

Functional Representations of Combinatorial Sets and Applications in Optimization

Oksana Pichugina

Brock university

Sergey Yakovlev

Ukraine, Kharkov

CONFERENCE ON OPTIMIZATION, TRANSPORTATION AND
EQUILIBRIUM IN ECONOMICS

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Problem Statement

$$f(x) \rightarrow \min, \quad (1)$$

$$x \in E \subset \mathbb{R}^n, \quad (2)$$

$$|E| = N < \infty, \quad (3)$$

E - Euclidean combinatorial set (ECS).

Considered ECS M classes:

1) Boolean vector set (ECS of permutation with repetitions from 0,1):

$$B_n = \left\{ x \in \mathbb{R}^n : x_i \in \{0,1\}, i \in J_n \right\}. \quad (4)$$

2) General permutations set from a multiset G:

$$E_{nk}(G), G = \{g_i\}_{i \in J_n}, |S(G)| = k, J_n = \{1, \dots, n\}. \quad (5)$$

Different Representation of ECS

1.

$$E \subseteq \mathbb{R}^n - \quad (1)$$

set representation of E (SR);

2.

$$E = \{x : f_j(x) = 0, j \in J_{m'}; f_{j+m'}(x) \leq 0, j \in J_{m''}\} - \quad (2)$$

functional representation of E (FR);

$m'' = 0$ - strict FR (SFR);

$m'' > 0$ - nonstrict FR (NSFR);

3. Mixture of (1)-(2)

$E = \{x \in E^* \subseteq \mathbb{R}^n : f_j(x) = 0, j \in J_{m'}; f_{j+m'}(x) \leq 0, j \in J_{m''}\}$ set-functional representation of E (SFR).

Some Representation of B_n

$$1. \text{ FR: } B_n = \{x \in \mathbb{R}^n : x_i^2 - x_i = 0, i \in J_n\}; \quad (1)$$

$$2. \text{ SFR: } B_n = \{x \in \Pi_n : \sum_{i=1}^n (1 - \cos(2\pi x_i)) = 0\}, \quad (2)$$

$$\Pi_n = \{x \in \mathbb{R}^n : \bar{0} \leq x \leq \bar{1}\}. \quad (3)$$

Penalty fuctions (PF) examples:

$$\circ \text{ FR (1) - } \Phi(x, \lambda) = f(x) + \lambda \cdot \sum_{j=1}^{m'} f_j^2(x) \xrightarrow{\lambda > 0} \min$$

\circ SFR (2) -

$$\Phi(x, \lambda) = f(x) + \lambda \cdot \left(\sum_{j=1}^{m'} f_j^2(x) + \sum_{j=1}^{m''} \max(0, -f_{j+m'}(x)) \right) \xrightarrow{\lambda > 0} \min$$

Approach 1: Combinatorial Cutting Plane Method for vertex located ECS (CCPM)

Let E be vertex located:

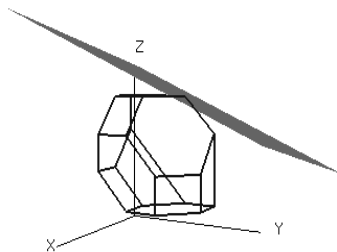
$$E = \text{vert } P, \quad (1)$$

$$P = \text{conv } E - \quad (2)$$

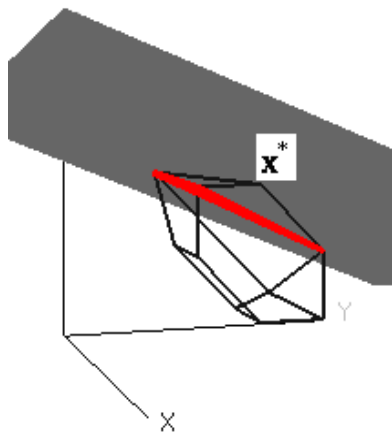
combinatorial polyhedron (CP)

