

**CONVEX FUNCTIONS ON NON-CONVEX DOMAINS**

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It is shown that a convex function, defined on an arbitrary, possibly finite, subset of a linear space, can be extended to the whole space. An application to decision making under risk is given.

**1. Introduction**

The definitions of convexity of a function, customary in literature, assume or imply convexity of the domain. In applications however, observations about the values of a function, and inequalities verified by these, are often available only on non-convex, e.g. finite, domains.

This paper considers a way to define convexity on non-convex domains. The main theorem, in section 2, shows that a function is convex, if and only if it is the restriction of a convex function defined on the whole space. Section 3 gives an application to decision making under risk, and section 4 concludes.

**2. The extension theorem**

Let  $V$  be a linear space over the reals. Let  $T$  be some arbitrary subset of  $V$ , and  $f$  a function from  $T$  to  $\mathbb{R} \cup \{-\infty, \infty\}$ . The following definition adapts the definitions of convexity, given in literature only for convex sets  $T$  [see for instance Rockafellar (1970, sect. 4)], to general, possibly finite, sets  $T$ . For compact sets it was given in Peters and Tijs (1981), and for general sets in Wakker, Peters and Van Riel (1985).

*Definition 1.* The function is *convex* if for all convex combinations  $\sum_{j=1}^n p_j x^j$  of elements  $x^j$  of  $T$ , for which not both  $-\infty$  and  $+\infty$  are contained in  $\{f(x^j)\}_{j=1}^n$ , we have

$$\sum_{j=1}^n p_j f(x^j) \geq f\left(\sum_{j=1}^n p_j x^j\right), \quad (1)$$

whenever  $\sum p_j x^j$  is in  $T$ .

As usual we take:  $\lambda \infty := \infty$  for  $\lambda \in \mathbb{R}_{++}$ ,  $\lambda \infty := 0$  for  $\lambda = 0$ ,  $\lambda \infty := -\infty$  for  $\lambda \in \mathbb{R}_{--}$ ;  $\lambda + \infty := \infty$  for  $\lambda \in \mathbb{R}$  or  $\lambda = \infty$ , and  $\lambda + \infty$  is undefined for  $\lambda = -\infty$ ;  $\lambda(-\infty) := -\infty$  for  $\lambda \in \mathbb{R}_{++}$ ,  $\lambda(-\infty) := 0$  for  $\lambda = 0$ ,  $\lambda(-\infty) := \infty$  for  $\lambda \in \mathbb{R}_{--}$ ;  $\lambda - \infty := -\infty$  for  $\lambda \in \mathbb{R}$  or  $\lambda = -\infty$ , and  $\lambda - \infty$  is left undefined for  $\lambda = \infty$ .

A function  $f$  is *concave* if  $-f$  is convex. All results, derived in the sequel for convex functions  $f$ , can be reformulated for concave functions  $g$ , by setting  $g := -f$ .

*Theorem 1 (extension theorem).* Let  $V$  be a linear space over  $\mathbb{R}$ ; let  $T \subset V$ . Let  $f: T \rightarrow \mathbb{R} \cup \{-\infty, \infty\}$  be convex. Then there exists a convex function  $\tilde{f}: V \rightarrow \mathbb{R} \cup \{-\infty, \infty\}$  which extends  $f$ .

*Proof.* We define  $\tilde{f}$  on  $V$  as follows:

$$\tilde{f}: x \mapsto \inf\{\mu : \mu = \sum_{j=1}^n p_j f(x^j), x \text{ is a convex combination } \sum p_j x^j \text{ of elements } x^j \text{ of } T \text{ such that not } f(x^j) = \infty \text{ for some } j, \text{ and } f(x^j) = -\infty \text{ for some other } j\}.$$
 (2)

Note that  $\tilde{f}(x) \leq f(x)$  for all  $x \in T$ ; by convexity of  $f$ ,  $\tilde{f}(x) \geq f(x)$  for all  $x \in T$ . So indeed  $\tilde{f}$  extends  $f$ . Convexity of  $\tilde{f}$  remains to be demonstrated. So let  $y$  be a convex combination  $\sum_{i=1}^m q_i y^i$ , with all  $q_i > 0$ . We have to prove

$$\tilde{f}(y) \leq \sum_{i=1}^m q_i \tilde{f}(y^i)$$
 (3)

whenever not  $\tilde{f}(y^i) = \infty$  for some  $i$ ,  $\tilde{f}(y^i) = -\infty$  for some other  $i$ .

