

Tight and Compact MILP Formulations for Unit Commitment Problems

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- 1 Introduction
- 2 Good and Ideal MIP formulations
 - The Ideal MIP Formulation
 - Good MIP Formulations
- 3 Tight & Compact (TC) UC Formulations
 - Traditional Formulation
 - Power-Based UC
 - Startup & Shutdown Ramps
- 4 Numerical Results & Further UC Extensions
 - Basic UC Formulations
 - Stochastic UC
 - Ramp-Based Scheduling
- 5 Conclusions

- Significant breakthroughs in Mixed-Integer Programming (MIP)
 - Solving MIP 100 million times faster than 20 years ago¹

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 - Computer power (e.g., clusters)
 - Solving algorithms (e.g., solvers, decomposition techniques)

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 - Computer power (e.g., clusters)
 - Solving algorithms (e.g., solvers, decomposition techniques)
 - Improving the MIP-Based UC formulation \Rightarrow \downarrow solving times

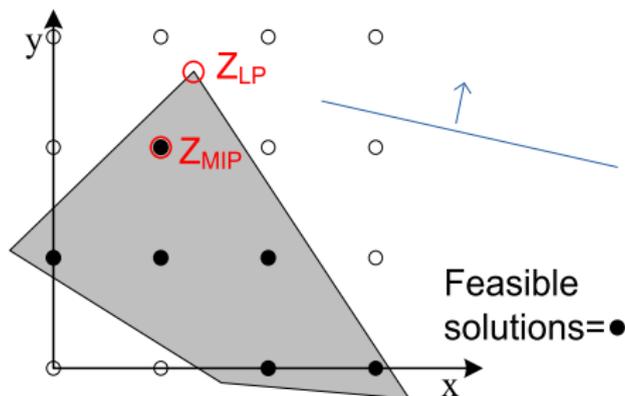
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Solving MIP Through The Powerful LP

Shaping the linear feasible region to arrive from vertex Z_{LP} to Z_{MIP}

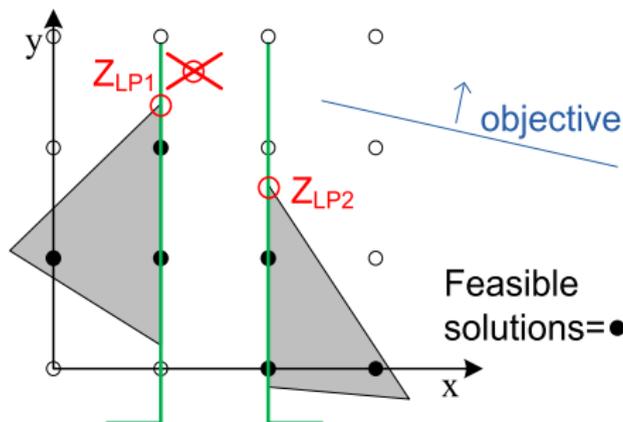


To prove optimality Z_{MIP} must become a vertex by:

- Branch and bound (divide and conquer)

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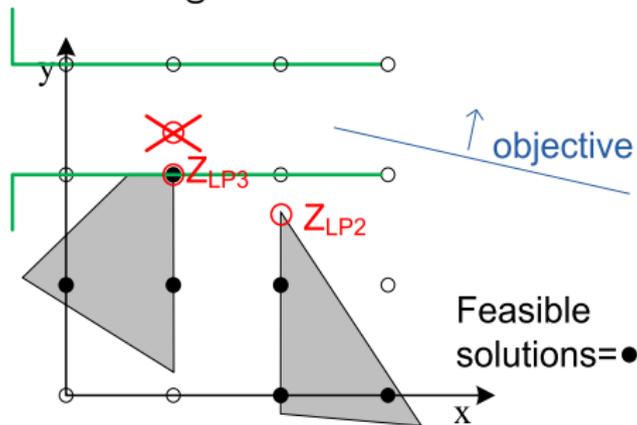


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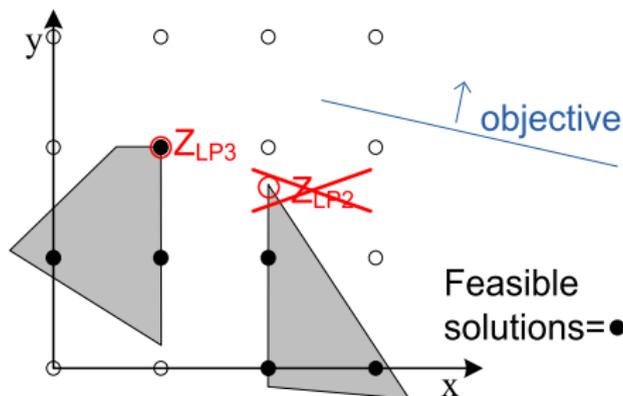


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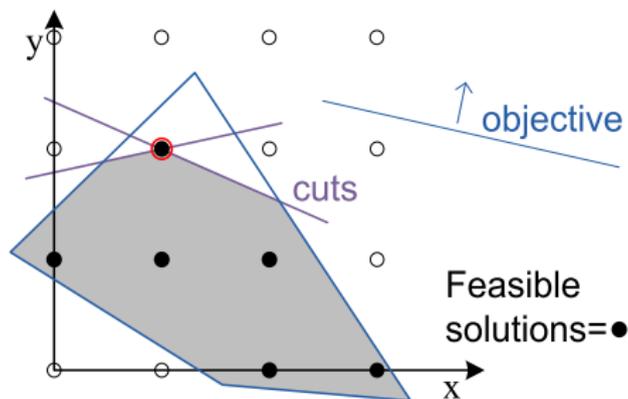


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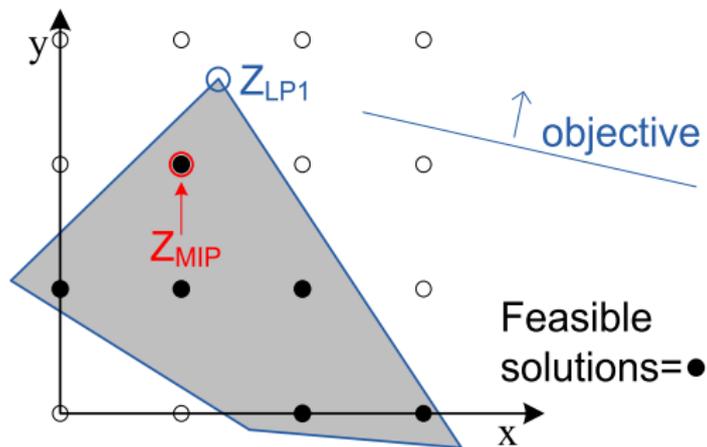
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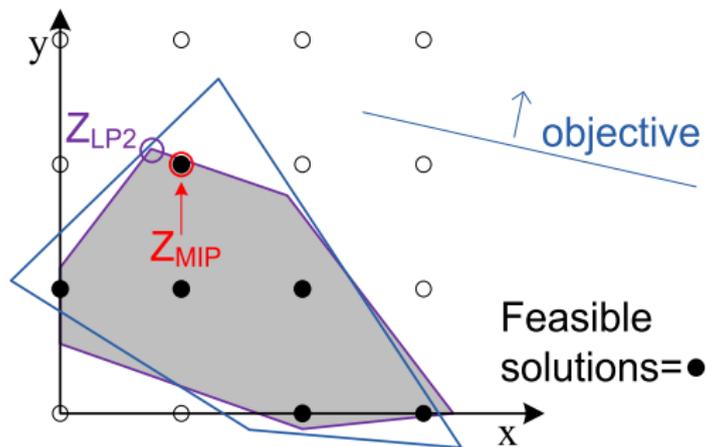
To prove optimality Z_{MIP} must become a vertex by:

