

# Decentralized Robust Optimization Algorithms for Tie-Line Scheduling of Multi-Area Grid with Variable Wind Energy

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Joint work with

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# Outline

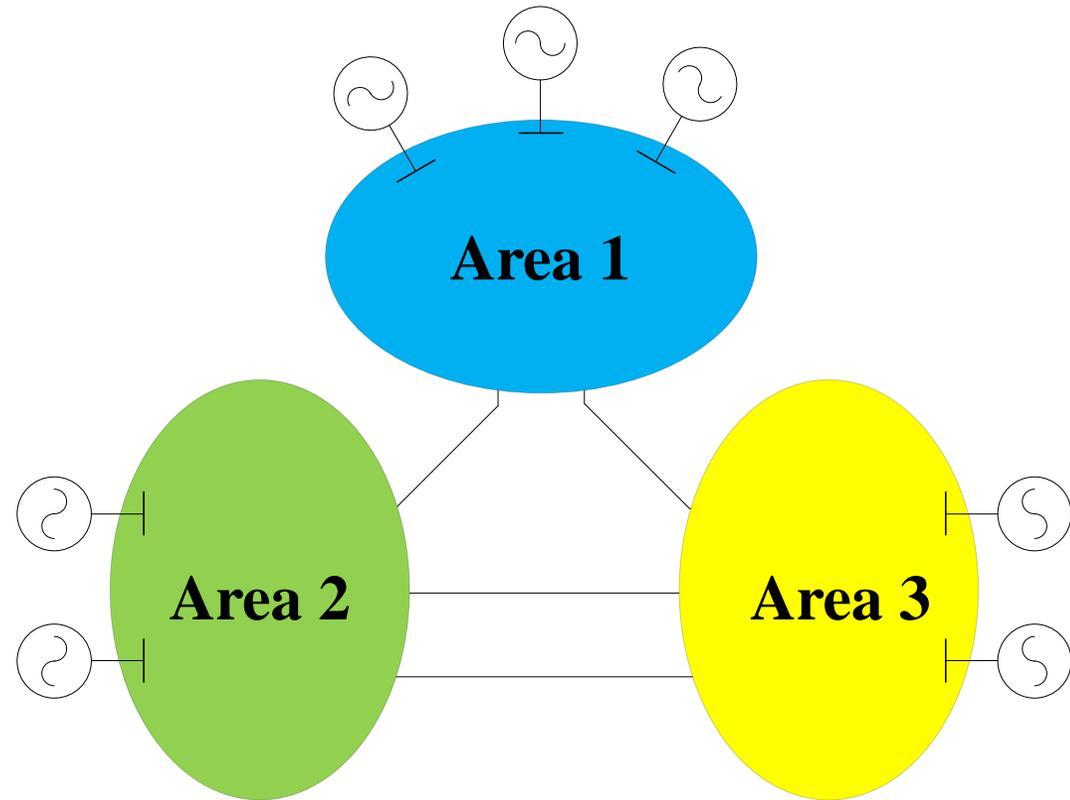
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- Background and Motivation
- Robust Optimization Formulations and Properties
- Distributed Computing Methods through Alternating Direction Method of Multipliers (ADMM)
- Numerical Study and Demonstrations
- Conclusions

# Part I: Background and Motivation

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- Multi-area power system
  - Physically separated regions (China)
  - Different electricity market
- Intermittent renewable energy
- Random contingencies
- Aligned LMPs for boundary nodes



# Challenges in Multi-Area Power Grids

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- Systems are cooperative but operated independently
  - Privacy
  - Commercial information
- Protocols between systems to support cooperation:
  - Coordinated Transaction Scheduling: NYISO and PJM
  - Interchange Optimization: PJM and MISO
  - Inter-Regional Interchange Scheduling: ISONE and NYISO
- Interfaces between multiple systems: **tie-lines**
  - Power flow decisions (day-ahead and intraday)
  - Unit commitment decisions (day-ahead)

# Decision Models for Multi-Area Power Grids

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- Lagrangian relaxation
  - Conejo and Aguado (1998), Multi-area coordinated decentralized DC optimal power flow
  - Aguado, Quintana, and Conejo (1999), Optimal power flows of interconnected power systems
- Augmented Lagrangian decomposition
  - Kim and Baldick (1997)
  - Ahmadi-Khatir, Conejo, and Cherkaoui (2014)
- Alternative Direction Method of Multipliers (ADMM)
  - Kim and R. Baldick (2000)
  - Zheng, Wu, Zhang, Sun, and Liu (2015)

