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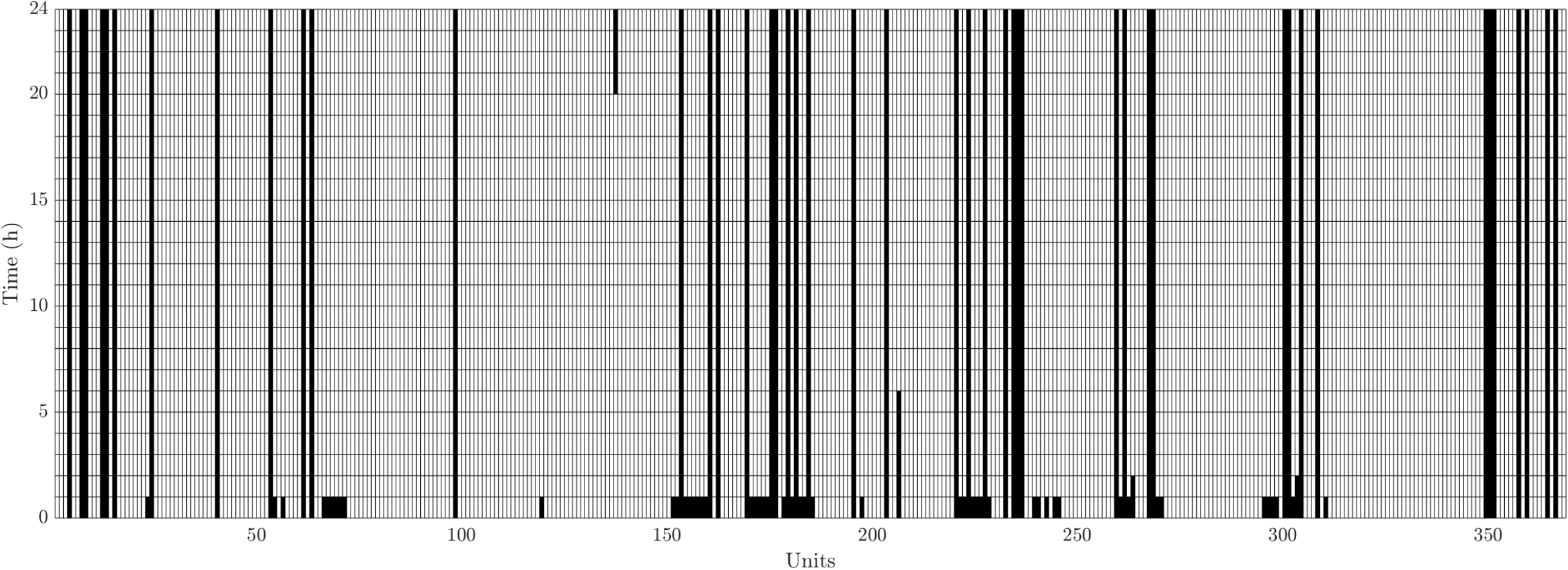


AC Network-Constrained Unit Commitment (AC-NCUC)

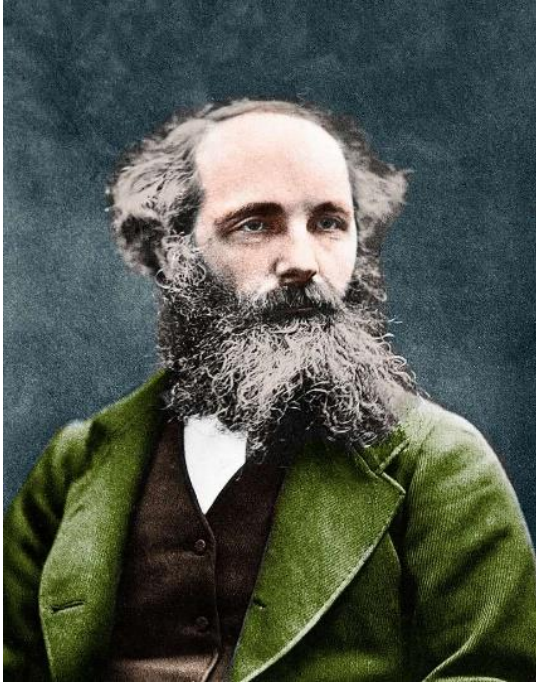
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April 20, 2021

Binary decisions!



Nonlinearities and non-convexities!

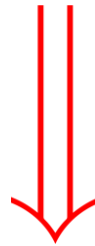


$$\nabla \cdot \mathbf{D} = \rho$$

$$\nabla \cdot \mathbf{B} = 0$$

$$\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t}$$

$$\nabla \times \mathbf{H} = \mathbf{J} + \frac{\partial \mathbf{D}}{\partial t}$$



$$p_{nm}(\cdot) = v_n^2 y_{Lnm} \cos(\theta_{Lnm}) - v_n v_m y_{Lnm} \cos(\delta_n - \delta_m - \theta_{Lnm}) + \frac{1}{2} v_n^2 y_{Snm} \cos(\theta_{Snm})$$

$$q_{nm}(\cdot) = -v_n^2 y_{Lnm} \sin(\theta_{Lnm}) - v_n v_m y_{Lnm} \sin(\delta_n - \delta_m - \theta_{Lnm}) - \frac{1}{2} v_n^2 y_{Snm} \sin(\theta_{Snm})$$

$$\begin{aligned}
 p_{nm}(\cdot) &= v_n^2 y_{Lnm} \cos(\theta_{Lnm}) - v_n v_m y_{Lnm} \cos(\delta_n - \delta_m - \theta_{Lnm}) + \frac{1}{2} v_n^2 y_{Snm} \cos(\theta_{Snm}) \\
 q_{nm}(\cdot) &= -v_n^2 y_{Lnm} \sin(\theta_{Lnm}) - v_n v_m y_{Lnm} \sin(\delta_n - \delta_m - \theta_{Lnm}) - \frac{1}{2} v_n^2 y_{Snm} \sin(\theta_{Snm})
 \end{aligned}$$



$$\begin{aligned}
 p_{nm}(\cdot) &= G_{nm}(e_m e_n + f_m f_n) - B_{nm}(e_m f_n - e_n f_m) - G_{nm}(e_n^2 + f_n^2) \\
 q_{nm}(\cdot) &= -G_{nm}(e_m f_n - e_n f_m) - B_{mn}(e_m e_n + f_m f_n) + (B_{nm} - b_{nm}^{\text{shunt}})(e_n^2 + f_n^2)
 \end{aligned}$$

The Unit Commitment (UC) problem consists of determining, for a planning horizon (typically 1 day), the hourly start-up and shut-down schedule of all production units so that the electric demand is supplied and total operating costs are minimized, while satisfying different technical and security constraints (network and production units).

