

# A sharp Blaschke–Santaló inequality for $\alpha$ -concave functions

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**Abstract** We define a new transform on  $\alpha$ -concave functions, which we call the  $\sharp$ -transform. Using this new transform, we prove a sharp Blaschke-Santaló inequality for  $\alpha$ -concave functions, and characterize the equality case. This extends the known functional Blaschke-Santaló inequality of Artstein-Avidan, Klartag and Milman, and strengthens a result of Bobkov.

Finally, we prove that the  $\sharp$ -transform is a duality transform when restricted to its image. However, this transform is neither surjective nor injective on the entire class of  $\alpha$ -concave functions.

**Keywords** Blaschke-Santaló inequality · convexity ·  $\alpha$ -concavity · log-concavity

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## 1 Blaschke–Santaló type inequalities

We begin by recalling the classic Blaschke-Santaló inequality. A *convex body* in  $\mathbb{R}^n$  is compact, convex set with non-empty interior. Such a convex body  $K$  is called *symmetric* if  $K = -K$ . We will denote by  $|K|$  the (Lebesgue) volume of  $K$ . Finally, we define the *polar body* of  $K$  to be

$$K^\circ = \{x \in \mathbb{R}^n : \langle x, y \rangle \leq 1 \text{ for all } y \in K\},$$

where  $\langle \cdot, \cdot \rangle$  is the standard scalar product on  $\mathbb{R}^n$ .

Polarity is a basic notion in convex geometry. It is easy to see that if  $K$  is symmetric, convex body, then so is  $K^\circ$ . The map  $K \mapsto K^\circ$  satisfies two fundamental properties:

1. It is *order reversing*: If  $K_1 \subseteq K_2$ , then  $K_1^\circ \supseteq K_2^\circ$ .
2. It is an *involution*: For every symmetric convex body  $K$  we have  $K = (K^\circ)^\circ$ .

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These two properties say that polarity is a *duality transform* (see, e.g., [2]).

The volume of  $K^\circ$  is related to the volume of  $K$  by the Blaschke-Santaló inequality

**Theorem 1 (Blaschke, Santaló)** *Assume  $K \subseteq \mathbb{R}^n$  is a symmetric, convex body. Then*

$$|K| \cdot |K^\circ| \leq |D|^2,$$

where  $D \subseteq \mathbb{R}^n$  is the unit Euclidean ball. Furthermore, we have an equality if and only if  $K$  is an ellipsoid.

This theorem was proven by Blaschke in dimensions 2 and 3, and Santaló extended the result to arbitrary dimensions. There exists a version of the inequality for non-symmetric bodies, but for simplicity we will only deal with the symmetric case. The generalized statement, proofs and further references can be found, e.g., in [12].

In [1], Artstein-Avidan, Klartag and Milman prove a functional extension of the Blaschke-Santaló inequality. To explain their result and put it in perspective, we will need to define  $\alpha$ -concave functions:

**Definition 1** Fix  $-\infty \leq \alpha \leq \infty$ . We say that a function  $f : \mathbb{R}^n \rightarrow [0, \infty)$  is  $\alpha$ -concave if  $f$  is supported on some convex set  $\Omega$ , and for every  $x, y \in \Omega$  and  $0 \leq \lambda \leq 1$  we have

$$f(\lambda x + (1 - \lambda)y) \geq [\lambda f(x)^\alpha + (1 - \lambda) f(y)^\alpha]^{\frac{1}{\alpha}}.$$

We will always assume that  $f$  is upper semicontinuous and that

$$\max_{x \in \mathbb{R}^n} f(x) = f(0) = 1$$

(this last condition is sometimes known as saying that  $f$  is *geometric*).

The class of all such  $\alpha$ -concave functions will be denoted by  $C_\alpha(\mathbb{R}^n)$ .

Remember that  $f : \mathbb{R}^n \rightarrow [0, \infty)$  is called upper semicontinuous if its upper level sets  $\{x \in \mathbb{R}^n : f(x) \geq t\}$  are closed for all  $t \geq 0$ .

$\alpha$ -concave functions were first defined by Avriel ([5]), and were studied by Borell ([8],[9]) and by Brascamp and Lieb ([10]). In the case  $\alpha = \infty$  we understand the definition in the limiting sense as

$$f(\lambda x + (1 - \lambda)y) \geq \max\{f(x), f(y)\}.$$

This just means that  $f$  is constant on its support, so functions in  $C_\infty(\mathbb{R}^n)$  are indicator functions of convex sets, which by our assumptions must also be closed and contain the origin. If we further assume that  $f \in C_\infty(\mathbb{R}^n)$  satisfy  $0 < \int f < \infty$ , then  $f$  must be the indicator function of a convex body, and we can identify these functions with the convex bodies themselves.

If  $\alpha_1 < \alpha_2$ , then  $C_{\alpha_1}(\mathbb{R}^n) \supseteq C_{\alpha_2}(\mathbb{R}^n)$ . This means that we can think of every class of functions  $C_\alpha(\mathbb{R}^n)$  as extending the class of convex bodies. Originally, this was done for the class  $C_0(\mathbb{R}^n)$  of *log-concave* functions. Again we interpret Definition 1 in the limiting sense, and say that a function  $f : \mathbb{R}^n \rightarrow [0, \infty)$  is log-concave if

$$f(\lambda x + (1 - \lambda)y) \geq f(x)^\lambda f(y)^{1-\lambda}.$$

for every  $x, y \in \mathbb{R}^n$  and  $0 \leq \lambda \leq 1$ .