# Uncertainty Consideration in Power Grids

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- Scenario based probabilistic models
  - Stochastic programming (SP) and chance constrained formulation
  - Explicit large-scale models but many existing computing methods: e.g., Ahmadi-Khatir, Conejo, and Cherkaoui (2013, 2014):
    - *SP+ augmented Lagrangian relaxation*
  - Issue: Inaccurate prediction of scenarios and probabilities -> infeasible solutions
- Robust optimization models
  - Uncertainty set based compact formulation
  - Produce highly feasible solutions by considering all possibilities inside uncertainty set
  - Two types of algorithms: Benders decomposition and *column-and-constraint generation*
  - **Our aim: integration of decentralized computing scheme + robust optimization**

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# Part II: Robust Optimization Formulations and Properties

# Multi-Area Robust Tie-Line Scheduling

- Two-stage decision making framework

$$\min_{\xi^f \in \Omega^f} \sum_{a \in \mathbf{A}} \left\{ \max_{\tilde{\mathbf{P}}_a^w \in \mathbf{U}_a^w} \left[ \min_{\xi_a^s \in \Omega_a^s(\xi_a^f, \tilde{\mathbf{P}}_a^w)} C_a^{ED}(\xi_a^s, \tilde{\mathbf{P}}_a^w) \right] \right\}$$

- Tie-line interchanges: the first-stage decisions before the availability of wind energy is known
  - Phase angles at boundary buses:

$$\xi^f = \{ \xi_a^f, \forall a \in \mathbf{A} \}, \xi_a^f = \{ \delta_{a,i,t}, \forall g \in \mathbf{G}_a, i \in \mathbf{N}_a^{BB} \cup \mathbf{N}_a'^{BB}, t \in \mathbf{T} \}$$

$$\Omega^f = \left\{ \xi^f \mid \delta_{\phi(i),i,t} = \delta_{\phi(j),i,t}, \delta_{\phi(i),j,t} = \delta_{\phi(j),j,t}, \left| \delta_{\phi(i),i,t} - \delta_{\phi(i),j,t} \right| / X_{i,j} \leq \bar{F}_{i,j}, \forall (i,j) \in \mathbf{E}^{tie}, i > j, t \in \mathbf{T}, \right.$$

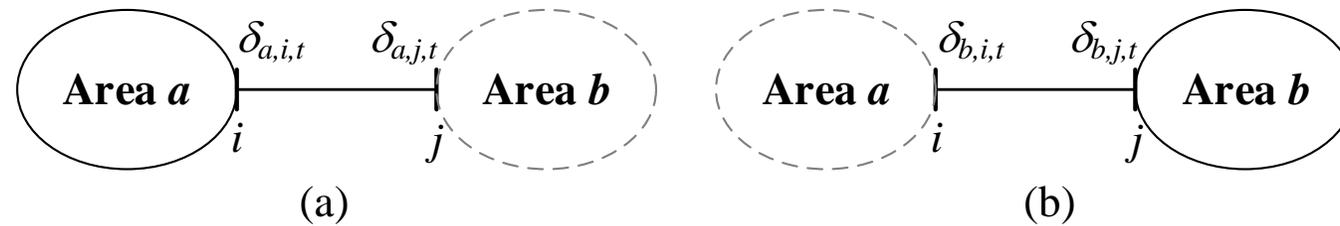
$$\left. \delta_{a',ref,t} = 0, \forall a' \in \Phi(ref), t \in \mathbf{T} \right\}$$

- Economic dispatch: the second-stage decision after wind energy is revealed.
  - Continuous model

# Inter-regional constraints

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- **Region-coupling constraints** - perceived phase angles are same at two ends of a tie-line



$$\delta_{a,i,t} = \delta_{b,i,t}, \delta_{a,j,t} = \delta_{b,j,t}, \forall t \in \mathbf{T}$$

# Economic Dispatch of Each Area

- Dispatch of conventional generation units, wind farms, and phase angles of internal buses

$$\Omega_a^s(\xi_a^f, \tilde{\mathbf{P}}_a^w) = \left\{ \xi_a^s \mid \right.$$

$$\sum_{j \in \Psi_a^{IB}(i)} (\theta_{i,t} - \theta_{j,t}) / X_{i,j} + \sum_{j \in \Psi_a^{BB}(i)} (\theta_{i,t} - \delta_{a,j,t}) / X_{i,j}$$

$$= \sum_{g \in \Psi_a^G(i)} p_{g,t}^G + \sum_{k \in \Psi_a^W(i)} p_{k,t}^w - \sum_{d \in \Psi_a^D(i)} p_{d,t}^D, \forall i \in \mathbf{N}_a^{IB}, t,$$

$$\sum_{j \in \Psi_a^{IB}(i)} (\delta_{a,i,t} - \theta_{j,t}) / X_{i,j} + \sum_{j \in \Psi_a^{BB}(i)} (\delta_{a,i,t} - \delta_{a,j,t}) / X_{i,j}$$

$$= \sum_{g \in \Psi_a^G(i)} p_{g,t}^G + \sum_{k \in \Psi_a^W(i)} p_{k,t}^w - \sum_{d \in \Psi_a^D(i)} p_{d,t}^D, \forall i \in \mathbf{N}_a^{BB}, t,$$

