

# Extended and Embedding Formulations for MINLP

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Mathematisches Forschungsinstitut Oberwolfach,  
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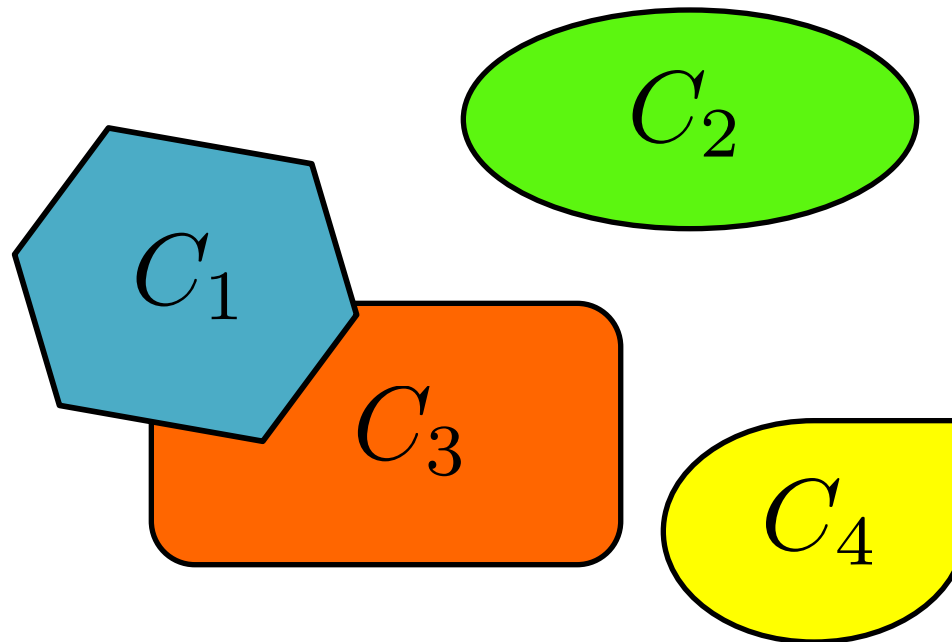
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# Nonlinear Mixed 0-1 Integer Formulations

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- Modeling Finite Alternatives = Unions of Convex Sets

$$x \in \bigcup_{i=1}^n C_i \subseteq \mathbb{R}^d$$



# Extended and Non-Extended Formulations for $\bigcup_{i=1}^n C_i$

$$C_i = \{x \in \mathbb{R}^d : f_i(x) \leq 0\}$$

Extended

$$\begin{aligned} \tilde{f}_i(x^i, y_i) &\leq 0 && \forall i \in [n] \\ \sum_{i=1}^n x^i &= x \\ \sum_{i=1}^n y_i &= 1 \\ y &\in \{0, 1\}^n \\ x, x^i &\in \mathbb{R}^d && \forall i \in [n] \end{aligned}$$

Strong, but large

Non-Extended

$$\begin{aligned} f_i(x) &\leq M_i(1 - y_i) && \forall i \in [n] \\ \sum_{i=1}^n y_i &= 1 \\ y &\in \{0, 1\}^n \\ x &\in \mathbb{R}^d && \forall i \in [n] \end{aligned}$$

Small, but weak?

# Outline

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- Extended Formulations
  - Conic formulations, stability and outer-approximation
- Non-Extended Formulations
  - Embedding formulations = Strong non-extended

Extended Formulations:

Birdmen: Or (The Unexpected Virtue of **Discipline**)

# Extended Formulations: Perspective “v/s” Cones

- e.g. Ceria and Soares '99

$$C_i = \{x \in \mathbb{R}^d : f_i(x) \leq 0\}$$

$$\tilde{f}(x, y) = \begin{cases} yf(x/y) & \text{if } y > 0 \\ \lim_{\alpha \downarrow 0} \alpha f(x' - x + x/\alpha) & \text{if } y = 0 \\ +\infty & \text{if } y < 0 \end{cases}$$

$$\begin{aligned} \tilde{f}_i(x^i, y_i) &\leq 0 && \forall i \in [n] \\ \sum_{i=1}^n x^i &= x \\ \sum_{i=1}^n y_i &= 1 \\ y &\in \{0, 1\}^n \\ x, x^i &\in \mathbb{R}^d && \forall i \in [n] \end{aligned}$$

- e.g. Ben-tal and Nemirovski '01, Helton and Nie '09

$$C_i = \left\{ x \in \mathbb{R}^d : \begin{array}{l} \exists u \in \mathbb{R}^{p_i} \text{ s.t.} \\ A^i x + D^i u - b \in K^i \end{array} \right\}$$

$K^i$  closed convex cone

$$\begin{aligned} A^i x^i + D^i u^i - b y_i &\in K^i && \forall i \in [n] \\ \sum_{i=1}^n x^i &= x \\ \sum_{i=1}^n y_i &= 1 \\ y &\in \{0, 1\}^n \\ x, x^i &\in \mathbb{R}^d && \forall i \in [n] \\ u^i &\in \mathbb{R}^{p_i} && \forall i \in [n] \end{aligned}$$

- Both formulations are **ideal** (extreme points of continuous relaxation satisfy integrality constraints)

# Cones Can Mitigate Unintended Numerical Issues

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- Let  $C_i = \{x \in \mathbb{R}^2 : f_i(x) \leq 0\}$   
where  $f_i(x) = x_1^2 - x_2 - 1$

$$\tilde{f}_i(x, y) = \begin{cases} y(x_1/y)^2 - x_2 - y & \text{if } y > 0 \\ -x_2 & \text{if } y = x_1 = 0 \\ +\infty & \text{if o.w.} \end{cases}$$

- Conic (SOCP) representation

$$C_i = \left\{ x \in \mathbb{R}^2 : \sqrt{x_2^2 + 4x_1^2} \leq 2 + x_2 \right\}$$

$$\sqrt{(x_2^i)^2 + 4(x_1^i)^2} \leq 2y_i + x_2$$

# Conic = Really Extended

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- Conic representation = additional auxiliary variables

$$C_i = \left\{ x \in \mathbb{R}^d : \begin{array}{l} \exists u \in \mathbb{R}^{p_i} \text{ s.t.} \\ A^i x + D^i u - b \in K^i \end{array} \right\}$$

