# On $\sigma$-convexity and $\sigma$-monotonicity 

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## Outline

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- Convex and $\sigma$-convex functions


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- Topological properties of $\sigma$-convex functions
- Subdifferential and $\sigma$-Subdifferential
- conjugate and ( $\sigma, y$ )-conjugate
- Monotone and $\sigma$-Monotone operators


## Convex functions

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D(f)=\operatorname{dom} f=\{x \in X: f(x)<\infty\}
$$

The function $f$ is called proper if $\operatorname{dom} f \neq \emptyset$. In addition, $f$ is said to be convex when for all $x, y \in X$ and for each $t \in[0,1]$,

$$
f((1-t) x+t y) \leq(1-t) f(x)+t f(y)
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If $f(x) \in \mathbb{R}$, then the subdifferential of $f$ at $x$ is denoted by $\partial f(x)$ and is defined as the set of all $x^{*} \in X^{*}$ satisfying

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for all $y \in X$. When $f(x) \notin \mathbb{R}$ we define $\partial f(x)=\emptyset$. We say that $f$ is subdifferentiable at $x$ if $\partial f(x) \neq \emptyset$.

## Definition (M.H.A, Roohi, 2017)

Given a function $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ and a map $\sigma$ form $\operatorname{dom} f$ to $\mathbb{R}_{+}$, we say that $f$ is $\sigma$-convex if

$$
\begin{equation*}
f(t x+(1-t) y) \leq t f(x)+(1-t) f(y)+t(1-t) \min \{\sigma(x), \sigma(y)\}\|x-y\| \tag{1}
\end{equation*}
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for all $x, y \in X$, and $t \in] 0,1[$.
There are $\sigma$-convex functions which are not $\varepsilon$-convex for any $\varepsilon \geq 0$, as shown in the following example.

## Example (M.H.A, Roohi, 2017)

Consider the functions $\varphi, f, \sigma: \mathbb{R} \rightarrow \mathbb{R}$ defined by

$$
\begin{aligned}
& \varphi(x)=\left\{\begin{array}{cl}
x \sin ^{2} x & \text { if } x \geq 0 \\
0 & \text { if } x<0
\end{array}\right. \\
& \sigma(x)=\max \left\{\varphi(x), \max _{z \leq x} \varphi(z)-\varphi(x)\right\} \\
& f(x)=\int_{0}^{x} \varphi(t) d t
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This function $f$ is $\sigma$-convex, but it is not $\epsilon$-convex for any $\epsilon>0$.
Note that if $f$ is a $\sigma$-convex function, then $\operatorname{dom} f$ is a convex set.

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## Example (M.H.A, 2021)

Consider the function $f: X \rightarrow \mathbb{R}$ defined by $f(x)=-\|x\|$. Then $f$ is $\sigma$-convex for $\sigma \equiv 2$, but it does not have an affine minorant.

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Fix $a \in X$ and define the function $g: X \rightarrow \mathbb{R}$ by $g(x)=|\|x\|-\|a\||$. Then $g$ is $\sigma$-convex for $\sigma \equiv 2$, and $y=0$ is an affine minorant of $g$.

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## Example (M.H.A, Hosseinabadi, 2023)

Fixed $a \in X$ and define $f_{0}(x)=|\|x\|-\|a\||$. For each $n \in \mathbb{N}$, define $f_{n}$ recursively by $f_{n}(x)=\mid f_{n-1}(x)-\|a\| \|$. Then $f_{n}$ for all $n \in \mathbb{N} \cup\{0\}$ is $\sigma$-convex with $\sigma \equiv 2$, and $y=0$ is an affine minorant of it.


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## Elementary properties

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\begin{equation*}
f(t x+(1-t) y) \leq t f(x)+(1-t) f(y)+t(1-t) \sigma(x)\|x-y\| \tag{2}
\end{equation*}
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(iii) Let $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ be a function. Then $f$ is convex if and only if it is $\sigma$-convex for every $\sigma: \operatorname{dom} f \rightarrow \mathbb{R}_{+}$.