$$-\bar{F}_{i,j} \leq (\theta_{i,t} - \theta_{j,t}) / X_{i,j} \leq \bar{F}_{i,j}, \forall i \in \mathbf{N}_a^{IB}, j \in \Psi_a^{IB}(i), j > i,$$

$$-\bar{F}_{i,j} \leq (\delta_{a,i,t} - \theta_{j,t}) / X_{i,j} \leq \bar{F}_{i,j}, \forall i \in \mathbf{N}_a^{BB}, j \in \Psi_a^{IB}(i),$$

$$p_{g,t}^G + r_{g,t}^{G+} \leq P_g^G, \forall g \in \mathbf{G}_a, t \in \mathbf{T},$$

$$r_{g,t}^{G-} \geq p_{g,t}^G - \underline{P}_g, \forall g \in \mathbf{G}_a, t \in \mathbf{T},$$

$$\sum_{g \in \mathbf{G}_a} r_{g,t}^{G+} \geq SR_{a,t}^+, \sum_{g \in \mathbf{G}_a} r_{g,t}^{G-} \geq SR_{a,t}^-, \forall t \in \mathbf{T},$$

$$0 \leq r_{g,t}^{G+} \leq RU_g^G, 0 \leq r_{g,t}^{G-} \leq RD_g^G, \forall g \in \mathbf{G}_a, t \in \mathbf{T},$$

$$0 \leq p_{k,t}^w \leq \tilde{P}_{k,t}^w, \forall k \in \mathbf{W}_a, t \in \mathbf{T} \}$$

# Observation:

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- If the uncertainty sets  $U_a^w$  are polyhedra, the robust multi-area tie-line schedule problem is a convex optimization (an extremely large-scale linear program)
- Idea: enumerating all extreme points of  $U_a^w$  to construct the equivalent formulation, which is an LP

# Multi-Area Generation Unit and Tie-line Scheduling

- Two-stage decision making framework

$$\min_{\xi^f \in \Omega^f} \sum_{a \in \mathbf{A}} \left\{ C_a^{UD}(\xi_a^f) + \max_{\tilde{\mathbf{P}}_a^w \in \mathbf{U}_a^w} \left[ \min_{\xi_a^s \in \Omega_a^s(\xi_a^f, \tilde{\mathbf{P}}_a^w)} C_a^{ED}(\xi_a^s, \tilde{\mathbf{P}}_a^w) \right] \right\}$$

- First-stage decisions: unit commitments and tie-line interchanges

$$\xi^f = \{ \xi_a^f, \forall a \in \mathbf{A} \}, \xi_a^f = \{ u_{g,t}^G, x_{g,t}^G, y_{g,t}^G, \delta_{a,i,t}, \forall g \in \mathbf{G}_a, i \in \mathbf{N}_a^{BB} \cup \mathbf{N}_a'^{BB}, t \in \mathbf{T} \}$$

$$\Omega^f = \left\{ \xi^f \mid \forall a \in \mathbf{A}, g \in \mathbf{G}_a, t \in \mathbf{T}, \delta_{\phi(i),i,t} = \delta_{\phi(j),i,t}, \delta_{\phi(i),j,t} = \delta_{\phi(j),j,t}, \left| \delta_{\phi(i),i,t} - \delta_{\phi(i),j,t} \right| / X_{i,j} \leq \bar{F}_{i,j}, \forall (i,j) \in \mathbf{E}^{tie}, i > j, \right.$$

$$\delta_{a',ref,t} = 0, \forall a' \in \Phi(ref), t \in \mathbf{T},$$

$$u_{g,t}^G - u_{g,t-1}^G = x_{g,t}^G - y_{g,t}^G, \sum_{\tau=\max\{1,t-MU_g^G+1\}}^t x_{g,\tau}^G \leq u_{g,t}^G, \sum_{\tau=\max\{1,t-MD_g^G+1\}}^t y_{g,\tau}^G \leq 1 - u_{g,t}^G, \forall g, t$$

$$u_{g,t}^G \in \{0,1\}, 0 \leq x_{g,t}^G \leq 1, 0 \leq y_{g,t}^G \leq 1\}$$

# Multi-Area Generation Unit and Tie-line Scheduling (Cont'd)

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- Second-stage decisions: economic dispatch after the available wind power is revealed and unit status are determined
- Observation:
  - Due to binary variables for unit scheduling, the robust formulation is equivalent to a non-convex and discrete mixed integer program
- **Challenge:** augmented Lagrangian methods typically do not converge

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# Part III: Distributed Computing Methods through ADMM

# Augmented Lagrangian Relaxation

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□ Relaxing  $\delta_{a,i,t} = \delta_{b,i,t}, \delta_{a,j,t} = \delta_{b,j,t}, \forall t \in \mathbf{T}$