- Branch and bound (divide and conquer)
- and/or by adding cuts

An MIP Has Infinite LP Formulations

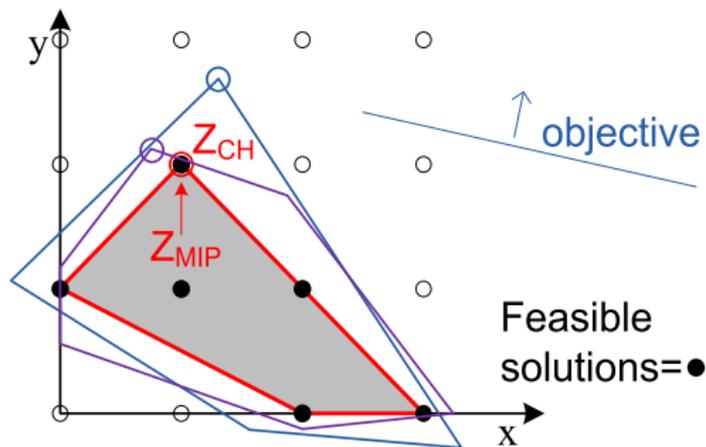


An MIP Has Infinite LP Formulations



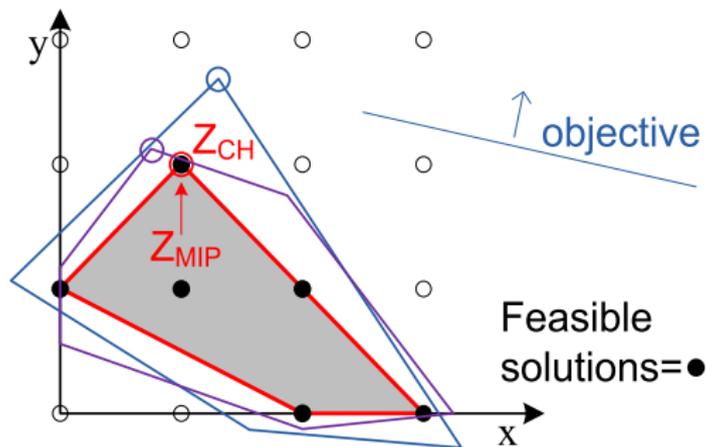
LP1 and LP2 represent the same MIP problem

An MIP Has Infinite LP Formulations



LP1, LP2 and CH represent the same MIP problem

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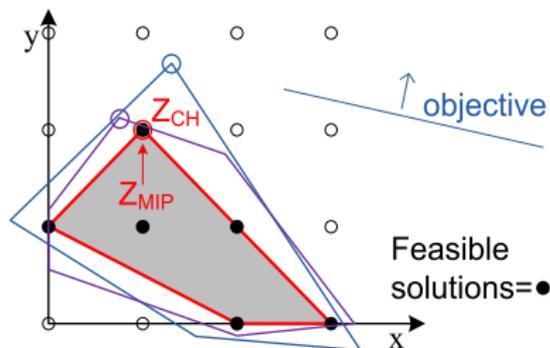
LP1, LP2 and CH represent the same MIP problem

which one to choose?

Convex Hull: The Tightest Formulation

Convex Hull (CH)

Smallest convex feasible region containing all the feasible integer points²

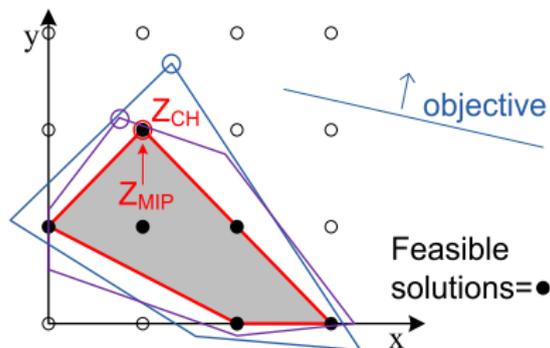


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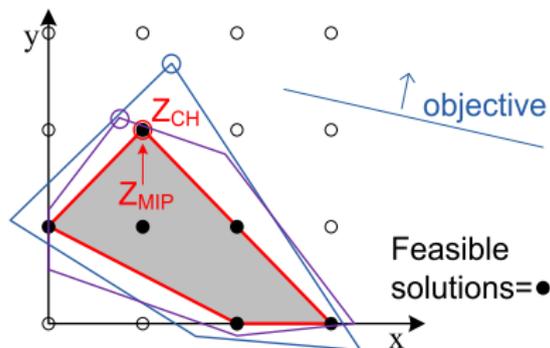
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 - Each vertex satisfies the integrality constraints
- Unfortunately...

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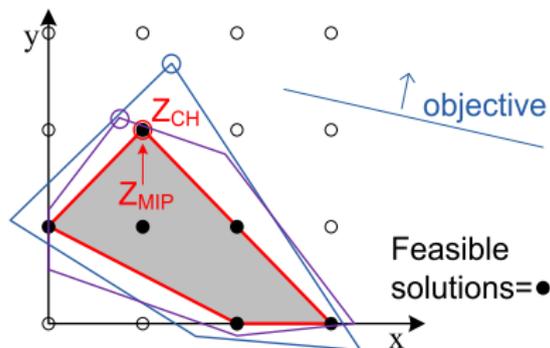
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- The *convex hull* problem solves an MIP as an LP
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- **Unfortunately**, the *convex hull* is typically too difficult to obtain^{2,3}
 - An enormous (exponential) number of inequalities is needed

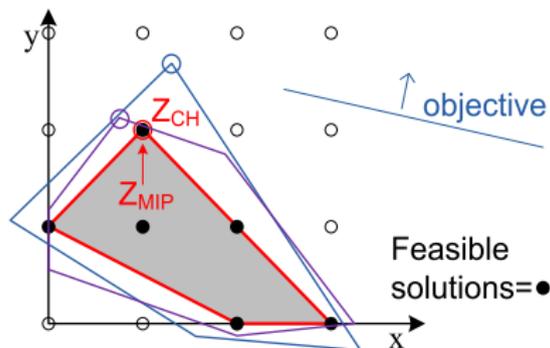
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Smallest convex feasible region containing all the feasible integer points²



- The *convex hull* problem solves an MIP as an LP
 - Each vertex satisfies the integrality constraints
 - So an LP optimum is also an MIP optimum
- **Unfortunately**, the *convex hull* is typically too difficult to obtain^{2,3}
 - An enormous (exponential) number of inequalities is needed
 - To solve an MIP is usually easier than trying to find its *convex hull*

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Choosing The Best Formulation

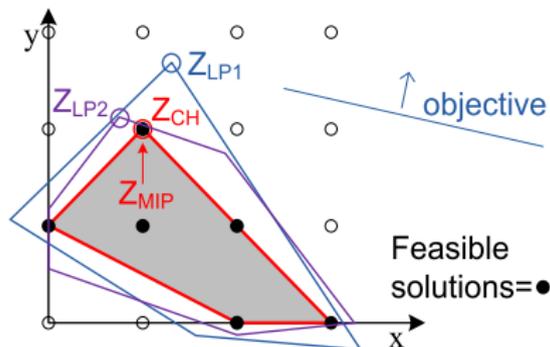
Measuring The Tightness



Integrality Gap (IGap)

Relative distance between MIP
and LP optima

$$\text{IGap}_{\text{LP1}} = \frac{Z_{\text{MIP}} - Z_{\text{LP1}}}{Z_{\text{MIP}}}$$