Objective function (cost)

minimize

$$\sum_{t \in \mathcal{T}} \left(\sum_{j \in \mathcal{G}^c \cup \mathcal{G}^g} (c_j(p_j(t)) + c_j^U y_j(t) + c_j^D z_j(t)) + \sum_{j \in \mathcal{G}^w} c_j(p_j^W(t)) + \sum_{j \in \mathcal{G}^{\text{ng}}} c_j(p_j(t)) + \sum_{n \in \mathcal{N}} (c_n^{\text{LOL}} D_n^{\text{shed}}(t) + c_n^{\text{shed}} Q_n^{\text{shed}}(t)) \right) \quad (1)$$

subject to:

Start-up and shut-down logic, ramping

$$v_j(t-1) - v_j(t) + y_j(t) - z_j(t) = 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, t = 2, \dots, |\mathcal{T}| \quad (2)$$

$$p_j(t) - p_j(t-1) \leq R_j^U v_j(t-1) + S_j^U y_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, t = 2, \dots, |\mathcal{T}| \quad (3)$$

$$p_j(t-1) - p_j(t) \leq R_j^D v_j(t) + S_j^D z_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, t = 2, \dots, |\mathcal{T}| \quad (4)$$

$$v_j^{\text{ini}} - v_j(1) + y_j(1) - z_j(1) = 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (5)$$

$$p_j(1) - p_j^{\text{ini}} \leq R_j^U v_j^{\text{ini}} + S_j^U y_j(1); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (6)$$

$$p_j^{\text{ini}} - p_j(1) \leq R_j^D v_j(1) + S_j^D z_j(1); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (7)$$

Power buffer

$$\underline{P}_j v_j(t) \leq p_j(t) \leq \bar{p}_j(t) \leq \bar{P}_j v_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \forall t \in \mathcal{T} \quad (8)$$

$$\underline{P}_j \leq p_j(t) \leq \bar{p}_j(t) \leq \bar{P}_j; \quad \forall j \in \mathcal{G}^{\text{ng}}, \forall t \in \mathcal{T} \quad (9)$$

$$\bar{p}_j(t) \leq p_j(t-1) + R_j^{\text{U}} v_j(t-1) + S_j^{\text{U}} y_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, t = 2, \dots, |\mathcal{T}| \quad (10)$$

$$\bar{p}_j(t) \leq p_j(t-1) + R_j^{\text{U}}; \quad \forall j \in \mathcal{G}^{\text{ng}}, \quad t = 2, \dots, |\mathcal{T}| \quad (11)$$

$$\bar{p}_j(t) \leq \bar{P}_j [v_j(t) - z_j(t+1)] + z_j(t+1) S_j^{\text{D}}; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \forall t \in \mathcal{T} \quad (12)$$

$$\bar{p}_j(1) \leq p_j^{\text{ini}} + R_j^{\text{U}} v_j^{\text{ini}} + S_j^{\text{U}} y_j(1); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (13)$$

$$\bar{p}_j(1) \leq p_j^{\text{ini}} + R_j^{\text{U}}; \quad \forall j \in \mathcal{G}^{\text{ng}} \quad (14)$$

Reactive power bounds

$$\underline{Q}_j v_j(t) \leq q_j(t) \leq \overline{Q}_j v_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \forall t \in \mathcal{T} \quad (15)$$

$$\underline{Q}_j \leq q_j(t) \leq \overline{Q}_j; \quad \forall j \in \mathcal{G}^{\text{ng}}, \forall t \in \mathcal{T} \quad (16)$$

$$\underline{Q}_j^{\text{N}} \leq q_j^{\text{N}}(t) \leq \overline{Q}_j^{\text{N}}; \quad \forall j \in \mathcal{G}^{\text{nuc}}, \forall t \in \mathcal{T} \quad (17)$$

Minimum up time

$$\sum_{t=1}^{L_j} [1 - v_j(t)] = 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (18)$$

$$\sum_{i=t}^{k+T_j^U-1} v_j(t) \geq T_j^U y_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \quad \forall t \in [L_j + 1, \dots, |\mathcal{T}| - T_j^U + 1] \quad (19)$$

$$\sum_{i=t}^{|\mathcal{T}|} [v_j(t) - y_j(t)] \geq 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \quad \forall t \in [|\mathcal{T}| - T_j^U + 2, \dots, |\mathcal{T}|] \quad (20)$$

Minimum down time

$$\sum_{t=1}^{F_j} v_j(t) = 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g \quad (21)$$

$$\sum_{i=t}^{k+T_j^D-1} [1 - v_j(t)] \geq T_j^D z_j(t); \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \quad \forall t \in [F_j + 1, \dots, |\mathcal{T}| - T_j^D + 1] \quad (22)$$

$$\sum_{i=t}^{|\mathcal{T}|} [1 - v_j(t) - z_j(t)] \geq 0; \quad \forall j \in \mathcal{G}^c \cup \mathcal{G}^g, \quad \forall t \in [|\mathcal{T}| - T_j^{rmD} + 2, \dots, |\mathcal{T}|] \quad (23)$$

Weather-dependent units

$$0 \leq p_j^W(t) \leq \bar{P}_j^W(t); \quad \forall j \in \mathcal{G}^W, \forall t \in \mathcal{T} \quad (24)$$

$$\underline{Q}_j^W(t) \leq q_j^W(t) \leq \bar{Q}_j^W(t); \quad \forall j \in \mathcal{G}^W, \forall t \in \mathcal{T} \quad (25)$$

Reserve

$$\sum_{j \in \Omega_r^G} \bar{p}_j(t) - p_j(t) \geq R_r(t); \quad \forall r \in \mathcal{R}, \forall t \in \mathcal{T} \quad (26)$$

Power balance

$$\begin{aligned} \sum_{j \in \Lambda_n^G} p_j(t) + \sum_{j \in \Lambda_n^W} p_j^W(t) + \sum_{j \in \Lambda_n^N} p_j^N(t) + D_n^{\text{shed}}(t) - \sum_{i \in \Lambda_n^D} D_i(t) = \\ \sum_{m \in \Lambda_n} [G_{nm} c_{nm}(t) - B_{nm} s_{nm}(t) - G_{nm} c_{nn}(t)]; \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \end{aligned} \quad (27)$$

$$\begin{aligned} \sum_{j \in \Lambda_n^G} q_j(t) + \sum_{j \in \Lambda_n^W} q_j^W(t) + \sum_{j \in \Lambda_n^N} q_j^N(t) + Q_n^{\text{shed}}(t) - \sum_{i \in \Lambda_n^D} Q_i(t) = \\ \sum_{m \in \Lambda_n} [-G_{nm} s_{nm}(t) - B_{mn} c_{nm}(t) + (B_{nm} - b_{nm}^{\text{shunt}}) c_{nn}(t)]; \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \end{aligned} \quad (28)$$

Transmission and voltage bounds

$$\begin{aligned} & (G_{nm}c_{nm}(t) - B_{nm}s_{nm}(t) - G_{nm}c_{nn}(t))^2 + \left(-G_{nm}s_{nm}(t) - B_{nm}c_{nm}(t) \right. \\ & \left. + (B_{nm} - b_{nm}^{\text{shunt}})c_{nn}(t) \right)^2 \leq \bar{S}_{nm}^2; \forall n \in \mathcal{N}, \forall m \in \Lambda_n, \forall t \in \mathcal{T} \end{aligned} \quad (29)$$

$$\underline{V}_n^2 \leq c_{nn}(t) \leq \bar{V}_n^2; \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \quad (30)$$

Non-convexities

Nonlinear and non-convex!

$$c_{nn}(t) = e_n^2(t) + f_n^2(t); \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \quad (31)$$

$$c_{nm}(t) = e_m(t)e_n(t) + f_m(t)f_n(t); \quad \forall n \in \mathcal{N}, \forall m \in \Lambda_n, \forall t \in \mathcal{T} \quad (32)$$

$$s_{nm}(t) = e_m(t)f_n(t) - e_n(t)f_m(t); \quad \forall n \in \mathcal{N}, \forall m \in \Lambda_n, \forall t \in \mathcal{T} \quad (33)$$

