

# MAXIMALITY OF SUMS OF TWO MAXIMAL MONOTONE OPERATORS IN GENERAL BANACH SPACE

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ABSTRACT. We combine methods from convex analysis, based on a function of Simon Fitzpatrick, with a fine recent idea due to Voisei, to prove maximality of the sum of two maximal monotone operators in Banach space under various natural transversality conditions.

## 1. INTRODUCTION AND PRELIMINARIES

The results of this paper, especially Theorem 9, marry recent work by Voisei [12] with additional convex analysis described in [1, 2], see also [4, §5.1] or [3, §8.3], to provide an accessible short proof of the maximality of the sum of two maximal monotone operators under domain conditions such as  $D(B) \cap \text{core}D(A) \neq \emptyset$ , while either  $D(B)$  is closed and convex or  $\text{core} \text{conv}D(B) \neq \emptyset$ .

Recall that the *domain* of an extended-valued convex function,  $\text{dom}(f)$ , is the set of points with value less than  $+\infty$ , and that a point  $s$  is in the *core* of a set  $S$  (denoted by  $s \in \text{core} S$ ) provided that  $s$  lies in  $S$  and  $X = \bigcup_{\lambda>0} \lambda(S - s)$ . Recall that  $x^* \in X^*$  is a *subgradient* of  $f : X \rightarrow (-\infty, +\infty]$  at  $x \in \text{dom} f$  provided that  $f(y) - f(x) \geq \langle x^*, y - x \rangle$ . The set of subgradients of  $f$  at  $x$  is the *subderivative* or *subdifferential* of  $f$  at  $x$  and is denoted  $\partial f(x)$ .

We shall need the *indicator* function  $\iota_C(x)$  which is zero for  $x$  in  $C$  and  $+\infty$  otherwise, the *Fenchel conjugate*  $f^*(x^*) := \sup_x \{\langle x, x^* \rangle - f(x)\}$  and the *infimal convolution*  $f \square g(x) := \inf\{f(y) + g(z) : x = y + z\}$ . The central examples of the *normal cone* to  $C$  at  $x$  and the *distance function*  $d_C$ , are covered by  $N_C(x) = \partial \iota_C$  and  $d_C = \iota_C \square \|\cdot\|$ .

We say a multifunction  $T : X \mapsto 2^{X^*}$  is *monotone* provided that for any  $x, y \in X$ ,  $x^* \in T(x)$  and  $y^* \in T(y)$ ,

$$\langle y^* - x^*, y - x \rangle \geq 0,$$

and we say that  $T$  is *maximal monotone* if its graph is not properly included in any other monotone graph. The subdifferential of a convex lower semicontinuous (lsc) function on a Banach space is a fine example of a maximal monotone multifunction (see [3, 4, 10] wherein other notation and usage may be also followed up).

## 2. REPRESENTATIVE FUNCTIONS

For any monotone mapping  $T$ , we associate the *Fitzpatrick function* introduced by Simon Fitzpatrick in [6] but then neglected for many years until re-popularized

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in papers by Penot [8], Buracik-Svaiter [5], and others. Some more of the related history may be found in [2]. *Fitzpatrick's function* is

$$\mathcal{F}_T(x, x^*) := \sup\{\langle x, y^* \rangle + \langle x^*, y \rangle - \langle y, y^* \rangle : y^* \in T(y), y \in \text{dom } T\},$$

which is clearly lower semicontinuous and convex as an affine supremum.