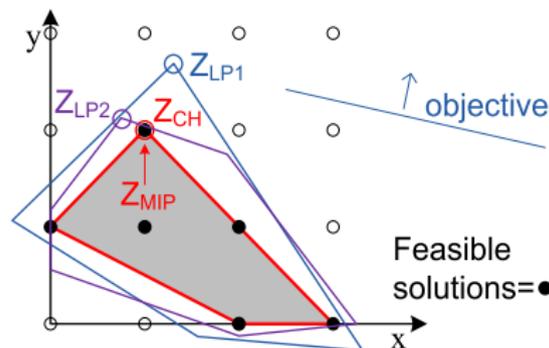


Choosing The Best Formulation

Measuring The Tightness

Integrality Gap (IGap)

Relative distance between MIP
and LP optima



$$\text{IGap}_{\text{LP1}} = \frac{Z_{\text{MIP}} - Z_{\text{LP1}}}{Z_{\text{MIP}}} > \text{IGap}_{\text{LP2}} = \frac{Z_{\text{MIP}} - Z_{\text{LP2}}}{Z_{\text{MIP}}}$$

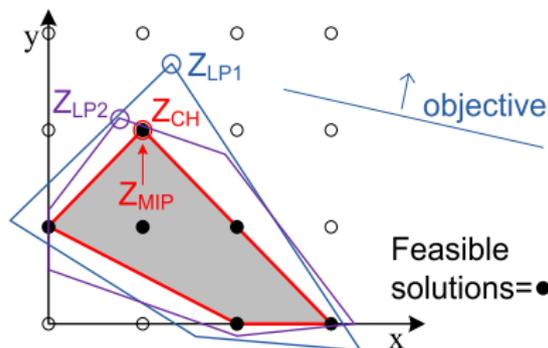
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Integrality Gap (IGap)

Relative distance between MIP
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$$\text{IGap}_{\text{LP1}} = \frac{Z_{\text{MIP}} - Z_{\text{LP1}}}{Z_{\text{MIP}}} > \text{IGap}_{\text{LP2}} = \frac{Z_{\text{MIP}} - Z_{\text{LP2}}}{Z_{\text{MIP}}} > \text{IGap}_{\text{CH}} = \frac{Z_{\text{MIP}} - Z_{\text{CH}}}{Z_{\text{MIP}}} = 0$$



As an MIP problem:

LP2 is expected to be solved faster than LP1

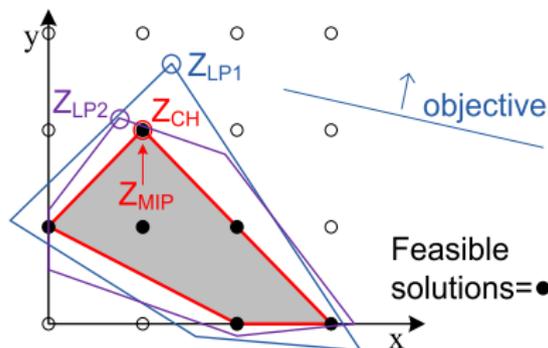
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Integrality Gap (IGap)

Relative distance between MIP
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$$\text{IGap}_{LP1} = \frac{Z_{MIP} - Z_{LP1}}{Z_{MIP}} > \text{IGap}_{LP2} = \frac{Z_{MIP} - Z_{LP2}}{Z_{MIP}} > \text{IGap}_{CH} = \frac{Z_{MIP} - Z_{CH}}{Z_{MIP}} = 0$$



As an MIP problem:

LP2 is expected to be solved faster than LP1

CH will be solved way faster than LP2

Concepts: Tightness and Compactness

- Tightness: defines the search space (relaxed feasible region) that the solver needs to explore to find the solution
- Compactness (problem size): defines the searching speed (data to process) that the solver takes to find the solution

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- Tightness: defines the search space (relaxed feasible region) that the solver needs to explore to find the solution
- Compactness (problem size): defines the searching speed (data to process) that the solver takes to find the solution
- *Convex hull*: The tightest formulation \Rightarrow MIP solved as LP

Tightening an MIP Formulation

- The most common strategy is adding cuts
 - In fact, this is the most effective strategy of current MIP solvers⁴
 - This may add a huge number of inequalities $\Rightarrow \uparrow$ Time

⁴R. Bixby and E. Rothberg, "Progress in computational mixed integer programming—A look back from the other side of the tipping point," *Annals of Operations Research*, vol. 149, no. 1, pp. 37–41, Jan. 2007

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 - This may add a huge number of inequalities $\Rightarrow \uparrow$ Time
 - Trade-off: Tightness vs. Compactness
- Improving the MIP formulation
 - Provide the *convex hull* for some set of constraints
 - If available, use the *convex hull* for some set of constraints

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Tight and Compact (TC) Formulation

- The whole formulation can be found in the paper [TC-UC](#)⁵

⁵G. Morales-España, J. M. Latorre, and A. Ramos, "Tight and compact MILP formulation for the thermal unit commitment problem," *IEEE Transactions on Power Systems*, 2013, Special Section on Analysis and Simulation of Very Large Power Systems. In Press

Tight and Compact (TC) Formulation

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- Let's focus on the core of UC formulations:
 - Min/max outputs
 - SU & SD capabilities
 - Minimum up/down (TU/TD) times

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- Let's focus on the core of UC formulations:
 - Min/max outputs
 - SU & SD capabilities
 - Minimum up/down (TU/TD) times, *convex hull* already available⁶

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⁶D. Rajan and S. Takriti, "Minimum Up/Down polytopes of the unit commitment problem with start-up costs," IBM, Research Report RC23628, Jun. 2005

Formulation (II)

- Generation limits taking into account: maximum \bar{P} and minimum \underline{P} output, as well as maximum SU and SD capabilities:

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} - \max(SD - SU, 0) v_t \quad \forall t \quad (1)$$

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t - \max(SU - SD, 0) w_{t+1} \quad \forall t \quad (2)$$

Total generation = $\underline{P} \cdot u_t + p_t$.

Formulation (III)

- Logical relationship: commitment u_t , startup v_t and shutdown w_t :

$$u_t - u_{t-1} = v_t - w_t \quad \forall t \quad (3)$$

$$v_t \leq u_t \quad \forall t \quad (4)$$

$$w_t \leq 1 - u_t \quad \forall t \quad (5)$$

where (4) and (5) avoid the simultaneous startup and shutdown.

- Variable bounds

$$p_t \geq 0 \quad \forall t \quad (6)$$

$$0 \leq u_t, v_t, w_t \leq 1 \quad \forall t \quad (7)$$

Tightness of the Formulation

Let's study the polytope (1)-(7) using PORTA⁷:

- PORTA enumerates all vertices of a convex feasible region

⁷T. Christof and A. Löbel, "PORTA: POLYhedron representation transformation algorithm, version 1.4.1,"
Konrad-Zuse-Zentrum für Informationstechnik Berlin, Germany, 2009

Tightness of the Formulation

Let's study the polytope (1)-(7) using PORTA⁷:

- PORTA enumerates all vertices of a convex feasible region
- Example: 3 periods and $\bar{P} = 200, \underline{P} = SU = SD = 100$ for:
 - Case 1: $TU = TD = 1$
 - Case 2: $TU = TD = 2$
- For the complete and detailed formulation and a comprehensive study, see paper [Tight LP-UC](#)⁸

⁷T. Christof and A. Löbel, "PORTA: POLYhedron representation transformation algorithm, version 1.4.1," *Konrad-Zuse-Zentrum für Informationstechnik Berlin, Germany*, 2009

⁸G. Morales España, C. Gentile, and A. Ramos, "Tight LP formulation of the unit commitment problem presenting integer solutions," *IEEE Transactions on Power Systems*, 2013, Under Review