- Bad for NLP solvers, but good for polyhedral approximations in MINLP solvers:
  - Separable = (Tawarmalani and Sahinidis '05, Hijazi et al. '12)
  - SOCP = (V. et al. '15) = If SOCP representable use MI-SOCP solver!
  - General MINLP? = **Disciplined** Convex Programming (DCP) (Grant et al. 06): Implemented in CVX
    - Systematic way to prove convexity
    - Yields extended conic representation (MINLPLib2 -1)



# Solving Mixed Integer Disciplined Convex Programs

- **Pajarito** solver!
  - Lubin, Yamangil, Bent and V. '15
  - SLayer10M ( **it** / **time** ):
    - Bonmim = **69** it / **1,379** s
    - Hijazi et al. = **23** it / **14** s
    - Pajarito = **5** it / **12** s  
(automatic separability)
  - Solved gams01 from MINLPLIB2 (prev 91% GAP)
  - Solved t1s5-6 (prev 25%-29% GAP) = Just SOCP + Gurobi!
  - ~200 lines of Julia code



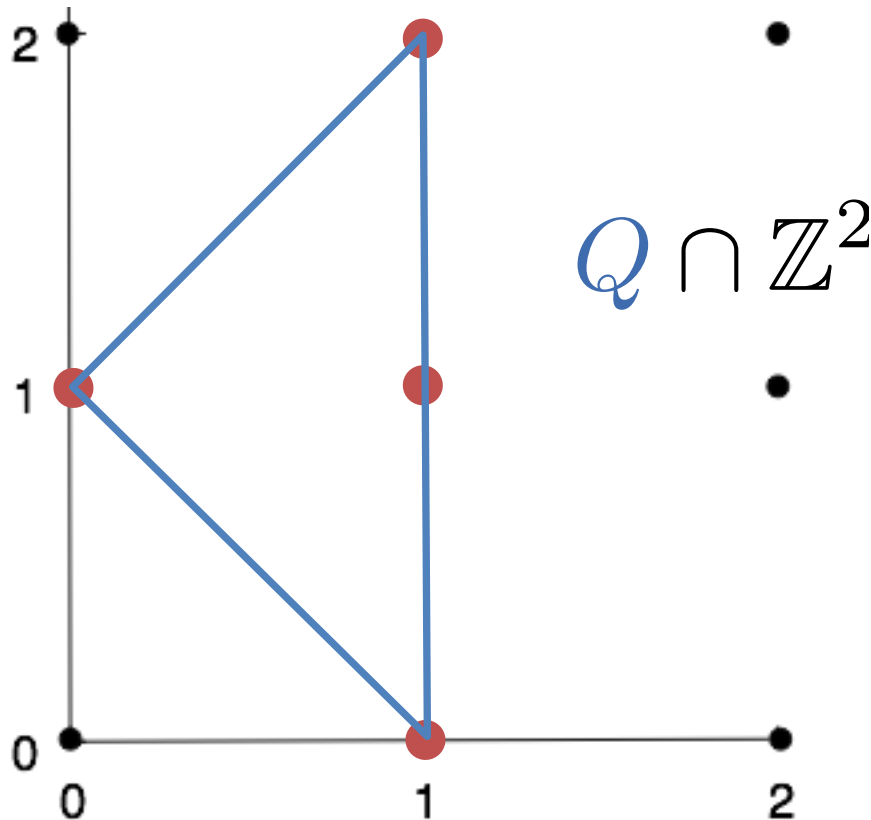
Miles Lubin and Emre Yamangil

Strong Non-Extended Formulations:  
Minkowski Sums, Good or Evil

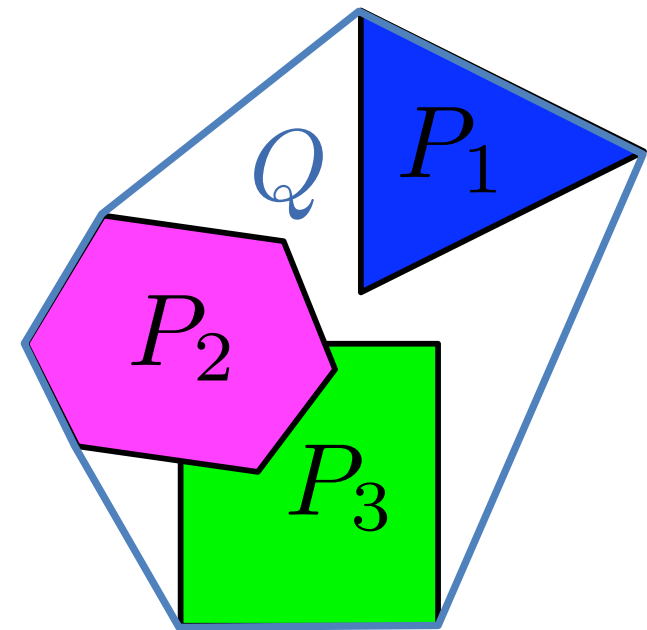
# Constructing Non-extended Ideal Formulations

- Pure Integer:

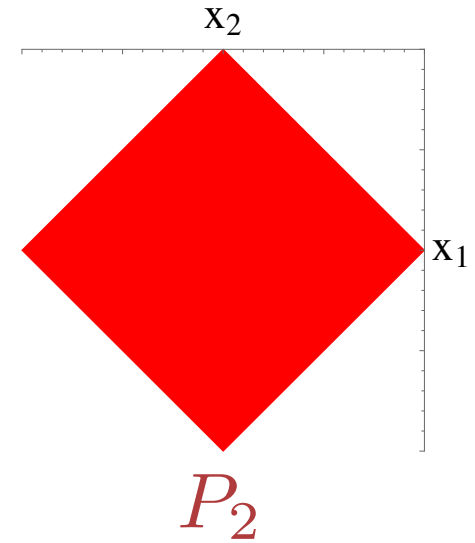
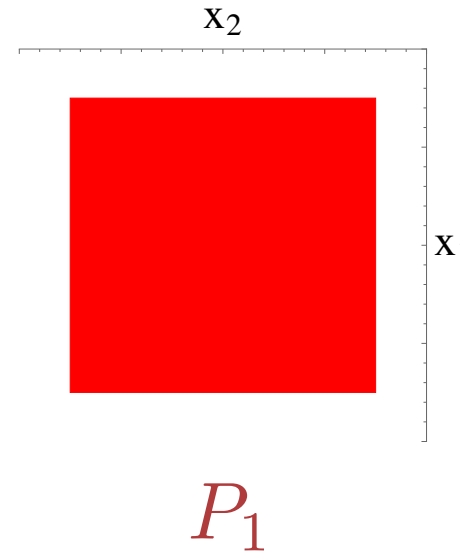
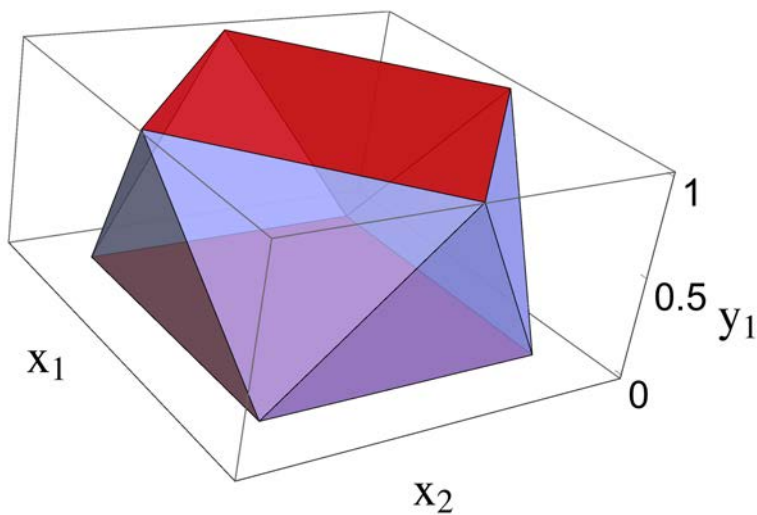
$$Q := \text{conv} \left( \{p^i\}_{i=1}^n \right)$$



- Mixed Integer:



# Embedding Formulation = Ideal non-Extended



$$Q(H) := \text{conv} \left( \bigcup_{i=1}^n P_i \times \{h^i\} \right)$$

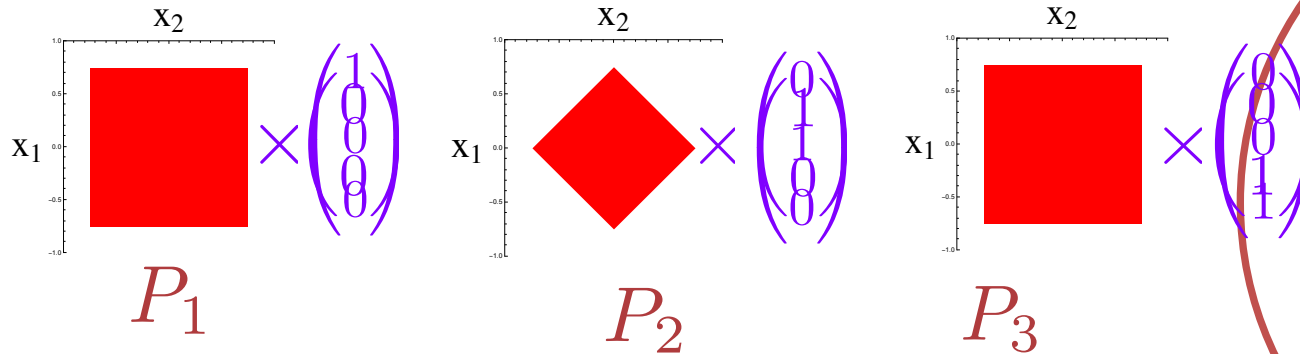
$$(x, y) \in Q \cap (\mathbb{R}^d \times \mathbb{Z}^k) \iff y = h^i \wedge x \in P_i$$

$$\text{ext}(Q) \subseteq \mathbb{R}^d \times \mathbb{Z}^k$$

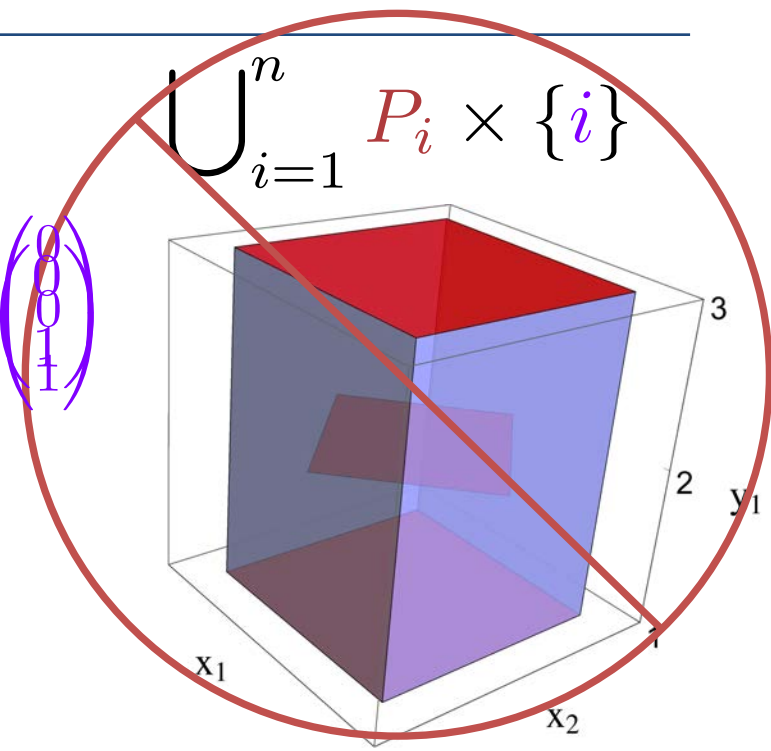
$$H := \{h^i\}_{i=1}^n \subseteq \{0, 1\}^k, \quad h^i \neq h^j$$