**Proposition 1.** [6, 4] *For a maximal monotone operator  $T$*

$$\mathcal{F}_T(x, x^*) \geq \langle x, x^* \rangle$$

*with equality if and only if  $x^* \in T(x)$ .*

Correspondingly *Penot's function* is given as the (closed) convexification

$$\mathcal{P}_T(x, x^*) := \inf \left\{ \sum_{i=1}^N \lambda_i \langle x_i, x_i^* \rangle : \sum_i \lambda_i \langle x_i, x_i^* \rangle = \langle x, x^* \rangle, x_i^* \in T(x_i), \sum \lambda_i = 1, \lambda_i \geq 0 \right\}$$

It is easy to see that  $\mathcal{P}_T$  is convex and that, with the appropriate ordering of variables  $x$  and  $x^*$  (and the conjugate restricted to  $X \times X^*$ ) we have

$$\mathcal{P}_T^* = \mathcal{F}_T \quad \text{while} \quad \mathcal{F}_T^* = \mathcal{P}_T^{**} = \overline{\mathcal{P}_T},$$

where  $\overline{\mathcal{P}_T}$  is the lower-semicontinuous hull of  $\mathcal{P}_T$ . Note that

More generally, we say that a lower-semicontinuous convex function  $\mathcal{H}_T$  *represents* a monotone operator  $T$  if

$$\mathcal{H}_T(x, x^*) \geq \langle x, x^* \rangle$$

with equality when  $x^* \in T(x)$ . We say a representative is *exact* if  $\mathcal{H}_T(x, x^*) = \langle x, x^* \rangle$  exactly on the graph of  $T$ . Now we may check that:

**Proposition 2.** ([2, 8]) *Let  $T$  be monotone on a Banach space  $X$ . Then*

- i.) *Penot's function  $\overline{\mathcal{P}_T}$  represents  $T$ .*
- ii.) *If  $\mathcal{H}_T$  represents  $T$ , then  $\mathcal{H}_T \leq \overline{\mathcal{P}_T}$  pointwise.*
- iii.) *If  $T$  is maximal then  $\mathcal{F}_T \leq \mathcal{H}_T \leq \overline{\mathcal{P}_T}$ .*
- iv.)  *$\mathcal{F}_T(x, x^*) \leq \langle x, x^* \rangle$  iff  $(x, x^*)$  is monotonically related to the graph of  $T$ .*
- v.) *Suppose  $\mathcal{F}_T$  represents  $T$ . Then  $\mathcal{F}_T(x, x^*) = \langle x, x^* \rangle$  iff  $\overline{\mathcal{P}_T}(x, x^*) = \langle x, x^* \rangle$ .*

**Proof.** (i.) is an easy computation performed in [2, 8]. (ii.) a direct consequence of  $\overline{\mathcal{P}_T} = (c_T)^{**}$  and that  $\mathcal{H}_T(x, x^*) \leq c_T$ , where  $c_T(x, x^*) := \langle x, x^* \rangle + \iota_{\text{Gr}(T)}(x, x^*)$ . (iii.) The lefthand inequality is established in [6, 8]. (iv.) is a direct computation. (v.) By (iv.)—as  $\mathcal{F}_T$  is representative—we need only show the ‘if’. We observe that if  $\mathcal{F}_T(x, x^*) = \langle x, x^* \rangle$ , then minorizing  $\mathcal{F}_T(x + t(y - x), x^* + t(y^* - x^*))$  by  $\langle x + t(y - x), x^* + t(y^* - x^*) \rangle$  we have

$$\mathcal{F}_T(y, y^*) - \mathcal{F}_T(x, x^*) \geq d^+ \mathcal{F}_T((x, x^*); (y - x, y^* - x^*)) \geq \langle x, y^* - x^* \rangle + \langle y - x, x^* \rangle$$

for all  $y, y^*$ . This shows  $(x^*, x) \in \partial \mathcal{F}_T(x, x^*)$ . Equivalently,

$$2\langle x, x^* \rangle = \mathcal{F}_T(x, x^*) + \mathcal{F}_T^*(x, x^*) = \mathcal{F}_T(x, x^*) + \overline{\mathcal{P}_T}(x, x^*)$$

and so  $\overline{\mathcal{P}_T}(x, x^*) = \langle x, x^* \rangle$ . □

Note that  $\mathcal{F}_T$  need not represent  $T$  if  $T$  is not maximal. The situation is however ameliorated when  $T = A + B$  is the sum of maximal monotone operators satisfying

