

# Coherence and Concentration in Tightly-Connected Networks

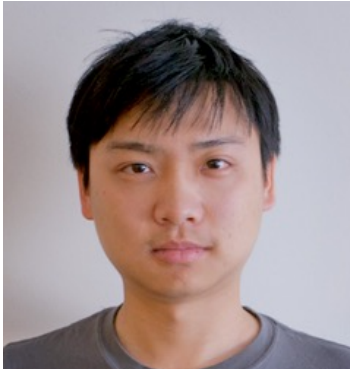
Model Reduction and Grid-Forming Freq. Shaping

**Enrique Mallada**



September 9, 2021

# Acknowledgements



Hancheng Min



Yan Jiang



Petr Vorobev



Andrey Bernstein

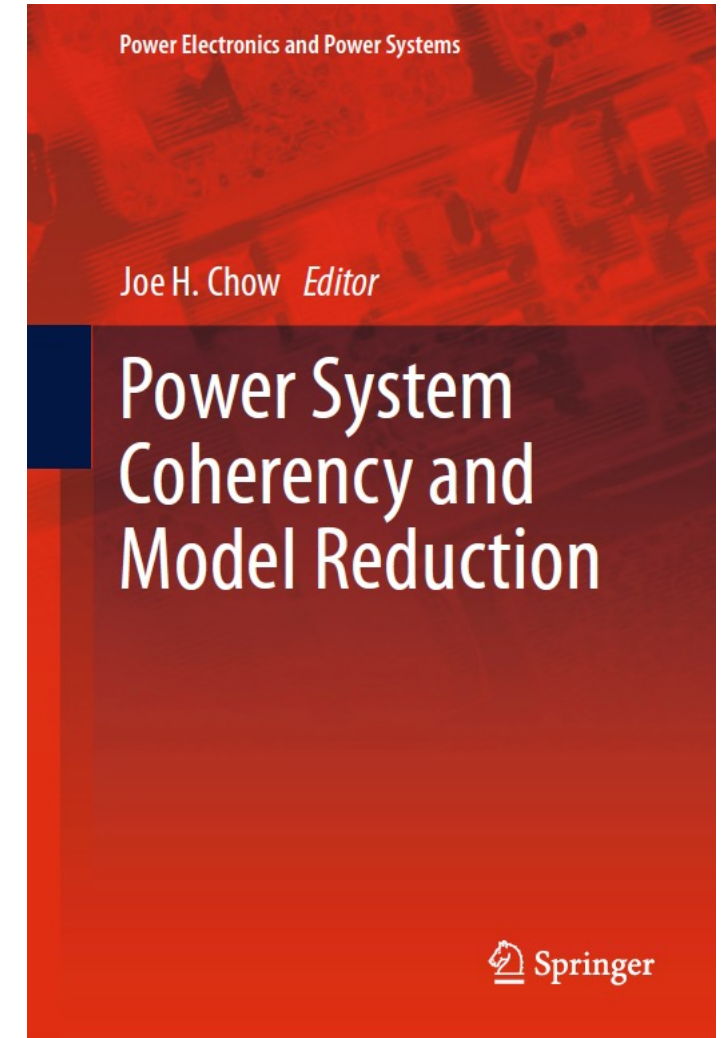


Fernando Paganini

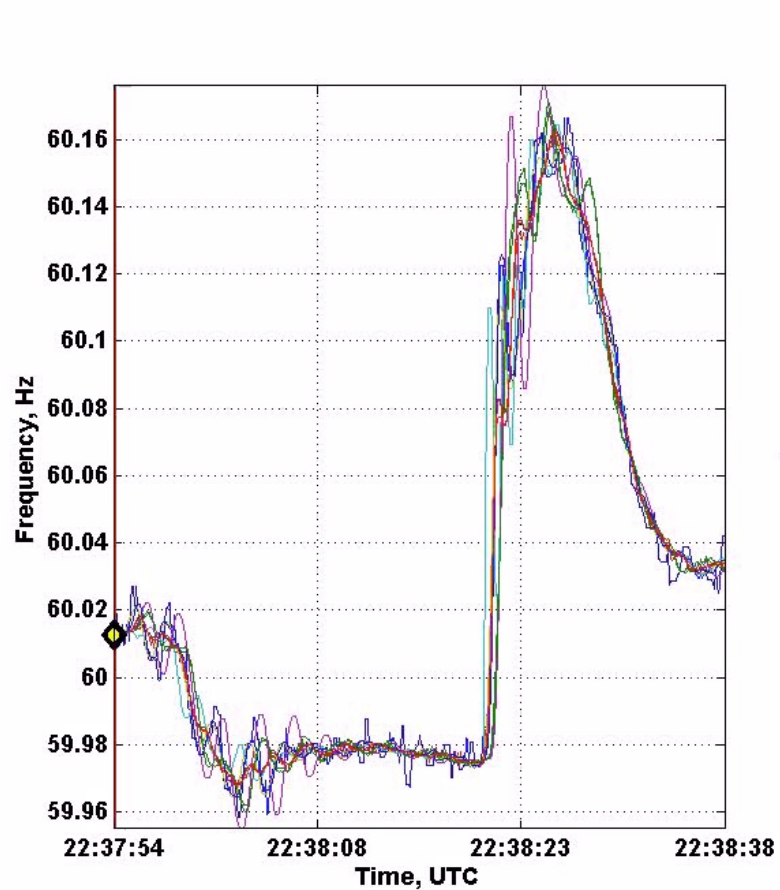


# Coherence in Power Networks

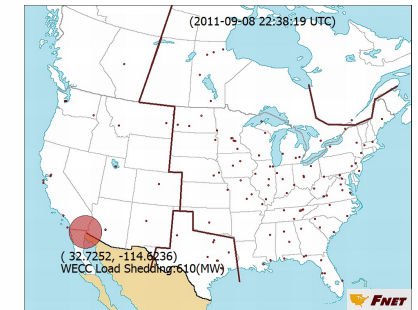
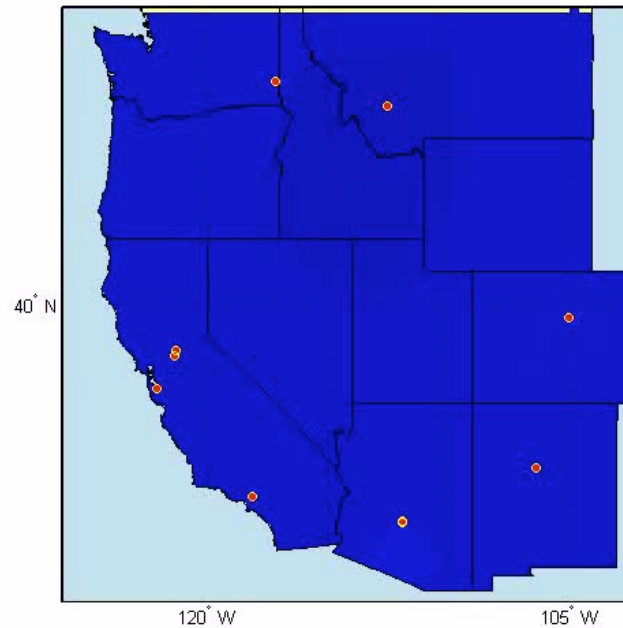
- Studied since the 70s
  - Podmore, Price, Chow, Kokotovic, Verghese, Pai, Schweppe,...
- Enables aggregation/model reduction
  - Speed up transient stability analysis
- Many important questions
  - How to identify coherent modes?
  - How to accurately reduce them?
  - What is the cause?
- Many approaches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



# This talk



FNET Data Display [9/8/2011 Southwest Blackout]  
Time: 22:37:54.0 UTC 60.0125 Hz



**Goal: Characterize the coherence response from a frequency domain perspective**

# Outline

- Characterization of Coherent Dynamics [Min, M '21]
- Reduced-Order Model of Coherent Response [Min, Paganini, M '21]
- Grid-forming Frequency Shaping Control [Jiang, Bernstein, Vorobev, M '21]

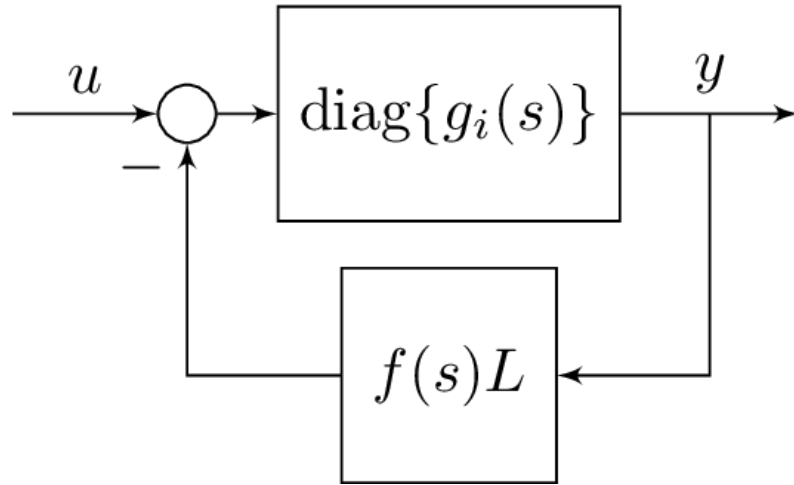
# Coherence and Concentration in Tightly-Connected Networks

Hancheng Min and Enrique Mallada

*ArXiv preprint: arXiv:2101.00981*

# Coherence in networked dynamical systems

## Block Diagram:



Node dynamics:  $g_i(s), i = 1, 2, \dots, n$

Symmetric Real Network Laplacian:  $L$

$$L = V\Lambda V^T, \quad V = [\mathbf{1}/\sqrt{n}, V_{\perp}]$$

$$\Lambda = \text{diag}\{0, \lambda_2(L), \dots, \lambda_n(L)\}$$

Coupling dynamics:  $f(s)$

## Examples:

- Consensus Networks:

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

$$f(s) = 1$$

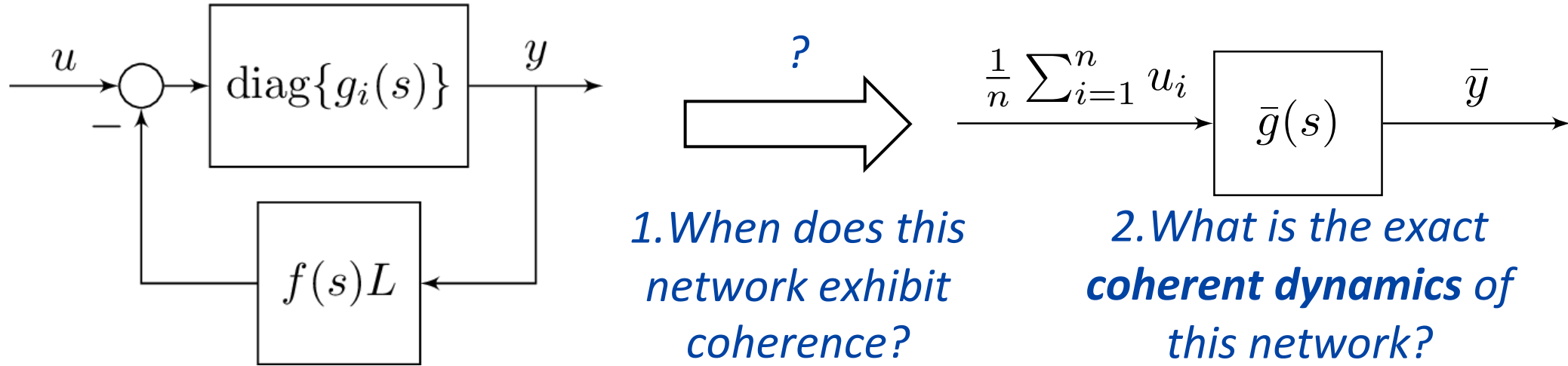
- Power Networks (2<sup>nd</sup> order generator):

