

## The Fenchel conjugate

---

*Ουκ ἐστ' ἐραστής ὅστις οὐκ αἰεὶ φιλεῖ*  
*(He is not a lover who does not love forever)*  
 (Euripides, “The Trojan Women”)

In the study of a (constrained) minimum problem it often happens that another problem, naturally related to the initial one, is useful to study. This is the so-called duality theory, and will be the subject of the next chapter.

In this one, we introduce a fundamental operation on convex functions that allows building up a general duality theory. Given an extended real valued function  $f$  defined on a Banach space  $X$ , its Fenchel conjugate  $f^*$  is a convex and lower semicontinuous function, defined on the dual space  $X^*$  of  $X$ . After defining it, we give several examples and study its first relevant properties. Then we observe that we can apply the Fenchel conjugation to  $f^*$  too, and this provides a new function, again defined on  $X$ , and minorizing everywhere the original function  $f$ . It coincides with  $f$  itself if and only if  $f \in \Gamma(X)$ , and is often called the convex, lower semicontinuous relaxation (or regularization) of  $f$ . Moreover, there are interesting connections between the subdifferentials of  $f$  and  $f^*$ ; we shall see that the graphs of the two subdifferentials are the same. Given the importance of this operation, a relevant question is to evaluate the conjugate of the sum of two convex functions. We then provide a general result in this sense, known as the Attouch–Brézis theorem.

### 5.1 Generalities

As usual, we shall denote by  $X$  a Banach space, and by  $X^*$  its topological dual.

**Definition 5.1.1** Let  $f: X \rightarrow (-\infty, \infty]$  be an arbitrary function. The *Fenchel conjugate* of  $f$  is the function  $f^*: X^* \rightarrow [-\infty, \infty]$  defined as

$$f^*(x^*) := \sup_{x \in X} \{\langle x^*, x \rangle - f(x)\}.$$

We have that

$$(x^*, \alpha) \in \text{epi } f^* \iff f(x) \geq \langle x^*, x \rangle - \alpha, \forall x \in X,$$

which means that the points of the epigraph of  $f^*$  parameterize the affine functions minorizing  $f$ . In other words, if the affine function  $l(x) = \langle x^*, x \rangle - \alpha$  minorizes  $f$ , then the affine function  $m(x) = \langle x^*, x \rangle - f^*(x^*)$  fulfills

$$l(x) \leq m(x) \leq f(x).$$

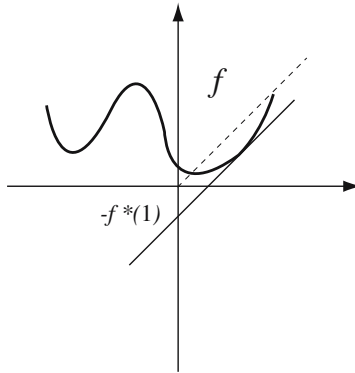
We also have that

$$\text{epi } f^* = \bigcap_{x \in X} \text{epi}\{\langle \cdot, x \rangle - f(x)\}.$$

Observe that even if  $f$  is completely arbitrary, its conjugate is a convex function, since  $\text{epi}\{\langle \cdot, x \rangle - f(x)\}$  is clearly a convex set for every  $x \in X$ . Furthermore, as  $\text{epi}\{\langle \cdot, x \rangle - f(x)\}$  is for all  $x$ , a closed set in  $X^* \times \mathbb{R}$  endowed with the product topology inherited by the weak\* topology on  $X^*$  and the natural topology on  $\mathbb{R}$ , it follows that for any arbitrary  $f$ ,  $\text{epi } f^* \subset X^* \times \mathbb{R}$  is a closed set in the above topology.

A geometrical way to visualize the definition of  $f^*$  can be captured by observing that

$$-f^*(x^*) = \sup\{\alpha : \alpha + \langle x^*, x \rangle \leq f(x), \forall x \in X\}.$$



**Figure 5.1.**

For,

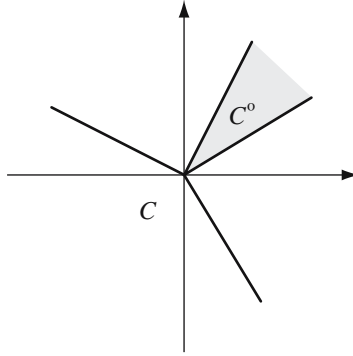
$$\begin{aligned} f^*(x^*) &= \inf\{-\alpha : \alpha + \langle x^*, x \rangle \leq f(x), \forall x \in X\} \\ &= -\sup\{\alpha : \alpha + \langle x^*, x \rangle \leq f(x), \forall x \in X\}. \end{aligned}$$

**Example 5.1.2** Here we see some examples of conjugates.

- (a) The conjugate of an affine function: for  $a \in X^*$ ,  $b \in \mathbb{R}$ , let  $f(x) = \langle a, x \rangle + b$ ; then

$$f^*(x^*) = \begin{cases} -b & \text{if } x^* = a, \\ \infty & \text{otherwise.} \end{cases}$$

- (b)  $f(x) = \|x\|$ ,  $f^*(x^*) = I_{B^*}(x^*)$ .  
 (c) Let  $X$  be a Hilbert space and  $f(x) = \frac{1}{2}\|x\|^2$ , then  $f^*(x^*) = \frac{1}{2}\|x^*\|_*^2$ , as one can see by looking for the maximizing point in the definition of the conjugate.  
 (d)  $f(x) = I_C(x)$ ,  $f^*(x^*) = \sup_{x \in C} \langle x^*, x \rangle := \sigma_C(x^*)$ ;  $\sigma_C$  is a positively homogeneous function, called the *support* function of  $C$ . If  $C$  is the unit ball of the space  $X$ , then  $f^*(x^*) = \|x^*\|_*$ . If  $C$  is a cone, the support function of  $C$  is the indicator function of the cone  $C^\circ$ , the polar cone of  $C$ , which is defined as  $C^\circ = \{x^* \in X^* : \langle x^*, x \rangle \leq 0, \forall x \in C\}$ . Observe that  $C^\circ$  is a weak\*-closed convex cone.



**Figure 5.2.** A cone  $C$  and its polar cone  $C^\circ$ .

**Exercise 5.1.3** Find  $f^*$ , for each  $f$  listed: (a)  $f(x) = e^x$ , (b)  $f(x) = x^4$ , (c)  $f(x) = \sin x$ , (d)  $f(x) = \max\{0, x\}$ , (e)  $f(x) = -x^2$ , (f)  $f(x, y) = xy$ ,

(g)  $f(x) = \begin{cases} e^x & \text{if } x \geq 0, \\ \infty & \text{otherwise,} \end{cases}$  (h)  $f(x) = \begin{cases} x \ln x & \text{if } x \geq 0, \\ \infty & \text{otherwise,} \end{cases}$

(i)  $f(x) = \begin{cases} 1 & \text{if } x \geq 0, \\ -1 & \text{otherwise;} \end{cases}$  (j)  $f(x) = (x^2 - 1)^2$ ,

(k)  $f(x) = \begin{cases} 0 & \text{if } |x| \leq 1, \\ (x^2 - 1)^2 & \text{otherwise.} \end{cases}$

The next proposition summarizes some elementary properties of  $f^*$ ; we leave the easy proofs as an exercise.