$$(1) \quad 0 \in \text{core} \{ \text{conv} D(A) - \text{conv} D(B) \}.$$

We next define two partial infimal convolutions:

$$\mathcal{V}_{A,B}(x, x^*) := \inf \{ \mathcal{F}_A(x, u^*) + \mathcal{F}_B(x, v^*) : u^* + v^* = x^* \},$$

and

$$\mathcal{W}_{A,B}(x, x^*) := \inf \{ \mathcal{P}_A(x, u^*) + \mathcal{P}_B(x, v^*) : u^* + v^* = x^* \}.$$

The first result is very interesting in its own right it is a lovely observation first exploited by Voisei:

**Theorem 3. (Partial Convolution, [12].)** *Suppose  $A$  and  $B$  are maximal monotone and satisfy the transversality condition (1). Then  $\mathcal{V}_{A,B}(x, x^*) = \mathcal{W}_{A,B}^*(x, x^*)$  is norm-weak-star lower-semicontinuous and is attained when finite.*

In consequence

$$\mathcal{V}_{A,B}(x, x^*) \geq \langle x, x^* \rangle$$

with equality if and only if  $x^* \in (A + B)(x)$ . In particular,  $\mathcal{V}_{A,B}$  represents  $A + B$  and so  $\mathcal{V}_{A,B} \leq \overline{\mathcal{P}}_{A+B}$ .

**Proof.** The argument—based on a conjugate formula of Penot [8, Prop. 13]—as in Vosei [12] and in [2, §5], or a direct Lagrangian calculation, shows  $\mathcal{V}_{A,B}(x, x^*) = \mathcal{W}_{A,B}^*(x, x^*)$  and is attained when finite. The rest follows since  $\mathcal{P}_A^* = \mathcal{F}_A$  and  $\mathcal{P}_B^* = \mathcal{F}_B$  have the representative properties of Proposition 1.

Indeed,  $\mathcal{V}_{A,B}(x, x^*) \geq \langle x, x^* \rangle$  follows directly from the definition of convolution as does  $\mathcal{V}_{A,B}(x, x^*) = \langle x, x^* \rangle$  when  $x^* \in (A + B)(x)$ . Finally if  $\mathcal{V}_{A,B}(x, x^*) = \langle x, x^* \rangle$ , we let  $x^* = u^* + v^*$  be the attaining values, as assured by the conjugacy formula. Then

$$0 = \mathcal{V}_{A,B}(x, x^*) - \langle x, x^* \rangle = \{ \mathcal{F}_A(x, u^*) - \langle x, u^* \rangle \} + \{ \mathcal{F}_B(x, v^*) - \langle x, v^* \rangle \}.$$

As the bracketed terms are non-negative we deduce that they are both zero and so  $u^* \in A(x)$ ,  $v^* \in B(x)$ ; and we are done.  $\square$

Let us say that a monotone operator  $T$  is *almost maximal* if

$$\mathcal{F}_T(x, x^*) \geq \langle x, x^* \rangle$$

for all  $x \in X, x^* \in X^*$ . This is to say that  $\mathcal{F}_T$  represents  $T$ . The name is justified since Proposition 1 assures that every maximal monotone operator is almost maximal. Also, if  $T$  is maximal and  $\overline{\text{Gr } S} = \text{Gr}(T)$  then  $S$  is almost maximal.