Given a function $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$, we define the $\operatorname{map} \sigma_{f}: \operatorname{dom} f \rightarrow \mathbb{R}_{+} \cup\{+\infty\}$ by

$$
\begin{aligned}
\sigma_{f}(x) & =\inf \left\{a \in \mathbb{R}_{+}: \frac{f(t x+(1-t) y)-t f(x)-(1-t) f(y)}{t(1-t)}\right. \\
& \leq a\|x-y\|, \forall y \in \operatorname{dom} f, t \in] 0,1[ \}
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It should be noticed that if $f$ is $\sigma^{\prime}$-convex for some $\sigma^{\prime}: \operatorname{dom} f \rightarrow \mathbb{R}_{+}$, then

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\begin{equation*}
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In this case, $\sigma_{f}$ is finite and $f$ is $\sigma_{f}$-convex. Note that $\sigma_{f}$ is the minimal $\sigma$ such that $f$ is $\sigma$-convex.

## Explicit formula

## Proposition (M.H.A, 2020)

Suppose that $f$ is $\sigma$-convex for some $\sigma$. Then

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\begin{equation*}
\sigma_{f}(x)=\max \left\{0, \sup _{t \in[0,1[y \in \operatorname{dom} f \backslash\{x\}} \frac{f(t x+(1-t) y)-t f(x)-(1-t) f(y)}{t(1-t)\|x-y\|}\right\} . \tag{4}
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## Property B

We introduce the following assumption:


We say that the function $\sigma$ has the property B , if for every $x \in \operatorname{int} \operatorname{dom} f$ and every $\varepsilon>0$ sufficiently small, $\sigma$ is bounded on the sphere $S(x, \varepsilon)=\{y \in X:\|x-y\|=\varepsilon\}$.

## Theorem (M.H.A, 2020)

Let $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ be a $\sigma$-convex function. Assume that $f$ is locally bounded from above in the interior of its domain. If $\sigma$ satisfies property $B$, then $f$ is locally Lipschitz in the interior of its domain.

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## Corollary

Every proper, $\sigma$-convex function $f: \mathbb{R} \rightarrow \mathbb{R} \cup\{+\infty\}$ is locally Lipschitz in the interior of its domain.

## Clarke-Rockafellar Directional Derivative

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For a proper function $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ the Clarke-Rockafellar generalized directional derivative at $x$ in a direction $z \in X$ is defined by

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f^{\uparrow}(x, z)=\sup _{\delta>0} \limsup _{(y, \alpha) \xrightarrow[\rightarrow]{f} x, \lambda \searrow 0} \inf _{u \in B(z, \delta)} \frac{f(y+\lambda u)-\alpha}{\lambda}
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where $(y, \alpha) \xrightarrow{f} x$ means that $y \rightarrow x, \alpha \rightarrow f(x)$ and $\alpha \geq f(y)$.
If $f$ is lsc at $x$, the above definition coincides with

$$
f^{\uparrow}(x, z)=\sup _{\delta>0} \limsup _{\substack{f \\ y \rightarrow x, \lambda \searrow 0}} \inf _{u \in B(z, \delta)} \frac{f(y+\lambda u)-f(y)}{\lambda} .
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Here, $y \xrightarrow{f} x$ means that $y \rightarrow x \quad$ and $\quad f(y) \rightarrow f(x)$.

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The Clarke-Rockafellar subdifferential of $f$ at $x \in \operatorname{dom} f$ is defined by

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In the following we introduce the notion of $\sigma$-subdifferential.

## $\sigma$-Subdifferential

## Definition (M.H.A, 2020)

Suppose that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is a proper function. The $\sigma$-subdifferential of $f$ is the multivalued operator $\partial^{\sigma} f: X \rightarrow 2^{X^{*}}$ defined by

$$
\partial^{\sigma} f(x):=\left\{x^{*}:\left\langle x^{*}, z\right\rangle \leq f(x+z)-f(x)+\min \{\sigma(x), \sigma(z+x)\}\|z\| \quad \forall z \in X\right\}
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if $x \in \operatorname{dom} f$; otherwise it is empty.

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It follows from the above definition that $\partial f \subset \partial^{\sigma} f$ and so $D(\partial f) \subset D\left(\partial^{\sigma} f\right) \subset \operatorname{dom} f$.

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if $x \in \operatorname{dom} f$; otherwise it is empty.
It follows from the above definition that $\partial f \subset \partial^{\sigma} f$ and so $D(\partial f) \subset D\left(\partial^{\sigma} f\right) \subset \operatorname{dom} f$. In the next proposition, we find a relationship between $\partial^{C R} f(x)$ and $\partial^{\sigma} f(x)$.