□ Averaging  $\bar{\delta}_{i,t} = \sum_{a \in \Phi(i)} \delta_{a,i,t} / |\Phi(i)|$

□ Augmented model

$$\begin{aligned}
 \min_{\xi_a^f} L_a(\xi_a^f, \lambda_a, \bar{\delta}) &= \sum_{i \in \mathbf{N}_a^{BB} \cup \mathbf{N}_a^{iBB}} \sum_{t \in \mathbf{T}} \left[ \lambda_{a,i,t} (\delta_{a,i,t} - \bar{\delta}_{i,t}) + \frac{\rho}{2} (\delta_{a,i,t} - \bar{\delta}_{i,t})^2 \right] \\
 &+ \max_{\tilde{\mathbf{P}}_a^w \in \mathbf{U}_a^w} \left[ \min_{\xi_a^s \in \Omega_a^s(\xi_a^f, \tilde{\mathbf{P}}_a^w)} C_a^{ED}(\xi_a^s, \tilde{\mathbf{P}}_a^w) \right] \\
 &= C'_a(\xi_a^f, \lambda_a, \bar{\delta}) + \max_{\tilde{\mathbf{P}}_a^w \in \mathbf{U}_a^w} \left[ \min_{\xi_a^s \in \Omega_a^s(\xi_a^f, \tilde{\mathbf{P}}_a^w)} C_a^{ED}(\xi_a^s, \tilde{\mathbf{P}}_a^w) \right]
 \end{aligned}$$

# Overall Algorithm Scheme

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- Using ADMM, each area can be computed independently
  - distributed computing and privacy protection
- For a single area problem: two-stage robust optimization
  - column and constraint generation method
  - finitely convergent for a polyhedron uncertainty set
- **Integrated ADMM+CCG (IAC) Solution Method**
  - Multi-area robust tie-line scheduling: **ADMM+CCG converges to optimal value**
  - Multi-area robust generation unit and tie-line scheduling: **convergence is NOT guaranteed**
  - **Computational enhancements?** Speed and convergence

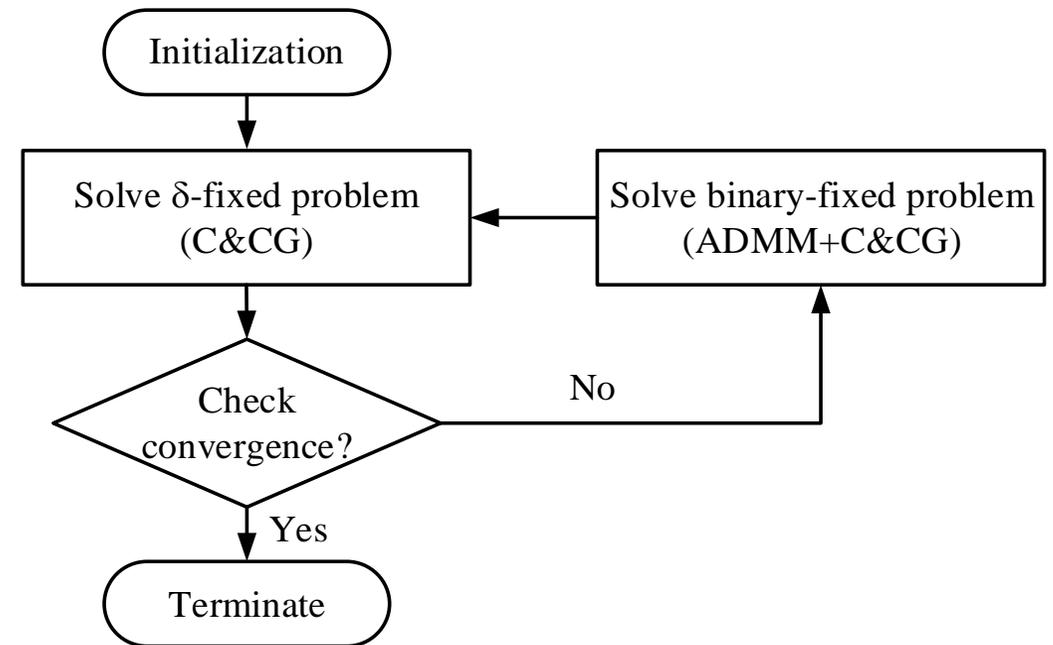
# Fast Computing

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- *Warm Start (WS)*:
  - select initial values of the first-stage variables and dual variables using the deterministic version
- *Scenario Retaining (SR)*:
  - CCG is repeatedly called within ADMM framework
  - Keep and re-use existing scenarios generated in previous CCG calls
- *Scenario Discard (SD)*:
  - Remove non-critical scenarios to maintain a small pool
  - Dynamically manage a scenario pool through a changing threshold
- SR and SD are key steps in distributed computation of Robust Optimization

