

# An Algorithm for Solving Inverse Integer Programs

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- 1 Inverse IP
- 2 Complexity
- 3 Algorithm
- 4 Computational Results



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- 3 Algorithm
- 4 Computational Results



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- 2 Complexity
- 3 Algorithm
- 4 Computational Results



# Definitions

We consider an IP

$$z_{IP} = \min_{x \in \mathcal{P}} d^T x, \quad (1)$$

where  $d \in \mathbb{R}^n$  and

$$\mathcal{P} = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\} \cap (\mathbb{Z}^r \times \mathbb{R}^{n-r}).$$

For a given  $c \in \mathbb{R}^n$ ,  $x^0 \in \mathcal{P}$ , the inverse problem is defined as follows.

$$\begin{aligned} & \min \|c - d\| \\ & \text{s.t.} \\ & d^T x^0 \leq d^T x \quad \forall x \in \mathcal{P}. \end{aligned} \quad (2)$$

Assumption:  $\mathcal{P}$  is bounded.



$$\begin{aligned} & \min \|c - d\| \\ & s.t. \\ & d^T x^0 \leq d^T x \quad \forall x \in \mathcal{P}. \end{aligned} \tag{4}$$

- Model can be linearized for  $l_1$  and  $l_\infty$  norms.
- Convex hull of  $\mathcal{P}$  is a polytope.
- Last constraint set can be represented with the set of extreme points of convex hull of  $\mathcal{P}$ .
- Let  $\mathcal{E}$  be the set of extreme points of convex hull of  $\mathcal{P}$ ,  $\mathcal{E}$  is finite.



# Inverse IP with $l_1$ norm

$$z_{IP}^1 = \min \sum_{i=1}^n \theta_i$$

*s.t.*

$$c_i - d_i \leq \theta_i$$

$$\forall i \in \{1, 2, \dots, n\} \quad (3)$$

$$d_i - c_i \leq \theta_i$$

$$\forall i \in \{1, 2, \dots, n\}$$

$$d^T x^0 \leq d^T x$$

$$\forall x \in \mathcal{E}.$$





# Inverse IP with $l_\infty$ norm

$$z_{IP}^\infty = \min y$$

s.t.

$$c_i - d_i \leq y \quad \forall i \in \{1, 2, \dots, n\} \quad (4)$$

$$d_i - c_i \leq y \quad \forall i \in \{1, 2, \dots, n\}$$

$$d^T x^0 \leq d^T x \quad \forall x \in \mathcal{E}.$$

For the remainder of the presentation, we deal with the case of  $l_\infty$  norm. Let  $\mathcal{S}$  represent feasible set of the inverse IP, defined as

$$\mathcal{S} = \{(y, d) \in \mathbb{R} \times \mathbb{R}^n \mid y \geq \|c - d\|_\infty, d^T(x^0 - x) \leq 0 \forall x \in \mathcal{E}\}.$$

Note that  $\mathcal{S}$  is a polyhedron.



# Polynomially Solvable Forward Problems

Ahuja and Orlin [1] determines the complexity of inverse problem when the forward problem is polynomially solvable.

## Theorem

*(Ahuja and Orlin [1]) If a forward problem is polynomially solvable for each linear cost function, then corresponding inverse problems under  $l_1$  and  $l_\infty$  are polynomially solvable.*



# Forward Problems

Define the following problems related to IP and inverse IP.

## Definition

*IP decision problem:* Given  $\gamma \in \mathbb{Q}$  decide whether  $d^T x \leq \gamma$  holds for some  $x \in \mathcal{P}$ .

## Definition

*IP optimization problem:* Find solution vector  $x^*$  such that  $x^* \in \operatorname{argmin}_{x \in \mathcal{P}} d^T x$  or decide the problem is unbounded or decide the problem is infeasible.



## Definition

*Inverse IP decision problem:* Given  $\gamma \in \mathbb{Q}$  decide whether  $y \leq \gamma$  holds for some  $(y, d) \in \mathcal{S}$ .

## Definition

*Inverse IP optimization problem:* Find solution vector  $(y^*, d^*)$ , such that  $(y^*, d^*) \in \operatorname{argmin}_{(y,d) \in \mathcal{S}} y$ .