## Proposition (M.H.A, 2020)

Assume that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is lsc and $\sigma$-convex. Then $\partial^{C R} f(x) \subset \partial^{\sigma} f(x)$.

## Example

Note that the function $f(x)=-|x|$ is $\sigma$-convex with $\sigma \equiv 2$. Then $\partial f(0)=\emptyset$, and $\partial^{C R}(f(0))=[-1,1]$

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On the other hand, if we take $\sigma^{\prime} \equiv 4$, then $f$ is $\sigma^{\prime}$-convex and $\partial^{\sigma^{\prime}} f(0)=[-3,3]$.
Thus the inclusion in the above proposition can be equality or strict.

## Proposition (M.H.A, 2021)

Let $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ be a proper and convex function. Suppose that $f$ and $\sigma: \operatorname{dom} f \rightarrow \mathbb{R}_{+}$are Lipschitz near $x$ and $f(\cdot)+\sigma(\cdot)\|\cdot-x\|$ is convex. Then

$$
\partial^{\sigma} f(x) \subset \partial f(x)+\sigma(x) B^{*}
$$

## Proposition (M.H.A, 2021)

Assume that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is a proper $\sigma$-convex function. If $f$ is Gateaux differentiable at $x \in X$, then $f^{\prime}(x) \in \partial^{\sigma} f(x)$ i.e., $\partial^{\sigma} f(x)$ is nonempty.

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## Proposition

Suppose that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is proper. If $\partial^{\sigma} f(x) \neq \emptyset$ and $\lim \sup _{y \rightarrow x} \sigma(y)<\infty$, then $f$ is lsc at $x$.

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## Proposition

Suppose that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is a proper $\sigma$-convex function and $x \in \operatorname{dom} f$. If $\operatorname{int} \operatorname{dom} f \neq \emptyset$ and $x \in \operatorname{bd}(\operatorname{dom} f)$, then $\partial^{\sigma} f(x)$ is either empty or unbounded.

## Proposition

Let $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ be a proper $\sigma$-convex function and $x_{0} \in \operatorname{dom} f$ be a local minimizer of $f$. Set

$$
\begin{equation*}
\varphi(x):=f(x)+\min \left\{\sigma(x), \sigma\left(x_{0}\right)\right\}\left\|x-x_{0}\right\| . \tag{5}
\end{equation*}
$$

Then $\varphi$ attains its global minimum at $x_{0}$.

## Sigma-subdifferential of sum

## Proposition (M.H.A., Zanjani, 2024)

Suppose $f, g: X \rightarrow \mathbb{R} \cup\{+\infty\}$ are proper, lsc and $\sigma$-convex functions and $\sigma$ has the property B. If $\bar{x} \in \operatorname{dom} g \cap \operatorname{int}(\operatorname{dom} f)$ is a local minimum point of the function $f+g-\left\langle x^{*}, \cdot\right\rangle$ for all $x^{*} \in \partial^{2 \sigma}(f+g)(\bar{x})$, then

## Sigma-subdifferential of sum

## Proposition (M.H.A., Zanjani, 2024)

Suppose $f, g: X \rightarrow \mathbb{R} \cup\{+\infty\}$ are proper, lsc and $\sigma$-convex functions and $\sigma$ has the property B. If $\bar{x} \in \operatorname{dom} g \cap \operatorname{int}(\operatorname{dom} f)$ is a local minimum point of the function $f+g-\left\langle x^{*}, \cdot\right\rangle$ for all $x^{*} \in \partial^{2 \sigma}(f+g)(\bar{x})$, then

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\partial^{\sigma} f(x)+\partial^{\sigma} g(x)=\partial^{2 \sigma}(f+g)(x)
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## Lemma

Let $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ be a proper and convex function. Then $\partial^{\sigma} f(x)=\partial(f+\sigma(\cdot)\|\cdot-x\|)(x)$. If, in addition, $f(\cdot)+\sigma(\cdot)\|\cdot-x\|$ is convex, then $\partial^{\sigma} f(x)=\partial^{C R}(f+\sigma(\cdot)\|\cdot x\|)(x)$.

