

Grid Shaping Control for High-IBR Power Systems

Stability Analysis and Control Design

Enrique Mallada



RSRG Seminar, Caltech

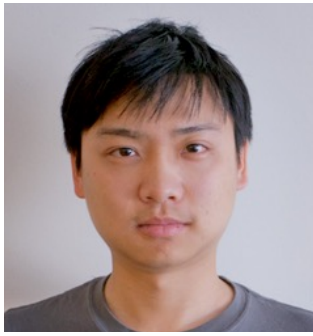
April 3, 2025

Acknowledgements

Students



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Eliza Cohn



Collaborators



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Richard Pates



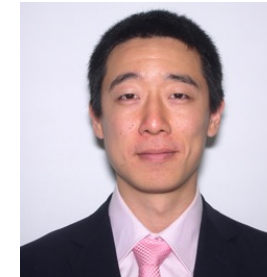
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Decarbonization of electricity is key to mitigate climate change

California lifts renewable energy target to 73% by 2032

The California Public Utilities Commission raised renewable energy procurement targets, plans for a more aggressive decarbonization plan, and includes increased reliability provisions.

FEBRUARY 14, 2022 **RYAN KENNEDY**

Vermont House passes 75% by 2032 renewable energy mandate

Published March 11, 2015

ENVIRONMENT

Maryland bill mandating 50% renewable energy by 2030 to become law, but without Gov. Larry Hogan's signature

By Scott Dance
Baltimore Sun • May 22, 2019 at 6:40 pm

New York mandates 70% renewable energy by 2030

By Kelsey Misbrener | October 15, 2020

Oregon bill targets 100% clean power by 2040, with labor and environmental justice on board

After Democratic cap-and-trade bills faltered in the face of GOP revolts, an electricity-focused, consensus-driven bill gains ground in Oregon.

23 June 2021

Virginia becomes the first state in the South to target 100% clean power

The state's Democratic majority is doing what Democratic majorities do.

By David Roberts | @drvlt | Updated Apr 13, 2020, 2:56pm EDT

Decarbonization of electricity is key to mitigate climate change

California lifts renewable energy target to 73% by 2032

The California Public Utilities Commission has lifted its renewable energy targets, plans for a more aggressive decarbonization and reliability provisions.

FEBRUARY 14, 2022 RYAN KENNEDY

Vermont Hopes to reach 100% renewable energy

Published March 11, 2015

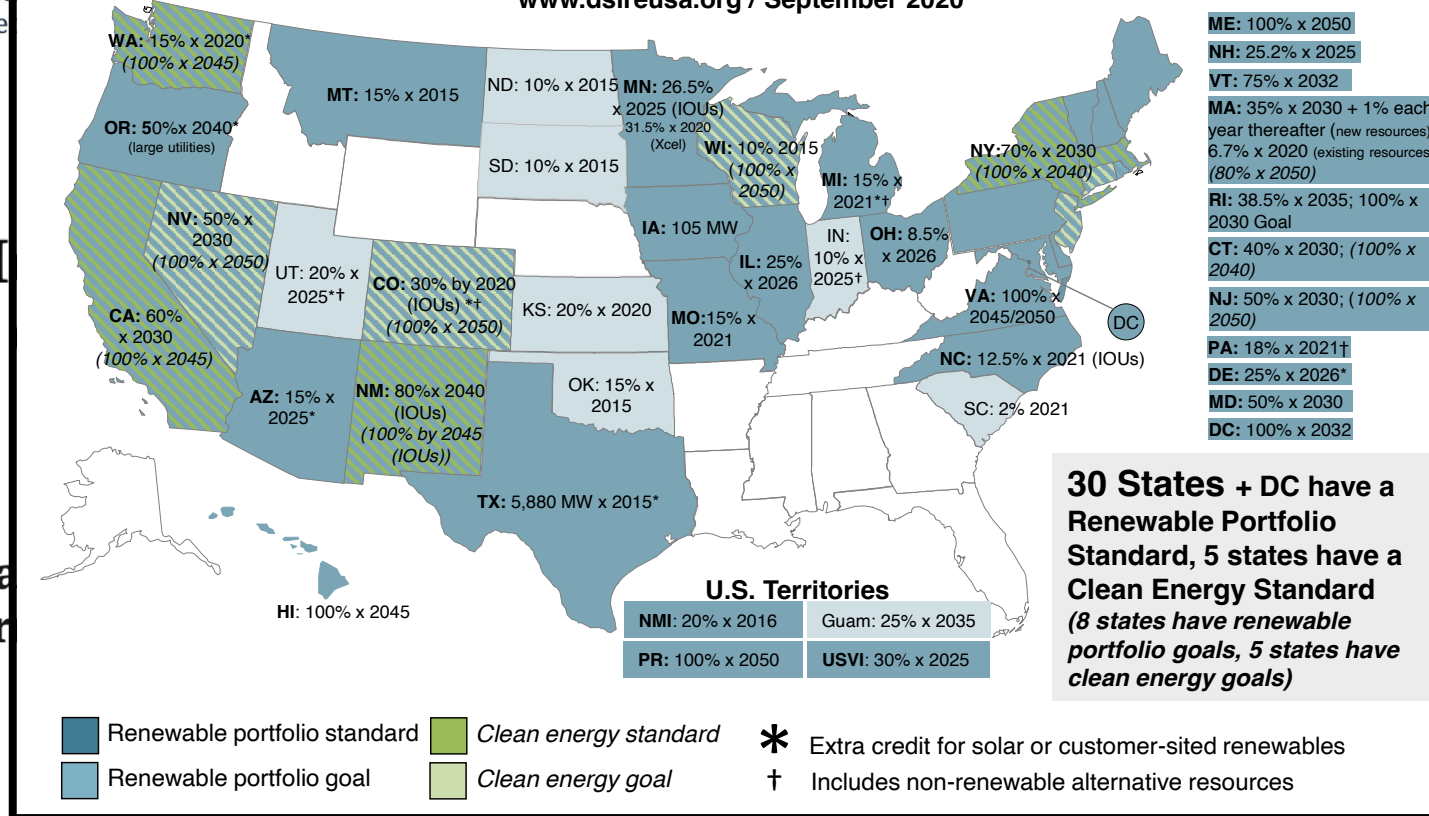
ENVIRONMENT

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Renewable & Clean Energy Standards

www.dsireusa.org / September 2020

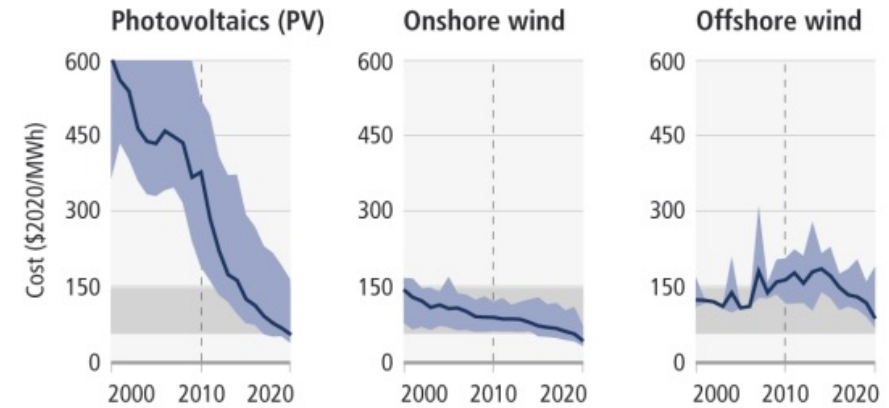
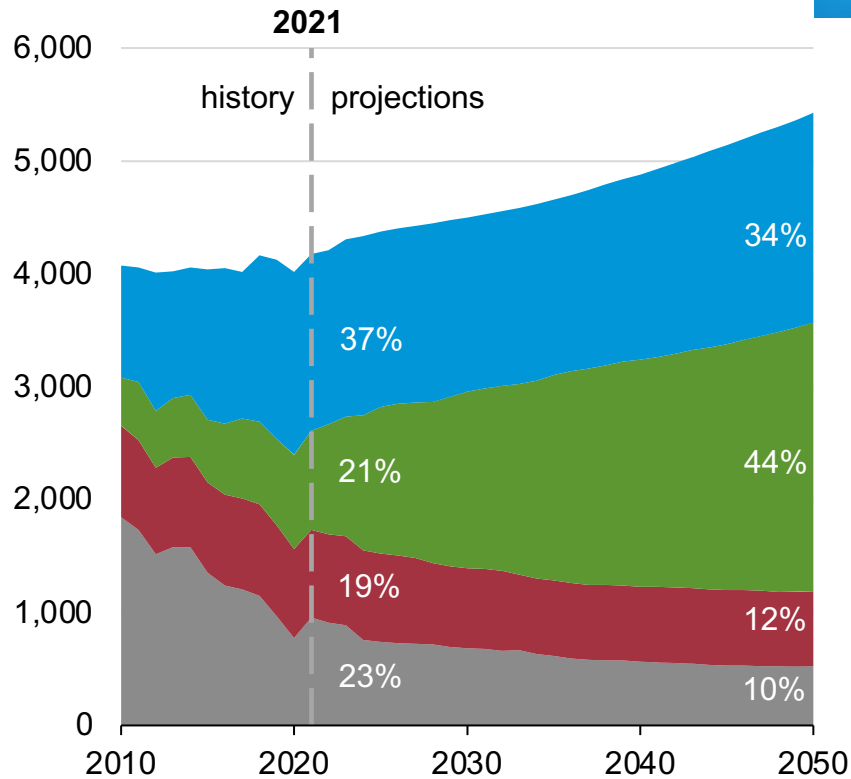


The state's Democratic majority is doing what Democratic majorities do.

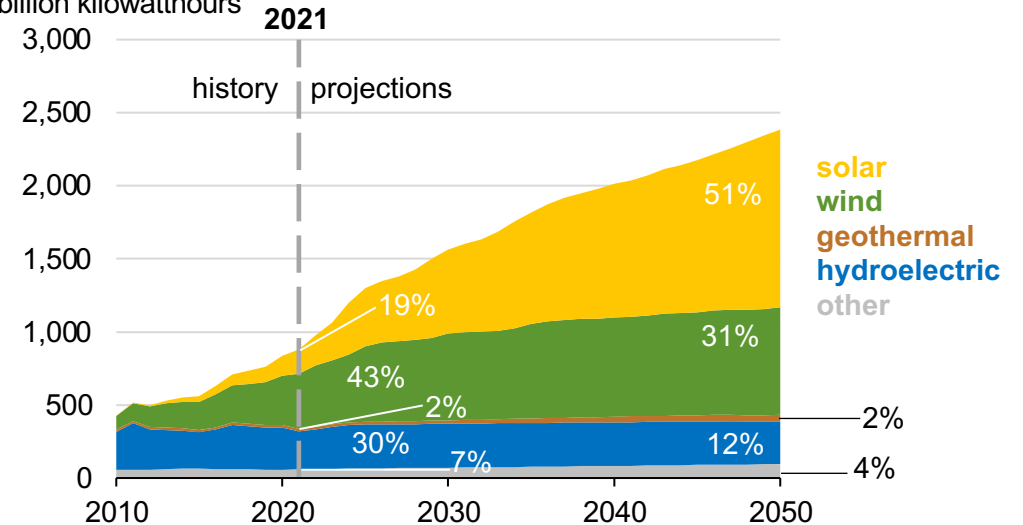
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Decarbonization of electricity is key to mitigate climate change

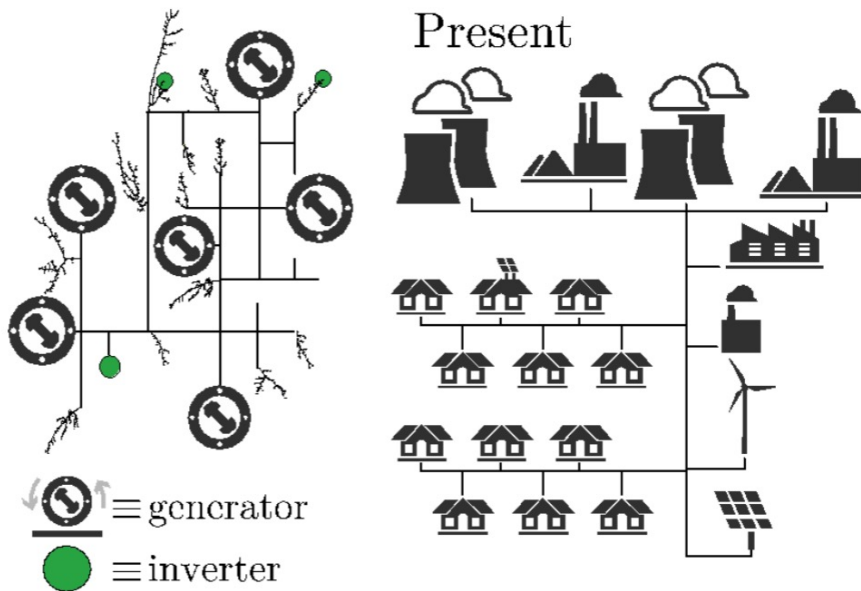
U.S. electricity generation from selected fuels
AEO2022 Reference case
 billion kilowatthours



U.S. renewable electricity generation, including end use
AEO2022 Reference case
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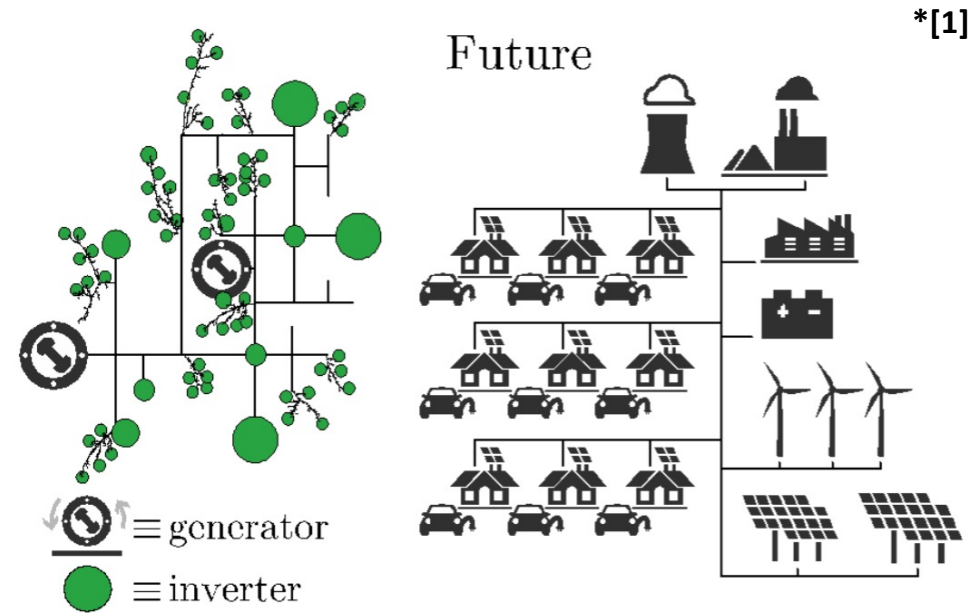


The Future Grid



Present grid

- dispatchable generation
- high inertial response
- strong voltage support
- well known physics

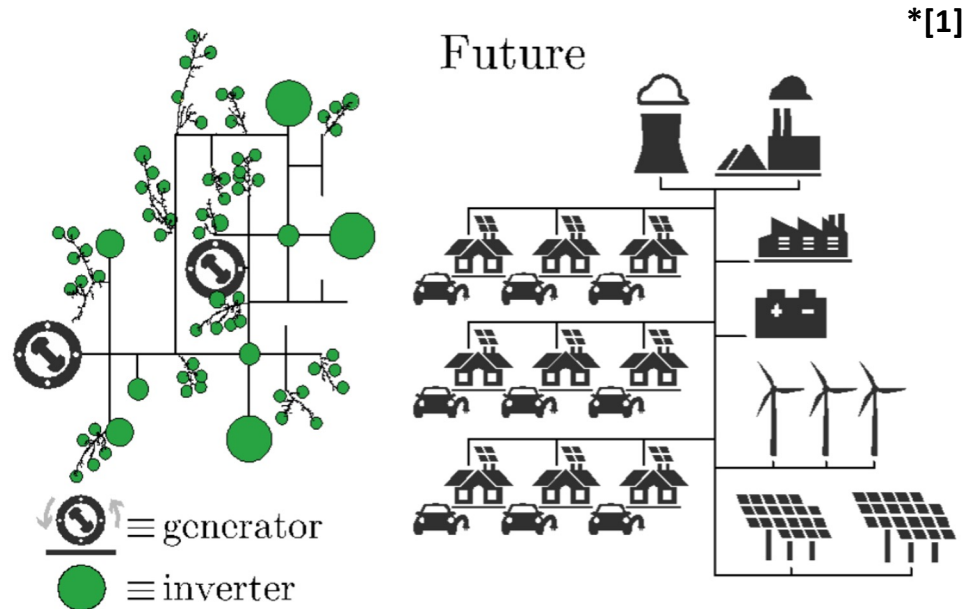


Future

- variable and distributed generation
- limited inertia levels
- weak voltage support
- proprietary control laws (black box)

[1] Lin et al. Research roadmap on grid-forming inverters. Technical report, National Renewable Energy Lab.(NREL), Golden CO, 2020

The Future Grid



Future

- variable and distributed generation
- limited inertia levels
- weak voltage support
- proprietary control laws (black box)

Selected challenges

- increased system **uncertainty**
- **sensitivity** to disturbances
- new forms of **instabilities**, induced by inverter-based resources
- need to compensate for **reduced inertia grid strength**

Research questions:

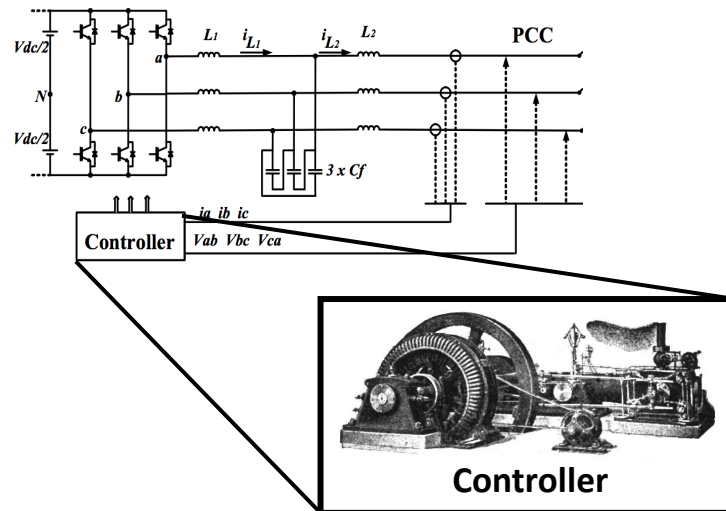
- How should we control a grid with limited inertial/voltage support?
- Should we try to mimic SGs response? Or find new and more efficient control paradigms, suitable for IBRs?

[1] Lin et al. Research roadmap on grid-forming inverters. Technical report, National Renewable Energy Lab.(NREL), Golden CO, 2020

Inverter-based Control

Current approach: Use inverter-based control to mimic generators response

Virtual Synchronous Generator



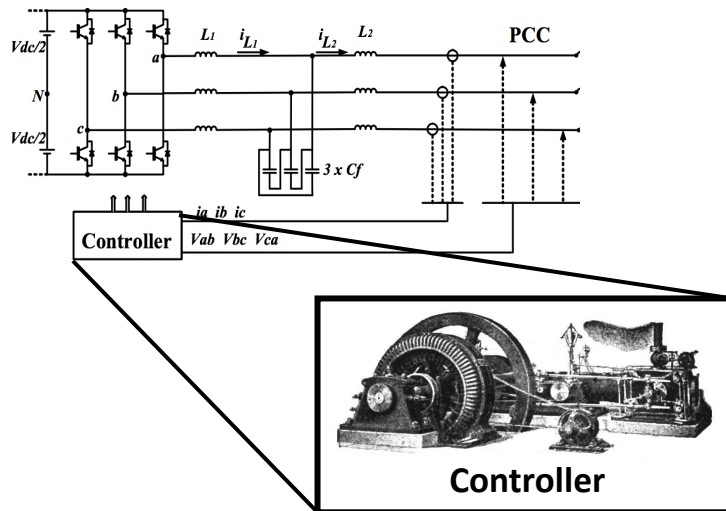
Telecom Analogy



Inverter-based Control

Current approach: Use inverter-based control to mimic generators response

Virtual Synchronous Generator



Telecom Analogy



It works, but perhaps
there is
something better...

