

# Accelerating the Convergence of MIP-based Unit Commitment Problems

The Impact of High Quality MIP Formulations

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Optimization on Friday  
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# Outline

- 1 Short-term Planning (UC)
  - Unit Commitment
  - Why MIP?
  - Improving MIP Formulations
- 2 Case Studies
  - Deterministic Self-UC
  - Stochastic UCs: Different Solvers
  - Stochastic UCs: IEEE-118 Bus System
- 3 Conclusions & Collaboration?

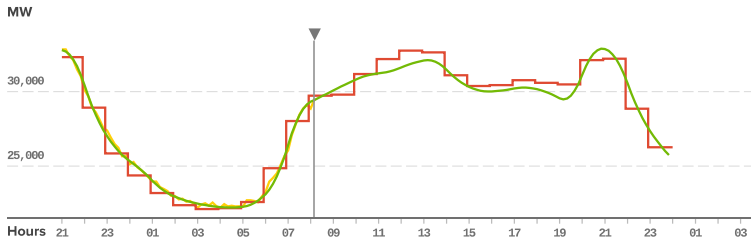
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# Day-ahead Planning

## GENERATION AND CONSUMPTION

29 / 09 / 2015 | 08:10



 REAL DEMAND

 SCHEDULED DEMAND

 FORECASTED DEMAND

**29,532** MW

29,725 MW

29,448 MW

The system **has to** respond to instantaneous variations of the demand

# Unit Commit for Short-term Planning

## ■ Objective:

- To schedule generating units (thermal, nuclear, hydro, etc.) at **minimum cost** (bids) over the study period subject to:
  - Meet system demand for electricity
  - Satisfy the system constraints (e.g., technical, environmental)
  - Provide a level of **flexibility (reserves) to accommodate RES energy**

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## ■ Planning:

- the physical operation
  - To make **Startup** and **shutdown** decisions
  - To obtain hourly schedules (production) for all generating units
- the economic issues
  - **Give market signals** (prices)
  - Forecast operational cost

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- Non-linear generation range
  - $\text{MinGeneration} \cdot u \leq p \leq \text{MaxGeneration} \cdot u$
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- **MIP: Global optimum or within a tolerance**

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  - California **7-day** (UC) model:  
48939 constraints, 25755 variables (2856 binary)
    - **Reported results 1989** – machine unknown
    - 2 day model: **8 hours, no progress**
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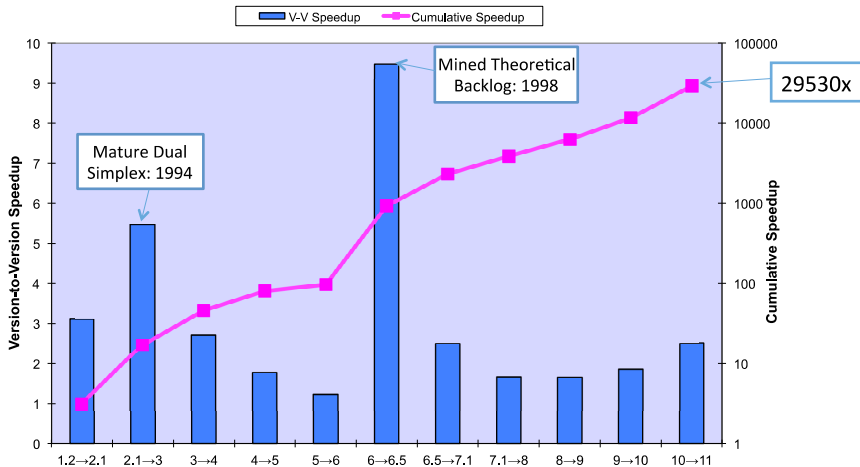


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- California 7-day model (on a desktop PC)
  - CPLEX 6.5 (1999): 22 minutes, optimal
  - CPLEX 11.0 (2007): 71 seconds, optimal

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- $\sim$  **3x** faster year-to-year

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- **So, what is left to do?**

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- How to reduce solving times?
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  - Improving the MIP-Based UC formulation  $\Rightarrow$   $\downarrow$  solving times



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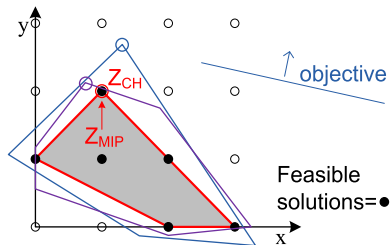
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## 3 Conclusions & Collaboration?