# Alternative Encodings

- “Only” use 0-1 encodings



$$\bigcup_{i=1}^n P_i \times \{i\}$$



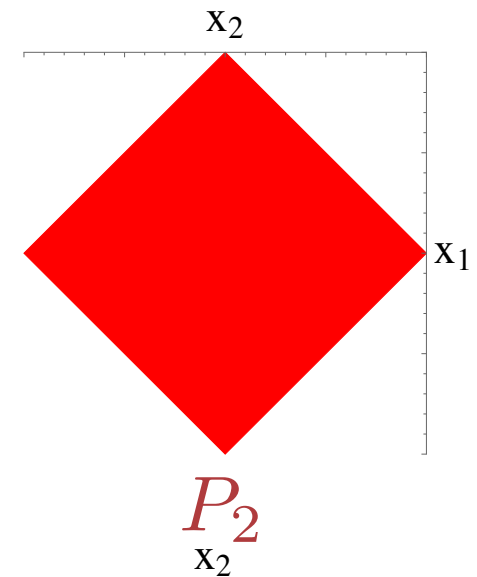
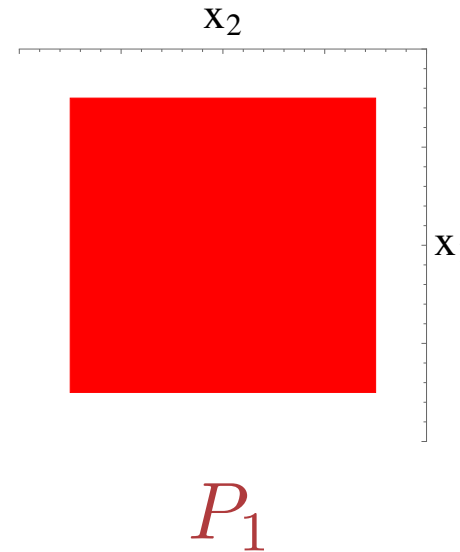
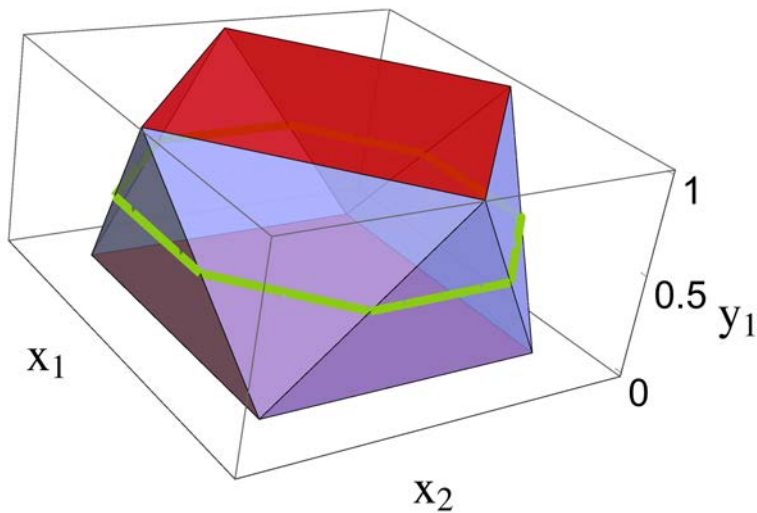
- Options for 0-1 encodings:
  - Traditional or **Unary** encoding

$$H = \left\{ y \in \{0, 1\}^n : \sum_{i=1}^n y_i = 1 \right\} = \{e^i\}_{i=1}^n$$

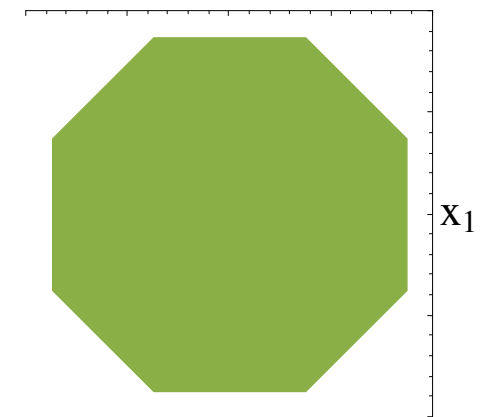
$$e_j^i = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$

- **Binary** encodings:  $H \equiv \{0, 1\}^{\log_2 n}$
- Others (e.g. **incremental** encoding  $\equiv$  unary)

# Unary Encoding, Minkowski Sum and Cayley Trick



$$Q \cap (\mathbb{R}^2 \times \{0.5\}) \equiv P_1 + P_2 =$$



For traditional or unary encoding:

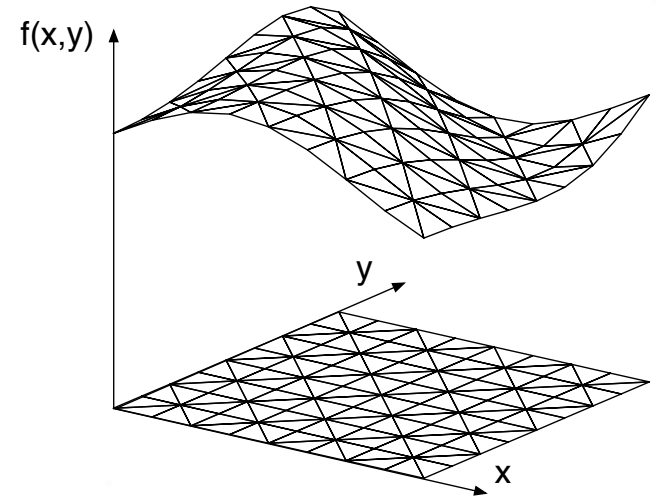
$$Q(H) \cap (\mathbb{R}^d \times \left\{ \frac{1}{n} \sum_{i=1}^n \mathbf{e}^i \right\}) \equiv \sum_{i=1}^n P_i$$

# Encoding Selection Matters: Evil Minkowski Sum

- Size of unary formulation is:  
(Lee and Wilson '01)

$$\left( \begin{array}{c} 2\sqrt{n/2} \\ \sqrt{n/2} \end{array} \right) + \left( \sqrt{n/2} + 1 \right)^2$$

↑ General Inequalities      ↑ Variable Bounds



- Size of one binary formulation:  
(V. and Nemhauser '08)

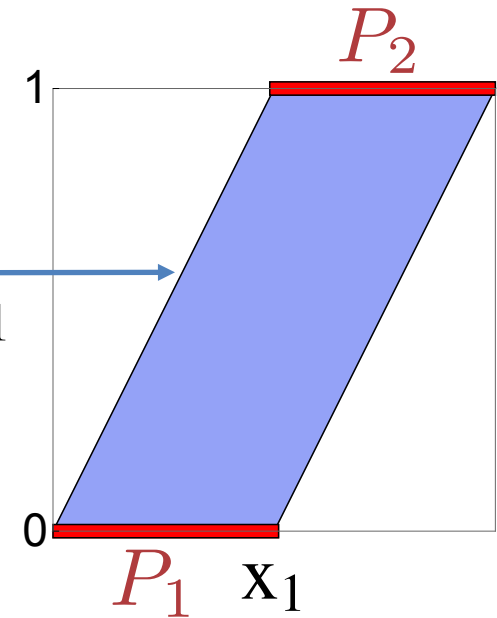
$$4 \log_2 \sqrt{n/2} + 2 + \left( \sqrt{n/2} + 1 \right)^2$$

- Right embedding = significant computational advantage over alternatives (Extended, Big-M, etc.)