The case where  $\tilde{f}(y^i) = \infty$  for some  $i$  is immediate anyhow, so we suppose  $\tilde{f}(y^i) < \infty$  for all  $i$ . This implies that every  $y^i$  is in  $\text{conv}(T)$ . Say  $y^i$  is the convex combination  $\sum_k p_{ik} x^{ik}$  of elements  $x^{ik}$  of  $T$ . Then  $y = \sum_{i=1}^m \sum_k q_i p_{ik} x^{ik}$ . By definition of  $\tilde{f}$ ,

$$\tilde{f}(y) \leq \sum_{i=1}^m q_i \sum_k p_{ik} f(x^{ik}).$$
 (4)

Now first suppose  $\tilde{f}(y^l) = -\infty$ , for some  $l$ . Then, for every  $M \in \mathbb{N}$ , we can take the  $x^{lk}$  above such that  $\sum_k p_{lk} f(x^{lk}) < -M$ . Since  $q_l > 0$ , by (4) we get  $\tilde{f}(y) = -\infty$ , and (3) follows.

Next suppose  $\tilde{f}(y^i) > -\infty$  for all  $i$ . Then, for any  $\epsilon > 0$  and every  $i$ , we can take the  $x^{ik}$  such that  $\sum_k p_{ik} f(x^{ik}) \leq \tilde{f}(y^i) + \epsilon$ . This again implies (3). Q.E.D.

Note that the  $\tilde{f}$  defined in (2) is the maximal convex extension. If  $f$  in Theorem 1 is real-valued, then even on  $\text{conv}(T)$ , it may be impossible to have  $\tilde{f}$  real-valued, as the following example shows.

*Example 1.* Let  $V = \mathbb{R}^2$ ,  $T = \{(-1, -j)\}_{j \in \mathbb{N}} \cup \{(1, j)\}_{j \in \mathbb{N}}$ ,  $f(x_1, x_2) = -|x_2|$  for all  $(x_1, x_2) \in T$ . Then considering the pair  $(-1, -j)$  and  $(1, j)$ , one sees that  $\tilde{f}(0, 0) \leq -j$  must hold, for all  $j \in \mathbb{N}$ .

For bounded  $f$ ,  $\tilde{f}$  can be taken real-valued on  $\text{conv}(T)$ :

*Corollary 1.* For the  $\tilde{f}$ , defined in (2),  $\sup \tilde{f}(\text{conv}(T)) = \sup(f(T))$ , and  $\inf \tilde{f}(\text{conv}(T)) = \inf f(T)$ .

*Proof.* Obvious from (2).

If  $V = \mathbb{R}$ , then a real-valued  $f$  in Theorem 1 can be extended to a real-valued  $\tilde{f}$  on  $\text{conv}(T)$ .

*Corollary 2.* Let  $V = \mathbb{R}$ ,  $T \subset V$ ,  $f: T \rightarrow \mathbb{R}$  convex. Then  $f$  has a convex extension  $\tilde{f}: \text{conv}(T) \rightarrow \mathbb{R}$ .

*Proof.* Obvious if  $T$  contains no more than two elements, then  $\tilde{f}$  can be taken affine (i.e., both convex and concave). So let  $x^1 > x^2 > x^3$  in  $T$ . Let  $l^1$  be the affine function through  $(x^1, f(x^1))$  and  $(x^2, f(x^2))$ ,  $l^2$  be the one through  $(x^2, f(x^2))$  and  $(x^3, f(x^3))$ . Then, with  $\tilde{f}$  as in (2), for all  $x \in \mathbb{R}$  we have  $f(x) \geq \min\{l^1(x), l^2(x)\} > -\infty$ .

Further, for all  $x \in \text{conv}(T)$ ,  $x \in [x^4, x^5]$  for some  $x^4, x^5$  in  $T$ . Hence  $\tilde{f}(x) \leq \max\{f(x^4), f(x^5)\} < \infty$  for such  $x$ . Q.E.D.