## Definition

*Inverse IP separation problem:* Given a vector  $(\bar{y}, \bar{d}) \in \mathbb{Q} \times \mathbb{Q}^n$ , decide whether  $(\bar{y}, \bar{d})$  is in  $\mathcal{S}$ , and if not, find a hyperplane that separates  $(\bar{y}, \bar{d})$  from  $\mathcal{S}$ , i.e., find  $\pi \in \mathbb{Q}^{n+1}$  such that  $\pi^T \begin{bmatrix} \bar{y} \\ \bar{d} \end{bmatrix} < \min \left\{ \pi^T \begin{bmatrix} y \\ d \end{bmatrix} \mid (y, d) \in \mathcal{S} \right\}$ .



# Component 1—An Observation

Recall inverse IP optimization problem.

$$z_{IP}^{\infty} = \min y$$

s.t.

$$c_i - d_i \leq y$$

$$\forall i \in \{1, 2, \dots, n\}$$

$$d_i - c_i \leq y$$

$$\forall i \in \{1, 2, \dots, n\}$$

$$d^T x^0 \leq d^T x$$

$$\forall x \in \mathcal{E}.$$

## Definition

*Inverse IP separation problem:* Given a vector  $(\bar{y}, \bar{d}) \in \mathbb{Q} \times \mathbb{Q}^n$ , decide whether  $(\bar{y}, \bar{d})$  is in  $\mathcal{S}$ , and if not, find a hyperplane that separates  $(\bar{y}, \bar{d})$  from  $\mathcal{S}$ , i.e., find  $\pi \in \mathbb{Q}^{n+1}$  such that  $\pi^T \begin{bmatrix} \bar{y} \\ \bar{d} \end{bmatrix} < \min \left\{ \pi^T \begin{bmatrix} y \\ d \end{bmatrix} \mid (y, d) \in \mathcal{S} \right\}$ .



## Component 2–GLS Theorem

The following theorem by Grötschel et al. indicates the relationship between separation and optimization problems.

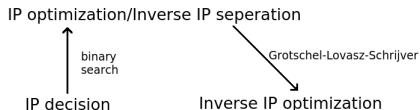
### Theorem

(Grötschel et al. [2]) *Given an oracle for the separation problem, the optimization problem over a given polyhedron with linear objective can be solved in time, polynomial in  $\varphi$ ,  $n$  and the **encoding length of objective coefficient vector**, where facet complexity of polyhedron is at most  $\varphi$ .*



# Complexity of IP optimization/decision problems

Figure: Problem relations



## Theorem

*Inverse IP optimization problem under  $l_\infty/l_1$  norm is solvable in time polynomial of  $\varphi$ ,  $n + 1/2n$ , and encoding length of  $(1, 0, \dots, 0)/(1, \dots, 1, 0, \dots, 0)$ , given an oracle for the IP decision problem.*

## Corollary

*Inverse IP decision problem is in  $\Delta_2^P$ .*

# Proposed Algorithm

First, we define two parametric problems named  $P_k$  and  $InvP_k$  as follows

$$\min_{x \in \mathcal{P}} d^{kT} x \quad (P_k)$$

$\min y$

*s.t.*

$$c_i - d_i \leq y \quad \forall i \in \{1, 2, \dots, n\} \quad (InvP_k)$$

$$d_i - c_i \leq y \quad \forall i \in \{1, 2, \dots, n\}$$

$$d^T x^0 \leq d^T x \quad \forall \mathcal{E}^{k-1}$$

where  $\mathcal{E}^{k-1}$  is the set of extreme points found so far.





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## Algorithm 1

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$k \leftarrow 1$

$d^k \leftarrow c$

Solve  $P_k, x^k \leftarrow x^*$

**while**  $d^{kT}(x^0 - x^k) > 0$  **do**

$k \leftarrow k + 1$

    Solve  $InvP_k, d^k \leftarrow d^*$

    Solve  $P_k, x^k \leftarrow x^*$

**end while**

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- $x^k$  is an optimal solution to  $P_k$ , whereas  $d^k$  is an optimal solution to  $invP_k$ .
- The algorithm stops when a generated cut is not violated by the current solution.