Many definitions and theorems of convex geometry were generalized to the class of log-concave functions. Except the usual aspiration for generality, the developed theory helped to prove new deep theorems in convexity and asymptotic geometric analysis. For a survey of such results and their importance, see [16].

One of the first results in this new direction was the functional Santaló theorem. In order to state it we will begin with a convenient definition:

**Definition 2** For every  $f \in C_0(\mathbb{R}^n)$  we define

$$f^* = \exp(-\mathcal{L}(-\log f)) \in C_0(\mathbb{R}^n),$$

where  $\mathcal{L}$  is the Legendre transform, defined by

$$(\mathcal{L}\varphi)(x) = \sup_{y \in \mathbb{R}^n} (\langle x, y \rangle - \varphi(y)).$$

This definition just means that if  $f = e^{-\varphi}$  for a convex function  $\varphi$ , then  $f^* = e^{-\mathcal{L}\varphi}$ . It turns out that the map  $f \mapsto f^*$  is a duality transform on  $C_0(\mathbb{R}^n)$ , and we have the following theorem:

**Theorem 2** Assume  $f \in C_0(\mathbb{R}^n)$  is even and  $0 < \int f < \infty$ . Then

$$\int f \cdot \int f^* \leq \left( \int G \right)^2 = (2\pi)^n,$$

where

$$G(x) = e^{-\frac{|x|^2}{2}}.$$

Equality occurs if and only if  $f = G \circ T$  for an invertible linear map  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ .

This inequality, in the even case, is originally due to Ball ([6]). In [1], Artstein-Avidan, Klartag and Milman present this result as a Blaschke–Santaló type inequality, extend the result to the non-even case and characterize the equality case.

As a side note, let us mention that given our current knowledge, the form of Theorem 2 is somewhat surprising. Following a series of works by Artstein-Avidan and Milman, we now understand that even though the map  $f \mapsto f^*$  is a duality transform, it is *not* the correct extension of the classic notion of polarity (see [3]). The correct extension is another duality transform, usually denoted  $f \mapsto f^\circ$ , which is based on the so-called  $\mathcal{A}$ -transform. Hence we expect the functional Blaschke–Santaló inequality to bound an expression of the form

$$\int f \cdot \int f^\circ, \tag{1}$$

which is not what we see in Theorem 2. In fact, a sharp upper bound on (1) is not known, even though an asymptotic result was recently found by Artstein-Avidan and Slomka ([4]).

The goal of this paper is to discuss Blaschke–Santaló inequalities for  $\alpha$ -concave functions, for values of  $\alpha$  different from 0. The case  $\alpha > 0$  was resolved already in [1], so we will assume from now on that  $\alpha \leq 0$ . The following definition appeared in [17]:

**Definition 3** Assume  $-\infty < \alpha < 0$ . The convex base of a function  $f \in C_\alpha(\mathbb{R}^n)$  is

$$\text{base}_\alpha f = \frac{1 - f^\alpha}{\alpha}.$$

Put differently,  $\varphi = \text{base}_\alpha f$  is the unique convex function such that

$$f = \left(1 + \frac{\varphi}{\beta}\right)^{-\beta}.$$

Here and after, we use the parameter  $\beta = -\frac{1}{\alpha}$ . As  $\alpha \leq 0$  and  $\beta \geq 0$ , it is often less confusing to use  $\beta$  instead of  $\alpha$ . In the limiting case  $\alpha = 0$  we define  $\text{base}_0(f) = -\log f$ . Using the notion of a convex base we can extend Definition 2 to the general case:

**Definition 4** The dual of a function  $f \in C_\alpha(\mathbb{R}^n)$  is the function  $f^* \in C_\alpha(\mathbb{R}^n)$  defined by relation

$$\text{base}_\alpha(f^*) = \mathcal{L}(\text{base}_\alpha(f)).$$

Note that the operation  $*$  depends on  $\alpha$ . Remember that if  $f \in C_\alpha(\mathbb{R}^n)$  then  $f \in C_{\alpha'}(\mathbb{R}^n)$  for all  $\alpha' < \alpha$ . Thinking of  $f$  as an element in  $C_{\alpha'}(\mathbb{R}^n)$  will yield a different  $f^*$  than thinking about  $f$  as function in  $C_\alpha(\mathbb{R}^n)$ . Therefore, strictly speaking, we should use a notation like  $f^{*\alpha}$ . However, this notation is extremely cumbersome, so we will use the simpler notation  $f^*$ , and keep in mind the implicit dependence on  $\alpha$ .

