



# LMI-Based Model Reduction with Structural Constraints

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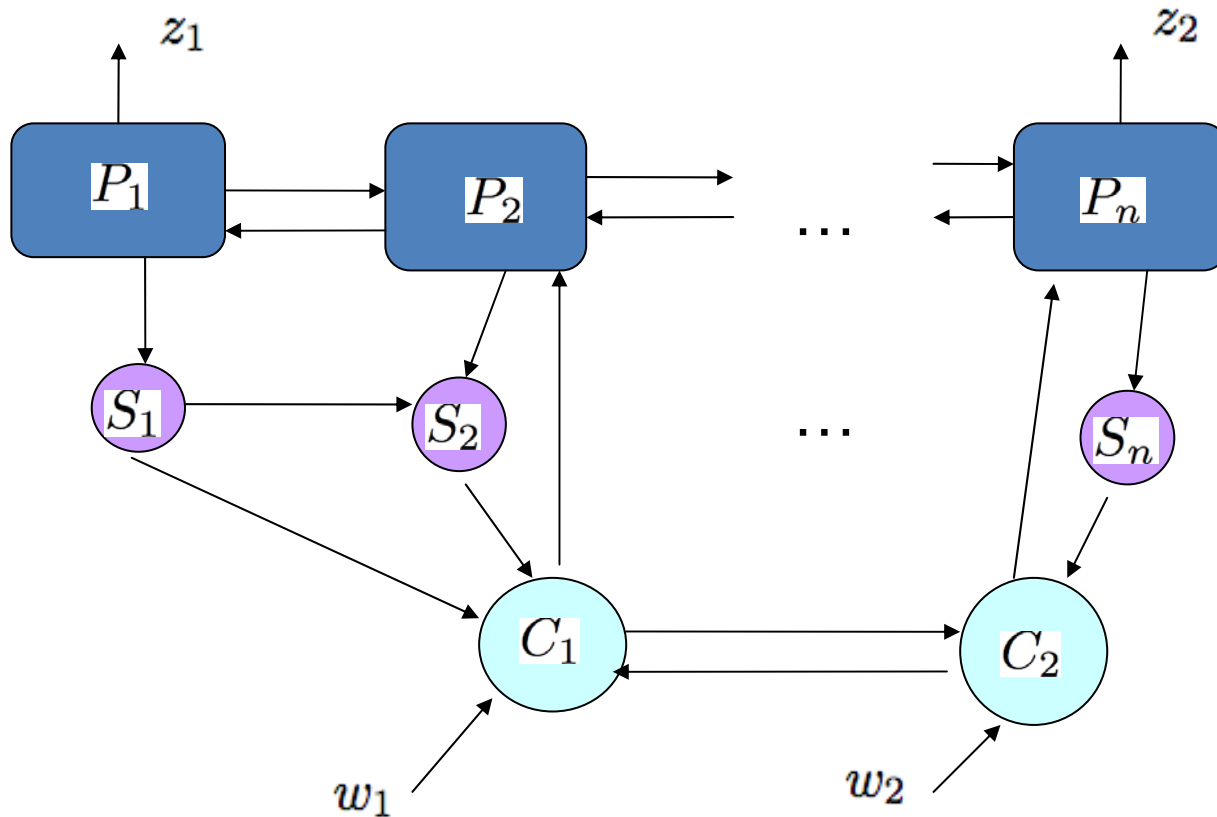
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# Outline

- Structured model reduction
- Structured (generalized) Gramians
- Model structures we can preserve:
  - Series/cascade and parallel systems
  - Feedback of passive systems
  - Port-Hamiltonian systems
- Example
  - Model reduction of boiler-header system