**Proposition 5.1.4** *We have:*

- (i)  $f^*(0) = -\inf f$ ;
- (ii)  $f \leq g \Rightarrow f^* \geq g^*$ ;
- (iii)  $(\inf_{j \in J} f_j)^* = \sup_{j \in J} f_j^*$ ;
- (iv)  $(\sup_{j \in J} f_j)^* \leq \inf_{j \in J} f_j^*$ ;
- (v)  $\forall r > 0, (rf)^*(x^*) = rf^*(\frac{x^*}{r})$ ;
- (vi)  $\forall r \in \mathbb{R}, (f+r)^*(x^*) = f^*(x^*) - r$ ;
- (vii)  $\forall \hat{x} \in X$ , if  $g(x) := f(x - \hat{x})$ , then  $g^*(x^*) = f^*(x^*) + \langle x^*, \hat{x} \rangle$ .

**Example 5.1.5** Let  $f(x) = x$ ,  $g(x) = -x$ . Then  $(\max\{f, g\})^*(x^*) = I_{[-1,1]}$ ,  $\min\{f^*, g^*\}(x^*) = 0$  if  $|x| = 1$ ,  $\infty$  elsewhere. Thus the inequality in the fourth item above can be strict, which is almost obvious from the fact that in general  $\inf_{j \in J} f_j^*$  need not be convex.

**Example 5.1.6** Let  $g: \mathbb{R} \rightarrow (-\infty, \infty]$  be an even function. Let  $f: X \rightarrow \mathbb{R}$  be defined as  $f(x) = g(\|x\|)$ . Then

$$f^*(x^*) = g^*(\|x^*\|_*).$$

For,

$$\begin{aligned} f^*(x^*) &= \sup_{x \in X} \{\langle x^*, x \rangle - g(\|x\|)\} = \sup_{t \geq 0} \sup_{\|x\|=t} \{\langle x^*, x \rangle - g(t)\} \\ &= \sup_{t \geq 0} \{t\|x^*\|_* - g(t)\} = \sup_{t \in \mathbb{R}} \{t\|x^*\|_* - g(t)\} = g^*(\|x^*\|_*). \end{aligned}$$

**Exercise 5.1.7** Let  $X$  be a Banach space,  $f(x) = \frac{1}{p}\|x\|^p$ , with  $p > 1$ . Then  $f^*(x^*) = \frac{1}{q}\|x^*\|^q$  ( $\frac{1}{p} + \frac{1}{q} = 1$ ).

The case  $p = 2$  generalizes Example 5.1.2 (c).

**Exercise 5.1.8** Let  $X$  be a Banach space, let  $A: X \rightarrow X$  be a linear, bounded and invertible operator. Finally, let  $f \in \Gamma(X)$  and  $g(x) = f(Ax)$ . Evaluate  $g^*$ .

*Hint.*  $g^*(x^*) = f^*((A^{-1})^*)(x^*)$ .

**Exercise 5.1.9** Evaluate  $f^*$  when  $f$  is

$$f(x) = \begin{cases} -\sqrt{x} & \text{if } x \geq 0, \\ \infty & \text{otherwise,} \end{cases} \quad f(x, y) = \begin{cases} -2\sqrt{xy} & \text{if } x \geq 0, y \geq 0, \\ \infty & \text{otherwise.} \end{cases}$$

**Exercise 5.1.10** Let  $X$  be a Banach space. Suppose  $\lim_{\|x\| \rightarrow \infty} \frac{f(x)}{\|x\|} = \infty$ . Prove that  $\text{dom } f^* = X^*$  and that the supremum in the definition of the conjugate of  $f$  is attained if  $X$  is reflexive.

**Exercise 5.1.11** Let  $X$  be a Banach space and let  $f \in \Gamma(X)$ . Then the following are equivalent:

- (i)  $\lim_{\|x\| \rightarrow \infty} f(x) = \infty$ ;

- (ii) there are  $c_1 > 0, c_2$  such that  $f(x) \geq c_1\|x\| - c_2$ ;
- (iii)  $0 \in \text{int dom } f^*$ .

Find an analogous formulation for the function  $f(x) - \langle x^*, x \rangle$ , where  $x^* \in X^*$ .

*Hint.* Suppose  $f(0) = 0$ , and let  $r$  be such that  $f(x) \geq 1$  if  $\|x\| \geq r$ . Then, for  $x$  such that  $\|x\| > r$ , we have that  $f(x) \geq \frac{\|x\|}{r}$ . Moreover, there exists  $\hat{c} < 0$  such that  $f(x) \geq \hat{c}$  if  $\|x\| \leq r$ . Then  $f(x) \geq \frac{\|x\|}{r} + \hat{c} - 1$  for all  $x$ . This shows that (i) implies (ii).

**Exercise 5.1.12** Let  $f \in \Gamma(X)$ . Then  $\lim_{\|x^*\|_* \rightarrow \infty} \frac{f^*(x^*)}{\|x^*\|_*} = \infty$  if and only if  $f$  is upper bounded on all the balls. In particular this happens in finite dimensions, if and only if  $f$  is real valued. On the contrary, in infinite dimensions there are continuous real valued convex functions which are not bounded on the unit ball.

*Hint.* Observe that the condition  $\lim_{\|x^*\|_* \rightarrow \infty} \frac{f^*(x^*)}{\|x^*\|_*} = \infty$  is equivalent to having that for each  $k > 0$ , there is  $c_k$  such that  $f^*(x^*) \geq k\|x^*\|_* - c_k$ . On the other hand,  $f$  is upper bounded on  $kB$  if and only if there exists  $c_k$  such that  $f(x) \leq I_{kB}(x) + c_k$ .

## 5.2 The bijection between $\Gamma(X)$ and $\Gamma^*(X^*)$

Starting from a given arbitrary function  $f$ , we have built its conjugate  $f^*$ . Of course, we can apply the same conjugate operation to  $f^*$ , too. In this way, we shall have a new function  $f^{**}$ , defined on  $X^{**}$ . But we are not interested in it. We shall instead focus our attention to its *restriction* to  $X$ , and we shall denote it by  $f^{**}$ . Thus

$$f^{**}: X \rightarrow [-\infty, \infty]; f^{**}(x) = \sup_{x^* \in X^*} \{\langle x^*, x \rangle - f^*(x^*)\}.$$

In this section, we study the connections between  $f$  and  $f^{**}$ .

