# Convex analysis applied to location theory

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### Justo Puerto

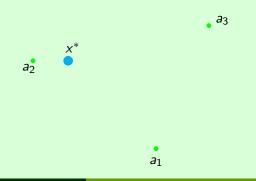
Universidad de Sevilla



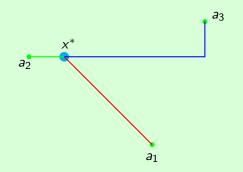
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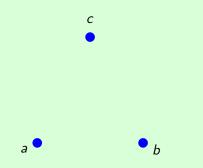
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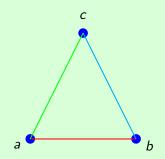
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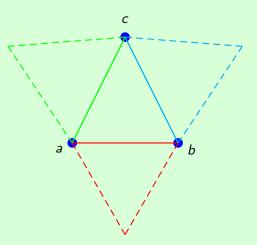
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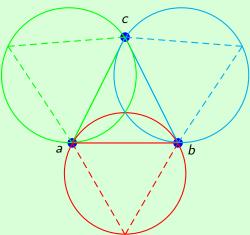
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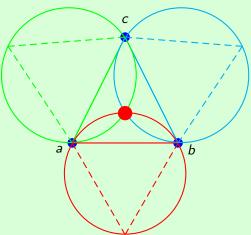
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#### **Elements of a location problem**

Support space
 ● Continuous Spaces (ℝ<sup>n</sup>).
 ● Sphere

Networks

Discrete spaces

## Elements of a location problem

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- Continuous Spaces  $(\mathbb{R}^n)$ 
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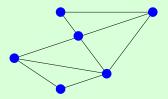
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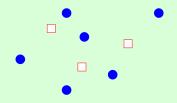


• Discrete spaces

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## **Objective Function**

- Fermat-Weber Problem (Minisum)
- Center Problem (minimax)
- Cent-dian Problem
- o⊳*k*-centnum Problem
- Ordered Median Problem.
- Multiobjective Problem

 $\min\{F_1(x),\ldots,F_k(x)\}$ 

## **Objective Function**

• Fermat-Weber Problem (Minisum)

$$\sum_{a\in A} w_a d_a(x,a) \quad \text{con} \quad i=1,\ldots,M$$

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### **Objective Function**

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$$\max_{a \in A} u_a d_a(x, a) \quad \text{con} \quad i = 1, \dots, M$$

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#### **Objective Function**

- Fermat-Weber Problem (Minisum)
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$$\alpha \sum_{\mathbf{a} \in A} w_{\mathbf{a}} d_{\mathbf{a}}(x, \mathbf{a}) + (1 - \alpha) \max_{\mathbf{a} \in A} u_{\mathbf{a}} d_{\mathbf{a}}(x, \mathbf{a})$$

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# $\min\{F_1(x),\ldots,F_k(x)\}$

## **Objective Function**

- Fermat-Weber Problem (Minisum)
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- k-centrum Problem

$$\sum_{j=M-k}^{M} w_j d_{\sigma_j}(x, a_{\sigma_j})$$
  
where  $d_{\sigma_1}(x, a_{\sigma_1}) \leq \ldots \leq d_{\sigma_M}(x, a_{\sigma_M})$ 

Ordered Median Problem

Multiobjective Problem

 $\min\{F_1(x),\ldots,F_k(x)\}$ 

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## **Objective Function**

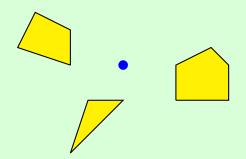
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- Facilities are represented by isolated points?
- Introducing sets instead of points introduces important differences in the mathematical analysis of these problems.
- Our approach: minimization of expected distances

# Introduction

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- Our approach: minimization of expected distances
- Application point of view:
  - Q Case of the stationing of rescue helicopters [Ehrgott 02].
  - Location of planes used to extinguish fires in reserves or natural parks.
  - Locating a read/write head of a computer hard-disk to easily access the stored data. [Vickson, Gerchak and Rotem 95] and [Puerto and Rodríguez-Chía 03].