# Structured Model Reduction

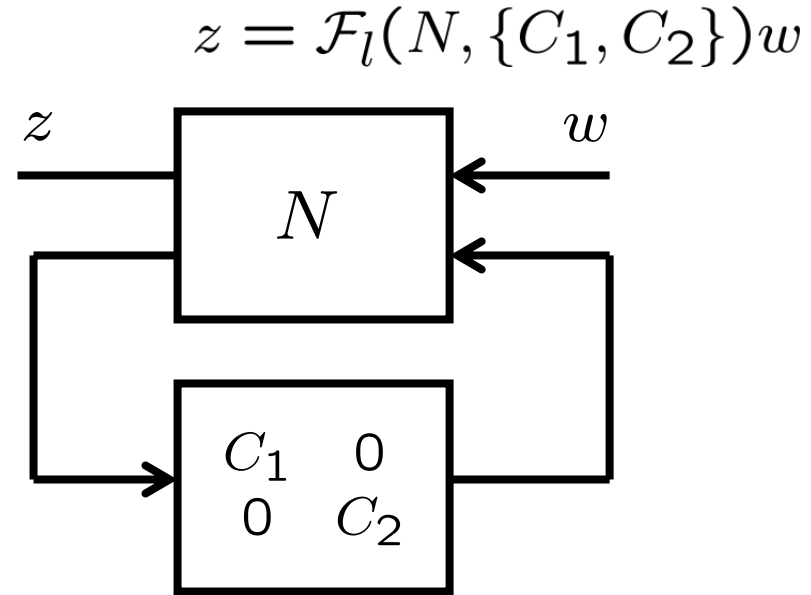
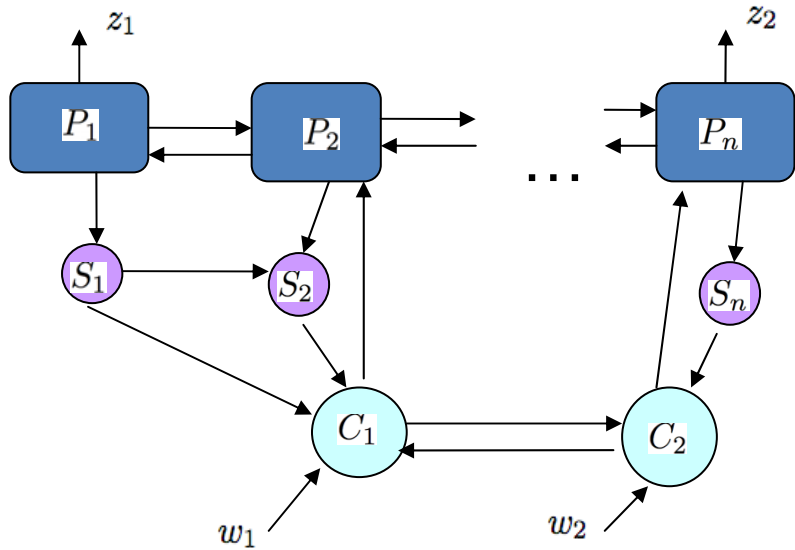


- Example: Distributed controller  $C_1, C_2$  in a networked control system
- **Problem:** Find (local) approximations of  $C_1, C_2$  such that (global) mapping  $(w_1, w_2) \mapsto (z_1, z_2)$  is well preserved

# Related Work

- Controller reduction
  - Survey by Anderson and Liu in IEEE TAC, 1989
  - Closed-loop model reduction by Schelfout and De Moor in IEEE TAC, 1996
- Problem often encountered outside the control area
  - *Industrial processes*
  - Physical models
  - Chemical models
  - Biological models
- The reduction method should take *topological structure constraints* into account
- Hard problems! Do not expect optimal solutions
- But what can we do?

# Equivalent Problems



$$\begin{bmatrix} \dot{x}_N \\ \dot{x}_{C_1} \\ \dot{x}_{C_2} \end{bmatrix} = \underbrace{\begin{bmatrix} A_N & A_{NC_1} & A_{NC_2} \\ A_{C_1N} & A_{C_1} & A_{C_1C_2} \\ A_{C_2N} & A_{C_2C_1} & A_{C_2} \end{bmatrix}}_{=A} \begin{bmatrix} x_N \\ x_{C_1} \\ x_{C_2} \end{bmatrix} + \underbrace{\begin{bmatrix} B_N \\ B_{C_1} \\ B_{C_2} \end{bmatrix}}_{=B} w$$

$$z = \underbrace{\begin{bmatrix} C_N & C_{C_1} & C_{C_2} \end{bmatrix}}_{=C} x$$

# A Bad Heuristic

- Ignore surrounding subsystems and compute local Gramians

$$\begin{cases} A_{C_1}P_{C_1} + P_{C_1}A_{C_1}^T + \bar{B}_{C_1}\bar{B}_{C_1}^T = 0 \\ A_{C_1}^TQ_{C_1} + Q_{C_1}A_{C_1} + \bar{C}_{C_1}^T\bar{C}_{C_1} = 0 \end{cases} \quad \begin{cases} A_{C_2}P_{C_2} + P_{C_2}A_{C_2}^T + \bar{B}_{C_2}\bar{B}_{C_2}^T = 0 \\ A_{C_2}^TQ_{C_2} + Q_{C_2}A_{C_2} + \bar{C}_{C_2}^T\bar{C}_{C_2} = 0 \end{cases}$$