# Convergence Issue from UC

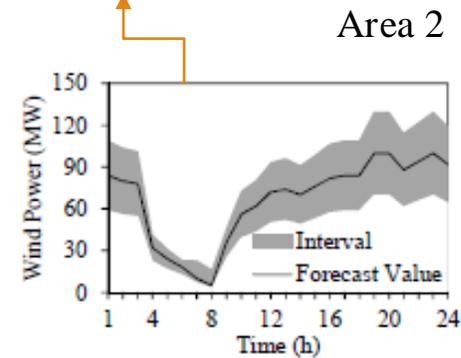
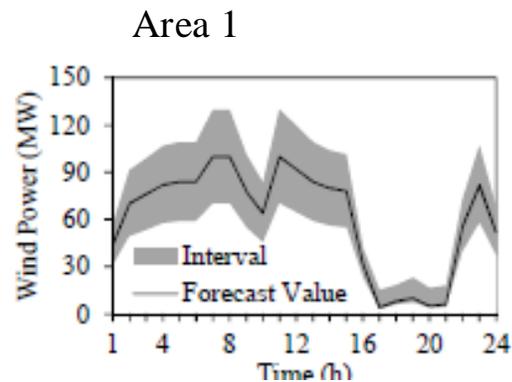
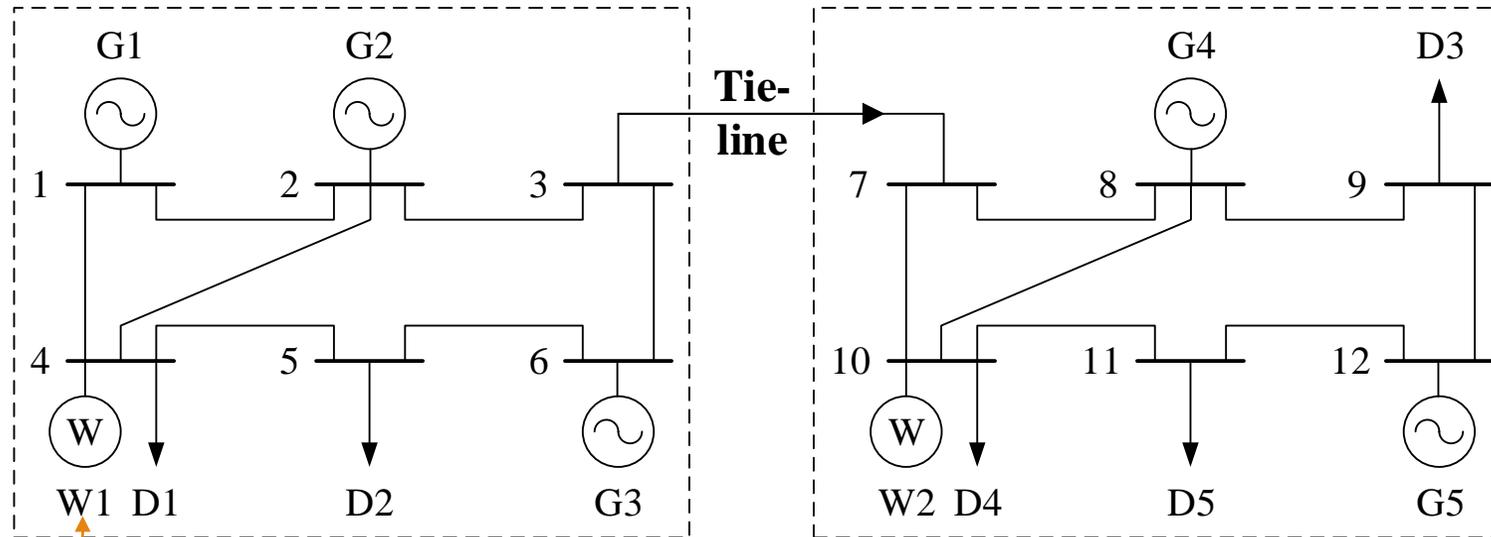
- Non-convergence due to non-convex structure from binary commitment decisions
- Alternating optimization procedure to ensure convergence (**heuristically**)
  - Alternatively computing with boundary phase angles or commitment status are fixed
  - A repeated commitment status indicates termination
  - **Finitely converged**