**Proposition 5.2.1** *We have  $f^{**} \leq f$ .*

*Proof.*  $\forall x \in X, \forall x^* \in X^*$ ,

$$\langle x^*, x \rangle - f^*(x^*) \leq f(x).$$

Taking the supremum over  $x^* \in X^*$  in both sides provides the result. □

**Definition 5.2.2** We define the *convex, lower semicontinuous regularization* of  $f: X \rightarrow (-\infty, \infty]$  to be the function  $\hat{f}$  such that

$$\text{epi } \hat{f} = \text{cl co epi } f.$$

The definition is consistent because the convex hull of an epigraph is still an epigraph. Clearly,  $\hat{f}$  is the largest convex (the closure of a convex set is convex) and lower semicontinuous function minorizing  $f$ : if  $g \leq f$  and  $g$  is convex and lower semicontinuous, then  $g \leq \hat{f}$ . For,  $\text{epi } g$  is a closed convex set containing  $\text{epi } f$ , hence it contains  $\text{cl co epi } f$ .

**Remark 5.2.3** If  $f$  is convex, then  $\hat{f} = \bar{f}$ . If  $f \in \Gamma(X)$ , then  $f = \hat{f}$ . This easily follows from

$$\text{epi } f = \text{cl co epi } f.$$

Observe that we always have  $\hat{f} \geq f^{**}$ , as  $f^{**} \leq f$  and  $f^{**}$  is convex and lower semicontinuous.

The next theorem provides a condition to ensure that  $\hat{f}$  and  $f^{**}$  coincide. Exercise 5.2.5 shows that such a condition is not redundant.

**Theorem 5.2.4** Let  $f: X \rightarrow (-\infty, \infty]$  be such that there are  $x^* \in X^*$ ,  $\alpha \in \mathbb{R}$  with  $f(x) \geq \langle x^*, x \rangle + \alpha, \forall x \in X$ . Then  $\hat{f} = f^{**}$ .

*Proof.* The claim is obviously true if  $f$  is not proper, as in such a case, both  $f^{**}$  and  $\hat{f}$  are constantly  $\infty$ . Then we have that  $\forall x \in X$ ,

$$\hat{f}(x) \geq f^{**}(x) \geq \langle x^*, x \rangle + \alpha.$$

The last inequality follows from the fact that  $f \geq g \implies f^{**} \geq g^{**}$  and that the biconjugate of an affine function coincides with the affine function itself. Thus  $f^{**}(x) > -\infty$  for all  $x$ . Let us suppose now, for the sake of contradiction, that there is  $x_0 \in X$  such that  $f^{**}(x_0) < \hat{f}(x_0)$ . It is then possible to separate  $(x_0, f^{**}(x_0))$  and  $\text{epi } \hat{f}$ . If  $\hat{f}(x_0) < \infty$ , we then get the existence of  $y^* \in X^*$  such that

$$\langle y^*, x_0 \rangle + f^{**}(x_0) < \langle y^*, x \rangle + \hat{f}(x) \leq \langle y^*, x \rangle + f(x), \forall x \in X.$$

(To be sure of this, take a look at the proof of Theorem 2.2.21). This implies

$$f^{**}(x_0) < \langle -y^*, x_0 \rangle - \sup_{x \in X} \{ \langle -y^*, x \rangle - f(x) \} = \langle -y^*, x_0 \rangle - f^*(-y^*),$$

which is impossible. We then have to understand what is going on when  $\hat{f}(x_0) = \infty$ . In the case that the separating hyperplane is not vertical, one concludes as before. In the other case, we have the existence of  $y^* \in X^*$ ,  $c \in \mathbb{R}$  such that

- (i)  $\langle y^*, x \rangle - c < 0 \forall x \in \text{dom } f$ ;
- (ii)  $\langle y^*, x_0 \rangle - c > 0$ .

Then

$$f(x) \geq \langle x^*, x \rangle + \alpha + t(\langle y^*, x \rangle - c), \forall x \in X, t > 0,$$

and this in turn implies, by conjugating twice, that

$$f^{**}(x) \geq \langle x^*, x \rangle + \alpha + t(\langle y^*, x \rangle - c), \forall x \in X, t > 0.$$

But then

$$f^{**}(x_0) \geq \langle x^*, x_0 \rangle + \alpha + t(\langle y^*, x_0 \rangle - c), \forall t > 0,$$

which implies  $f^{**}(x_0) = \infty$ . □

**Exercise 5.2.5** Let

$$f(x) = \begin{cases} -x^2 & \text{if } x \leq 0, \\ \infty & \text{otherwise.} \end{cases}$$

Find  $f^{**}$  and  $\hat{f}$ .

**Proposition 5.2.6** *Let  $f: X \rightarrow [-\infty, \infty]$  be a convex function and suppose  $f(x_0) \in \mathbb{R}$ . Then  $f$  is lower semicontinuous at  $x_0$  if and only if  $f(x_0) = f^{**}(x_0)$ .*

*Proof.* We always have  $f^{**}(x_0) \leq f(x_0)$  (Proposition 5.2.1). Now, suppose  $f$  is lower semicontinuous at  $x_0$ . Let us see first that  $\bar{f}$  cannot assume value  $-\infty$  at any point. On the contrary, suppose there is  $z$  such that  $\bar{f}(z) = -\infty$ . Then  $\bar{f}$  is never real valued, and so  $\bar{f}(x_0) = -\infty$ , against the fact that  $f$  is lower semicontinuous and real valued at  $x_0$ . It follows that  $\bar{f}$  has an affine minorizing function; thus

$$\bar{f} = \hat{\bar{f}} = (\bar{f})^{**} \leq f^{**}.$$

As  $\bar{f}(x_0) = f(x_0)$ , we finally have  $f(x_0) = f^{**}(x_0)$ . Suppose now  $f(x_0) = f^{**}(x_0)$ . Then

$$\liminf f(x) \geq \liminf f^{**}(x) \geq f^{**}(x_0) = f(x_0),$$

and this shows that  $f$  is lower semicontinuous at  $x_0$ . □

The function

$$f(x) = \begin{cases} -\infty & \text{if } x = 0, \\ \infty & \text{otherwise,} \end{cases}$$

shows that the assumption  $f(x_0) \in \mathbb{R}$  is *not* redundant in the above proposition. A more sophisticated example is the following one. Consider an infinite dimensional Banach space  $X$ , take  $x^* \in X^*$  and a linear discontinuous functional  $l$  on  $X$ . Define

$$f(x) = \begin{cases} l(x) & \text{if } \langle x^*, x \rangle \geq 1, \\ \infty & \text{otherwise.} \end{cases}$$

Then  $f$  is continuous at zero, and it can be shown that  $f^{**}(x) = -\infty$  for all  $x$ . Observe that  $f$  is lower semicontinuous at no point of its effective domain. This is the case because it can be shown that if there is at least a point of the effective domain of  $f$  where  $f$  is lower semicontinuous, then  $f(x) = f^{**}(x)$

for all  $x$  such that  $f$  is lower semicontinuous (not necessarily real valued) at  $x$  ([Si2, Theorem 3.4]).