Unserviced energy

$$0 \leq D_n^{\text{shed}}(t) \leq \sum_{i \in \Lambda_n^{\text{D}}} D_i(t); \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \quad (34)$$

$$0 \leq Q_n^{\text{shed}}(t) \leq \sum_{i \in \Lambda_n^{\text{D}}} Q_i(t); \quad \forall n \in \mathcal{N}, \forall t \in \mathcal{T} \quad (35)$$

1. Why?
2. A three-step solution approach
3. Benders (if needed)
4. Two case studies

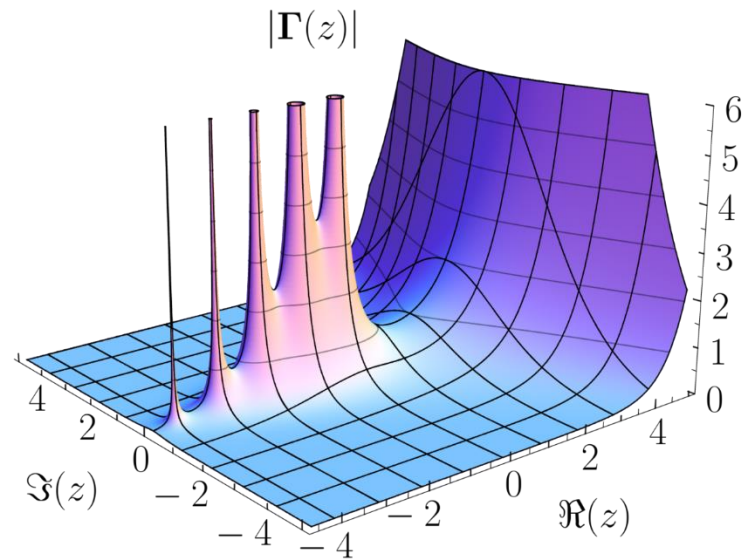
Why?

⇒ Increasing integration of weather-dependent renewable power.

⇒ Increasing demand (electric vehicles, data centers).

Reactive power / voltage constraints may alter commitment decisions.

$$p_{nm}(\cdot) = v_n^2 y_{Lnm} \cos(\theta_{Lnm}) - v_n v_m y_{Lnm} \cos(\delta_n - \delta_m - \theta_{Lnm}) + \frac{1}{2} v_n^2 y_{Snm} \cos(\theta_{Snm})$$
$$q_{nm}(\cdot) = -v_n^2 y_{Lnm} \sin(\theta_{Lnm}) - v_n v_m y_{Lnm} \sin(\delta_n - \delta_m - \theta_{Lnm}) - \frac{1}{2} v_n^2 y_{Snm} \sin(\theta_{Snm})$$



$$\min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{B}^m}$$

s.t.

$$f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y})$$

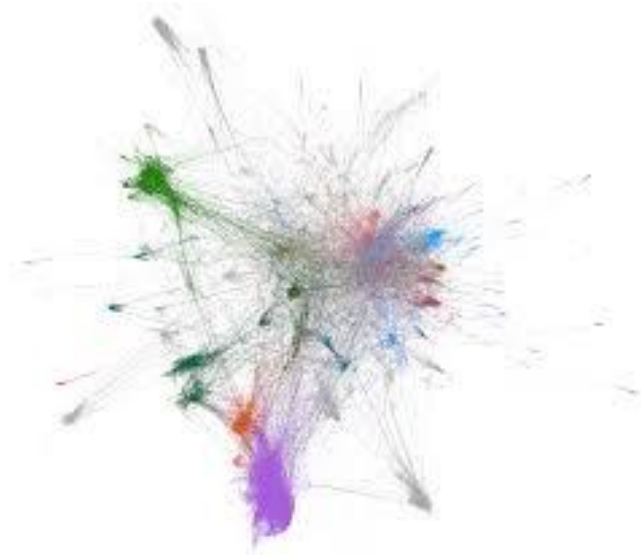
$$\mathbf{h}(\mathbf{x}, \mathbf{y}) = \mathbf{0}$$

$$\mathbf{g}(\mathbf{x}, \mathbf{y}) \leq \mathbf{0}$$



Large-scale

Nonlinear and nonconvex

Binary decisions



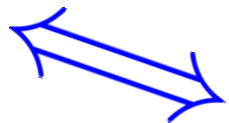
A three-step approach:

1. Linear NCUC (CPLEX, GUROBI)
2. Second order conic NCUC (CPLEX, GUROBI) 
3. Convex OPF (MOSEK) 

1. Solve a mixed-integer linear **DC-NCUC** and get (i) an initial solution and (ii) the set of potentially congested transmission lines.
2. Solve a mixed-integer second-order **conic relaxation of the AC-NCUC** using as initial solution the solution in 1 and an **active set strategy** on transmission line constraints (bounds).
3. Fixing the commitment decisions to those in 2, solve a sequence of **continuous convex problems** to ensure AC feasibility.

1

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{B}^m} \quad & f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}) \\ \text{s.t.} \quad & \mathbf{h}(\mathbf{x}, \mathbf{y}) = \mathbf{0} \\ & \mathbf{g}(\mathbf{x}, \mathbf{y}) \leq \mathbf{0} \end{aligned}$$



$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{B}^m} \quad & \mathbf{c}^\top [\mathbf{x} \ \mathbf{y}] \\ \text{s.t.} \quad & \mathbf{A}[\mathbf{x} \ \mathbf{y}] = \mathbf{b} \end{aligned}$$



$$[\mathbf{x}^0 \ \mathbf{y}^0]$$

Potentially congested lines

Large-scale

Nonlinear and nonconvex

Binary decisions

2

$$[x^0 \ y^0]$$

Potentially congested lines



$$\min_{x \in \mathbb{R}^n, y \in \mathbb{B}^m}$$

s.t.

$$f^{\text{CONVEX}}(x, y)$$

$$h^{\text{SOC}}(x, y) = 0$$

$$g_{\text{REDUCED}}^{\text{SOC}}(x, y) \leq 0$$



$$[x^c \ y^c]$$

Before 3

$\min_{\mathbf{x} \in \mathbb{R}^n}$
s.t.

$$f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^c)$$

$$h(\mathbf{x}, \mathbf{y}^c) = \mathbf{0}$$

$$g(\mathbf{x}, \mathbf{y}^c) \leq \mathbf{0}$$



$\min_{\mathbf{x} \in \mathbb{R}^n}$
s.t.

$$f^{\text{CONVEX}}(\mathbf{x})$$

$$\mathbf{u}^{\text{CONVEX}}(\mathbf{x}) - \mathbf{v}^{\text{CONVEX}}(\mathbf{x}) \leq \mathbf{0}$$



min
 $\mathbf{x} \in \mathbb{R}^n$
s.t.

$$f^{\text{CONVEX}}(\mathbf{x})$$

$$\mathbf{u}^{\text{CONVEX}}(\mathbf{x}) - \mathbf{v}^{\text{CONVEX}}(\mathbf{x}) \leq \mathbf{0}$$

3

\mathbf{y}^c



min
 $\mathbf{x} \in \mathbb{R}^n$
s.t.

$$f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^c)$$

$$\mathbf{u}^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^c) - \mathbf{v}^{\text{CONVEX LINEARIZED}}(\mathbf{x}, \mathbf{y}^c) \leq \mathbf{0}$$



$[\mathbf{x}^* \ \mathbf{y}^*]$

Checking AC feasibility

3

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n} & f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^c) \\ \text{s.t.} & \mathbf{h}(\mathbf{x}, \mathbf{y}^c) = \mathbf{0} \\ & \mathbf{g}(\mathbf{x}, \mathbf{y}^c) \leq \mathbf{0} \end{array}$$