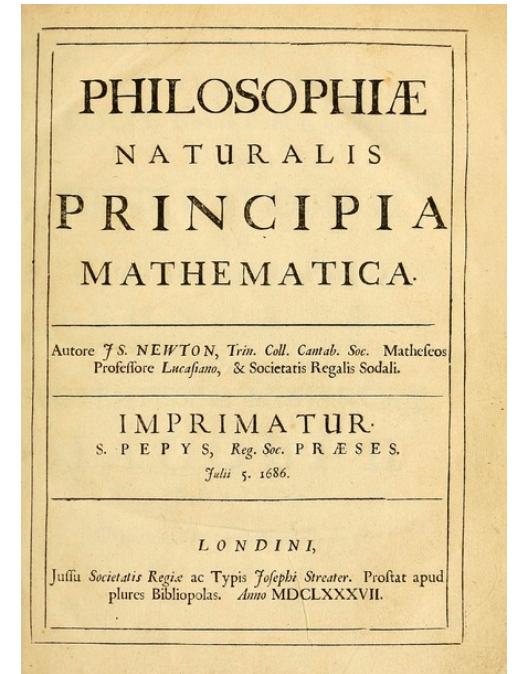
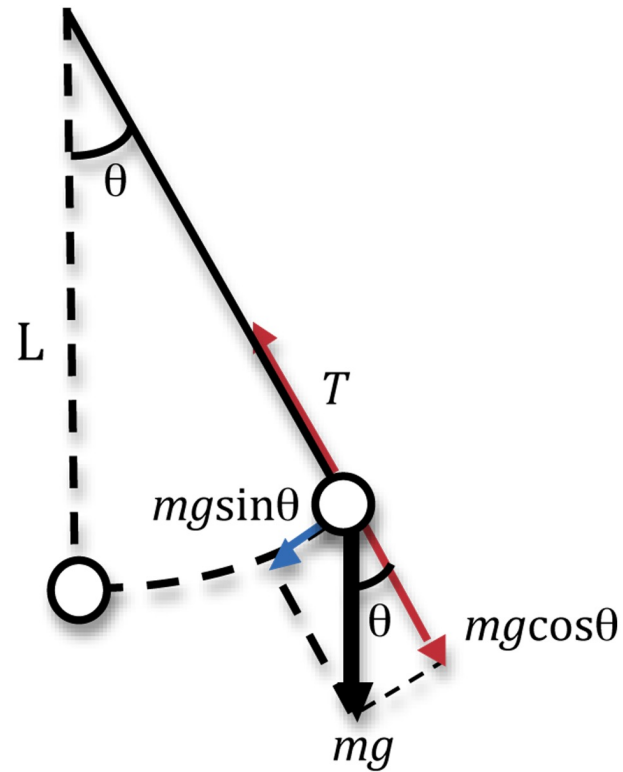
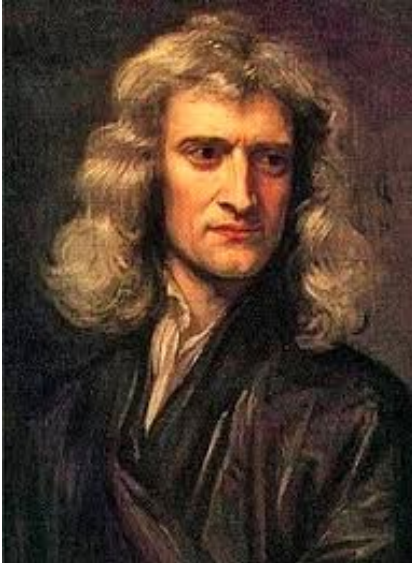
Outline

- **Merits and trade-offs of low inertia**
 - Control Perspective: Lighter systems are easier to control!
- **Scale-free Stability Analysis of Grids**
 - Generalizes passivity notions using network information
- **Analysis of Weakly-Connected Coherent Networks**
 - Generalized Center of Inertia captures IBR dynamics
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 - Grid-following/forming control framework for controlling future grids

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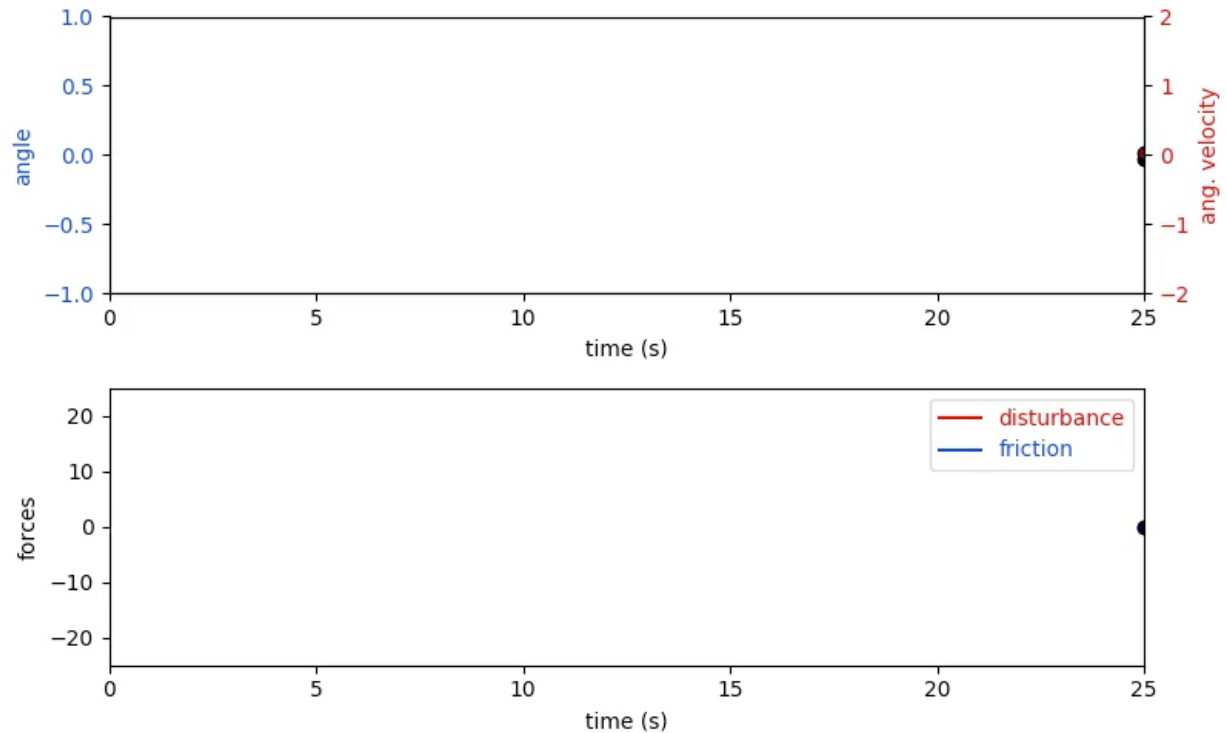
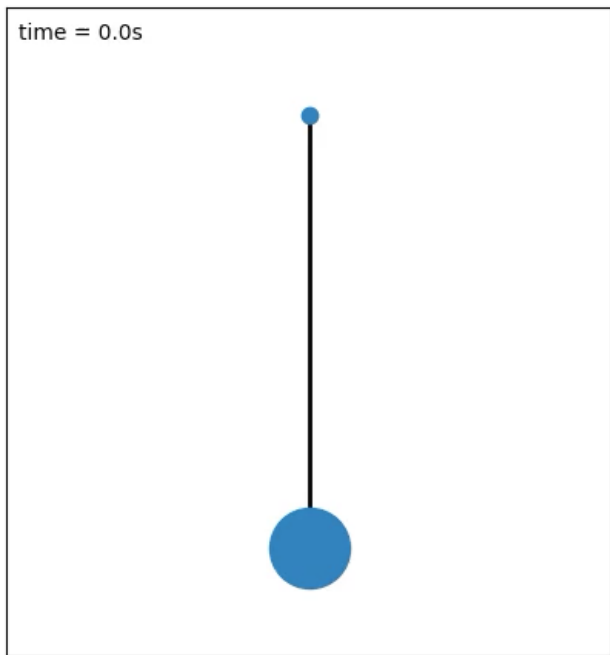
Merits and Trade-offs of Inertia



$$\ddot{\theta} = -\frac{d}{m}\dot{\theta} - g \sin \theta + \frac{f}{m}$$

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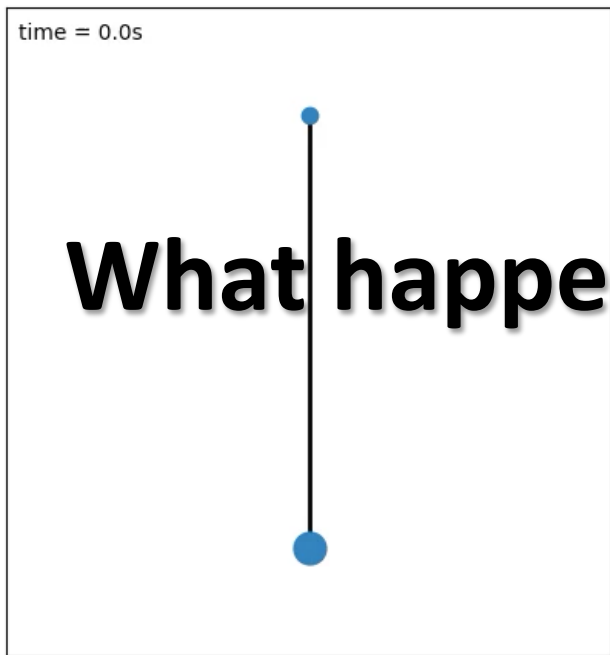


Pros: Provides natural disturbance rejection

Cons: Hard to regain steady-state

Merits and Trade-offs of **Low** Inertia

$$\ddot{\theta} = -\frac{d}{m}\dot{\theta} - g \sin \theta + \frac{f}{m}$$



What happens when one adds **control**?

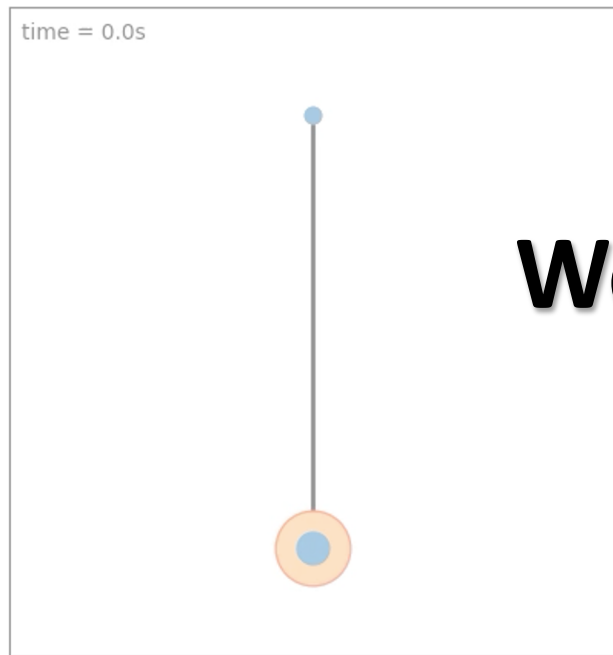


Cons: Susceptible to disturbances

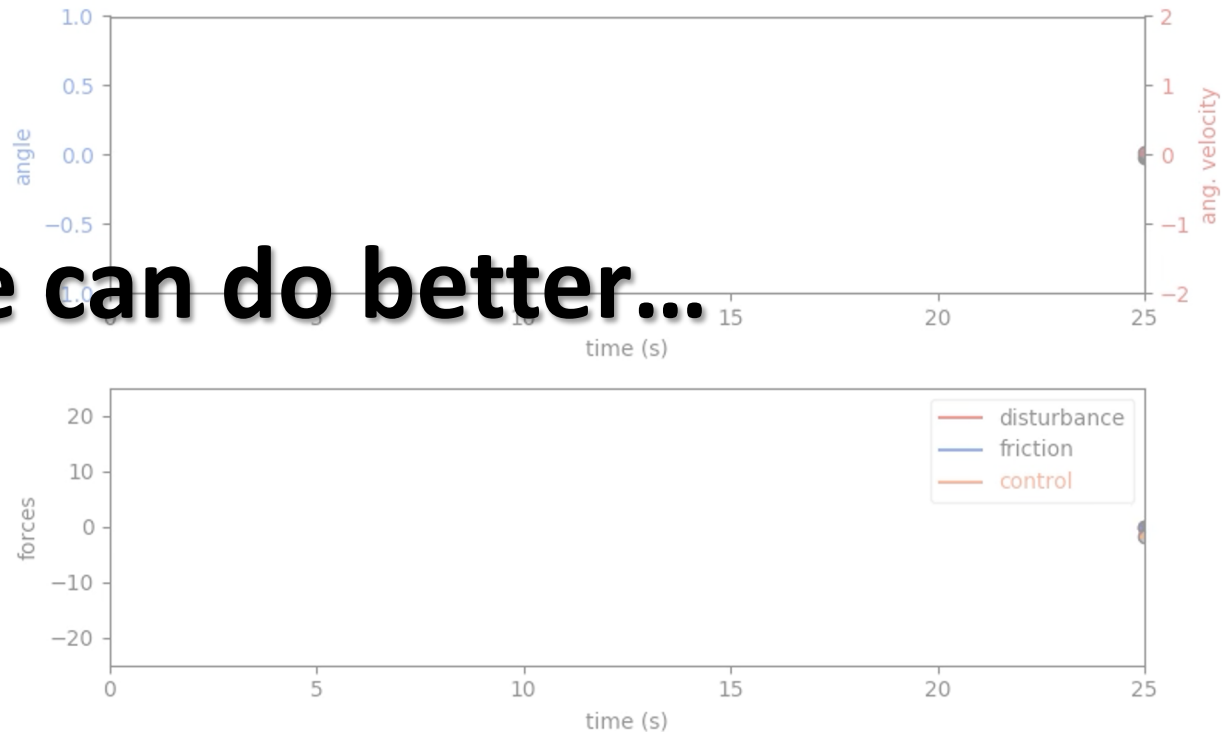
Pros: Regains steady-state faster

Control of **Low** Inertia Pendulum

Virtual **Mass** Control: $m\ddot{\theta} = -d\dot{\theta} - mg \sin \theta + f - \nu\ddot{\theta}$



We can do better...



Pros:

Provides disturbance rejection

Cons:

Hard to regain steady-state + **excessive control effort**

Control of **Low** Inertia Pendulum

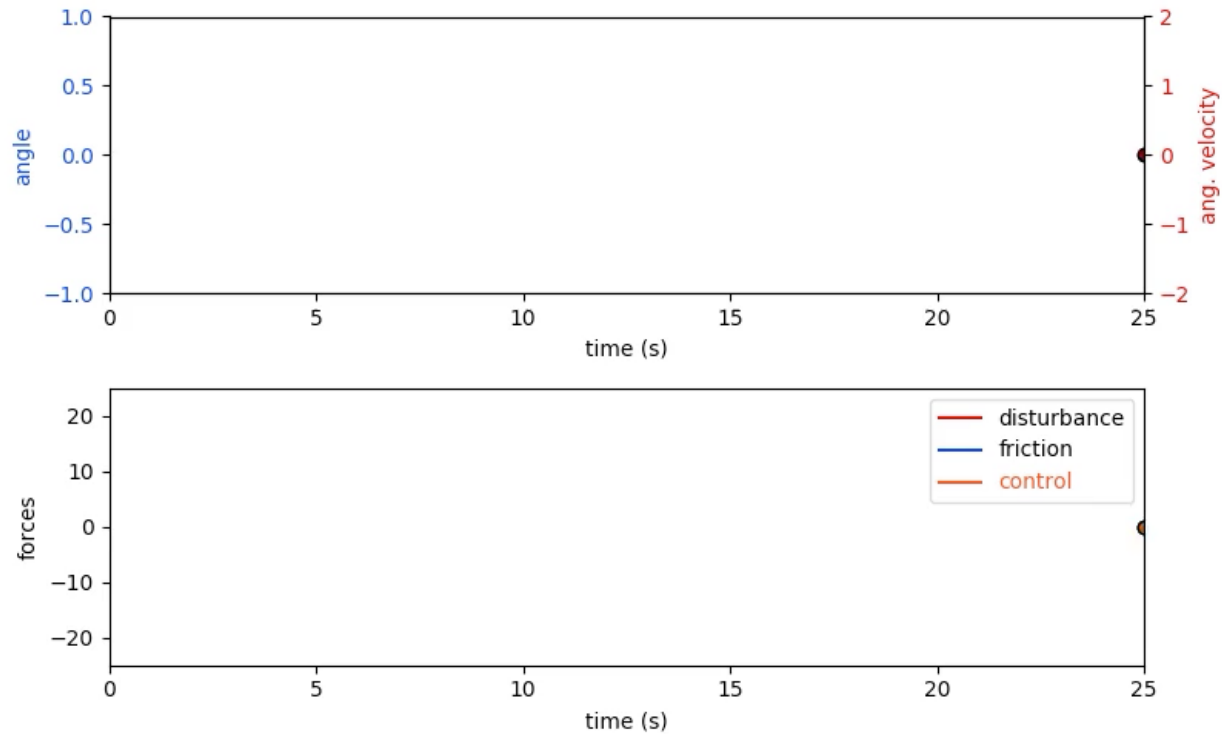
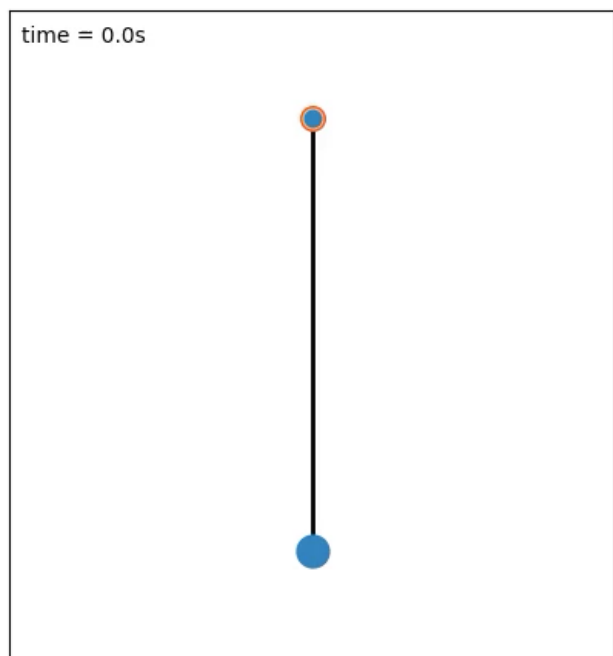


Yan Jiang



Richard Pates

Dynamic Droop:
$$m\ddot{\theta} = -d\dot{\theta} - mg \sin \theta + f + x$$
$$\tau' \dot{x} = -x - (r_r^{-1} \dot{\theta} + \tau' \nu' \ddot{\theta})$$



Control of **Low** Inertia Pendulum

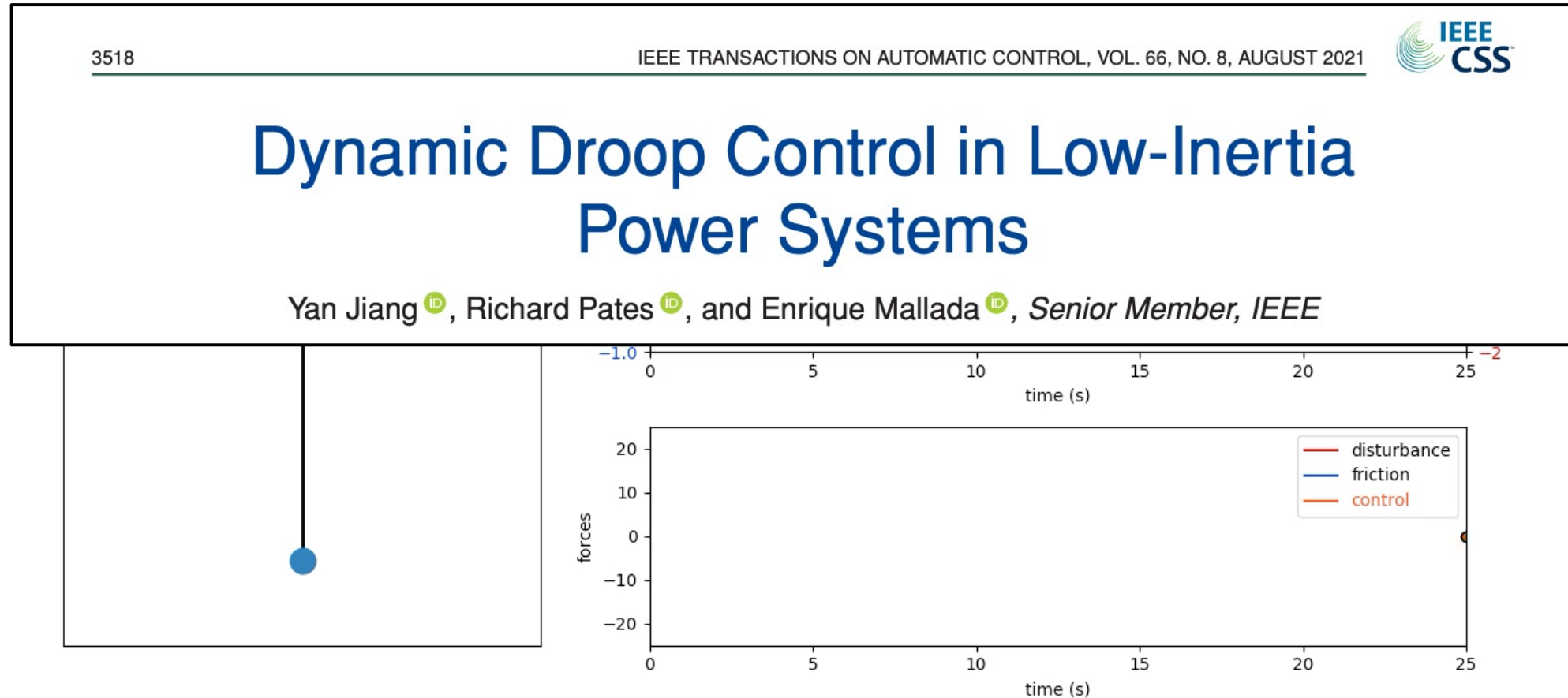


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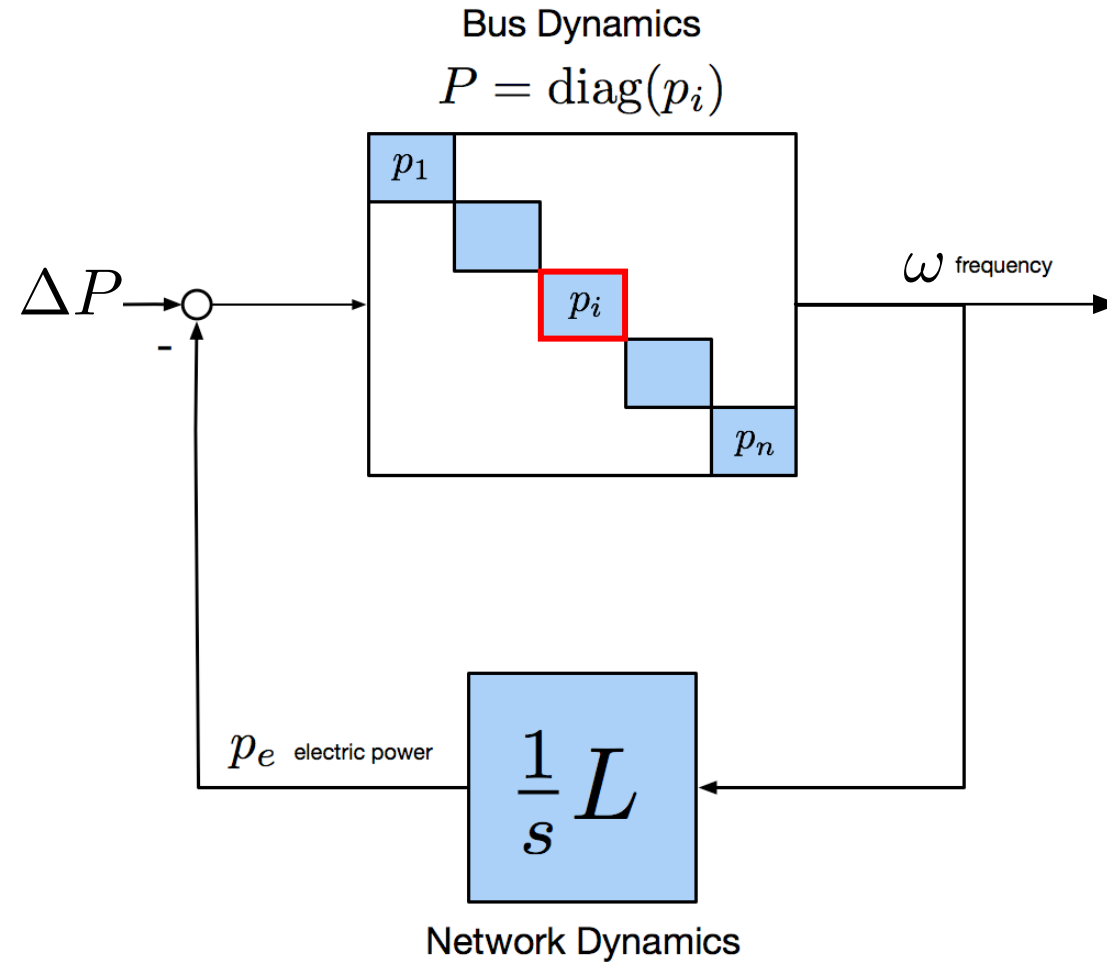


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Dynamic Droop: $m\ddot{\theta} = -d\dot{\theta} - mg \sin \theta + f + x$



Power Network Model



Laplacian Matrix

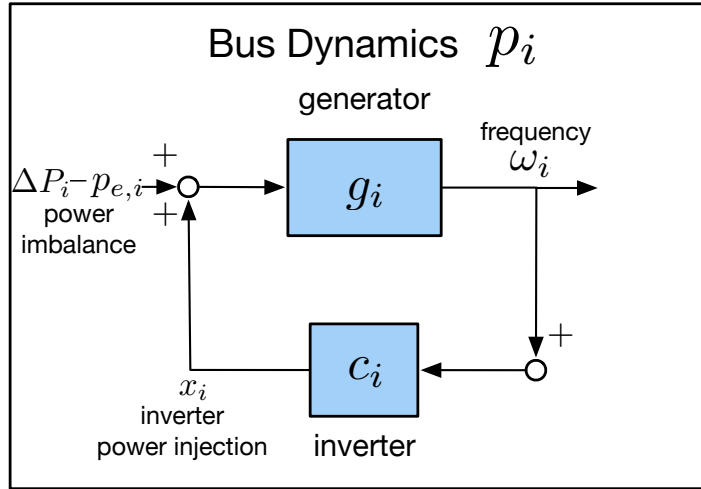
$$L_{ij} = \begin{cases} -B_{ij} & \text{if } ij \in E \\ \sum_k B_{ik} & \text{if } i = j \\ 0 & \text{o.w.} \end{cases}$$