The next proposition shows that iterated application of the conjugation operation *does not* provide new functions.

**Proposition 5.2.7** *Let  $f: X \rightarrow (-\infty, \infty]$ . Then  $f^* = f^{***}$ .*

*Proof.* As  $f^{**} \leq f$ , one has  $f^* \leq f^{***}$ . On the other hand, by definition of  $f^{***}$ , we have  $f^{***}(x^*) = \sup_x \{\langle x^*, x \rangle - f^{**}(x)\}$ , while, for all  $x \in X$ ,  $f^*(x^*) \geq \langle x^*, x \rangle - f^{**}(x)$ , and this allows to conclude.  $\square$

Denote by  $\Gamma^*(X^*)$  the functions of  $\Gamma(X^*)$  which are conjugate of some function of  $\Gamma(X)$ . Then, from the previous results we get:

**Theorem 5.2.8** *The operator  $*$  is a bijection between  $\Gamma(X)$  and  $\Gamma^*(X^*)$ .*

*Proof.* If  $f \in \Gamma(X)$ ,  $f^*$  cannot be  $-\infty$  at any point. Moreover,  $f^*$  cannot be identically  $\infty$  as there is an affine function  $l(\cdot)$  of the form  $l(x) = \langle x^*, x \rangle - r$  minorizing  $f$  (Corollary 2.2.17), whence  $f^*(x^*) \leq r$ . These facts imply that  $*$  actually acts between  $\Gamma(X)$  and  $\Gamma^*(X^*)$ . To conclude, it is enough to observe that if  $f \in \Gamma(X)$ , then  $f = f^{**}$  (Proposition 5.2.4).  $\square$

**Remark 5.2.9** If  $X$  is not reflexive, then  $\Gamma^*(X^*)$  is a proper subset of  $\Gamma(X^*)$ . It is enough to consider a linear functional on  $X^*$  which is the image of no element of  $X$  via the canonical embedding of  $X$  into  $X^{**}$ ; it belongs to  $\Gamma(X^*)$ , but it is not the conjugate of any function  $f \in \Gamma(X)$ .

### 5.3 The subdifferentials of $f$ and $f^*$

Let us see, by a simple calculus in a special setting, how it is possible to evaluate the conjugate  $f^*$  of a function  $f$ , and the connection between the derivative of  $f$  and that of  $f^*$ . Let  $f: \mathbb{R}^n \rightarrow (-\infty, \infty]$  be a convex function. Since  $f^*(x^*) = \sup_{x \in X} \{\langle x^*, x \rangle - f(x)\}$ , we start by supposing that  $f$  is superlinear ( $\lim_{\|x\| \rightarrow \infty} \frac{f(x)}{\|x\|} = \infty$ ) and thus we have that the supremum in the definition of the conjugate is attained, for every  $x^*$ . To find a maximum point, like every student we assume that the derivative of  $f$  is zero at the maximum point, called  $\bar{x}$ . We get  $x^* - \nabla f(\bar{x}) = 0$ . We suppose also that  $\nabla f$  has an inverse. Then  $\bar{x} = (\nabla f)^{-1}(x^*)$ . By substitution we get

$$f^*(x^*) = \langle x^*, (\nabla f)^{-1}(x^*) \rangle - f((\nabla f)^{-1}(x^*)).$$

We try now to determine  $\nabla f^*(x^*)$ . We get

$$\begin{aligned} \nabla f^*(x^*) &= (\nabla f)^{-1}(x^*) + \langle J_{(\nabla f)^{-1}}(x^*), x^* \rangle - \langle J_{(\nabla f)^{-1}}(x^*), \nabla f((\nabla f)^{-1}(x^*)) \rangle \\ &= (\nabla f)^{-1}(x^*), \end{aligned}$$



where  $J_{(\nabla f)^{-1}}$  denotes the jacobian matrix of the function  $(\nabla f)^{-1}$ . Then we have the interesting fact that the derivative of  $f$  is the inverse of the derivative of  $f^*$ . This fact can be fully generalized to subdifferentials, as we shall see in a moment.

**Proposition 5.3.1** *Let  $f: X \rightarrow (-\infty, \infty]$ . Then  $x^* \in \partial f(x)$  if and only if  $f(x) + f^*(x^*) = \langle x^*, x \rangle$ .*

*Proof.* We already know that

$$f(x) + f^*(x^*) \geq \langle x^*, x \rangle, \forall x \in X, x^* \in X^*.$$

If  $x^* \in \partial f(x)$ , then

$$f(y) - \langle x^*, y \rangle \geq f(x) - \langle x^*, x \rangle, \forall y \in X,$$

whence,  $\forall y \in X$ ,

$$\langle x^*, y \rangle - f(y) + f(x) \leq \langle x^*, x \rangle.$$

Taking the supremum over all  $y$  in the left side provides one implication. As to the other one, if  $f(x) + f^*(x^*) = \langle x^*, x \rangle$ , then from the definition of  $f^*$ , we have that

$$f(x) + \langle x^*, y \rangle - f(y) \leq \langle x^*, x \rangle, \forall y \in X,$$

which shows that  $x^* \in \partial f(x)$ . □

Proposition 5.3.1 has some interesting consequences. At first,

**Proposition 5.3.2** *Let  $f: X \rightarrow (-\infty, \infty]$ . If  $\partial f(x) \neq \emptyset$ , then  $f(x) = f^{**}(x)$ . If  $f(x) = f^{**}(x)$ , then  $\partial f(x) = \partial f^{**}(x)$ .*

*Proof.*  $\forall x \in X, \forall x^* \in X^*$ , we have

$$f^*(x^*) + f^{**}(x) \geq \langle x^*, x \rangle.$$

If  $x^* \in \partial f(x)$ , by Proposition 5.3.1 we get

$$f^*(x^*) + f(x) = \langle x^*, x \rangle.$$

It follows that  $f^{**}(x) \geq f(x)$ , and this shows the first part of the claim. Suppose now  $f(x) = f^{**}(x)$ . Then, using the equality  $f^* = (f^{**})^*$ ,

$$\begin{aligned} x^* \in \partial f(x) &\iff \langle x^*, x \rangle = f(x) + f^*(x^*) = f^{**}(x) + f^{***}(x^*) \\ &\iff x^* \in \partial f^{**}(x). \end{aligned}$$

□

Another interesting consequence is the announced connection between the subdifferentials of  $f$  and  $f^*$ .