A nice consequence of the definition is:

**Corollary 4.** *Suppose  $T$  is almost maximal monotone. Then a closed convex function  $\mathcal{H}$  represents  $T$  if and only if*

$$\mathcal{F}_T \leq \mathcal{H} \leq \overline{\mathcal{P}}_T.$$

*Proof.* By Proposition 2 we need only show that each representative function  $\mathcal{H}$  is minorized by  $\mathcal{F}_T$ . Suppose we show that  $\mathcal{H}^*$  is also a representative. Then  $\mathcal{H}^* \leq \overline{\mathcal{P}}_T \Rightarrow \mathcal{H} \geq \mathcal{F}_T$  as required. To show  $\mathcal{H}^*$  is a representative, since  $\mathcal{H}^* \geq \mathcal{F}_T$  we need only show that  $\mathcal{H}^*(x, x^*) = \langle x, x^* \rangle$  on  $\text{Gr}(T)$ . This is the case by the argument of Proposition 2 v.) applied to  $\mathcal{H}$ .  $\square$

We offer further justification of the term in the final preliminary result.

**Proposition 5.** *Suppose that  $A$  and  $B$  are maximal monotone operators on a Banach space  $X$  and that the transversality condition (1) holds. Then  $A + B$  is maximal as soon as it is almost maximal.*

**Proof.** Suppose that  $(x, x^*)$  is monotonically related to the graph of  $A + B$ . Then  $\mathcal{F}_{A+B}(x, x^*) \leq \langle x, x^* \rangle$ . As  $A + B$  is almost maximal we deduce that  $\mathcal{F}_{A+B}(x, x^*) = \langle x, x^* \rangle$  and so, by part v.) of Proposition 2, we see that  $\mathcal{P}_{A+B}(x, x^*) = \langle x, x^* \rangle$ . Consequently, an appeal to Theorem 3 shows  $\mathcal{V}_{A,B}(x, x^*) = \langle x, x^* \rangle$  and so that  $x^* \in (A + B)(x)$ , which completes the proof.  $\square$

The next corollary shows that topologically  $A + B$  is close-to-maximal. By ‘ $bdw^*$ ’ we denote weak\*-convergence for bounded nets (and hence include all weak\*-convergent sequences).

**Corollary 6. (Graph Closedness.)** *Suppose that  $A$  and  $B$  are maximal monotone in Banach space and that the transversality condition (1) holds. Then  $A + B$  has a  $\|\cdot\| \times bdw^*$  closed graph and consequently has weak\*-closed convex images.*

**Proof.** Clearly,

$$\{(x, x^*) : \mathcal{V}_{A,B}(x, x^*) - \langle x, x^* \rangle \leq 0\} = \text{Gr}(A + B)$$

is  $\|\cdot\| \times bdw^*$  closed in the product space, since  $\mathcal{V}_{A,B}$  is  $\|\cdot\| \times bdw^*$  lower-semicontinuous while the bilinear form is  $\|\cdot\| \times bdw^*$  continuous.  $\square$

Observe that the graph of a maximal monotone operator need not be  $\|\cdot\| \times w^*$  closed (or even  $bw^*$ ) already for a subgradient [4, Example 5.2.31]. The  $w^*$  closure of  $(A + B)(x)$  was first proven in [11].

### 3. OUR MAIN RESULTS

We first provide two useful criteria for almost maximality.

**Proposition 7.** *Assume that  $S$  is monotone and that either  $S$  is surjective or has full domain. Then  $S$  is almost maximal.*

**Proof.** Fix  $x^* \in X^*$  and  $x \in X$ . Suppose  $S$  is surjective and write  $x^* = s^* \in S(s)$ . Then, by definition

$$\mathcal{F}_S(x, x^*) \geq \langle x, s^* \rangle + \langle s, x^* \rangle - \langle s, s^* \rangle = \langle x, x^* \rangle.$$

The other case is similar.  $\square$

We denote the *algebraic closure* of a set at  $x \in C$  by  $C^{alg}(x) := \{d : t_n d + (1 - t_n)x \in C, \exists t_n < 1, t_n \rightarrow 1\}$ . We write  $C^{alg} := \bigcap_{x \in C} C^{alg}(x)$ .