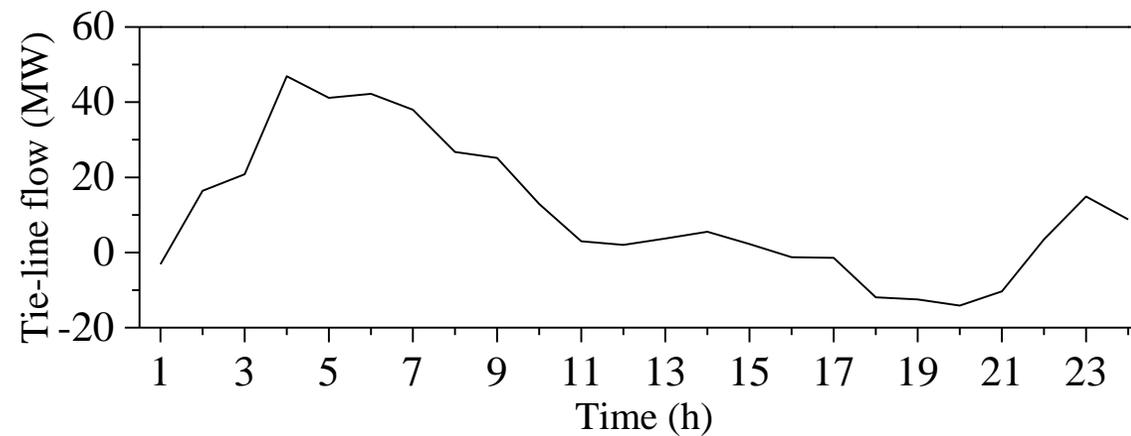
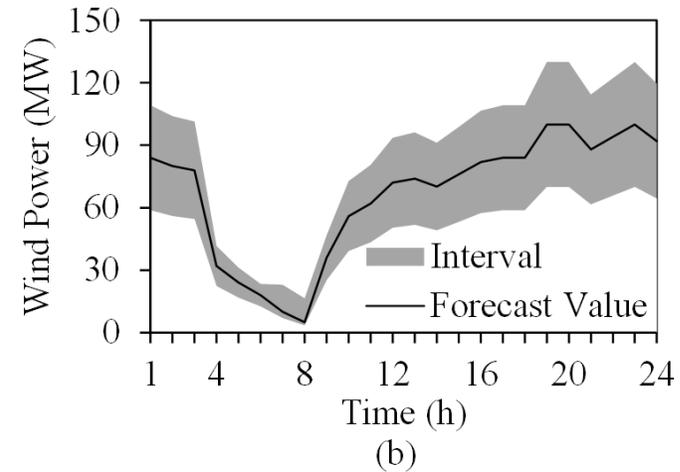
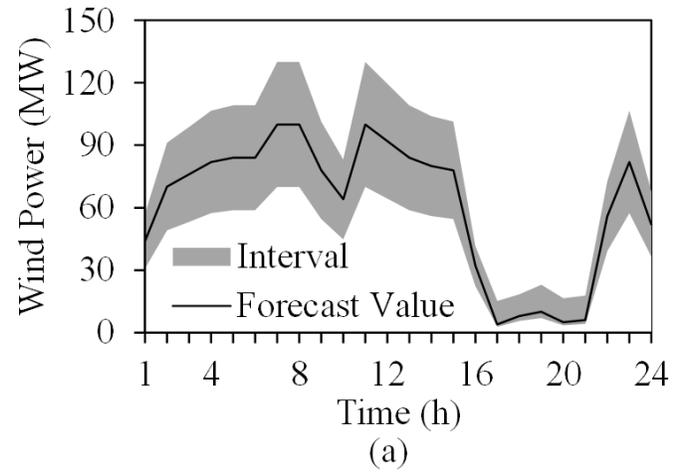
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# Part IV: Numerical Study and Demonstrations