**Corollary 5.3.3** *Let  $f: X \rightarrow (-\infty, \infty]$ . Then*

$$x^* \in \partial f(x) \implies x \in \partial f^*(x^*).$$

*If  $f(x) = f^{**}(x)$ , then*

$$x^* \in \partial f(x) \text{ if and only if } x \in \partial f^*(x^*).$$

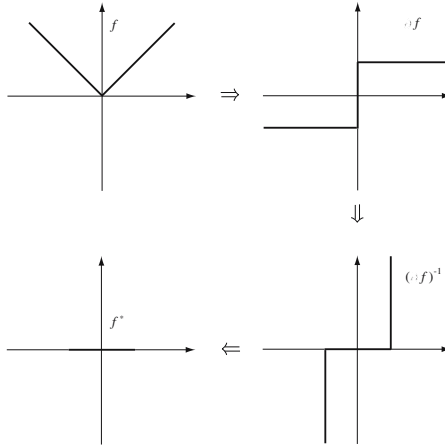
*Proof.*  $x^* \in \partial f(x) \iff \langle x^*, x \rangle = f(x) + f^*(x^*)$ . Thus  $x^* \in \partial f(x)$  implies  $f^{**}(x) + f^*(x^*) \leq \langle x^*, x \rangle$ , and this is equivalent to saying that  $x \in \partial f^*(x^*)$ . If  $f(x) = f^{**}(x)$ ,

$$\begin{aligned} x^* \in \partial f(x) &\iff \langle x^*, x \rangle = f(x) + f^*(x^*) = f^{**}(x) + f^*(x^*) \\ &\iff x \in \partial f^*(x^*). \end{aligned}$$

□

Thus, for a function  $f \in \Gamma(X)$ , it holds that  $x^* \in \partial f(x)$  if and only if  $x \in \partial f^*(x^*)$ .

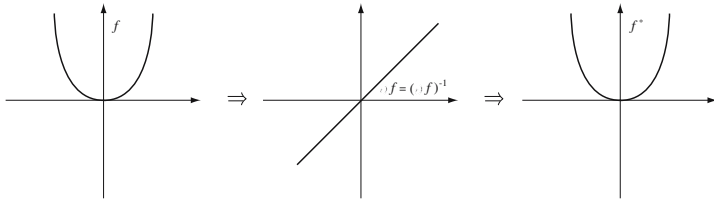
The above conclusion suggests how to draw the graph of the conjugate of a given function  $f: \mathbb{R} \rightarrow \mathbb{R}$ . We can construct the graph of its subdifferential, we “invert” it and we “integrate”, remembering that, for instance,  $f^*(0) = -\inf f$ . See Figures 5.3–5.5 below.



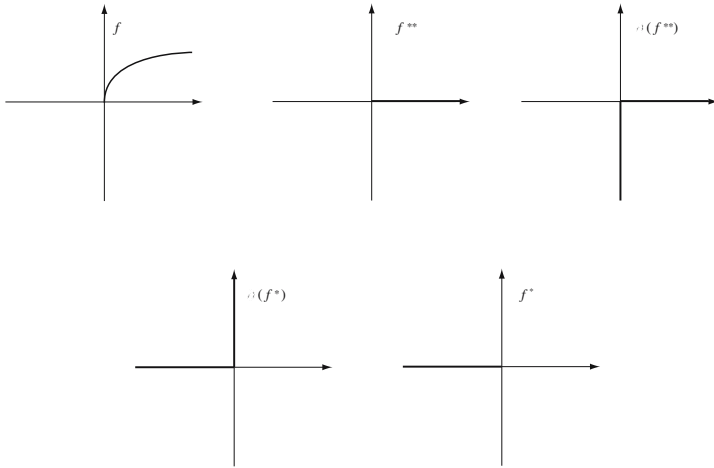
**Figure 5.3.** From the function to its conjugate through the subdifferentials.

A similar relation holds for approximate subdifferentials. For the following generalization of Proposition 5.3.1 holds:

**Proposition 5.3.4** *Let  $f \in \Gamma(X)$ . Then  $x^* \in \partial_\epsilon f(x)$  if and only if  $f^*(x^*) + f(x) \leq \langle x^*, x \rangle + \epsilon$ . Hence,  $x^* \in \partial_\epsilon f(x)$  if and only if  $x \in \partial_\epsilon f^*(x^*)$ .*



**Figure 5.4.** Another example.



**Figure 5.5.** ... and yet another one.

*Proof.*  $x^* \in \partial_\varepsilon f(x)$  if and only if

$$f(x) + \langle x^*, y \rangle - f(y) \leq \langle x^*, x \rangle + \varepsilon, \forall y \in X,$$

if and only if  $f(x) + f^*(x^*) \leq \langle x^*, x \rangle + \varepsilon$ . The second claim follows from  $f = f^{**}$ . □

The previous proposition allows us to show that only in exceptional cases can the approximate subdifferential be a singleton (a nonempty, small set indeed).

**Proposition 5.3.5** *Let  $f \in \Gamma(X)$  and suppose there are  $x \in \text{dom } f$ ,  $x^* \in X^*$  and  $\bar{\varepsilon} > 0$  such that  $\partial_{\bar{\varepsilon}} f(x) = \{x^*\}$ . Then  $f$  is an affine function.*

*Proof.* As a first step one verifies that  $\partial_\varepsilon f(x) = \{x^*\}$  for all  $\varepsilon > 0$ . This is obvious if  $\varepsilon < \bar{\varepsilon}$ , because  $\partial_\varepsilon f(x) \neq \emptyset$ , and due to monotonicity. Furthermore, the convexity property described in Theorem 3.7.2 implies that  $\partial_\varepsilon f(x)$  is a singleton also for  $\varepsilon > \bar{\varepsilon}$ . For, take  $\sigma < \bar{\varepsilon}$  and suppose  $\partial_\varepsilon f(x) \ni y^* \neq x^*$ , for some  $\varepsilon > \bar{\varepsilon}$ . An easy but tedious calculation shows that being  $\partial_\sigma f(x) \ni x^*$ ,

$\partial_\varepsilon f(x) \ni \frac{\varepsilon - \bar{\varepsilon}}{\varepsilon - \sigma} x^* + \frac{\bar{\varepsilon} - \sigma}{\varepsilon - \sigma} y^* \neq x^*$ , a contradiction. It follows, by Proposition 5.3.4, that if  $y^* \neq x^*$ ,

$$f^*(y^*) > \langle y^*, x \rangle - f(x) + \varepsilon, \forall \varepsilon > 0,$$

and this implies  $\text{dom } f^* = \{x^*\}$ . We conclude that  $f$  must be an affine function.  $\square$