( $\bar{B}_{C_1}, \bar{C}_{C_1}, \bar{B}_{C_2}, \bar{C}_{C_2}$  denote local input and output matrices)

- Balance the Gramians  $(P_{C_1}, Q_{C_1}), (P_{C_2}, Q_{C_2})$ , then truncate
- Well known from controller reduction this can give very bad/conservative *global* approximations

# A Good Heuristic

- Compute Gramians 
$$\begin{cases} AP + PA^T + BB^T = 0 \\ A^T Q + QA + C^T C = 0 \end{cases}$$

$$P = \begin{bmatrix} P_N & * & * \\ * & P_{C_1} & * \\ * & * & P_{C_2} \end{bmatrix}, \quad Q = \begin{bmatrix} Q_N & * & * \\ * & Q_{C_1} & * \\ * & * & Q_{C_2} \end{bmatrix}$$

- Balance the sub-Gramians  $(P_{C_1}, Q_{C_1}), (P_{C_2}, Q_{C_2})$
- A *local* coordinate transformation in  $C_1, C_2$  with a *global* objective
- Often works very well. No stability and error bounds, however... How can we get bounds?

# Structured (Generalized) Gramians

- Compute generalized Gramians (LMIs):

$$\min \text{trace } P : \quad AP + PA^T + BB^T \leq 0 \quad (1a)$$

$$\min \text{trace } Q : \quad A^T Q + QA + C^T C \leq 0$$

- Enforce structure constraints (“structured Gramians”)

$$P = \begin{bmatrix} P_N & 0 & 0 \\ 0 & P_{C_1} & 0 \\ 0 & 0 & P_{C_2} \end{bmatrix}, \quad Q = \begin{bmatrix} Q_N & 0 & 0 \\ 0 & Q_{C_1} & 0 \\ 0 & 0 & Q_{C_2} \end{bmatrix} \quad (1b)$$

- *Local* approximations, *global* error bound:

$$\|z - \hat{z}\|_2 \leq 2 \left( \sum_{i=r_1+1}^{n_1} \sigma_{C_1,i} + \sum_{i=r_2+1}^{n_2} \sigma_{C_2,i} \right) \|w\|_2, \quad \sigma_{C_k,i} = \sqrt{\lambda_i(P_{C_k} Q_{C_k})}$$

**Key problem:** When are the LMIs (1a)-(1b) feasible?



# Key Problem: When Can We Ensure Existence of Structured Gramians?

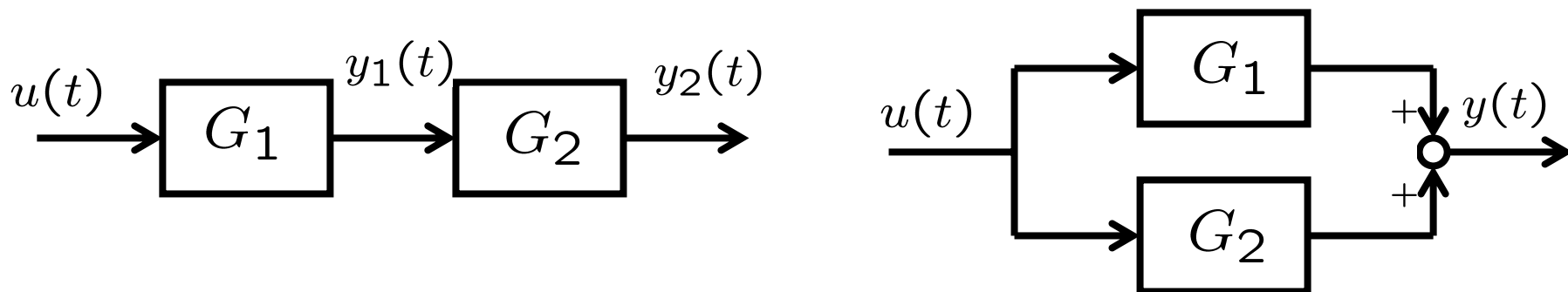
Existence can be proven for the following structures:

1. Series/cascade and parallel systems
2. Feedback of passive systems
3. Port-Hamiltonian systems
4. ...

General feedback interconnections remain hard!