We can now state what appears to be the natural extension of Theorem 2 to the  $\alpha$ -concave case:

**Theorem 3** Fix  $-\frac{1}{n} < \alpha \leq 0$ . For every even function  $f \in C_\alpha(\mathbb{R}^n)$  such that  $0 < \int f < \infty$  we have

$$\int f \cdot \int f^* \leq \left(\int H_\alpha\right)^2, \quad (2)$$

where

$$H_\alpha(x) = \left(1 + \frac{|x|^2}{\beta}\right)^{-\frac{\beta}{2}} \in C_\alpha(\mathbb{R}^n).$$

This theorem was proven by Bobkov ([7]) using a general result by Fradelizi and Meyer ([11]), which we will cite in the next section as Theorem 5. The condition  $\alpha > -\frac{1}{n}$  is necessary, because the function  $H_\alpha$  is no longer integrable for  $\alpha \leq -\frac{1}{n}$ . In fact, it is not hard to check that

$$\int H_\alpha \cdot \int H_\alpha^* \rightarrow \infty$$

as  $\alpha \rightarrow -\frac{1}{n}$ , so it is not possible to find a finite upper bound for  $\int f \cdot \int f^*$  whenever  $\alpha \leq -\frac{1}{n}$ . Notice that as  $\alpha \rightarrow 0$  the functions  $H_\alpha$  converge to the Gaussian  $G$ , and we obtain Theorem 2 as a special case of Theorem 3.

Surprisingly, as was already observed by Bobkov, Theorem 3 is not sharp when  $\alpha < 0$ . Given the previously described results, it is very natural to expect an equality in (2) when  $f = H_\alpha$ . However, an explicit (yet tedious) calculation shows that  $H_\alpha^*(x) < H_\alpha(x)$  for all  $x \neq 0$ , so

$$\int H_\alpha \cdot \int H_\alpha^* < \left(\int H_\alpha\right)^2.$$

In fact, there exists a unique function  $G_\alpha \in C_\alpha(\mathbb{R}^n)$  such that  $G_\alpha^* = G_\alpha$ , and this function is

$$G_\alpha(x) = \left(1 + \frac{|x|^2}{2\beta}\right)^{-\beta}.$$

Notice that in the limiting case  $\alpha \rightarrow 0$  we have  $H_0 = G_0 = G$ , but for other values of  $\alpha$  these functions are quite different. Inspired by Theorems 1 and 2, it may seem reasonable to conjecture that

$$\int f \cdot \int f^* \leq \left(\int G_\alpha\right)^2$$

for all even  $f \in C_\alpha(\mathbb{R}^n)$ . Such an inequality, if true, will obviously be sharp. Unfortunately, this inequality is false for  $\alpha < 0$ . As one possible counterexample, take  $f = \mathbf{1}_D$ , the indicator function of the ball. In this case

$$f^* = \left(1 + \frac{|x|}{\beta}\right)^{-\beta},$$

and a direct computation of the integrals show that this is indeed a counterexample if the dimension  $n$  is large enough compared to  $\beta$  (For concreteness, it is enough to take  $n = \lceil \frac{\beta}{2} \rceil$  for all large enough  $\beta$ ).

In the next sections we will show the reason inequality (2) is not sharp is that the transform  $*$  is not the correct extension of polarity to use in the functional Blaschke-Santaló inequality. In section 2 we will define a new transform on  $C_\alpha(\mathbb{R}^n)$ , which we call  $\sharp$ -transform. We will then use this  $\sharp$ -transform to prove a sharp version of Theorem 3. Finally, in section 3, we will discuss further properties of  $\sharp$  which are not directly related to the Blaschke-Santaló inequality, and give a geometric interpretation of this transform.

## 2 A new transform on $\alpha$ -concave functions

In [1], Artstein-Avidan, Klartag and Milman obtain Theorem 2 as the limit of Blaschke-Santaló type inequalities for  $\alpha$ -concave functions,  $\alpha \geq 0$ . Let us warn the reader that [1] uses a slightly different notation than the one used in this paper: an  $\alpha$ -concave function in this paper is the same as a  $\frac{1}{\alpha}$ -concave function in [1], and vice versa.

Inspired by the transforms in [1], we define the following:

**Definition 5** For  $f \in C_\alpha(\mathbb{R}^n)$  we define

$$f^\sharp(x) = \inf_y \frac{1}{f(y) \cdot \left(1 + \frac{\langle x, y \rangle}{\beta}\right)^\beta} = \frac{1}{\sup_y \left[ f(y) \cdot \left(1 + \frac{\langle x, y \rangle}{\beta}\right)^\beta \right]},$$

where the infimum is taken over all points  $y \in \mathbb{R}^n$  such that  $f(y) > 0$  and  $\langle x, y \rangle > -\beta$ .

Like the  $*$  transform, the  $\sharp$  transform also depends on  $\alpha$ , so in principle we should write  $f^{\sharp\alpha}$ . Nonetheless, we opt for the simpler notation  $f^\sharp$ .

Let us begin by checking that  $f^\sharp \in C_\alpha(\mathbb{R}^n)$ :

**Proposition 1** For every  $f \in C_\alpha(\mathbb{R}^n)$  we have  $f^\sharp \in C_\alpha(\mathbb{R}^n)$ . If  $f$  is even, so is  $f^\sharp$ .

*Proof* For a fixed  $y \in \mathbb{R}^n$  with  $f(y) > 0$  the function

$$f_y(x) = \begin{cases} \frac{1}{f(y)} \cdot \left(1 + \frac{\langle x, y \rangle}{\beta}\right)^{-\beta} & \text{if } \langle x, y \rangle > -\beta, \\ \infty & \text{otherwise} \end{cases}$$

is upper semicontinuous and  $\alpha$ -concave (except the fact it can attain the value  $+\infty$ , which we usually exclude from the definition). Now we can write

$$f^\sharp(x) = \inf_{y: f(y) > 0} f_y(x),$$

so  $f^\sharp$  is  $\alpha$ -concave as the infimum of a family of  $\alpha$ -concave functions. Similarly  $f^\sharp$  is upper semicontinuous, as the infimum of a family of upper semicontinuous functions.