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

$$f(s) = \frac{1}{s}$$

# Coherence in networked dynamical systems

## Block Diagram:



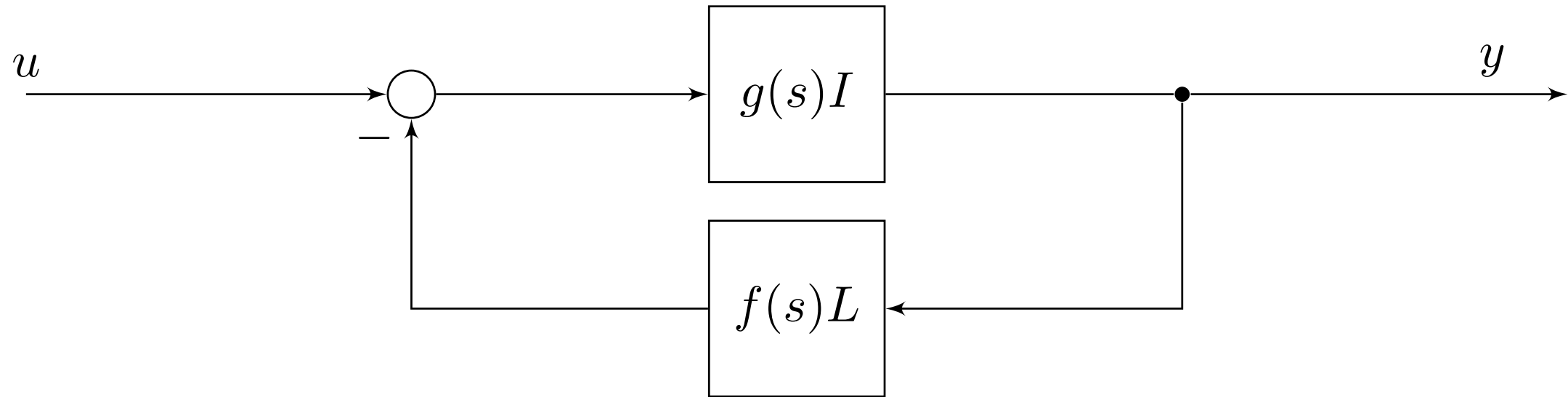
1. Coherence can be understood as a **low rank** property the **closed-loop transfer matrix**
2. It emerges as the **effective algebraic connectivity** increases
3. The coherent dynamics is given by the **harmonic mean** of nodal dynamics

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



# Network Coherence: Homogeneous Case

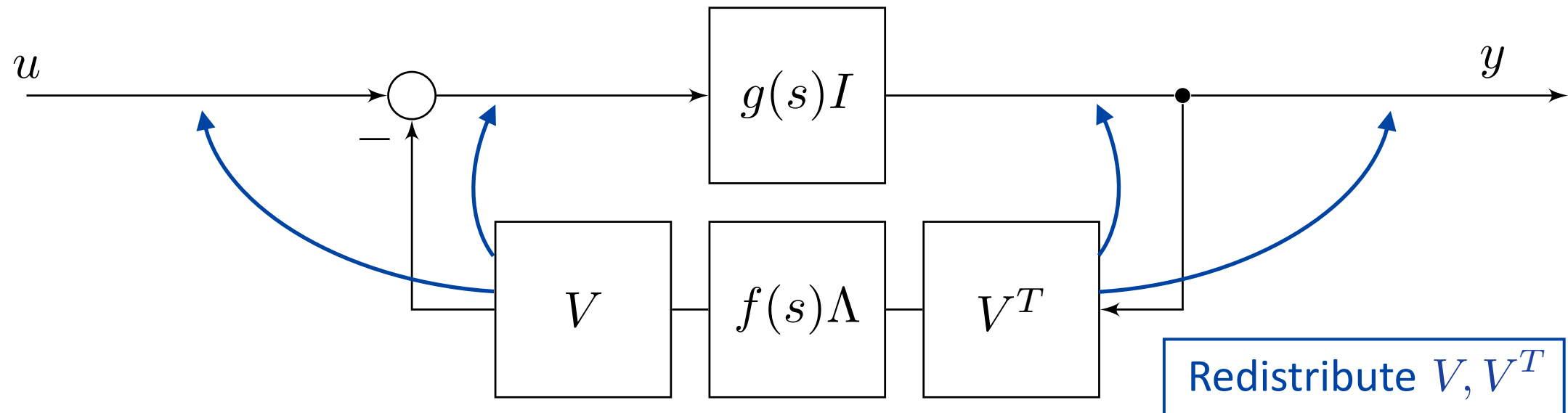
Assume homogeneity:  $g_i(s) = g(s), i = 1, \dots, n$



Eigendecomposition  $L = V\Lambda V^T$

# Network Coherence: Homogeneous Case

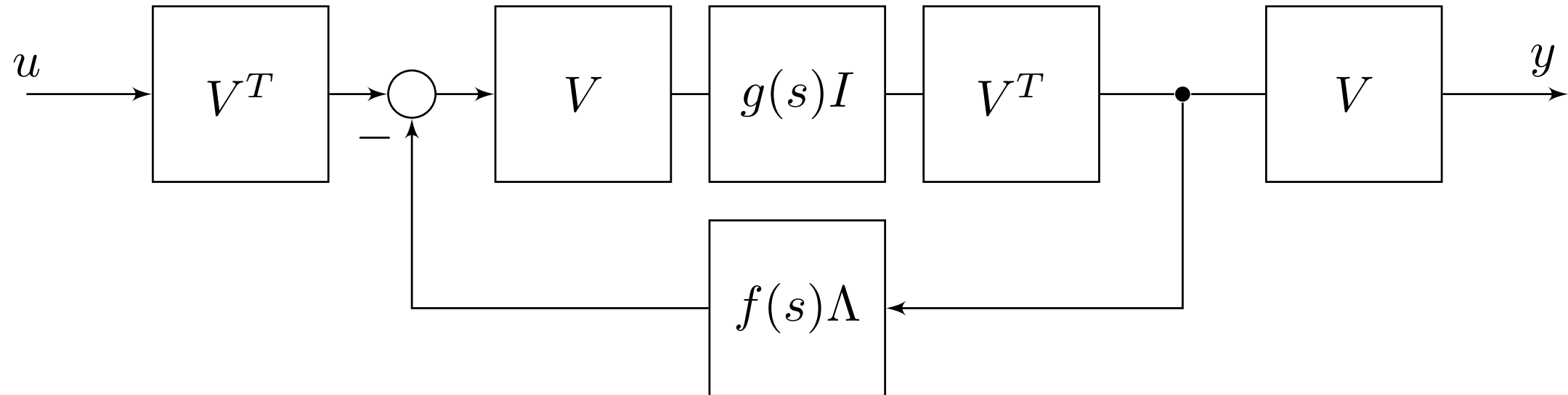
Assume homogeneity:  $g_i(s) = g(s), i = 1, \dots, n$



# Network Coherence: Homogeneous Case

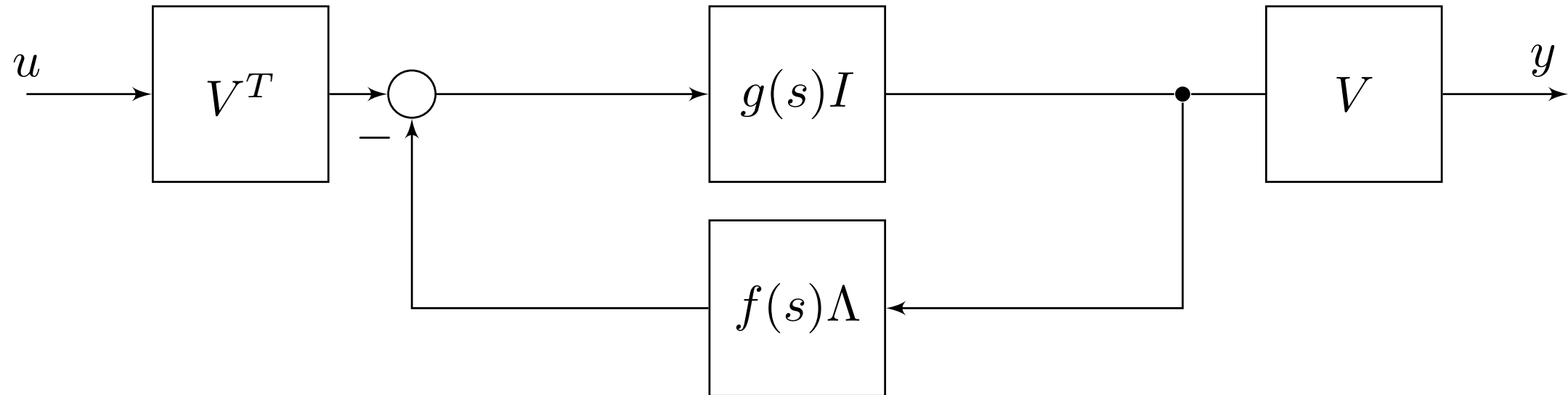
Assume homogeneity:  $g_i(s) = g(s), i = 1, \dots, n$

Merge forward path  
 $V^T V = I$



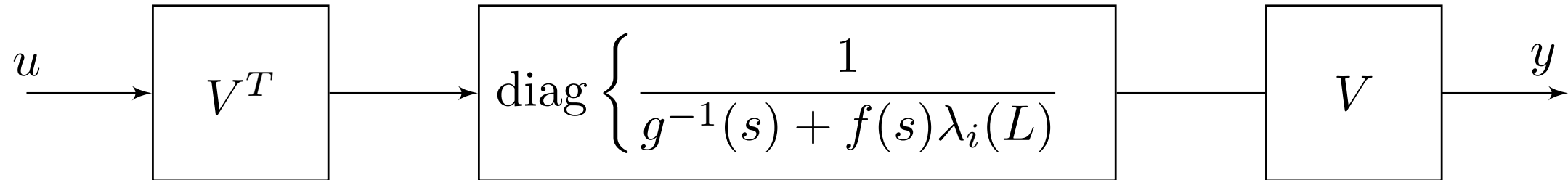
# Network Coherence: Homogeneous Case

Assume homogeneity:  $g_i(s) = g(s), i = 1, \dots, n$



# Network Coherence: Homogeneous Case

Assume homogeneity:  $g_i(s) = g(s)$ ,  $i = 1, \dots, n$

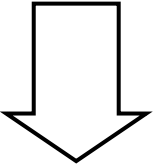


# Network Coherence: Homogeneous Case

Assume homogeneity:  $g_i(s) = g(s)$ ,  $i = 1, \dots, n$

The transfer matrix from input  $u$  to output  $y$  :

$$T(s) = V \operatorname{diag} \left\{ \frac{1}{g^{-1}(s) + f(s)\lambda_i(L)} \right\}_{i=1}^n V^T$$

$$V = [\mathbb{1}/\sqrt{n}, V_{\perp}], \lambda_1(L) = 0$$


$$T(s) = \frac{1}{n} g(s) \mathbb{1} \mathbb{1}^T + V_{\perp} \operatorname{diag} \left\{ \frac{1}{g^{-1}(s) + f(s)\lambda_i(L)} \right\}_{i=2}^n V_{\perp}^T$$

**Coherent dynamics**  
independent of the  
network structure

**Dynamics dependent of**  
the network structure

# Network Coherence: Homogeneous Case

$$T(s) = \frac{1}{n}g(s)\mathbb{1}\mathbb{1}^T + V_{\perp} \text{diag} \left\{ \frac{1}{g^{-1}(s) + f(s)\lambda_i(L)} \right\} V^T$$

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

- The **algebraic connectivity**  $\lambda_2(L)$  of the network increases
- The point of interest gets close to a **pole** of  $f(s)$

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

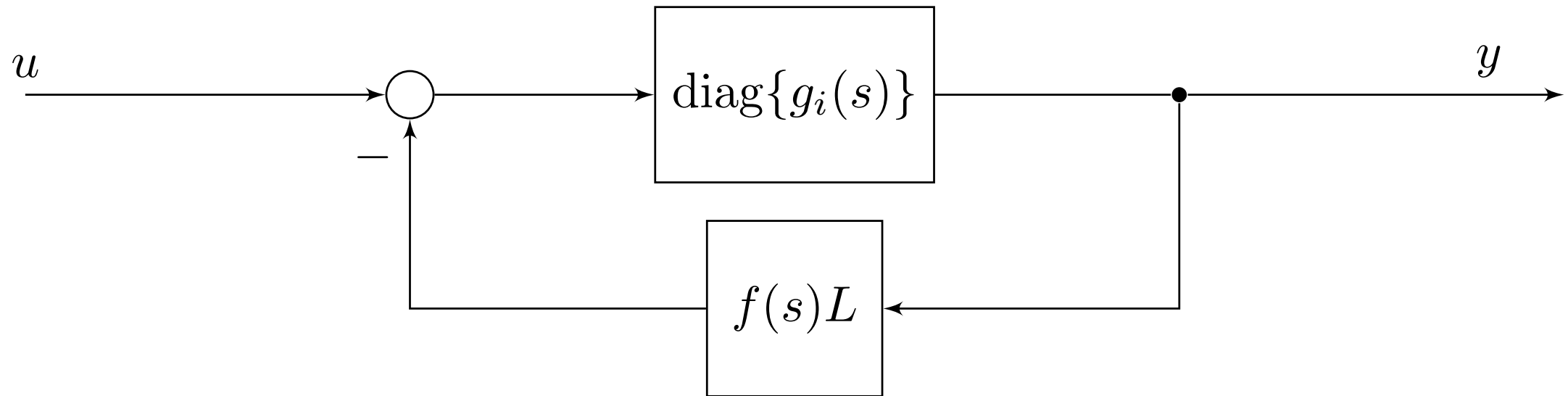
$$\lim_{\lambda_2(L) \rightarrow +\infty} \left\| T(s_0) - \frac{1}{n}g(s_0)\mathbb{1}\mathbb{1}^T \right\| = 0$$