# Convex Hull: The Tightest Formulation

## Convex Hull (CH)

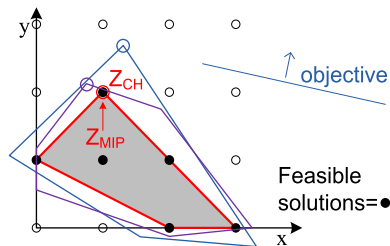
Smallest convex feasible region containing all feasible integer points



# Convex Hull: The Tightest Formulation

## Convex Hull (CH)

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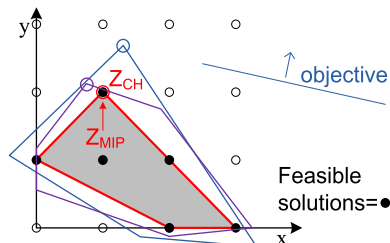


- The *convex hull* problem solves an MIP as an LP
  - Each vertex satisfies the integrality constraints
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## Convex Hull (CH)

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- 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
  - To solve an MIP is usually easier than trying to find its *convex hull*

## 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

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  - 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

# Tight and Compact (TC) Formulation

- Let's focus on the core of UC formulations:
    - Min/max outputs
    - SU & SD ramps
    - Minimum up/down ( $TU/TD$ ) times
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- Let's focus on the core of UC formulations:
  - Min/max outputs
  - SU & SD ramps
  - Minimum up/down ( $TU/TD$ ) times, *convex hull already available*<sup>1</sup>
- The whole formulation can be found in the paper **TC-UC**<sup>2</sup> and the convex hull proof in *gentile et al.*<sup>3</sup>

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<sup>1</sup>D. Rajan and S. Takriti, "Minimum Up/Down Polytopes of the Unit Commitment Problem with Start-Up Costs," IBM, Research Report RC23628, Jun. 2005

<sup>2</sup>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*, vol. 28, no. 4, pp. 4897–4908, Nov. 2013

<sup>3</sup>C. Gentile, G. Morales-España, and A. Ramos, "A tight MIP formulation of the unit commitment problem with start-up and shut-down constraints," *en, EURO Journal on Computational Optimization*, vol. 5, no. 1, pp. 177–201, Apr. 2016

## Formulation for a generating unit (I)

- Generation limits taking into account:

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

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

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

Variables		Parameters	
$p_t$	Energy production above $\underline{P}$	$\underline{P}$	Minimum power output
$u_t$	Commitment status	$\overline{P}$	Maximum power output
$v_t$	Startup status	$SU$	Startup ramp
$w_t$	Shutdown status	$SD$	Shutdown ramp

## Formulation for a generating unit (II)

- Logical relationship:

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

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

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

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 PORTA<sup>4</sup>:

- PORTA enumerates all vertices of a convex feasible region

---

<sup>4</sup>T. 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 PORTA<sup>4</sup>:

- PORTA enumerates all vertices of a convex feasible region
- Example: 3 periods and  $\overline{P} = 200, \underline{P} = SU = SD = 100$  for:
  - Case 1:  $TU = TD = 1$
  - Case 2:  $TU = TD = 2$

---

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## Case 1: Providing The Convex Hull

Formulation:

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

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

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

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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 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
( 7) 1 1 0 0 0 0 1 0 0 0 0
( 8) 1 1 0 0 0 0 1 100 0 0
( 9) 1 1 1 0 0 0 0 0 0 0 0
(10) 1 1 1 0 0 0 0 0 0 0 100
(11) 1 1 1 0 0 0 0 0 0 100 0
(12) 1 1 1 0 0 0 0 0 0 100 100
(13) 1 1 1 0 0 0 0 100 0 0
(14) 1 1 1 0 0 0 0 100 0 100
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(17) 1 0 1 0 1 1 0 0 0 0
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All vertices are integer



Convex Hull

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END

## Case 2: Providing and Using Convex Hulls (I)

Formulation + *TU/TD Convex hull*:

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

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

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

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

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

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

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	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  $TU \geq 2$ :