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## Theorem

Suppose that $f, g: X \rightarrow \mathbb{R} \cup\{+\infty\}$ are proper, lsc and $\sigma$-convex functions. Assume that $\sigma$ is Lipschitz, $\sigma(0)=0, x_{0} \in \operatorname{int}(\operatorname{dom} f) \cap \operatorname{dom} g$ and $\sigma\left(x_{0}+z\right)=\sigma(z)$ for all $z \in X$. Then

$$
\partial^{\sigma} f\left(x_{0}\right)+\partial^{\sigma} g\left(x_{0}\right)=\partial^{2 \sigma}(f+g)\left(x_{0}\right) .
$$

## $\sigma$-conjugate

## Definition (M.H.A., 2021)

Suppose that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is a $\sigma$-convex function and $y \in X$ is fixed. Then the $\operatorname{map} f_{\sigma, y}^{*}: X^{*} \rightarrow \mathbb{R} \cup\{+\infty\}$ defined by

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\begin{equation*}
f_{\sigma, y}^{*}\left(x^{*}\right)=\sup _{x \in X}\left\{\left\langle x^{*}, x\right\rangle-f(x)-\sigma(x)\|x-y\|\right\}, \quad \forall x^{*} \in X^{*} \tag{6}
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As for the convex case, the function $f_{\sigma, y}^{*}: X \rightarrow \mathbb{R} \cup\{ \pm \infty\}$ defined by

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## Proposition

Suppose that $f: X \rightarrow \mathbb{R} \cup\{+\infty\}$ is a $\sigma$-convex function. Then for every $x \in \operatorname{dom} f$, $x^{*} \in X^{*}$,

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\begin{equation*}
x^{*} \in \partial^{\sigma} f(x) \Longleftrightarrow f_{\sigma, x}^{*}\left(x^{*}\right)+f(x)=\left\langle x^{*}, x\right\rangle . \tag{7}
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In particular, $\operatorname{gr}\left(\partial^{\sigma} f\right) \subset \operatorname{dom} f \times \operatorname{dom} f_{\sigma, x}^{*}$.

Theorem (M.H.A., 2021)
Suppose that $f, g: X \rightarrow \mathbb{R} \cup\{+\infty\}$ are lsc. Let $f$ be $\sigma$-convex, $g$ be $\sigma^{\prime}$-convex. Then (i) for every $y \in X$ and $x^{*} \in X^{*}$ one has

$$
\begin{equation*}
(f+g)_{\sigma+\sigma^{\prime}, y}^{*}\left(x^{*}\right) \leq\left(f_{\sigma, y}^{*}(\cdot) \square g_{\sigma^{\prime}, y}^{*}(\cdot)\right)\left(x^{*}\right) ; \tag{8}
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(ii) if $\sigma$ satisfies property $B, y \in \operatorname{int} \operatorname{dom} f$ and it is a local minimum point of the function $f+g-\left\langle x^{*}, \cdot\right\rangle$, then the equality holds.

## Monotone Operatos

Let $T$ be a set-valued map from $X$ to $X^{*}$. The domain and the graph of $T$ are, respectively, defined by

$$
\begin{gathered}
D(T)=\{x \in X: T(x) \neq \emptyset\} \\
\operatorname{gr} T=\left\{\left(x, x^{*}\right) \in X \times X^{*}: x \in D(T), \text { and } x^{*} \in T(x)\right\} .
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For two set-valued operators $T$ and $S$, we write $T \subseteq S$ if $S$ is an extension of $T$, i.e., $\operatorname{gr} T \subseteq \operatorname{gr} S$.
We recall that $T$ is monotone if

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\left\langle x-y, x^{*}-y^{*}\right\rangle \geq 0
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A monotone operator is called maximal monotone if it has no monotone extension other than itself.

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## $\sigma$-Monotonicity

## Definition

(i) Given an operator $T: X \rightarrow 2^{X^{*}}$ and a map $\sigma: D(T) \rightarrow \mathbb{R}_{+}, T$ is said to be $\sigma$-monotone if for every $x, y \in D(T), x^{*} \in T(x)$ and $y^{*} \in T(y)$,

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\left\langle x^{*}-y^{*}, y-x\right\rangle \leq \min \{\sigma(x), \sigma(y)\}\|x-y\| \tag{9}
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(ii) A $\sigma$-monotone operator $T$ is called maximal $\sigma$-monotone, if for every operator $T^{\prime}$ which is $\sigma^{\prime}$-monotone with $\operatorname{gr} T \subseteq \operatorname{gr} T^{\prime}$ and $\sigma^{\prime}$ an extension of $\sigma$, one has $T=T^{\prime}$.