**Proposition 8.** *Assume that  $A$  and  $B$  are maximal monotone and that (1) holds. Assume also that*

$$(2) \quad \overline{\text{conv}} D(A) \cap \overline{\text{conv}} D(B) = \overline{D(A) \cap D(B)}^{alg}.$$

*Then  $A + B$  is almost maximal.*

**Proof.** Assume that  $0 \in D(A) \cap D(B)$ . Let  $U := \overline{\text{conv}} D(A)$  and  $V := \overline{\text{conv}} D(B)$ . Note also that (1) implies  $N_U + N_V = N_{U \cap V}$ . Much as in [12] we argue by maximality that  $A = A + N_U$  and  $B = B + N_V$ . Thus, using (2) shows

$$A + B = (A + B) + N_{\overline{D(A+B)}}.$$

Now suppose that  $\mathcal{F}_{A+B}(x, x^*) \leq \langle x, x^* \rangle$ . This implies that for every  $n^* \in N_{\overline{D(A+B)}}(y)$  and  $s^* \in (A + B)(y)$  we have

$$\langle s^* + tn^* - x^*, y - x \rangle \geq 0,$$

for all  $t > 0$ . Hence  $0 \in N_{\overline{D(A+B)}}(x)$  by maximality of the normal cone.

Thus,  $x \in \overline{D(A+B)}$ . Now (2) implies that  $x$  lies in  $\overline{D(A+B)}^{alg}$ . Thus we can select  $z^* \in (A+B)(\alpha x)$ , for  $0 < \alpha < 1$ . Then, by convexity and by definition

$$\begin{aligned} \alpha \mathcal{F}_{A+B}(x, x^*) &= \alpha \mathcal{F}_{A+B}(x, x^*) + (1 - \alpha) \mathcal{F}_{A+B}(0, 0) \\ &\geq \mathcal{F}_{A+B}(\alpha x, \alpha x^*) \\ &\geq \langle \alpha x, \alpha x^* \rangle + \langle \alpha x, z^* \rangle - \langle \alpha x, z^* \rangle = \alpha^2 \langle x, x^* \rangle. \end{aligned}$$

Thus  $\mathcal{F}_{A+B}(x, x^*) \geq \alpha \langle x, x^* \rangle$ . Letting  $\alpha \uparrow 1$  completes the proof.  $\square$

**Theorem 9. (Maximality of Sums, I.)** *Suppose that  $A$  and  $B$  are maximal monotone on a Banach space. Suppose also that either*

- i.) *The set  $\text{int } D(A) \cap \text{int } D(B)$  is nonempty; or*
- ii.)  *$D(A) \cap \text{int } D(B) \neq \emptyset$  while  $D(A)$  is closed and convex; or*
- iii.) *Both  $D(A), D(B)$  are closed and convex and  $0 \in \text{core conv } \{D(A) - D(B)\}$ .*

*Then  $A + B$  is maximal monotone.*

**Proof.** Each of the hypotheses leads to  $\overline{\text{conv}} D(A) \cap \overline{\text{conv}} D(B) \subset \overline{D(A+B)}^{alg}$ , since  $\overline{D(A)}$  is convex when  $D(A)$  has nonempty interior, see [2, 9, 10]. More over the hypotheses of i.) and ii.) imply (1). Thus, Proposition 8 applies as then does Proposition 5.  $\square$

Part iii.) of Theorem 9 is the main result in [12]. In [13] corresponding results are given for compositions with closed convex domain—such also extend as above. A quite different proof of the Theorem 9 i.) follows from results in [2]:

**Theorem 10. (Maximality of Sums, II.)** *Suppose  $A$  and  $B$  are maximal monotone on a Banach space. Suppose also that  $\text{core conv } D(A) \cap \text{core conv } D(B) \neq \emptyset$ .*