- Facilities are represented by isolated points?
- Introducing sets instead of points introduces important differences in the mathematical analysis of these problems.
- Our approach: minimization of expected distances
- Our goal: Geometrical characterization of the solution set for a single facility location model with sets as demand facilities using average distances. Networks: [Hakimi64, Hooker91]. Continuous location problems: [Durier and Michelot 85, Durier95, PF00, NPR03].
  - Basic model
  - General model
  - Discretization result.

#### **Basic tools and definitions**

- X is a real separable Banach space.
- $\gamma$  norm with unit ball *B*.
- $f: X \to \mathbb{R} \cup \{+\infty\}$  convex function.

 $p \in \partial f(x) \subseteq X^* ext{ iff } f(y) \geq f(x) + \langle p, y - x \rangle \quad ext{ for each } y \in X.$ 

• Conjugate function:

 $f^*(p) = \sup\{\langle p, x \rangle - f(x) : x \in \text{dom } f\} \ \forall p \in X^*.$ 

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• Conjugate function:  $f^*(p) = \sup\{\langle p, x \rangle - f(x) : x \in \text{dom } f\} \forall p \in X^*.$ 

**Result:** For a closed and proper convex function, [Barbu, Precupanu 75]:

$$p \in \partial f(x), x \in X, p \in X^*$$
 iff  $x \in \partial f^*(p)$ .

#### **Remarks:**

• 
$$\partial \gamma(x) = \{ p \in B^o : \gamma(x) = \langle p, x \rangle \}.$$
  
•  $\partial \gamma^*(p) = \{ x \in X : \gamma(x) = \langle p, x \rangle \}.$ 

**Optimality** Conditions

#### The basic model

$$\inf_{x \in X} \phi(x) := \int_{\mathcal{T}} \varphi_t(x) \ d\mu(t), \qquad (P_{\phi}(\mathcal{T}))$$

where  $\varphi_t(x) := \gamma_t(x - t)$ ,  $\mu$  a  $\sigma$ -finite, positive measure.

Optimality Conditions

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where  $\varphi_t(x) := \gamma_t(x - t)$ ,  $\mu$  a  $\sigma$ -finite, positive measure.

**Result:**  $\phi$  is convex on X. **Result:**  $F : G \longrightarrow L^1(X, \mathbb{R})$  such that  $(F(x))(t) = \varphi_t(x)$ . If X is separable or T is countable, then  $\phi$  is continuous at  $x_o$  and

$$\partial \phi(\mathbf{x}_o) = \int_{\mathcal{T}} \partial \varphi_t(\mathbf{x}_o) \mu(dt) = \int_{\mathcal{T}} \partial F(\mathbf{x}_o) \mu(dt).$$

Optimality Conditions

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**Remark:**  $\mu$  is concentrated on a finite set of points,  $P_{\phi}(T)$  reduces to the classical Fermat-Weber problem.

Optimality Conditions

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where  $\varphi_t(x) := \gamma_t(x - t)$ ,  $\mu$  a  $\sigma$ -finite, positive measure.

## Existence results: [Garkavi and Smatkov 74],

- X is finite dimension and φ<sub>t</sub> are lower-semicontinuous (l.s.) in the t argument;
- X is reflexive and φ<sub>t</sub> are sequentially l.s. in the t argument for the weak topology;
- X is the dual space to a separable space and φ<sub>t</sub> are sequentially l.s. in the t argument for the weak topology
- X is a dual space, φ<sub>t</sub> are l.s. in the t argument for the weak\* topology and T is μ-separable.

Optimality Conditions

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where  $\varphi_t(x) := \gamma_t(x - t)$ ,  $\mu$  a  $\sigma$ -finite, positive measure.