Existence is in fact equivalent to system being *strongly stable*, see [Beck, 2007].

# 1. Series/Cascade and Parallel Systems

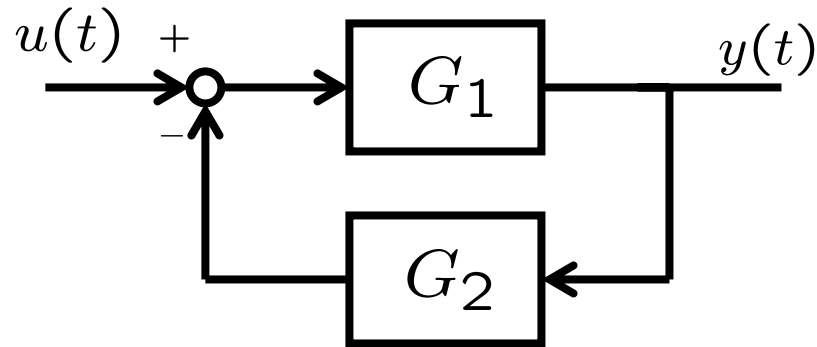


Assume  $A = \begin{pmatrix} A_1 & 0 \\ A_{21} & A_2 \end{pmatrix}$ , where  $A_1$  and  $A_2$  are stable. Then there exists a structured controllability Gramian  $P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix} > 0$  and a

structured observability Gramian  $Q = \begin{pmatrix} Q_1 & 0 \\ 0 & Q_2 \end{pmatrix} >$

$0$  satisfying  $AP + PA^T + BB^T < 0$  and  $A^T Q + QA + C^T C < 0$ , respectively.

## 2. Feedback of (Strictly) Passive Systems



The negative feedback interconnection of two strictly positive real systems with dimensions  $n_1, n_2$  admits a structured controllability Gramian

$P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}$  and a structured observabil-

ity Gramian  $Q = \begin{pmatrix} Q_1 & 0 \\ 0 & Q_2 \end{pmatrix}$ , where  $P_1, Q_1 \in R^{n_1 \times n_1}$  and  $P_2, Q_2 \in R^{n_2 \times n_2}$ .

# 3. Port-Hamiltonian Systems

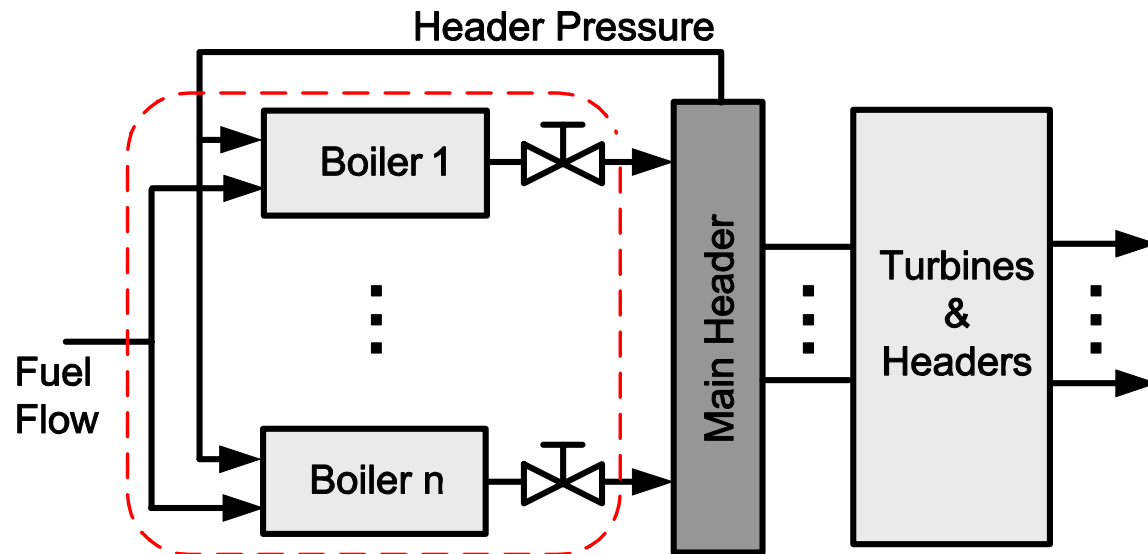
$$\begin{aligned}\dot{x} &= (J - R)Hx + Bu, & J &= -J^T, R = R^T \geq 0 \\ y &= B^T Hx, & H &= H^T > 0\end{aligned}$$