For every  $x \in \mathbb{R}^n$  we have

$$f^\sharp(x) = \inf_y \frac{1}{f(y) \cdot \left(1 + \frac{\langle x, y \rangle}{\beta}\right)^\beta} \leq \frac{1}{f(0) \left(1 + \frac{\langle x, 0 \rangle}{\beta}\right)^\beta} = 1.$$

Additionally,

$$f^\sharp(0) = \inf_y \frac{1}{f(y) \left(1 + \frac{\langle 0, y \rangle}{\beta}\right)_+^\beta} = \inf_y \frac{1}{f(y)} = \frac{1}{\sup_y f(y)} = 1.$$

and we see that  $f^\sharp$  is geometric. Hence we have  $f^\sharp \in C_\alpha(\mathbb{R}^n)$  like we wanted.

Finally if  $f$  is even then

$$\begin{aligned} f^\sharp(-x) &= \inf_y \frac{1}{f(y) \cdot \left(1 + \frac{\langle -x, y \rangle}{\beta}\right)^\beta} = \inf_y \frac{1}{f(-y) \cdot \left(1 + \frac{\langle -x, -y \rangle}{\beta}\right)^\beta}, \\ &= \inf_y \frac{1}{f(y) \cdot \left(1 + \frac{\langle x, y \rangle}{\beta}\right)^\beta} = f^\sharp(x), \end{aligned}$$

so  $f^\sharp$  is even as well.

Our main goal in this section is to prove the following theorem:

**Theorem 4** Fix  $-\frac{1}{n} < \alpha \leq 0$ . For every even  $f \in C_\alpha(\mathbb{R}^n)$  such that  $0 < \int f < \infty$  we have

$$\int f \cdot \int f^\sharp \leq \left( \int H_\alpha \right)^2,$$

with equality if and only if  $f = H_\alpha \circ T$  for an invertible linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ .

Here we have  $H_\alpha(x) = \left(1 + \frac{|x|^2}{\beta}\right)^{-\frac{\beta}{2}}$ , just like in Theorem 3 of Bobkov. For  $\alpha \leq -\frac{1}{n}$  one cannot hope for a finite upper bound on the product  $\int f \cdot \int f^\sharp$ . This can be seen by choosing  $f = H_\alpha$ , and considering the following proposition:

**Proposition 2**  $H_\alpha^\sharp = H_\alpha$ , and  $H_\alpha$  is the only function with this property.

*Proof* First we calculate  $H_\alpha^\sharp$  explicitly and show that  $H_\alpha^\sharp = H_\alpha$ . By definition,

$$\begin{aligned} H_\alpha^\sharp(x) &= \left[ \sup_y H_\alpha(y) \left( 1 + \frac{\langle x, y \rangle}{\beta} \right)^\beta \right]^{-1} \\ &= \left[ \sup_y \underbrace{\left( 1 + \frac{|y|^2}{\beta} \right)^{-\frac{\beta}{2}}}_{(\star)} \left( 1 + \frac{\langle x, y \rangle}{\beta} \right)^\beta \right]^{-1}. \end{aligned}$$

Notice that if we take a vector  $y$  and rotate it to have the same direction as  $x$ , we can only increase the expression  $(\star)$ . Hence

$$\begin{aligned} H_\alpha^\sharp(x) &= \left[ \sup_{\lambda > 0} \left( 1 + \frac{|\lambda x|^2}{\beta} \right)^{-\frac{\beta}{2}} \left( 1 + \frac{\langle x, \lambda x \rangle}{\beta} \right)^\beta \right]^{-1} \\ &= \left[ \sup_{\lambda > 0} \frac{\left( 1 + \frac{\lambda |x|^2}{\beta} \right)^2}{1 + \frac{\lambda^2 |x|^2}{\beta}} \right]^{-\frac{\beta}{2}}. \end{aligned}$$

It is now an exercise in calculus to differentiate and check that the supremum is actually a maximum, which is obtained for  $\lambda = 1$ . Hence

$$H_\alpha^\sharp(x) = \left[ \frac{\left( 1 + \frac{|x|^2}{\beta} \right)^2}{1 + \frac{|x|^2}{\beta}} \right]^{-\frac{\beta}{2}} = \left( 1 + \frac{|x|^2}{\beta} \right)^{-\frac{\beta}{2}} = H_\alpha(x)$$

which is what we needed to show.

Now assume that  $f \in C_\alpha(\mathbb{R}^n)$  is any function such that  $f^\sharp = f$ . For every  $x \in \mathbb{R}^n$  with  $f(x) > 0$  we have

$$f(x) = f^\sharp(x) = \inf_y \frac{1}{f(y) \cdot \left( 1 + \frac{\langle x, y \rangle}{\beta} \right)^\beta} \leq \frac{1}{f(x) \cdot \left( 1 + \frac{|x|^2}{\beta} \right)^\beta},$$

so multiplying by  $f(x)$  and taking a square root we get  $f(x) \leq H_\alpha(x)$ . If  $f(x) = 0$  then  $f(x) \leq H_\alpha(x)$  holds trivially, so the inequality is true for all  $x \in \mathbb{R}^n$ .