## 5.4 The conjugate of the sum

**Proposition 5.4.1** *Let  $f, g \in \Gamma(X)$ . Then*

$$(f \nabla g)^* = f^* + g^*.$$

*Proof.*

$$\begin{aligned} (f \nabla g)^*(x^*) &= \sup_{x \in X} \{ \langle x^*, x \rangle - \inf_{x_1 + x_2 = x} \{ f(x_1) + g(x_2) \} \} \\ &= \sup_{\substack{x_1 \in X \\ x_2 \in X}} \{ \langle x^*, x_1 \rangle + \langle x^*, x_2 \rangle - f(x_1) - g(x_2) \} = f^*(x^*) + g^*(x^*). \end{aligned}$$

$\square$

Proposition 5.4.1 offers a good idea for evaluating  $(f + g)^*$ . By applying the above formula to  $f^*, g^*$  and conjugating, we get that

$$(f^* \nabla g^*)^{**} = (f^{**} + g^{**})^* = (f + g)^*.$$

So that if  $f^* \nabla g^* \in \Gamma(X^*)$ , then

$$(f + g)^* = f^* \nabla g^*.$$

Unfortunately we know that the inf-convolution operation between functions in  $\Gamma(X)$  does not always produce a function belonging to  $\Gamma(X)$ ; besides the case when at some point it is valued  $-\infty$ , it is not always lower semicontinuous. The next important theorem, due to Attouch–Brézis (see [AB]), provides a sufficient condition to get the result.

**Theorem 5.4.2** *Let  $X$  be a Banach space and  $X^*$  its dual space. Let  $f, g \in \Gamma(X)$ . Moreover, let*

$$F := \mathbb{R}^+(\text{dom } f - \text{dom } g)$$

*be a closed vector subspace of  $X$ . Then*

$$(f + g)^* = f^* \nabla g^*,$$

*and the inf-convolution is exact.*

*Proof.* From the previous remark, it is enough to show that the inf-convolution is lower semicontinuous; in proving this we shall also see that it is exact (whence, in particular, it never assumes the value  $-\infty$ ). We start by proving the claim in the particular case when  $F = X$ . From Exercise 2.2.4 it is enough to show that the level sets  $(f^* \nabla g^*)^a$  are weak\* closed for all  $a \in \mathbb{R}$ . On the other hand,

$$(f^* \nabla g^*)^a = \bigcap_{\varepsilon > 0} C_\varepsilon := \{y^* + z^* : f^*(y^*) + g^*(z^*) \leq a + \varepsilon\}.$$

It is then enough to show that the sets  $C_\varepsilon$  are weak\* closed. Fixing  $r > 0$ , let us consider

$$K_{\varepsilon r} := \{(y^*, z^*) : f^*(y^*) + g^*(z^*) \leq a + \varepsilon \text{ and } \|y^* + z^*\|_* \leq r\}.$$

Then  $K_{\varepsilon r}$  is a closed set in the weak\* topology. Setting  $T(y^*, z^*) = y^* + z^*$ , we have that

$$C_\varepsilon \cap rB_{X^*} = T(K_{\varepsilon r}).$$

Since  $T$  is continuous from  $X^* \times X^*$  to  $X^*$  (with the weak\* topologies), if we show that  $K_{\varepsilon r}$  is bounded (hence weak\* compact), then  $C_\varepsilon \cap rB_{X^*}$  is a weak\* compact set, for all  $r > 0$ . The Banach–Dieudonné–Krein–Smulian theorem then guarantees that  $C_\varepsilon$  is weak\* closed (See Theorem A.2.1 in Appendix B). Let us then show that  $K_{\varepsilon r}$  is bounded. To do this, we use the uniform boundedness theorem. Thus, it is enough to show that  $\forall y, z \in X$ , there is a constant  $C = C(y, z)$  such that

$$|\langle (y^*, z^*), (y, z) \rangle| = |\langle y^*, y \rangle + \langle z^*, z \rangle| \leq C, \forall (y^*, z^*) \in K_{\varepsilon r}.$$

By assumption there is  $t \geq 0$  such that  $y - z = t(u - v)$  with  $u \in \text{dom } f$  and  $v \in \text{dom } g$ . Then

$$\begin{aligned} |\langle y^*, y \rangle + \langle z^*, z \rangle| &= |t\langle y^*, u \rangle + t\langle z^*, v \rangle + \langle y^* + z^*, z - tv \rangle| \\ &\leq |t(f(u) + f^*(y^*) + g(v) + g^*(z^*))| + r\|z - tv\| \\ &\leq |t(a + \varepsilon + f(u) + g(v))| + r\|z - tv\| = C(y, z). \end{aligned}$$

The claim is proved in the case when  $F = X$ . Let us now turn to the general case. Suppose  $u \in \text{dom } f - \text{dom } g$ . Then  $-u \in F$  and so there are  $t \geq 0$  and  $v \in \text{dom } f - \text{dom } g$  such that  $-u = tv$ . It follows that

$$0 = \frac{1}{1+t}u + \frac{t}{1+t}v \in \text{dom } f - \text{dom } g.$$

Hence  $\text{dom } f \cap \text{dom } g \neq \emptyset$  and after a suitable translation, we can suppose that  $0 \in \text{dom } f \cap \text{dom } g$ , whence  $\text{dom } f \subset F$ ,  $\text{dom } g \subset F$ . Let  $i: F \rightarrow X$  be the canonical injection of  $F$  in  $X$  and let  $i^*: X^* \rightarrow F^*$  be its adjoint operator:  $\langle i^*(x^*), d \rangle = \langle x^*, i(d) \rangle$ . Let us consider the functions

$$\tilde{f}: F \rightarrow (-\infty, \infty], \tilde{f} := f \circ i, \quad \tilde{g}: F \rightarrow (-\infty, \infty], \tilde{g} := g \circ i.$$

We can apply the first step of the proof to them. We have

$$(\tilde{f} + \tilde{g})^*(z^*) = (\tilde{f}^* \nabla \tilde{g}^*)(z^*),$$

for all  $z^* \in F^*$ . It is now easy to verify that if  $x^* \in X^*$ ,

$$\begin{aligned} f^*(x^*) &= \tilde{f}^*(i^*(x^*)), & g^*(x^*) &= \tilde{g}^*(i^*(x^*)), \\ (f + g)^*(x^*) &= (\tilde{f} + \tilde{g})^*(i^*(x^*)), & (f^* \nabla g^*)(x^*) &= (\tilde{f}^* \nabla \tilde{g}^*)(i^*(x^*)), \end{aligned}$$

(in the last one we use that  $i^*$  is onto).