*Then  $A + B$  is maximal monotone.*

**Proof.** Suppose  $(x, x^*)$  is monotonically related to the graph of  $A + B$ . Let  $W$  be an arbitrary basic weak-star zero neighbourhood. Fix a finite dimensional subspace  $F$  of  $X$  containing both  $x$  and the vectors defining  $W$ . By translation we may assume that  $0 \in \text{core conv } D(A) \cap \text{core conv } D(B) \neq \emptyset$ . Hence, by the composition result in [2, §5], both  $A_F$  and  $B_F$  are maximal monotone; and  $0 \in \text{core conv } \{D(A_F) - D(B_F)\}$ . Thus, by the reflexive (or finite-dimensional) sum theorem

$$A_F + B_F = (A + B)_F$$

is maximal monotone. Since  $x \in F$  and  $(x, x^*)$  is monotonically related to the graph of  $(A + B)$  we observe that  $(x, x^*|_F)$  is monotonically related to the graph of  $(A + B)_F$ . Hence by maximality  $x^*|_F \in (A + B)_F(x)$ . In consequence,

$$x^* \in (A + B)(x) + F^\perp \subset (A + B)(x) + W.$$

Since  $W$  is arbitrary, applying Corollary 6, we deduce that

$$x^* \in \overline{(A + B)(x)}^* = (A + B)(x),$$

by the Veronas' part of Corollary 6, and we are done.  $\square$

**Remark 11.** This argument works under the Brezis-Attouch condition (1) if we can ensure that for each finite dimensional space  $F$  there is a reflexive superspace,  $R$ , such that  $A_R$  and  $B_R$  are both maximal. This is the case, for example, if after translation  $0 \in D(A) \cap \text{core } D(B) \neq \emptyset$  and for each finite dimensional  $F$  there is a reflexive subspace  $R$  containing  $F$  such that  $A|_R$  is maximal. Thus, any counter-example to the sum theorem has to have quite messy domains.

Another proof of this result can be obtained from Asplund's decomposition of a maximal monotone operator as the sum of cyclic and acyclic operators, as described in [2, §3].

Note also that maximality of  $T_Y$  and that of  $T + \partial\iota_Y$  are equivalent for a closed subspace  $Y$ .  $\square$

**Remark 12.** Suppose  $f$  is lower-semicontinuous, proper and convex in Banach space. We observe—without appealing to maximality—that the representative function  $(f \oplus f^*)(x, x^*) := f(x) + f^*(x^*)$  coincides with  $\langle x, x^* \rangle$  exactly for  $x^*$  in  $\partial f(x)$ . As in Proposition 5, to prove  $\partial f$  maximal it thus suffices to show  $\partial f$  is almost maximal. Can this be done any more efficiently than directly proving maximality via an approximate mean-value theorem, as say in [4, Thm. 3.4.6]?

**Remark 13.** We can probably significantly improve the result in the case where one operator is a subgradient because the representative function  $(f \oplus f^*)(x, x^*) := f(x) + f^*(x^*)$  is exact. We define

$$\mathcal{F}_{T,f}(x, x^*) := f(x) + \sup_{y^* \in T(y)} \{\langle x, y^* \rangle + \langle y, x^* \rangle - \langle y, y^* \rangle - f(y)\},$$

with conjugate

$$\mathcal{P}_{T,f}(x, x^*) := f(x) + \overline{\text{conv}}_{y_i^* \in T(y_i)} \{\langle y_i, y_i^* \rangle - f(y_i)\},$$

and note that  $\mathcal{F}_{T,0} = \mathcal{F}_T$  and  $\mathcal{P}_{T,0} = \overline{\mathcal{P}}_T$ , and as before  $\mathcal{P}_{T,f}^* = \mathcal{F}_{T,f} \leq \mathcal{P}_{T,f}$ . Likewise,  $\mathcal{F}_{0,f} = f \oplus f^*$ .

