

# Using BCD-CTA for difficult tables: a practical experiment with a real Eurostat table

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- 2 Outline of MILP-CTA
- 3 BCD-CTA
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# Protection of tables published at the EU-level

- ▷ The goal:
  - ① At national level: NSIs collect and process the information of each member state (primary and secondary suppressions).
  - ② The NSIs send their tables to Eurostat.
  - ③ Eurostat compiles the complete European table, with the obligation to guarantee the confidentiality of the national data.
- ▷ The solution approach of S. Giessing, A.H., J.C. (2009) was to compute rounded figures for the EU-cells at risk.
- ▷ CTA was used in one of the steps of this solution approach.

# The instance that motivated this work

- ▷ A.H. working on real Eurostat dataset (EU structural business statistics).
- ▷ Three-dimensional table.
  - First: EU-member state (27 plus the EU-total).
  - Second: NACE hierarchical classification (120 codes).
  - Third: size-class (5 codes plus a total)
- ▷ Number of cells:  $28 \cdot 120 \cdot 6 = 20160$ .
- ▷ Number of constraints: 8280.
- ▷ A.H. found unacceptable long running CTA times for this instance.
- ▷ A.H. came to J.C. and J.A.G.: what to do?
- ▷ Solution: either use BCD-CTA or try different weights.

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## MILP-CTA: Example

ORIGINAL TABLE. Protection levels:  $x_{23} \geq 45$  or  $x_{23} \leq 35$

	$Z_1$	$Z_2$	$Z_3$	TOTAL
$E_1$	20	24	28	72
$E_2$	38	38	40	116
$E_3$	40	39	42	121
TOTAL	98	101	110	309

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PROTECTED TABLE: either ... or ...

	$Z_1$	$Z_2$	$Z_3$		$Z_1$	$Z_2$	$Z_3$	
$E_1$	25	24	23	72	15	24	33	72
$E_2$	33	38	45	116	43	38	35	116
$E_3$	40	39	42	121	40	39	42	121
TOTAL	98	101	110	309	98	101	110	309



# Parameters of MILP-CTA

- Set of cells  $a_i, i = 1, \dots, n$ .
- Set  $\mathcal{S} = \{i_1, i_2, \dots, i_s\} \subseteq \{1, \dots, n\}$  of indices of sensitive cells.
- Linear relations  $Aa = b$ .
- Lower and upper protection level for each sensitive cell  $i \in \mathcal{S}$ :  
 $lpl_i$  and  $upl_i$ .
- Lower and upper bound for each cell:  $l_{x_i}$  and  $u_{x_i}$ .
- Cell weights  $w_i$  for cost of adjustment of each cell.

# Aim of CTA

Find released values  $x$  such that:

- Remain **near  $a$  for some distance  $\ell$** .
- Satisfy the linear relations  **$Ax = b$**
- Satisfy the bounds:  **$l_x \leq x \leq u_x$**
- Satisfy the protection levels: either  **$x_i \geq a_i + upl_i$**  or  **$x_i \leq a_i - lpl_i$** .

The optimization problem is:

$$\begin{array}{ll}
 \min_x & \|x - a\|_\ell \\
 \text{subject to} & Ax = b \\
 & l_x \leq x \leq u_x \\
 & x_i \leq a_i - lpl_i \quad \text{or} \quad x_i \geq a_i + upl_i \quad i \in \mathcal{S}.
 \end{array}$$

# The MILP-CTA problem for $\ell_1$

- Defining *cell deviations* as:  $z = x - a$ ,
- and introducing binary variables for sensitive cells:  $y_i, i \in \mathcal{S}$   
 $y_i = 1 \Leftrightarrow$  *upside* protected:  $x_i \geq a_i + \text{up}l_i$ ;  
 $y_i = 0 \Leftrightarrow$  *downside* protected:  $x_i \leq a_i - \text{lp}l_i$ .

MILP-CTA model:

$$\begin{array}{ll}
 \min_{z^+, z^-, y} & \sum_{i=1}^n w_i (z_i^+ + z_i^-) \\
 \text{subject to} & A(z^+ - z^-) = 0 \\
 & 0 \leq z^+ \leq u_z, \quad 0 \leq z^- \leq -l_z \\
 & y \in \{0, 1\}^{\mathcal{S}} \\
 & \left. \begin{array}{l} \text{up}l_i y_i \leq z_i^+ \leq u_{z_i} y_i \\ \text{lp}l_i (1 - y_i) \leq z_i^- \leq -l_{z_i} (1 - y_i) \end{array} \right\} i \in \mathcal{S}
 \end{array}$$