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The operator $T$ is called premonotone, if it is $\sigma$-monotone for some $\sigma$.

## $\sigma$-Monotone

Theorem (Rockafellar for $\sigma$-monotonicity, M.H.A., Hadjisavvas, Roohi, 2012)
Suppose that $X$ is a Banach space and $T: X \rightarrow 2^{X^{*}}$ is a premonotone operator. Then $T$ is locally bounded at every point of int $D(T)$.

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Theorem (Libor Veselý for $\sigma$-monotonicity, M.H.A., Hadjisavvas, Roohi, 2012)
Suppose that $T$ is maximal $\sigma$-monotone, $\sigma$ is usc and $x_{0} \in \overline{D(T)}$. If $T$ is locally bounded at $x_{0}$, then $x_{0} \in D(T)$. If in addition $\overline{D(T)}$ is convex, then $x_{0} \in \operatorname{int} D(T)$.

## $\sigma$-Monotonicity

We recall that an operator $T: \mathbb{R}^{n} \rightarrow 2^{\mathbb{R}^{n}}$ is bounded on bounded sets if $\cup_{x \in B} T(x)$ is bounded for all bounded set $B \subset \mathbb{R}^{n}$.

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For a bounded set $B \subset \mathbb{R}^{n}$ and any positive $\varepsilon \in \mathbb{R}$, we define $B^{\varepsilon} \subset \mathbb{R}^{n}$ as

$$
\begin{equation*}
B^{\varepsilon}=\left\{x \in \mathbb{R}^{n}: \operatorname{dist}(x, B) \leq \varepsilon\right\} . \tag{10}
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## Theorem (M.H.A. Iusem, Sosa, 2024)

Consider an operator $T: \mathbb{R}^{n} \rightarrow 2^{\mathbb{R}^{n}}$, with convex domain which is bounded on bounded sets and monotone outside some bounded set $B \subset \mathbb{R}^{n}$. Then $T$ is premonotone, with

$$
\sigma(y)=\alpha+\sup _{v \in T(y)}\|v\|
$$

where

$$
\alpha=\sup _{u \in T(x), x \in B^{\varepsilon}}\|u\|
$$

for an arbitrary $\varepsilon>0$.

## $\sigma$-Monotonicity

## Proposition

Let $T: \mathbb{R}^{n} \rightarrow 2^{\mathbb{R}^{n}}$ be bounded on bounded sets and $\sigma$-monotone. Consider a bounded set $B \subset \mathbb{R}^{n}$ and an operator $\bar{T}: \mathbb{R}^{n} \rightarrow 2^{\mathbb{R}^{n}}$ with convex domain which is bounded on bounded sets and such that $\bar{T}(x)=T(x)$ for all $x \notin B$. Then $\bar{T}$ is $\bar{\sigma}$-monotone, with

$$
\bar{\sigma}(y)=\max \{\sigma(y), \widehat{\sigma}(y)\},
$$

where

$$
\widehat{\sigma}(y)=\widehat{\alpha}+\sup _{v \in \bar{T}(y)}\|v\|,
$$

and

$$
\widehat{\alpha}=\sup _{u \in \bar{T}(x), x \in B^{\varepsilon}}\|u\| .
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and

$$
\widehat{\alpha}=\sup _{u \in \bar{T}(x), x \in B^{\varepsilon}}\|u\|
$$

## Corollary

If $T: \mathbb{R}^{n} \rightarrow \mathbb{R}^{n}$ is point to point, continuous and monotone outside some bounded set $B \subset \mathbb{R}^{n}$, then $T$ is $\sigma$-monotone, with $\sigma(y)=\alpha+\|T(y)\|$, where $\alpha=\sup _{x \in B^{\varepsilon}}\|T(x)\|$.

## $\sigma$-Monotonicity

## Corollary

If $p: \mathbb{R} \rightarrow \mathbb{R}$ is a polynomial of odd degree, then $p$ is $\sigma$-monotone.

## $\sigma$-Monotonicity

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If $p: \mathbb{R} \rightarrow \mathbb{R}$ is a polynomial of odd degree, then $p$ is $\sigma$-monotone.
Conjecture (Iusem Sosa, 2020, JNVA ): Every maximal premonotone operator contains a maximal monotone one.

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