It is obvious from the definition that  $\sharp$  is order reversing, so we may apply it on both sides and obtain

$$f = f^\sharp \geq H_\alpha^\sharp = H_\alpha,$$

so  $f = H_\alpha$  like we wanted.

Theorem 4, like Theorem 3, will follow from a general result of Fradelizi and Meyer ([11]). We state their result here:

**Theorem 5** Let  $f_1, f_2 : \mathbb{R}^n \rightarrow \mathbb{R}_+$  and  $\rho : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  be measurable functions, such that

$$f_1(x)f_2(y) \leq \rho^2(\langle x, y \rangle)$$

for every  $x, y \in \mathbb{R}^n$  such that  $\langle x, y \rangle > 0$ . If, additionally,  $f_1$  is even, then

$$\int f_1 \cdot \int f_2 \leq \left( \int \rho(|x|^2) dx \right)^2.$$

Assume further that  $\rho$  is continuous. Then equality will occur if and only if:

1.  $\sqrt{\rho(s)\rho(t)} \leq \rho(\sqrt{st})$  for every  $s, t \geq 0$ .
2. If  $n \geq 2$  then either  $\rho(0) > 0$  or  $\rho \equiv 0$ .
3. There exists a positive definite matrix  $T$  and a constant  $d > 0$  such that

$$f_1(x) = d \cdot \rho(|Tx|^2), \quad f_2(x) = \frac{1}{d} \cdot \rho(|T^{-1}x|^2)$$

almost everywhere.

Let us use this result to prove Theorem 4:

*Proof (Proof of Theorem 4)* Define a function  $\rho : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  by

$$\rho(t) = \left( 1 + \frac{t}{\beta} \right)^{-\beta/2}.$$

Fix  $x, y \in \mathbb{R}^n$  with  $\langle x, y \rangle > 0$ . If  $f(x) = 0$  then obviously  $f(x)f^\sharp(y) \leq \rho^2(\langle x, y \rangle)$ . If, on the other hand,  $f(x) > 0$  then

$$\begin{aligned} f(x) \cdot f^\sharp(y) &= \inf_z \frac{f(x)}{f(z) \cdot \left( 1 + \frac{\langle y, z \rangle}{\beta} \right)_+^\beta} \leq \frac{f(x)}{f(x) \cdot \left( 1 + \frac{\langle y, x \rangle}{\beta} \right)^\beta} \\ &= \left( 1 + \frac{\langle x, y \rangle}{\beta} \right)^{-\beta} = \rho^2(\langle x, y \rangle). \end{aligned}$$

From Theorem 5 we conclude that indeed

$$\int f \cdot \int f^\sharp \leq \left( \int \rho(|x|^2) dx \right)^2 = \left( \int H_\alpha \right)^2.$$

Next we analyze the equality case. From Theorem 5 we see that a necessary condition to have equality is

$$f(x) = d \cdot \rho(|Tx|^2) = d \cdot H_\alpha(Tx)$$

for a constant  $d > 0$  and a linear map  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  (which we may take to be positive definite if we want). Since  $f \in C_\alpha(\mathbb{R}^n)$  we know that

$$1 = f(0) = d \cdot H_\alpha(0) = d \cdot 1 = d,$$

so we must have  $f = H_\alpha \circ T$ .



To see that this condition is also sufficient, notice that for every  $f \in C_\alpha(\mathbb{R}^n)$  and every invertible linear map  $T$  we have

$$\begin{aligned} (f \circ T)^\sharp(x) &= \left[ \sup_y f(Ty) \left( 1 + \frac{\langle x, y \rangle}{\beta} \right)^\beta \right]^{-1} = \left[ \sup_y f(y) \left( 1 + \frac{\langle x, T^{-1}y \rangle}{\beta} \right)^\beta \right]^{-1} \\ &= \left[ \sup_y f(y) \left( 1 + \frac{\langle (T^{-1})^* x, y \rangle}{\beta} \right)^\beta \right]^{-1} = f^\sharp((T^{-1})^* x). \end{aligned}$$

Using a simple change of variables and Proposition 2 we get that if  $f = H_\alpha \circ T$  then

$$\begin{aligned} \int f \cdot \int f^\sharp &= \int (H_\alpha \circ T) \cdot \int (H_\alpha^\sharp \circ (T^{-1})^*) \\ &= \frac{1}{\det(T) \cdot \det((T^{-1})^*)} \int H_\alpha \int H_\alpha^\sharp = \left( \int H_\alpha \right)^2 \end{aligned}$$

so we are done.

*Remark 1* For simplicity, Theorem 4 is only stated for even functions. Fradelizi and Meyer also proved in [11] a generalization of Theorem 5 for non-even functions, which can be used to extend Theorem 4 to the non-even case. The proof remains essentially the same, so we leave the details to the interested reader.