⇓ Reformulate

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n} & f^{\text{CONVEX}}(\mathbf{x}) \\ \text{s.t.} & \mathbf{u}^{\text{CONVEX}}(\mathbf{x}) - \mathbf{v}^{\text{CONVEX}}(\mathbf{x}) \leq \mathbf{0} \end{array}$$

⇓ Convexify

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n} & f^{\text{CONVEX}}(\mathbf{x}) \\ \text{s.t.} & \mathbf{u}^{\text{CONVEX}}(\mathbf{x}) - \mathbf{v}^{\text{CONVEX}}_{\text{LINEARIZED}}(\mathbf{x}) \leq \mathbf{0} \end{array}$$



Checking AC feasibility 3

Power balance

$$\sum_{j \in \Lambda_n^G} p_j - \sum_{i \in \Lambda_n^D} D_i = \sum_{m \in \Lambda_n} \left[G_{nm}(e_m e_n + f_m f_n) - B_{nm}(e_m f_n - e_n f_m) - G_{nm}(e_n^2 + f_n^2) \right]$$

$$\sum_{j \in \Lambda_n^G} q_j - \sum_{i \in \Lambda_n^D} Q_i = \sum_{m \in \Lambda_n} \left[-G_{nm}(e_m f_n - e_n f_m) - B_{nm}(e_m e_n + f_m f_n) + (B_{nm} - b_{nm}^{\text{shunt}})(e_n^2 + f_n^2) \right]$$

⇓ Lift!

$$\sum_{j \in \Lambda_n^G} p_j - \sum_{i \in \Lambda_n^D} D_i = \sum_{m \in \Lambda_n} \left[G_{nm} c_{nm} - B_{nm} s_{nm} - G_{nm} c_{nn} \right]$$

$$\sum_{j \in \Lambda_n^G} q_j - \sum_{i \in \Lambda_n^D} Q_i = \sum_{m \in \Lambda_n} \left[-G_{nm} s_{nm} - B_{nm} c_{nm} + (B_{nm} - b_{nm}^{\text{shunt}}) c_{nn} \right]$$

$$c_{nn} = e_n^2 + f_n^2$$

$$c_{nm} = e_m e_n + f_m f_n$$

$$s_{nm} = e_m f_n - e_n f_m$$

Checking AC feasibility

3

$$c_{nn} = e_n^2 + f_n^2$$

$$c_{nm} = e_m e_n + f_m f_n$$

$$s_{nm} = e_m f_n - e_n f_m$$

$$\Downarrow xy = \frac{1}{4} [(x+y)^2 - (x-y)^2]$$

$$e_n^2 + f_n^2 \leq c_{nn}$$

$$c_{nn} \leq e_n^2 + f_n^2$$

$$(e_n + e_m)^2 + (f_n + f_m)^2 - 4c_{nm} \leq (e_n - e_m)^2 + (f_n - f_m)^2$$

$$(e_n - e_m)^2 + (f_n - f_m)^2 + 4c_{nm} \leq (e_n + e_m)^2 + (f_n + f_m)^2$$

$$(e_n - f_m)^2 + (e_m + f_n)^2 + 4s_{nm} \leq (e_n + f_m)^2 + (e_m - f_n)^2$$

$$(e_n + f_m)^2 + (e_m - f_n)^2 - 4s_{nm} \leq (e_n - f_m)^2 + (e_m + f_n)^2$$

Linearize!

1. Linear NCUC



2. Second order conic NCUC



3. Convex OPF



$$\min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{B}^m}$$

s.t.

$$f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y})$$

$$\mathbf{h}^{\text{SOC}}(\mathbf{x}, \mathbf{y}) = \mathbf{0}$$

$$\mathbf{g}_{\text{REDUCED}}^{\text{SOC}}(\mathbf{x}, \mathbf{y}) \leq \mathbf{0}$$

Too much?

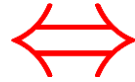
2

Benders

$$\begin{array}{ll} \min_{\mathbf{x}_1 \in \mathbb{R}^{n_1}, \mathbf{y} \in \mathbb{B}^m, \eta \in \mathbb{R}} & \mathbf{c}^\top [\mathbf{x}_1 \ \mathbf{y} \ \eta] \\ \text{s.t.} & \mathbf{A}[\mathbf{x}_1 \ \mathbf{y} \ \eta] = \mathbf{b} \end{array}$$

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{B}^m} & \\ \text{s.t.} & \end{array}$$

$$\begin{array}{l} f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}) \\ \mathbf{h}^{\text{SOC}}(\mathbf{x}, \mathbf{y}) = \mathbf{0} \\ \mathbf{g}_{\text{REDUCED}}^{\text{SOC}}(\mathbf{x}, \mathbf{y}) \leq \mathbf{0} \end{array}$$



Commitment
decisions, \mathbf{y}^k



Benders' cuts
Network information

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n} & f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^k) \\ \text{s.t.} & \mathbf{h}^{\text{SOC}}(\mathbf{x}, \mathbf{y}^k) = \mathbf{0} \\ & \mathbf{g}_{\text{REDUCED}}^{\text{SOC}}(\mathbf{x}, \mathbf{y}^k) \leq \mathbf{0} \end{array}$$

Benders

Trick: Include “physical” cuts in the master problem!

$$\begin{array}{ll} \min_{\mathbf{x}_1 \in \mathbb{B}^{n_1}, \mathbf{y} \in \mathbb{B}^m, \eta \in \mathbb{R}} & \mathbf{c}^\top [\mathbf{x}_1 \ \mathbf{y} \ \eta] \\ \text{s.t.} & \mathbf{A}[\mathbf{x}_1 \ \mathbf{y} \ \eta] = \mathbf{b} \end{array}$$

Commitment
decisions, \mathbf{y}^k



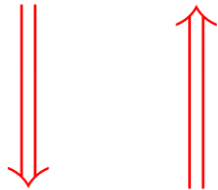
Benders' cuts
Network information

Benders' approach

$$\begin{array}{ll} \min_{\mathbf{x}_1 \in \mathbb{R}^{n_1}, \mathbf{y} \in \mathbb{B}^m, \eta \in \mathbb{R}} & \mathbf{c}^\top [\mathbf{x}_1 \ \mathbf{y} \ \eta] \\ \text{s.t.} & \mathbf{A}[\mathbf{x}_1 \ \mathbf{y} \ \eta] = \mathbf{b} \end{array}$$

→ Unit-related constraints

Commitment decisions, \mathbf{y}^k



Benders' cuts
Network information

Can we make the master network-aware?

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^n} & f^{\text{CONVEX}}(\mathbf{x}, \mathbf{y}^k) \\ \text{s.t.} & \mathbf{h}^{\text{SOC}}(\mathbf{x}, \mathbf{y}^k) = \mathbf{0} \\ & \mathbf{g}_{\text{REDUCED}}^{\text{SOC}}(\mathbf{x}, \mathbf{y}^k) \leq \mathbf{0} \end{array}$$

→ Network-related constraints

Adding “physical” cuts



- More expensive master problem
- Scalable and linear
- Does not cut feasible solutions



- Less iterations
- Better convergence



For each node



$$\sum_{j \in \Lambda_n^G} p_j(t) - \sum_{i \in \Lambda_n^D} D_i(t) = \sum_{m \in \Lambda_n} \left[G_{nm} c_{nm}(t) - B_{nm} s_{nm}(t) - G_{nm} c_{nn}(t) \right]$$

$$\sum_{j \in \Lambda_n^G} q_j(t) - \sum_{i \in \Lambda_n^D} Q_i(t) = \sum_{m \in \Lambda_n} \left[-G_{nm} s_{nm}(t) - B_{mn} c_{nm}(t) + (B_{nm} - b_{nm}^{\text{shunt}}) c_{nn}(t) \right]$$

$$\underline{V}_n^2 \leq c_{nn}(t) \leq \overline{V}_n^2$$

2000 × 4 × 24 = 192,000 constraints!