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

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

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

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

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

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

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
(15)	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}
 & \cancel{p_t \leq (\bar{P} - \underline{P}) u_t - (\bar{P} - SD) w_{t+1}} \\
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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
(15)	1	1	1	0	0	0	0	100	100	100

END

⇒ Convex Hull

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## Self-UC case Study

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  - $TC^5$  Tight & Compact: a  $2^{3 \times 10 \times 512}$  combinatorial problem
  - $1bin^6$  1-binary variable ( $u$ ):  $2^{10 \times 512}$  combinations
  - $3binTUTD^7$ : 3-binary variable version ( $u, v, w$ ) +  $TU/TD$  convex hull:  $2^{3 \times 10 \times 512}$  combinations

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<sup>5</sup>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*, vol. 28, no. 4, pp. 4897–4908, Nov. 2013

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- All results are expressed as percentages of  $1bin$  results

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## Case Study: Self-UC (I)

Results presented as percentages of *1bin*:

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



*TC* is more Compact

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Integrality Gap	34	=0



*TC* is Tighter **and Simultaneously** more Compact

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	<i>3binTUTD</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 runtime (speedup)	4.9 (20x)	0.107 (995x)



*TC* is Tighter **and Simultaneously** more Compact

## Case Study: Self-UC (II)

Results presented as percentages of *1bin*:

	<i>3binTUTD</i> (%)	<i>TC</i> (%)
Constraints	<78	<48
Nonzeros	89	72
Real Vars	33.3	33.3
Bin Vars	=300	=300
Integrality Gap	34	<b>=0</b>
MIP runtime	4.9	0.107
LP runtime	80	<b>49.8</b>



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

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- 10 generating units for a time span of 2 days
  - 10 to 200 scenarios in demand
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<sup>9</sup>J. Arroyo and A. Conejo, "Optimal response of a thermal unit to an electricity spot market," *Power Systems, IEEE Transactions on*, vol. 15, no. 3, pp. 1098–1104, 2000

# Stochastic UC: Case Study

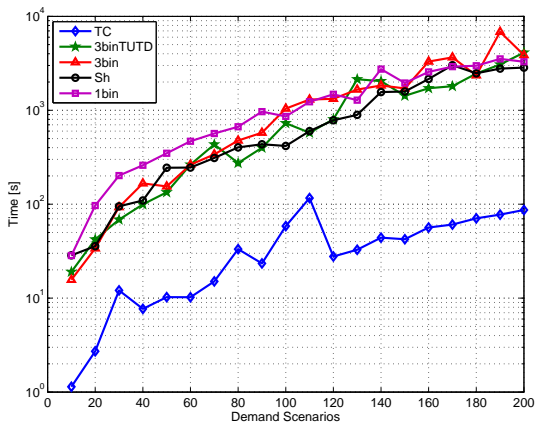
- 10 generating units for a time span of 2 days
- 10 to 200 scenarios in demand
- Comparing *TC*, *1bin*, *3binTUTD* and Two additional formulations *Sh*<sup>8</sup> and *3bin*<sup>9</sup>
- Different Solvers
  - Cplex 12.6.0
  - Gurobi 5.6.2
  - XPRESS 25.01.07
- Stop criteria:
  - Time limit: 5 hours or
  - Optimality tolerance: 0.1 %

---

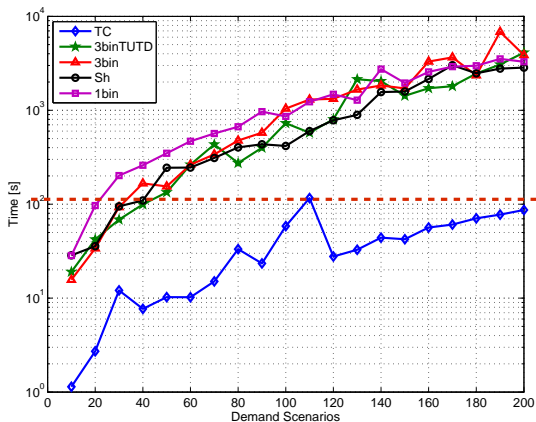
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# Stochastic: Cplex