To conclude this section, let us compare Theorem 4 with Theorem 3. We have the following proposition:

**Proposition 3** *For every  $f \in C_\alpha(\mathbb{R}^n)$  we have  $f^\sharp \geq f^*$ .*

*Proof* Denote  $\varphi = \text{base}_\alpha(f)$ . We need to prove that for every  $x \in \mathbb{R}^n$  we have  $f^\sharp(x) \geq f^*(x)$ , which is equivalent to

$$\inf_y \left( \frac{1 + \frac{\langle x, y \rangle}{\beta}}{1 + \frac{\varphi(y)}{\beta}} \right)^{-\beta} \geq \inf_y \left( 1 + \frac{\langle x, y \rangle - \varphi(y)}{\beta} \right)^{-\beta}.$$

Choose a sequence  $\{y_n\}$  such that

$$\left( \frac{1 + \frac{\langle x, y_n \rangle}{\beta}}{1 + \frac{\varphi(y_n)}{\beta}} \right)^{-\beta} \rightarrow f^\sharp(x).$$

We claim it is always possible to choose this sequence in such a way that  $\langle x, y_n \rangle \geq \varphi(y_n) \geq 0$  for every  $n$ . Indeed, we know that  $f^\sharp(x) \leq 1$ . If  $f^\sharp(x) < 1$  we will automatically have  $\langle x, y_n \rangle > \varphi(y_n) \geq 0$  for large enough  $n$ . If  $f^\sharp(x) = 1$ , just take  $y_n = 0$  for all  $n$ .

For every two numbers  $B \geq A \geq 0$  we have

$$\frac{1+B}{1+A} \leq 1+B-A,$$

as one easily checks. Applying this to  $B = \frac{\langle x, y_n \rangle}{\beta}$  and  $A = \frac{\varphi(y_n)}{\beta}$  we see that

$$\left( \frac{1 + \frac{\langle x, y_n \rangle}{\beta}}{1 + \frac{\varphi(y_n)}{\beta}} \right)^{-\beta} \geq \left( 1 + \frac{\langle x, y_n \rangle - \varphi(y_n)}{\beta} \right)^{-\beta} \geq f^*(x)$$

for all  $n$ . Sending  $n \rightarrow \infty$  we see that  $f^\sharp(x) \geq f^*(x)$  like we wanted.

This means that for every value of  $\alpha$  Theorem 3 follows from Theorem 4. When  $\alpha \rightarrow 0$  the transforms  $*$  and  $\sharp$  coincide, so both theorems reduce to same result - Theorem 2.

### 3 Further properties of $\sharp$ -transform

In this section we will discuss further properties of the new  $\sharp$ -transform,  $\sharp : C_\alpha(\mathbb{R}^n) \rightarrow C_\alpha(\mathbb{R}^n)$  for  $\alpha < 0$ . We already used the simple fact that  $\sharp$  is order reversing: if  $f, g \in C_\alpha(\mathbb{R}^n)$  and  $f \leq g$  (pointwise), then  $f^\sharp \geq g^\sharp$ . Surprisingly, however,  $\sharp$  is *not* a duality transform, as it is not an involution.

One simple way of verifying the last assertion is by computing a few examples:

*Example 1* Let  $K$  be a convex body containing the origin. Remember that the gauge function of  $K$  is defined by

$$\|x\|_K = \inf \left\{ r > 0 : \frac{x}{r} \in K \right\}.$$

Let us denote by  $\mathbf{1}_K$  the indicator function of  $K$ . Then a simple calculation gives

$$\mathbf{1}_K^\sharp = \mathbf{1}_K^* = \left( 1 + \frac{\|x\|_{K^\circ}}{\beta} \right)^{-\beta},$$

and

$$\left[ \left( 1 + \frac{\|x\|_K}{\beta} \right)^{-\beta} \right]^\sharp = \min \left\{ \frac{1}{\|x\|_{K^\circ}^\beta}, 1 \right\}.$$

In particular, we see that  $\mathbf{1}_K^{\sharp\sharp} \neq \mathbf{1}_K$  for every  $\alpha < 0$  (equivalently, for every  $\beta < \infty$ ).

In order to prove more delicate properties of the  $\sharp$ -transform, we will need to examine it from a different point view. Denote by  $\text{Cvx}_0(\mathbb{R}^n)$  the class of all convex functions  $\varphi : \mathbb{R}^n \rightarrow [0, \infty]$  such that  $\varphi$  is lower semicontinuous and  $\varphi(0) = 0$ . The map  $\text{base}_\alpha : C_\alpha(\mathbb{R}^n) \rightarrow \text{Cvx}_0(\mathbb{R}^n)$  from Definition 3 is easily seen to be an order reversing bijection. Hence, if we wish to understand the  $\sharp$ -transform, it is enough to study its conjugate  $\mathcal{T}_\alpha : \text{Cvx}_0(\mathbb{R}^n) \rightarrow \text{Cvx}_0(\mathbb{R}^n)$  which is defined by

$$\mathcal{T}_\alpha = \text{base}_\alpha \circ \sharp \circ (\text{base}_\alpha^{-1}).$$

The transform  $\mathcal{T}_\alpha$  can be written down explicitly:

**Proposition 4** *For every  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$  and every  $x \in \mathbb{R}^n$  we have*

$$(\mathcal{T}_\alpha \varphi)(x) = \sup_{y \in \mathbb{R}^n} \frac{\langle x, y \rangle - \varphi(y)}{1 - \alpha \varphi(y)} = \sup_{y \in \mathbb{R}^n} \frac{\langle x, y \rangle - \varphi(y)}{1 + \frac{\varphi(y)}{\beta}}. \quad (3)$$

In particular we see that  $\mathcal{T}_0 = \mathcal{L}$  is the Legendre transform. This also follows from the fact that on  $C_0(\mathbb{R}^n)$  the  $\sharp$ -transform and the  $*$ -transform coincide.

