# Convexity on groups and semigroups

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## Outline

- Convex sets/functions
- Examples
- Convex Analysis
- Future directions

#### General theme

Many known results hold assuming only an additive structure

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Definitions

## Definition (Convex set in vector spaces)

X a vector space.  $A \subseteq X$  is convex if  $x_1, \ldots, x_n \in A$ ,  $\alpha_i > 0$ ,  $\sum_{i=1}^{n} \alpha_i = 1 \Longrightarrow \sum_{i=1}^{n} \alpha_i x_i \in A$ .

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If 
$$\alpha_i \in \mathbb{Q}$$
, write  $\alpha_i = \frac{m_i}{m}$ . Then  $\sum_{i=1}^n \alpha_i = 1 \Longrightarrow \sum_{i=1}^n m_i = m$ .

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Definition (Convex set in monoids/groups)

X a monoid.  $A \subseteq X$  is said to be convex if  $x_1, \ldots, x_n \in A$ ,  $m_1, \ldots, m_n, m \in \mathbb{N}$ ,  $m_X = \sum_{i=1}^n m_i x_i$ ,  $m = \sum_{i=1}^n m_i \Longrightarrow x \in A$ .

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#### Definition (Convex hull)

For  $A \subseteq X$ , conv(A) is the smallest convex set that contains A.

#### Convex function

## Definition (Convex function on vector spaces)

X a vector space.  $f : X \to \mathbb{R}$  is convex if  $f(x) \le \sum_{i=1}^{n} \alpha_i f(x_i)$ , whenever  $x = \sum_{i=1}^{n} \alpha_i x_i$ ,  $\alpha_i > 0$ ,  $\sum_{i=1}^{n} \alpha_i = 1$ . f is concave if -f is convex

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X a monoid.  $f : X \to \mathbb{R}$  is convex if  $mf(x) \le \sum_{i=1}^{n} m_i f(x_i)$ , whenever  $mx = \sum_{i=1}^{n} m_i x_i$ ,  $m = \sum_{i=1}^{n} m_i$ . f is concave if -f is convex

This can be done in a more general setting (X is a module,  $\pm \infty$ ...)

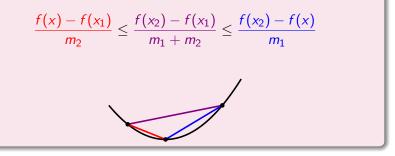
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Definitions

#### Example of some basic properties

#### Example: three slopes lemma

X a monoid,  $f: X \to \mathbb{R}$  convex,  $m, m_1, m_2 \in \mathbb{N}$ ,  $x, x_1, x_2 \in X$  such that  $mx = m_1x_1 + m_2x_2$ . Then



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Definitions

Divisible and semidivisible monoids/groups

In some cases, want to solve the equation  $mx = \sum_{i=1}^{n} m_i x_i$ , at least for some  $m \in \mathbb{N}$ .

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#### Definition (Divisible, semidivisible monoids/groups)

A monoid/group X is p-semidivisible if pX = X. X is semidivisible if it is p-semidivisible for some p. X is divisible if it is p-semidivisible for every p.

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*p*-semidivisible: same as saying that for every  $x \in X$ , there exists  $y \in X$  such that x = py.

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#### Convex sets in certain groups

#### Finite groups

# For every $x \in X$ , there is *m* such that $mx = 0 \Longrightarrow \operatorname{conv}(\{0\}) = X$ . The only convex sets are X and $\emptyset$ .

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#### Circle group

 $X = \mathbb{R}/\mathbb{Z}$ . Then  $\operatorname{conv}(\{x\}) = \{x + y \mid y \in \mathbb{Q}\} \Longrightarrow$  no convex singletons in X.

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Integer lattice  $\mathbb{Z}^d$ 

For 
$$A \subseteq \mathbb{Z}^d$$
,  $\operatorname{conv}_{\mathbb{Z}^d}(A) = \operatorname{conv}_{\mathbb{R}^d}(A) \cap \mathbb{Z}^d$ .

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#### Convex sets in certain groups

#### Arctan semigroup

 $X = [0, \infty)$  with the addition  $a \oplus b = \frac{a+b}{1+ab}$ . If  $a, b \neq 0$  then  $\frac{1}{a} \oplus \frac{1}{b} = a \oplus b$ . Thus, if  $a \neq 1$ , then  $\frac{1}{a} \in \operatorname{conv}(\{a\})$ , and so  $\{0\}$ ,  $\{1\}$  are the only convex singletons. X is 3-semidivisible but not 2-semidivisible.