Linearized Power Flows

$$B_{ij} = v_i v_j b_{ij} \cos(\theta_i^* - \theta_j^*)$$

[Bergen Hill '81]

Bus Dynamics

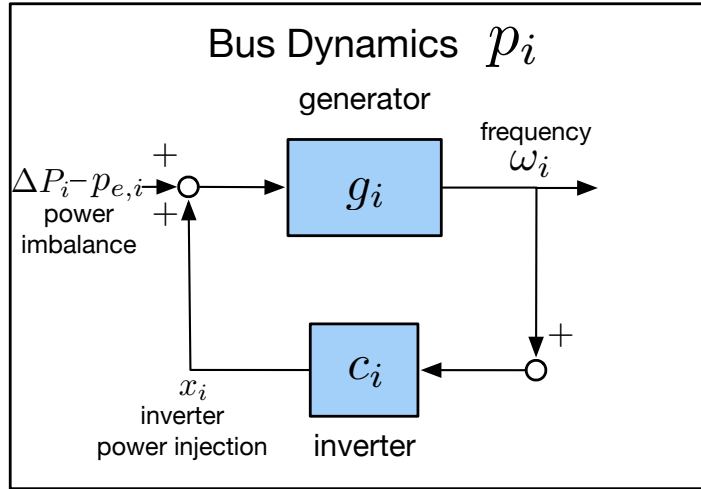


Generator: $g_i : (\Delta P_i - p_{e,i} + x_i) \mapsto \omega_i$

Model: Swing Equations + Turbine

$$g_i : \begin{cases} \dot{\theta}_i = \omega_i \\ M_i \dot{\omega}_i = -D_i \omega_i + q_i + (\Delta P_i - p_{e,i} + x_i) \\ \tau_i \dot{q}_i = -R_{g,i}^{-1} \omega_i - q_i \end{cases}$$

Bus Dynamics



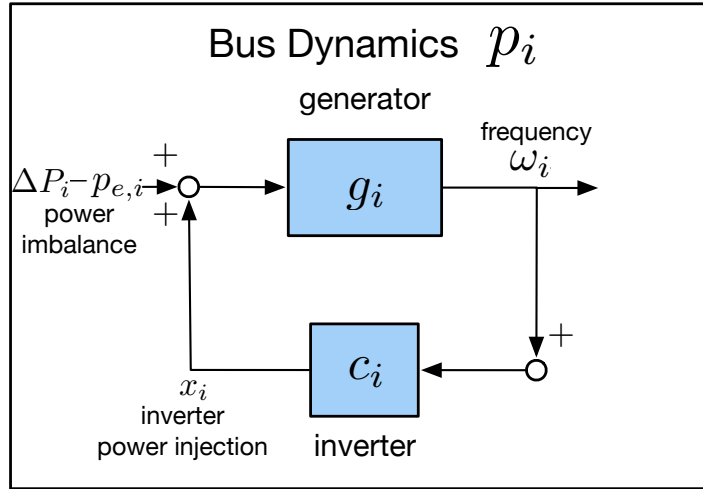
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$$g_i(s) = \frac{1}{M_i s + D_i + \frac{R_{g,i}^{-1}}{\tau_i s + 1}}$$

Bus Dynamics



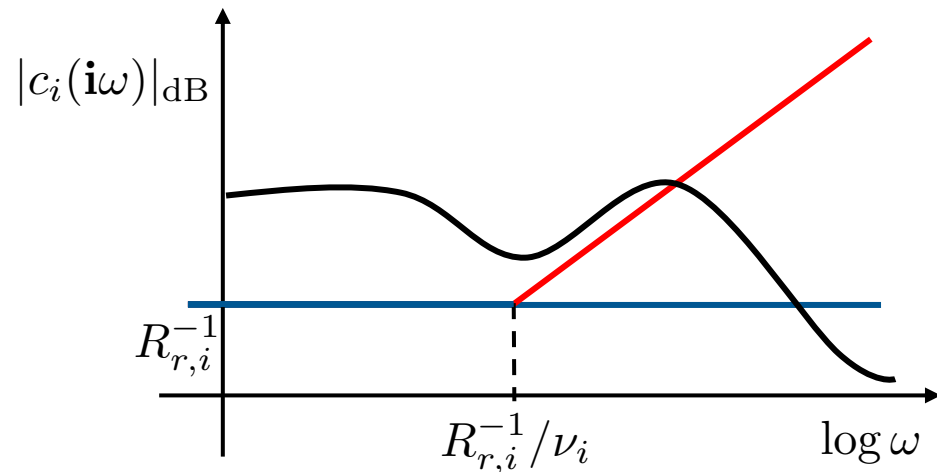
Grid Following Inverter: $c_i : \omega_i \mapsto x_i$

Droop Control and Virtual Inertia:

$$c_i : \begin{cases} x_i = -(\nu_i \dot{\omega}_i + R_{r,i}^{-1} \omega_i), \\ c_i(s) = -(\nu_i s + R_{r,i}^{-1} \omega_i) \end{cases}$$

Closed-loop Bus Dynamics:

$$p_i : \begin{cases} \dot{\theta}_i = \omega_i \\ (M_i + \nu_i) \dot{\omega}_i = -(D_i + R_{r,i}^{-1}) \omega_i + q_i + (\Delta P_i - p_{e,i}) \\ \tau_i \dot{q}_i = -q_i - R_{g,i}^{-1} \omega_i \end{cases}$$

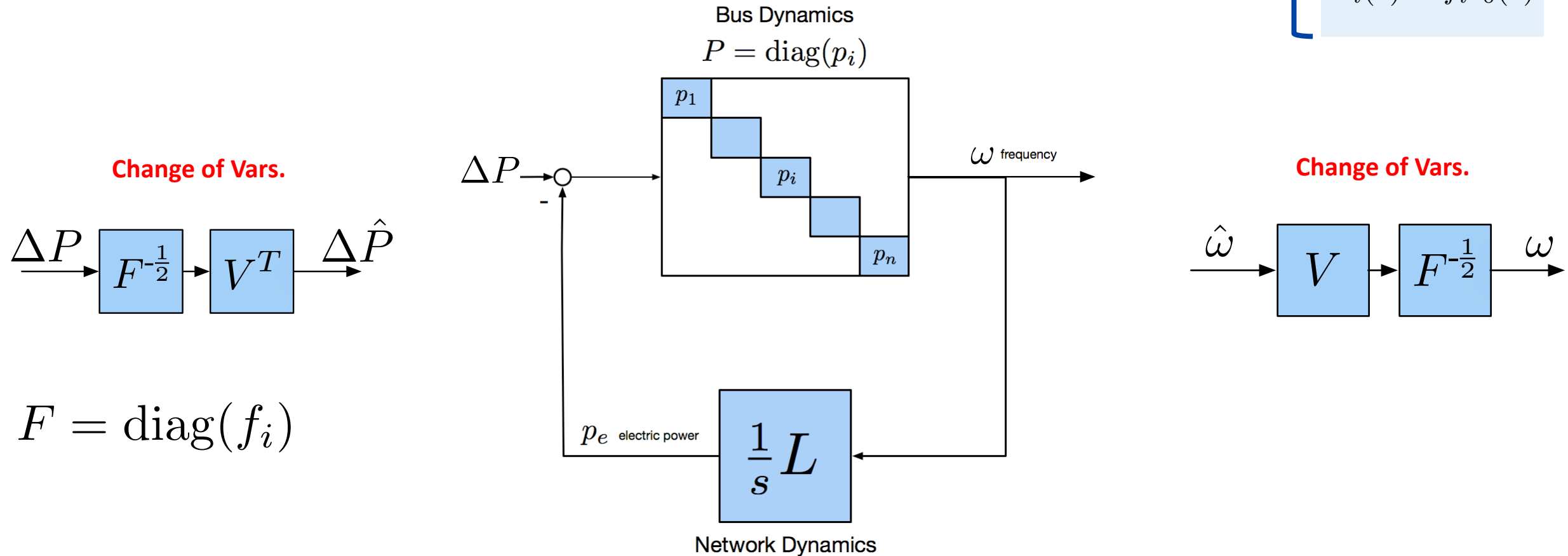


Modal Decomposition for Multi-Rated Machines

Assumption: Let f_i be the machine relative inertia ($f_i = \frac{M_i}{\max_j M_j}$), and

$$g_i(s) = \frac{1}{f_i} g_0(s)$$

$$c_i(s) = f_i c_0(s)$$



[Paganini M '17 , Guo Low 18']

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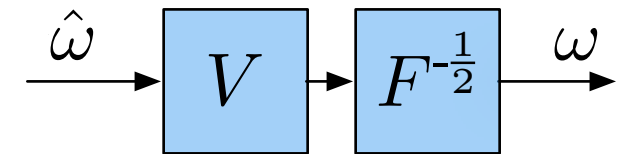
$$g_i(s) = \frac{1}{f_i} g_0(s)$$

$$c_i(s) = f_i c_0(s)$$

Center of Inertia

$$\omega_{\text{CoI}}(t) = \frac{\sum_{i=1}^n M_i \omega_i(t)}{\sum_{i=1}^n M_i}$$

Change of Vars.

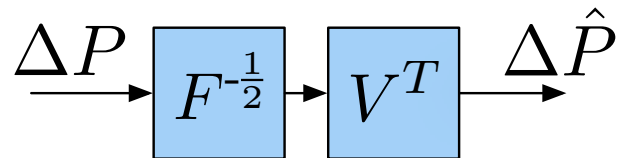


Sync Error

$$\tilde{\omega}_i(t) = \omega_i(t) - \omega_{\text{CoI}}(t)$$

[Paganini M '17 , Guo Low 18']

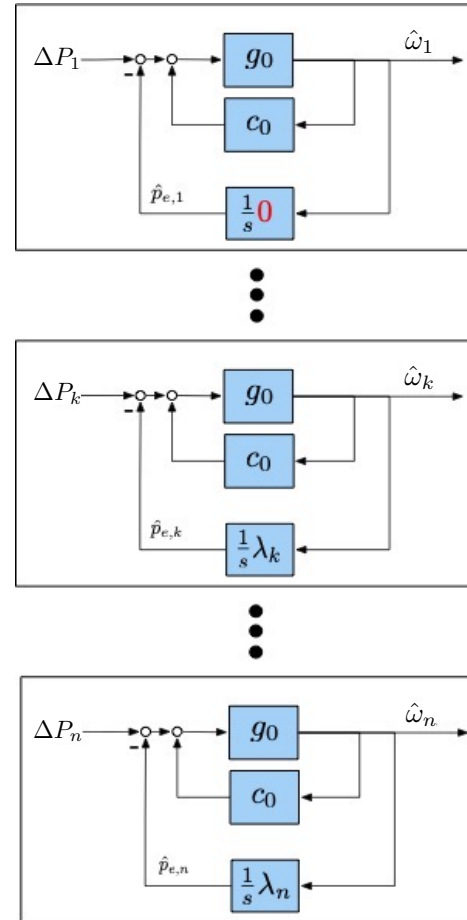
Change of Vars.



$$F = \text{diag}(f_i)$$

Eigenvalues of: $L_F = F^{-\frac{1}{2}} L F^{-\frac{1}{2}}$

$$0 = \lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_{n-1}$$



Control of **Low** Inertia Pendulum

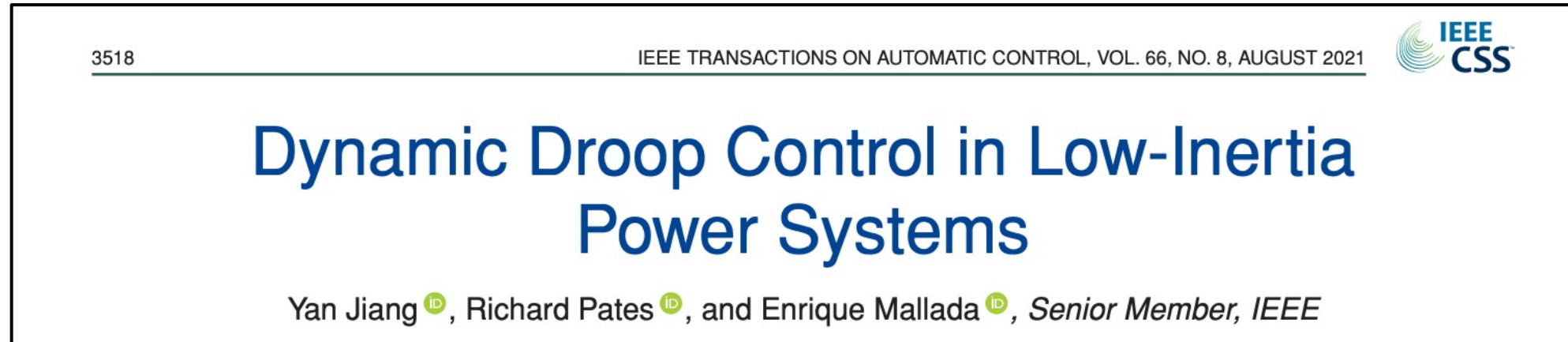


Yan Jiang



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Dynamic Droop: $m\ddot{\theta} = -d\dot{\theta} - mg \sin \theta + f + x$



Dynamic Droop Benefits

- Overshoot Elimination in Nadir
- Disturbance Rejection
- Noise Attenuation
- Reduce Inter-area Oscillations

Caveat

- Control design limited to co-located resources (SGs and **GFL**-IBRs)
- Restrictive assumptions: Proportional dynamics ($p_i(s) = f_i p_0(s)$)

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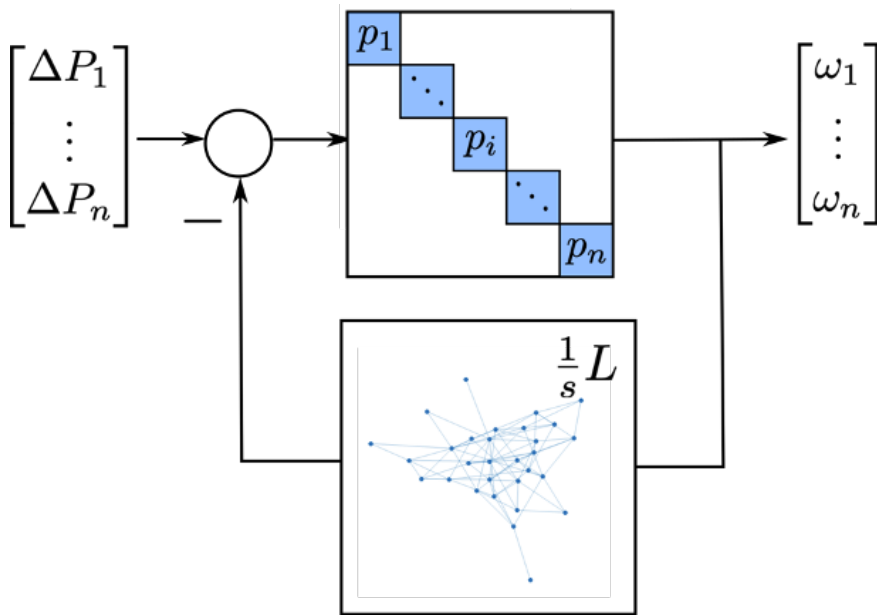
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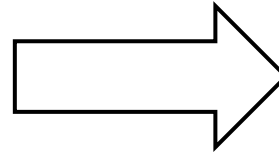
Decentralized Stability Analysis in Power Grids [TCNS 19]



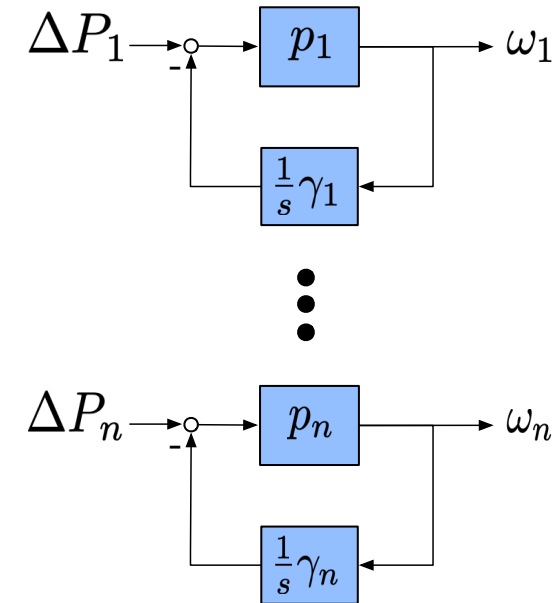
Richard Pates



1. When does this interconnection is stable?



2. Can we analysis and control design based on **local** rules?



Problem Setup:

- Linearized power flows, lossless

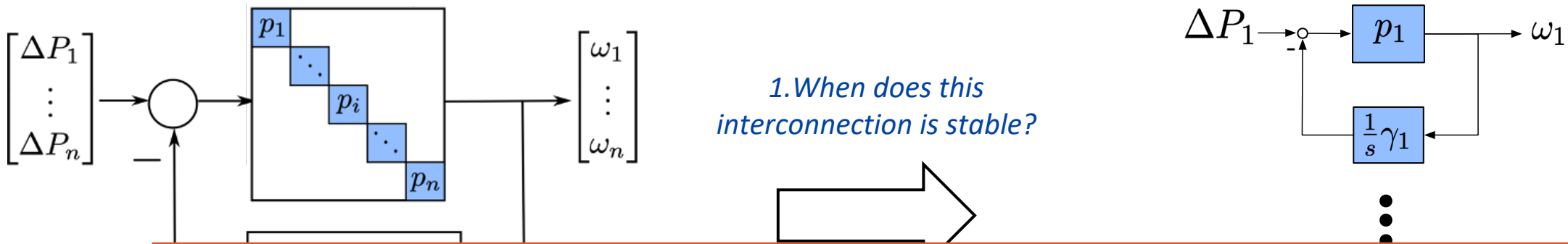
$$L_{ij} = -b_{ij}v_i v_j \cos(\theta_i^* - \theta_j^*)$$
- Bus i : arbitrary siso transfer function:

$$\omega_i = p_i(s) \Delta P_i \quad (\text{SGs or GFM-IBRs})$$

Decentralized Stability Analysis in Power Grids [TCNS 19]



Richard Pates



1. When does this interconnection is stable?

Can we use **network information** to relax passivity conditions?

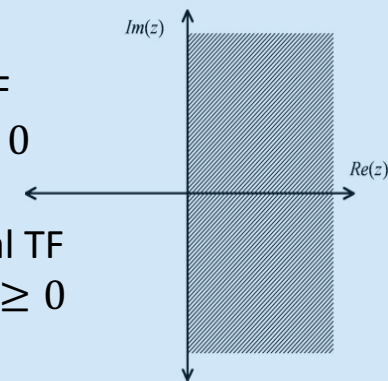
Standard Approach: Passivity

- If $p_i(s)$ is strictly positive real (SPR), then the interconnection is stable for **all networks L** !

Converse: for **unknown network (L)**, passivity is also **necessary**. [TCNS 19]

Positive Real (PR) TF
 $\text{Re}[p_i(s)] \geq 0$

Strictly Positive Real TF
 $\text{Re}[p_i(s - \varepsilon)] \geq 0$



Classical Result: Absolute Stability

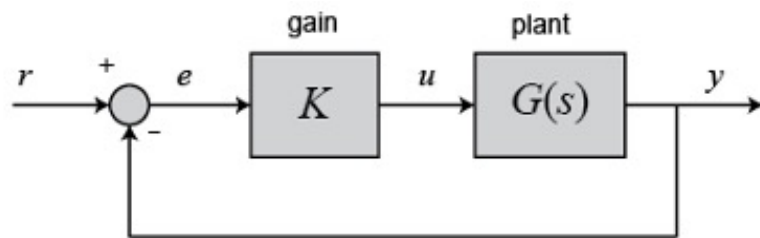
IEEE TRANSACTIONS ON AUTOMATIC CONTROL

Frequency Domain Stability Criteria—Part I

R. W. BROCKETT, MEMBER, IEEE AND J. L. WILLEMS

Abstract—The objective of this paper is to illustrate the limita-

II. THE GENERALIZED POPOV THEOREM

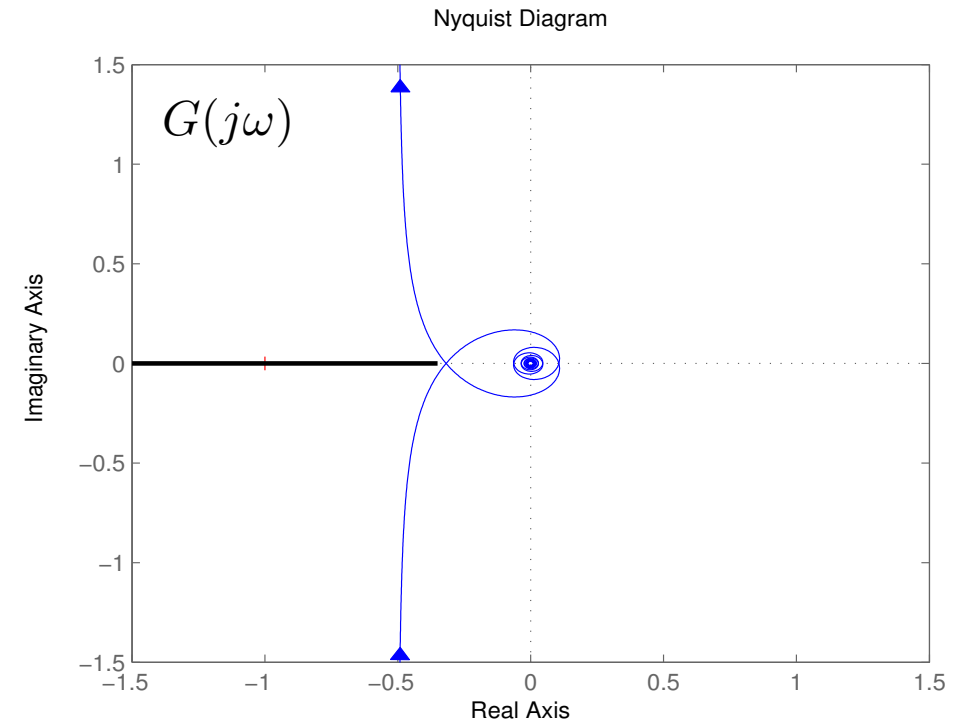


Stable for $0 \leq K \leq k^*$?