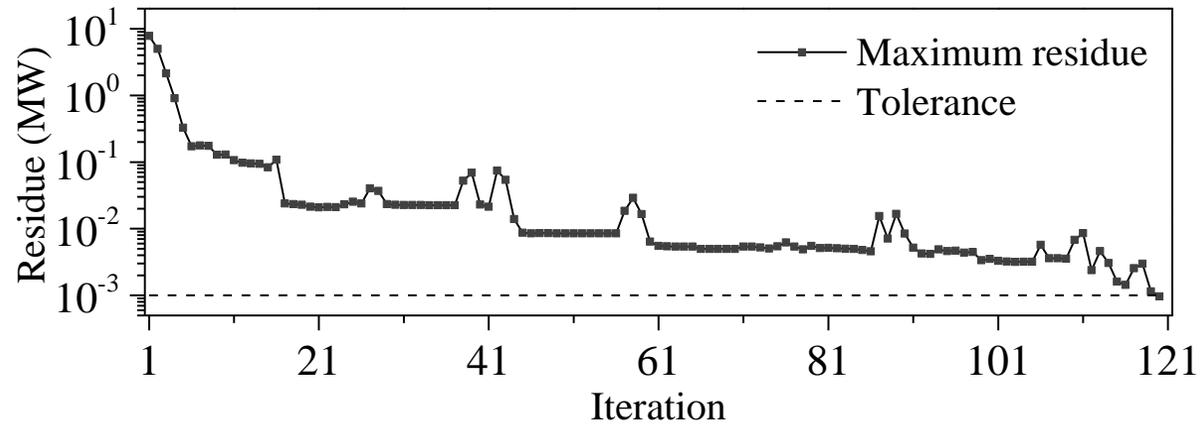
# Two-Area 12-Bus Interconnected System



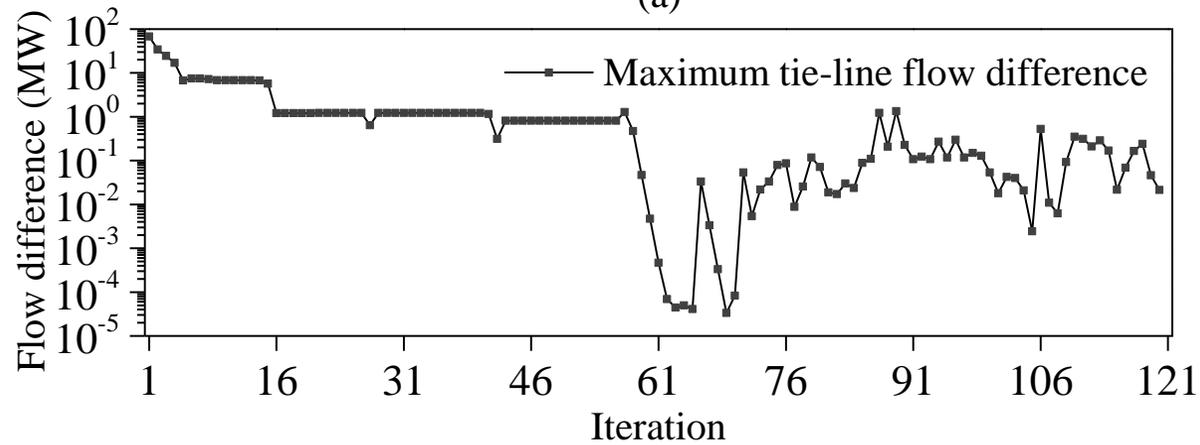
# Tie-line Flows



# IAC Performance for Tie-line Scheduling



(a)



(b)

# Computational Enhancement Strategies

Case	Strategy			Initialization		ADMM					Total time (s)	
	WS	SR	SD	# iter.	Time (s)	# iter.	C&CG					
							# solution		Time (s)			
							MP	SP	MP	SP		SD
M0	-	-	-	-	-	120	516	469	20.1	33.3	-	53.4
M1	-	✓	-	-	-	120	151	152	223.2	11.0	-	234.2
M2	-	✓	✓	-	-	120	179	179	16.6	13.7	5.8	36.1
M3	✓	-	-	21	5.1	41	82	82	0.9	4.3	-	10.3
M4	✓	✓	✓	21	5.2	41	41	41	0.5	1.5	0.5	7.7

\*M0: IAC without enhancement

# Coordination Effect with Unit Commitment

slightly higher than the centralized solution by 0.11%

SCHEDULING RESULTS ON 12-BUS SYSTEM

Hour	Proposed IAC					Centralized C&CG					Non-coordinated							
	Area 1			Area 2		Tie (MW)	Area 1			Area 2		Tie (MW)	Area 1			Area 2		Tie (MW)
	G1	G2	G3	G4	G5		G1	G2	G3	G4	G5		G1	G2	G3	G4	G5	
1	1	0	0	1	0	-10.4	1	0	0	1	0	-7.9	1	0	0	1	0	0.0
2	1	0	0	1	0	-1.9	1	0	0	1	0	16.1	1	0	0	1	0	0.0
3	1	0	0	1	0	5.4	1	0	0	1	0	18.1	1	0	0	1	0	0.0
4	1	0	0	1	0	11.8	1	0	0	1	0	24.4	1	0	0	1	0	0.0
5	1	0	0	1	0	11.6	1	0	0	1	0	25.3	1	0	0	1	0	0.0
6	1	0	0	1	0	12.6	1	0	0	1	0	21.4	1	0	0	1	0	0.0
7	1	0	0	1	0	25.9	1	0	0	1	0	25.9	1	0	0	1	1	0.0
8	1	0	0	1	0	16.7	1	0	0	1	0	16.5	1	0	0	1	1	0.0
9	1	0	0	1	0	-0.3	1	0	0	1	0	19.9	1	0	0	1	1	0.0
10	1	0	0	1	0	-3.7	1	0	0	1	0	12.7	1	0	0	1	1	0.0
11	1	0	1	1	1	3.0	1	0	1	1	0	11.0	1	0	1	1	1	0.0
12	1	0	1	1	1	2.0	1	0	1	1	0	7.6	1	0	1	1	1	0.0
13	1	0	1	1	1	2.7	1	0	1	1	0	6.7	1	0	0	1	1	0.0
14	1	0	1	1	1	3.4	1	0	1	1	0	7.6	1	0	0	1	1	0.0
15	1	0	1	1	1	1.7	1	0	1	1	0	6.1	1	0	1	1	1	0.0
16	1	0	0	1	0	-8.4	1	0	0	1	0	-0.6	1	0	0	1	0	0.0
17	1	0	0	1	0	-11.4	1	0	0	1	0	-8.4	1	0	0	1	0	0.0
18	1	0	0	1	0	-2.7	1	0	0	1	0	-2.0	1	0	0	1	0	0.0
19	1	0	0	1	0	-5.2	1	0	0	1	0	-9.9	1	0	0	1	0	0.0
20	1	0	0	1	0	-2.5	1	0	0	1	0	-34.3	1	0	0	1	0	0.0
21	1	0	0	1	0	-6.8	1	0	0	1	0	-9.0	1	0	0	1	0	0.0
22	1	0	0	1	0	9.3	1	0	0	1	0	4.6	1	0	0	1	0	0.0
23	1	0	0	1	0	7.1	1	0	0	1	0	0.4	1	0	0	1	0	0.0
24	1	0	0	1	0	-2.0	1	0	0	1	0	4.4	1	0	0	1	0	0.0
Obj.	\$150,419						\$150,243						\$153,645					