# Complexity of Family of Polyhedra $\mathcal{P} := \{P_i\}_{i=1}^n$

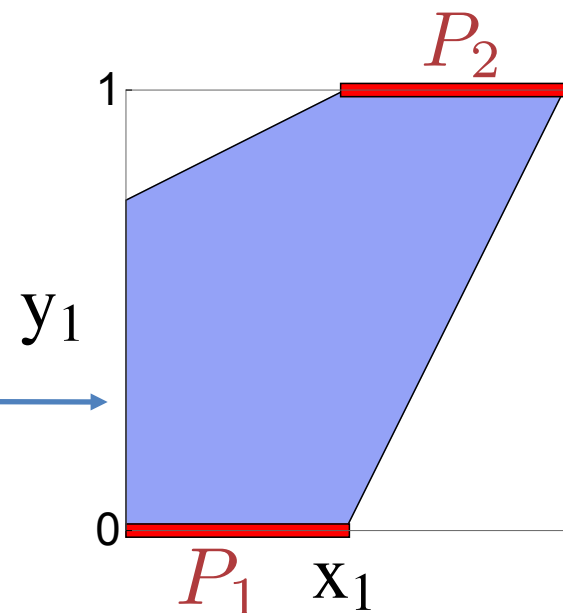
- Embedding complexity = smallest **ideal** formulation

$$\text{mc}(\mathcal{P}) := \min_H \{\text{size}(Q(H))\}$$



- Relaxation complexity = smallest formulation

$$\text{rc}(\mathcal{P}) := \min_{Q,H} \{\text{size}(Q)\}$$





# Complexity Results

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- Lower and Upper bounds for special structures:
  - e.g. for Special Order Sets of Type 2 (SOS2) on  $n$  variables

- Embedding complexity (ideal)

$$2^{\lceil \log_2 n \rceil} \longleftarrow \text{General Inequalities}$$

$$n + 1 \leq \dots \leq n + 1 + 2^{\lceil \log_2 n \rceil} \longleftarrow \text{Total}$$

- Relaxation complexity (non-ideal)

$$2 \leq \dots \leq 4 \longleftarrow \text{General Inequalities}$$

$$2 \leq \dots \leq 5 + 2n \longleftarrow \text{Total}$$

- Relation to other complexity measures

$$\text{hc}(\mathcal{P}) := \text{size} \left( \text{conv} \left( \bigcup_{i=1}^n P_i \right) \right)$$

$$\text{xc}(\mathcal{P}) := \min_R \left\{ \text{size}(R) : \text{proj}_x(R) = \text{conv} \left( \bigcup_{i=1}^n P_i \right) \right\}$$

- Still open questions (see V. 2015)

# Faces for Unary Encoding: Good Minkowski Sum

- Two types of facets (or faces):

- $P_1 \times \{0\} \equiv y_i \geq 0$

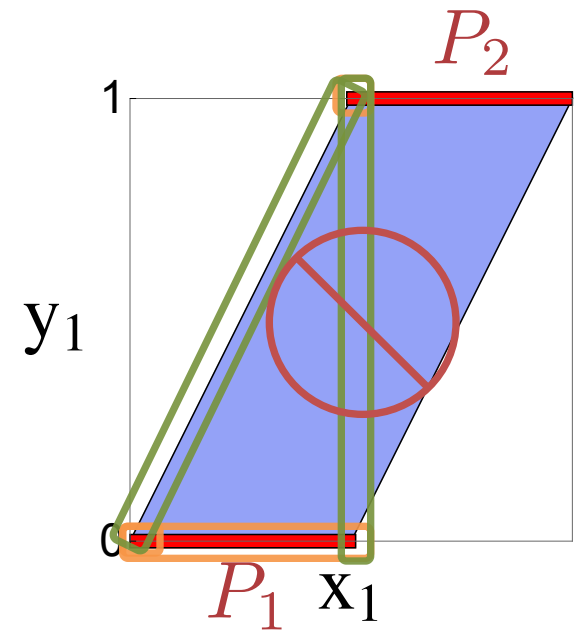
- $\text{conv}((F_1 \times 0) \cup (F_2 \times 1))$

$F_i$  proper face of  $P_i$

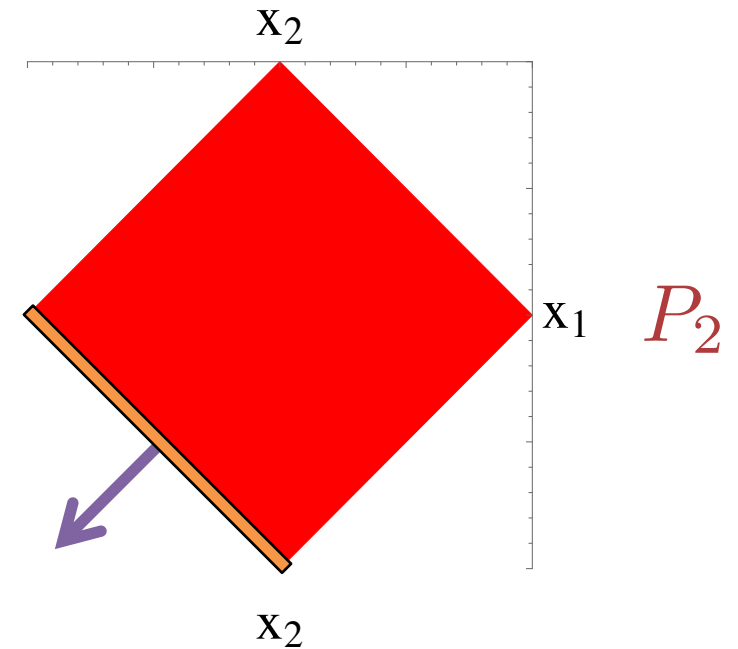
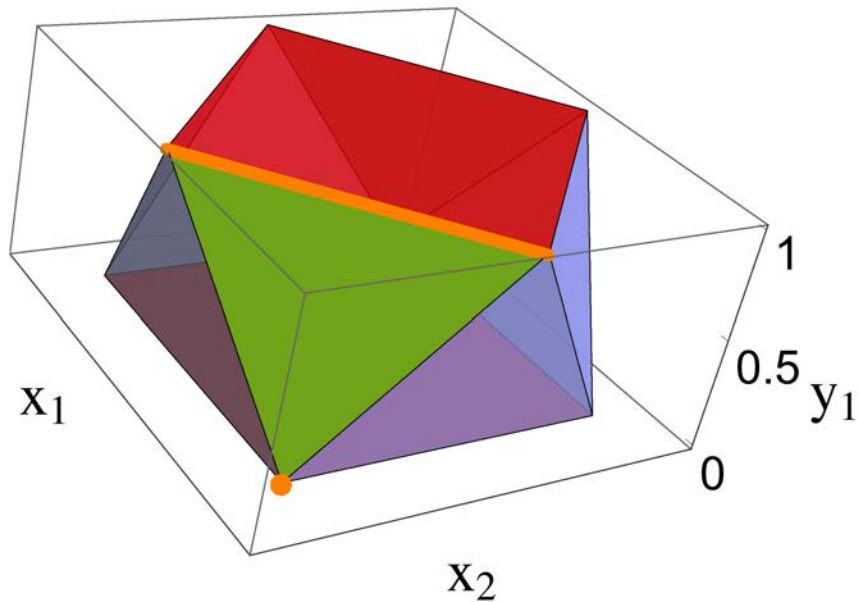
- Not all combinations of faces