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

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

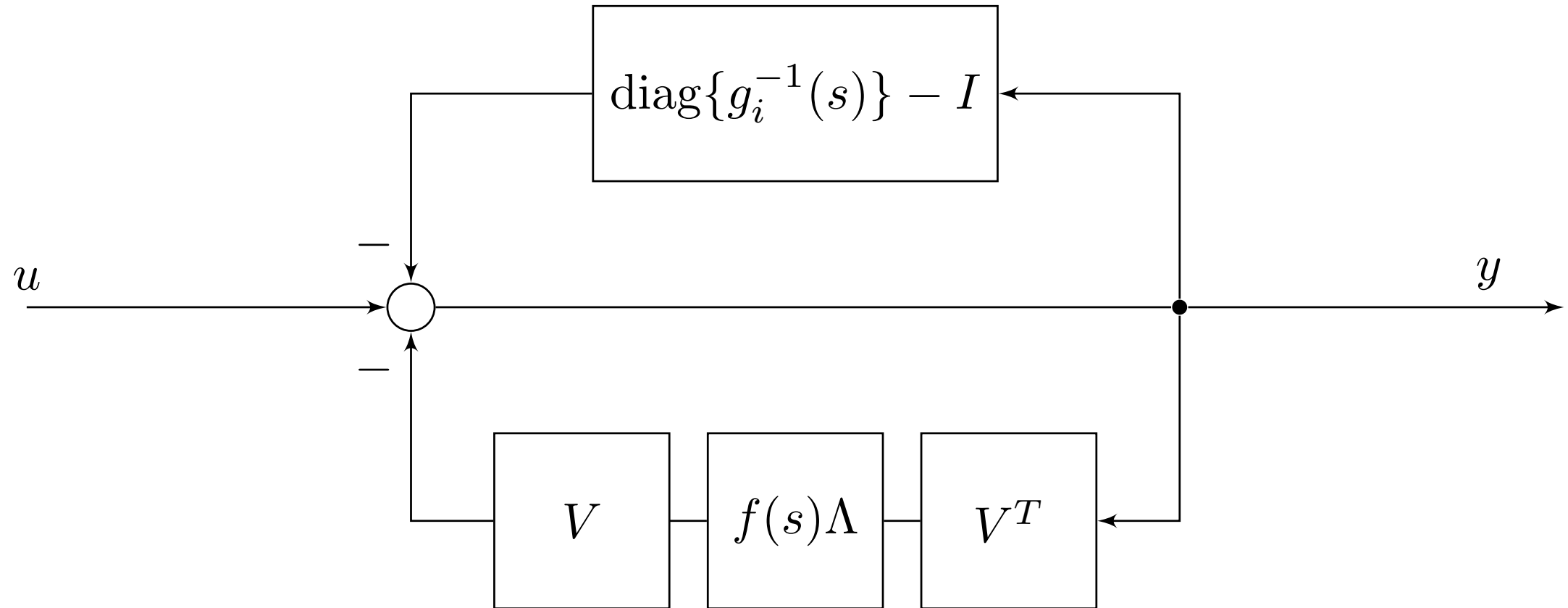
Our **frequency-dependent** coherence measure  $\left\| T(s) - \frac{1}{n}g(s)\mathbb{1}\mathbb{1}^T \right\|$  is controlled by the **effective algebraic connectivity**  $|f(s)|\lambda_2(L)$

# Network Coherence: Heterogeneous Case

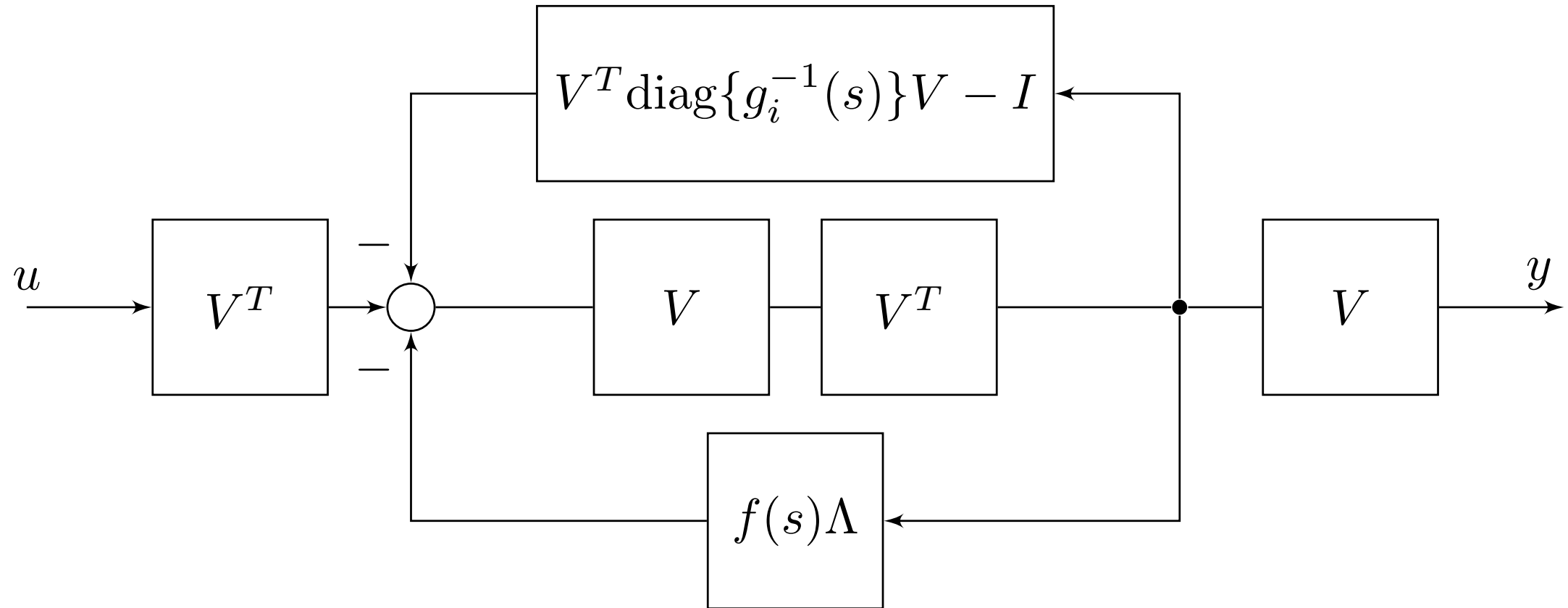




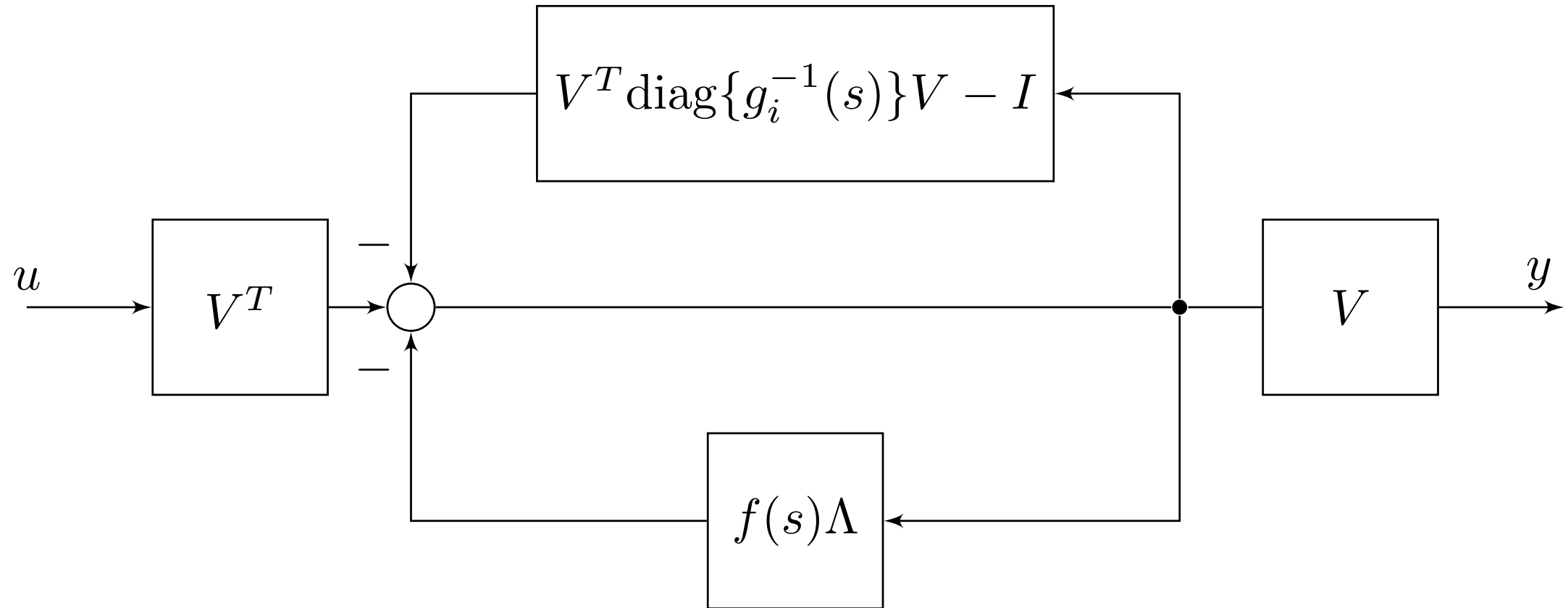
# Network Coherence: Heterogeneous Case



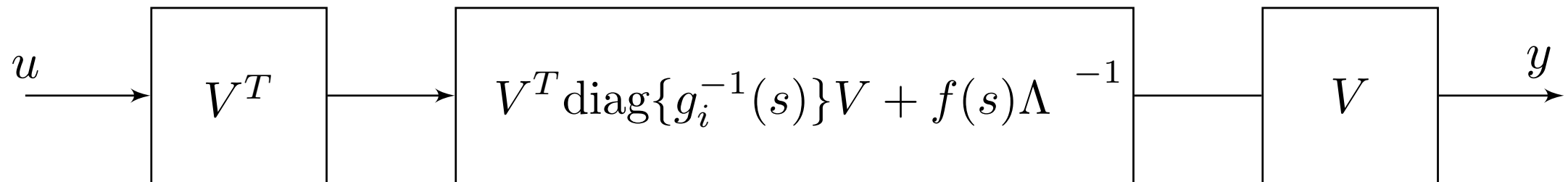
# Network Coherence: Heterogeneous Case



# Network Coherence: Heterogeneous Case



# Network Coherence: Heterogeneous Case



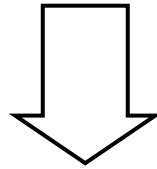
The transfer matrix from input  $u$  to output  $y$  :

$$T(s) = V \left[ V^T \text{diag}\{g_i^{-1}(s)\}V + f(s)\Lambda^{-1} \right] V^T$$

# Network Coherence: Heterogeneous Case

The transfer matrix from input  $u$  to output  $y$  :