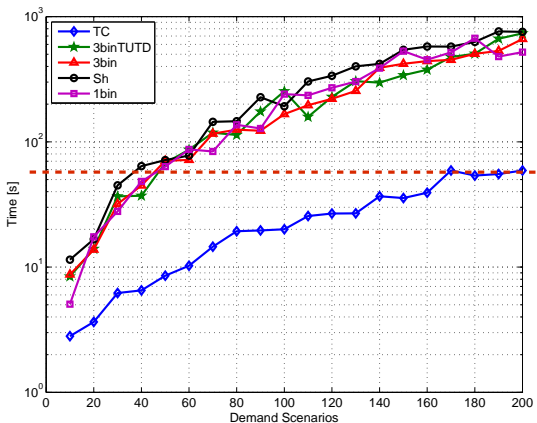


# Stochastic: Cplex



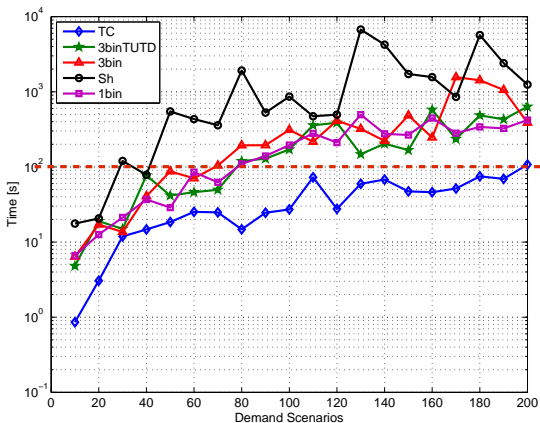
TC deals with 200 scenarios within the time that others deal with 40

# Stochastic: Gurobi



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

## Stochastic: XPRESS



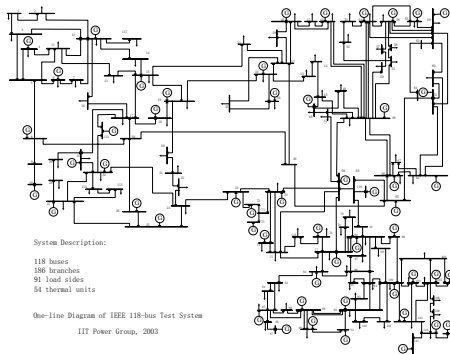
TC deals with 200 scenarios within the time that others deal with 80

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# IEEE-118 Bus System



- 54 thermal units; 118 buses; 186 transmission lines; 91 loads
  - 24 hours time span
  - 3 wind farms, 20 wind power scenarios
  - Stop Criteria in Cplex 12.6.0
    - 0.05% opt. tolerance or 24h time limit

## UC performance comparisons (I)

	Traditional Energy-Block Scheduling	
	$3binTUTD^{10}$	$TC$
o.f. [k\$]	829.04	829.02
opt.tol [%]	0.224	0.023
IntGap [%]	1.27	0.58

- Compared with  $3binTUTD$ ,  $TC$ :
  - lowered IntGap by 53.3%

<sup>10</sup>FERC, "RTO Unit Commitment Test System," Federal Energy and Regulatory Commission, Washington DC, USA, Tech. Rep., Jul. 2012, p. 55

# UC performance comparisons (I)

	Traditional Energy-Block Scheduling	
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o.f. [k\$]	829.04	829.02
opt.tol [%]	0.224	0.023
IntGap [%]	1.27	0.58
MIP runtime [s]	<b>86400</b>	<b>206.5</b>

- Compared with *3binTUTD*, *TC*:
  - lowered IntGap by 53.3%
  - is more than 420x faster

<sup>10</sup>FERC, "RTO Unit Commitment Test System," Federal Energy and Regulatory Commission, Washington DC, USA, Tech. Rep., Jul. 2012, p. 55

## UC performance comparisons (II)

	Traditional Energy-Block Scheduling	
	<i>3binTUTD</i> <sup>11</sup>	<i>TC</i>
o.f. [k\$]	829.04	829.02
opt.tol [%]	<b>0.224</b>	0.023
IntGap [%]	1.27	0.58
MIP runtime [s]	86400	<b>206.5</b>
LP runtime [s]	<b>246.76</b>	22.03

- *TC* solved the MIP before *3binTUTD* solved the LP
  - within the required opt. tolerance (0.05%)

<sup>11</sup>FERC, "RTO Unit Commitment Test System," Federal Energy and Regulatory Commission, Washington DC, USA, Tech. Rep., Jul. 2012, p. 55