Assume: $G(s)$ is stable

Define: $h(s) \in PR$ (passive)

Test: If $h(s)(1 + k^*G(s)) \in SPR$ (strictly)
then, **yes!**



Classical Result: Absolute Stability

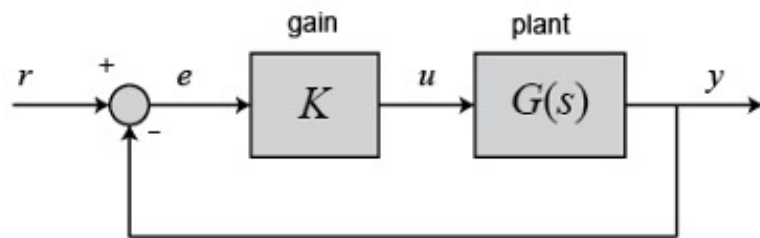
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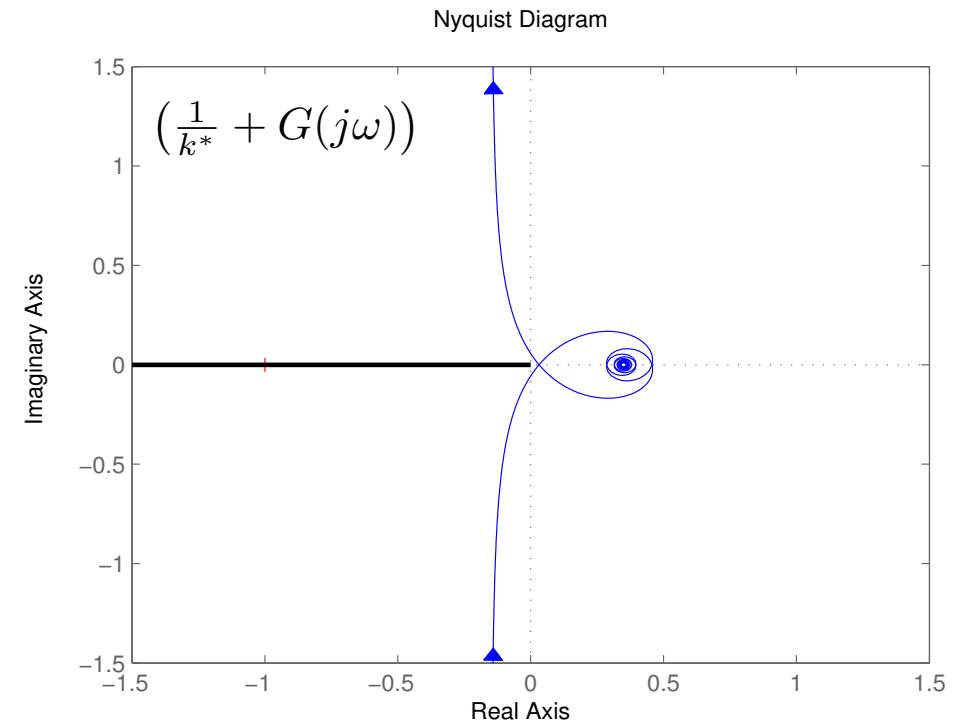


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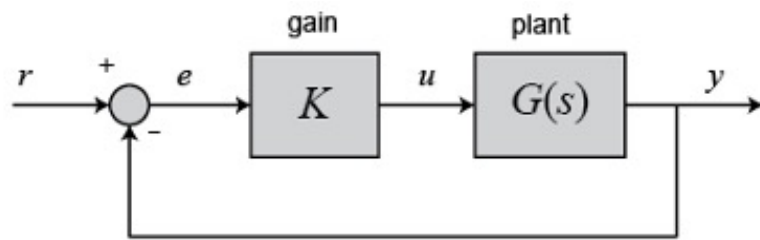
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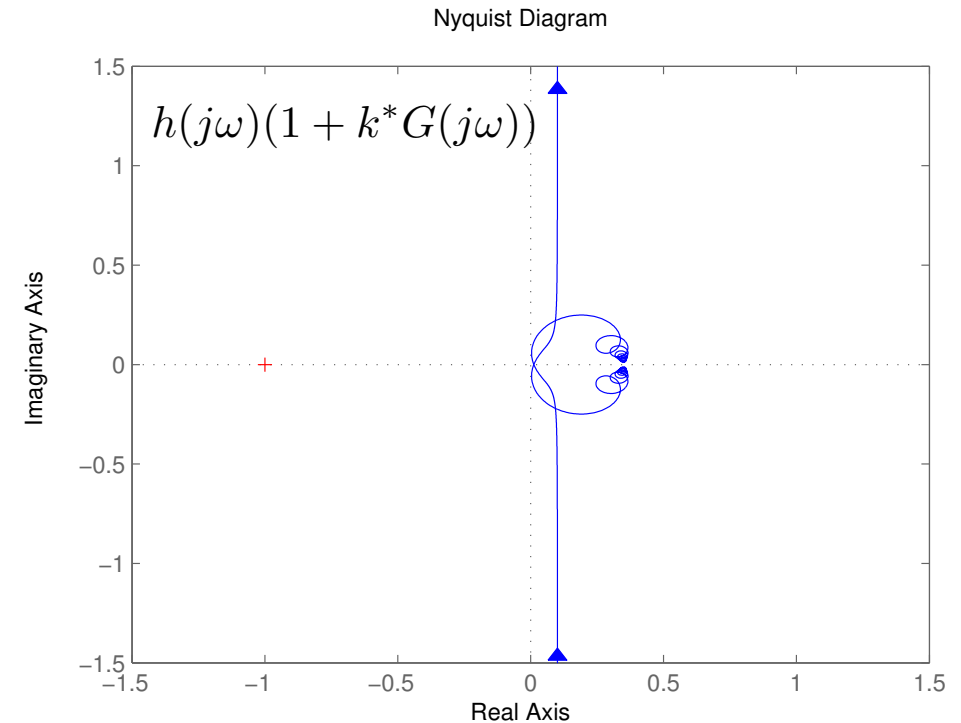


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Scale-free Stability Analysis

Key Idea: Exploit limited network information to relax passivity condition

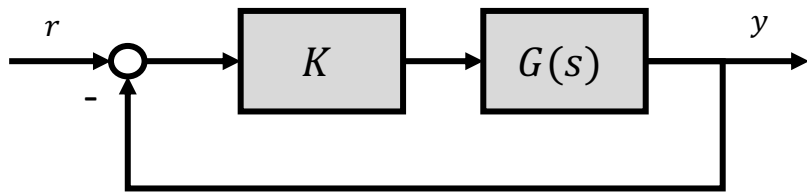
- Let γ_i be a local connectivity bound: $\sum_{j \in N_i} L_{ij} \leq \frac{\gamma_i}{2}$ $L_{ij} = -b_{ij}v_i v_j \cos(\theta_i^* - \theta_j^*)$

Brockett & Willems '65

Assume: $G(s)$ is stable

Define: $h(s) \in PR$ (passive)

Test: If $h(s)(1 + k^*G(s)) \in SPR$ (strictly) then system is stable for all $0 \leq K \leq k^*$

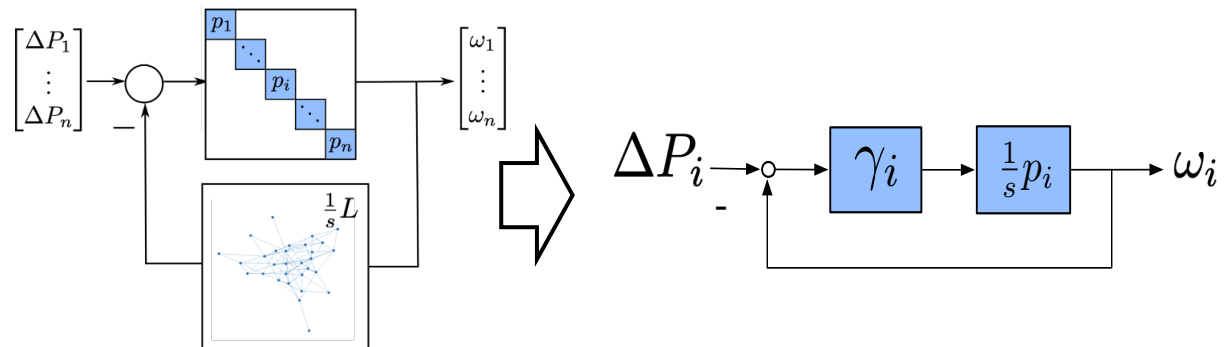


Pates & M 2019

Assume: $p_i(s)$ is stable

Define: $h(s) \in PR$ (passive)

Test: If $h(s)\left(1 + \gamma_i \frac{1}{s} p_i(s)\right) \in SPR, \forall i$, then system stable for networks $\sum_{j \in N_i} L_{ij} \leq \frac{\gamma_i}{2}, \forall i$



Scale-free Stability Analysis

Key Idea: Exploit limited network information to relax passivity condition

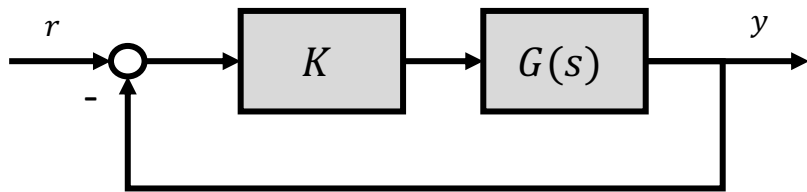
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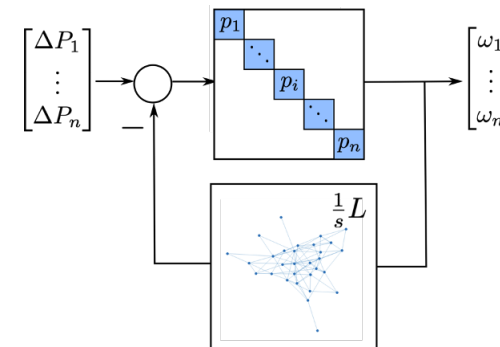


Pates & M 2019

Assume: $p_i(s)$ is stable

Define: $h(s) \in PR$ (passive)

Test: If $h(s)\left(1 + \gamma_i \frac{1}{s} p_i(s)\right) \in SPR, \forall i$, then
system stable for networks $\sum_{j \in N_i} L_{ij} \leq \frac{\gamma_i}{2}, \forall i$



Scale-free Stability Analysis

Key Idea: Exploit limited network information to relax passivity condition

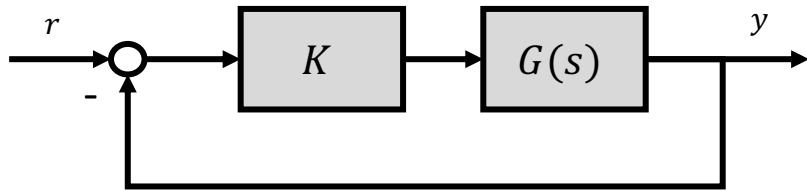
- Let γ_i be a local connectivity bound: $\sum_{j \in N_i} L_{ij} \leq \frac{\gamma_i}{2}$ $L_{ij} = -b_{ij}v_i v_j \cos(\theta_i^* - \theta_j^*)$

Brockett & Willems '65

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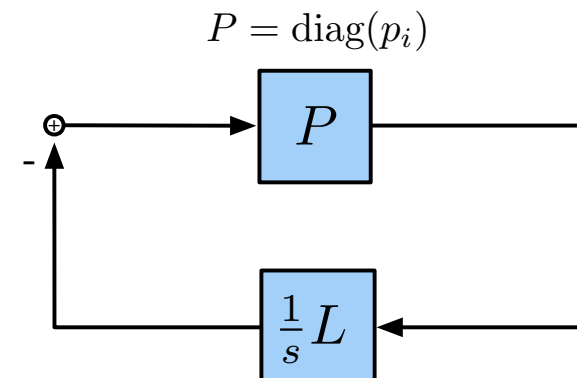


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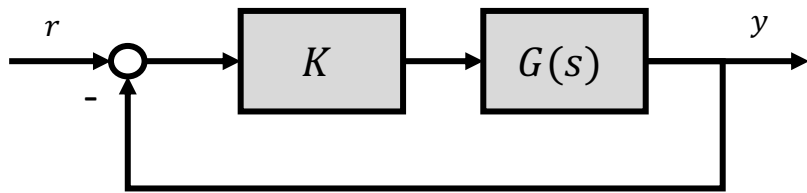
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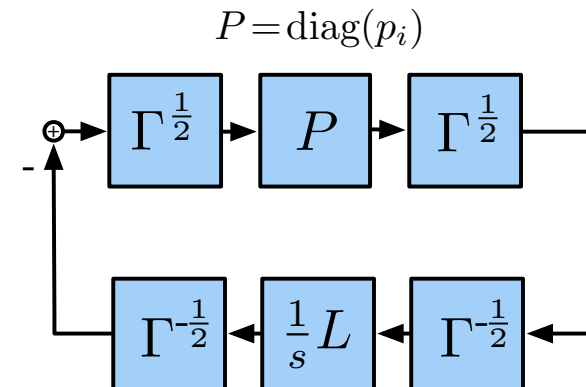


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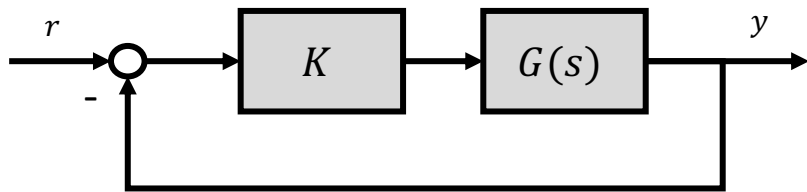
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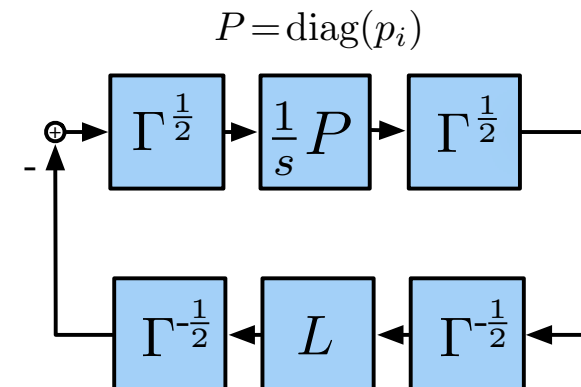


Pates & M 2019

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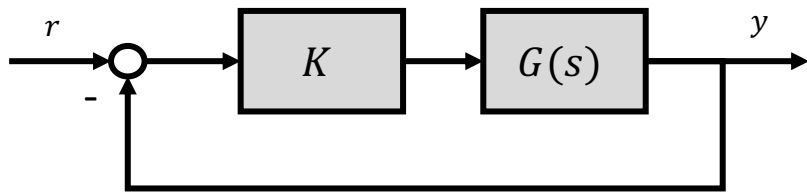
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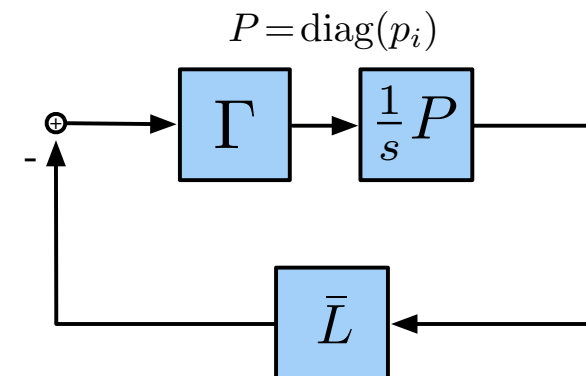


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$$0 \preceq \bar{L} := \Gamma^{-\frac{1}{2}} L \Gamma^{-\frac{1}{2}} \preceq I$$

Examples

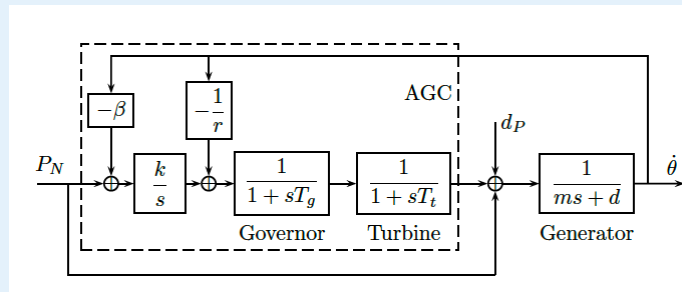
Delay Robustness of Swing Equations

Let
$$p_i(s) = \frac{1}{M_i s + D_i e^{-\tau_i s}}$$

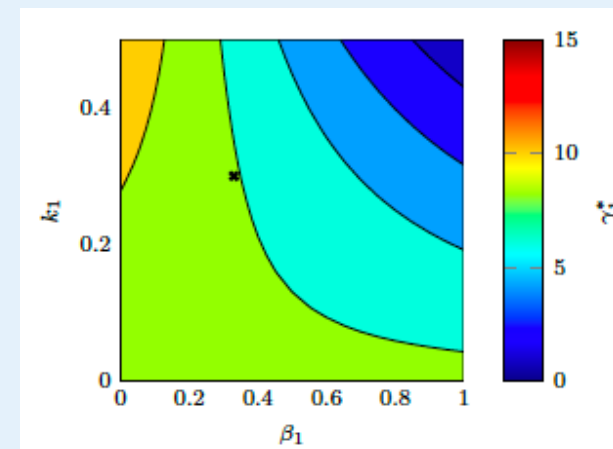
Given $\tau^* < \frac{\pi}{2}$, then, for any network such that $2 \sum_{j \in N_i} L_{ij} < \gamma^*$ with $\gamma^* \approx \frac{\pi M_i (\frac{\pi}{2} - \tau^*)}{2 \left(\frac{M_i \tau^*}{D_i} \right)^2}$

the delayed swing equations are stable for whenever $\tau_i \leq \tau^* \frac{M_i}{D_i}$

Automatic Generation Control



m	d	T_g	T_t	r	β	k
0.16	0.02	0.08	0.40	3.00	0.33	0.30



Outline

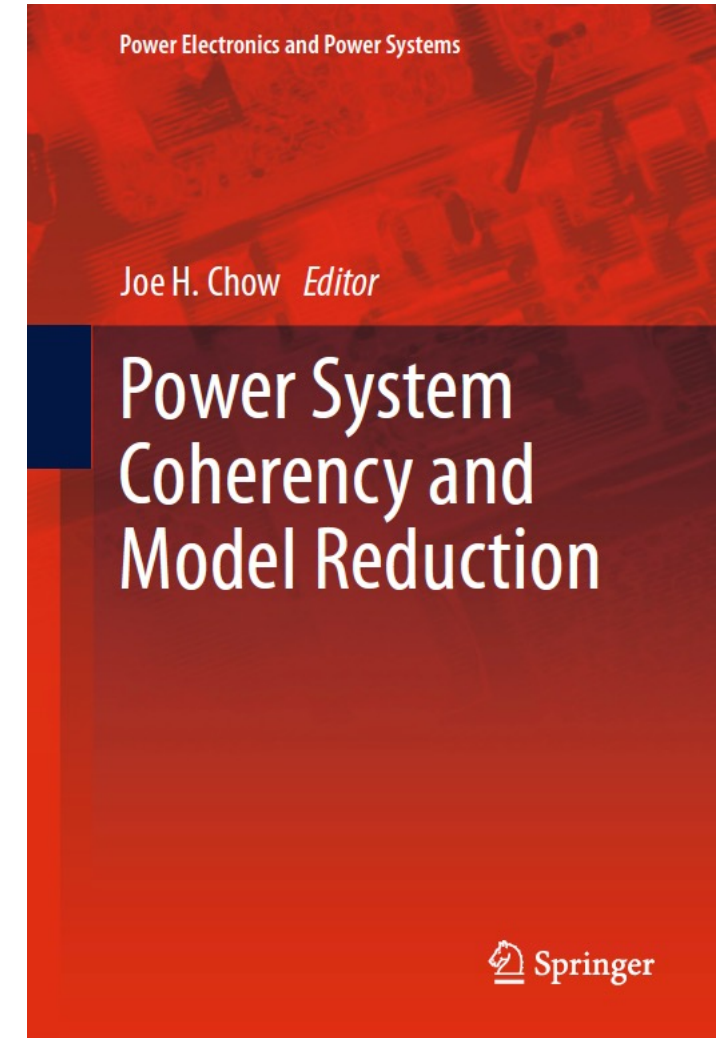
- **Merits and trade-offs of low inertia**
 - Control Perspective: Lighter systems are easier to control!
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 - Generalizes passivity notions using network information
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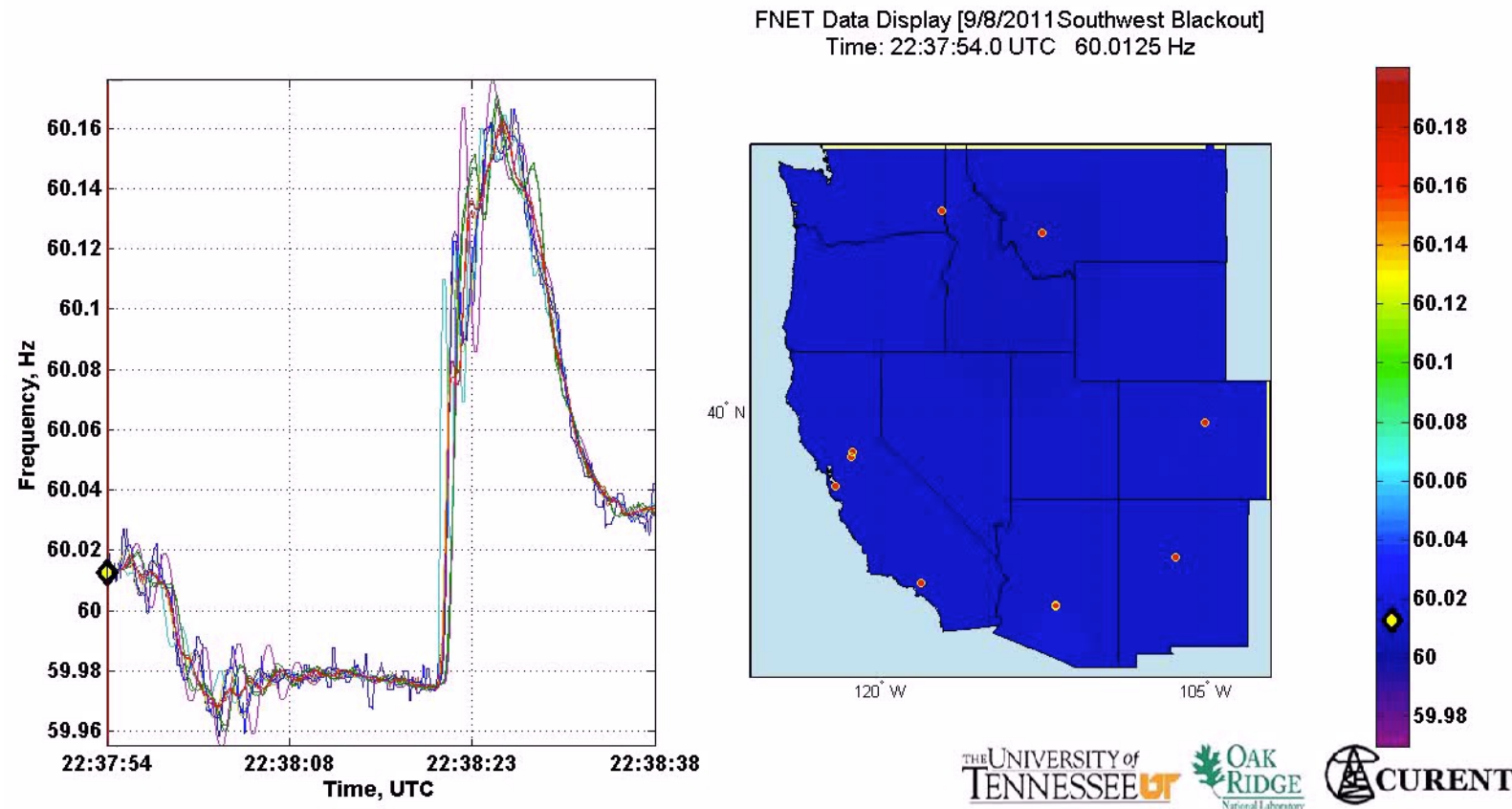
Coherence in Power Networks