Uniqueness results: [Garkavi and Smatkov 74]

- μ(T) < +∞ and X is a strictly normed space. P<sub>φ</sub>(T) has a unique solution iff T does not contain two nonintersecting subsets T<sub>1</sub> and T<sub>2</sub> such that μ(T<sub>1</sub>) = μ(T<sub>2</sub>) = μ(T)/2, being T<sub>1</sub> and T<sub>2</sub> enclosed in nonintersecting rays ℓ<sub>1</sub> and ℓ<sub>2</sub>, respectively and lying in the same straight line.
- $\mu(T)$  is not finite.

Let dim $(X) \ge 2$ . If  $\gamma_t$  are strict norms and  $\mu$  is absolutely continuous with respect to any measure that assigns null measure to any subspace of dimension less than or equal to 1 then the considered problem has a unique optimal solution.

**Optimality** Conditions

### The basic model

$$\inf_{x \in X} \phi(x) := \int_{T} \varphi_t(x) \ d\mu(t), \qquad (P_{\phi}(T))$$

where  $\varphi_t(x) := \gamma_t(x - t)$ ,  $\mu$  a  $\sigma$ -finite, positive measure.

**Result:**  $\varphi_t(x) = \gamma(x-t) \ \forall t \in T$ . The closure of co(T) contains at least an optimal solution of Problem  $P_{\phi}(T)$  if dim(X) = 2 or  $\gamma$  is a norm derived from an inner product when  $dim(X) \ge 3$ .

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**Remark:** These results extend some results proved in [Carrizosa et al. 95] for 2-dimensional spaces and in [Durier85] for finite set of points in  $\mathbb{R}^n$ .

## **Optimality Conditions:**

**Result:** Let X be a separable Banach space, then:

If  $M_{\phi}(T) \neq \emptyset$ ,  $\exists q \in L^{1}(X, X^{*})$  such that  $\int_{T'} q(t)\mu(dt) = 0$ and  $M_{\phi}(T) = \bigcap_{t \in T} \partial \varphi_{t}^{*}(q(t)) = \bigcap_{t \in T} (t + N_{t}(q(t))) := C_{q}(T).$ If  $\exists q \in L^{1}(X, X^{*})$ , such that  $\int_{T} q(t)\mu(dt) = 0$  and  $\bigcap_{t \in T'} \partial \varphi_{t}^{*}(q(t)) \neq \emptyset$  then  $M_{\phi}(T) = \bigcap_{t \in T'} \partial \varphi_{t}^{*}(q(t)) = \bigcap_{t \in T'} (t + N_{t}(q(t))).$ 

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If  $\exists q \in L^1(X, X^*)$ , such that

 $\int_{\mathcal{T}} q(t)\mu(dt) = 0 \text{ and } \bigcap_{\substack{t \in \mathcal{T}' \\ M_{\phi}(\mathcal{T}) = \bigcap_{t \in \mathcal{T}} \partial \varphi_t^*(q(t)) = \bigcap_{t \in \mathcal{T}} (t + N_t(q(t)))}$ 

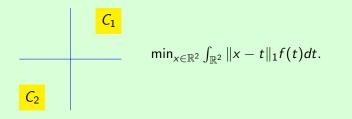
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**Optimality** Conditions

# **Example:** $X = \mathbb{R}^2$ , $\ell_1$ -norm and $f(t) = 1/2\delta_{C_1}(t) + 1/2\delta_{C_2}(t)$ ,

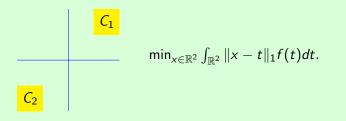


For

$$q(t) = \begin{cases} (-1, -1) & \text{if } t \in C_1 \\ (1, 1) & \text{if } t \in C_2 \\ (0, 0) & \text{if } t \notin C_1 \cup C_2 \end{cases} \cdot \int_{\mathbb{R}^2} q(t)f(t)dt = (0, 0).$$
  
Moreover,  $t + N_{B^0}(q(t)) = \begin{cases} t - \mathbb{R}^2_+ & \text{if } t \in C_1 \\ t + \mathbb{R}^2_+ & \text{if } t \in C_2 \end{cases}$ ; and thus  
$$\bigcap_{t \in C_1 \cup C_2} (t + N_{B^0}(q(t)) = conv\{(-1, -1), (-1, 1), (1, -1), (1, 1)\}.$$