Case 1: Providing The Convex Hull

Formulation:

PORTA results for ($TU = TD = 1$)

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} - \max(SD - SU, 0) v_t \quad (1)$$

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t - \max(SU - SD, 0) w_{t+1} \quad (2)$$

$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

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$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

$$v_t \leq u_t \quad (4)$$

$$w_t \leq 1 - u_t \quad (5)$$

PORTA results for ($TU = TD = 1$)

$u_1, u_2, u_3, v_2, v_3, w_2, w_3, p_1, p_2, p_3$:

DIM = 10

CONV_SECTION

(1)	0	0	0	0	0	0	0	0	0	0	0	0
(2)	0	0	1	0	1	0	0	0	0	0	0	0
(3)	1	0	0	0	0	1	0	0	0	0	0	0
(4)	0	1	0	1	0	0	1	0	0	0	0	0
(5)	0	1	1	1	0	0	0	0	0	0	0	0
(6)	0	1	1	1	0	0	0	0	0	0	100	0
(7)	1	1	0	0	0	0	1	0	0	0	0	0
(8)	1	1	0	0	0	0	1	100	0	0	0	0
(9)	1	1	1	0	0	0	0	0	0	0	0	0
(10)	1	1	1	0	0	0	0	0	0	0	100	0
(11)	1	1	1	0	0	0	0	0	100	0	0	0
(12)	1	1	1	0	0	0	0	0	0	100	100	0
(13)	1	1	1	0	0	0	0	100	0	0	0	0
(14)	1	1	1	0	0	0	0	100	0	100	0	0
(15)	1	1	1	0	0	0	0	100	100	0	0	0
(16)	1	1	1	0	0	0	0	100	100	100	0	0
(17)	1	0	1	0	1	1	0	0	0	0	0	0

END

Case 1: Providing The Convex Hull

Formulation:

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} - \max(SD - SU, 0) v_t \quad (1)$$

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t - \max(SU - SD, 0) w_{t+1} \quad (2)$$

$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

$$v_t \leq u_t \quad (4)$$

$$w_t \leq 1 - u_t \quad (5)$$

All vertices are integer



Convex Hull

PORTA results for ($TU = TD = 1$)

$u_1, u_2, u_3, v_2, v_3, w_2, w_3, p_1, p_2, p_3$:

DIM = 10

CONV_SECTION

(1)	0	0	0	0	0	0	0	0	0	0	0	0
(2)	0	0	1	0	1	0	0	0	0	0	0	0
(3)	1	0	0	0	0	1	0	0	0	0	0	0
(4)	0	1	0	1	0	0	1	0	0	0	0	0
(5)	0	1	1	1	0	0	0	0	0	0	0	0
(6)	0	1	1	1	0	0	0	0	0	0	100	0
(7)	1	1	0	0	0	0	1	0	0	0	0	0
(8)	1	1	0	0	0	0	1	100	0	0	0	0
(9)	1	1	1	0	0	0	0	0	0	0	0	0
(10)	1	1	1	0	0	0	0	0	0	0	100	0
(11)	1	1	1	0	0	0	0	0	100	0	0	0
(12)	1	1	1	0	0	0	0	0	100	100	0	0
(13)	1	1	1	0	0	0	0	100	0	0	0	0
(14)	1	1	1	0	0	0	0	100	0	100	0	0
(15)	1	1	1	0	0	0	0	100	100	0	0	0
(16)	1	1	1	0	0	0	0	100	100	100	0	0
(17)	1	0	1	0	1	1	0	0	0	0	0	0

END

Case 2: Providing and Using Convex Hulls (I)

Formulation + *TU/TD Convex hull*:

PORTA results for ($TU=TD=2$)

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} - \max(SD - SU, 0) v_t \quad (1)$$

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t - \max(SU - SD, 0) w_{t+1} \quad (2)$$

$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

$$\sum_{i=t-TU+1}^t v_i \leq u_t \quad (4)$$

$$\sum_{i=t-TD+1}^t w_i \leq 1 - u_t \quad (5)$$

Case 2: Providing and Using Convex Hulls (I)

Formulation + *TU/TD Convex hull*:

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} - \max(SD - SU, 0) v_t \quad (1)$$

$$p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t - \max(SU - SD, 0) w_{t+1} \quad (2)$$

$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

$$\sum_{i=t-TU+1}^t v_i \leq u_t \quad (4)$$

$$\sum_{i=t-TD+1}^t w_i \leq 1 - u_t \quad (5)$$

How to remove the fractional vertices?

PORTA results for ($TU=TD=2$)

$u_1, u_2, u_3, v_2, v_3, w_2, w_3, p_1, p_2, p_3$:

DIM = 10

CONV_SECTION

(1)	0	0	0	0	0	0	0	0	0	
(2)	1/2	1	1/2	1/2	0	0	1/2	0	50	0
(3)	1/2	1	1/2	1/2	0	0	1/2	0	50	50
(4)	1/2	1	1/2	1/2	0	0	1/2	50	50	0
(5)	1/2	1	1/2	1/2	0	0	1/2	50	50	50
(6)	0	0	1	0	1	0	0	0	0	0
(7)	1	0	0	0	0	1	0	0	0	0
(8)	0	1	1	1	0	0	0	0	0	0
(9)	0	1	1	1	0	0	0	0	0	100
(10)	1	1	0	0	0	0	1	0	0	0
(11)	1	1	0	0	0	0	1	100	0	0
(12)	1	1	1	0	0	0	0	0	0	0
(13)	1	1	1	0	0	0	0	0	0	100
(14)	1	1	1	0	0	0	0	0	100	0
(15)	1	1	1	0	0	0	0	0	100	100
(16)	1	1	1	0	0	0	0	100	0	0
(17)	1	1	1	0	0	0	0	100	0	100
(18)	1	1	1	0	0	0	0	100	100	0
(19)	1	1	1	0	0	0	0	100	100	100

END

Case 2: Providing and Using Convex Hulls (II)

Reformulating (1) and (2) for $TU \geq 2$:

$$\begin{aligned}
 p_t &\leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1} \\
 &\quad - \max(SD - SU, 0) v_t \quad (1)
 \end{aligned}$$

$$\begin{aligned}
 p_t &\leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t \\
 &\quad - \max(SU - SD, 0) w_{t+1} \quad (2)
 \end{aligned}$$

$$\begin{aligned}
 p_t &\leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SU) v_t \\
 &\quad - (\bar{P} - SD) w_{t+1} \quad (8)
 \end{aligned}$$

$$u_t - u_{t-1} = v_t - w_t \quad (3)$$

$$\sum_{i=t-TU+1}^t v_i \leq u_t \quad (4)$$

$$\sum_{i=t-TD+1}^t w_i \leq 1 - u_t \quad (5)$$

PORTA results for $(TU = TD = 2)$

$u_1, u_2, u_3, v_2, v_3, w_2, w_3, p_1, p_2, p_3$:

DIM = 10

CONV_SECTION

(1)	0	0	0	0	0	0	0	0	0	0
(2)	0	0	1	0	1	0	0	0	0	0
(3)	1	0	0	0	0	1	0	0	0	0
(4)	0	1	1	1	0	0	0	0	0	0
(5)	0	1	1	1	0	0	0	0	0	100
(6)	1	1	0	0	0	0	1	0	0	0
(7)	1	1	0	0	0	0	1	100	0	0
(8)	1	1	1	0	0	0	0	0	0	0
(9)	1	1	1	0	0	0	0	0	0	100
(10)	1	1	1	0	0	0	0	0	100	0
(11)	1	1	1	0	0	0	0	0	100	100
(12)	1	1	1	0	0	0	0	100	0	0
(13)	1	1	1	0	0	0	0	100	0	100
(14)	1	1	1	0	0	0	0	100	100	0
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(1)	0	0	0	0	0	0	0	0	0	0
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(4)	0	1	1	1	0	0	0	0	0	0
(5)	0	1	1	1	0	0	0	0	0	100
(6)	1	1	0	0	0	0	1	0	0	0
(7)	1	1	0	0	0	0	1	100	0	0
(8)	1	1	1	0	0	0	0	0	0	0
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(10)	1	1	1	0	0	0	0	0	100	0
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(12)	1	1	1	0	0	0	0	100	0	0
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(15)	1	1	1	0	0	0	0	100	100	100