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## Hyperbolic group

Let X be the matrices of the form  $\pm \begin{bmatrix} \cosh(\theta) & \sinh(\theta) \\ \sinh(\theta) & \cosh(\theta) \end{bmatrix}$ ,  $\theta \in \mathbb{R}$ , with the matrix multiplication. Then  $2nX \neq X$  and (2n+1)X = X. If  $f : \mathbb{R} \to \mathbb{R}$  is convex then  $F\left(\pm \begin{bmatrix} \cosh(\theta) & \sinh(\theta) \\ \sinh(\theta) & \cosh(\theta) \end{bmatrix}\right) = f(\theta)$  is convex.

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Interpolation of convex functions Maximum formula Optimisation

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## Interpolation of convex functions

 $f: X \to \mathbb{R}$  is subadditive if  $f(x + y) \leq f(x) + f(y)$ .

#### Theorem (Kaufman)

X a monoid,  $f, -g : X \to \mathbb{R}$  subadditive, and  $g \leq f$ . Then there exists  $a : X \to \mathbb{R}$  additive such that  $g \leq a \leq f$ .

This is a monoid version of result by Mazur-Orlicz

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Image: A math a math

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 $f: X \to \mathbb{R}$  is affine if it is both convex and concave.

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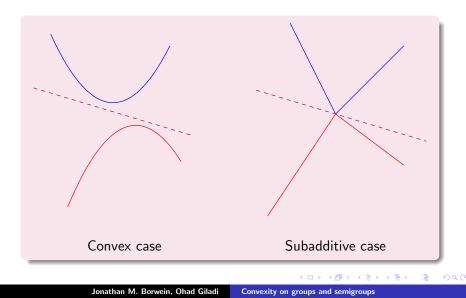
 $f: X \to \mathbb{R}$  is affine if it is both convex and concave.

#### Theorem

X is a semidivisible monoid,  $f, -g : X \to \mathbb{R}$  convex, and  $g \le f$ . Then there exists  $a : X \to \mathbb{R}$  affine such that  $g \le a \le f$ .

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## Picture: interpolation of subadditive/convex functions



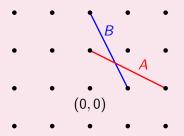
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## Example: nondivisible case

#### Failure in the nondivisible case

$$X = \mathbb{Z}^2$$
,  $f(x) = 5d_A(x) - 1$  and  $g = -5d_B(x) + 1$ .



f, -g are convex,  $g \leq f$ , but there is no affine a s.t.  $g \leq a \leq f$ .

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#### Directional derivative and subgradient

Definition (Directional derivative)

$$f_x(h) = \inf \left\{ n \big( f(x+g) - f(x) \big) \mid ng = h \right\}$$

If f is convex: n(f(x+g) - f(x)) is decreasing in n.

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Definition (Subgradient)

$$\partial f(x) = \left\{ a : X \to \mathbb{R} \mid f(x) + a(h) \leq f(x+h), a \text{ additive} \right\}$$

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#### Theorem (Max formula)

X a semidivisible group and  $f : X \to \mathbb{R}$  convex. Then

$$f_x(h) = \max \left\{ a(h) \mid a \in \partial f(x) \right\}$$

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# Consequences of the max formula

## Definition (Sublinear function)

 $f: X \to \mathbb{R}$  is sublinear if f(nx) = nf(x) and f is subadditive.

#### Theorem (Hahn-Banach for groups)

X a group and  $Y \subseteq X$  a subgroup.  $f : X \to \mathbb{R}$  is sublinear and  $h : Y \to \mathbb{R}$  is additive such that  $h \leq f$  on Y. Then there exists  $\overline{h} : X \to \mathbb{R}$  additive such that  $\overline{h} \leq f$  and  $\overline{h} = h$  on Y.

Did not use semidvisibility since the functions are sublinear.

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# Convex optimisation on groups

#### Consider the constrained problem

$$\inf \left\{ f(x) \mid g_1(x) \leq 0, \dots, g_k(x) \leq 0 \right\}$$

#### Theorem (Subgradient of max function)

X semidivisible group and  $f_1, \ldots, f_k : X \to \mathbb{R}$  convex. Let  $g(x) = \max_{1 \le i \le k} f_i(x)$ . Then

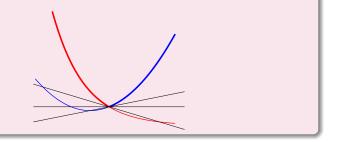
$$\partial g(x) = \operatorname{conv}\Big(\bigcup_{f_i(x)=g(x)} \partial f_i(x)\Big)$$

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# Convex optimisation on groups

## Subgradient of max function

$$\partial g(x) = \operatorname{conv}\Big(\bigcup_{f_i(x)=g(x)} \partial f_i(x)\Big), \ \ g = \max_{1 \le i \le k} f_i$$



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## **Future Directions**

- Noncommutative groups
- Questions in topological groups (continuity, differentiability...)
- Applications in integer programming

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# The End

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