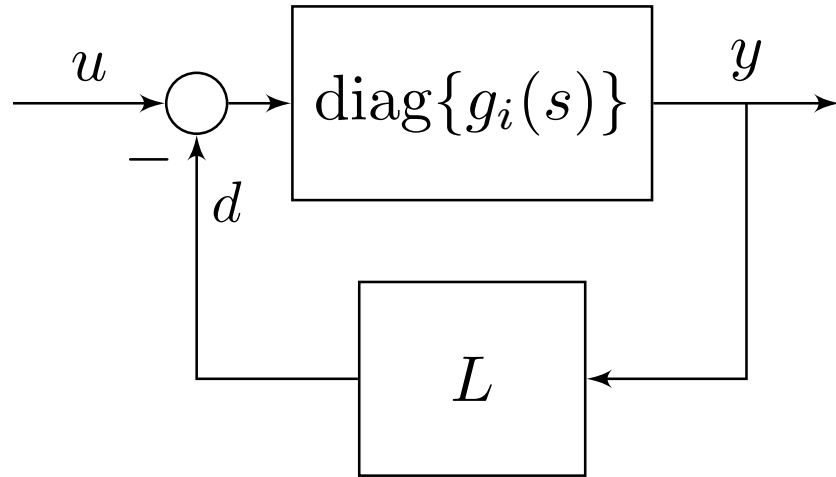
$$T(s) = V V^T \text{diag}\{g_i^{-1}(s)\} V + f(s) \Lambda^{-1} V^T$$



$$T(s) = \underbrace{\frac{1}{n} \bar{g}(s) \mathbb{1} \mathbb{1}^T}_{\text{Coherent Dynamics?}} + \underbrace{N(s)}_{\text{Network dependent?}}$$

# Informed guess for coherent dynamics: $\bar{g}(s)$

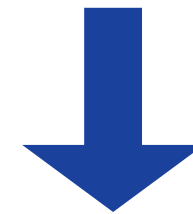
Block Diagram:



Dynamics for node  $i$

$$y_i(s) = g_i(s)(u_i(s) - d_i(s)), \quad i = 1, \dots, n$$

Assume all nodes  
output are **identical**  
as the result of  
**coherence**



$$y_i(s) = \bar{y}(s)$$

$$g_i^{-1}(s)\bar{y}(s) = u_i(s) - d_i(s), \quad i = 1, \dots, n$$

**Coherent Dynamics:**

$$\bar{y}(s) = \left( \frac{1}{n} \sum_{i=1}^n g_i^{-1}(s) \right)^{-1} \frac{1}{n} \sum_{i=1}^n u_i(s)$$

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

Harmonic mean of all  $g_i(s)$

Average equations from  $i = 1$  to  $n$ :

$$\frac{1}{n} \sum_{i=1}^n \left( g_i^{-1}(s) \right) \bar{y}(s) = \frac{1}{n} \sum_{i=1}^n u_i(s) - \underbrace{\frac{1}{n} \sum_{i=1}^n d_i(s)}_{=0}$$

$$\mathbf{1}^T L = 0$$

# 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  $s_0 \in \mathbb{C}$

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

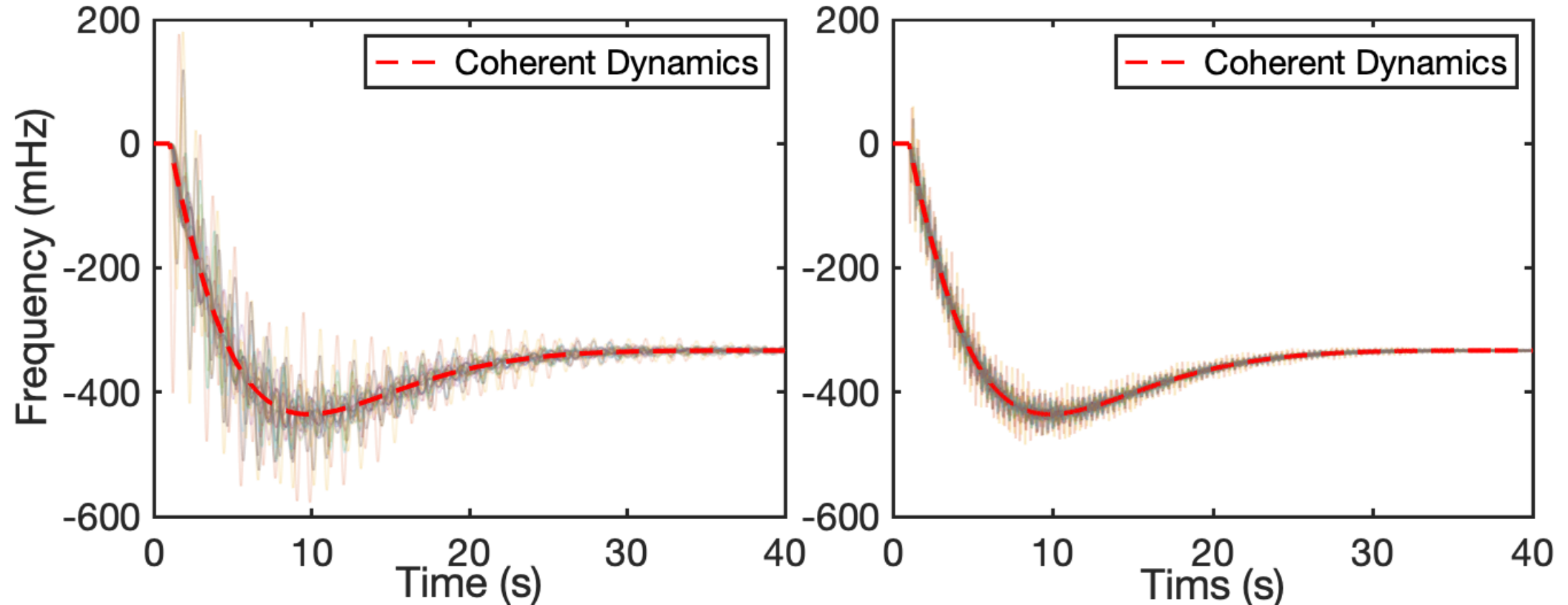
- For  $s_0 \in \mathbb{C}$ , a pole of  $f(s)$

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

- 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  $\bar{g}(s)$
- Convergence of transfer matrix is **related to time-domain response** by Inverse Laplace Transform
- Extensions for random network ensembles  $\bar{g}(s) = (E_w [g^{-1}(s, w)])^{-1}$

# Effect of Network Algebraic Connectivity

$$\lambda_2(L) \uparrow$$

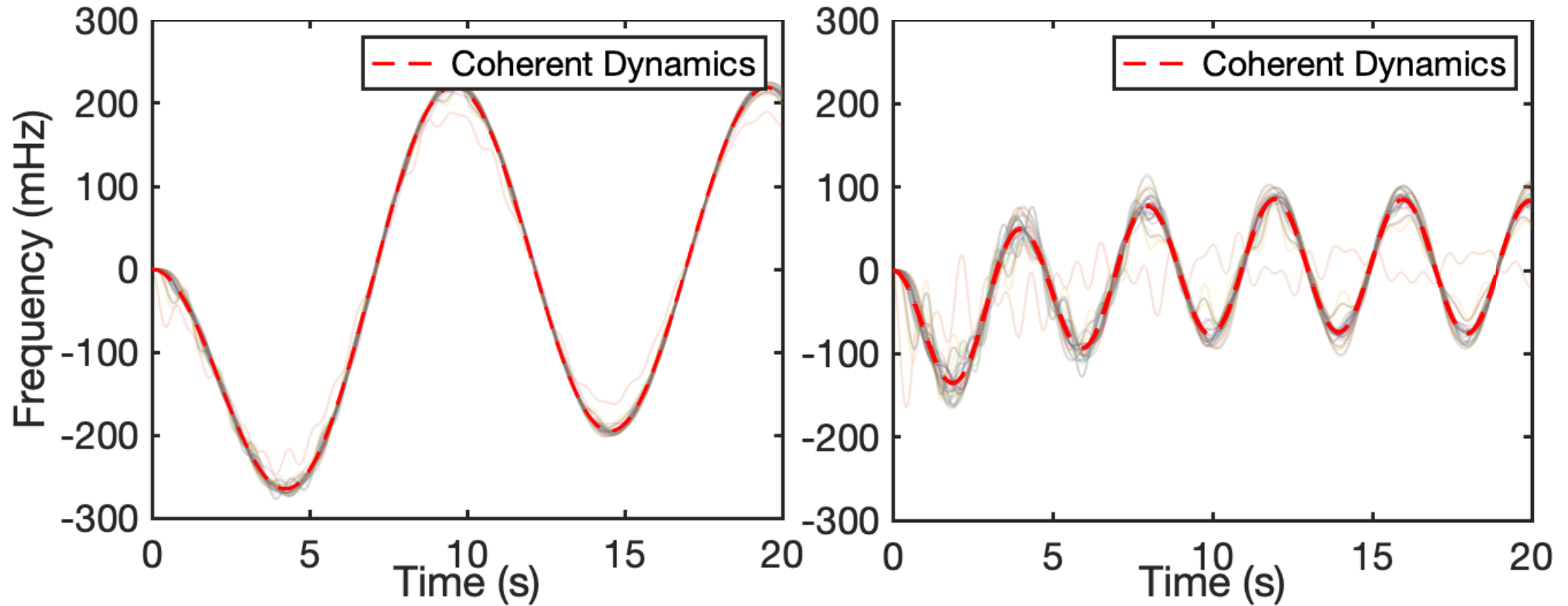


Coherent dynamics acts as a more accurate version of the Center of Inertia (CoI)



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

$\omega_d \uparrow$



# Outline

- Characterization of Coherent Dynamics [Min, M '21]
- Reduced-Order Model of Coherent Response [Min, Paganini, M '21]
- Grid-forming Frequency Shaping Control [Jiang, Bernstein, Vorobev, M '21]

# Accurate Reduced-Order Models for Heterogeneous Coherent Generators

Hancheng Min, Fernando Paganini, and Enrique Mallada

*IEEE Control Systems Letters, 2021*

# Aggregation of Coherent Generators

▪

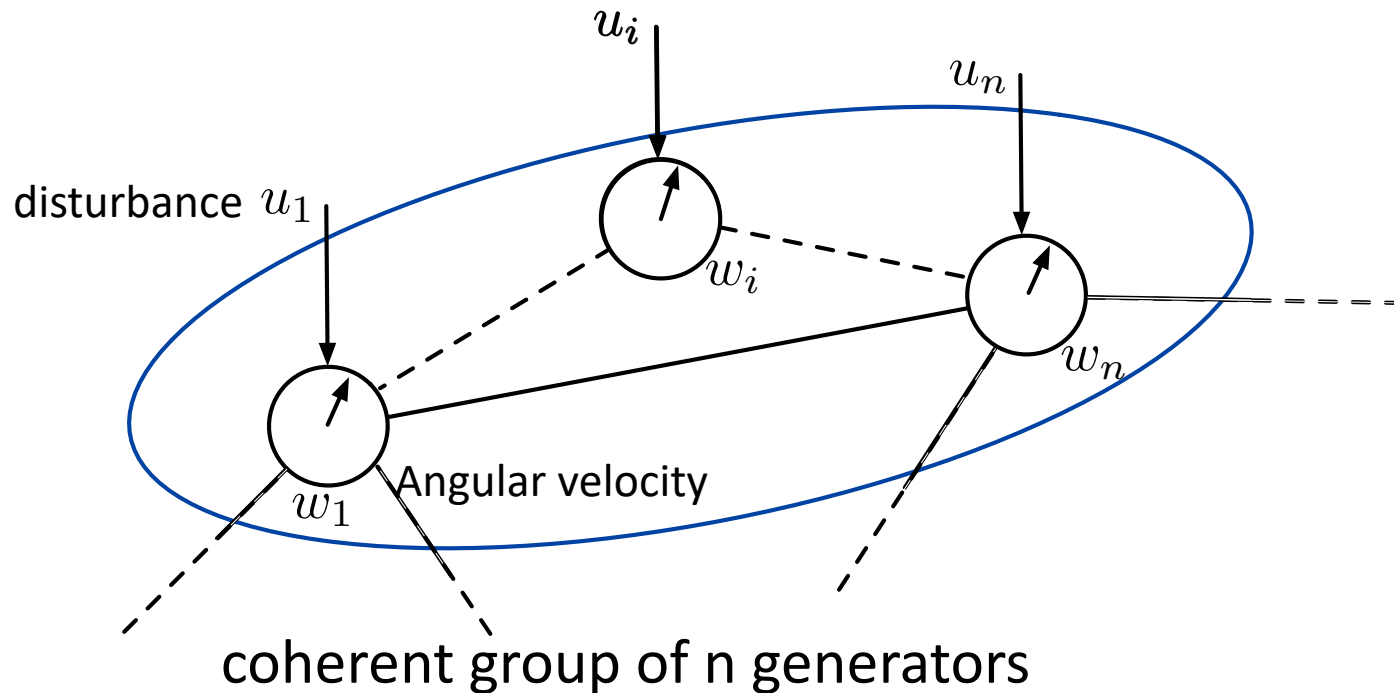
$$g_i(s) = \frac{1}{m_i s + d_i + \frac{r_i^{-1}}{\tau_i s + 1}}$$