# Performance in Large Systems

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System	Areas	Units	Buses	Int. Lines	Tie-lines
2A-RTS	2	66	48	76	3
3A-RTS	3	99	73	115	5
118-Bus	3	79	118	174	12

# IAC for Tie-line Scheduling

SIMULATION RESULTS ON LARGE-SCALE TEST SYSTEMS

Uncertainty Budget			$\Gamma = 0$	$\Gamma = 6$	$\Gamma = 12$	$\Gamma = 24$
2A-RTS	Centr.	Obj. (\$)	5,982,249	6,104,215	6,182,390	6,203,326
		Time (s)	2.4	15.2	24.0	9.8
	IAC	Obj. (\$)	5,982,497	6,118,620	6,198,730	6,226,089
		# iter.	101	103	103	107
		Time (s)	17.9	92.1	79.4	98.7
		Error (%)	0.00	0.24	0.26	0.37
3A-RTS	Centr.	Obj. (\$)	9,151,653	9,333,234	9,459,759	9,510,397
		Time (s)	3.6	33.6	106.7	9.9
	IAC	Obj. (\$)	9,151,694	9,348,872	9,468,978	9,512,639
		# iter.	226	320	316	339
		Time (s)	34.4	830.4	501.9	676.5
		Error (%)	0.00	0.17	0.10	0.02
118-Bus	Centr.	Obj. (\$)	2,249,048	2,250,762	2,252,142	2,255,044
		Time (s)	4.4	4.6	4.7	4.2
	IAC	Obj. (\$)	2,249,120	2,251,392	2,253,339	2,257,712
		# iter.	491	535	573	600
		Time (s)	197.1	285.1	325.2	348.3
		Error (%)	0.00	0.03	0.05	0.12

# IAC for Generation Unit and Tie-line Scheduling

SIMULATION RESULTS ON LARGE-SCALE TEST SYSTEMS

Uncertainty Budget			$\Gamma = 0$	$\Gamma = 6$	$\Gamma = 12$	$\Gamma = 24$
2A-RTS	Centr.	Obj. (\$)	1,014,164	1,018,237	1,039,095	1,044,114
		Time (s)	597	3,005	1,617	1,451
	IAC	Obj. (\$)	1,019,117	1,034,468	1,055,462	1,061,706
Time (s)		127	3,642	4,186	788	
Gap (%)		0.49	1.59	1.58	1.66	
3A-RTS	Centr.	Obj. (\$)	1,529,977	1,535,347	1,565,097	1,576,184
		Time (s)	25	4,196	5,876	4,821
	IAC	Obj. (\$)	1,541,769	1,562,666	1,583,698	1,591,784
Time (s)		125	2,360	1,757	3,486	
Gap (%)		0.77	1.78	1.19	0.99	
118-Bus	Centr.	Obj. (\$)	1,101,662	1,111,969	1,112,181	1,112,431
		Time (s)	19	922	316	867
	IAC	Obj. (\$)	1,111,116	1,113,739	1,117,006	1,119,490
Time (s)		172	4,775	918	437	
Gap (%)		0.85	1.06	1.34	1.54	

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# Part V: Conclusions



# Observations and Conclusions

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- An integrated ADMM+CCG computing method
  - Supporting information and privacy protection in handling uncertainties
  - Advanced enhancement strategies for fast computation
  - New strategies to ensure convergence in non-convex structures
- Coordination plays a critical role in multi-area grid performance
  - For tie-line scheduling, IAC performs (almost) the same as centralized method
  - For commitment and tie-line scheduling, IAC significantly outperforms non-coordinated control
- Future Improvement
  - Economic implications from IAC computation
  - Novel algorithmic improvement to support fast computing