- Studied since the 70s
 - Podmore, Price, Chow, Kokotovic, Verghese, Pai, Schweppe,...
- Enables aggregation/model reduction
 - Speed up transient stability analysis
- Many important questions
 - How to identify coherent modes?
 - How to accurately reduce them?
 - What is the cause?
- Many approaches
 - Timescale separations (Chow, Kokotovic,)
 - Krylov subspaces (Chaniotis, Pai '01)
 - Balanced truncation (Liu et al '09)
 - Selective Modal Analysis (Perez-Arriaga, Verghese, Schweppe '82)



Goal: Understand how IBR presence affect classical coherence studies

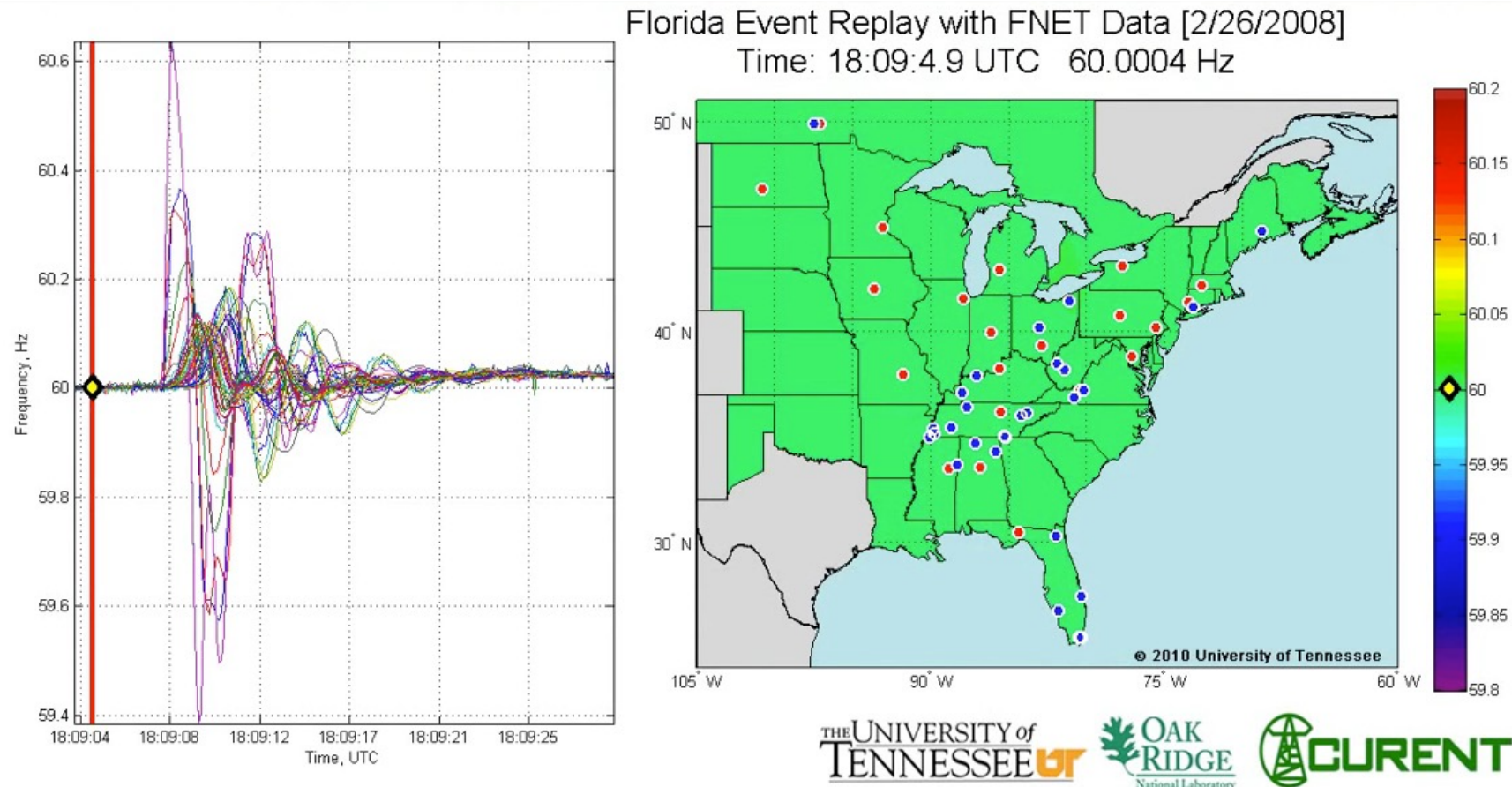
Case Study 1: Network Coherence



Key Questions:

- How does coherence emerge, and what does it depend on?
- How to characterize the coherent response in the presence of IBRs?

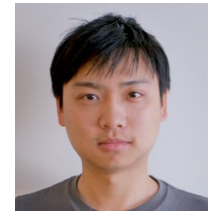
Case Study 2: Coherent Inter-area Modes



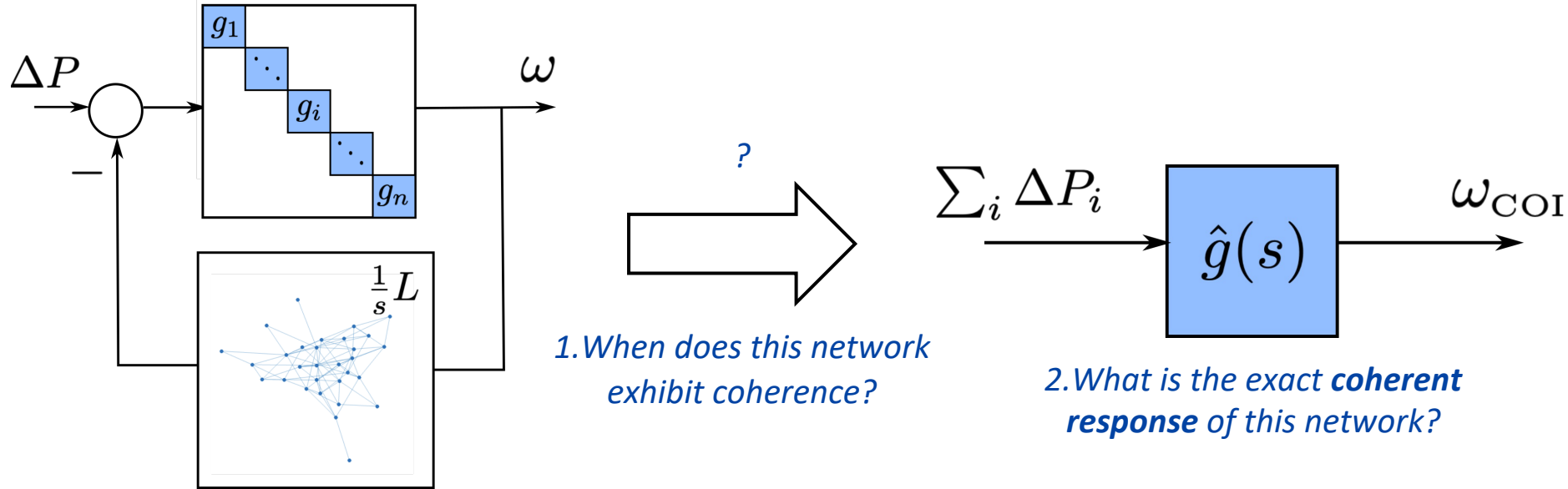
Key Questions:

- How to identify coherent areas?
- Can we model the inter-area oscillations?

Analysis of Coherent Dynamics [CDC 19, Auto 25]



Hancheng Min Richard Pates



• Problem Setup:

- Linearized power flows L_{ij}
- Bus i : arbitrary siso tf:
 $\omega_i = g_i(s) \Delta P_i$ (SGs or IBRs)

Example I: SG + Turbine

$$g_i(s) = \frac{1}{m_i s + d_i + \frac{R_i^{-1}}{\tau s + 1}}$$

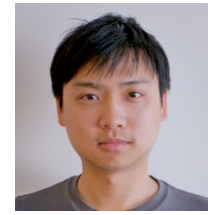
Example II: IBRs

$$g_i(s) = \frac{1}{v_i s + R_i^{-1}}$$

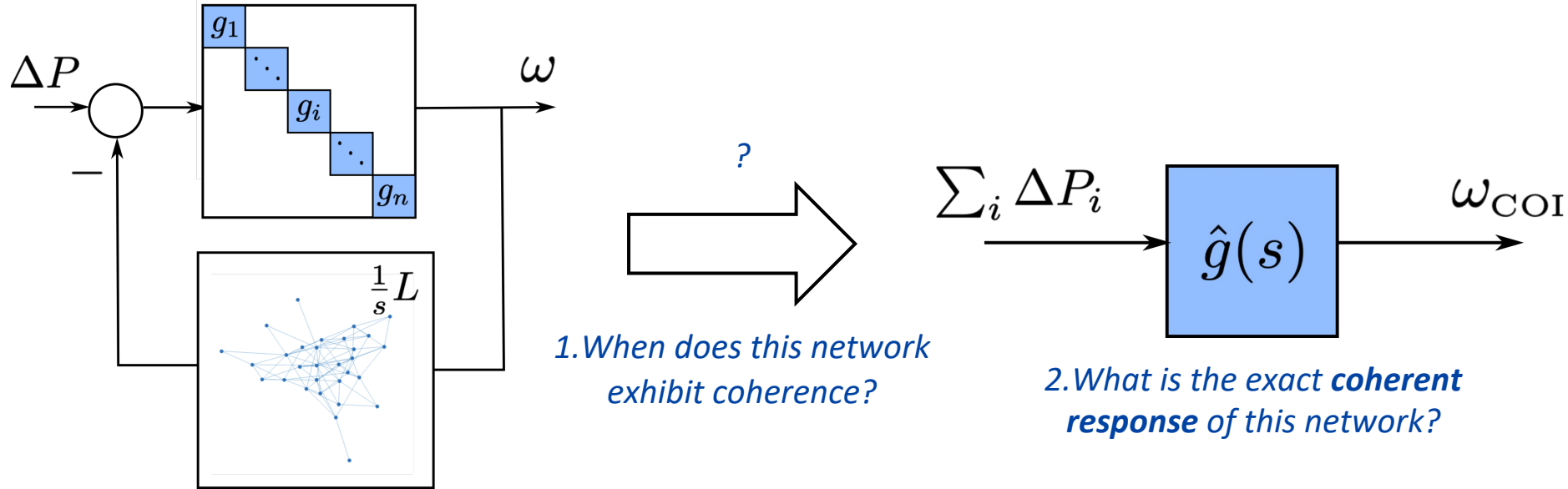
[CDC 19] Min, M. Dynamics concentration of large-scale tightly-connected networks. **Conference on Decision and Control 2019**

[Automatica 25] Min, Pates, M. A frequency domain analysis of slow coherency in networked systems. **Automatica 2025**

Analysis of Coherent Dynamics [CDC 19, Auto 25]



Hancheng Min Richard Pates



1. When does this network exhibit coherence?

2. What is the exact **coherent response** of this network?

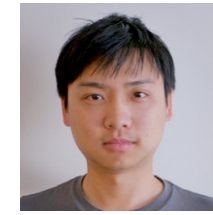
1. Coherence can be understood as a **low rank** property the **closed-loop transfer matrix**
2. It emerges as the **effective algebraic connectivity** $\left| \frac{1}{s_0} \lambda_2(L) \right|$ increases
3. The coherent dynamics is given by the **harmonic sum** of bus dynamics

$$\hat{g}(s) = \left(\sum_{i=1}^n g_i^{-1}(s) \right)^{-1}$$

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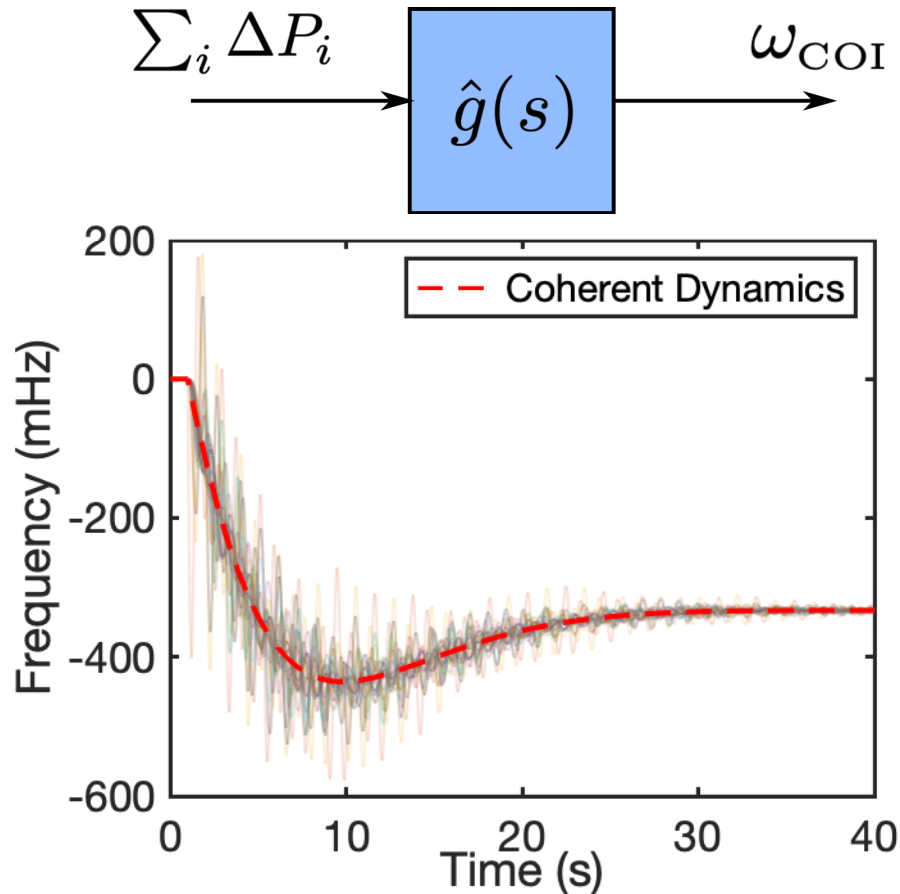
Generalized Center of Inertia [CDC 19, Auto 25]



Hancheng Min



Richard Pates



$$\hat{g}(s) = \left(\sum_{i=1}^n g_i^{-1}(s) \right)^{-1}$$

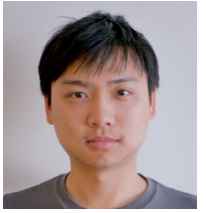
- **Coherent Dynamics: $\hat{g}(s)$**
- Representation of aggregate response
- Accuracy of approximation:
 - is frequency dependent
 - increases with network connectivity
- Provides excellent template for reduced order models (via balance-truncations)
- More details [LCSS 20]

[CDC 19] Min, M. Dynamics concentration of large-scale tightly-connected networks. **Conference on Decision and Control 2019**

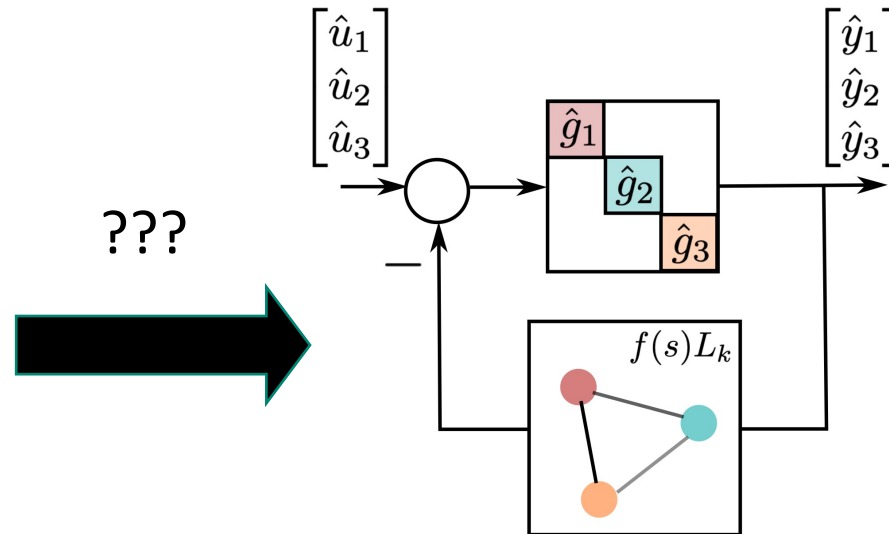
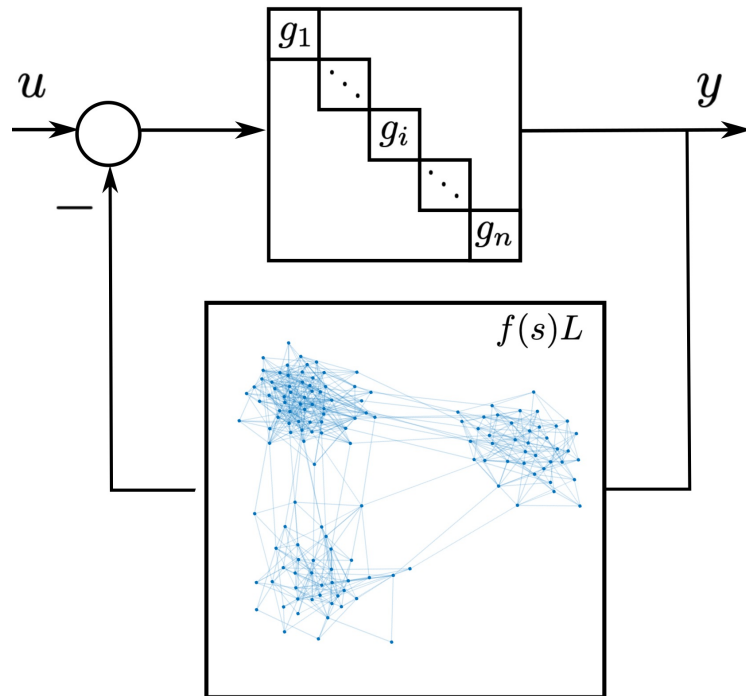
[LCSS 20] Min, Paganini, M. Accurate reduced-order models for heterogeneous coherent generators. **IEEE LCSS 2020**

[Auto 25] Min, Pates, M. A frequency domain analysis of slow coherency in networked systems. **Automatica 2025**

Weakly-Connected Coherent Networks



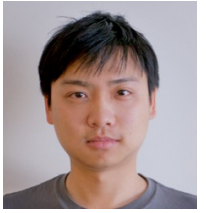
Hancheng Min



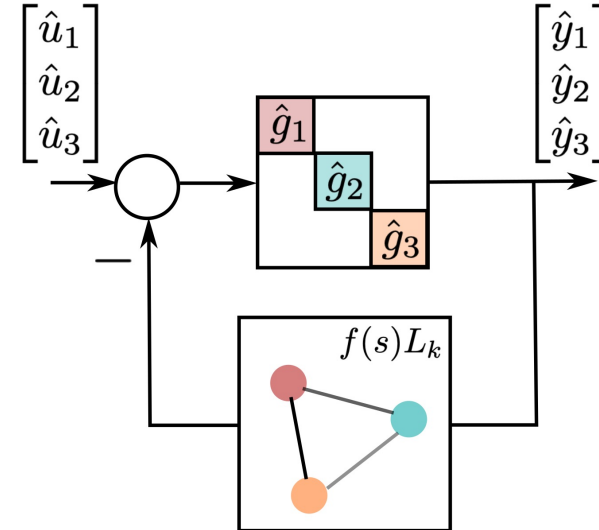
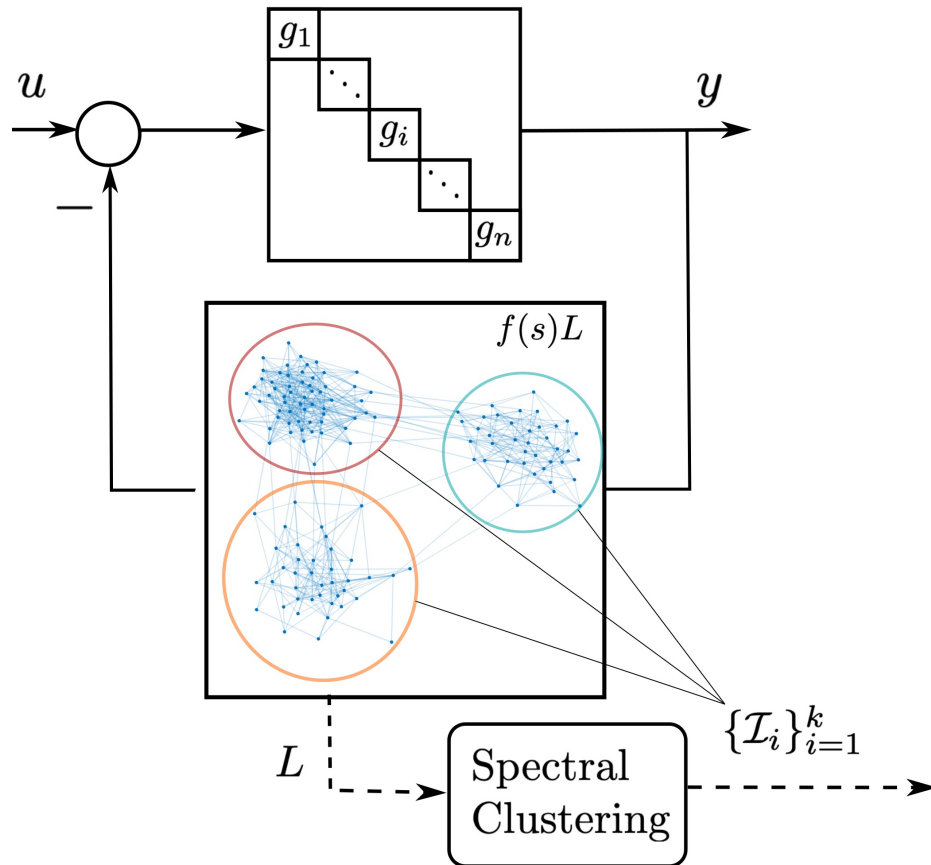
- Aggregate each coherent area
- Inter-area oscillation can be modeled as the interaction among aggregate nodes

Structure-preserving Network Reduction

Step 1: Identifying coherent areas



Hancheng Min



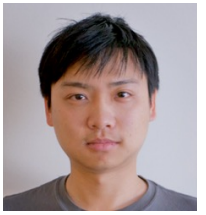
*Tightly-connected
Networks are coherent*