END

⇒ Convex Hull



Outline

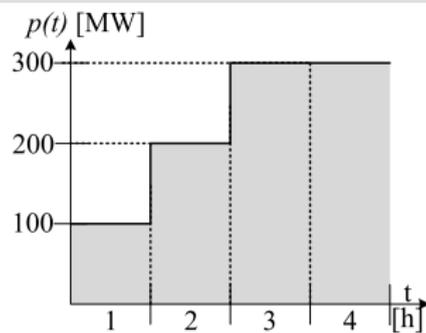
- 1 Introduction
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Energy vs Power

Generation levels are usually considered as energy blocks.

Example: $\bar{P} = 300\text{MW}$; $\underline{P} = 100\text{MW}$; Up/Down ramp rate: 100 MW/h

Traditional UC

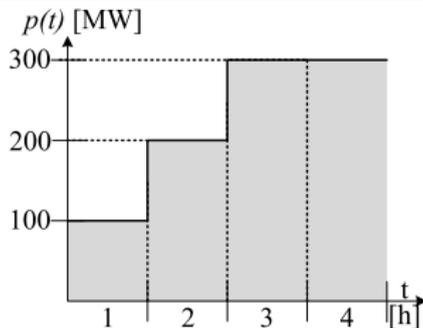


Energy vs Power

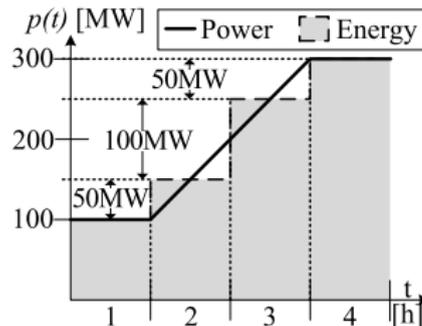
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Traditional UC



Feasible energy profile

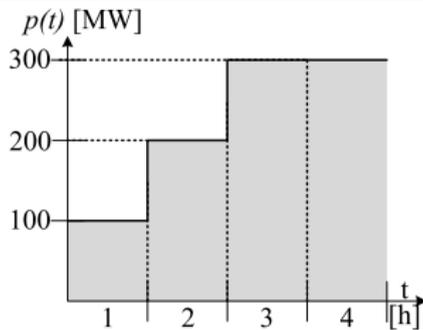


Energy vs Power

Generation levels are usually considered as energy blocks.

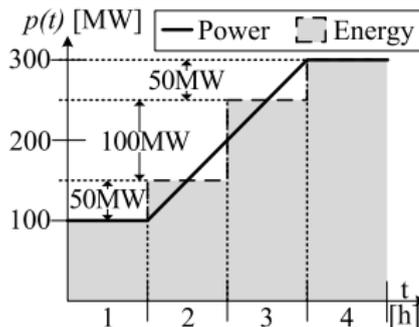
Example: $\bar{P} = 300\text{MW}$; $\underline{P} = 100\text{MW}$; Up/Down ramp rate: 100 MW/h

Traditional UC



Infeasible energy delivery⁹

Feasible energy profile



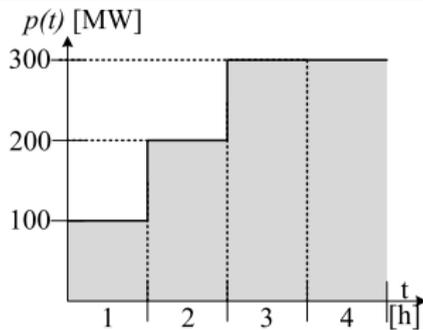
⁹X. Guan, F. Gao, and A. Svoboda, "Energy delivery capacity and generation scheduling in the deregulated electric power market," *IEEE Transactions on Power Systems*, vol. 15, no. 4, pp. 1275–1280, Nov. 2000

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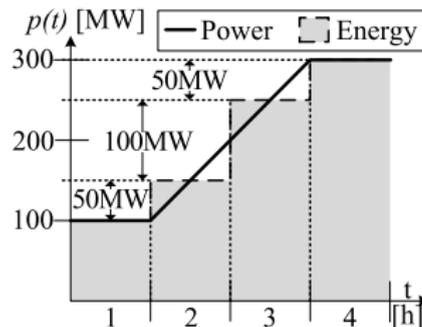
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Traditional UC



Infeasible energy delivery⁹
Overestimated ramp availability

Feasible energy profile



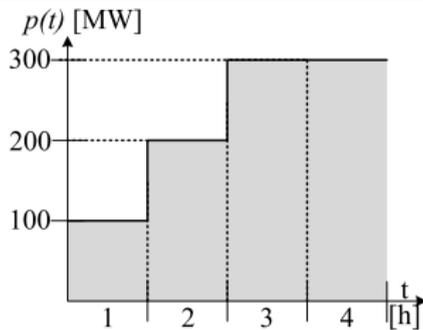
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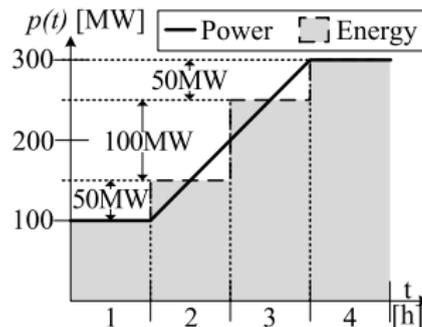
Example: $\bar{P} = 300\text{MW}$; $\underline{P} = 100\text{MW}$; Up/Down ramp rate: 100 MW/h

Traditional UC



Infeasible energy delivery⁹
Overestimated ramp availability

Feasible energy profile



A clear difference between power and energy is required in an UC

⁹X. Guan, F. Gao, and A. Svoboda, "Energy delivery capacity and generation scheduling in the deregulated electric power market," *IEEE Transactions on Power Systems*, vol. 15, no. 4, pp. 1275–1280, Nov. 2000

TC P-Based UC: Providing The Convex Hull

Power-Based UC formulation for
 $SU = SD = \underline{P}$:

PORTA results for $(TU = TD = 2)$

$$p'_t \leq (\overline{P} - \underline{P}) (u_t - w_{t+1}) \quad (9)$$

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$$u_1, u_2, u_3, v_2, v_3, w_2, w_3, p'_1, p'_2, p'_3:$$

DIM = 10

CONV_SECTION

(1)	0	0	0	0	0	0	0	0	0	0	0	0
(2)	0	0	1	0	1	0	0	0	0	0	0	0
(3)	0	0	1	0	1	0	0	0	0	0	100	0
(4)	1	0	0	0	0	1	0	0	0	0	0	0
(5)	0	1	1	1	0	0	0	0	0	0	0	0
(6)	0	1	1	1	0	0	0	0	0	0	100	0
(7)	0	1	1	1	0	0	0	0	0	100	0	0
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(16)	1	1	1	0	0	0	0	100	0	100	0	0
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END

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p'_t = power over \underline{P} at the end of period t