- Which ones are valid?

- Minkowski to the rescue!



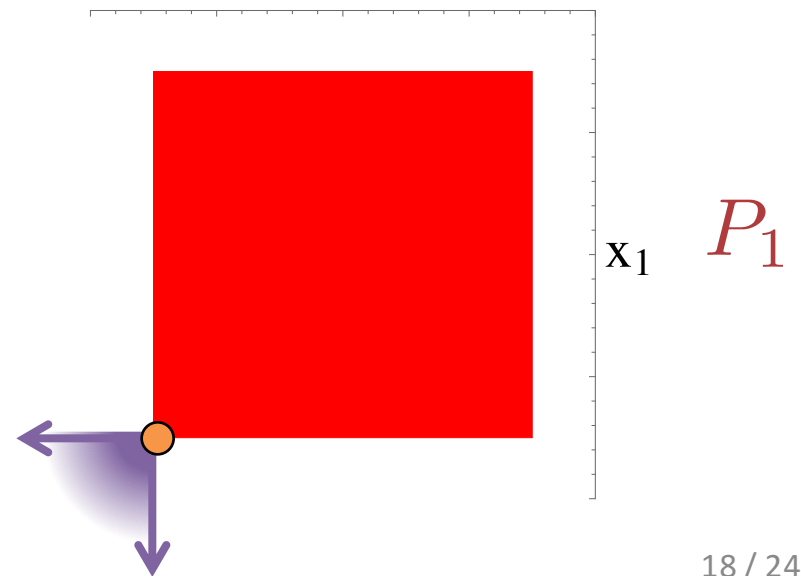
# Valid Combinations = Common Normals



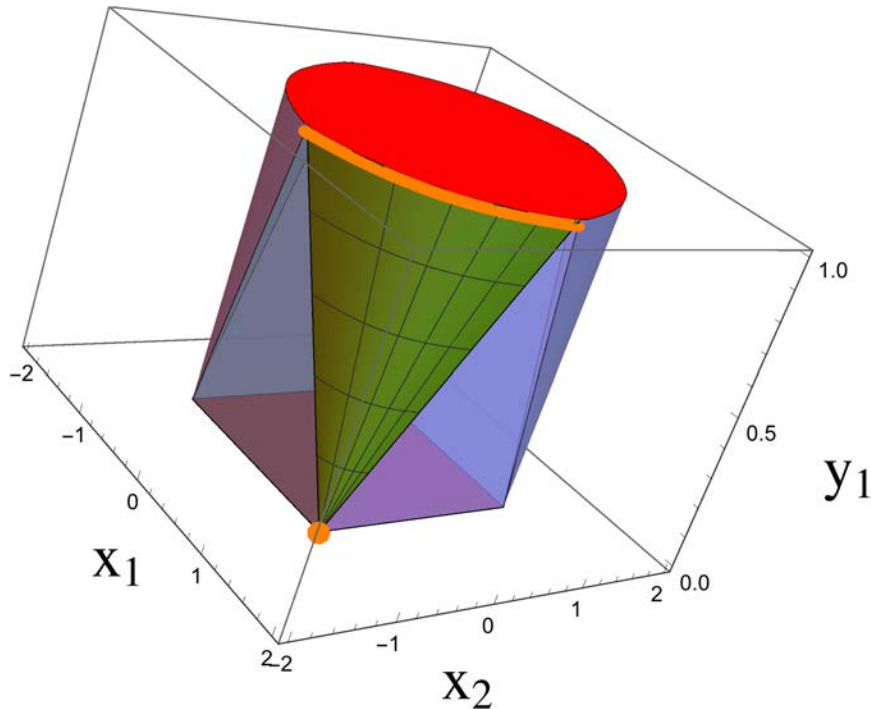
$$N(F_1) \cap N(F_2) \neq \emptyset$$



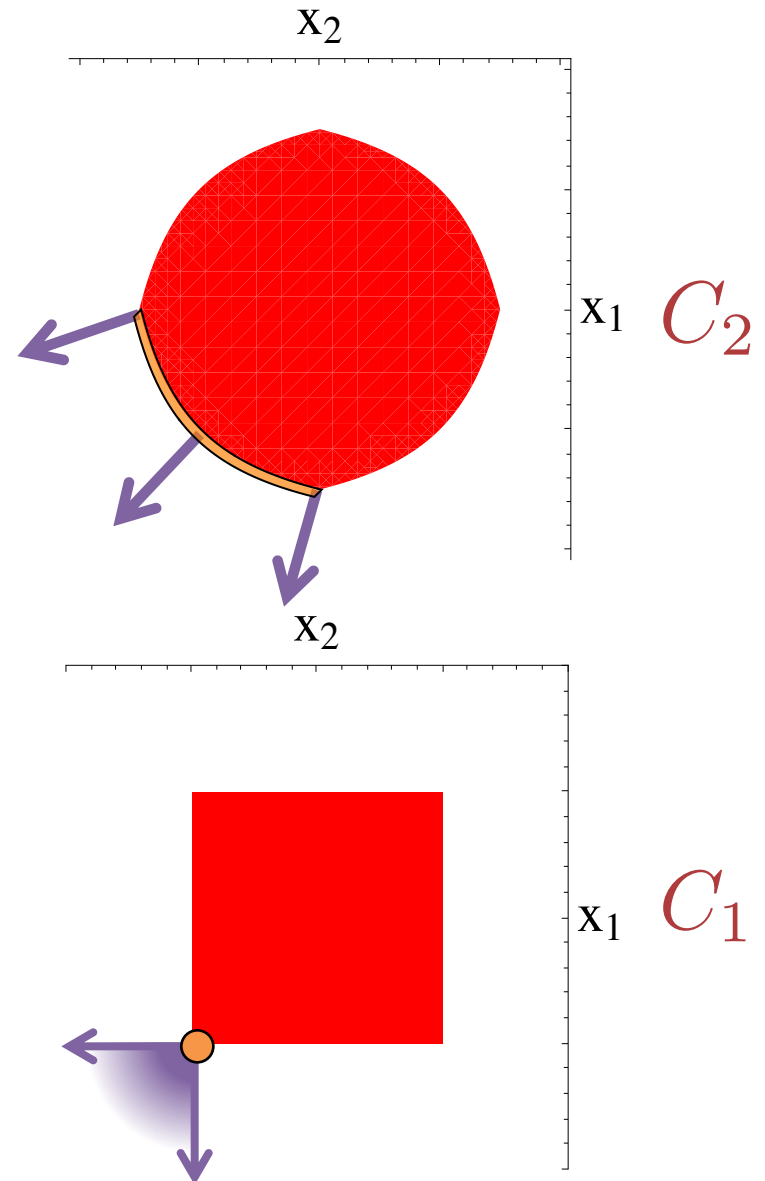
$\text{conv}((F_1 \times 0) \cup (F_2 \times 1))$   
is face of  $Q(H)$



# Unary Embedding for Unions of Convex Sets

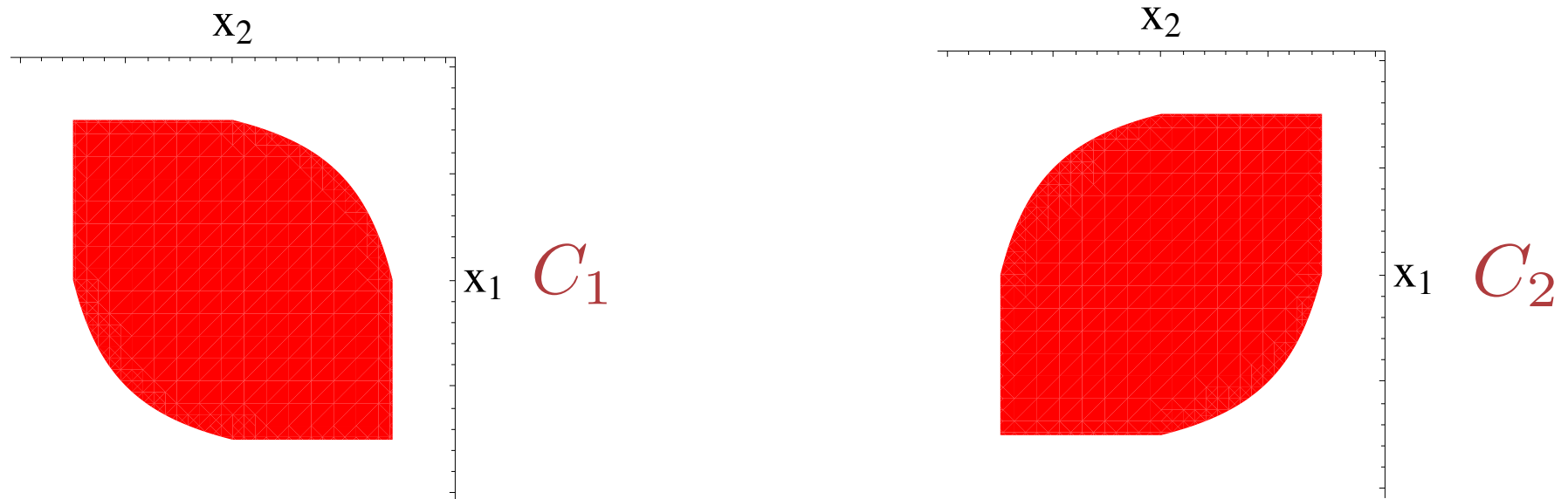