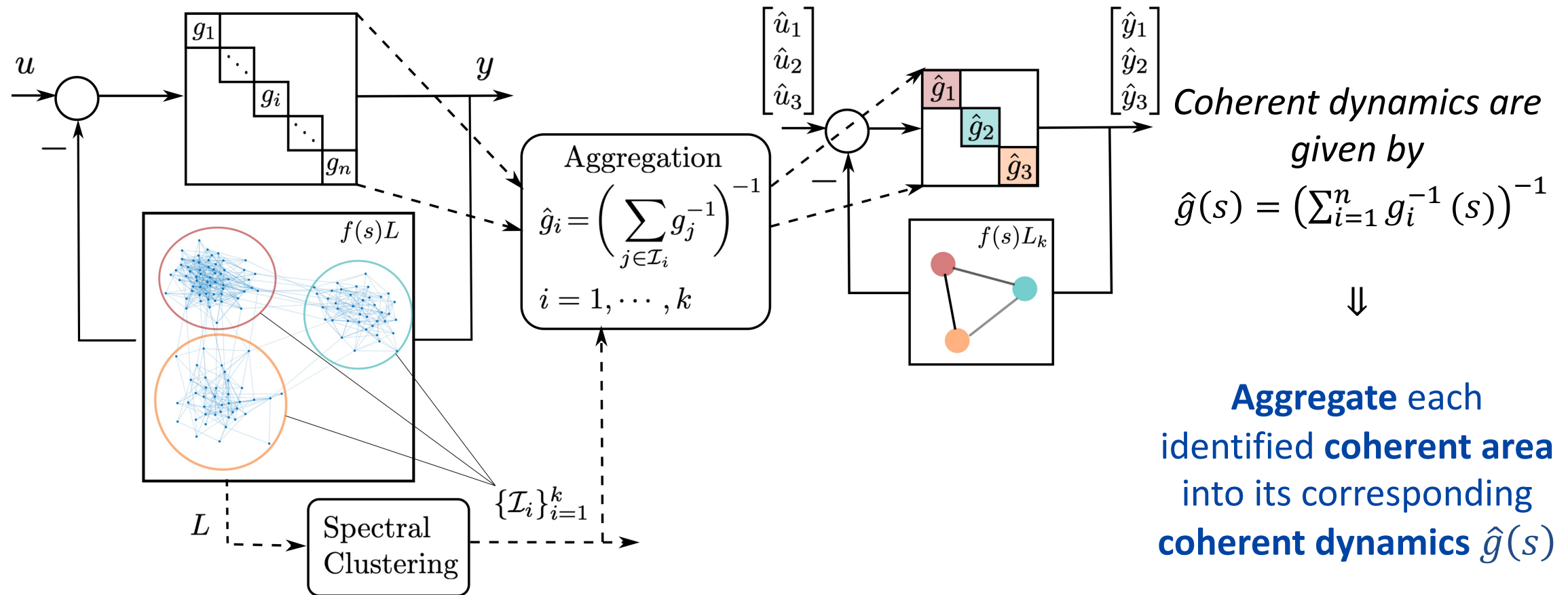
Use **spectral clustering**
algorithm to **find**
tightly-connected
subnetworks/areas

Structure-preserving Network Reduction

Step 2: Aggregate coherent areas

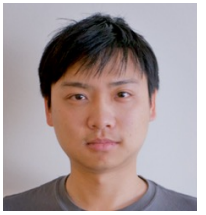


Hancheng Min

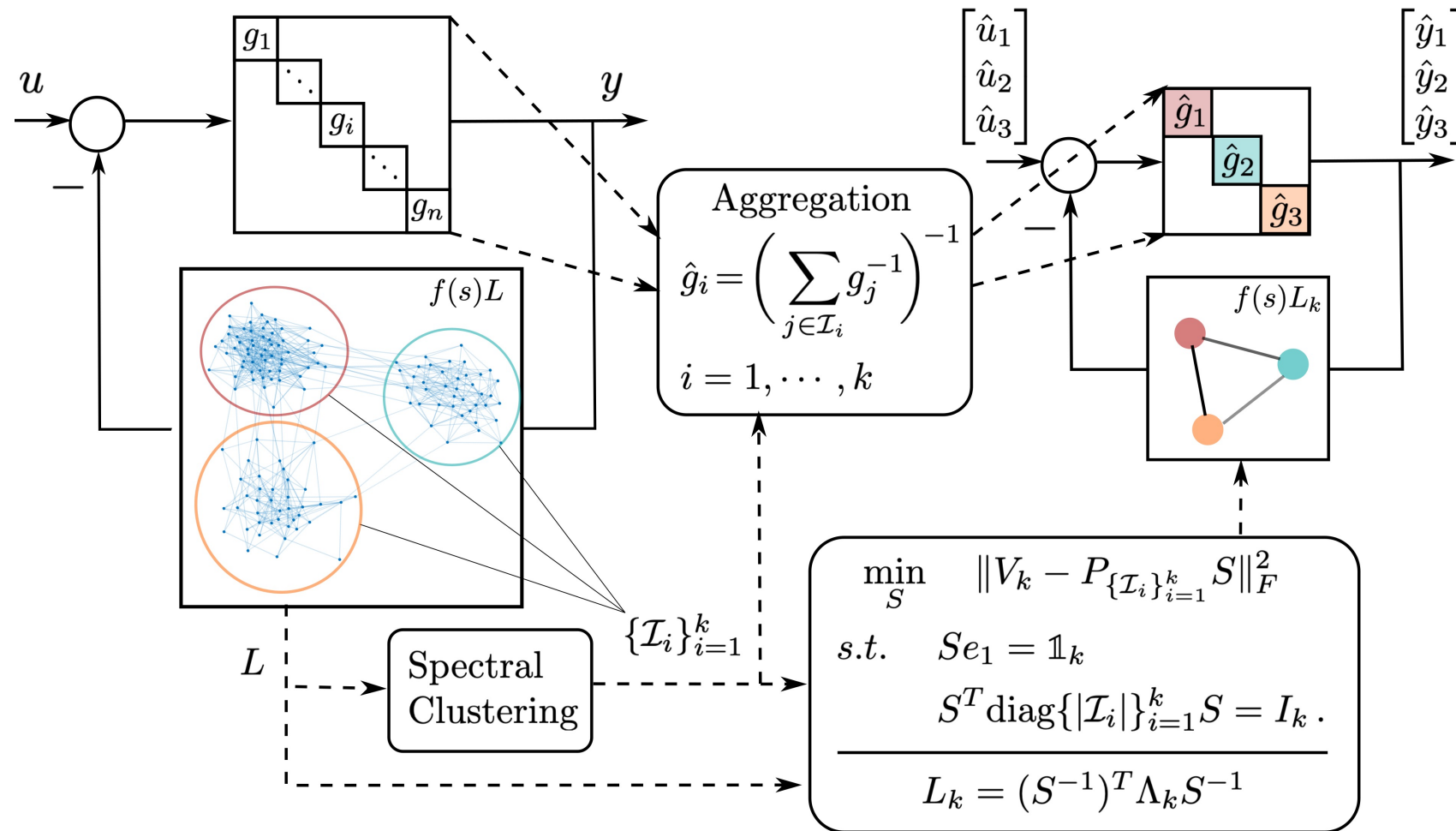


Structure-preserving Network Reduction

Step 3: Model the **interaction** among aggregate nodes

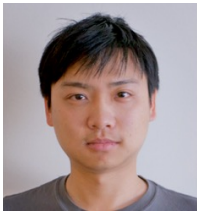


Hancheng Min



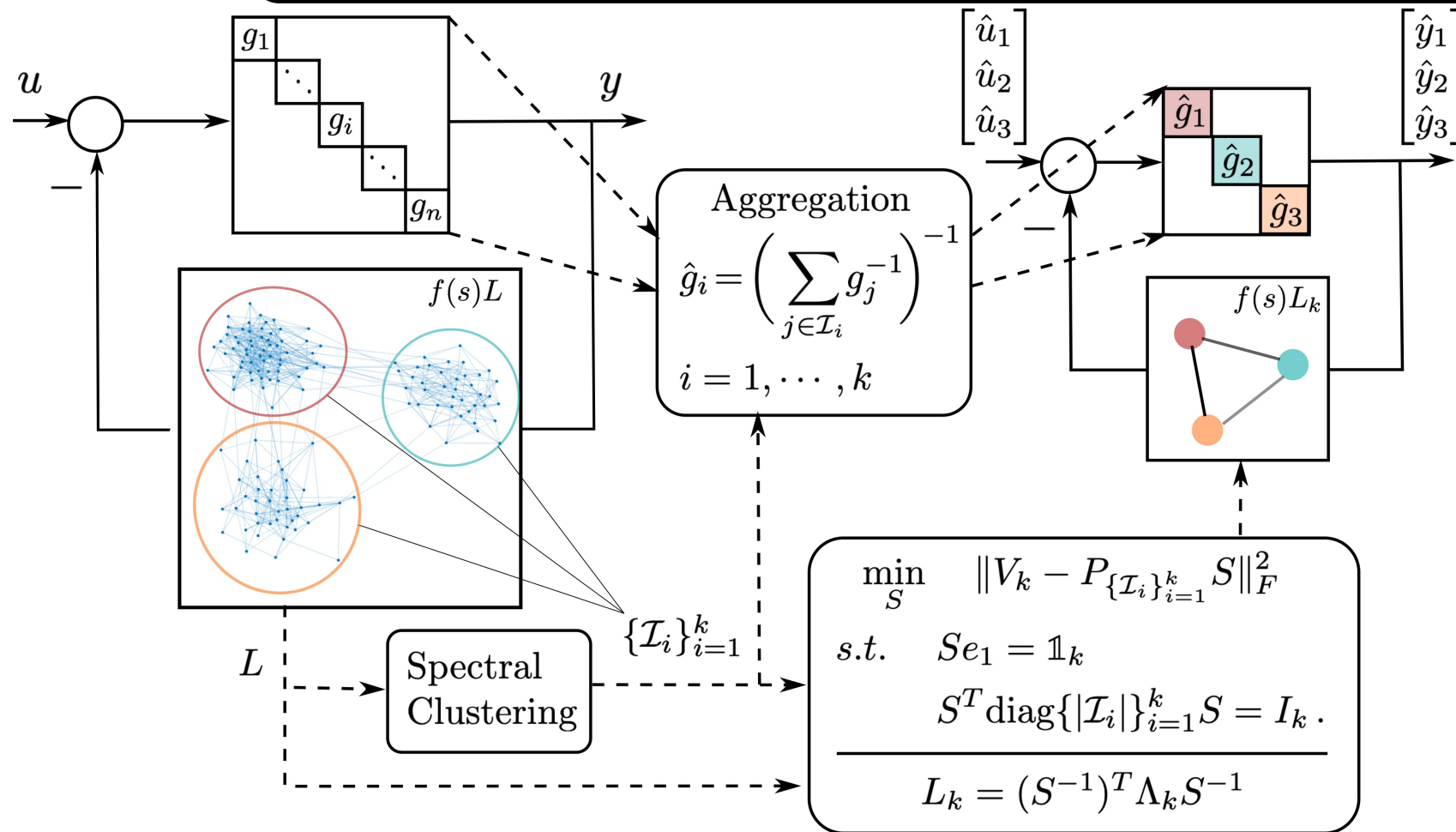
Construct the reduced network L_k by solving an optimization problem (it has closed-form solution)

Approximation Errors

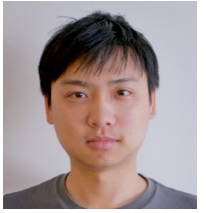


Hancheng Min

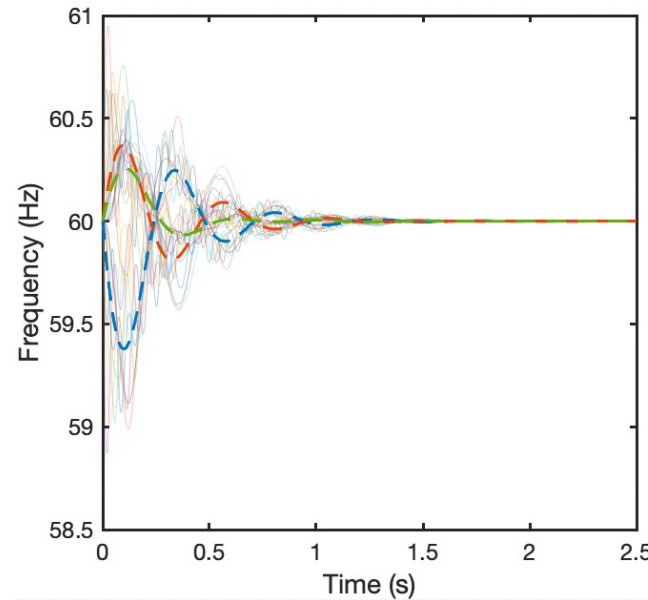
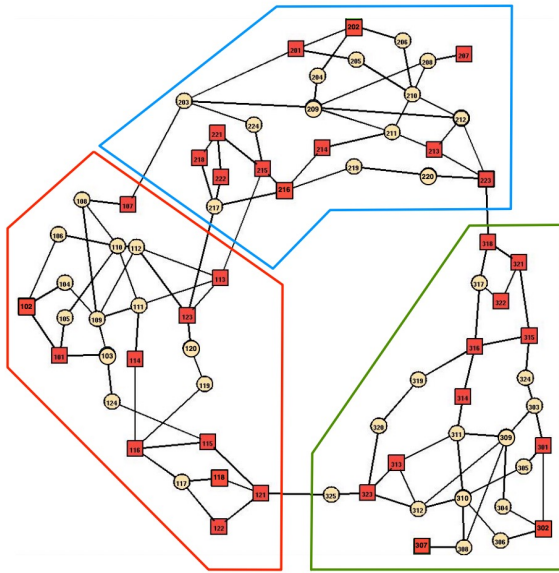
$$\|T(s_0) - \hat{T}_k(s_0)\|_2 = \mathcal{O}\left(\frac{1}{\lambda_{k+1}(L)}\right) + \mathcal{O}\left(\|V_k(L) - P_{\{I_i\}_{i=1}^k} S\|_2\right)$$



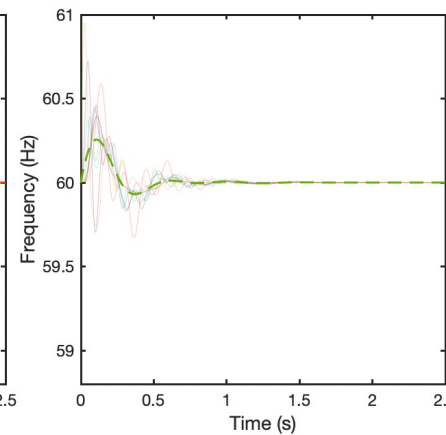
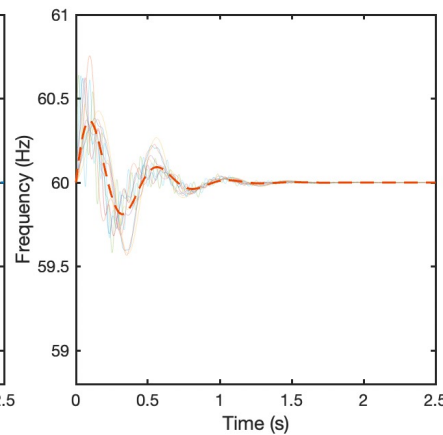
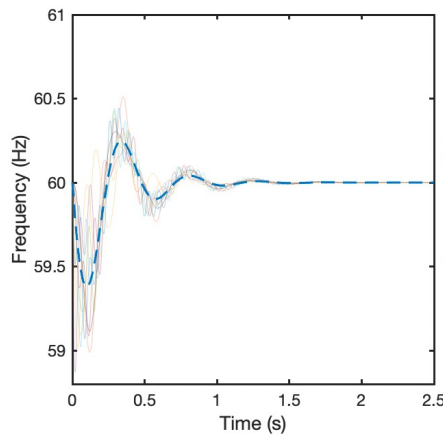
Numerical validation – RTS 96 test case



Hancheng Min



- The IEEE reliability test system: 1996
- 3 areas, 33 generators in total
- Different rotor angles across each area at initialization
- Solid lines: actual frequency response
Dashed lines: reduced model



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Grid Shaping Control

Use model matching control to shape system response

Grid-following IBRs

Grid-forming IBRs

Grid-shaping with GFL IBRs [TPS 21]



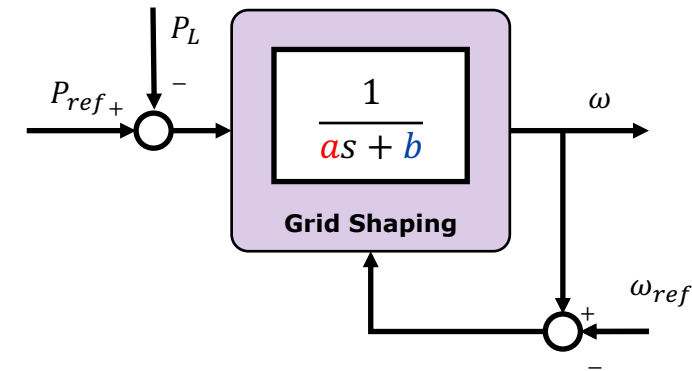
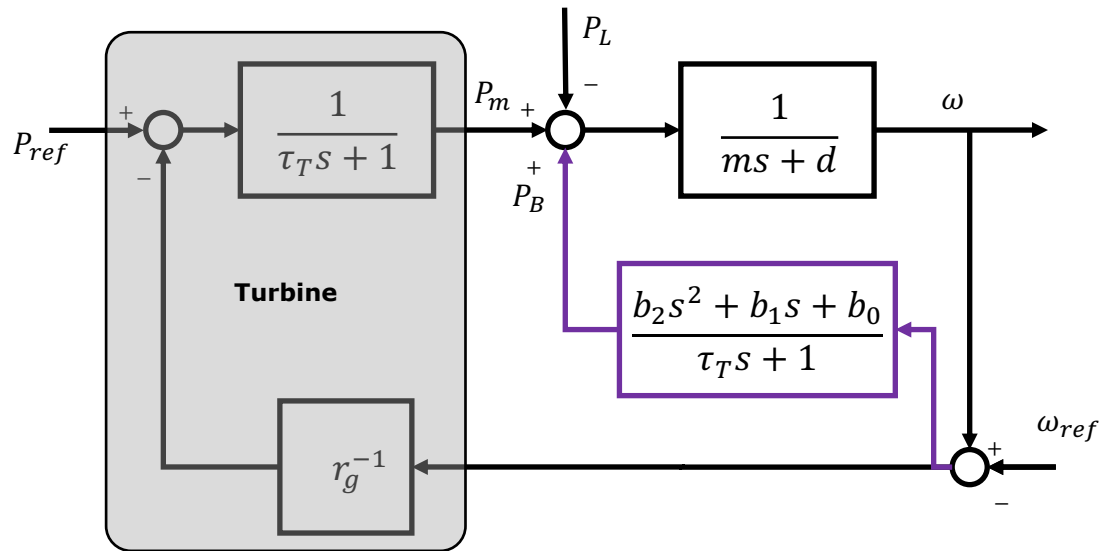
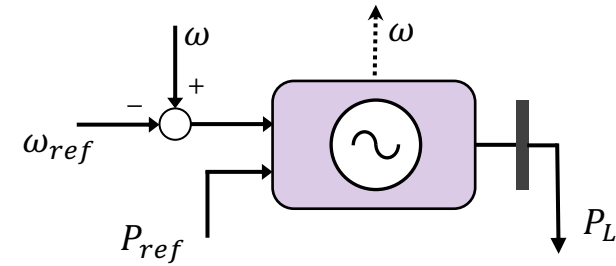
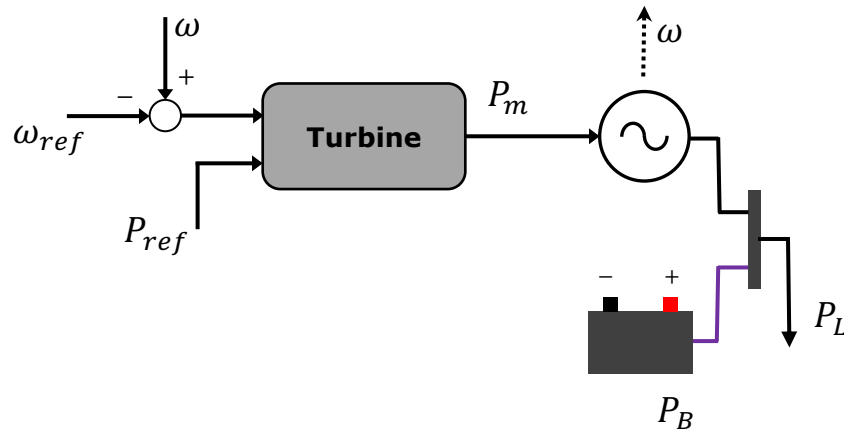
Yan Jiang



Eliza Cohn



Petr Vorobev



Tunable Performance:

$$\text{RoCoF} = \frac{1}{a} \Delta P, \quad \Delta \omega = \frac{1}{b} \Delta P$$

Grid-shaping with GFL IBRs [TPS 21]



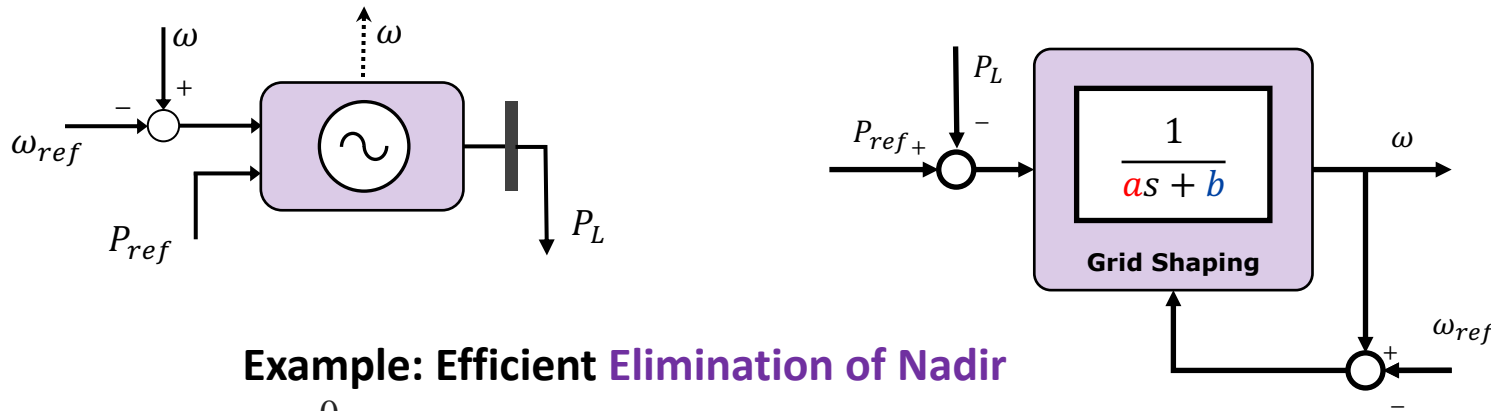
Yan Jiang



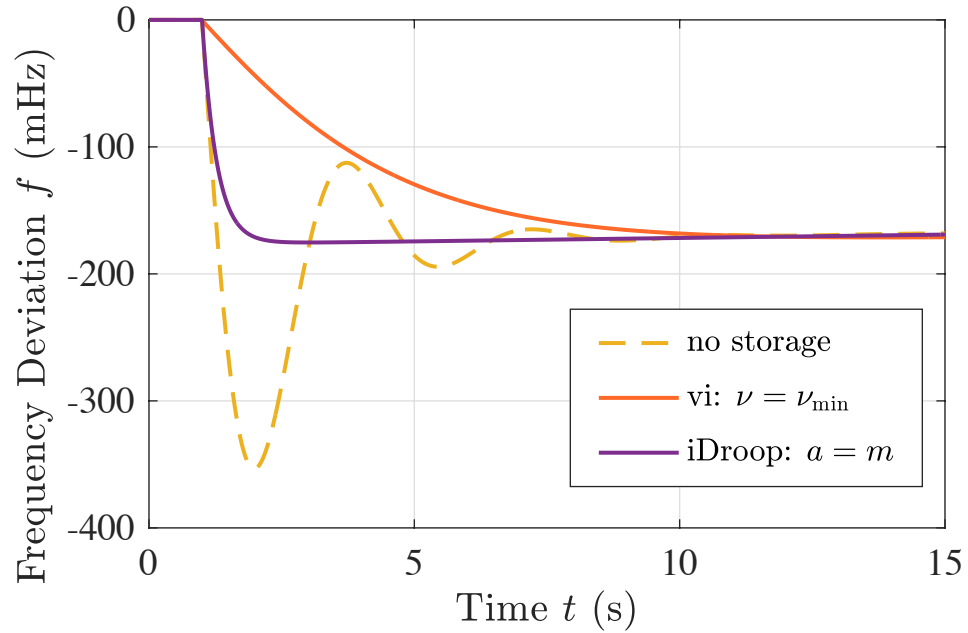
Eliza Cohn



Petr Vorobev

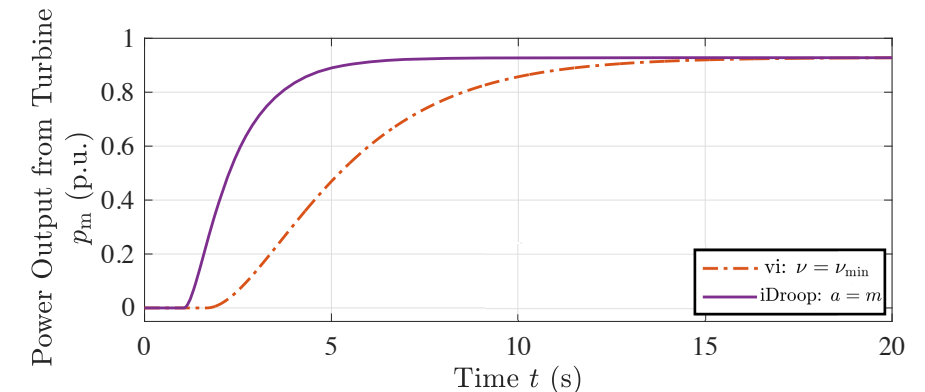
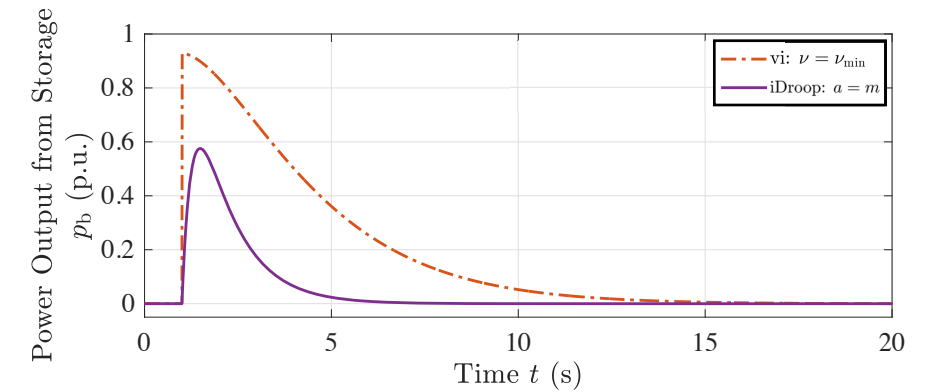


