Proof.* First we prove that when  $v_1 - v_{k+1}, \dots, v_k - v_{k+1}$  are linearly independent, that each element of  $C$  has a unique convex combination of elements  $v_1, \dots, v_{k+1}$ .

Let  $v \in C$  and let  $v = \lambda_1 v_1 + \dots + \lambda_{k+1} v_{k+1} = \mu_1 v_1 + \dots + \mu_{k+1} v_{k+1}$  be convex combinations of  $v$ . Then  $v - v_{k+1} = \lambda_1(v_1 - v_{k+1}) + \dots + \lambda_k(v_k - v_{k+1}) = \mu_1(v_1 - v_{k+1}) + \dots + \mu_k(v_k - v_{k+1})$ , so

$$\lambda_1(v_1 - v_{k+1}) + \dots + \lambda_k(v_k - v_{k+1}) = \mu_1(v_1 - v_{k+1}) + \dots + \mu_k(v_k - v_{k+1}).$$

It follows from the fact that  $v_1 - v_{k+1}, \dots, v_k - v_{k+1}$  are linearly independent, that  $\lambda_i = \mu_i$  for all  $i = 1, \dots, k$ . Finally,  $\lambda_{k+1} = 1 - \lambda_1 - \dots - \lambda_k = 1 - \mu_1 - \dots - \mu_k = \mu_{k+1}$ . So  $v$  has a unique convex combination.

For the proof in the other direction, suppose  $v_1 - v_{k+1}, \dots, v_k - v_{k+1}$  are not linearly independent. We will show, that there is an element in the convex hull that does not have a unique convex combination.

From the linear dependence of  $v_1 - v_{k+1}, \dots, v_k - v_{k+1}$  follows that there are  $\alpha_1, \dots, \alpha_k$ , not all zero, so that  $\alpha_1(v_1 - v_{k+1}) + \dots + \alpha_k(v_k - v_{k+1}) = 0$ . Let  $I = \{i : \alpha_i > 0\}$  and  $J = \{i : \alpha_i \leq 0\}$ . So

$$\sum_{i \in I} \alpha_i(v_i - v_{k+1}) = \sum_{i \in J} -\alpha_i(v_i - v_{k+1}).$$

As at least one  $\alpha_i \neq 0$ ,  $i \in \{1, \dots, k\}$ , at least one of  $\sum_{i \in I} \alpha_i$  or  $\sum_{i \in J} -\alpha_i$  is positive, and both are non-negative. Let  $M = \max\{\sum_{i \in I} \alpha_i, \sum_{i \in J} -\alpha_i\} > 0$ . Let  $\beta = M - \sum_{i \in I} \alpha_i$  and  $\gamma = M - \sum_{i \in J} -\alpha_i$ . Note that  $\beta, \gamma \geq 0$ , and that  $\beta + \sum_{i \in I} \alpha_i = \gamma + \sum_{i \in J} -\alpha_i = M$ . As  $v_{k+1} - v_{k+1} = 0$ , we have

$$\frac{\beta}{M}(v_{k+1} - v_{k+1}) + \sum_{i \in I} \frac{\alpha_i}{M}(v_i - v_{k+1}) = \frac{\gamma}{M}(v_{k+1} - v_{k+1}) + \sum_{i \in J} \frac{-\alpha_i}{M}(v_i - v_{k+1}).$$

Using that  $\frac{\beta}{M} + \sum_{i \in I} \frac{\alpha_i}{M} = \frac{\gamma}{M} + \sum_{i \in J} \frac{-\alpha_i}{M} = 1$ , adding  $v_{k+1}$  on both sides gives

$$\frac{\beta}{M}v_{k+1} + \sum_{i \in I} \frac{\alpha_i}{M}v_i = \frac{\gamma}{M}v_{k+1} + \sum_{i \in J} \frac{-\alpha_i}{M}v_i.$$

As  $I$  and  $J$  are disjoint, and at least one of  $\alpha_i \neq 0$ , it follows that this are two different convex combinations of  $\frac{\beta}{M}v_{k+1} + \sum_{i \in I} \frac{\alpha_i}{M}v_i$ .  $\square$