All vertices are integer



Convex Hull

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DIM = 10

CONV_SECTION

```
( 1) 0 0 0 0 0 0 0 0 0 0 0 0 0
( 2) 0 0 1 0 1 0 0 0 0 0 0
( 3) 0 0 1 0 1 0 0 0 0 0 100
( 4) 1 0 0 0 0 0 1 0 0 0 0
( 5) 0 1 1 1 0 0 0 0 0 0 0
( 6) 0 1 1 1 0 0 0 0 0 0 100
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( 8) 0 1 1 1 0 0 0 0 0 100 100
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END

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Ignoring SU and SD Power Trajectories

Generation output below minimum output is usually ignored in UCs

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Ignoring SU and SD Power Trajectories

Generation output below minimum output is usually ignored in UCs

- SU & SD ramps are deterministic events in day-ahead UCs
- Ignoring them change commitment decisions and increase costs^{10,11}
- This energy must be optimally allocated by day-ahead UCs

¹⁰G. Morales-España, J. M. Latorre, and A. Ramos, "Tight and compact MILP formulation of start-up and shut-down ramping in unit commitment," *IEEE Transactions on Power Systems*, vol. 28, no. 2, pp. 1288–1296, 2013

¹¹G. Morales-España, A. Ramos, and J. Garcia-Gonzalez, "An MIP formulation for joint market-clearing of energy and reserves based on ramp scheduling," *IEEE Transactions on Power Systems*, 2013, Special Section on Electricity Markets Operation. In Press

Modeling SU and SD Power Trajectories

Total unit's production including SU & SD power trajectories¹²:

$$\hat{p}_t = \underbrace{\sum_{i=1}^{SU^D} P_i^{SU} v_{(t-i+SU^D+2)}}_{\text{SU trajectory}} + \underbrace{\sum_{i=2}^{SD^D+1} P_i^{SD} w_{(t-i+2)}}_{\text{SD trajectory}} + \underbrace{\underline{P}(u_t + v_{t+1}) + p_t}_{\text{Traditional Output}}$$

- Key aspects to not destroy the *convex hull*:
 - Output above and below \underline{P} are managed independently
 - Overlapping is avoided by using the min up/down constraints

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- Different SU ramps can be easily included in a similar way, for either energy- or power-based formulations¹²

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Case Studies

- Formulations tested –modeling the same MIP problem:
 - TC^{13} : Proposed Tight & Compact
 - $1bin^{14}$: 1-binary variable (u)
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All results are expressed as percentages of $1bin$ results

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Case Study A: Self-UC (I)

Results presented as percentages of *1bin*:

	<i>3bin</i> (%)	<i>TC</i> (%)
Constraints	<78	<48
Nonzeros	89	72
Real Vars	33.3	33.3
Bin Vars	=300	=300

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Integrality Gap	34	=0



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Bin Vars	=300	=300
Integrality Gap	34	=0
MIP Sol.Time	6.3	0.192
MIP Sol.Time (best-worst)	0.46 - 87	0.001 - 3.3



TC is Tighter and **Simultaneously** more Compact

Case Study A: Self-UC (I)

Results presented as percentages of *1bin*:

	<i>3bin</i> (%)	<i>TC</i> (%)
Constraints	<78	<48
Nonzeros	89	72
Real Vars	33.3	33.3
Bin Vars	=300	=300
Integrality Gap	34	=0
MIP Sol.Time	6.3	0.192
MIP Sol.Time (best-worst)	0.46 - 87	0.001 - 3.3
LP Sol.Time	80	49.8



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Case Study A: Self-UC (II)

Performance of the Energy-Based formulations:

	<i>3bin (%)</i>	<i>TC (%)</i>	<i>TC SU&SD</i>
Constraints	<78	<48	<59
Nonzeros	89	72	95
Real Vars	33.3	33.3	66.7
Bin Vars	=300	=300	=300
Integrality Gap	34	=0	=0
MIP Sol.Time	6.3	0.192	0.196
MIP Sol.Time (best-worst)	0.46 - 87	0.001 - 3.3	0.001 - 3.3
LP Sol.Time	80	49.8	54.1