- Description of boundary of  $Q(H)$  is easy if “normals condition” yields convex hull of **1** nonlinear constraint and point(s)



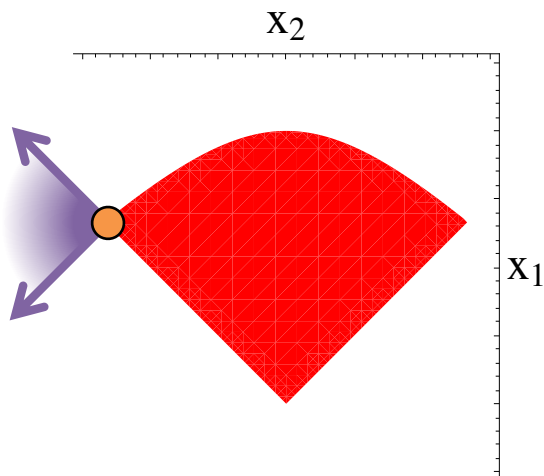
# Easy to Recover and Generalize Existing Results

- Isotone function results from Hijazi et al. '12 and Bonami et al. '15 ( $n=1, 2$ ):
  - $C_i = \{x \in \mathbb{R}^d : l^i \leq x \leq u^i, f_i(x) \leq 0\}$
- Can generalize to  $n \geq 3$  and **two** functions per set:



- Other special cases (previous slide)