**Lemma 3.** *Let  $V$  be a vector space, and  $C$  the convex hull of  $v_1, \dots, v_m \in V$ . When  $v \in C$  has two different convex combinations of  $v_1, \dots, v_m$ , then  $v$  has infinitely many convex combinations of  $v_1, \dots, v_m$ .*

*Proof.* Suppose

$$v = \sum_{i=1}^m \lambda_i v_i = \sum_{i=1}^m \mu_i v_i$$

are two different convex combinations of  $v$ . So for some  $i_0 \in \{1, \dots, m\}$ ,  $\lambda_{i_0} \neq \mu_{i_0}$ . Let  $\alpha \in [0, 1]$ . Note that

$$v = \sum_{i=1}^m (\alpha \lambda_i + (1 - \alpha) \mu_i) v_i =: \sum_{i=1}^m \nu_i(\alpha) v_i,$$

is also a convex combination of  $v$ . When  $\alpha_1 \neq \alpha_2$ ,  $\nu_{i_0}(\alpha_1) - \nu_{i_0}(\alpha_2) = (\alpha_1 - \alpha_2)(\lambda_{i_0} - \mu_{i_0}) \neq 0$ . Hence there are infinitely many convex combinations of  $v$ .  $\square$

**Definition 4.** Let  $C$  be a convex set. A convex combination

$$v = \sum_{i=1}^m \lambda_i v_i$$

is open when for all  $i \in \{1, \dots, m\}$ ,  $\lambda_i > 0$ .

**Definition 5.** Let  $V$  be a real vector space and let  $v_1, \dots, v_m \in V$ . We define the open convex set generated by  $v_1, \dots, v_m$  to be the set of all open convex combinations of  $v_1, \dots, v_m$ .

**Lemma 6.** *Let  $V$  be a real vector space and let  $C^\circ$  be the open convex set generated by  $v_1, \dots, v_m \in V$ . Let  $C$  be the convex set generated by  $v_1, \dots, v_m$ . Then  $C^\circ$  is convex and  $\emptyset \neq C^\circ \subseteq C$ .*

*Proof.* It is obvious that  $C^\circ$  is contained in the closed convex set generated by  $v_1, \dots, v_m$ . We have that  $(1/m)v_1 + \dots + (1/m)v_m \in C^\circ$ , so  $C^\circ$  is not empty.

Let  $v = \sum_{i=1}^m \lambda_i v_i, w = \sum_{i=1}^m \mu_i v_i \in C^\circ$ ,  $\lambda_i, \mu_i > 0$  for all  $i$ . Let  $\alpha \in [0, 1]$ . Then

$$\alpha v + (1 - \alpha)w = \sum_{i=1}^m (\alpha \lambda_i + (1 - \alpha)\mu_i)v_i.$$

Note that  $\sum_{i=1}^m (\alpha \lambda_i + (1 - \alpha)\mu_i) = 1$ , and  $\alpha \lambda_i + (1 - \alpha)\mu_i > 0$ , for all  $i \in \{1, \dots, m\}$ . Hence  $\alpha v + (1 - \alpha)w \in C^\circ$ . So  $C^\circ$  is convex.  $\square$

**Lemma 7.** *Let  $V$  be a vector space and let  $C^\circ$  (resp.  $C$ ) be the open (resp. closed) convex set generated by  $v_1, \dots, v_m \in V$ . The following statements are equivalent:*

- (i) *There is an element  $v \in C$  that does not have a unique convex combination of  $v_1, \dots, v_m$ .*
- (ii) *Every element of  $v \in C^\circ$  does not have a unique convex combination of  $v_1, \dots, v_m$ .*
- (iii) *For every element  $v \in C^\circ$  there are infinitely many convex combinations of  $v_1, \dots, v_m$ .*