$$cx \rightarrow \min_E$$

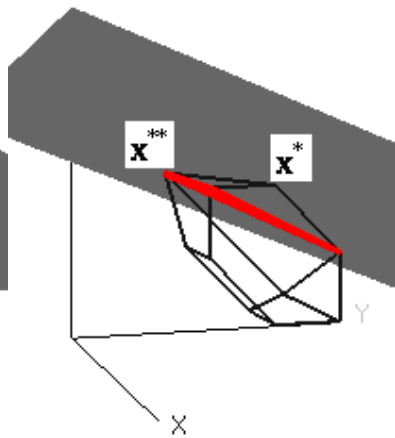
$$Ax \leq b, A = (a_i)_{i \in J_m}.$$



CCPM (Part 2)



Pic.1



Pic.2

Representations of ECS inscribed into a sphere

$$E \subset S_r(a), \quad (1)$$

$$S_r(a) = \{x \in \mathbb{R}^n : (x - a)^2 = r^2\}. \quad (2)$$

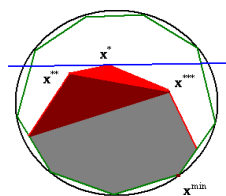
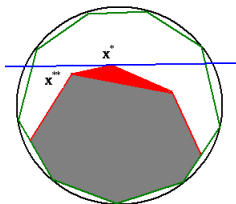
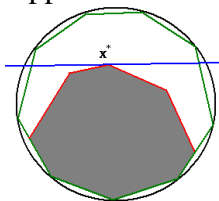
$$E = P \cap S_r(a) \quad (3)$$

$$P = \{x : A'x \leq b'\} \quad (4)$$

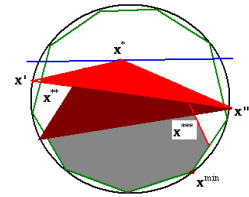
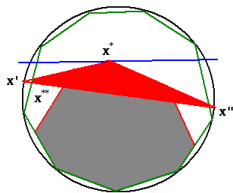
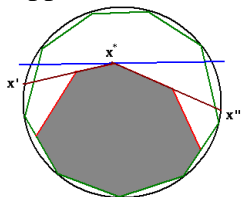
- (3)- SR of E;
- (2)-(3), (3)-(4)- SFR of E;
- (2)-(4)- FR of E.

Approach 2: Modification of CCPM for ECS inscribed into a sphere

Approach 1



Approach 2



Approach3: Polyhedral Spherical Method (PSM) for ECS inscribed into a sphere

Representation 1- $f(x) \rightarrow \min_{x \in E}$ (1)

Representation 2- $f(x) \rightarrow \min_{x \in P \cap S_r(a) \subset \mathbb{R}^n}$ (2)

Relaxation 1

$$f(x) \rightarrow \min_{x \in P \subset \mathbb{R}^n} \quad (3)$$

Relaxation 2

$$f(x) \rightarrow \min_{x \in S_r(a) \subset \mathbb{R}^n} \quad (4)$$

Remark. We can think that $f(x)$ is convex, because it is known that any optimization problem on the vertex located set can be converted to the optimization of its convex extension in \mathbb{R}^n .

PSM (Part 2 – Additional problems)

Subproblem 1.

$$cx \xrightarrow{E} \min \left(cx \xrightarrow{P} \min \right) \quad (1)$$

Subproblem2.

$$f(x) \xrightarrow{P} \min \quad (2)$$

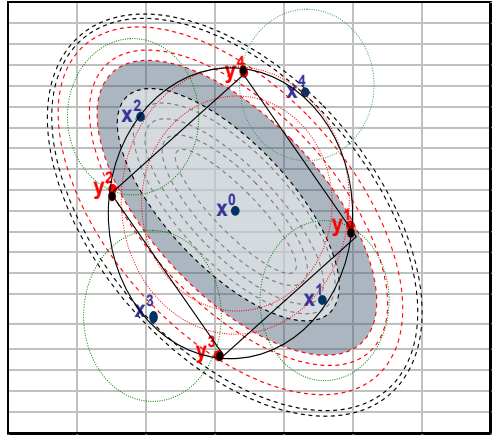
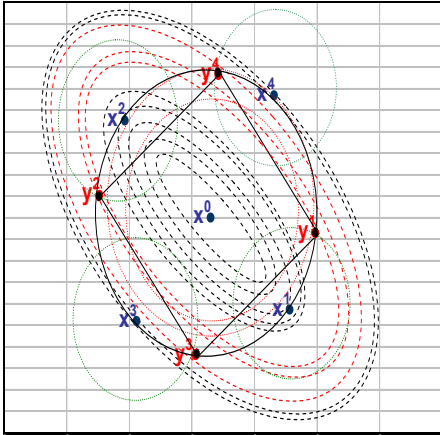
Subproblem 3.