For instance, we have

$$\begin{aligned} \tilde{f}^*(i^*(x^*)) &= \sup_{z \in F} \{ \langle i^*(x^*), z \rangle - \tilde{f}(z) \} = \sup_{z \in F} \{ \langle x^*, i(z) \rangle - f(i(z)) \} \\ &= \sup_{x \in X} \{ \langle x^*, x \rangle - f(x) \}, \end{aligned}$$

where the last inequality holds as  $\text{dom } f \subset F$ . The others follow in the same way. Finally, the exactness at a point  $x^* \in \text{dom } f^* \nabla g^*$  follows from the compactness, previously shown, of  $K_{\varepsilon, \|x^*\|_*}$ , with  $a = (f^* \nabla g^*)(x^*)$  and  $\varepsilon > 0$  arbitrary. This allows us to conclude.  $\square$

Besides its intrinsic interest, the previous theorem yields the following sum rule for the subdifferentials which generalizes Theorem 3.4.2.

**Theorem 5.4.3** *Let  $f, g \in \Gamma(X)$ . Moreover, let*

$$F := \mathbb{R}_+(\text{dom } f - \text{dom } g)$$

*be a closed vector space. Then*

$$\partial(f + g) = \partial f + \partial g.$$

*Proof.* Let  $x^* \in \partial(f + g)(x)$ . We must find  $y^* \in \partial f(x)$  and  $z^* \in \partial g(x)$  such that  $y^* + z^* = x^*$ . By the previous result there are  $y^*, z^*$  such that  $y^* + z^* = x^*$  and fulfilling  $f^*(y^*) + g^*(z^*) = (f + g)^*(x^*)$ . As  $x^* \in \partial(f + g)(x)$  we have (Proposition 5.3.1)

$$\begin{aligned} \langle y^*, x \rangle + \langle z^*, x \rangle &= \langle x^*, x \rangle = (f + g)(x) + (f + g)^*(x^*) \\ &= f(x) + f^*(y^*) + g(x) + g^*(z^*). \end{aligned}$$

This implies (why?)

$$\langle y^*, x \rangle = f(x) + f^*(y^*) \text{ and } \langle z^*, x \rangle = g(x) + g^*(z^*),$$

and we conclude.  $\square$

The previous generalization is useful, for instance, in the following situation: suppose we have a Banach space  $Y$ , a (proper) closed subspace  $X$  and two continuous functions  $f, g \in \Gamma(X)$  fulfilling the condition  $\text{int dom } f \cap \text{dom } g \neq \emptyset$ . It can be useful sometimes to consider the natural extensions  $\tilde{f}, \tilde{g} \in \Gamma(Y)$  of  $f$  and  $g$  (by defining them  $\infty$  outside  $X$ ). In such a case the previous theorem can be applied, while Theorem 3.4.2 obviously cannot.

**Exercise 5.4.4** Let

$$f(x, y) = \begin{cases} -\sqrt{xy} & \text{if } x \leq 0, y \leq 0, \\ \infty & \text{otherwise,} \end{cases}$$

$$g(x, y) = \begin{cases} -\sqrt{-xy} & \text{if } x \geq 0, y \leq 0, \\ \infty & \text{otherwise.} \end{cases}$$

Find  $(f + g)^*$  and  $f^* \nabla g^*$ .

**Exercise 5.4.5** Given a nonempty closed convex set  $K$ ,

$$d^*(\cdot, K) = \sigma_K + I_{B^*}.$$

*Hint.* Remember that  $d(\cdot, K) = (\|\nabla I_K)(\cdot)$  and apply Proposition 5.4.1.

**Exercise 5.4.6** Let  $X$  be a reflexive Banach space. Let  $f, g \in \Gamma(X)$ . Let

$$\lim_{\|x\| \rightarrow \infty} \frac{f(x)}{\|x\|} = \infty.$$

Then  $(f \nabla g) \in \Gamma(X)$ .

*Hint.* Try to apply the Attouch–Brézis theorem to  $f^*, g^*$ .

## 5.5 Sandwiching an affine function between a convex and a concave function

In this section we deal with the following problem: suppose we are given a Banach space  $X$  and two convex, lower semicontinuous extended real valued functions  $f$  and  $g$  such that  $f(x) \geq -g(x) \forall x \in X$ . The question is: when is it possible to find an affine function  $m$  with the property that

$$f(x) \geq m(x) \geq -g(x),$$

for all  $x \in X$ ? It is clear that the problem can be restated in an equivalent, more geometric, way: suppose we can separate the sets  $\text{epi } f$  and  $\text{hyp}(-g)$  with a nonvertical hyperplane. With a standard argument this provides the affine function we are looking for. And, clearly, the condition  $f \geq -g$  gives some hope to be able to make such a separation.

In order to study the problem, let us first observe the following simple fact.

**Proposition 5.5.1** *Let  $y^* \in X^*$ . Then  $y^* \in \{p : f^*(p) + g^*(-p) \leq 0\}$  if and only if there exists  $a \in \mathbb{R}$  such that*

$$f(x) \geq \langle y^*, x \rangle + a \geq -g(x),$$

for all  $x \in X$ .

*Proof.* Suppose  $f^*(y^*) + g^*(-y^*) \leq 0$ . Then, for all  $x \in X$ ,

$$\langle y^*, x \rangle - f(x) + g^*(-y^*) \leq 0,$$

i.e.,

$$f(x) \geq \langle y^*, x \rangle + a,$$

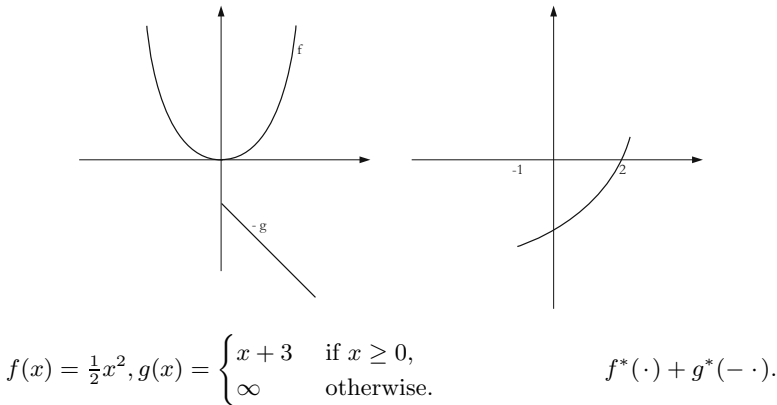
with  $a = g^*(-y^*)$ . Moreover

$$a = g^*(-y^*) \geq \langle -y^*, x \rangle - g(x),$$

for all  $x \in X$ , implying  $\langle y^*, x \rangle + a \geq -g(x)$ , for all  $x \in X$ . Conversely, if  $f(x) \geq \langle y^*, x \rangle + a$  and  $\langle y^*, x \rangle + a \geq -g(x)$  for all  $x$ , then

$$-a \geq f^*(y^*), \quad a \geq \langle -y^*, x \rangle - g(x),$$

for all  $x$ , implying  $f^*(y^*) + g^*(-y^*) \leq 0$ . □



**Figure 5.6.**

It follows in particular that the set of the “slopes” of the affine functions sandwiched between  $f$  and  $-g$  is a weak\* closed and convex set, as it is the zero level set of the function  $h(\cdot) = f^*(\cdot) + g^*(-\cdot)$ . Now, observe that  $\inf_x (f + g)(x) \geq 0$  if and only if  $(f + g)^*(0^*) \leq 0$ . Thus, if



$$(f + g)^*(0^*) = (f^* \nabla g^*)(0^*)$$

and the epi-sum is exact, then  $\inf_x (f + g)(x) \geq 0$  is equivalent to saying that there exists  $y^* \in X^*$  such that

$$(f^* \nabla g^*)(0^*) = f^*(y^*) + g^*(-y^*) \leq 0.$$

Thus a sufficient condition to have an affine function sandwiched between  $f$  and  $-g$  is that the assumption of the Attouch–Brezis theorem be satisfied.