*Proof.* Obviously, (iii)  $\implies$  (ii). As  $C^\circ$  is not empty (lemma 6), (ii)  $\implies$  (i). The implication (ii)  $\implies$  (iii) follows from lemma 3. We are only left to prove (i)  $\implies$  (ii). Let  $v \in C$  be an element so that

$$v = \sum_{i=1}^m \lambda_i v_i = \sum_{i=1}^m \mu_i v_i$$

are two different convex combinations of  $v$ . Note that

$$0 = \sum_{i=1}^m (\lambda_i - \mu_i)v_i.$$

Let  $w \in C^\circ$  have an open convex combination

$$w = \sum_{i=1}^m \nu_i v_i.$$

Let  $\alpha = \min_i \nu_i > 0$ . As  $\lambda_i - \mu_i \geq -1$ ,  $\nu_i + \alpha(\lambda_i - \mu_i) \geq 0$ , for all  $i$ , and  $\sum_{i=1}^m (\nu_i + \alpha(\lambda_i - \mu_i)) = \sum_{i=1}^m \nu_i + \alpha \sum_{i=1}^m (\lambda_i - \mu_i) = 1 + 0 = 1$ . So

$$w = \sum_{i=1}^m (\nu_i + \alpha(\lambda_i - \mu_i))v_i.$$

is another convex combination of  $w$ , because for at least one  $i \in \{1, \dots, m\}$ ,  $\lambda_i \neq \mu_i$ .  $\square$

## 2 Theorem of Carathéodory (1907)

**Theorem 8** (Carathéodory). *Let  $V$  be a real  $n$ -dimensional vector space. Let  $C$  be the convex hull of a set  $S$ . Then each element  $x$  is the convex combination of at most  $n + 1$  elements in  $S$ .*

*Proof.* Let  $y \in C$ . Let  $m$  be the smallest integer so that there are  $x_1, \dots, x_m \in C$  and  $\lambda_1, \dots, \lambda_m > 0$ ,  $\sum_{i=1}^m \lambda_i = 1$  so that

$$y = \sum_{i=1}^m \lambda_i x_i.$$

Suppose  $m > n + 1$ . As  $V$  is  $n$ -dimensional, there are scalars  $\alpha_2, \dots, \alpha_m$ , at least one of them positive, so that

$$0 = \sum_{i=2}^m \alpha_i (x_i - x_1).$$

Let  $\alpha_1 = -\sum_{i=2}^m \alpha_i$ . Then  $\sum_{i=1}^m \alpha_i = 0$ . It follows that

$$0 = \sum_{i=2}^m \alpha_i (x_i - x_1) = \alpha_1 x_1 + \sum_{i=2}^m \alpha_i x_i = \sum_{i=1}^m \alpha_i x_i.$$

Let  $\mu = \min_{i:\alpha_i > 0} \frac{\lambda_i}{\alpha_i}$ , and let  $j \in \{1, \dots, m\}$  be such that  $\mu = \frac{\lambda_j}{\alpha_j}$ . For all  $i \in \{1, \dots, m\}$ ,  $\lambda_i - \mu\alpha_i \geq 0$ , and  $\lambda_j - \mu\alpha_j = 0$ . Moreover, we have

$$\sum_{i \neq j} (\lambda_i - \mu\alpha_i) = \sum_{i=1}^m \lambda_i - \mu \sum_{i=1}^m \alpha_i = 1 - 0 = 1.$$

It follows that

$$\sum_{i \neq j} (\lambda_i - \mu\alpha_i) x_i = \sum_{i=1}^m \lambda_i x_i - \mu \sum_{i=1}^m \alpha_i x_i = y - 0 = y$$

is a convex combination of  $y$  with less than  $m$  elements, which is in contradiction with our assumption that  $m$  was the minimum number of elements of  $S$  needed to represent  $y$  as a convex combination of elements in  $S$ . It follows that each element in  $C$  can be represented with at most  $n + 1$  elements from  $S$ .  $\square$

### 3 Minimal sets and extreme elements

**Definition 9.** Let  $S$  be a subset of a vector space, and let  $C$  be the convex space generated by  $S$ . We call  $S$  minimal, when for every  $x \in S$ ,  $C \neq \text{co}(S \setminus \{x\})$ .

**Example 10.** Not all convex sets are generated by minimal sets. Take for example  $\mathbb{R}$ , which is convex, however, there is no finite subset  $S$  that generates  $\mathbb{R}$ , because when  $S$  is finite, set  $r = \max_{x \in S} |x|$ , then  $\text{co}(S) \subseteq [-r, r] \neq \mathbb{R}$ .  $S = \{\pm n : n \in \mathbb{N}\}$  is an example of a set that generates  $\mathbb{R}$ .