Example: Efficient Elimination of Nadir



Tunable Performance:

$$\text{RoCoF} = \frac{1}{a} \Delta P, \quad \Delta \omega = \frac{1}{b} \Delta P$$



Grid-shaping with GFL IBRs [TPS 21]



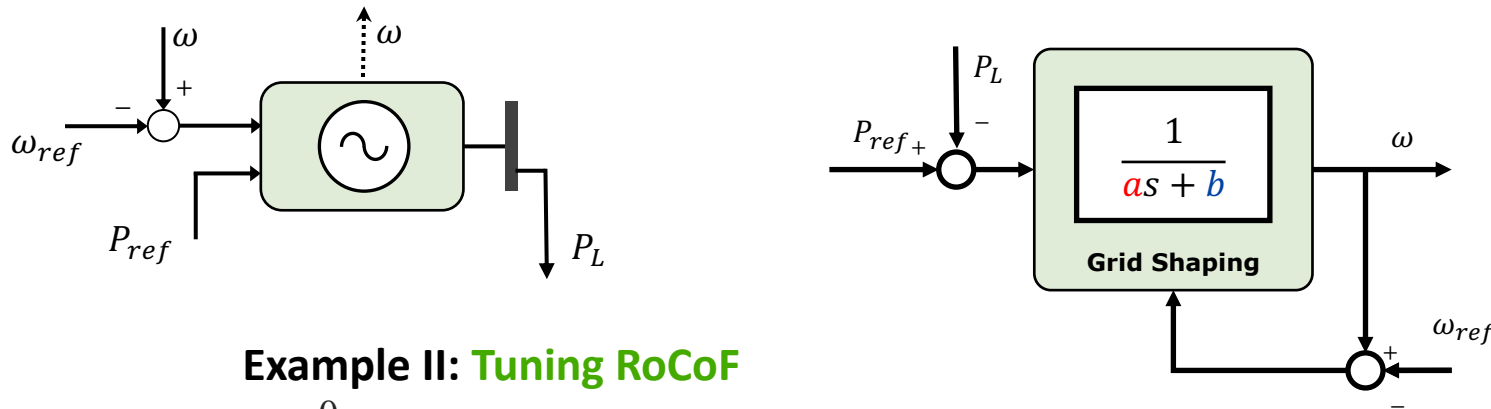
Yan Jiang



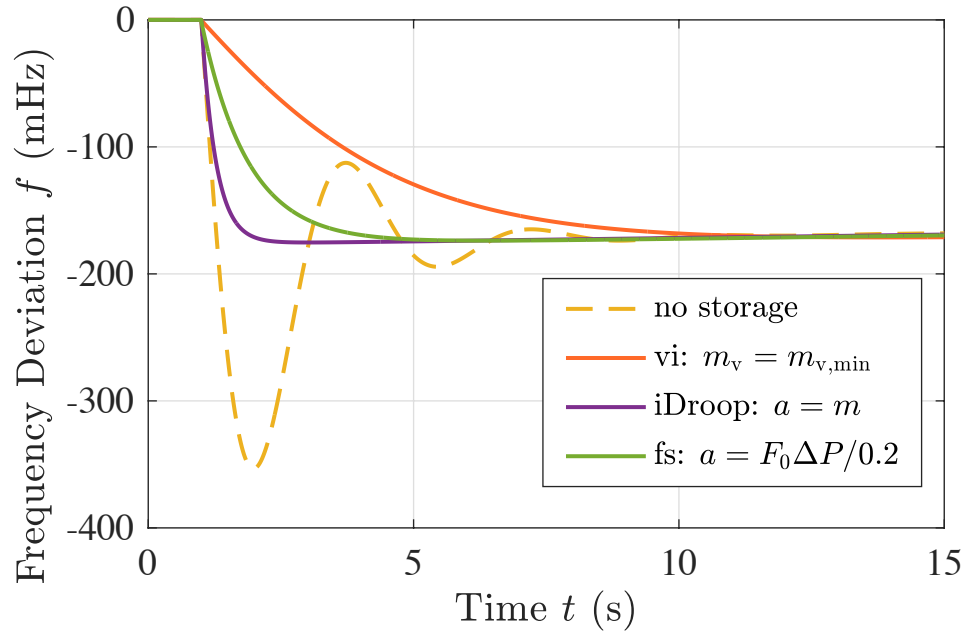
Eliza Cohn



Petr Vorobev



Example II: Tuning RoCoF



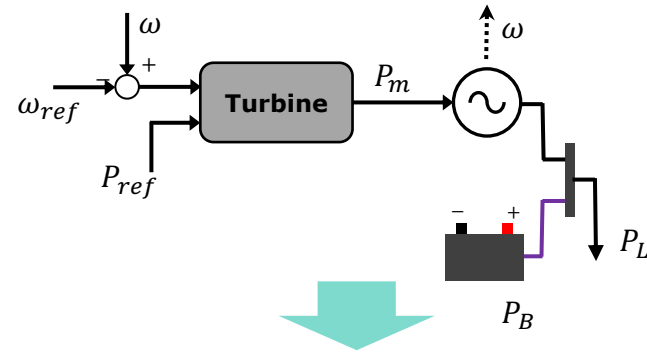
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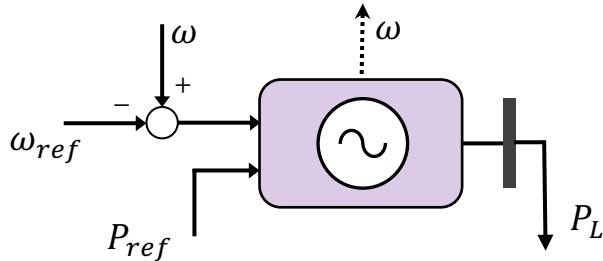
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Use model matching control to shape system response

Grid-following IBRs



Grid-forming IBRs



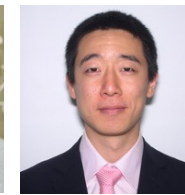
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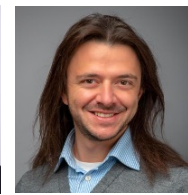
GFM Grid-shaping Through Lines [LCSS 23]



B. K. Poolla



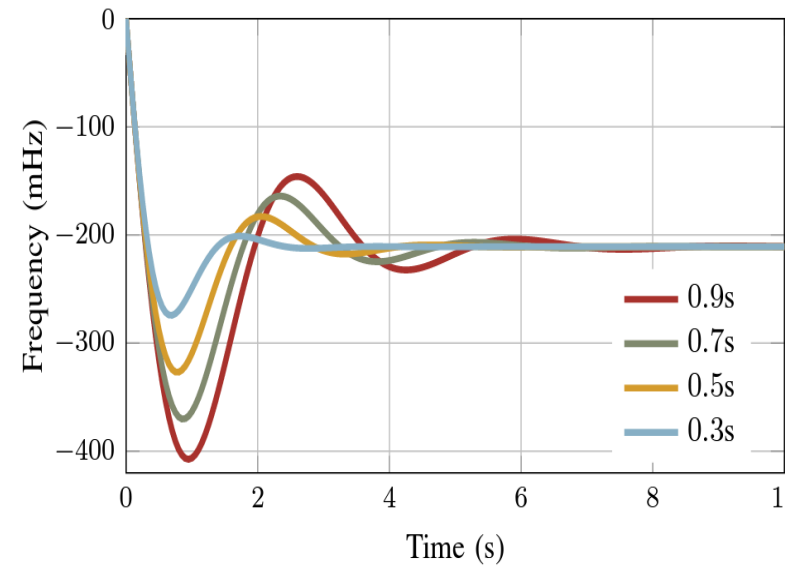
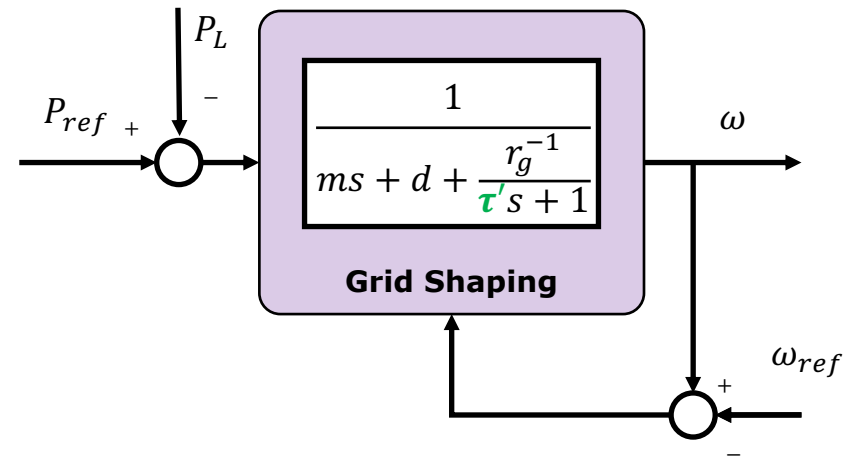
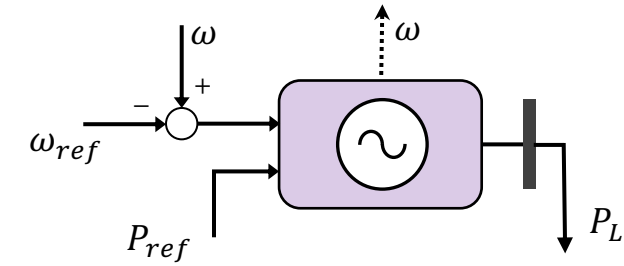
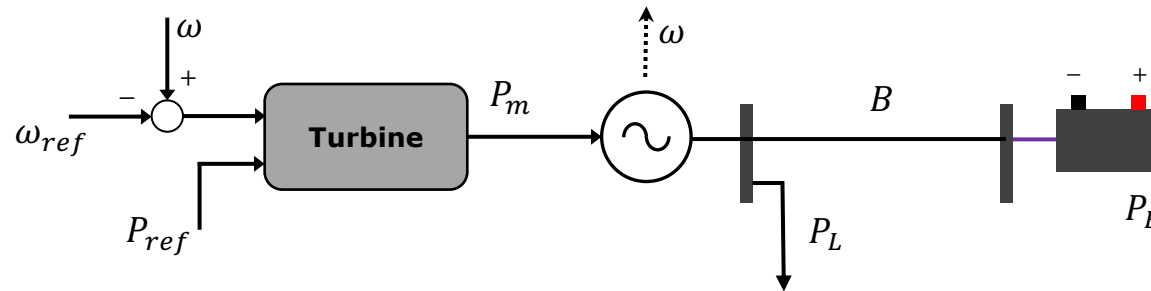
Y. Lin



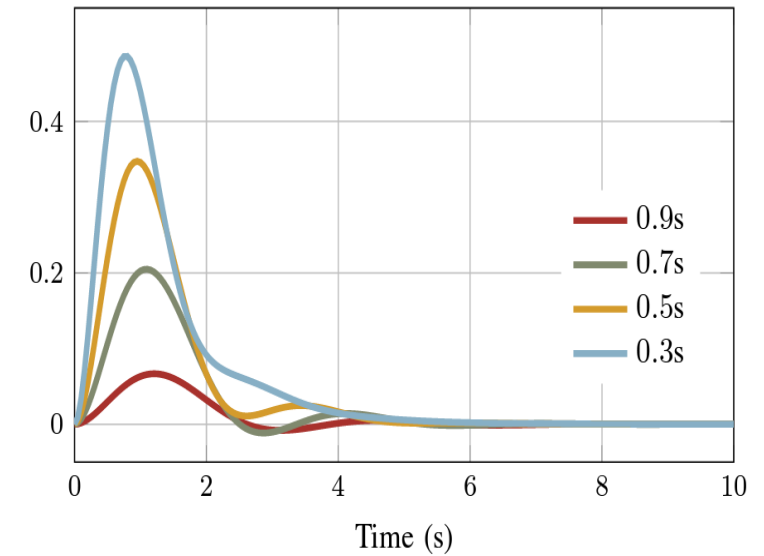
A. Bernstein



D. Groß



Frequency response for a 1 p.u. load step



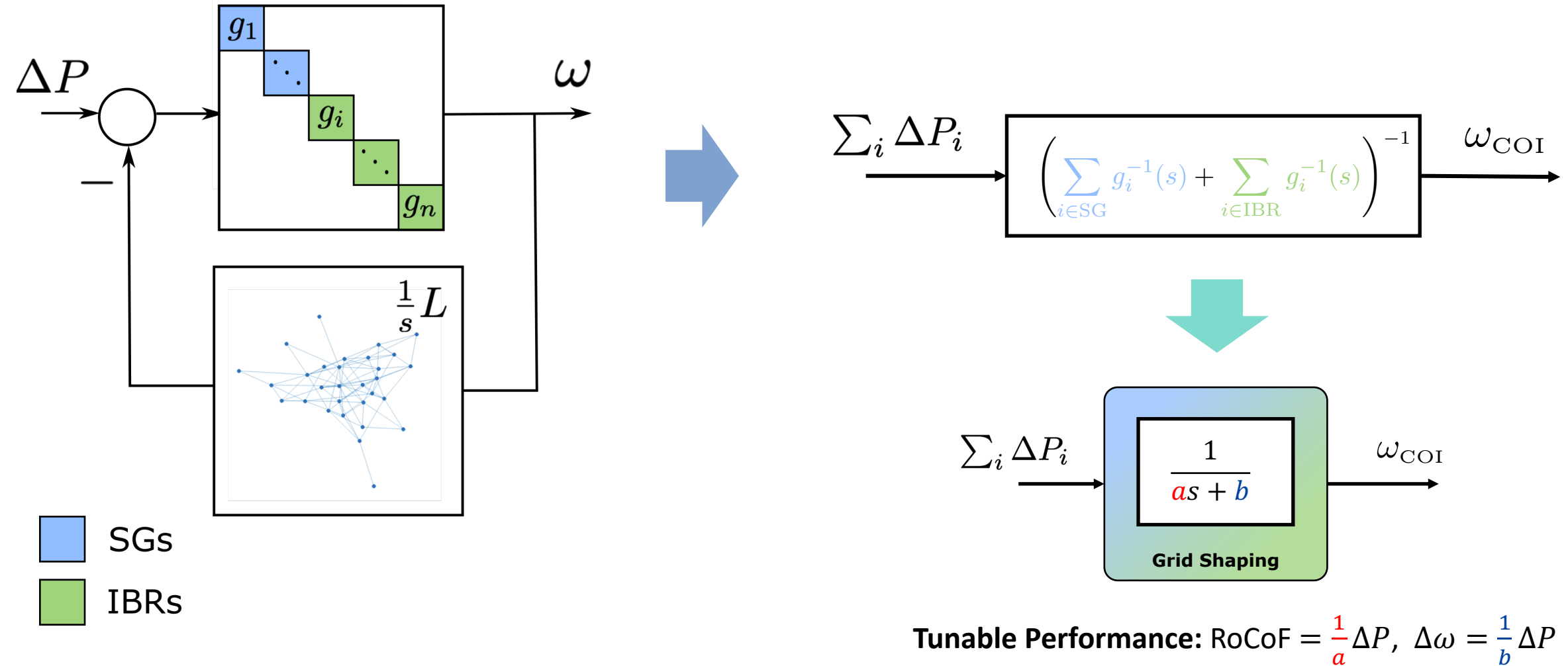
IBR power injection for a 1 p.u. load step

Tunable Performance:

E.g.: Turbine Time Constant = τ'

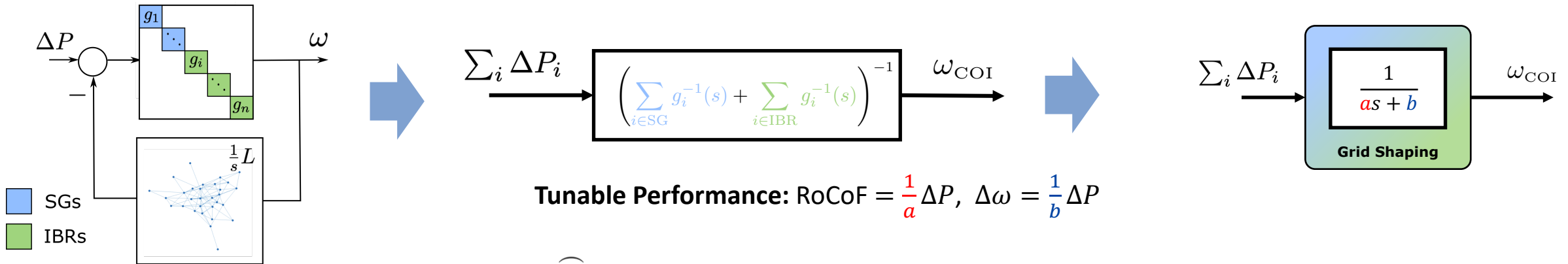


GFM System-wide Grid-shaping [LCSS 20]

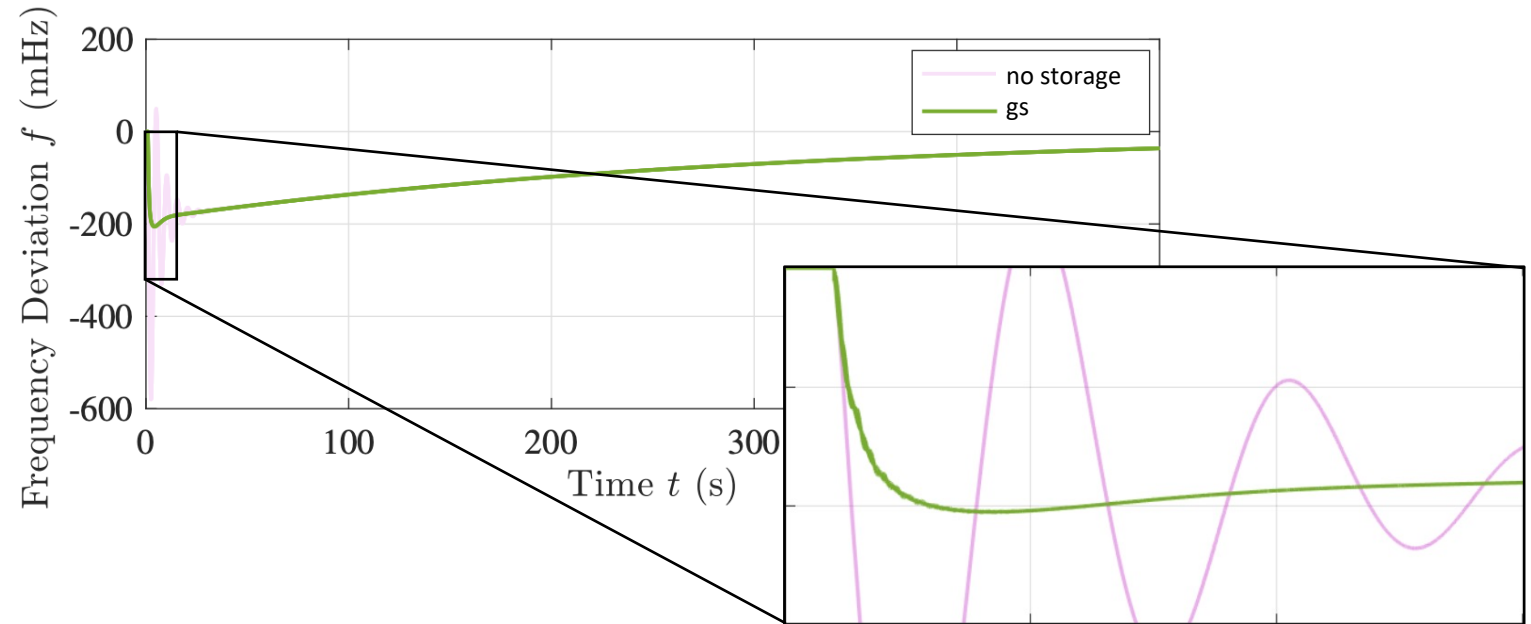
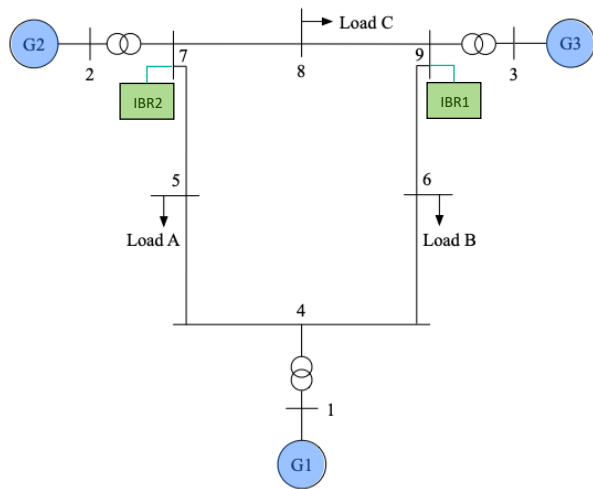




GFM System-wide Grid-shaping [LCSS 20]