*Proof* Let us use equation (3) as the definition of  $\mathcal{T}_\alpha$ , and check that under this definition we really have

$$\mathcal{T}_\alpha = \text{base}_\alpha \circ \sharp \circ (\text{base}_\alpha^{-1}).$$

This is of course the same as  $(\text{base}_\alpha^{-1}) \circ \mathcal{T}_\alpha = \sharp \circ (\text{base}_\alpha^{-1})$ . Plugging in all of the definitions, we need to prove that for every  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$  and every  $x \in \mathbb{R}^n$

$$\left(1 + \frac{1}{\beta} \cdot \sup_y \frac{\langle x, y \rangle - \varphi(y)}{1 + \frac{\varphi(y)}{\beta}}\right)^{-\beta} = \inf_y \frac{\left(1 + \frac{\langle x, y \rangle}{\beta}\right)^{-\beta}}{\left(1 + \frac{\varphi(y)}{\beta}\right)^{-\beta}},$$

and checking this equality involves nothing more than simple algebra.

Interestingly, the transforms  $\mathcal{T}_\alpha$  were introduced and studied by Milman around 1970 for very different applications in functional analysis (see [13], [15] for the original papers in Russian and section 3.3 of the survey [14] for a partial translation to English. The remark in the end of section 3 of [1] is also relevant, but inaccurate). The only result we will need from these works is the following geometric characterization of  $\mathcal{T}_\alpha$ :

Fix a function  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$ . We will use  $\varphi$  to construct a function  $\rho : \mathbb{R}^n \times \mathbb{R} \rightarrow [0, \infty]$  in the following way: first, we define

$$\rho(x, \sqrt{\beta}) = \frac{\beta + \varphi(x)}{\sqrt{\beta}}.$$

Next, we extend  $\rho$  by requiring it to be 1-homogeneous. Hence for every  $x \in \mathbb{R}^n$  and  $t \neq 0$  we define

$$\rho(x, t) = \rho\left(\frac{t}{\sqrt{\beta}} \cdot \left(\frac{x\sqrt{\beta}}{t}, \sqrt{\beta}\right)\right) = \frac{|t|}{\sqrt{\beta}} \cdot \frac{\beta + \varphi\left(\frac{x\sqrt{\beta}}{t}\right)}{\sqrt{\beta}} = |t| + \frac{|t|}{\beta} \varphi\left(\frac{x\sqrt{\beta}}{t}\right).$$

The values of  $\rho$  on the hyperplane  $t = 0$  are not so important, but for concreteness we will define  $\rho(x, 0) = \lim_{t \rightarrow 0^+} \rho(x, t)$  (the limit exists by the convexity of  $\varphi$ ).

The function  $\rho$  is 1-homogeneous by construction, but in general it will *not* be a norm on  $\mathbb{R}^{n+1}$ , since there is no reason for  $\rho$  to be convex. Nonetheless, we can define the “dual” norm

$$\rho^*(x, t) = \sup_{(y, s) \in \mathbb{R}^{n+1}} \frac{\langle x, y \rangle + ts}{\rho(y, s)},$$

which is always a proper norm on  $\mathbb{R}^{n+1}$ . Now if we restrict ourselves back to the hyperplane  $t = \beta$  a direct calculation gives

$$\rho^*(x, \sqrt{\beta}) = \frac{\beta + (\mathcal{T}_\alpha \varphi)(x)}{\sqrt{\beta}},$$

which shows the relation between the transform  $\mathcal{T}_\alpha$  and the classic notion of duality.

Using this construction we can prove several properties of the  $\sharp$ -transform. Specifically we have:

**Theorem 6** Fix  $-\infty < \alpha < 0$ , and let  $\sharp : C_\alpha(\mathbb{R}^n) \rightarrow C_\alpha(\mathbb{R}^n)$  be the  $\sharp$ -transform. Then:

1. For every  $f \in C_\alpha(\mathbb{R}^n)$  we have  $f^{\sharp\sharp} \geq f$ .
2. For every  $f \in C_\alpha(\mathbb{R}^n)$  we have  $f^{\sharp\sharp\sharp} = f^\sharp$ . In other words,  $\sharp$  is a duality transform on its image.
3.  $\sharp$  is neither injective nor surjective.

Theorem 6 is an immediate corollary of the following proposition, establishing the same properties for the transform  $\mathcal{T}_\alpha$ :

**Proposition 5** Fix  $-\infty < \alpha < 0$ , and let  $\mathcal{T}_\alpha : \text{Cvx}_0(\mathbb{R}^n) \rightarrow \text{Cvx}_0(\mathbb{R}^n)$  be the transform defined above. Then:

1. For every  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$  we have  $\mathcal{T}_\alpha^2 \varphi \leq \varphi$ .
2. For every  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$  we have  $\mathcal{T}_\alpha^3 \varphi = \mathcal{T}_\alpha \varphi$ . In other words,  $\mathcal{T}_\alpha$  is a duality transform on its image.
3.  $\mathcal{T}_\alpha$  is neither injective nor surjective.

*Proof* Fix  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$ , and let  $\rho : \mathbb{R}^{n+1} \rightarrow [0, \infty]$  we defined as above. It is well known that if  $\rho$  is any 1-homogenous function, which is not necessarily convex, then  $\rho^{**} \leq \rho$ . In particular

$$\frac{\beta + (\mathcal{T}_\alpha^2 \varphi)(x)}{\sqrt{\beta}} = \rho^{**}(x, \sqrt{\beta}) \leq \rho(x, \sqrt{\beta}) = \frac{\beta + \varphi(x)}{\sqrt{\beta}},$$

which proves (1).