Also note that every set  $S$  that generates  $\mathbb{R}$  is not minimal. We already saw that is infinite. When  $x, y, z \in S$  and  $x \leq y \leq z$ , then  $S \setminus \{y\}$  also generates  $\mathbb{R}$ .

**Definition 11.** Let  $C$  be convex and  $x \in C$ . We call  $x$  extreme, when there are no  $y, z \in C$ ,  $y \neq z$  and  $\alpha \in (0, 1)$  so that  $x = \alpha y + (1 - \alpha)z$ .

**Lemma 12.** Let  $C$  be a convex set generated by a minimal set  $S$ . Then  $S$  is the set of all extrema of  $C$ .

*Proof.* For an extremum  $x \in C$ , there are no  $y, z \in C$ ,  $y \neq z$  and  $\alpha \in (0, 1)$  so that  $x = \alpha y + (1 - \alpha)z$ . So  $\text{co}(S \setminus \{x\})$  does not contain  $x$ . Hence  $x \in S$ .

Suppose  $x \in C$  is not extreme. Then there are  $y, z \in C$ ,  $x \neq y$  and  $\alpha \in (0, 1)$  so that  $x = \alpha y + (1 - \alpha)z$ . Then there are elements  $x_1, \dots, x_m \in S$  so that  $y, z$  are convex combinations

$$y = \sum_{i=1}^m \beta_i x_i, \quad \text{and} \quad z = \sum_{i=1}^m \gamma_i x_i.$$

Note that we can choose this  $x_i$  so that at least one of  $\beta_i$  or  $\gamma_i$  is positive. So

$$x = \sum_{i=1}^m (\alpha\beta_i + (1 - \alpha)\gamma_i) x_i.$$

If all  $x_i \neq x$ , then  $x$  is a convex combination of other elements of  $S$ , and so  $C = \text{co}(S \setminus \{x\})$ , so  $S$  is not minimal. Contradiction. So  $x$  is equal to some  $x_i$ . After relabelling, if necessary, we may assume  $x = x_1$ . As  $y \neq z$ , either  $\beta_1 < 1$  or  $\gamma_1 < 1$ , or both. We already saw that  $\beta_1 > 0$  or  $\gamma_1 > 0$ . So  $0 < \alpha\beta_1 + (1 - \alpha)\gamma_1 < 1$ . So

$$(1 - (\alpha\beta_1 + (1 - \alpha)\gamma_1))x = \sum_{i=2}^m (\alpha\beta_i + (1 - \alpha)\gamma_i)x_i.$$

Note that  $\sum_{i=2}^m (\alpha\beta_i + (1 - \alpha)\gamma_i) = 1 - (\alpha\beta_1 + (1 - \alpha)\gamma_1)$ , so

$$x = \sum_{i=2}^m \frac{\alpha\beta_i + (1 - \alpha)\gamma_i}{1 - (\alpha\beta_1 + (1 - \alpha)\gamma_1)} x_i,$$

is a convex combination of elements from  $S \setminus \{x\}$ . So  $C = \text{co}(S \setminus \{x\})$ , so  $S$  is not minimal. Contradiction. It follows that all elements of  $S$  are minimal.  $\square$

**Corollary 13.** *A convex set  $C$  has at most one minimal set. When it exist, it is unique, which allows us to speak about the minimal set.*

*Remark 14.* When  $C$  is generated by a minimal set  $S$ , then there is not necessarily a unique convex decomposition. Take for instance  $C = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1\}$ , which has minimal set  $S = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 = 1\}$ . Then

$$(0, 0) = \frac{1}{2}(-1, 0) + \frac{1}{2}(1, 0) \quad \text{and} \quad (0, 0) = \frac{1}{2}(0, -1) + \frac{1}{2}(0, 1).$$

**Lemma 15.** *Let  $C$  be a convex set generated by  $x_1, \dots, x_{k+1}$  and  $x_1 - x_{k+1}, \dots, x_k - x_{k+1}$  are linearly independent. Then  $S = \{x_1, \dots, x_{k+1}\}$  is a minimal set.*