$m_i$ : inertia

$d_i$ : damping coefficient

$r_i^{-1}$ : droop coefficient

$\tau_i$ : turbine time constant

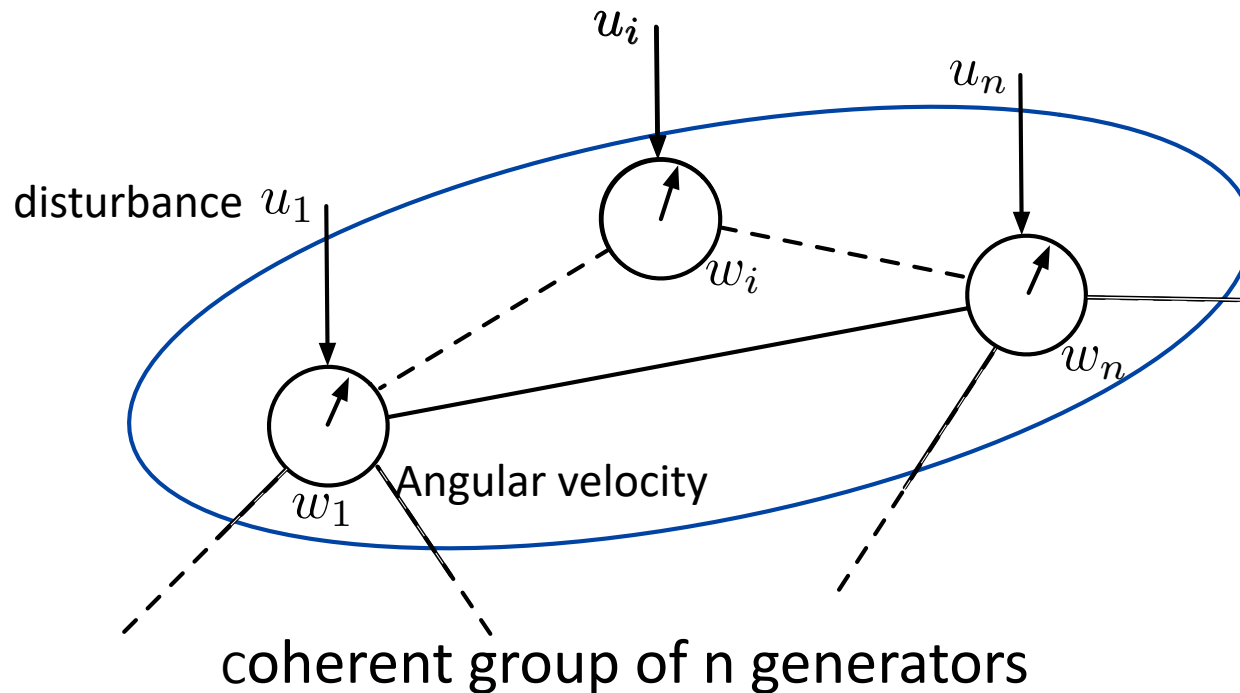


# Aggregation of Coherent Generators

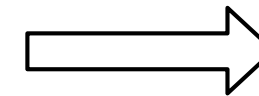
▪

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

**Question:** How to choose the different parameters of  $\hat{g}(s)$ ?

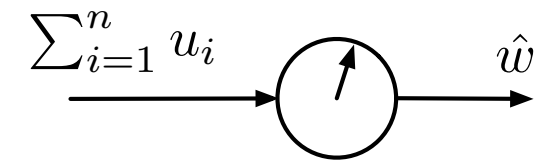


Aggregation



▪

$$\hat{g}(s) = \frac{1}{\hat{m}s + \hat{d} + \frac{\hat{r}^{-1}}{\hat{\tau}s + 1}}$$



**Answer:** Use instead

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

# Aggregation for Homogeneous $\tau_i = \tau$

▪

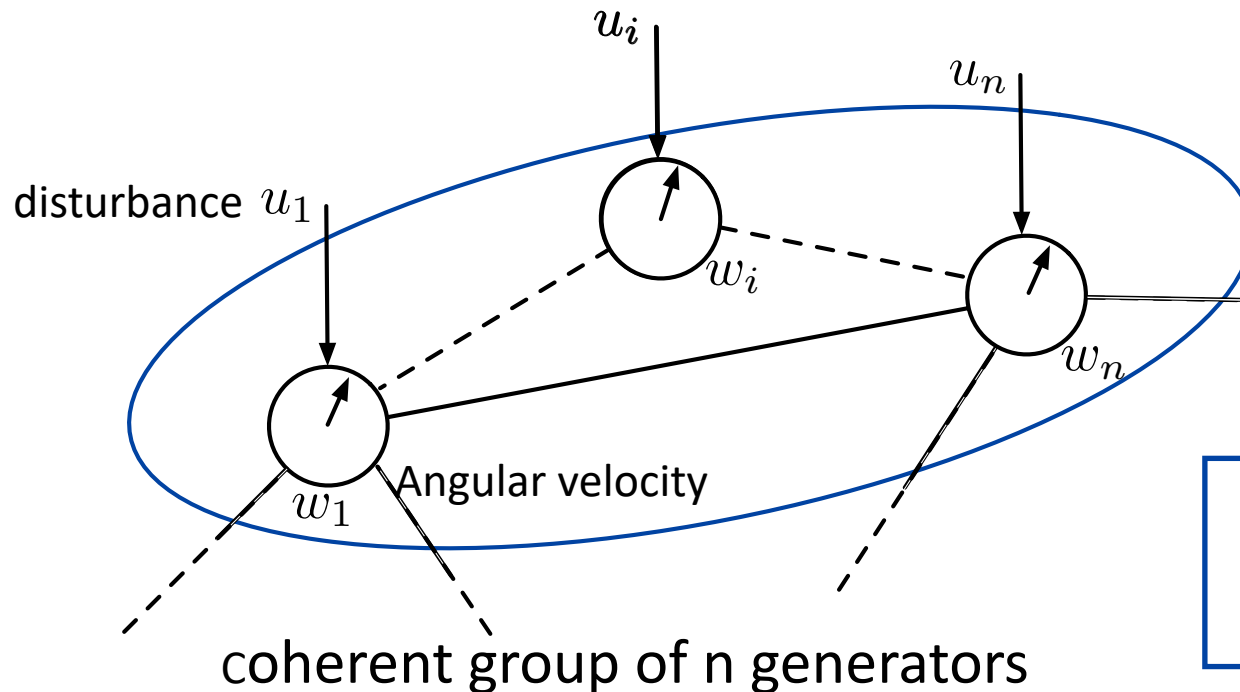
$$g_i(s) = \frac{1}{m_i s + d_i + \frac{r_i^{-1}}{\tau_i s + 1}}$$

then  $\hat{m} = \sum_{i=1}^n m_i, \quad \hat{d} = \sum_{i=1}^n d_i, \quad \hat{r}^{-1} = \sum_{i=1}^n r_i^{-1}$

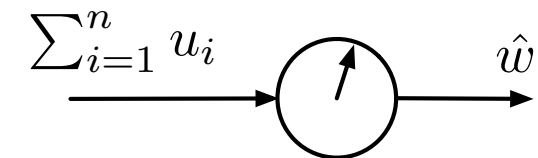
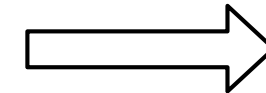
suppose  $\tau_i = \tau$

▪

$$\hat{g}(s) = \frac{1}{\hat{m}s + \hat{d} + \frac{\hat{r}^{-1}}{\hat{\tau}s + 1}}$$



Aggregation



$$\hat{g}(s) = \frac{1}{(\sum_{i=1}^n m_i)s + (\sum_{i=1}^n d_i) + \frac{1}{\tau s + 1} (\sum_{i=1}^n r_i^{-1})}$$

# Challenges on Aggregating Coherent Generators

For generator dynamics given by a swing model with turbine control:

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

The aggregate dynamics:

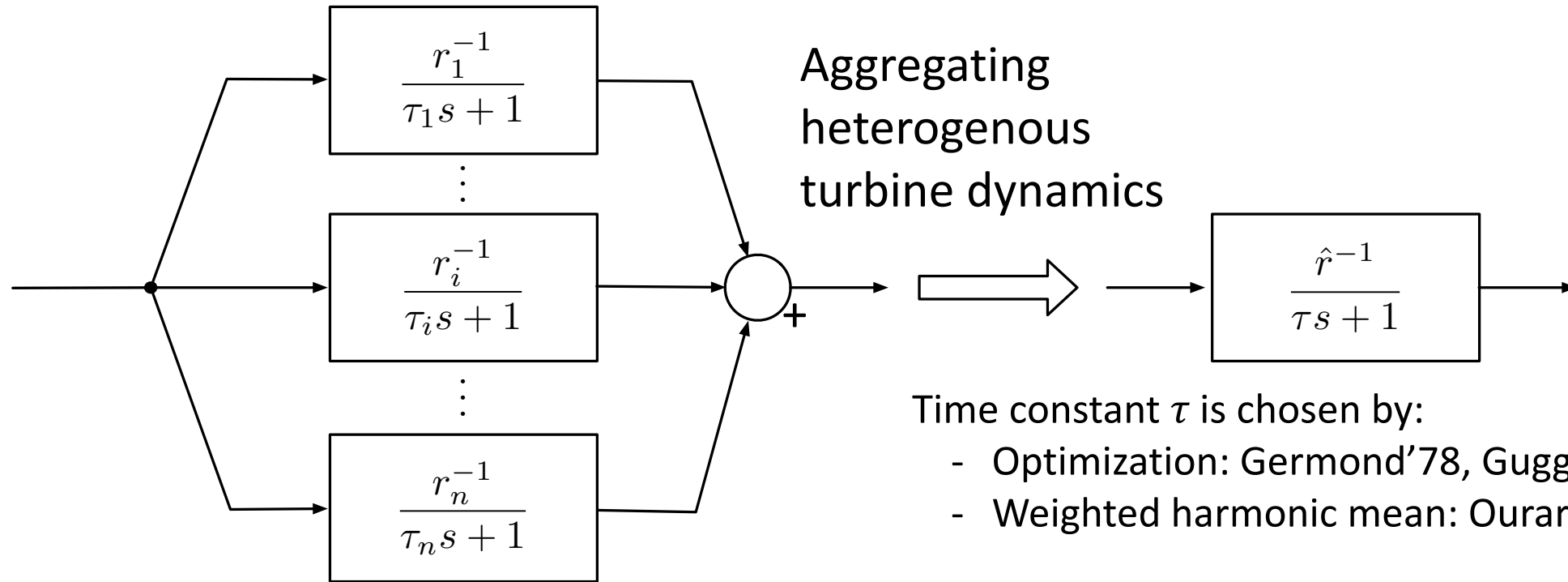
$$\hat{g}(s) = \frac{1}{\hat{m}s + \hat{d} + \boxed{\sum_{i=1}^n \frac{r_i^{-1}}{\tau_i s + 1}}}$$

high-order if  $\tau_i$  are heterogeneous

Need to find a low-order approximation of  $\hat{g}(s)$

# Prior Work: Aggregation for heterogeneous $\tau_i$ s

When time constants are **heterogenous**:



Time constant  $\tau$  is chosen by:

- Optimization: Germond'78, Guggilam'18
- Weighted harmonic mean: Ourari'06

## Drawbacks:

- the order of overall approximation model is restricted to 2nd order
- the only “decision variable” is the time constant
- does not consider the effect of inertia or damping in the approx.