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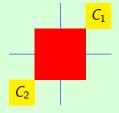


For

$$q(t) = \begin{cases} (-1,-1) & \text{if } t \in C_1 \\ (1,1) & \text{if } t \in C_2 \\ (0,0) & \text{if } t \notin C_1 \cup C_2 \end{cases} \cdot \int_{\mathbb{R}^2} q(t)f(t)dt = (0,0).$$
  
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# **Example:** $X = \mathbb{R}^2$ , $\ell_1$ -norm and $f(t) = 1/2\delta_{C_1}(t) + 1/2\delta_{C_2}(t)$ ,



$$\min_{x\in\mathbb{R}^2}\int_{\mathbb{R}^2} \|x-t\|_1 f(t) dt$$

For

$$\begin{aligned} q(t) &= \begin{cases} (-1,-1) & \text{if } t \in C_1 \\ (1,1) & \text{if } t \in C_2 \\ (0,0) & \text{if } t \notin C_1 \cup C_2 \end{cases} & \int_{\mathbb{R}^2} q(t)f(t)dt = (0,0). \end{aligned}$$

$$\begin{aligned} \text{Moreover, } t + N_{B^0}(q(t)) &= \begin{cases} t - \mathbb{R}^2_+ & \text{if } t \in C_1 \\ t + \mathbb{R}^2_+ & \text{if } t \in C_2 \end{cases}; \text{ and thus} \\ &\bigcap_{t \in C_1 \cup C_2} (t + N_{B^0}(q(t))) = \text{conv}\{(-1,-1),(-1,1),(1,-1),(1,1)\}. \end{aligned}$$

The polyhedral planar case

#### **Extended model**

- $\Phi(\cdot)$  which is a monotone norm on  $\mathbb{R}^M$ .
- $\mu_i \sigma$ -finite, positive measures and  $T \subseteq X$ .
- $\bar{d}_i(x) := \int_T \varphi_t(x) \mu_i(dt)$ , where  $\varphi_t(x) = \gamma_t(x-t)$ .

$$\inf_{x \in X} F(x) := \Phi((\bar{d}_1(x), \dots, \bar{d}_M(x))), \qquad (P_{\Phi}(\Upsilon))$$

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$$\inf_{x \in X} F(x) := \Phi((\bar{d}_1(x), \dots, \bar{d}_M(x))), \qquad (P_{\Phi}(\Upsilon))$$

#### **Properties:**

- Particular instances: center, cent-dian, k-centrum, etc.
- $F = \Phi \circ \overline{D}$  is convex on  $\mathbb{R}^M$ .
- Existence and uniqueness results are still valid.

The polyhedral planar case

**Result:** Let  $x \in X$  be such that  $\overline{D}(x) \neq 0 \in \mathbb{R}^M$ .

 $\beta \in \partial F(x)$  iff  $\exists p_i \in \partial \overline{d}_i(x), \forall i \text{ and } \delta \in \partial \Phi(\overline{D}(x))$ , such that, $\beta = \sum_{i=1}^M \delta_i p_i$ 

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**Definition:** 
$$p = (p_1, \dots, p_M) \in (X^*)^M$$
 and  $I \subseteq \{1, \dots, M\}$ .  
 $\overline{C}_I(p) := \bigcap_{i \in I} \partial \overline{d}_i^*(p_i).$ 

For any  $\delta = (\delta_1, \dots, \delta_M) \ge 0$  $\overline{D}_I(\delta) := \{x : \sum_{i \in I} \delta_i \overline{d}_i(x) = F(x)\}.$ 

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**Definition:**  $(I, \delta, p)$  is a suitable triplet if

#### **Result:**

- If  $M_{\Phi}(\Upsilon) \neq \emptyset$ , ∃(I, δ, p) such that  $M_{\Phi}(\Upsilon) = \overline{C}_{I}(p) \cap \overline{D}_{I}(\delta)$ .
- $(I, \delta, p) \text{ s.t. with } \overline{C}_I(p) \cap \overline{D}_I(\delta) \neq \emptyset, \ M_{\Phi}(\Upsilon) = \overline{C}_I(p) \cap \overline{D}_I(\delta).$

#### Remark:

- We only need to find a suitable triplet  $(I, \delta, p)$  such that  $\overline{C}_I(p) \cap \overline{D}_I(\delta) \neq \emptyset$ .
- From an application point of view, in the case of total polyhedrality this result is specially adequate.