The following example shows that, even if  $V = \mathbb{R}$ , and  $f$  is real valued and bounded on  $T$ , then still no convex real-valued extension  $\tilde{f}$  of  $f$  to all of  $V$  may exist.

*Example 2.* Let  $T = [0, 1]$ ,  $f: x \mapsto -\sqrt{x}$  on  $T$ . Then any convex extension  $\tilde{f}$  of  $f$  on  $V$  can be seen to assign  $\infty$  to all of  $\mathbb{R} \setminus T$ .

*Lemma 1.* Let  $V = \mathbb{R}$ , and  $f$  non-decreasing and convex. Then  $\tilde{f}$ , as defined by (2), is non-decreasing on  $\text{conv}(T)$ .

*Proof.* By Corollary 1,  $\tilde{f}$  has the same infimum on  $\text{conv}(T)$ , as  $f$  on  $T$ . Hence the infimum of  $\tilde{f}$  can be found on the ‘left side’ of  $\text{conv}(T)$ : the convex  $\tilde{f}$  must be non-decreasing. Q.E.D.

### 3. An application to risk aversion

Let  $\mathcal{C}$  be a non-empty set of consequences.  $\mathcal{L}^s(\mathcal{C})$  is the set of (simple) lotteries on  $\mathcal{C}$ ; a (simple) lottery on  $\mathcal{C}$  is a probability measure on  $(\mathcal{C}, 2^{\mathcal{C}})$ , assigning probability one to a finite subset of  $\mathcal{C}$ . By  $(p_j; x^j)_{j=1}^n$  we denote the simple lottery, assigning probability  $p_j$  to every  $x^j$ . Implicit in this is that  $x^j \in \mathcal{C}$  for all  $j$ ,  $p_j \geq 0$  for all  $j$ , and  $\sum p_j = 1$ . For any  $\alpha \in \mathcal{C}$ ,  $\bar{\alpha}$  denotes  $(1; \alpha)$ , i.e., the lottery which with probability one results in consequence  $\alpha$ .

We assume there are two decision makers (persons, players, etc.)  $T^k$ ,  $k = 1, 2$ , with preference relations  $\succeq^k$ , i.e., binary relations on  $\mathcal{L}^s(\mathcal{C})$ . The interpretation of  $l \succeq^k l'$  is that  $T^k$ , when having to choose one element from  $\{l, l'\}$ , is willing to choose  $l$ .

We assume that there exist von Neumann–Morgenstern (vNM) utility functions  $U^1, U^2$ , i.e., for  $k = 1, 2$ ,  $U^k: \mathcal{C} \rightarrow \mathbb{R}$  is such that

$$(p_j; x^j)_{j=1}^n \succeq (q_i; y^i)_{i=1}^m \Leftrightarrow \sum_{j=1}^n p_j U^k(x^j) \geq \sum_{i=1}^m q_i U^k(y^i). \tag{5}$$

Here  $\sum p_j U^k(x^j)$  is the expected utility of  $(p_j; x^j)_{j=1}^n$ .

The following definition gives a way to compare  $T^1$  and  $T^2$  with respect to their ‘risk aversion’. It has been introduced in Yaari (1969, p. 316, in terms of so-called ‘acceptance sets’), and is a variation on earlier definitions of Pratt (1964) and Arrow (1971).

*Definition 2.*  $T^1$  is more risk averse (MRA) than  $T^2$  if

$$(p_j; sx^j)_{j=1}^n \succeq^1 \bar{\alpha} \Rightarrow (p_j; x^j)_{j=1}^n \succeq^2 \bar{\alpha}. \tag{6}$$

The above definition applies if  $T^2$  is willing to take a risky lottery  $(p_j; x^j)_{j=1}^n$  instead of a certain consequence  $\alpha$ , whenever  $T^1$  is willing to do so.

