Complexity of branch-and-bound and cutting planes in mixed-integer optimization

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Joint work with Amitabh Basu Michele Conforti Hongyi Jiang

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 BB and CP are (among) the main general-purpose techniques for mixed-integer optimization, but little is known on their relative strength

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- Show that Branch-and-Cut (BC) can be "exponentially better" than BB and CP alone

#### Motivation:

- BB and CP are (among) the main general-purpose techniques for mixed-integer optimization, but little is known on their relative strength
- Computationally, BC tends to be far more efficient and effective than BB and CP alone

## The setting

 $\begin{array}{ll} \min \ c^{\mathsf{T}}x\\ \mathsf{s.t.} \ x\in C\\ x\in S \end{array}$ 

where

- $C \subseteq \mathbb{R}^n$  is a closed convex set
- $S \subseteq \mathbb{R}^n$  models some non-convexity

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Typical case: (mixed) integer linear programming:

 $\begin{array}{l} \min \ c^{\mathsf{T}}x \\ \text{s.t.} \ Ax \leq b \\ x \in \mathbb{Z}^n \end{array}$ 

C is a polyhedron {x ∈ ℝ<sup>n</sup> : Ax ≤ b}
S = ℤ<sup>n</sup>

#### Disjunctions

Variable disjunction:  $D = \{x : x_i \leq b \text{ or } x_i \geq b + 1\}$ , where  $b \in \mathbb{Z}$ 



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Split disjunction:  $D = \{x : a^{\mathsf{T}}x \leq b \text{ or } a^{\mathsf{T}}x \geq b+1\}$ , where  $a \in \mathbb{Z}^n$  and  $b \in \mathbb{Z}$ .



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General disjunction: a finite union of polyhedra that cover S.







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We assume that the **best-node** strategy is used: then the first feasible solution found is optimal.







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We compare the number of nodes (length) produced by these algorithms based on the same families of disjunctions, assuming optimal choices.

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Proofs are stronger than algorithms, even in dimension 2 (Owen & Mehrotra 2001).

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Variable dim.				
Fixed dim.				

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	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$			
Fixed dim.	BB <i>O</i> (1) CP <i>O</i> (1)		BB <i>O</i> (1) CP <i>O</i> (1)	

### 0/1 convex sets, variable disjunctions

#### Theorem (Dash 2003/Chvátal 1973)

Let  $P \subseteq [0,1]^n$  be a polytope. If a valid inequality for  $P \cap \mathbb{Z}^n$  has a BC proof/algorithm of length N based on variable disjunctions, then it has a CP proof/algorithm of length N based on variable disjunctions.
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### Theorem (BCDJ 2022)

Let  $C \subseteq [0,1]^n$  be a closed convex set. If a valid inequality  $cx \leq \gamma$  for  $C \cap \mathbb{Z}^n$  has a BC proof/algorithm of length N based on variable disjunctions, then  $cx \leq \gamma + \epsilon$  has a CP proof/algorithm of length N based on variable disjunctions, for any  $\epsilon > 0$ .

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#### Question

Can  $\epsilon$  be removed?

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	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$			
Fixed dim.	BB <i>O</i> (1) CP <i>O</i> (1)		BB <i>O</i> (1) CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable	$CP \leq BB$			
dim.	CP poly(n) vs BB exp(n)			
Fixed dim.	BB <i>O</i> (1)		BB <i>O</i> (1)	
	CP <i>O</i> (1)		CP <i>O</i> (1)	

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(This example can be made less pathological.)

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Variable	$CP \leq BB$			
dim.	CP poly(n) vs BB exp(n)			
Fixed dim.	BB <i>O</i> (1)		BB <i>O</i> (1)	
	CP <i>O</i> (1)		CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable	$CP \leq BB$			
dim.	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
Fixed dim.	BB <i>O</i> (1)		BB <i>O</i> (1)	
	CP <i>O</i> (1)		CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$	$\begin{array}{c} BB \ \mathcal{O}(1) \\ vs \\ CP \ \infty \end{array}$		
	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
Fixed dim.	BB <i>O</i> (1)		BB <i>O</i> (1)	
	CP <i>O</i> (1)		CP <i>O</i> (1)	

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- The best BB tree has 4 nodes.
- CP only converges in infinitely many iterations.

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$	BB $O(1)$ vs CP $\infty$		
	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
	BB <i>O</i> (1)		BB <i>O</i> (1)	
Fixed dim.	CP <i>O</i> (1)		CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$	BB $O(1)$ vs CP $\infty$		
	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
	BB <i>O</i> (1)		BB <i>O</i> (1)	
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$	BB $O(1)$ vs CP $\infty$		
	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
	BB <i>O</i> (1)	BB poly(CP)	BB <i>O</i> (1)	
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable	$CP \leq BB$	BB $O(1)$ vs CP $\infty$	$BB \leq 3 \cdot CP$	$BB \leq 3 \cdot CP$
dim.	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
	BB <i>O</i> (1)	BB poly(CP)	BB <i>O</i> (1)	$BB \leq 3 \cdot CP$
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	

# General split disjunctions

### Theorem (BCDJ 2022)

Let C be a closed convex set. If a valid inequality for C has a CP proof of length N based on general split disjunctions, then it has a BB proof of length 3N based on general split disjunctions.

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This can be extended to arbitrary disjunctions, provided that all split disjunctions are included.

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable	$CP \leq BB$	BB $O(1)$ vs CP $\infty$	BB ≤ 3⋅CP	$BB \leq 3 \cdot CP$
dim.	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		
	BB <i>O</i> (1)	BB poly(CP)	BB <i>O</i> (1)	$BB \leq 3 \cdot CP$
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	

	variable disjunctions		split disjunctions	
	0/1 sets	general sets	0/1 sets	general sets
Variable dim.	$CP \leq BB$	$\begin{array}{c} BB  \mathit{O}(1) \\ vs \\ CP  \infty \end{array}$	BB ≤ 3⋅CP	$BB \leq 3 \cdot CP$
	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		BB O(1) vs CP poly(data)
	BB <i>O</i> (1)	BB poly(CP)	BB <i>O</i> (1)	$BB \leq 3 \cdot CP$
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	BB O(1) vs CP poly(data)

### Theorem (Conforti, Del Pia, DS, Faenza, Grappe 2015)

For general polytopes and general split disjunctions in fixed dimension, there are examples in which BB takes in O(1) iterations while CP needs poly(data) iterations.

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#### Question

Is there an exponential-gap instance?

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	CP poly(n) vs BB exp(n)	CP poly(n) vs BB exp(n)		BB O(1) vs CP poly(data)
	BB <i>O</i> (1)	BB <i>poly</i> (CP)	BB <i>O</i> (1)	$BB \leq 3 \cdot CP$
Fixed dim.	CP <i>O</i> (1)	BB $O(1)$ vs CP $\infty$	CP <i>O</i> (1)	BB <i>O</i> (1) vs CP <i>poly</i> (data)

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Variable dim.	$CP \leq BB$	BB $O(1)$ vs CP $\infty$	$BB \leq 3 \cdot CP$	$BB \leq 3 \cdot CP$
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#### Definition

A branching scheme based on a family of disjunction  $\mathcal{D}$  and a CP paradigm are complementary if there is a family of instances where CP gives polynomial size proofs and the shortest BB proof based on  $\mathcal{D}$  is exponential, and there is another family where the opposite happens.

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Example: BB based on variable branching and Chvátal–Gomory cuts.

#### Theorem (BCDJ 2022; here informal)

Under the above complementarity assumption, there are instances where BC does exponentially better than BB and CP alone.

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