$$f(x) \xrightarrow{S_r(a)} \min \quad (3)$$

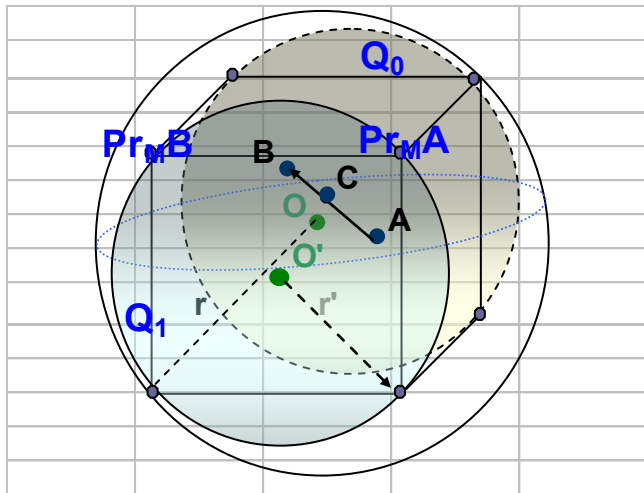
Subproblem4. Projection on ECS

$$x' = \text{Pr}_E y \quad (4)$$

PSM (Part 3 – Quadratic optimization in \mathbb{R}^2)



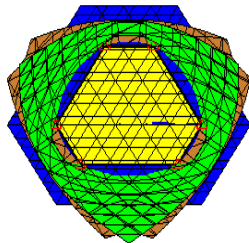
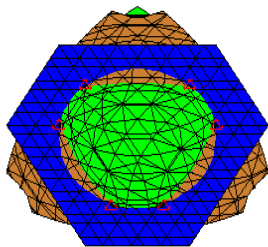
PSM (Part 4 – Optimization over B_3)



Functional Representation of $E_{nk}(G)$, $n=3$

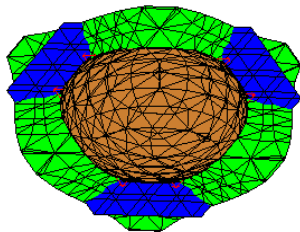
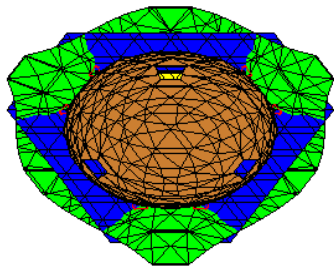
$$E = M, M = \left\{ x : \sum_{i=1}^n x_i^j = \sum_{i=1}^n g_i^j, j \in J_n \right\} \quad (1)$$

Example 1. $k=n=3$, $E=E_3(G)$, $G \geq 0$, convex FR (1)



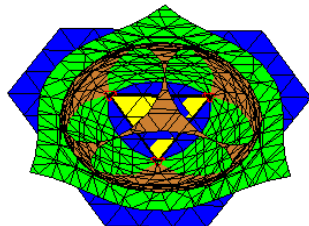
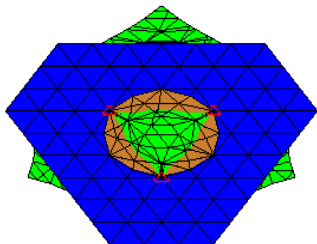
Functional Representation of $E_{nk}(G)$, $n=3$ (Part 2)