For each line



$$c_{nm}(t) = c_{mn}(t)$$

$$s_{nm}(t) = -s_{mn}(t)$$

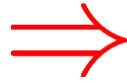
$$c_{nm}^2(t) + s_{nm}^2(t) \leq c_{nn}(t) c_{mm}(t)$$

$$p_{ij}^2(t) + q_{ij}^2(t) \leq \overline{S}_{ij}(t)$$

3206 × 4 × 24 = 307,776 constraints!

Copper plate relaxation

4 × 24 = 96 linear constraints!



$$\sum_{j \in \mathcal{G}} q_j(t) - \sum_{i \in \mathcal{D}} Q_i(t) \geq \sum_{n \in \mathcal{N}} B_{nn} c_{nn}(t) - \sum_{(nm) \in \mathcal{L}} b_{nm}^{\text{shunt}} (c_{nn}(t) + c_{mm}(t))$$

$$\underline{V}_n^2 \leq c_{nn}(t) \leq \bar{V}_n^2$$

$\min_{\mathbf{x}_2 \in \mathbb{R}^{m_2}, \mathbf{y} \in \mathbb{B}, \eta \in \mathbb{R}}$	$\mathbf{c}^\top [\mathbf{x}_2 \ \mathbf{y} \ \eta]$
s.t.	$\mathbf{A}[\mathbf{x}_2 \ \mathbf{y} \ \eta] = \mathbf{b}$
	$\mathbf{G}\mathbf{x}_2 \leq \mathbf{0}$

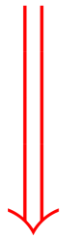


Unit-related constraints



Network-aware constraints



Commitment decisions, \mathbf{y}^k



Benders' cuts
Network information

C. Coffrin, H. Hijazi, and P. van Hentenryck, "Network Flow and Copper Plate Relaxations for AC Transmission Systems," <https://arxiv.org/abs/1506.05202>

Summary

1. **Good initial solution:** Mixed-integer linear NCUC
2. **Commitment solution:** Mixed-integer second order conic NCUC 
3. **AC feasibility:** Continuous convex OPF 

Case Studies

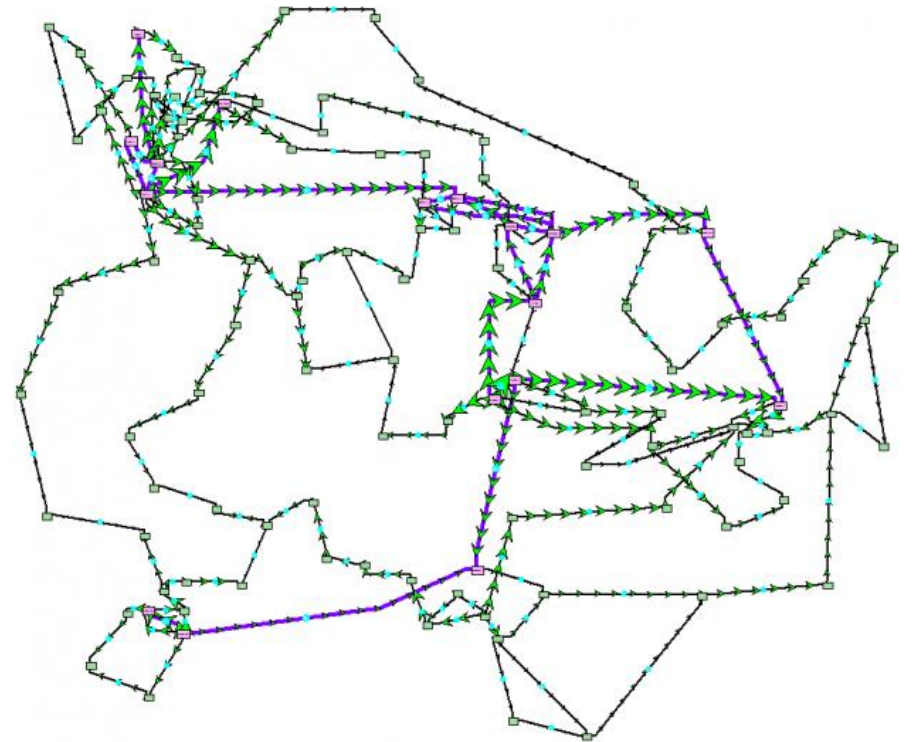
Illinois 200-bus 49-generator 245-line system



<https://electricgrids.engr.tamu.edu/electric-grid-test-cases/activsg200/>

Data

- Entirely synthetic
 - Based on real data
 - Preserves statistical properties
- 49 generating units
 - 1 nuclear (15% installed capacity)
 - 6 wind (15% installed capacity)
 - 42 thermal (70% installed capacity)
 - **33 units with commitment (binary) variables**
- 245 transmission lines