# Computational Results

- We used MIPLIB3 benchmark problems to create our inverse problems.
- We perturbed objective function coefficients, solved the problems and used the resulting solutions as  $x^0$ 's.
- Using generated  $x^0$ 's we created inverse IP problems.
- Experiments are conducted using COIN-OR Branch and Cut (Coin-Cbc) and COIN-OR Open Solver Interface (Coin-Osi) tools with Condor on a cluster running Debian operating system.



Table: Computational results— $l_\infty$

Problem name	Iter.	$\ c - d^*\ _\infty$	$\ c\ _\infty$	Time
air03	1	0	13746	05:05:24
bell3a	4	4.49554	60000	00:09:17
blend2	9	0.00508924	24.0142	00:05:00
cap6000	2	13.9	91110	01:40:01
dcmulti	867	7	1800	00:25:38
egout	7	1.38485	43.71	00:09:10
fiber	2	27.7829	729670	08:21:59
gesa3	1	0	1548890	00:00:37
gt2	1	0	7797	00:00:05
khb05250	6	0.75	2500000	00:01:05
l152lav	5	3.28571	268	00:02:02
lseu	1	0	517	00:00:17
mas74	1	0	1	02:28:05
misc03	1	0	1	00:00:02
misc06	1163	0.410302	1	01:01:38
mod008	2	4.625	87	00:00:11
mod010	4	3.78947	266	00:03:31
noswot	4	1	1	1-09:33:20
p0201	1	0	9600	00:00:00
p0282	1	0	160646	00:00:00
pk1	11	0.138533	1	01:16:51
pp08aCUTS	2380	4.69136	500	08:42:09
qiu	20	16.947	114.034	00:08:11
qnet1	6	0.41136	1	00:01:59
rgn	3	1	3	00:00:00
rout	10	0.555358	1	06:52:02
stein27	1	0	1	00:00:00
vpml	4	0.111111	1	1-02:55:03



# References



Ravindra K. Ahuja and James B. Orlin.

Inverse optimization.

*Operations Research*, 49(5):771–783, September/October 2001.



Martin Grötschel, László Lovász, and Alexander Schrijver.

*Geometric Algorithms and Combinatorial Optimization*, volume 2 of *Algorithms and Combinatorics*.

Springer, second corrected edition edition, 1993.



This is end of presentation!

Thank you for listening!



## A Small Example:

Define forward problem as follows and let  $x^0 = (2, 1)$ .

$$\min 2x_1 + x_2$$

*s.t.*

$$2x_1 + 3x_2 \geq 1$$

$$-x_1 + 2x_2 \leq 3$$

$$5x_1 + 5x_2 \leq 16$$

$$2x_1 - 4x_2 \leq 3$$

$$x_1, x_2 \geq 0$$

$$x_1, x_2 \text{ integer}$$



## A Small Example:

$k$ ,  $d^k$  and  $x^k$  values through iterations are given in Table 2.

Table:  $k$ ,  $d^k$  and  $x^k$  values through iterations

	$k$	$d^k$	$x^k$
initialization	1	(2, 1)	(0, 1)
iteration 1	2	(0, 3)	(1, 0)
iteration 2	3	(0, -1)	(1, 2)
iteration 3	4	(0, 0)	(0, 1)

Inverse IP optimal value is  $y^* = \|c - d^4\|_\infty = 2$ . Inverse IP optimal solution is  $d^4 = (0, 0)$ .



Table: Computational results— $l_1$

Problem name	Iter.	$\ c - d^*\ _1$	$\ c\ _1$	Time
bell3a	1	26.7727	1.78994e+06	
cap6000	2	19	1.29696e+07	
dcmulti	1652	193.312	30648.9	
egout	34	28.472	793.391	
enigma	1	0	45	
fiber	1	3.28735	4.80778e+07	
flugpl	1	0	25380	
gesa2	1	0	9.09347e+07	
gesa3	1	0	8.73813e+07	
gt2	1	0	291998	
khb05250	3	8	6.5264e+07	
l152lav	21	85	382524	
lseu	1	0	15494	
mas74	1	0	1.0015	
misc03	1	0	1	
misc06	10	1	1	
mod008	6	23	23554	
mod010	60	101	489211	
p0201	1	50	99900	
p0282	1	0	1.30262e+06	
pk1	1	1	1	
qnet1	25	2.07111	124	
rgn	53	20	160	
rout	7	1	1	
stein27	1	0	27	