Definition 2, and the results to be derived below, have first been studied in literature for the case  $\mathcal{C} = \mathbb{R}$ . Later Kihlstrom and Mirman (1974) extended these to the case  $\mathcal{C} = \mathbb{R}^n_+$ . All this was done under differentiability and monotonicity assumptions about the vNM utility functions. In Wakker, Peters and Van Riel (1985, Theorem 3.1.b) the results were extended to arbitrary spaces  $\mathcal{C}$ , without any restriction on the vNM utility functions. Motivation for this was, firstly, that differentiability of the utility functions does not always have a clear behavioural meaning; secondly, that it seems desirable for applications to be able to handle cases where one has information about the choices of decision makers, only with respect to finitely many consequences; and thirdly, that it seems desirable to be able to handle cases where one does not (yet) have a quantification of the consequences.

The following theorem characterizes the ‘more risk averse than’ relation.

*Theorem 2. The following two statements are equivalent:*

- (i)  $T^1$  is more risk averse than  $T^2$ .
- (ii)  $U^2 = \psi \circ U^1$  for a convex non-decreasing function  $\psi : U^1(\mathcal{C}) \rightarrow U^2(\mathcal{C})$ .

*Proof.* First suppose (i). Then  $\bar{\beta} \succcurlyeq^1 \bar{\alpha} \Rightarrow \bar{\beta} \succcurlyeq^2 \bar{\alpha}$ , so  $U^1(\beta) \geq U^1(\alpha) \Rightarrow U^2(\beta) \geq U^2(\alpha)$ . So there must exist a non-decreasing  $\psi : U^1(\mathcal{C}) \rightarrow U^2(\mathcal{C})$  such that  $U^2 = \psi \circ U^1$ . Now let some element  $U^1(\mu)$  of  $U^1(\mathcal{C})$  be a convex combination  $\sum_{j=1}^n p_j U^1(\mu^j)$  of other elements of  $U^1(\mathcal{C})$ . Then  $(p_j; \mu^j)_{j=1}^n \succcurlyeq^1 \bar{\mu}$ , so by (i):  $(\beta_j; \mu^j)_{j=1}^n \succcurlyeq^2 \bar{\mu}$ , i.e.,  $\sum p_j U^2(\mu^j) \geq U^2(\mu)$ . Substituting  $\psi$  gives

$$\sum p_j \psi(U^1(\mu^j)) \geq [\psi(U^1(\mu)) =] \psi(\sum p_j U^1(\mu^j)), \tag{7}$$

so  $\psi$  is convex on  $U^1(\mathcal{C})$ , and (ii) is proved.

Next suppose (ii). Let  $(p_j; \mu^j)_{j=1}^n \succcurlyeq^1 \bar{\mu}$ , so  $\sum_{j=1}^n p_j U^1(\mu^j) \geq U^1(\mu)$ . Let  $\tilde{\psi}$  be as defined in (2); by Lemma 1,  $\tilde{\psi}$  is non-decreasing on  $\text{conv}(T)$ . We get  $\sum_{j=1}^n p_j U^2(\mu^j) = \sum p_j \psi(U^1(\mu^j)) \geq \tilde{\psi}(\sum p_j U^1(\mu^j)) \geq \psi(U^1(\mu)) = U^2(\mu)$ . This implies  $(p_j; \mu^j)_{j=1}^n \succcurlyeq^2 \bar{\mu}$ , which is what (i) requires. Q.E.D.

Theorem 2 shows a natural occurrence of convex functions on non-convex domains: if observations of choices of  $T^1$  and  $T^2$  are available only with respect to consequences from  $\mathcal{C}$ , then (inequalities concerning)  $\psi$  will only be observed on  $U^1(\mathcal{C})$ , which may be any arbitrary subset of  $\mathbb{R}$ . In the last occurrence of  $\tilde{\psi}$  in the proof we essentially use the extension theorem: the argument of  $\tilde{\psi}$  there does not have to be in the domain of  $\psi$ . Because of this complication, in Wakker, Peters and Van Riel (1985) a more complicated proof (without the extension theorem available) was given, using first-order-difference results derived in the appendix there.

#### 4. Conclusion

Convexity and concavity have proved to be fruitful concepts in economics. Therefore, it seems to be worthwhile to define convexity and concavity of a function also on non-convex domains, since these may occur in economics in a natural way. The main theorem presented in this paper, which extends a convex function to at least the convex hull of its domain, enables one to apply results known for convex functions on convex domains to such functions on non-convex domains. An example of an application, to decision making under risk, has been given.

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