Now we specialize to the case when  $X$  is a Euclidean space. In this case the condition  $f \geq -g$  implies that

$$\text{ri epi } f \cap \text{ri hyp}(-g) = \emptyset.$$

Then we can apply Theorem A.1.13 to separate the sets  $\text{epi } f$  and  $\text{hyp}(-g)$ . However, this does not solve the problem, as it can happen that the separating hyperplane is vertical. So, let us now see a sufficient condition in order to assure that the separating hyperplane is not vertical, which amounts to saying that the affine function we are looking for is finally singled out.

**Proposition 5.5.2** *Suppose*

$$\text{ri dom } f \cap \text{ri dom}(-g) \neq \emptyset.$$

*Then there exists  $y^*$  such that  $f^*(y^*) + g^*(-y^*) \leq 0$ .*

*Proof.* Let us use the Attouch–Brezis theorem, as suggested at the beginning of the section. Thus, we must show that

$$F := \mathbb{R}_+(\text{dom } f - \text{dom } g)$$

is a subspace. As is suggested in the next exercise, it is enough to show that if  $x \in F$ , then  $-x \in F$ . We can suppose, without loss of generality, that  $0 \in \text{ri dom } f \cap \text{ri dom } g$ . As  $x \in F$ , there are  $l > 0$ ,  $u \in \text{dom } f$  and  $v \in \text{dom } g$  such that  $x = l(u - v)$ . As  $0 \in \text{ri dom } f \cap \text{ri dom } g$ , there is  $c > 0$  small enough such that  $-cu \in \text{dom } f$ ,  $-cv \in \text{dom } g$ . Thus  $-cu - (-cv) \in \text{dom } f - \text{dom } g$ . Then

$$\frac{l}{c}(-cu - (-cv)) = -x \in F.$$

□

**Exercise 5.5.3** Let  $A$  be a convex set containing zero. Then  $\bigcup_{\lambda > 0} \lambda A$  is a convex cone. Moreover, if  $x \in \bigcup_{\lambda > 0} \lambda A$  implies  $-x \in \bigcup_{\lambda > 0} \lambda A$ , then  $\bigcup_{\lambda > 0} \lambda A$  is a subspace.

*Hint.* Call  $F = \bigcup_{\lambda > 0} \lambda A$ . It has to be shown that  $x, y \in F$  implies  $x + y \in F$ . There are positive  $l_1, l_2$  and  $u, v \in A$  such that  $x = l_1 u$ ,  $y = l_2 v$ . Then  $x/l_1 \in A$ ,  $y/l_2 \in A$  and  $\frac{1}{l_1 + l_2}(x + y)$  is a convex combination of  $x/l_1$  and  $y/l_2$ .

We now give some pretty examples showing that the affine function separating  $\text{epi } f$  and  $\text{hyp}(-g)$  need not exist, unless some extra condition is imposed.

**Example 5.5.4**

$$f(x) = \begin{cases} -\sqrt{x} & \text{if } x \geq 0, \\ \infty & \text{otherwise,} \end{cases}$$

$$g(x) = \begin{cases} 0 & \text{if } x = 0, \\ \infty & \text{otherwise.} \end{cases}$$

Here  $\inf(f + g) = 0$ , and  $\text{ri}(\text{dom } f) \cap \text{ri}(\text{dom } g) = \emptyset$ .

**Example 5.5.5**

$$f(u, v) = \begin{cases} -1 & \text{if } uv \geq 1, u \geq 0, \\ \infty & \text{otherwise,} \end{cases}$$

$$g(u, v) = \begin{cases} 0 & \text{if } u \geq 0, v = 0, \\ \infty & \text{otherwise.} \end{cases}$$

Here we have  $\text{dom } f \cap \text{dom } g = \emptyset$ .

**Example 5.5.6**

$$f(u, v) = \begin{cases} u & \text{if } v = -1, \\ \infty & \text{otherwise,} \end{cases}$$

$$g(u, v) = \begin{cases} 0 & \text{if } v = 0, \\ \infty & \text{otherwise.} \end{cases}$$

The Example 5.5.4 can induce the idea that the separator must be vertical as the two effective domains do intersect at a point. So, it could be argued that, if the two domain are far apart, the property could hold. But in Example 5.5.6 the distance between  $\text{dom } f$  and  $\text{dom } g$  is 1.

In the last two examples the domains of  $f$  and  $g$  do not intersect, while in the first example a crucial role is played by the fact that  $\inf(f + g) = 0$ . In the following example  $\inf(f + g) > 0$ , and yet there is no affine separator. Observe that such example could not be provided in one dimension (see Remark 2.2.15).

**Example 5.5.7**

$$f(u, v) = \begin{cases} 1 - 2\sqrt{uv} & \text{if } u, v \geq 0, \\ \infty & \text{otherwise,} \end{cases}$$

$$g(u, v) = \begin{cases} 1 - 2\sqrt{-uv} & \text{if } u \leq 0, v \geq 0, \\ \infty & \text{otherwise.} \end{cases}$$

A straightforward calculation shows

$$f^*(u^*, v^*) = \begin{cases} -1 & \text{if } u^* \leq 0, u^*v^* \geq 1, \\ \infty & \text{otherwise,} \end{cases}$$
$$g^*(u^*, v^*) = \begin{cases} -1 & \text{if } u^* \geq 0, u^*v^* \leq -1, \\ \infty & \text{otherwise.} \end{cases}$$

Our finite dimensional argument actually holds, without any changes in the proof, provided we assume that at least one of the sets  $\text{epi } f$ ,  $\text{hyp}(-g)$  has an interior point. In particular, the assumption in Proposition 5.5.2 becomes, in infinite dimensions,  $\text{int dom } f \cap \text{dom } g \neq \emptyset$ . To conclude, let me mention that this section is inspired by my work with Lewis [LeL], where we studied the more general problem of giving sufficient conditions under which the slope of the affine function between  $f$  and  $-g$  is in the range (or in the closure of the range) of the Clarke subdifferential of a locally Lipschitz function  $h$  such that  $f \geq h \geq -g$ .