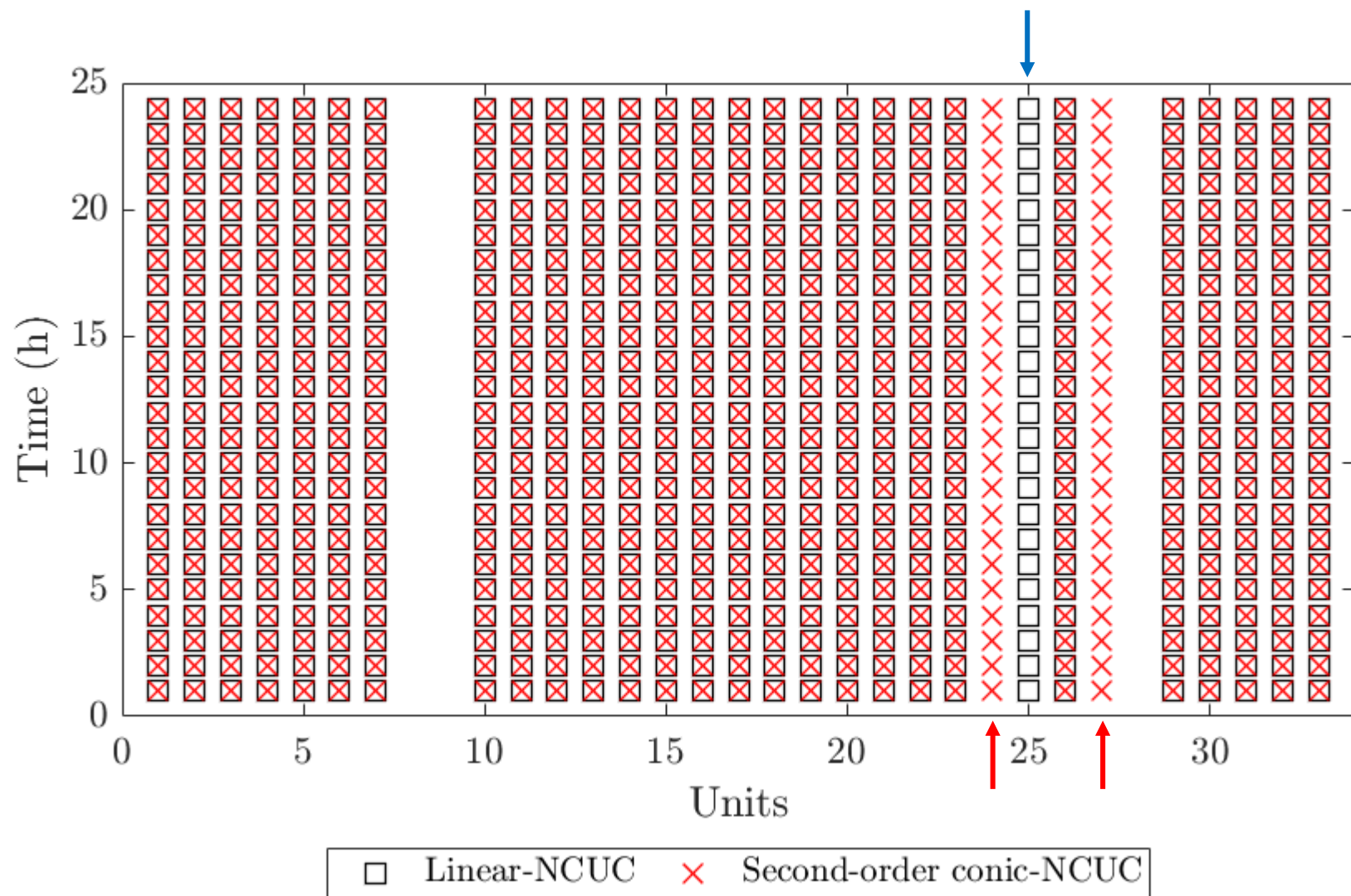
Results

Model	Solver	Loading level	# Start ups ¹	Generation (MW)	Objective (\$)	Time* (s)
Mixed-integer linear	CPLEX	Low	0	29344.51	482444.82	15.38
		Medium	0	39126.02	484188.40	25.39
		High	6	48907.52	509953.48	23.65
Mixed-integer second-order conic	CPLEX	Low	0	29889.96	482472.70	236.14
		Medium	0	39428.27	484247.05	313.66
		High	7	49396.99	512111.75	977.09
Continuous second-order conic	Mosek	Low	-	29528.43	482473.35	407.83
		Medium	-	39432.99	484248.75	434.29
		High	-	49360.09	512142.89	750.66

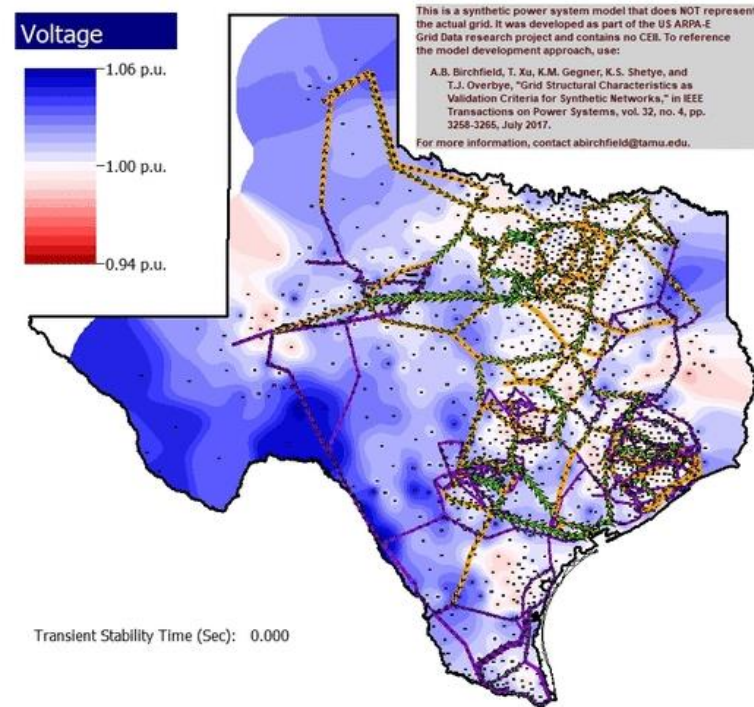
¹Number of units that start up/shutdown in the planning horizon.

*Windows-based laptop with a processor Intel Core i7 2.60 GHz and 16 GB of RAM.

Results



Texas 2000-bus 544-unit 3206-line system



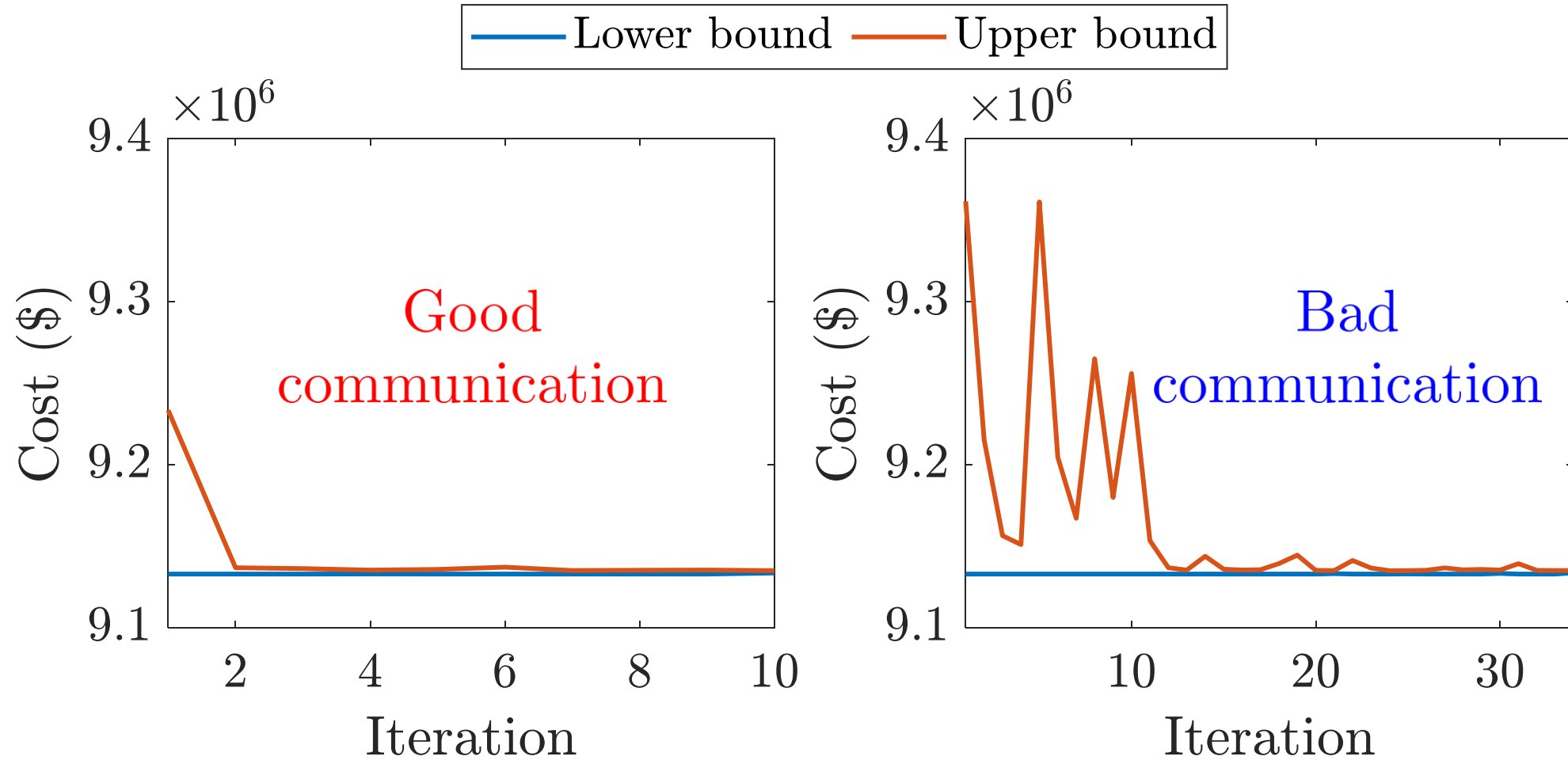
<https://my.syncplicity.com/share/wubzq34byp7h2g4/ACTIVSg2000>

