

Structural Analysis of Branch-and-Cut and the Learnability of Gomory Mixed Integer Cuts

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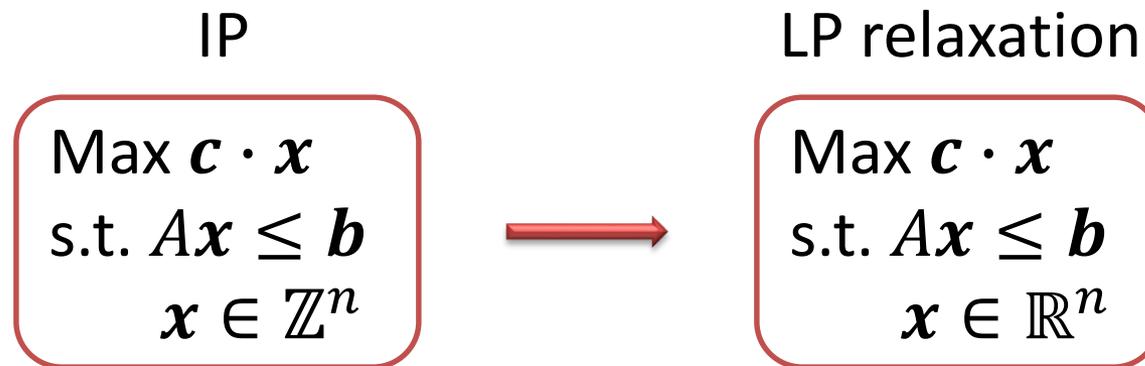
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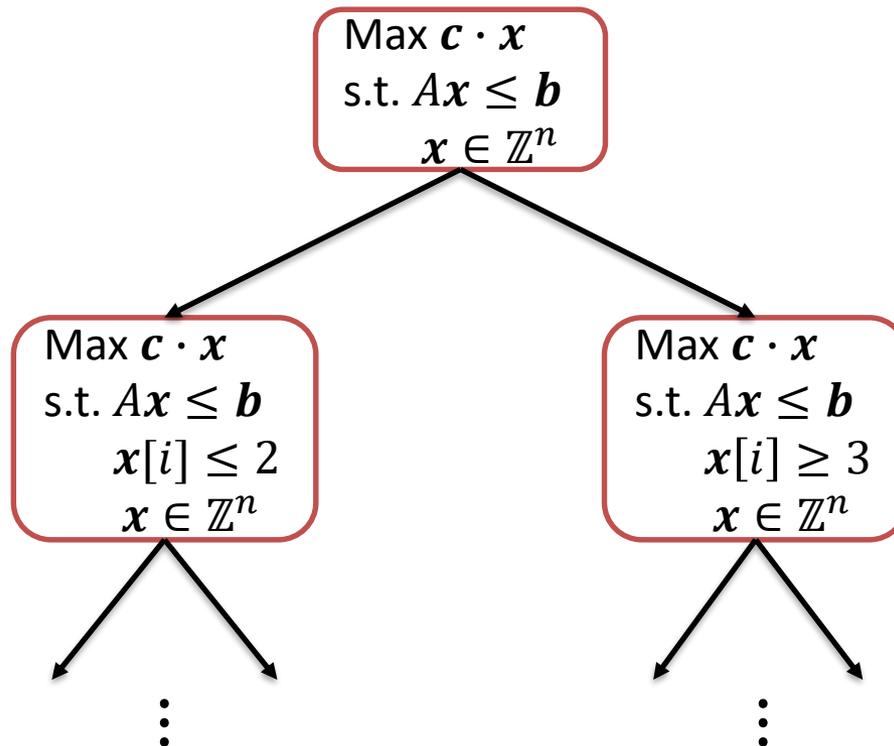
Branch-and-bound

- Powerful tree-search algorithm used to solve IPs in practice
- Uses the linear programming (LP) relaxation to do an informed search through the set of feasible integer solutions