I conjecture that

$$(3) \quad \mathcal{F}_{T,f}(x, x^*) \leq \langle x, x^* \rangle$$

for  $(x, x^*)$  monotonically related to  $\text{Gr}(T + \partial f)$  as holds in the extreme cases.  $\square$

**Theorem 14. (Maximality of Sums, III.)** *Suppose  $T$  is maximal monotone and  $f$  is convex and closed. Suppose that (3) holds and that*

$$(4) \quad \text{dom } f \cap \text{core conv } D(T)$$

*is nonempty. Then  $T + \partial f$  is maximal.*

**Proof.** *We define*

$$\mathcal{V}_{T,f}(x, x^*) := f(x) + (\mathcal{F}_{T,f}(x, \cdot) \square f^*)(x^*).$$

*Then*

$$\mathcal{V}_{T,f}(x, x^*) \leq \mathcal{P}_{T,f}(x, x^*).$$

*By assumption  $\mathcal{F}_{T,f}(x, x^*) \leq \langle x, x^* \rangle$  for  $(x, x^*)$  monotonically related to  $\text{Gr}(T + \partial f)$  while, as in Proposition 3,  $\mathcal{V}_{T,f}(x, x^*)$  is an exact representative for  $T + \partial f$ . Now*

the weakened constraint qualification (4) still ensures  $\mathcal{F}_{T,f}$  represents  $T + \partial f$  as it implies that  $\text{dom } f \cap \overline{\text{conv}}D(T) \subseteq \overline{\text{dom } f \cap \text{conv}D(T)}^{\text{alg}}$ . Much as before

$$\mathcal{F}_{T,f}(x, x^*) = \langle x, x^* \rangle \Rightarrow \mathcal{P}_{T,f}(x, x^*) = \langle x, x^* \rangle \Rightarrow \mathcal{V}_{T,f}(x, x^*) = \langle x, x^* \rangle,$$

and we are done.  $\square$

We finish with two especially nice consequences of part ii.) of Theorem 9.

**Corollary 15. (Normal Cones.)** *Suppose in an arbitrary Banach space that  $T$  is maximal monotone and  $C$  is closed and convex while  $C \cap \text{int } D(T) \neq \emptyset$ . Then  $T + N_C$  is maximal monotone.*

**Proof.** The maximality of  $T + N_C$  is an immediate consequence of part ii.) of Theorem 9.  $\square$

Recall that a maximal monotone mapping  $T$  is maximal monotone locally [11], or *type (FPV)*, if for every open convex set  $V$  in  $X$  with  $V \cap D(T) \neq \emptyset$  the following holds for every  $x \in V$ :  $\langle y^* - x^*, y - x \rangle \geq 0$  for all  $y^* \in T(y)$ , and all  $y \in V$  implies that  $x^* \in T(x)$ .

**Corollary 16.** *Suppose in an arbitrary Banach space that  $T$  is maximal monotone and  $D(T)$  has nonempty interior, or is closed and convex, then  $T$  is of type (FPV).*

**Proof.** We argue as follows. Fix  $x, V$  and  $x^*$  as in the definition of (FPV). We may select a closed convex set  $C$  such that  $x \in \text{int } C \subset V$  and  $\text{int } D(T) \cap C \neq \emptyset$ . It follows from Corollary 15 that  $T + N_C$  is maximal. Let  $y^* \in T(y), n^* \in N_C(y), y \in Y$  be given. Then  $\langle y^* + n^* - x^*, y - x \rangle = \langle y^* - x^*, y - x \rangle + \langle n^*, y - x \rangle \geq 0$  since  $x \in C$ . By maximality  $x^* \in T(x) + N_C(x) = T(x)$  since  $x \in \text{int } C$ .

The same argument shows that every maximal monotone mapping with closed convex domain is type (FPV).  $\square$

The case  $D(T) = X$  of Corollary 16 was first established in [7].

#### 4. FINAL REMARKS

As distinct from the reflexive case, we note that our arguments make no use of the duality map. Indeed, outside of reflexive space, the Rockafellar-Minty approach via surjectivity of  $T + J$  is of little use, see [2, §6] and [4, §5.1]. That said, Theorem 9 certainly suggests that  $A + B$  may well be maximal given only (1) and no auxiliary conditions.

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