Data

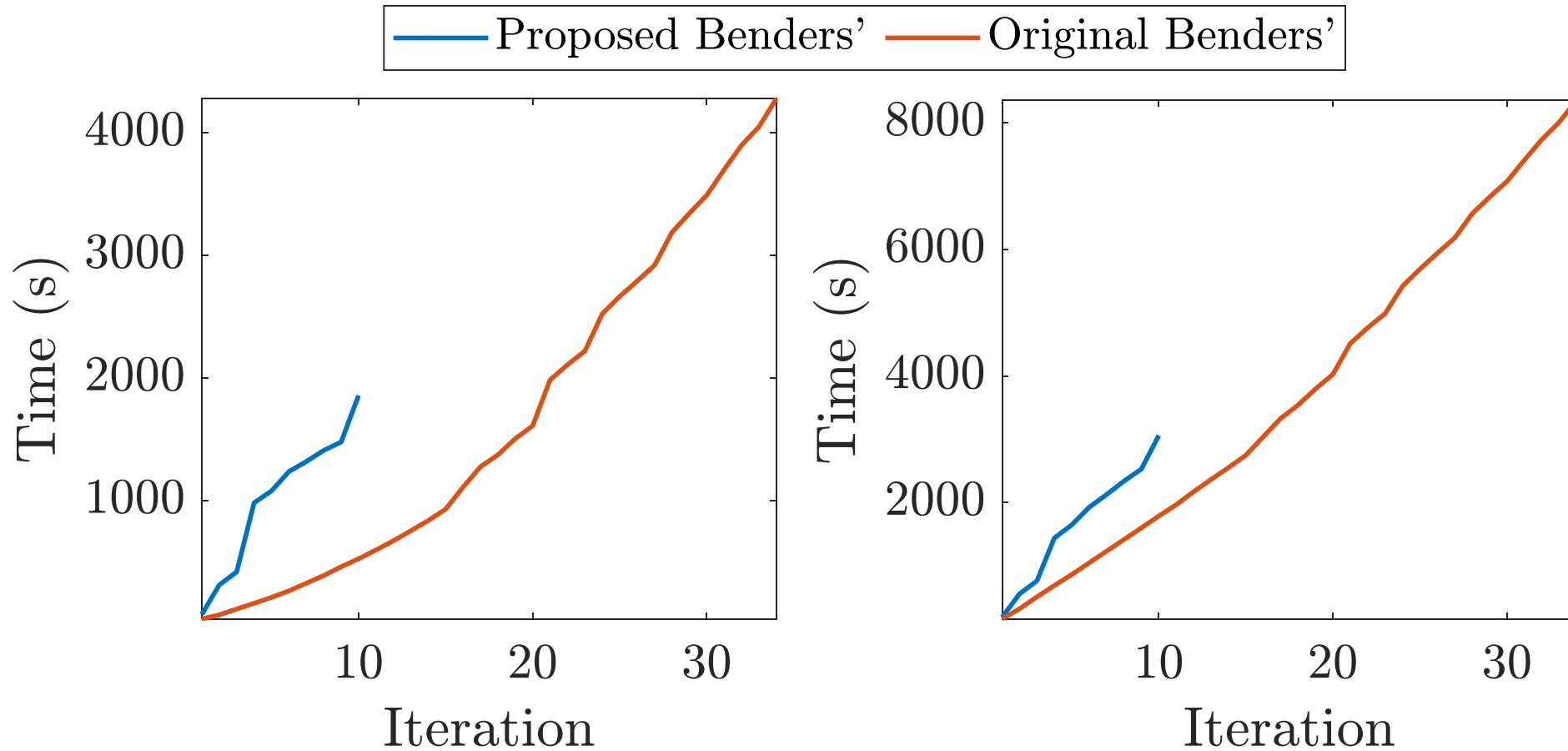
- Entirely synthetic
- 544 generating units
 - 39 coal units - 15.06% of total installed capacity
 - 367 gas units - 66.27% of total installed capacity
 - 25 run-of-the-river hydro units - 2.70% of total installed capacity
 - 87 wind units - 9.96% of total installed capacity
 - 22 solar units - 0.68% of total installed capacity
 - 4 nuclear units - 5.34% of total installed capacity
- 3206 transmission lines

⇒ 369 units with
commitment variables!