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The Energy-Based *TC* formulations describe the *convex hull* then solving MIP (non-convex) as LP (convex)

Case Study A: Self-UC (II)

Performance of the Power-Based formulations *P-TC*:

	<i>3bin</i> (%)	<i>P-TC</i> (%)	<i>P-TC SU&SD</i>
Constraints	<78	<45	<56
Nonzeros	89	67	92
Real Vars	33.3	33.3	66.7
Bin Vars	=300	=300	=300
Integrality Gap	34	=0	=0
MIP Sol.Time	6.3	0.18	0.191
MIP Sol.Time (best-worst)	0.46 - 87	0.001 - 3.2	0.001 - 3.2
LP Sol.Time	80	43.3	45.9

Case Study A: Self-UC (II)

Performance of the Power-Based formulations P -TC:

	$3bin$ (%)	P -TC (%)	P -TC $SU\&SD$
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Case Study B: UC for 40 power system mixes

Results presented as percentages of *1bin*:

	<i>3bin</i> (%)	<i>TC</i> (%)
Constraints	<99	<40
Nonzeros	~100	<35
Real Vars	75	50
Bin Vars	=300	<500
Integrality Gap	72	40

Case Study B: UC for 40 power system mixes

Results presented as percentages of *1bin*:

	<i>3bin</i> (%)	<i>TC</i> (%)
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TC is Tighter **and Simultaneously** more Compact

Case Study B: UC for 40 power system mixes

Results presented as percentages of *1bin*:

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Real Vars	75	50
Bin Vars	=300	<500
Integrality Gap	72	40
Total Average Sol.Time	71	7
Sol.Time (best-worst)	11 – 269	2 – 57



TC is Tighter **and Simultaneously** more Compact

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Total Average Sol.Time	71	7
Sol.Time (best-worst)	11 – 269	2 – 57
Sol.Time Small Cases	67	11
Sol.Time Large Cases	77	4.5



TC is Tighter **and Simultaneously** more Compact

Outline

- 1 Introduction
- 2 Good and Ideal MIP formulations
 - The Ideal MIP Formulation
 - Good MIP Formulations
- 3 Tight & Compact (TC) UC Formulations
 - Traditional Formulation
 - Power-Based UC
 - Startup & Shutdown Ramps
- 4 **Numerical Results & Further UC Extensions**
 - Basic UC Formulations
 - **Stochastic UC**
 - Ramp-Based Scheduling
- 5 Conclusions

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Stochastic UC: Case Study

- 10 generating units for a time span of 4 days
- 10 to 200 scenarios in demand
- New formulation included: Sh^{16}
- Different Solvers
 - Cplex 12.5.1
 - Gurobi 5.5
 - XPRESS 24.01.04

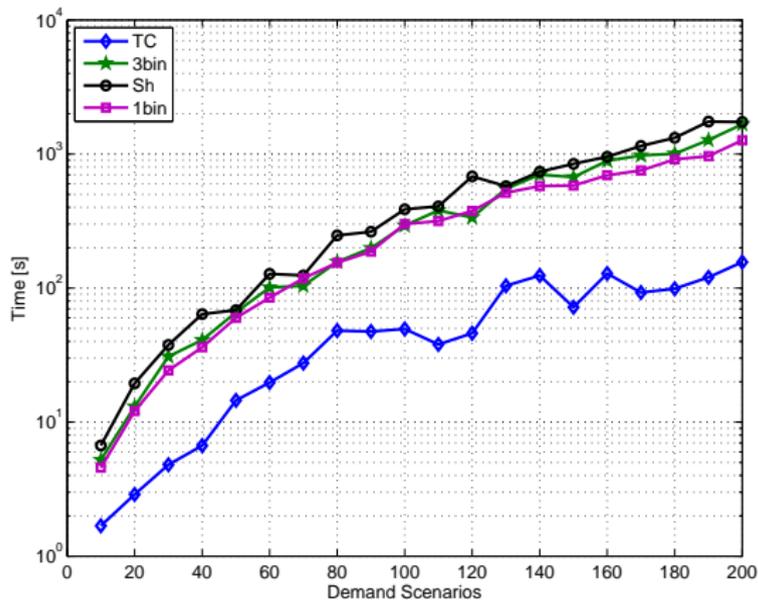
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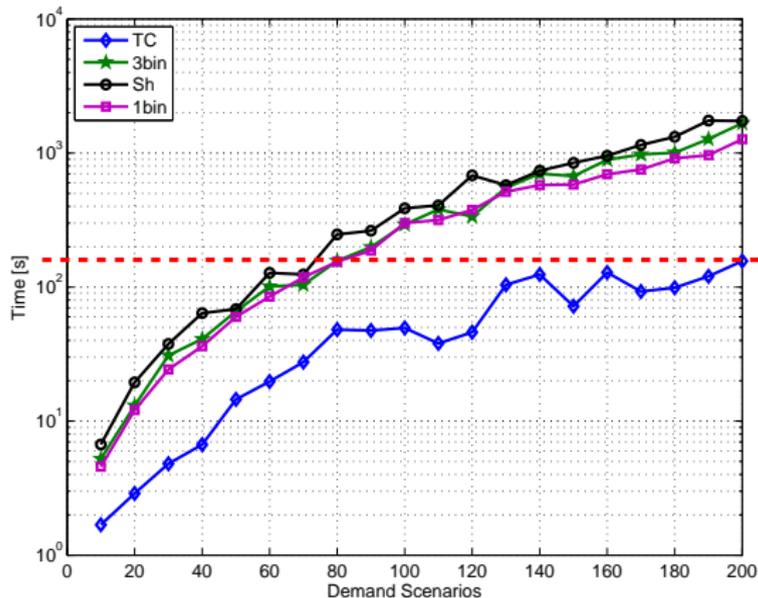
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- 10 to 200 scenarios in demand
- New formulation included: Sh^{16}
- Different Solvers
 - Cplex 12.5.1
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 - XPRESS 24.01.04
- Stop criteria:
 - Time limit: 5 hours or
 - Optimality tolerance: 0.01

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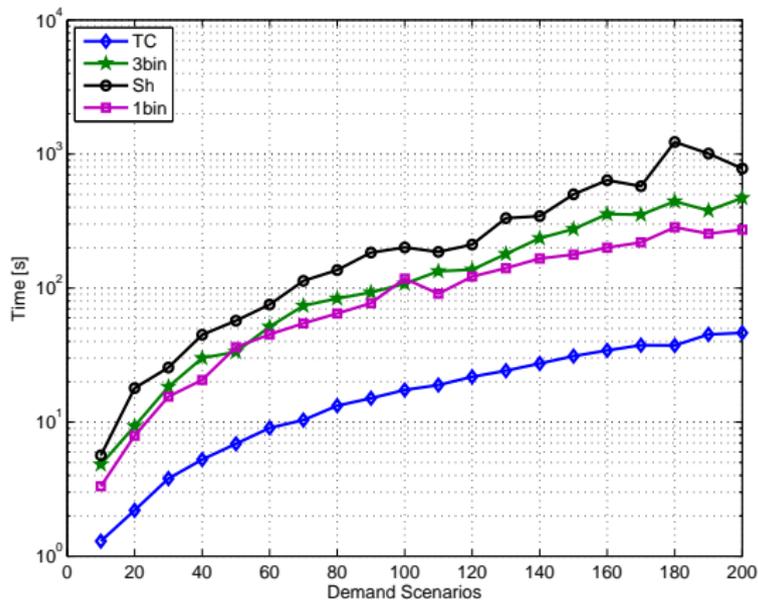


Stochastic: Cplex

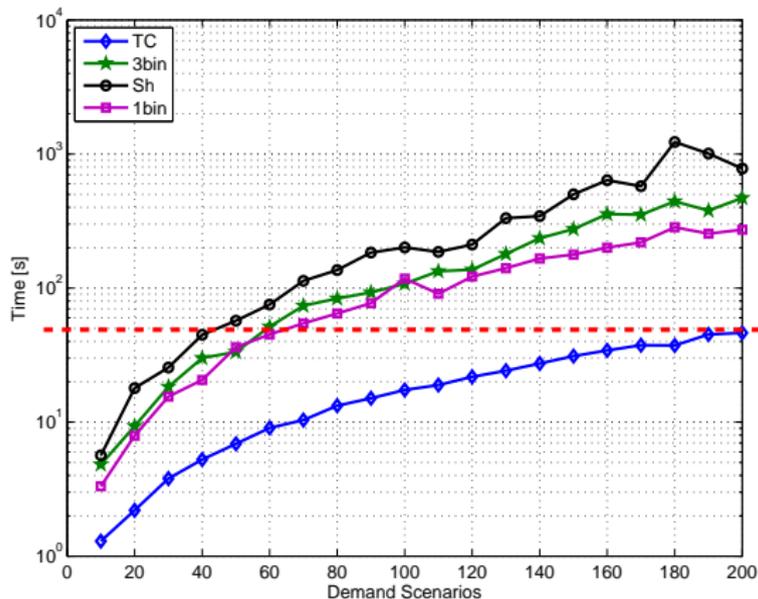


TC deals with 200 scenarios within the time that others deal with 80

Stochastic: Gurobi

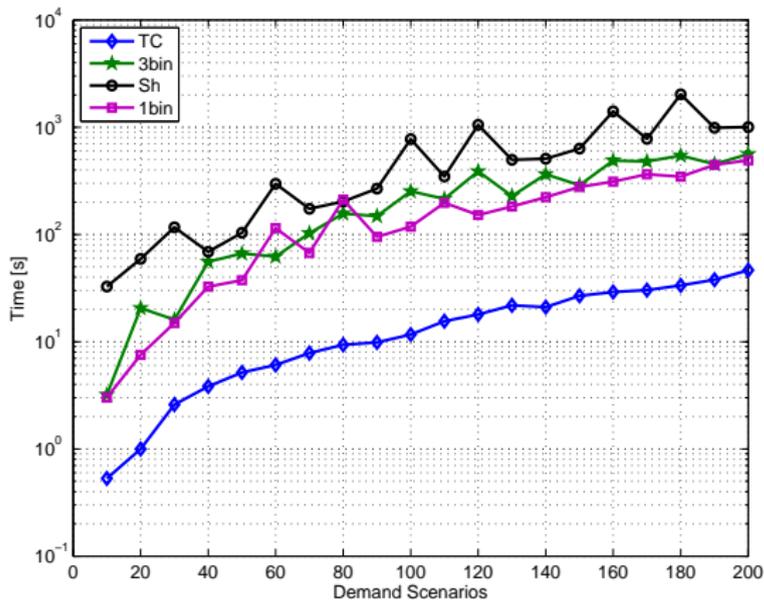


Stochastic: Gurobi

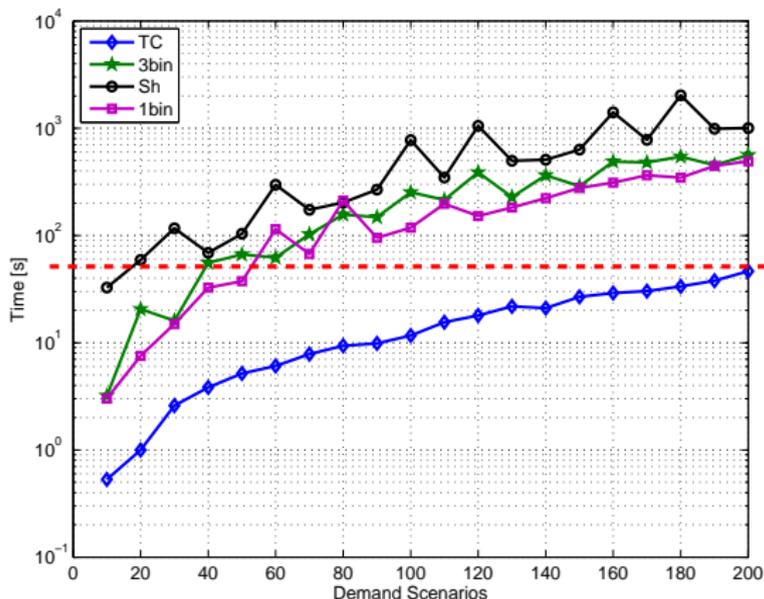


TC deals with 200 scenarios within the time that others deal with 60

Stochastic: XPRESS



Stochastic: XPRESS



TC deals with 200 scenarios within the time that others deal with 50

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Ramp-Based Scheduling Approach

- Some drawbacks of current UC formulations were identified¹⁷
- Then the TC-UC was reformulated for better scheduling (\downarrow costs)¹⁸, mainly by introducing new features, e.g.,
 - Linear piece-wise power scheduling
 - SU & SD power trajectories

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 - SU & SD power trajectories
- **The challenge:**