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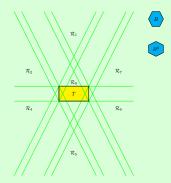
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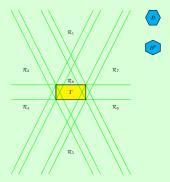
**Result:** There exists a finite partition of  $\mathbb{R}^2$  such that  $\overline{d}(x, T)$  has a common closed form expression on each element of the partition (linear or quadratic).

Example:



*Figura:* Partition of  $\mathbb{R}^2$  generated by the norm  $\gamma$ .

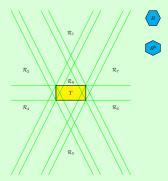
The polyhedral planar case



 $\mu$  is a uniform probability density on T.

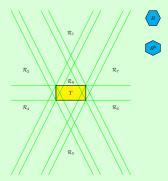
$$\overline{d}(x,T) = \int_T \gamma(x-t)\mu(dt) = \frac{1}{\mu(T)}\int_T \gamma(x-t)\,dt.$$

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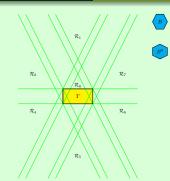
$$egin{aligned} &x\in\mathcal{R}_1:\ \gamma(x-t)=\langle(0,-1),(t_1-x_1,t_2-x_2)
angle.\ &ar{d}(x,T)=rac{1}{8}\int_T\gamma(x-t)\ dt=x_2. \end{aligned}$$

The polyhedral planar case



$$egin{aligned} &x\in \mathcal{R}_2: \ \gamma(x-t) = \langle (1,-0,5), (t_1-x_1,t_2-x_2) 
angle. \ &ar{d}(x,T) = rac{1}{8} \int_T \gamma(x-t) \ dt = -x_1 + rac{x_2}{2} \end{aligned}$$

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$$\begin{array}{ll} x \in \mathcal{R}_8: \ \gamma(x-t) = \\ \left\{ \begin{array}{ll} \langle (1,0,5), (x_1-t_1, x_2-t_2) \rangle & \text{ if } t \in T \text{ and } t_1 \leq \frac{t_2+2x_1-x_2}{2} \\ \langle (0,1), (x_1-t_1, x_2-t_2) \rangle & \text{ if } t \in T \text{ and } \frac{t_2+2x_1-x_2}{2} \leq t_1 \leq \frac{-t_2+2x_1+x_2}{2} \\ \langle (-1,0,5), (x_1-t_1, x_2-t_2) \rangle & \text{ if } t \in T \text{ and } t_1 \geq \frac{-t_2+2x_1+x_2}{2}. \end{array} \right. \\ \text{Thus,} \end{array}$$

$$\bar{d}(x,T) = \frac{1}{8} \int_{T} \gamma(x-t) \, dt = 2\left(x_1 + \frac{x_2}{2}\right) \left(x_1 - \frac{x_2}{2} - 2\right) - \frac{15}{2}$$

### Complexity analysis:

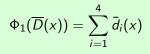
- $O(Mgk_{max})$  lines, the overall complexity of finding the arrangementis  $O((Mgk_{max})^2)$ . [Eldel87].
- *R<sub>j₀</sub>* with *j₀* ∈ *J*, the upper envelope defining Φ has a complexity of at most *O*(*λ*<sub>4</sub>(*r*<sup>0</sup>)).
- Subpartition  $\{\mathcal{R}'_j\}_{j\in J'}\cap \mathcal{R}_{j_0}$  has  $O(\lambda_4(r^0))$  elements.
- The number of elements in the partition induced by the family  $\overline{C}_I(p) \cap \overline{D}_I(\delta)$  is  $O((Mgk_{max})^2\lambda_4(r^0))$ . Moreover, it can be computed in  $O((Mgk_{max})^2\lambda_4(r^0)\log(r^0))$ .