**Inaccurate  
Approximation**



# Balanced Truncation

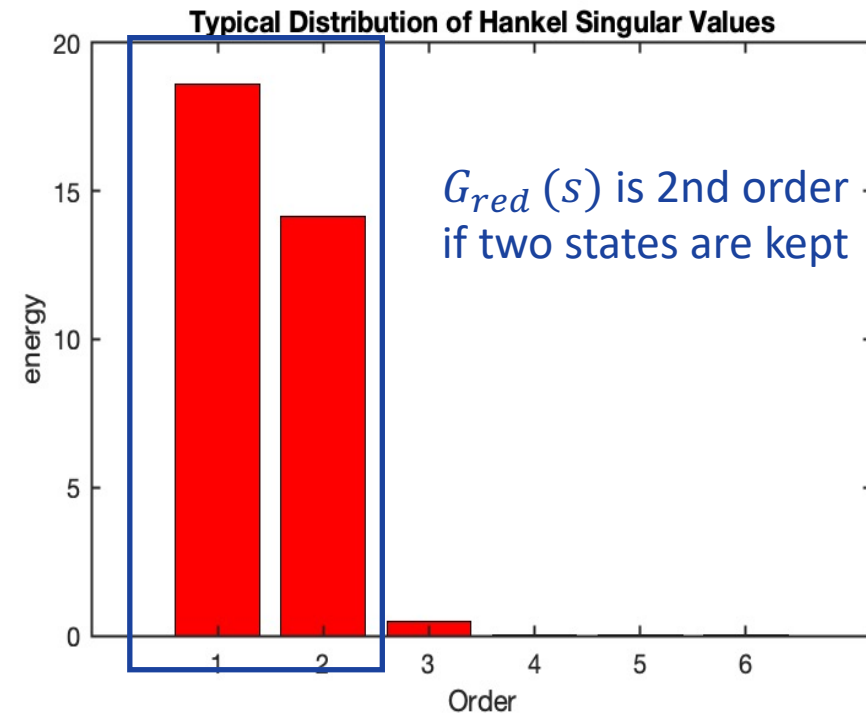
A model reduction method on stable system  $G(s)$  such that:

- The reduced model  $G_{red}(s)$  is stable
- The error in  $H_\infty$ -norm:

$$\|G(s) - G_{red}(s)\|_{\mathcal{H}_\infty}$$

is upper bounded by a small value that depends on  $G(s)$  and **the order of  $G_{red}(s)$**

$k$ -th order  $G_{red}(s)$  is obtained by only keeping states of  $G(s)$  associated with  $k$  largest Hankel Singular Value



**There is DC gain mismatch between  $G(s)$  and  $G_{red}(s)$ !!**

# Frequency Weighted Balanced Truncation

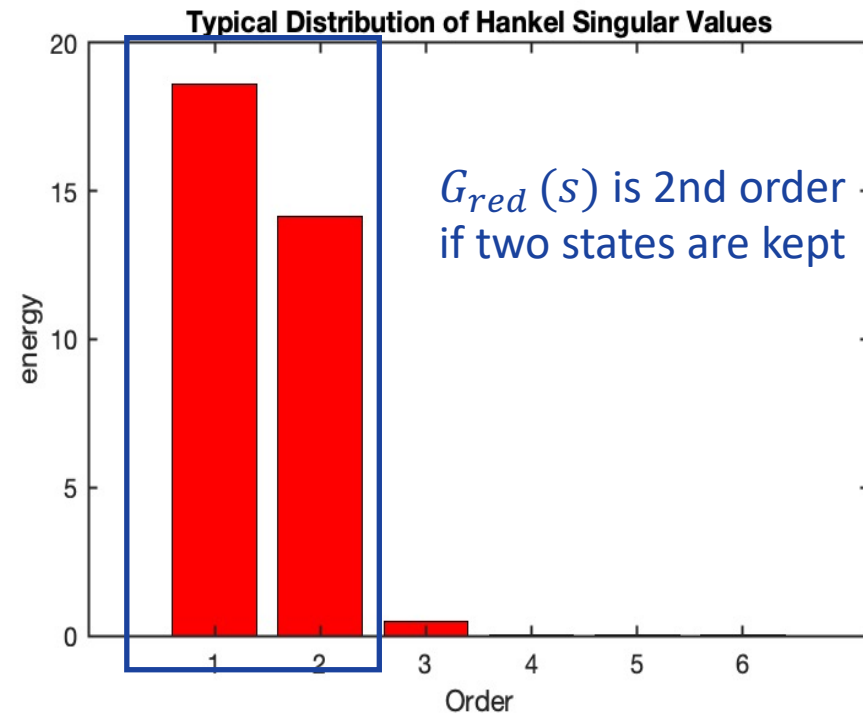
A **frequency weighted** model reduction method on stable system  $G(s)$  such that:

- The reduced model  $G_{red}(s)$  is stable
- The **frequency weighted** error in  $H_\infty$ -norm:

$$\|W(s)(G(s) - G_{red}(s))\|_{\mathcal{H}_\infty}$$

is upper bounded by a small value that depends on  $G(s)$  and **the order of  $G_{red}(s)$  ) and  $W(s)$**

k-th order  $G_{red}(s)$  is obtained by only keeping states of  $G(s)$  associated with k largest **frequency weighted** Hankel Singular Value



**The DC gain mismatch between  $G(s)$  and  $G_{red}(s)$  can be made arbitrarily small weighting higher low freqs.**

## Aggregation Model by Frequency **Weighted** Balanced Truncation

Two approaches to get a k-th order reduction model of aggregate dynamics  $\hat{g}(s)$ :

- (k-1)-th order balanced truncation on high-order turbine dynamics

$$\tilde{g}_k^{tb}(s) = \frac{1}{\hat{m}s + \hat{d} + \boxed{\tilde{g}_{t,k-1}(s)}}$$

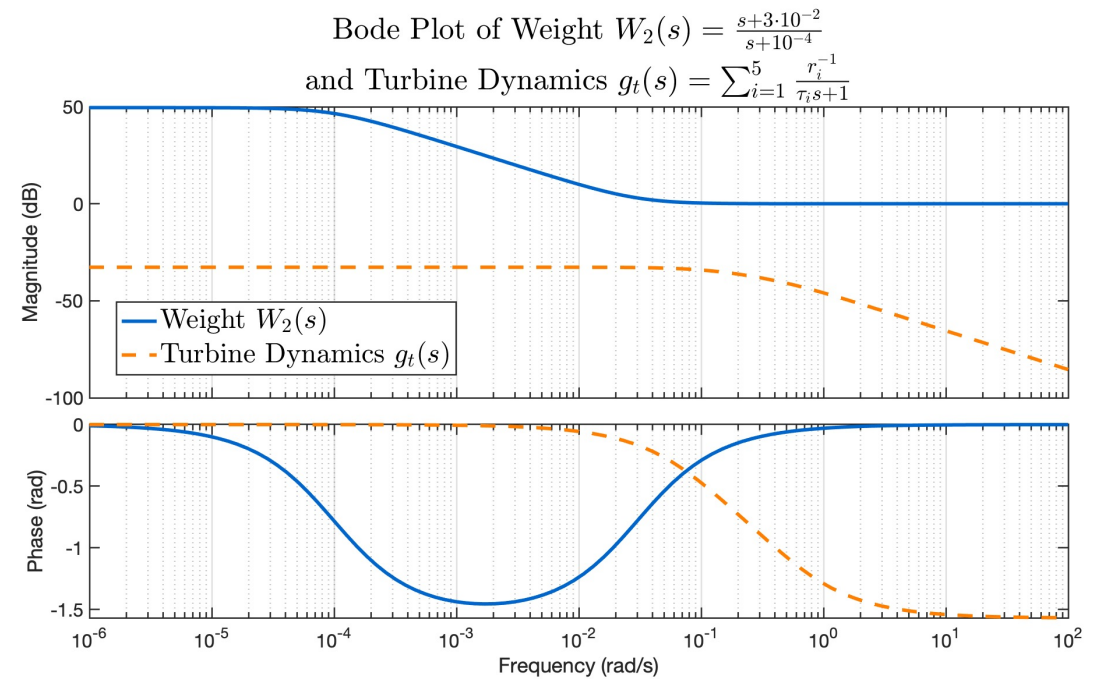
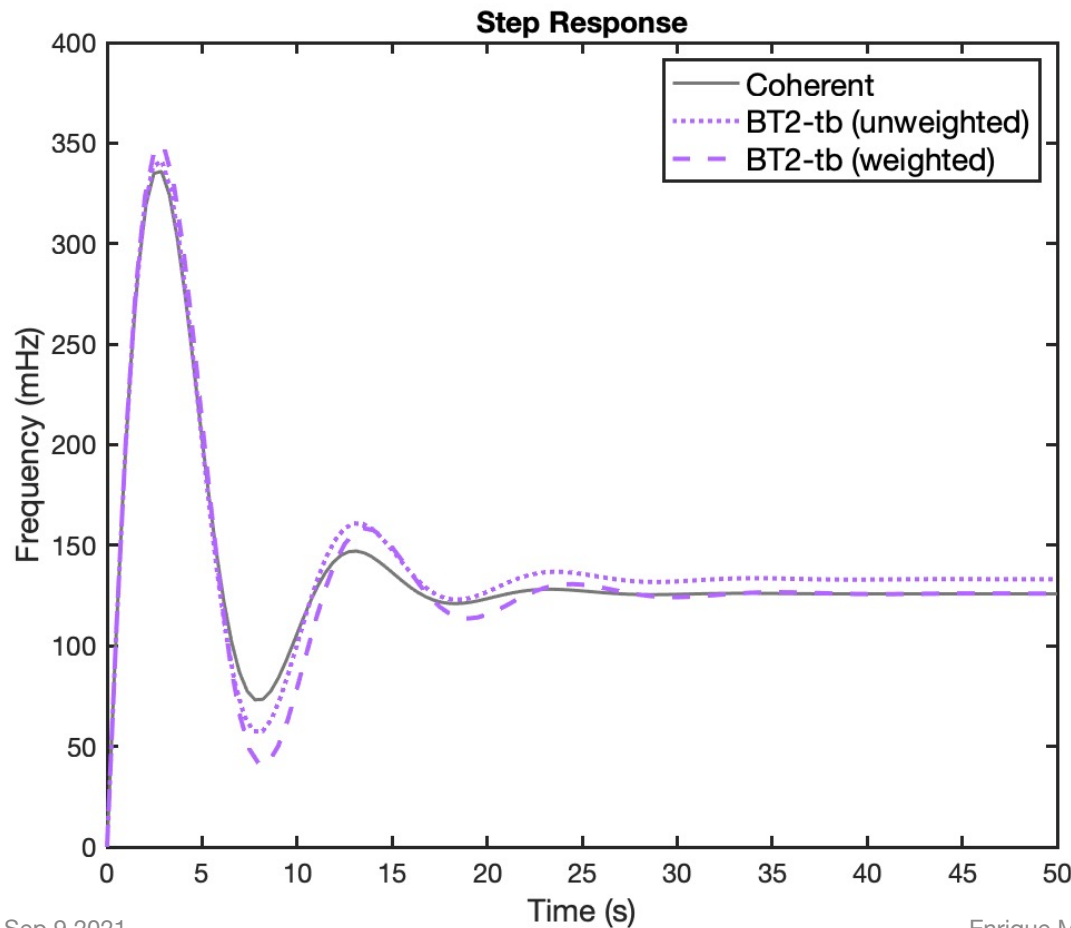
(k-1)-th reduction model on  $\sum_{i=1}^n \frac{r_i^{-1}}{\tau_i s + 1}$

- k-th order balanced truncation on closed-loop dynamics  $\hat{g}(s)$

# Numerical Simulation—Matching DC Gain in Balanced Truncation

Compare 2nd order model by balanced truncation on turbine dynamics

with different weights:  $W_1(s) = 1$  (unweighted)  $W_2(s) = \frac{s + 3 \cdot 10^{-2}}{s + 10^{-4}}$  (weighted)