Since  $\rho^*$  is already a norm, we must have  $\rho^{***} = (\rho^*)^{**} = \rho^*$ . Restricting again to the hyperplane  $t = \beta$  we see that  $\mathcal{T}_\alpha^3 \varphi = \mathcal{T}_\alpha \varphi$ , which proves (2).

Next we prove (3), and begin by showing that  $\mathcal{T}_\alpha$  is not surjective. If  $\varphi$  is in the image of  $\mathcal{T}_\alpha$ , then the above discussion implies that the corresponding  $\rho$  must be a norm on  $\mathbb{R}^{n+1}$ . In particular,  $\rho$  must be comparable to the Euclidean norm, i.e. there exists a constant  $C > 0$  such that

$$\rho(x, t) \leq C |(x, t)| = C \sqrt{|x|^2 + t^2}.$$

Therefore

$$\varphi(x) = \sqrt{\beta} \cdot \rho(x, \sqrt{\beta}) - \beta \leq C \sqrt{\beta} \sqrt{|x|^2 + \beta} \leq C (\sqrt{\beta} |x| + \beta),$$

and we see that every function  $\varphi$  in the image of  $\mathcal{T}_\alpha$  must grow at most linearly. In particular, the function  $\varphi(x) = |x|^2$  is not in the image of  $\mathcal{T}_\alpha$ , so  $\mathcal{T}_\alpha$  is not surjective.

Finally, we will show that  $\mathcal{T}_\alpha$  is also not injective. Take any  $\varphi \in \text{Cvx}_0(\mathbb{R}^n)$  which is not in the image of  $\mathcal{T}_\alpha$ . Then  $\mathcal{T}_\alpha(\mathcal{T}_\alpha^2 \varphi) = \mathcal{T}_\alpha^3 \varphi = \mathcal{T}_\alpha \varphi$ , even though  $\mathcal{T}_\alpha^2 \varphi \neq \varphi$ . This shows that  $\mathcal{T}_\alpha$  is not injective, and the proof is complete.

## References

1. Shiri Artstein-Avidan, Bo'az Klartag, and Vitali Milman. The Santaló point of a function, and a functional form of the Santaló inequality. *Mathematika*, 51(1-2):33, February 2010.
2. Shiri Artstein-Avidan and Vitali Milman. A characterization of the concept of duality. *Electronic Research Announcements in Mathematical Sciences*, 14:42–59, 2007.
3. Shiri Artstein-Avidan and Vitali Milman. Hidden structures in the class of convex functions and a new duality transform. *Journal of the European Mathematical Society*, 13(4):975–1004, 2011.
4. Shiri Artstein-Avidan and Boaz Slomka. A note on Santaló inequality for the polarity transform and its reverse. *arXiv:1303.3114*, 2013.
5. Mordecai Avriel.  $r$ -convex functions. *Mathematical Programming*, 2(1):309–323, February 1972.
6. Keith Ball. *Isometric problems in  $\ell_p$  and sections of convex sets*. PhD thesis, University of Cambridge, 1987.
7. Sergey G. Bobkov. Convex bodies and norms associated to convex measures. *Probability Theory and Related Fields*, 147(1-2):303–332, March 2009.
8. Christer Borell. Convex measures on locally convex spaces. *Arkiv för matematik*, 12(1-2):239–252, December 1974.
9. Christer Borell. Convex set functions in  $d$ -space. *Periodica Mathematica Hungarica*, 6(2):111–136, 1975.
10. Herm J. Brascamp and Elliott H. Lieb. On extensions of the Brunn–Minkowski and Prékopa–Leindler theorems, including inequalities for log concave functions, and with an application to the diffusion equation. *Journal of Functional Analysis*, 22(4):366–389, August 1976.
11. Matthieu Fradelizi and Mathieu Meyer. Some functional forms of Blaschke–Santaló inequality. *Mathematische Zeitschrift*, 256(2):379–395, December 2006.
12. Mathieu Meyer and Alain Pajor. On the Blaschke–Santaló inequality. *Archiv der Mathematik*, 55(1):82–93, July 1990.
13. Vitali Milman. A certain transformation of convex functions and a duality of the  $\beta$  and  $\delta$  characteristics of a  $B$ -space. *Doklady Akademii Nauk SSSR*, 187:33–35, 1969.
14. Vitali Milman. Geometric theory of Banach spaces, part II: Geometry of the unit sphere. *Russian Mathematical Surveys*, 26(6):79–163, December 1971.
15. Vitali Milman. Duality of certain geometric characteristics of a Banach space. *Teorija Funkcii, Funkcionalnyi Analiz i ih Prilozenija*, 18:120–137, 1973.
16. Vitali Milman. Geometrization of probability. In Mikhail Kapranov, Sergiy Kolyada, Yuri Ivanovich Manin, Pieter Moree, and Leonid Potyagailo, editors, *Geometry and Dynamics of Groups and Spaces*, volume 265 of *Progress in Mathematics*, pages 647–667. Birkhäuser, Basel, 2008.
17. Liran Rotem. Support functions and mean width for  $\alpha$ -concave functions. *Advances in Mathematics*, 243:168–186, August 2013.