*Proof.* Obviously,  $C$  is generated by  $S$ . Let  $x_j \in S$ . Suppose there are  $y, z \in C$  and  $\alpha \in (0, 1)$  so that

$$x_j = \alpha y + (1 - \alpha)z.$$

Then  $y$  and  $z$  have convex decompositions

$$y = \sum_{i=1}^m \beta_i x_i, \quad \text{and} \quad z = \sum_{i=1}^m \gamma_i x_i.$$

So

$$x_j = \sum_{i=1}^m (\alpha\beta_i + (1 - \alpha)\gamma_i)x_i,$$

is a convex decomposition of  $x_j$  in terms of  $x_1, \dots, x_m$ . As elements in  $C$  have a unique convex decomposition,  $\beta_i = \gamma_i = 0$  for all  $i \neq j$ , and  $\beta_j = \gamma_j = 1$ , so  $y = z$ , so  $x_j$  is extreme.  $\square$

**Corollary 16.**  *$C$  has a minimal set if and only if  $C$  is generated by its extreme points.*

**Lemma 17.** *When  $C$  is generated by a finite set, then  $C$  has a minimal set.*

*Proof.* Let  $S$  be a set of minimum cardinality that generates  $C$ , then  $S$  is minimal, because for  $x \in S$ ,  $S \setminus \{x\}$  has a lower cardinality than  $S$  and therefore does not generate  $C$ .  $\square$

**Corollary 18.** *When  $C$  does not have extreme points, then  $C$  is not generated by finite sets.*

*Remark 19.* A convex set might have extreme points, but not be generated by them, even if the convex space is precompact. Consider

$$C = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1, \text{ when } x, y \geq 0, \text{ otherwise } x^2 + y^2 < 1\}.$$

Then the set of extreme points is

$$S = \{(x, y) : x^2 + y^2 = 1, \text{ either } x < 0 \text{ or } y < 0\}.$$

## 4 Convex isomorphisms

**Definition 20.** Let  $C_1, C_2$  be convex sets. We call a map  $f : C_1 \rightarrow C_2$  a convex homomorphism when  $f(\alpha x + (1 - \alpha)y) = \alpha f(x) + (1 - \alpha)f(y)$  for all  $\alpha \in [0, 1]$ .

**Lemma 21.** When  $f : C_1 \rightarrow C_2$  is a bijective convex homomorphism,  $f^{-1}$  is also a convex homomorphism.

*Proof.* Let  $x, y \in C_2$  and  $\alpha \in [0, 1]$ . There are  $a, b \in C_1$  so that  $x = f(a), y = f(b)$ . So

$$\begin{aligned} f^{-1}(\alpha x + (1 - \alpha)y) &= f^{-1}(\alpha f(a) + (1 - \alpha)f(b)) \\ &= f^{-1}(f(\alpha a + (1 - \alpha)b)) \\ &= \alpha a + (1 - \alpha)b \\ &= \alpha f^{-1}(f(a)) + (1 - \alpha)f^{-1}(f(b)) \\ &= \alpha f^{-1}(x) + (1 - \alpha)f^{-1}(y). \end{aligned}$$

So  $f^{-1}$  is convex. □

**Definition 22.** A convex isomorphism is a bijective convex homomorphism.

**Definition 23.** Let  $C_i$  be a convex subset of a normed vector space  $V_i$ ,  $i = 1, 2$ . A convex isometrism is a convex isomorphism  $f : C_1 \rightarrow C_2$  so that for all  $x, y \in C_1$ ,  $\|x - y\| = \|f(x) - f(y)\|$ .

*Remark 24.* Note that a convex isomorphism  $f : C_1 \rightarrow C_2$  maps extreme points to extreme points. When  $C_1$  is generated by  $S_1$ , then  $C_2$  is generated by  $f(S_1)$ . When  $S_1$  is minimal, then  $f(S_1)$  is minimal. When every element in  $C_1$  has a unique convex decomposition in terms of  $S_1$ , then every element of  $C_2$  has a unique decomposition in terms of  $S_2$ .