 - Trade-off: Model detail vs. Computation burden
 - SU & SD power trajectories was the main challenge
 - So, a **Tight & Compact MIP formulation for SU & SD trajectories** was proposed¹⁹

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Some Details per Unit

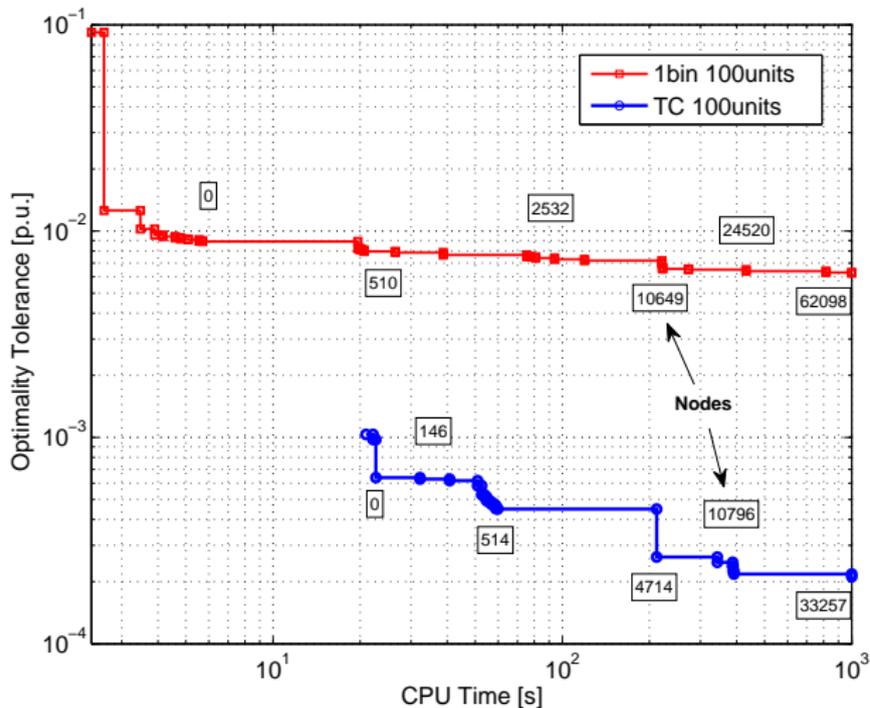
	<i>1bin</i>	<i>Ramp-Based</i>
Co-optimization	No	Yes
SU costs	3 types	3 types
SU ramps	–	3 types
Operating Ramps	2 types	6 types
Online reserves	1	4
Offline reserves	–	2

Problem Size

100 units for a time span of 24 hours

	<i>1bin</i>	<i>Ramp-Based</i>	Increase (%)
Constraints	40449	43271	6.97
Nonzeros	208445	217661	4.42
Real Vars	9624	15840	64.6
Bin Vars	2400	13650	468.8
Integrality Gap (%)	1.76	0.333	-81.1

Convergence Evolution



Conclusions

- **Beware** of what matters in good MIP formulations
 - **Tightness**: defines the search space
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- **Better UC core** \Rightarrow critical solving time reductions in further UC extensions

For Further Reading



R. Bixby and E. Rothberg, “Progress in computational mixed integer programming—A look back from the other side of the tipping point,” *Annals of Operations Research*, vol. 149, no. 1, pp. 37–41, Jan. 2007.



M. Carrion and J. Arroyo, “A computationally efficient mixed-integer linear formulation for the thermal unit commitment problem,” *IEEE Transactions on Power Systems*, vol. 21, no. 3, pp. 1371–1378, 2006.



T. Christof and A. Löbel, “PORTA: POLYhedron representation transformation algorithm, version 1.4.1,” *Konrad-Zuse-Zentrum für Informationstechnik Berlin, Germany*, 2009.



X. Guan, F. Gao, and A. Svoboda, “Energy delivery capacity and generation scheduling in the deregulated electric power market,” *IEEE Transactions on Power Systems*, vol. 15, no. 4, pp. 1275–1280, Nov. 2000.



T. Koch, T. Achterberg, E. Andersen, O. Bastert, T. Berthold, R. E. Bixby, E. Danna, G. Gamrath, A. M. Gleixner, S. Heinz, A. Lodi, H. Mittelmann, T. Ralphs, D. Salvagnin, D. E. Steffy, and K. Wolter, “MIPLIB 2010,” *Mathematical Programming Computation*, vol. 3, no. 2, pp. 103–163, Jun. 2011.

For Further Reading (cont.)



T. Li and M. Shahidehpour, “Price-based unit commitment: a case of lagrangian relaxation versus mixed integer programming,” *IEEE Transactions on Power Systems*, vol. 20, no. 4, pp. 2015–2025, Nov. 2005.



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G. Morales España, C. Gentile, and A. Ramos, “Tight LP formulation of the unit commitment problem presenting integer solutions,” *IEEE Transactions on Power Systems*, 2013, Under Review.



G. Morales-Espana, A. Ramos, and J. Garcia-Gonzalez, “An MIP formulation for joint market-clearing of energy and reserves based on ramp scheduling,” *IEEE Transactions on Power Systems*, 2013, Special Section on Electricity Markets Operation. In Press.



G. Morales-Espana, J. M. Latorre, and A. Ramos, “Tight and compact MILP formulation for the thermal unit commitment problem,” *IEEE Transactions on Power Systems*, 2013, Special Section on Analysis and Simulation of Very Large Power Systems. In Press.

For Further Reading (cont.)



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J. Ostrowski, M. F Anjos, and A. Vannelli, “Tight mixed integer linear programming formulations for the unit commitment problem,” *IEEE Transactions on Power Systems*, vol. 27, no. 1, pp. 39–46, Feb. 2012.



D. Rajan and S. Takriti, “Minimum Up/Down polytopes of the unit commitment problem with start-up costs,” IBM, Research Report RC23628, Jun. 2005.



H. P. Williams, *Model Building in Mathematical Programming*, 5th Edition. John Wiley & Sons Inc, Feb. 2013.



L. Wolsey, *Integer Programming*. Wiley-Interscience, 1998.