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# CD and BCD algorithms

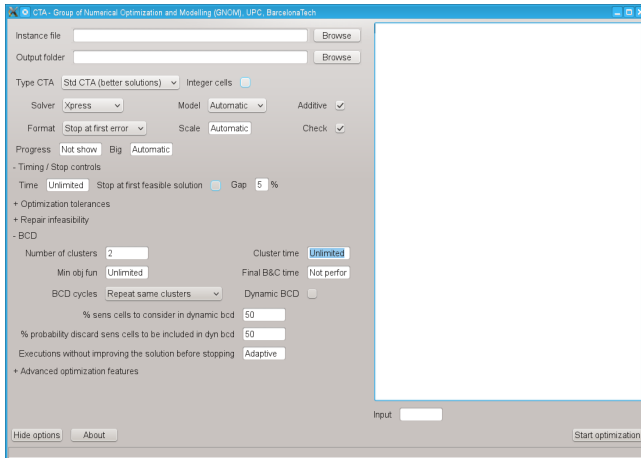
- ▷ **Coordinate descent**: family of optimization algorithms that successively optimize along coordinate directions.
- ▷ Very popular in 80s and 90s.
- ▷ Recently gained reputation for approximate solutions in big-data.
- ▷ **Block coordinate descent (BCD)**: optimize on a block of variables.
- ▷ BCD solves a sequence of subproblems on a subset of variables, the other variables kept fixed.

# BCD-CTA algorithm

- Step 0** Initialization. Set outer iteration counter:  $t = 0$ . Set initial values, hopefully feasible, to  $y$ .
- Step 1**  $t = t + 1$ . Set inner iteration counter  $i = 0$ .  
Divide  $y$  into  $k$  blocks:  $y = \{y^{1,i}, \dots, y^{k,i}\}$ , not necessarily of the same size.
- Step 1.1**  $i := i + 1$ . Solve **CTA** with respect to block  $y^{i,i}$ , taking into account that  $y^{j,i}$  is fixed for  $j \neq i$ .  
Let  $y^{i,i+1} = (y^{i,i})^*$  (the point at the optimum).  
Let  $y^{j,i+1} = y^{j,i}$  for  $j \neq i$ .
- Step 1.2** If  $i < k$  go to Step 1.1.
- Step 2** Check for end conditions: if apply, stop, and return the current best solution. Otherwise, go to Step 1

# BCD-CTA in $\tau$ -Argus

- **BCD-CTA** available in  $\tau$ -Argus.
- Implemented within DwB EU project.



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# Preliminary results

- ▷ Three-dimensional Eurostat SBS table.
- ▷ Used both MILP-CTA and BCD-CTA as implemented in  $\tau$ -Argus.
- ▷ CPLEX used as LP/MILP solver.
- ▷ Initially used  $w_i = 1$  and MILP-CTA: stopped after 7692 seconds.
- ▷ Then we tried BCD-CTA: decent solution in 19 seconds.
- ▷ We tried other weights: unexpectedly much faster.

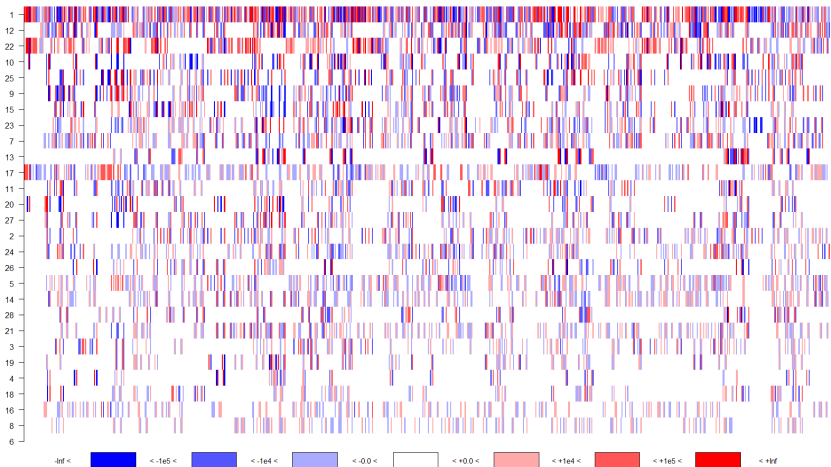
# Summary of results

- Results for MILP-CTA and BCD-CTA with three different weights.
- $\bar{x}$  CTA: average deviation of CTA solution.
- $\bar{x}$  published: average deviation of published tables with full approach.
- BCD-CTA did a good job for CPU time and “ $\bar{x}$  CTA”.
- Best combination for “ $\bar{x}$  published” is likely  $w_i = 1/\sqrt{a_i}$  with MILP-CTA.

$w_i$	MILP-CTA			BCD-CTA		
	CPU	$\bar{x}$ CTA	$\bar{x}$ published	CPU	$\bar{x}$ CTA	$\bar{x}$ published
1	7692	38933	64490	19	46443	91977
$1/a_i$	85	46904	89451	16	45900	93617
$1/\sqrt{a_i}$	179	40717	65297	16	42785	85098

# Graphical representation of solution

Solution with one minute of BCD-CTA followed by two hours of MILP-CTA with weights  $w_j = 1$ .



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# Conclusions and future work

- ▷ Cell weights have a large implication in solutions and CPU time.
- ▷ Cell weights based on cell hierarchy (J.C., S. Giessing 2006 paper).
- ▷ BCD-CTA has enormous gain in CPU time, with a price in solution.
- ▷ BCD-CTA available in  $\tau$ -Argus: it allows solution of very large and intractable tables by other approaches.
- ▷ From an optimization point of view: why the behaviour of the MILP solver changes so drastically with different weights?

Thanks for your attention!