**Lemma 25.** Let  $V, W$  be linear spaces and  $x_1, \dots, x_{k+1} \in V, y_1, \dots, y_{k+1} \in W$ , so that  $x_1 - x_{k+1}, \dots, x_k - x_{k+1}$  are linearly independent in  $V$  and  $y_1 - y_{k+1}, \dots, y_k - y_{k+1}$  are linearly independent in  $W$ . Let  $C$  be the convex hull of  $x_1, \dots, x_{k+1}$  and  $D$  the convex hull of  $y_1, \dots, y_{k+1}$ . Then  $f : C \rightarrow D$  defined by

$$f\left(\sum_{i=1}^{k+1} \alpha_i x_i\right) = \sum_{i=1}^{k+1} \alpha_i y_i$$

is an convex isomorphism.

When  $V$  and  $W$  are inner product spaces and  $x_1 - x_{k+1}, \dots, x_k - x_{k+1}$  are orthogonal in  $V$  and  $y_1 - y_{k+1}, \dots, y_k - y_{k+1}$  are orthogonal in  $W$ , and  $0 < \|x_i - x_{k+1}\| = \|y_i - y_{k+1}\|$ , for all  $i$ , then  $f$  is a convex isometrism.

*Proof.* As any element in  $C$  has a unique convex decomposition in terms of  $x_1, \dots, x_{k+1}$ ,  $f$  is well defined. By construction it is also a convex homomorphism and it is surjective. Now suppose  $f(a) = f(b)$ . There are unique convex decompositions

$$a = \sum_{i=1}^{k+1} \alpha_i x_i \quad \text{and} \quad b = \sum_{i=1}^{k+1} \beta_i x_i. \tag{1}$$

Using that  $f$  is a convex homomorphism, and that  $f(x_i) = y_i$  for all  $i$ ,

$$f(a) = \sum_{i=1}^{k+1} \alpha_i y_i = \sum_{i=1}^{k+1} \beta_i y_i = f(b).$$

As all elements in  $D$  have a unique convex decomposition,  $\alpha_i = \beta_i$ , for all  $i$ . It follows that  $a = b$ . So  $f$  is injective as well. So  $f$  is a convex isomorphism.

Now suppose that  $V$  and  $W$  are inner product spaces and  $x_1 - x_{k+1}, \dots, x_k - x_{k+1}$  are orthogonal in  $V$  and  $y_1 - y_{k+1}, \dots, y_k - y_{k+1}$  are orthogonal in  $W$ , and  $0 < \|x_i - x_{k+1}\| = \|y_i - y_{k+1}\|$ , for all  $i$ . In particular  $x_1 - x_{k+1}, \dots, x_k - x_{k+1}$  are linearly independent in  $V$ , and  $y_1 - y_{k+1}, \dots, y_k - y_{k+1}$  are linearly independent in  $W$ . So  $f : V \rightarrow W$  is a convex isomorphism. Let  $a, b \in C$ , with unique convex decompositions as in eq. (1). Then

$$\begin{aligned}
\|a - b\|^2 &= \left\| \sum_{i=1}^{k+1} \alpha_i x_i - \sum_{i=1}^{k+1} \beta_i x_i \right\|^2 \\
&= \left\| \sum_{i=1}^{k+1} \alpha_i x_i - x_{k+1} + x_{k+1} - \sum_{i=1}^{k+1} \beta_i x_i \right\|^2 \\
&= \left\| \sum_{i=1}^k \alpha_i (x_i - x_{k+1}) - \sum_{i=1}^k \beta_i (x_i - x_{k+1}) \right\|^2 \\
&= \left\| \sum_{i=1}^k (\alpha_i - \beta_i) (x_i - x_{k+1}) \right\|^2 \\
&= \sum_{i=1}^k (\alpha_i - \beta_i)^2 \|x_i - x_{k+1}\|^2.
\end{aligned}$$

similar,

$$\|f(a) - f(b)\|^2 = \sum_{i=1}^k (\alpha_i - \beta_i)^2 \|y_i - y_{k+1}\|^2.$$

So it follows from  $\|x_i - x_{k+1}\| = \|y_i - y_{k+1}\|$ , for all  $i$ , that  $\|f(a) - f(b)\| = \|a - b\|$ . Hence  $f$  is a convex isometrism.  $\square$