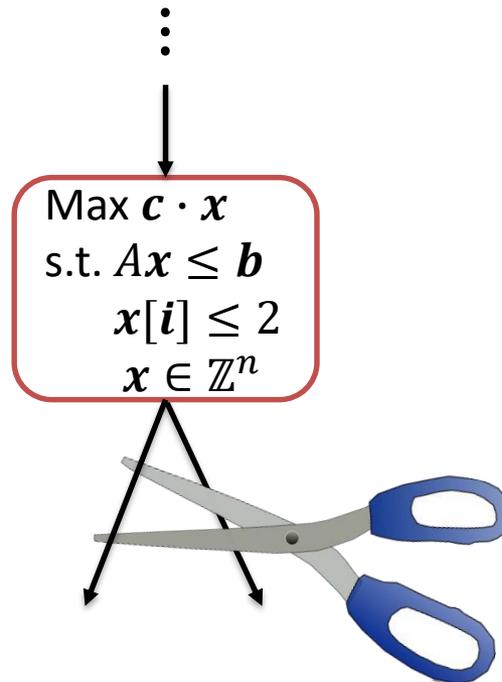
Branch-and-bound: branching

- Choose variable i to branch on.
- Generate one subproblem with $x[i] \leq \lfloor x_{LP}^*[i] \rfloor$ another with $x[i] \geq \lceil x_{LP}^*[i] \rceil$



Branch-and-bound: pruning

- Prune subtrees if
 - LP relaxation at a node is integral, infeasible, or
 - (Bounding) LP optimal *worse* than best feasible integer solution found so far

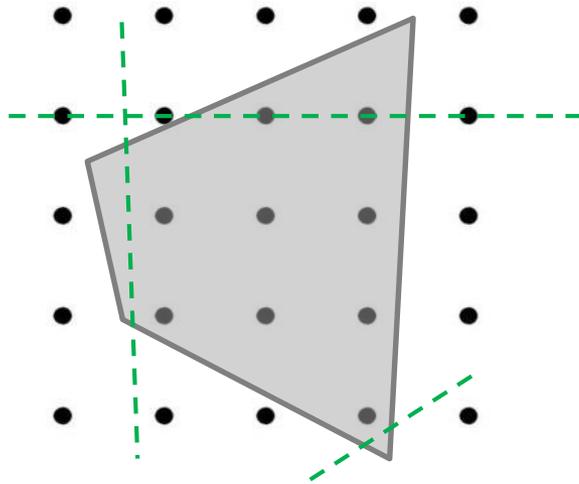


Branch-and-cut

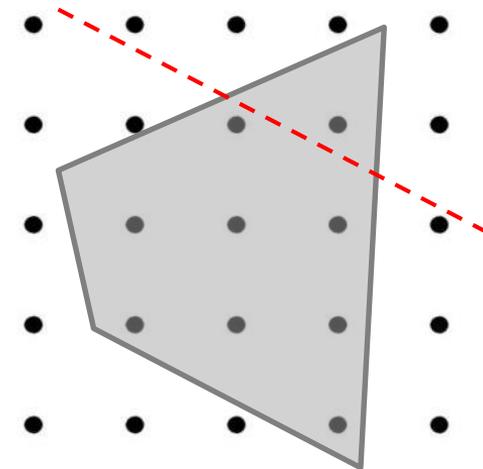
- Branch-and-bound, but at each node may add *cutting planes*
- Method of getting tighter LP relaxation bounds, and thus pruning subtrees sooner

Cutting planes

- Constraint $\alpha x \leq \beta$ is a *valid cutting plane* if it does not cut off any integer feasible points