## Remark:

- λ<sub>s</sub>(n) is the maximum length of a Davenport-Schinzel sequence of order s on n symbols. [Sharir and Agarwal].
- $\lambda_1(n) = O(n), \ \lambda_2(n) = O(n), \ \lambda_3(n) = \theta(n\alpha(n)), \ \text{and} \ \lambda_4(n) = \theta(n2^{\alpha(n)}), \ \text{where } \alpha(n) \ \text{is the inverse of the Ackermann function.}$

The polyhedral planar case

#### Example:



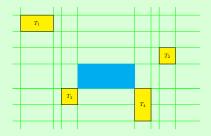
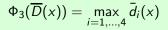


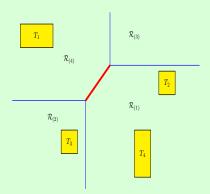
Figura: Minisum problem.

• Taking  $I = \{1, 2, 3, 4\}$ ,  $\delta = (1, 1, 1, 1)$ ,  $p_1 = (1, -1)$ ,  $p_2 = (-1, -1)$ ,  $p_3 = (1, 1)$ ,  $p_4 = (-1, 1)$ ,  $(I, \delta, p)$  is a suitable triplet.  $\overline{D}_I(\delta) = \mathbb{R}^2$ .

The polyhedral planar case

### Example:





• Taking  $I = \{1, 4\}$ ,  $\delta = (\frac{1}{2}, 0, 0, \frac{1}{2})$ ,  $p_1 = (1, -1)$ ,  $p_4 = (-1, 1)$ ;  $(I, \delta, p)$  is a suitable triplet.

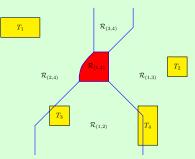
The polyhedral planar case

#### Example:

$$\Phi_4(\overline{D}(x)) = \sum_{i=3}^{4} \overline{d}_{(i)}(x)$$

Л

 $d_{(i)}(x) = d_{\sigma_i}(x)$  with  $\sigma$  a permutation of  $\{1, \ldots, 4\}$ , such that,  $d_{\sigma_1}(x) \leq \ldots \leq d_{\sigma_4}(x)$ .



• Taking  $I = \{1,4\}$ ,  $\delta = (1,0,0,1)$ ,  $p_1 = (1,-1)$ ,  $p_4 = (-1,1)$ ;  $\overline{C}_I(p)$  is the rectangle defined by the closest vertices of  $T_1$ and  $T_4$ .

# Discretization Result:

∀ε > 0 there exist countable sets A ⊆ T, {w<sub>a</sub> ≥ 0}<sub>a∈A</sub>, such that, the solutions of F<sup>\*</sup><sub>W</sub>(A) are ε-solution set of P<sub>φ</sub>(T), with

$$F_W^*(A) = \min_{x \in X} F_{W,A}(x) := \sum_{a \in A} w_a \gamma_a(x - a)$$

$$\min_{x\in X} \Phi(F_{W_1,A}(x),\ldots,F_{W_M,A}(x)), \qquad (P_{\Phi}(A))$$

are  $\varepsilon$ -solution set of Problem  $P_{\Phi}(\Upsilon)$ .

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**Proof:**  $(X, \gamma)$  separable Banach space has a de Possel net.[loffe72]

- $E_a \cap E_{a'} = \emptyset$ ,  $a \neq a'$ , and  $\bigcup_{a \in A} E_a = X$ ;
- $int(E_a) \neq \emptyset$ ,  $E_a \subset cl(int(E_a))$ ,  $a \in A$ ;
- $\sup_{a \in A} diam(E_a) < \varepsilon/(2\mu(T)).$

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**Remark:** If T were a compact set,  $|A| < \infty$ .