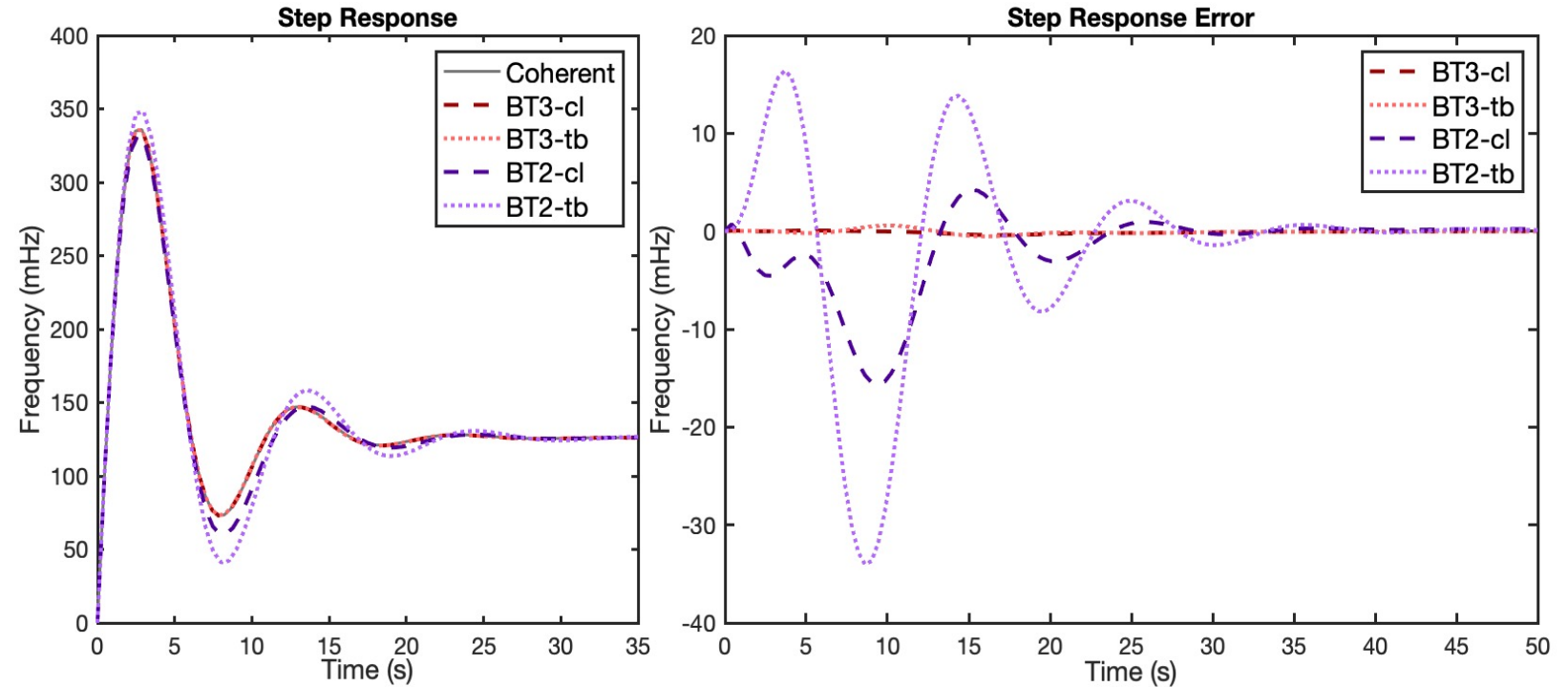


DC gain is matched by putting more weights on low frequency range

# Numerical Simulation—Compare Models by Balanced Truncation

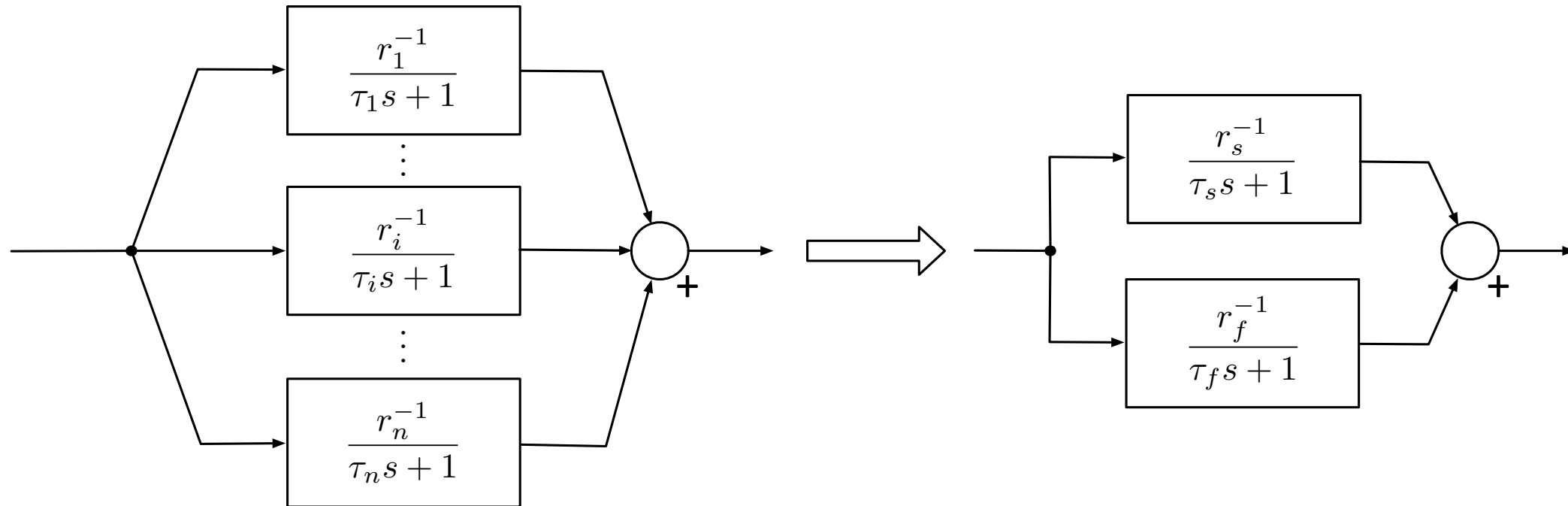
We compare the following 4 reduced order models:

- Balanced truncation on **turbine** dynamics with weight  $W_{tb}(s) = \frac{s+3 \cdot 10^{-2}}{s+10^{-4}}$ 
  - 2nd order (BT2-tb)
  - 3rd order (BT3-tb)
- Balanced truncation on **closed-loop** dynamics with weight  $W_{cl}(s) = \frac{s+8 \cdot 10^{-2}}{s+10^{-4}}$ 
  - 2nd order (BT2-cl)
  - 3rd order (BT3-cl)



- 3rd order models are almost accurate
- balanced truncation on closed-loop is better than on turbine dynamics, given the same order

# Interpretation of 3rd Order Reduced Model



- The high-order turbine dynamics can be **almost accurately** recovered by **two turbines** in parallel
- Such approximation works for aggregating even more turbines than in the test case

# Outline

- Characterization of Coherent Dynamics [Min, M '21]
- Reduced-Order Model of Coherent Response [Min, Paganini, M '21]
- Grid-forming Frequency Shaping Control [Jiang, Bernstein, Vorobev, M '21]

# Storage-Based Frequency Shaping Control

Yan Jiang, Eliza Cohn, Petr Vorobev, *Member, IEEE*, and Enrique Mallada, *Senior Member, IEEE*

[TPS 21]

*IEEE Transactions on Power Systems, 2021*

# Grid-forming frequency shaping control

Yan Jiang<sup>1</sup>, Andrey Bernstein<sup>2</sup>, Petr Vorobev<sup>3</sup>, and Enrique Mallada<sup>1</sup>

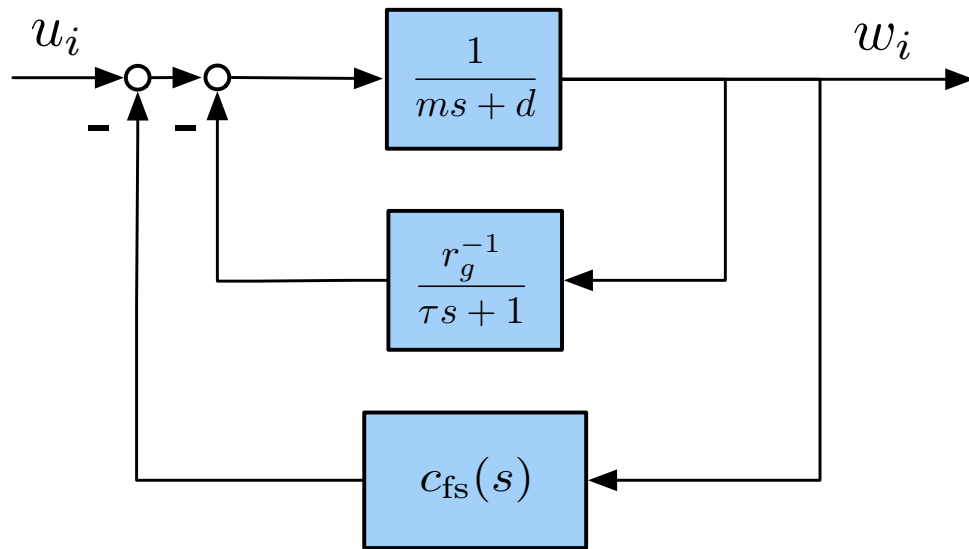
[L-CSS 21]

*IEEE Control Systems Letters, 2021*

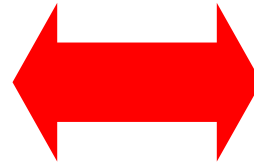


# Grid-following Frequency Shaping Control

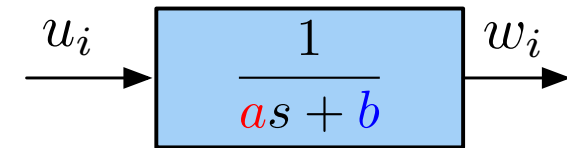
**Key idea:** use model matching control (at each bus)



$$c_{fs}(s) := \frac{A_1 s^2 + A_2 s + A_3}{\tau s + 1}$$



$$\begin{aligned} A_1 &= \tau (a - m) \\ A_2 &= b\tau + a - m \\ A_3 &= b - r_g - d \end{aligned}$$



Leads to Col Frequency  $\bar{w}$  with:

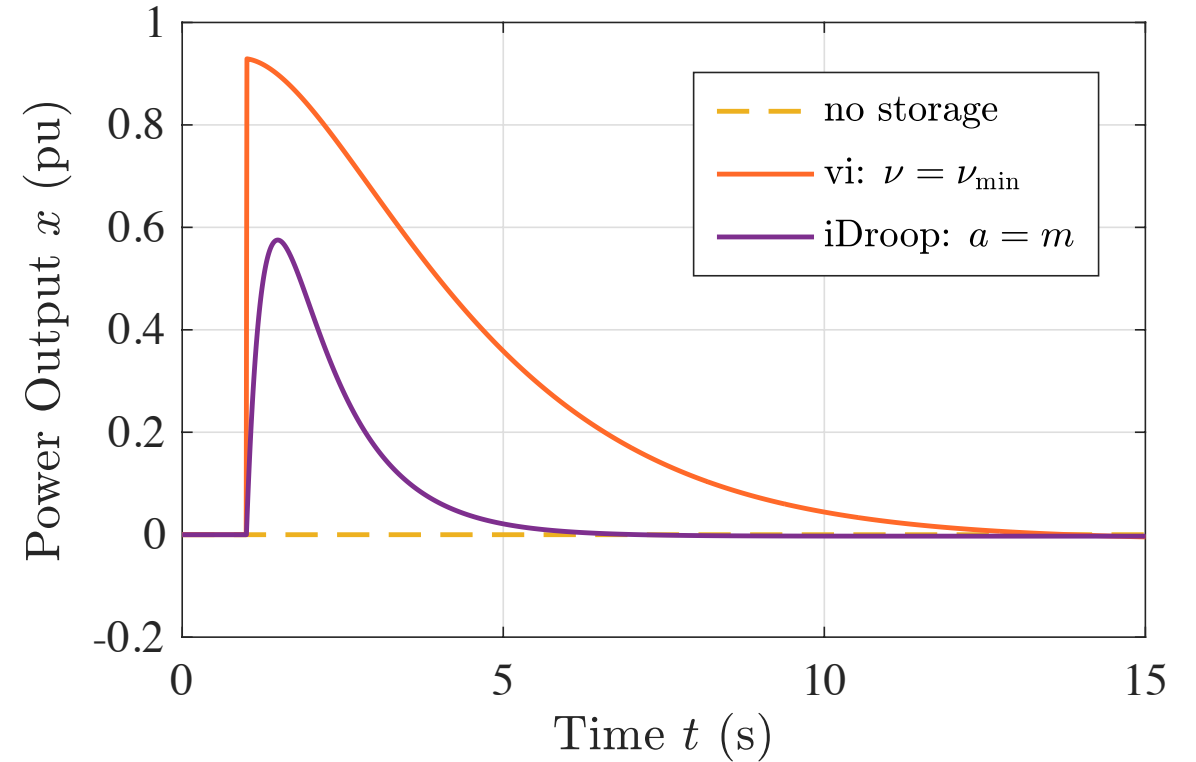
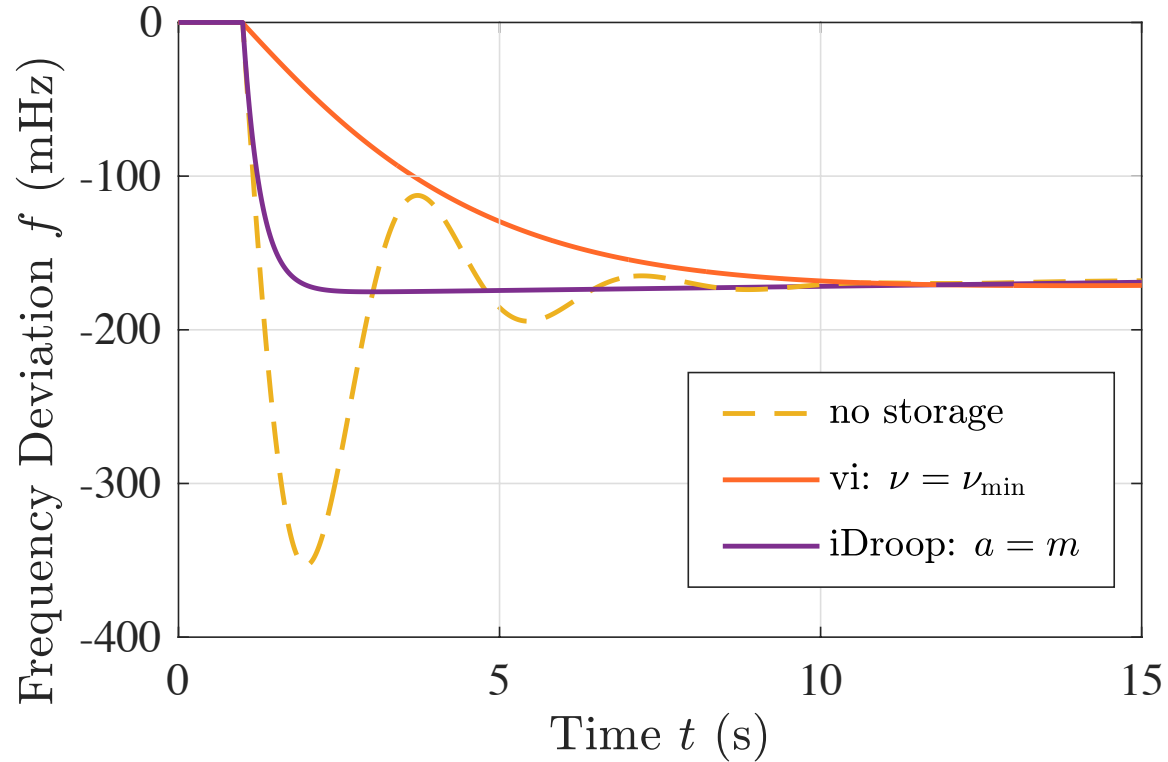
**RoCoF:**

$$\|\dot{\bar{w}}\|_{\infty} = \frac{|\sum_i u_{0i}|}{\sum_i f_i} \frac{1}{a}$$

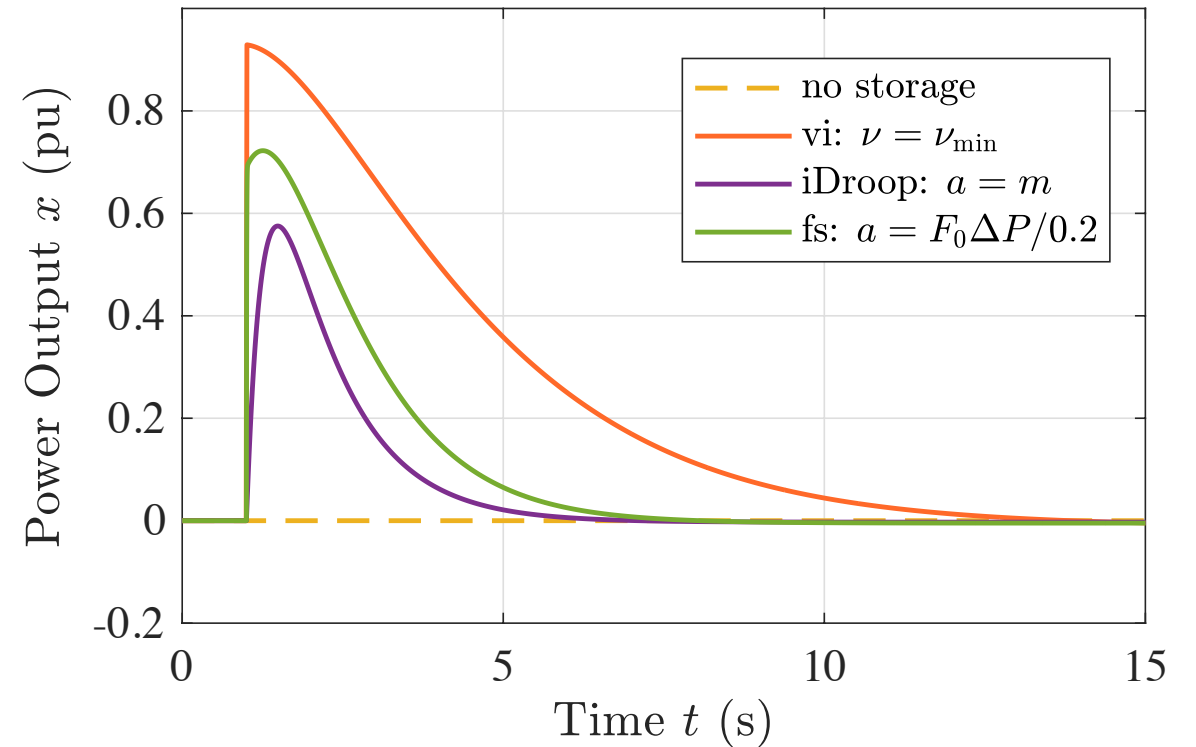
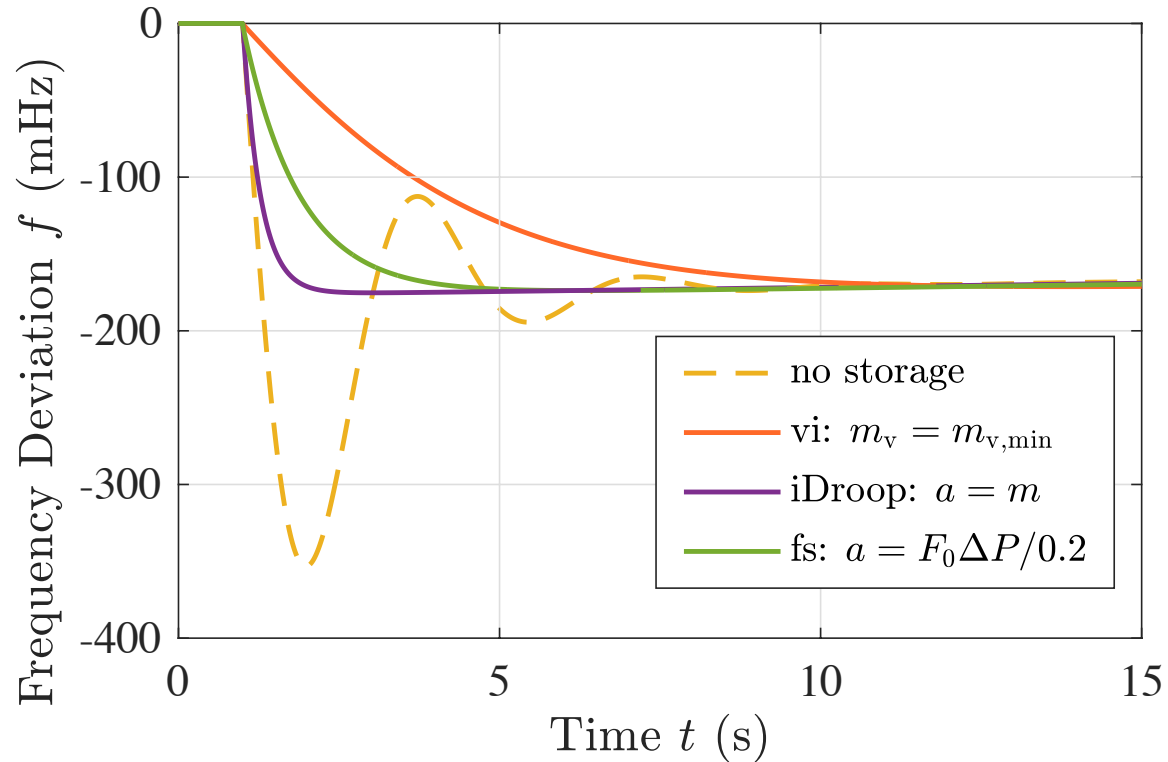
**Steady-state:**

$$\bar{w}(\infty) = \frac{\sum_i u_{0i}}{\sum_i f_i} \frac{1}{b}$$

# Trading off Control Effort and RoCoF



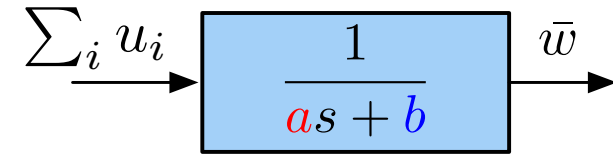
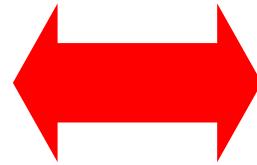
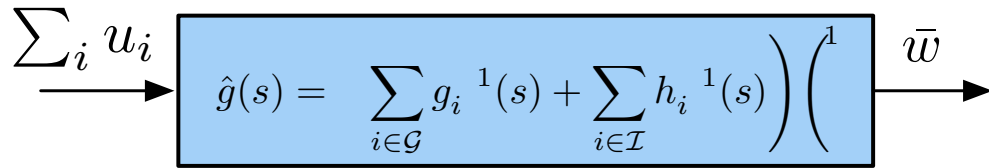
# Trading off Control Effort and RoCoF



**Challenge: Solution Limited to Grid-following Inverters**

# Grid-forming Frequency Shaping Control

**Key idea:** use model matching control on coherent dynamics



**Generation:**

$$g_i(s) = \frac{1}{m_i s + d_i + \frac{r_i^{-1}}{\tau_i s + 1}}, \quad i \in \mathcal{G}$$

$$a := \sum_{i \in \mathcal{G}} m_i + \sum_{i \in \mathcal{I}} m_i$$

$$b := \sum_{i \in \mathcal{G}} (d_i + r_i^{-1}) + \sum_{i \in \mathcal{I}} d_i$$

$$\sum_{i \in \mathcal{I}} c_i(s) = \sum_{i \in \mathcal{G}} \frac{r_i^{-1} \tau_i s}{\tau_i s + 1}$$

**Inverters:**

$$h_i(s) = \frac{1}{m_i s + d_i + c_i(s)}, \quad i \in \mathcal{I}$$

**RoCoF:**

$$\|\dot{\bar{w}}\|_{\infty} = \frac{|\sum_i u_{0i}|}{a}$$

**Steady-state:**

$$\bar{w}(\infty) = \frac{\sum_i u_{0i}}{b}$$

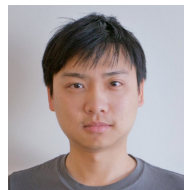
# Summary

- Frequency domain characterization of **coherent dynamics**, as a low rank property of the transfer function.
- **Coherence is a frequency dependent** property:
  - Effective algebraic connectivity  $f(s)\lambda_2(L)$
  - Disturbance frequency spectrum
- We use frequency **weighted balanced truncation** to suggest possible improvements to obtain accurate reduced order model of aggregated dynamics of coherent generators:
  - increase model complexity (3<sup>rd</sup> order/two turbines)
  - model reduction on closed-loop dynamics
- Grid-forming Frequency Shaping Control

# Thanks!

## Related Publications:

- Min, M, “Coherence and Concentration in Tightly Connected Networks,” **submitted**
- Min, Paganini, M, “Accurate Reduced Order Models for Coherent Synchronous Generators,” **L-CSS 2021**
- Jiang, Bernstein, Vorobev, M, “Grid-forming Frequency Shaping Control,” **L-CSS 2021**



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Petr Vorobev



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Fernando Paganini