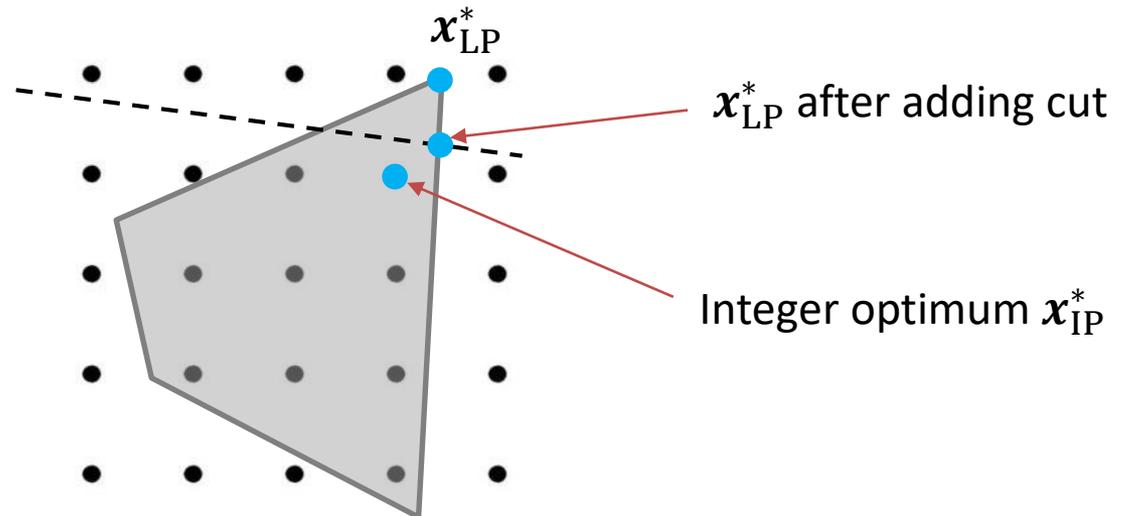
Valid cutting planes



An invalid cutting plane

Cutting planes

- If $\alpha x \leq \beta$ is valid and separates the LP optimum, can speed up B&C by pruning nodes sooner



Gomory Mixed Integer Cuts

$$\sum_{i:f_i \leq f_0} f_i x_i + \frac{f_0}{1 - f_0} \sum_{i:f_i > f_0} (1 - f_i) x_i \geq f_0$$

- *Gomory Cuts Revisited*, 1996. Balas, Ceria, Cornuéjols, Natraj
 - achieved remarkable speedups by effectively integrating these cuts into branch-and-bound (believed to be practically useless before)
- Today: Gomory mixed integer cuts a crucial component of commercial solvers like CPLEX and Gurobi



Main contributions

- The first generalization guarantees for using machine learning to add Gomory mixed integer cuts
- A novel structural analysis of the branch-and-cut algorithm that pins down its possible behaviors

Generalization guarantees for cutting planes

Distribution-dependent cut selection

Learning to cut

If a cut yields small average branch-and-cut tree size over IP samples...

$$\begin{array}{c} \text{Max } \mathbf{c}_1 \cdot \mathbf{x} \\ \text{s.t. } A_1 \mathbf{x} \leq \mathbf{b}_1 \\ \mathbf{x} \in \mathbb{Z}^n \end{array} \quad \dots \quad \begin{array}{c} \text{Max } \mathbf{c}_N \cdot \mathbf{x} \\ \text{s.t. } A_N \mathbf{x} \leq \mathbf{b}_N \\ \mathbf{x} \in \mathbb{Z}^n \end{array} \quad \sim D$$

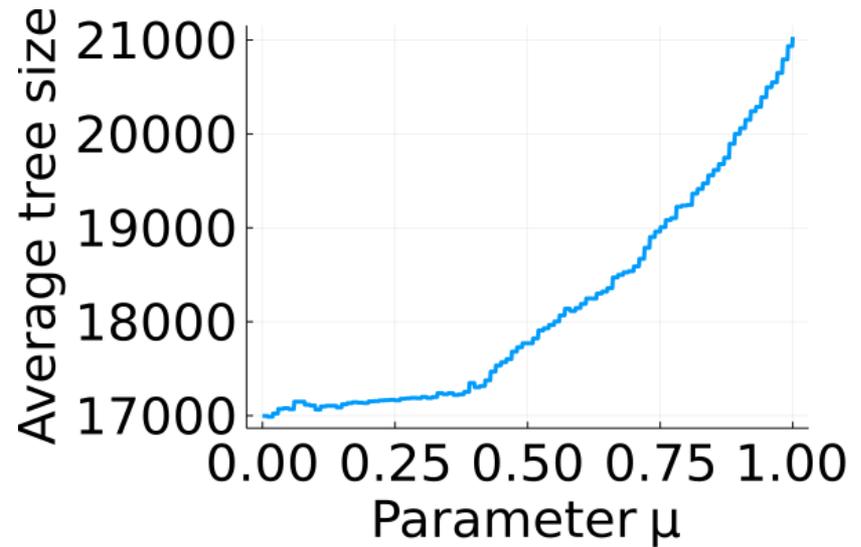
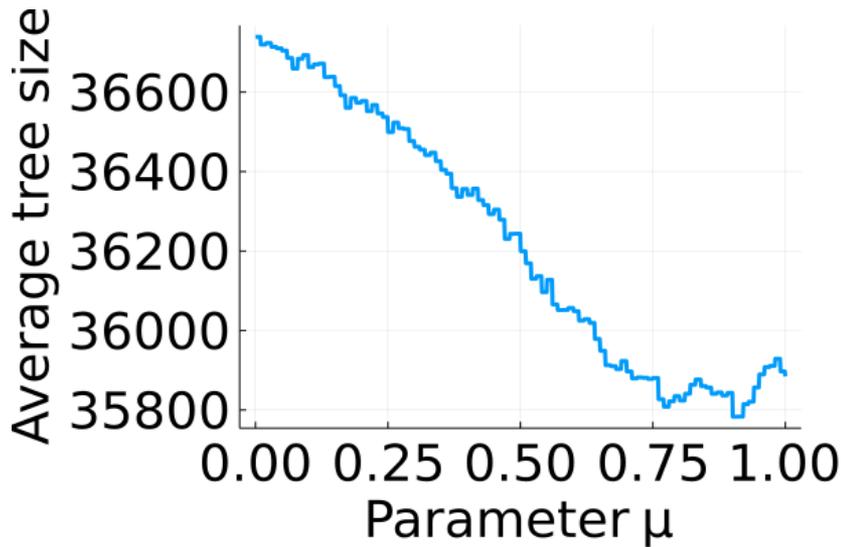
IP 1 IP N