Example 2. $k=n=3$, $E=E_3(G)$, $G \geq 0$, nonconvex FR with r_{\min}



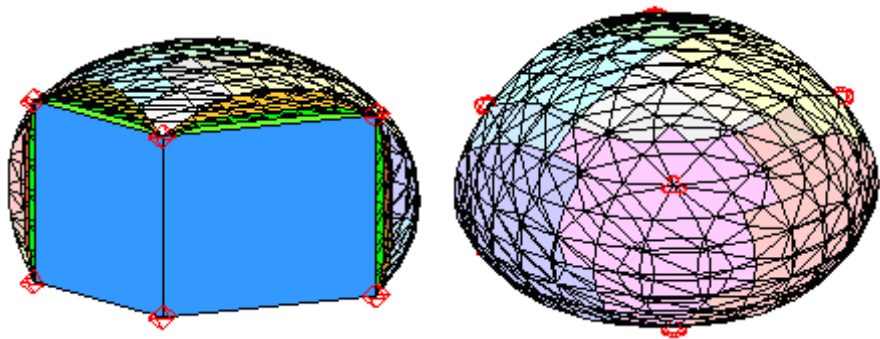
Functional Representation of $E_{nk}(G)$, $n=3$ (Part 3)

Example 3. $k=2$, $E=E_{32}(G)$, $G \geq 0$, nonconvex FR with r_{\min}



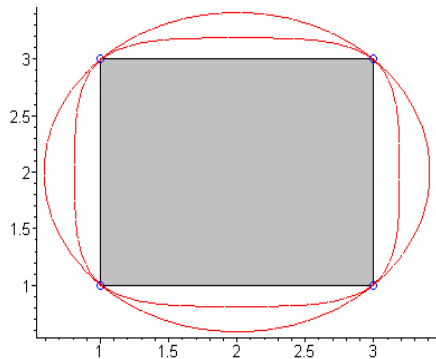
Functional Representation of B_n

$$E = M, M = \left\{ x : \sum_{i=1}^n (x_i - \frac{1}{2})^{2j} = \frac{n}{2^{2j}}, j \in J_n \right\}$$

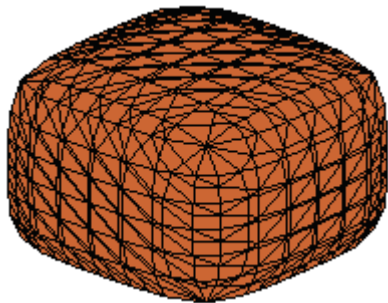


Irredundant Functional Representation of B_n

$$E = M, M = \left\{ x : \sum_{i=1}^n (x_i - \frac{1}{2})^2 = \frac{n}{4}, \sum_{i=1}^n (x_i - \frac{1}{2})^4 = \frac{n}{16} \right\}$$



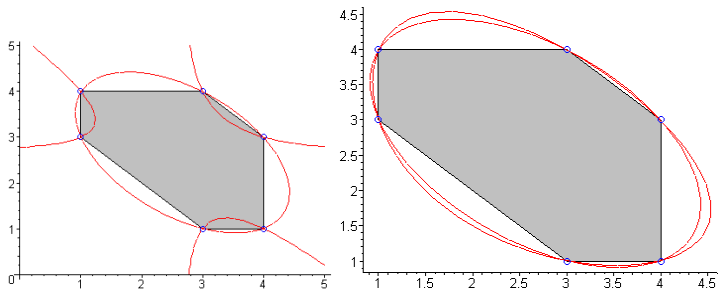
Pic. 1 $n = 2$



Pic. 2 $n = 3$ Inscribed surface in $S_r(a)$

Intersectional and Tangent FR

1. Intersectional FR $m' = n, m \geq n$



2. Tangent FR: $E = S' \cap S_r(a)$.

$$S' = \{x \in \mathbb{R}^n : h(x) = 0\}$$

$$\Phi(x, \lambda) = f(x) + \lambda \cdot \left(\left((x - a)^2 - r^2 \right)^2 + h^2(x) \right) \xrightarrow{\lambda > 0} \min \cdot$$

Thank you for your attention!