## UC performance comparisons (III)

	Traditional Energy-based UC		Power-Based UC
	<i>3binTUTD</i>	<i>TC</i>	<i>P-TC</i>
o.f. [k\$]	829.04	829.02	<b>818.13</b>
opt.tol [%]	0.224	0.023	0.049
IntGap [%]	1.27	0.58	
MIP runtime [s]	<b>86400</b>	206.5	
LP runtime [s]	246.76	22.03	

- *P-TC*<sup>12</sup> has a more detailed and accurate UC representation

<sup>12</sup>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*, vol. 29, no. 1, pp. 476–488, 2014

## UC performance comparisons (III)

	Traditional Energy-based UC		Power-Based UC
	<i>3binTUTD</i>	<i>TC</i>	<i>P-TC</i>
o.f. [k\$]	829.04	829.02	<b>818.13</b>
opt.tol [%]	0.224	0.023	0.049
IntGap [%]	1.27	0.58	0.74
MIP runtime [s]	<b>86400</b>	206.5	<b>867.9</b>
LP runtime [s]	246.76	22.03	38.1

- *P-TC*<sup>12</sup> has a more detailed and accurate UC representation
  - it solved 100x faster than *3binTUTD*
  - its UC core is also a **convex hull**<sup>13</sup>

<sup>12</sup>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*, vol. 29, no. 1, pp. 476–488, 2014

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# Conclusions (I)

- Beware of what matters in good MIP formulations
    - Tightness & Compactness
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# Conclusions (I)

- Beware of what matters in good MIP formulations
  - Tightness & Compactness
  - $\uparrow$  Binaries  $\Rightarrow$   $\uparrow$  Solving time **False myth**
- Use the *convex hull* of some set of constraints
  - Minimum up/down times<sup>14</sup>
  - Unit operation in Energy-based UC<sup>15</sup>
  - Unit operation in Power-based UC<sup>16</sup>
  - $\Rightarrow$   $\downarrow$  solving time by **simultaneously T&Cing** the final UCs<sup>17,18</sup>

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<sup>14</sup>D. Rajan and S. Takriti, "Minimum Up/Down Polytopes of the Unit Commitment Problem with Start-Up Costs," IBM, Research Report RC23628, Jun. 2005

<sup>15</sup>C. Gentile, G. Morales-España, and A. Ramos, "A tight MIP formulation of the unit commitment problem with start-up and shut-down constraints," *en, EURO Journal on Computational Optimization*, vol. 5, no. 1, pp. 177–201, Apr. 2016

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<sup>18</sup>G. 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

# Conclusions

- Better UC core in stochastic UCs  $\Rightarrow$ 
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# Conclusions

- **Better UC core** in stochastic UCs  $\Rightarrow$ 
  - Critical solving time reductions
- If **convex hulls** are not available  $\Rightarrow$ 
  - Create simultaneously tight and compact models
  - by reformulating the problem, e.g., CCGTs<sup>19</sup>
  - **Key hint:** start removing all **big-M parameters**

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<sup>19</sup>G. Morales-España, C. M. Correa-Posada, and A. Ramos, "Tight and Compact MIP Formulation of Configuration-Based Combined-Cycle Units," *IEEE Transactions on Power Systems*, vol. 31, no. 2, pp. 1350–1359, Mar. 2016

## Possible collaboration?

- **With PhD Students** (papers envisioned)
  - Convex hull for variable SU costs
    - Almost there (just one constraint needs to be added)
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- **Master theses** with **Jedlix**
  - Robust smart charging of EVs
  - Aggregated EV models to deal with uncertainty



# Questions

Thank you for your attention

Contact Information:  
g.a.moralesespama@tudelft.nl

## For Further Reading



J. Arroyo and A. Conejo, “Optimal response of a thermal unit to an electricity spot market,” *Power Systems, IEEE Transactions on*, vol. 15, no. 3, pp. 1098–1104, 2000.



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T. Christof and A. Löbel, “PORTA: POLYhedron Representation Transformation Algorithm, Version 1.4.1,” *Konrad-Zuse-Zentrum für Informationstechnik Berlin, Germany*, 2009.



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D. Rajan and S. Takriti, "Minimum Up/Down Polytopes of the Unit Commitment Problem with Start-Up Costs," IBM, *Research Report RC23628*, Jun. 2005.