...is it likely to yield a small branch-and-cut tree on a fresh IP?

$$\begin{array}{c} \text{Max } \mathbf{c} \cdot \mathbf{x} \\ \text{s.t. } A \mathbf{x} \leq \mathbf{b} \\ \mathbf{x} \in \mathbb{Z}^n \end{array} \quad \sim D$$

Tuning a GMI cut selection parameter

- E.g. mixture of $d = 2$ scores
 $\mu \cdot \text{parallelism} + (1 - \mu) \cdot \text{efficacy}$



Two different distributions over facility location IPs.

Generalization for Gomory mixed integer cuts

Theorem [Balcan, Prasad, Sandholm, Vitercik NeurIPS'22]: for all GMI cuts $\mathbf{u} \in [-U, U]^m$, difference between average training performance over N samples and expected performance is (whp)

$$\tilde{O} \left(H \sqrt{\frac{mn^3 \log(mn\tau U \|A\|_{1,1} \|\mathbf{b}\|_1)}{N}} \right)$$

Proof uses our structural analysis of branch-and-cut

A structural analysis of branch-and-cut

- Given two valid cutting planes

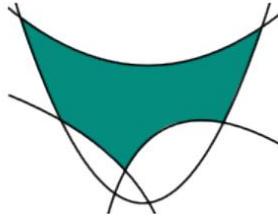
$$\alpha_1 x \leq \beta_1 \text{ and } \alpha_2 x \leq \beta_2$$

- When does B&C behave identically on the following IPs?

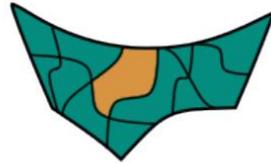
$$\begin{array}{l} \text{Max } c \cdot x \\ \text{s.t. } Ax \leq b \\ \alpha_1 x \leq \beta_1 \\ x \in \mathbb{Z}^n \end{array}$$

$$\begin{array}{l} \text{Max } c \cdot x \\ \text{s.t. } Ax \leq b \\ \alpha_2 x \leq \beta_2 \\ x \in \mathbb{Z}^n \end{array}$$

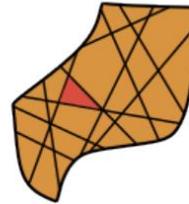
Branch-and-cut piecewise invariance



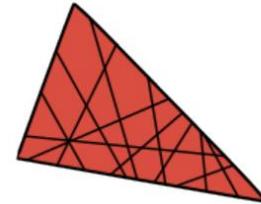
(a) LP optimum closed form



(b) Invariant branching



(c) Invariant LP integrality



(d) Invariant B&C execution

Theorem [BPSV NeurIPS'22]: Given IP $(\mathbf{c}, A, \mathbf{b})$, there are $O(14^n (m + 2n)^{3n^2} \tau^{5n^2})$ polynomial hypersurfaces of degree ≤ 5 that partition \mathbb{R}^{n+1} into connected components such that the branch-and-cut tree built after adding the cut $\alpha \mathbf{x} \leq \beta$ is invariant over all (α, β) within a given component.

$$\tau := \left\lceil \max_{\mathbf{x} \in P} \|\mathbf{x}\|_\infty \right\rceil \leq \|A\|_{\infty, \infty}^n n^{n/2}$$

If A has a row with all positive entries, then $\tau \leq \|\mathbf{b}\|_\infty$