Benders' bounds

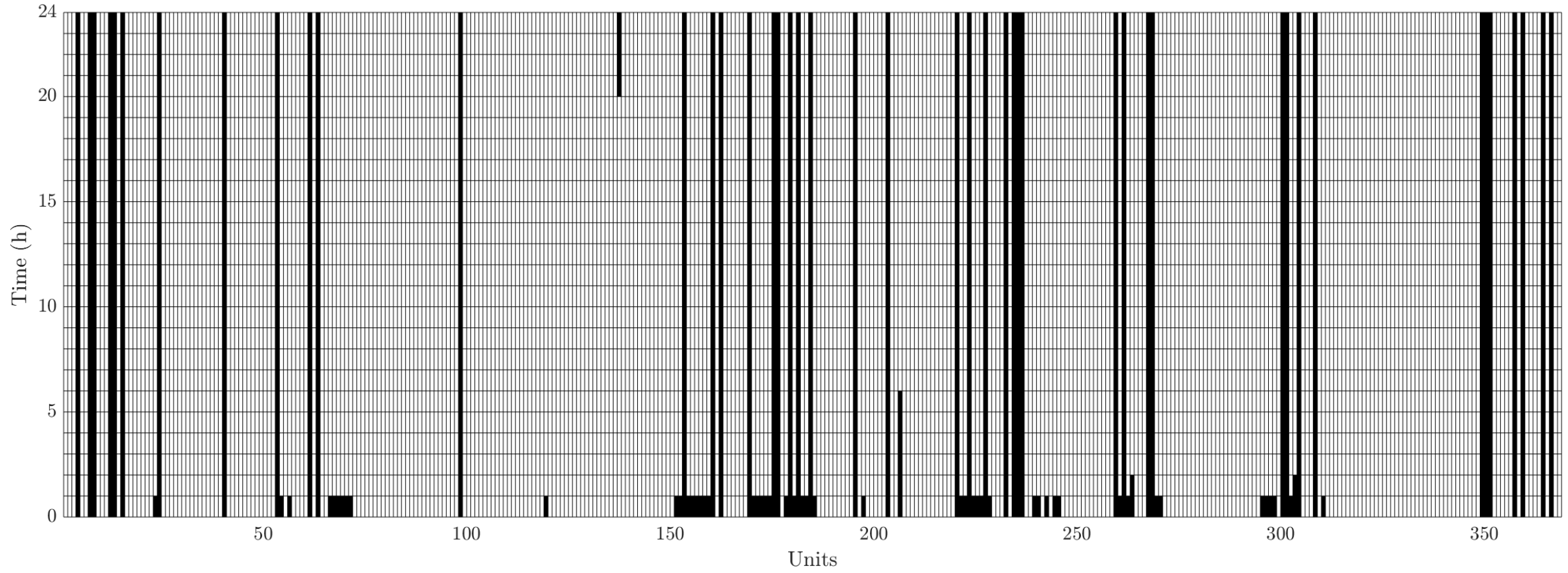


Computing time

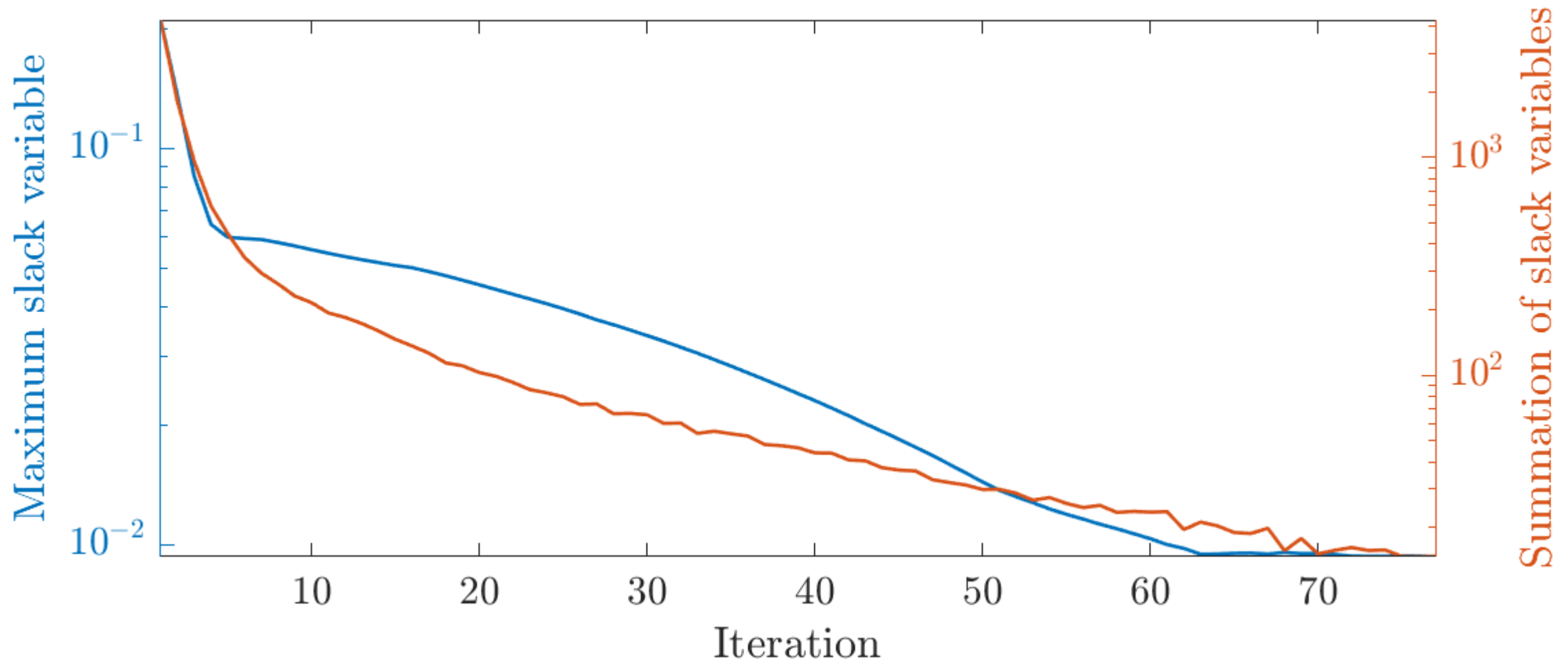


> 60%
improvement!

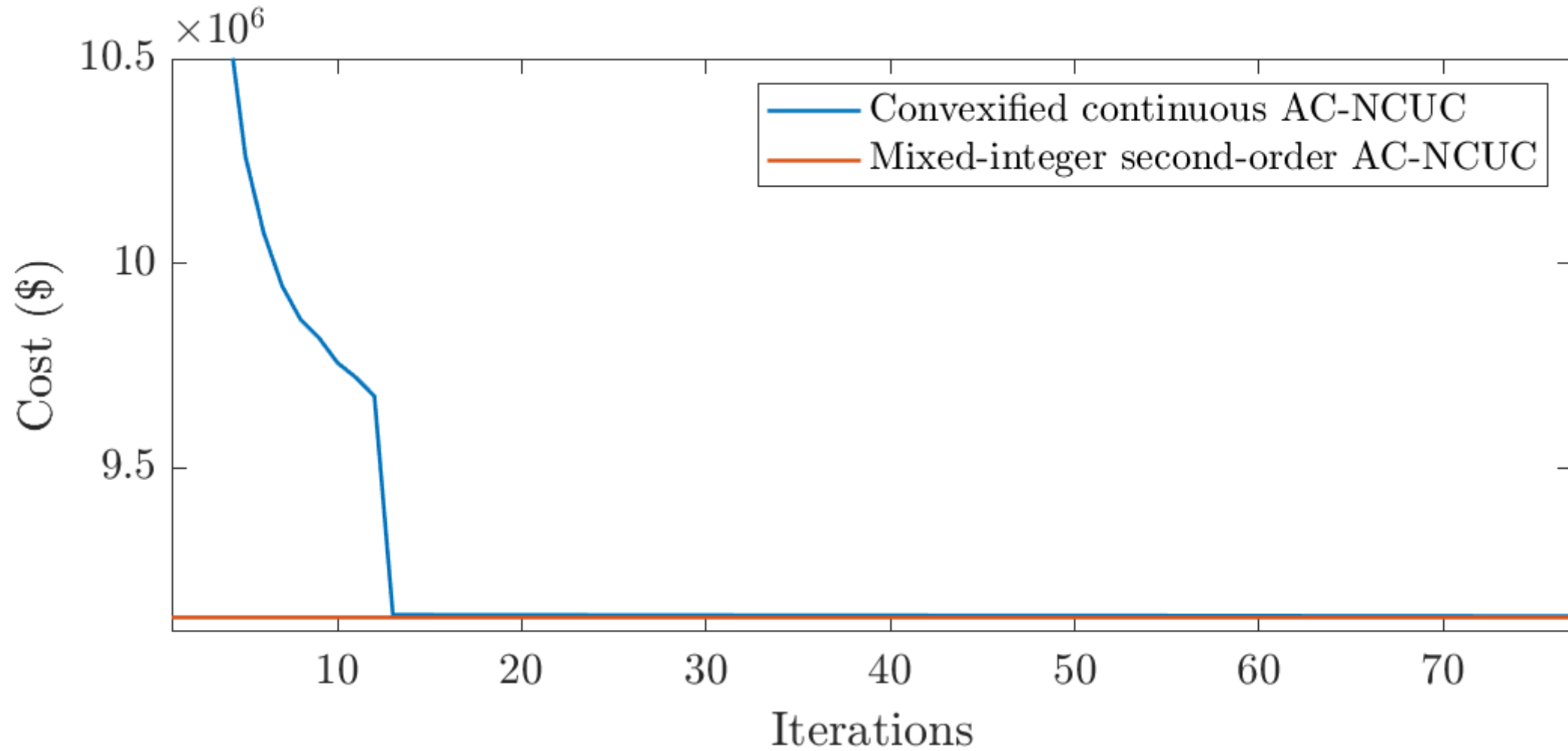
Commitment



Feasibility step



Feasibility step



Thank you!