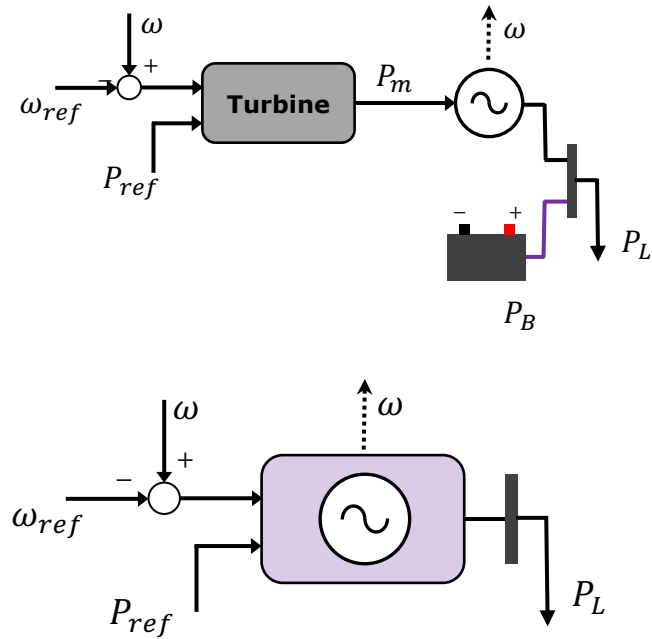
Tunable Performance: $\text{RoCoF} = \frac{1}{a} \Delta P$, $\Delta \omega = \frac{1}{b} \Delta P$



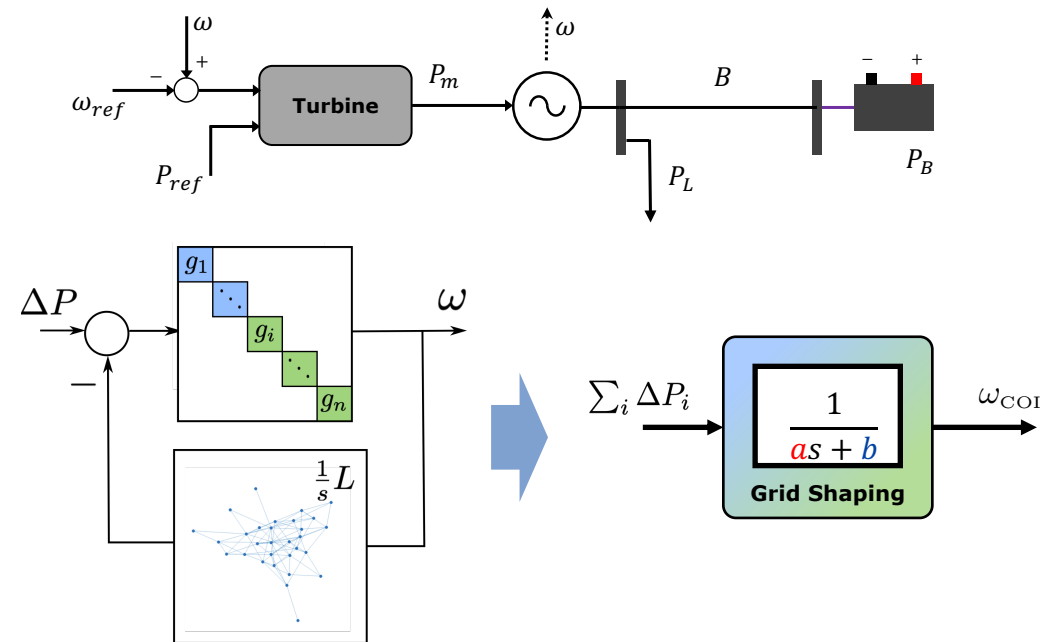
Grid Shaping Control

Use model matching control to shape system response

Grid-following IBRs



Grid-forming IBRs



Tunable Performance: $\text{RoCoF} = \frac{1}{a} \Delta P$, $\Delta\omega = \frac{1}{b} \Delta P$, τ' , ...

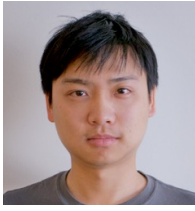
Summary

- **Merits and trade-offs of low inertia**
 - Control Perspective: Lighter systems are easier to control!
 - Smarter controller can provide multiple benefits in Nadir, RoCoF, inter-area oscillations, and disturbance rejection, with less effort
- **Scale-free Stability Analysis of Grids**
 - Generalizes passivity notions using network information
 - Decentralized test based on local models
 - Compatible with H_∞ -synthesis methods
- **Analysis of Weakly-Connected Coherent Networks**
 - Generalized Center of Inertia captures IBR dynamics
 - Provide a new tunable target to meet system specs
 - Coherent modes identified via spectral clustering
- **Grid Shaping Control**
 - Grid-following/forming control framework for future grids
 - Leverages IBRs to *shape* the coherent response

Thanks!



Yan Jiang



Hancheng Min



Eliza Cohn



Petr Vorobev



Richard Pates



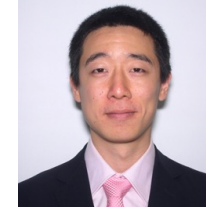
Fernando Paganini



Dominic Groß



Bala K. Poolla



Yashen Lin



Andrey Bernstein

Merits and trade-offs of low inertia

[TAC 21] Jiang, Pates, M, Dynamic droop control in low inertia power systems. **Transactions on Automatic Control**, 2021

Analysis of Weakly-Connected Coherent Networks

[CDC 19] Min, M. Dynamics concentration of large-scale tightly-connected networks. **Conference on Decision and Control 2019**

[LCSS 20] Min, Paganini, M. Accurate reduced-order models for heterogeneous coherent generators. **IEEE Control Systems Letters** 2020

[L4DC 23] Min, M. Learning coherent clusters in weakly-connected network systems. **Learning for Dynamics and Control** 2023

[Auto 25] Min, Pates, M. A frequency domain analysis of slow coherency in networked systems. **Automatica** 2025

Scale-free Stability Analysis

[TCNS 19] Pates, M, Robust scale-free synthesis for frequency control in power systems. **Transactions on Control of Network Systems**, 2019

[GM 24] Siahaan, M, Geng. Decentralized Stability Criteria for Grid-Forming Control in Inverter-Based Power Systems. **IEEE PES GM** 2024

Grid Shaping Control

[LCSS 20] Jiang, Bernstein, Vorobev, M. Grid-forming frequency shaping control for low-inertia power systems. **Control Systems Letters** 2020

[TPS 21] Jiang, Cohn, Vorobev, M. Storage-based frequency shaping control. **Transactions on Power Systems** 2021

[LCSS 23] Poolla, Lin, Bernstein, M, Groß. Frequency shaping control for weakly-coupled grid-forming IBRs. **IEEE Control Systems Letters** 2023

Backup Slides

Network Coherence: Heterogeneous Case

$$T(s) = \frac{1}{n} \bar{g}(s) \mathbb{1} \mathbb{1}^T + \boxed{T(s) - \frac{1}{n} \bar{g}(s) \mathbb{1} \mathbb{1}^T} \quad \bar{g}(s) = \left(\frac{1}{n} \sum_{i=1}^n g_i^{-1}(s) \right)^{-1}$$

The effect of **non-coherent dynamics** vanishes as:

- For almost any $s_0 \in \mathbb{C}$

$$\lim_{\lambda_2(L) \rightarrow +\infty} \left\| T(s_0) - \frac{1}{n} \bar{g}(s_0) \mathbb{1} \mathbb{1}^T \right\| = 0$$

- For $s_0 \in \mathbb{C}$, a pole of $f(s)$

$$\lim_{s \rightarrow s_0} \left\| T(s) - \frac{1}{n} \bar{g}(s) \mathbb{1} \mathbb{1}^T \right\| = 0$$

- Excluding zeros: the limit holds at zero, but by different convergence result
- We can further prove **uniform convergence** over a compact subset of complex plane, if it doesn't contain any zero nor pole of $\bar{g}(s)$
- Extensions for random network ensembles, $g_i(s) := g(s, w_i)$ (w_i random), then $\bar{g}(s) = (E_w[g^{-1}(s, w)])^{-1}$
- Convergence of transfer matrix is **related to time-domain response** by Inverse Laplace Transform

Connection to Time Domain

If $\bar{g}(s)$ and $T(s)$ stable ($\|\bar{g}\|_\infty, \|T\|_\infty \leq \gamma$), then there is $\bar{\lambda} = O(\gamma/\varepsilon)$ such that:

- **ε -approximation**, for any network L , with $\lambda_2(L) \geq \bar{\lambda}$

$$\sup_{t>0} |y_i(t) - \bar{y}(t)| \leq \varepsilon$$

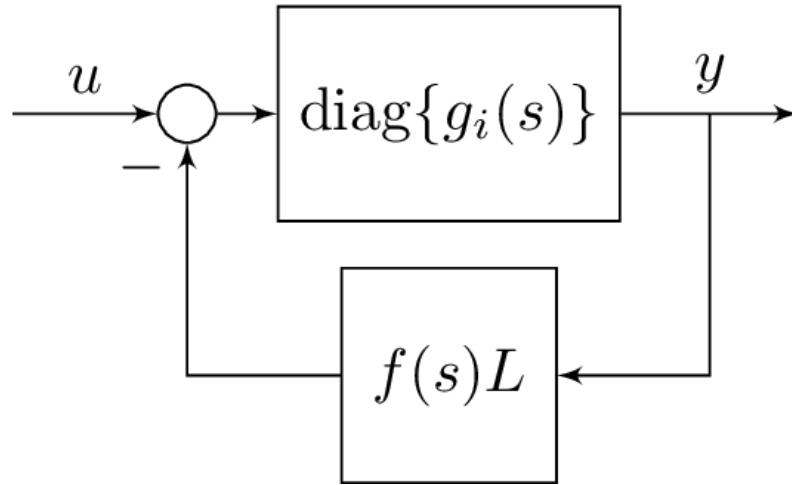
where $\bar{y}(t)$ is the coherence dynamics response: $y(s) = \bar{g}(s) \frac{1}{n} \sum_{i=1}^n u_i(s)$

- **element-wise coherence**, for any pair of nodes i and j

$$\sup_{t>0} |y_i(t) - y_j(t)| \leq 2\varepsilon$$

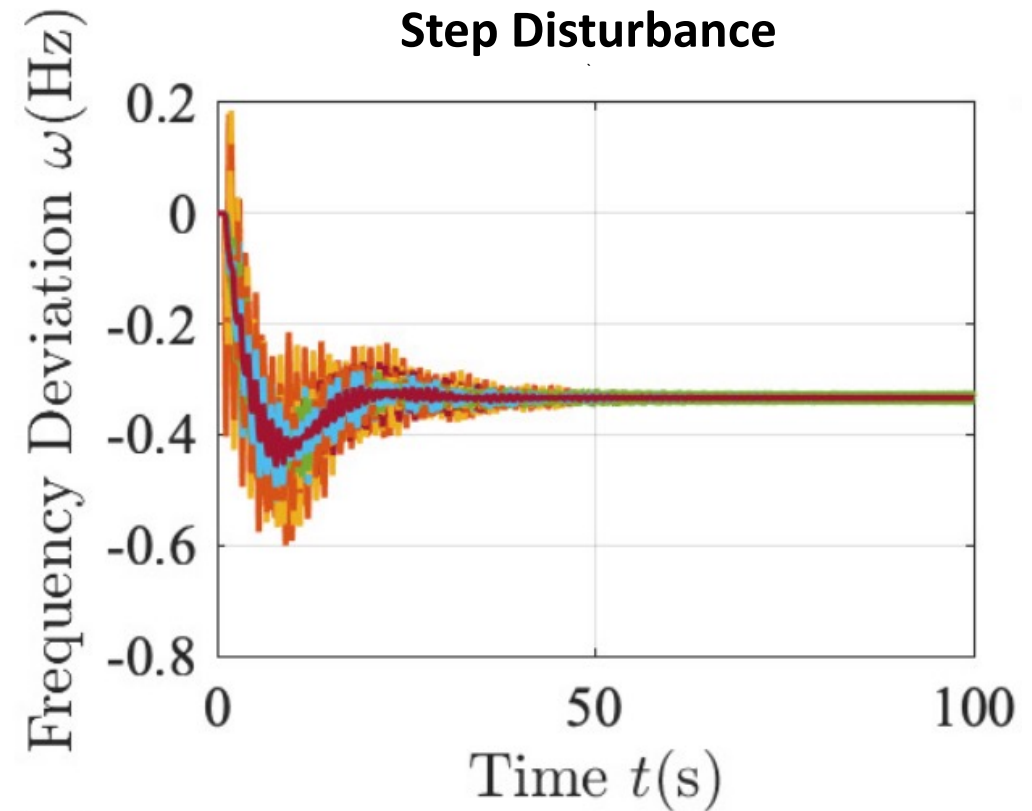
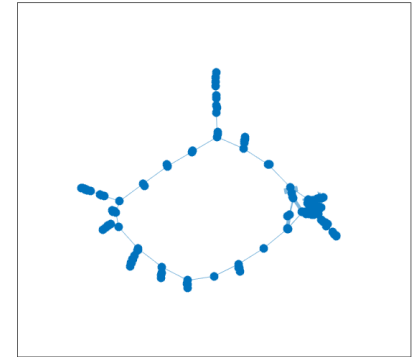
Example: Icelandic Power Grid

- Iceland power network: 189 buses, 35 generators, load 1.3GW (PSAT)

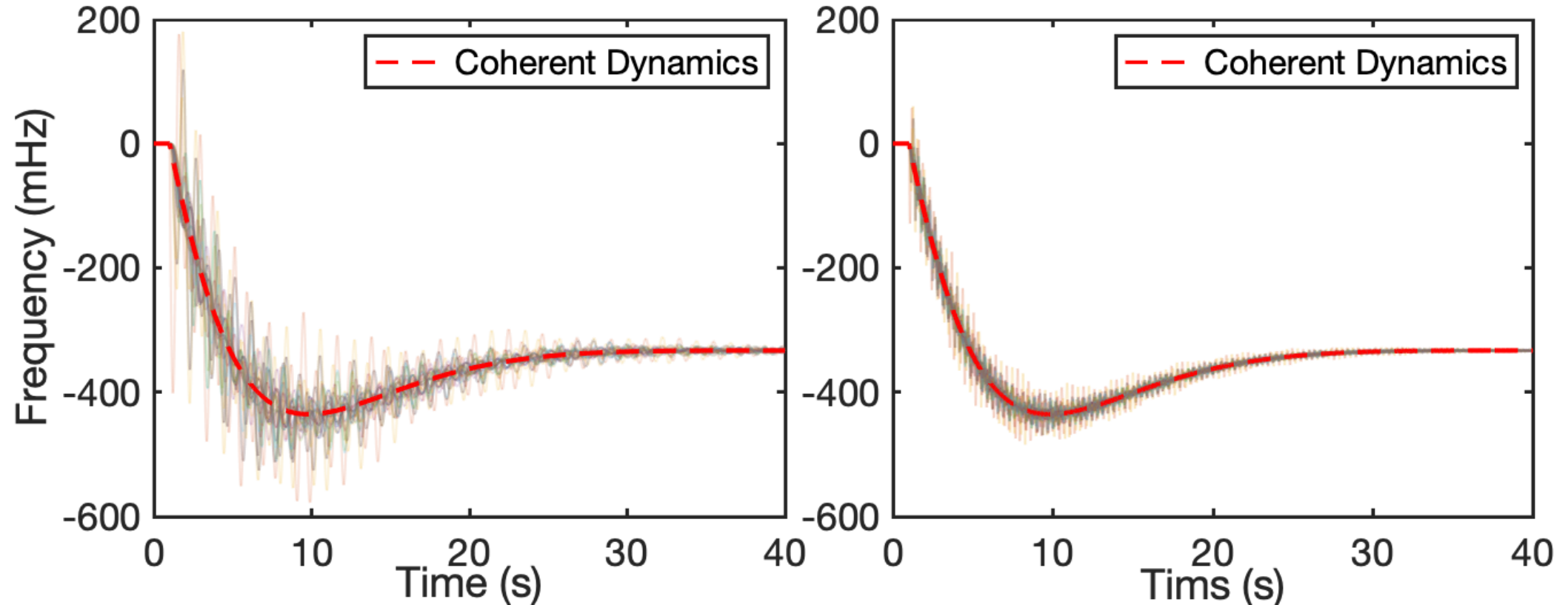


$$g_i(s) = \frac{1}{m_i s + d_i + \frac{r_i^{-1}}{\tau_i s + 1}}$$
$$f(s) = \frac{1}{s}$$

Icelandic Grid



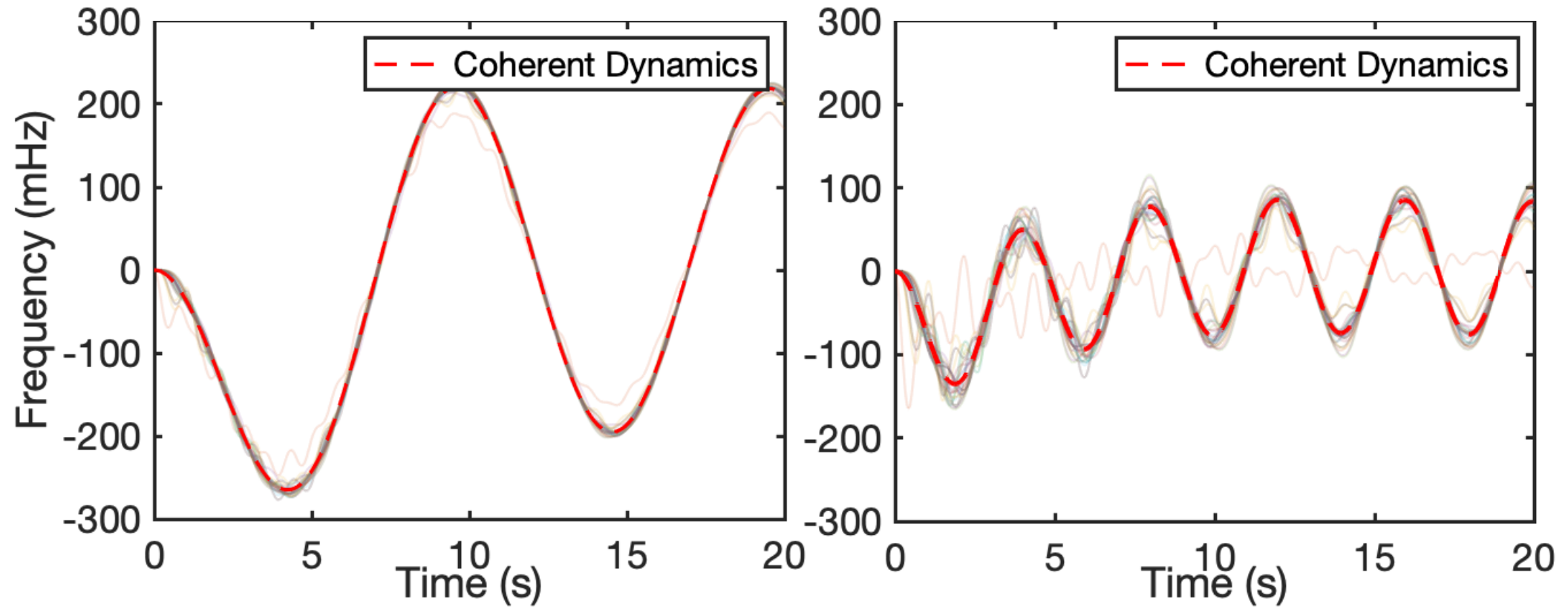
Example: Effect of Network Algebraic Connectivity $\lambda_2(L) \uparrow$



Coherent dynamics acts as a more accurate version of the Center of Inertia (CoI)

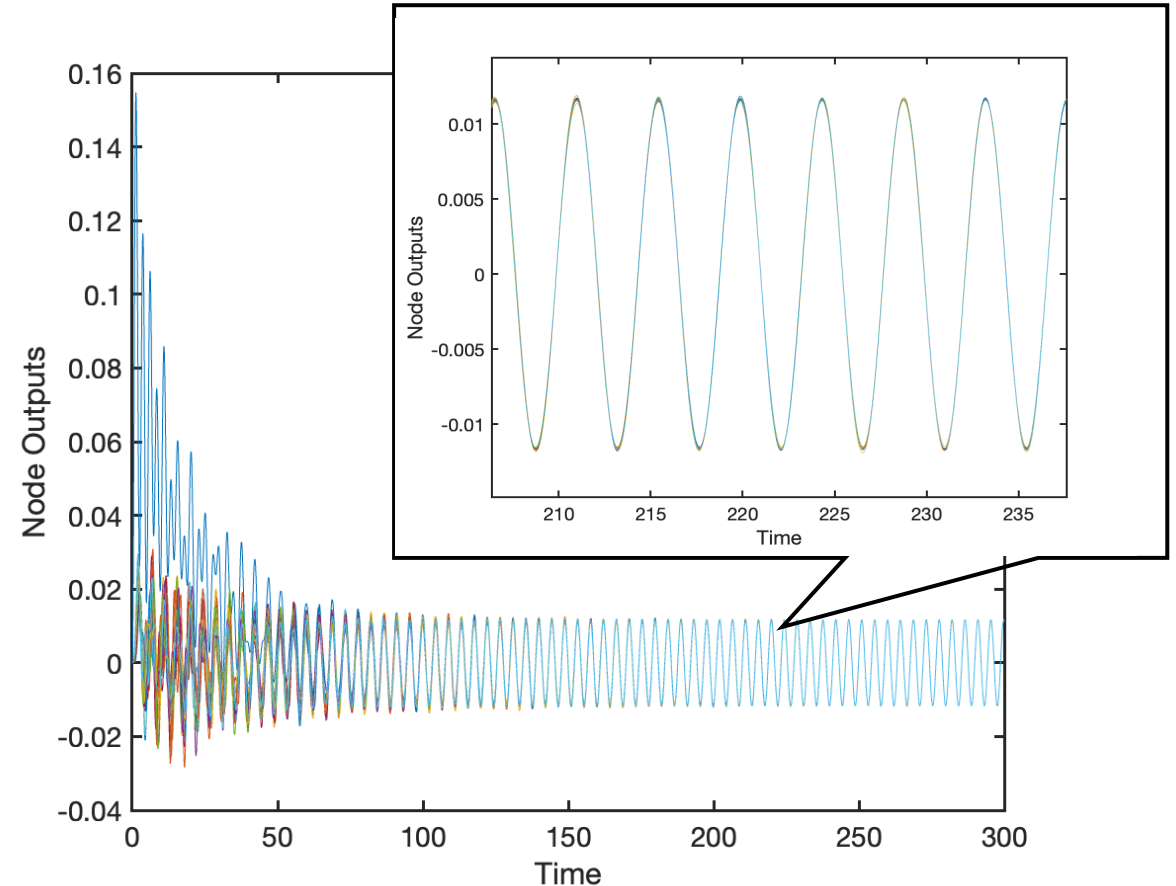
Example: Sinusoidal Disturbances: $\sin(\omega_d t)$

$\omega_d \uparrow$



Frequency-dependent Coherence from Coupling Dynamics

- Frequency dependent coherence:
A stable network responds coherently when subject to signal with its frequency component concentrated around pole of $f(s)$
- An Artificial Example:
A stable heterogeneous network with $f(s) = \frac{s}{s^2 + \omega_0^2}$ is “synchronized” by external sinusoidal input $\sin \omega_0 t$
(Such coherence is robust to small changes in input frequency)



First order nodal dynamics $g_i(s) = \frac{1}{m_i s + d_i}$
20 nodes with $m_i \sim \text{Unif}(1, 5)$, $d_i \sim \text{Unif}(0.1, 0.5)$
12-regular graph with unit weights
Sin input to the first node (shown in blue) only