# Also “Non-isotone” Results: Pizza Slices

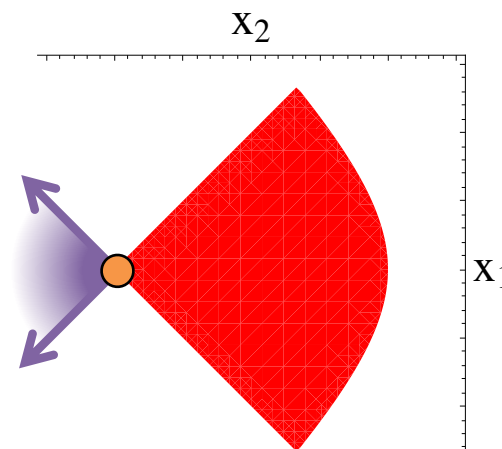


$C_1 :$

$$\sqrt{1 + x_1^2} \leq 2 - x_2$$

$$x_1 - x_2 \leq 1$$

$$-x_2 - x_1 \leq 1$$



$C_2 :$

$$\sqrt{1 + x_2^2} \leq 2 - x_1$$

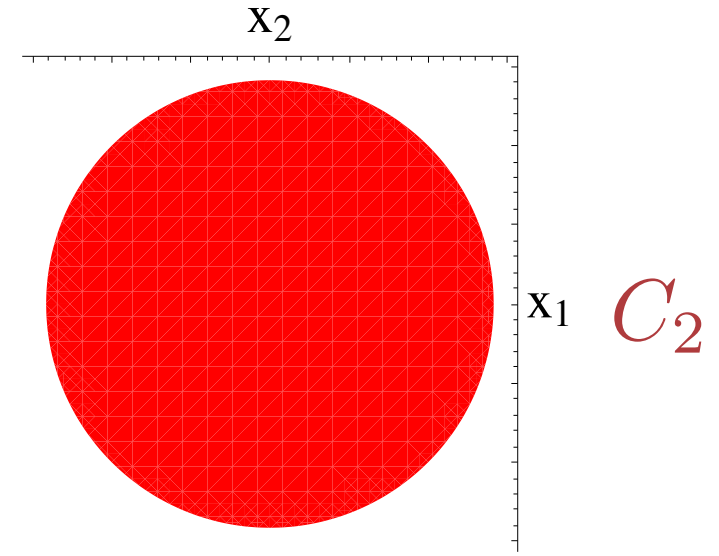
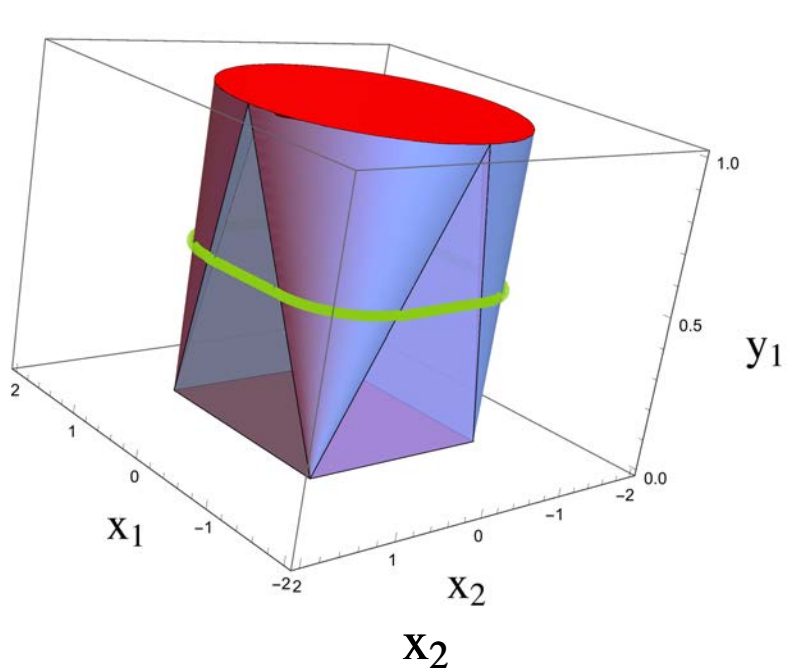
$$x_2 - x_1 \leq 1$$

$$-x_2 - x_1 \leq 1$$

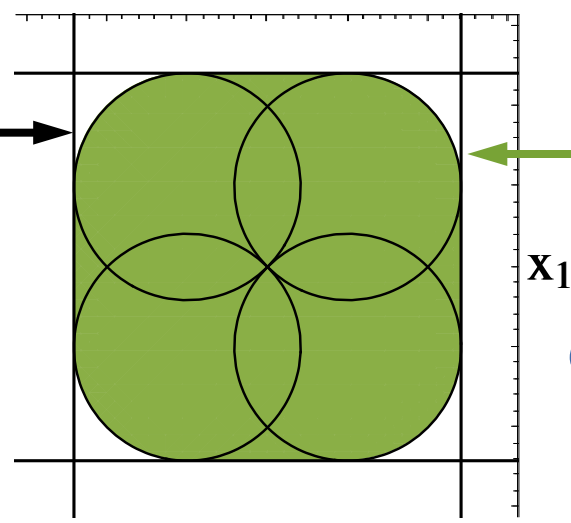
$$\sqrt{y_4^2 + \left(x_2 - \frac{1}{3}y_1 + \frac{1}{3}y_3\right)^2} \leq 2y_4 + x_1 + \frac{4}{3}y_1 + y_2 + \frac{4}{3}y_3$$

$$\text{conv} \left( \bigcup_{i=1}^4 (C_i \times \{e^i\}) \right) = 4 \text{ conic} + 4 \text{ linear inequalities}$$

# Bad Example: Representability Issues



Zariski closure of boundary? →



Description with finite number of (quadratic) polynomial inequalities?

$Q := \text{conv}((C_1 \times \{0\}) \cup (C_2 \times \{1\}))$   
 can fail to be basic semi-algebraic

# Final Positive Results

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- Unions of Homothetic Convex Bodies  $C_i = \lambda_i C + b^i$   
(all extreme points exposed)

$$\text{conv} \left( \bigcup_{i=1}^n (C_i \times \{e^i\}) \right) =$$

$$\gamma_C \left( x - \sum_{i=1}^n y_i b^i \right) \leq \sum_{i=1}^n \lambda_i y_i$$

$$\sum_{i=1}^n y_i = 1$$

$$y \geq 0 \quad \forall i \in [n]$$

$$\gamma_C(x) := \inf \{ \lambda > 0 : x \in \lambda C \}$$

- Generalizes polyhedral results from Balas '85, Jeroslow '88 and Blair '90



# Summary

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- Extended formulations
  - Really extended formulations through SOCP or DCP
  - Pajarito = MIDCP / MINLP extended polyhedral solver
- Embedding formulations = systematic procedure for ideal non-extended formulations
  - Polyhedral case = Formulations and complexity
  - Non-Polyhedral 1 = Simplified proofs, extensions and new formulations
  - Non-Polyhedral 2 = Representability issues