Suppose the Hamiltonian has the structure  $H = H^T = \begin{pmatrix} H_1 & 0 \\ 0 & H_2 \end{pmatrix}$ , and that the input matrix  $B$  lies in the range of the dissipation matrix  $R$  (see note), i.e.,  $B = \mathcal{R}(R)$ . Then the port-Hamiltonian system admits structured Gramians  $P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}$  and  $Q = \begin{pmatrix} Q_1 & 0 \\ 0 & Q_2 \end{pmatrix}$ .

Note: This is often ensured by the *fluctuation-dissipation theorem*.

# Example of Structured Model Reduction: Boiler-Header System

- Multiple configurations of boilers
- MPC requires models for each configuration
- Combine identified boiler and header models to form the full system  $\mathcal{F}_l(N, G)$



[Sturk et al., 2010]

Honeywell

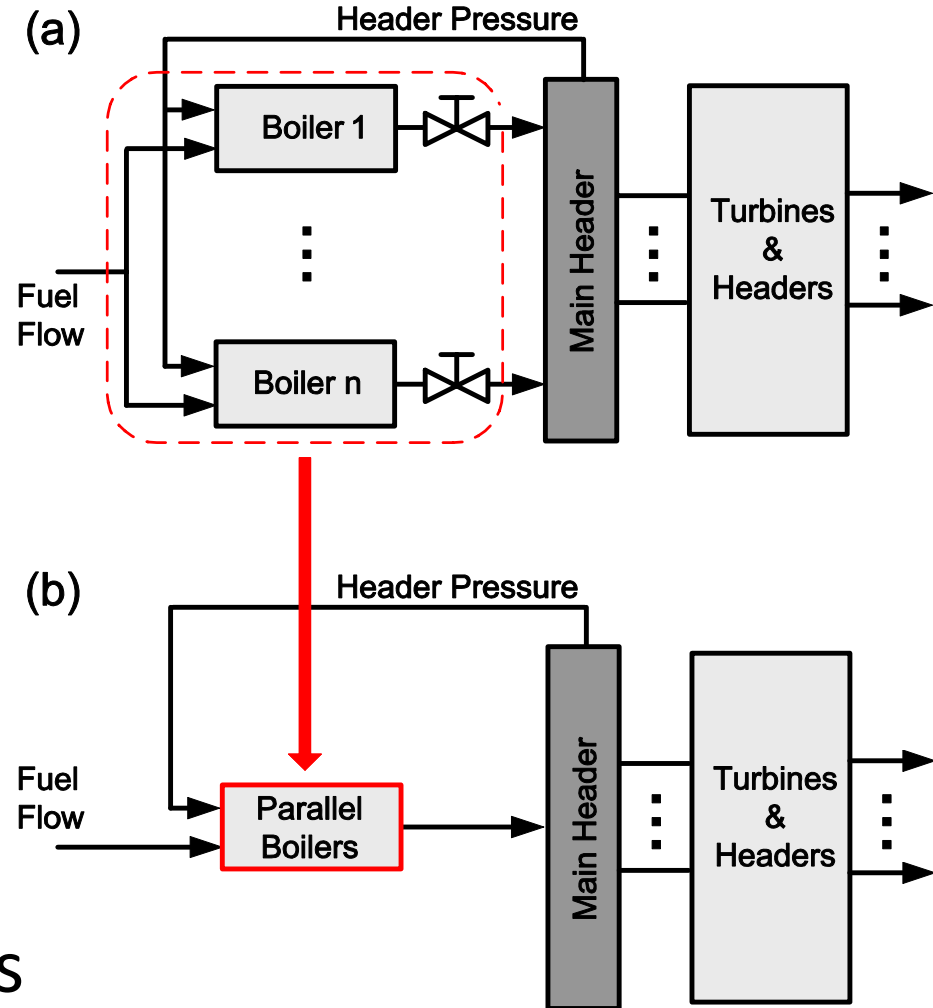


# Problem Formulation

- Inputs and outputs

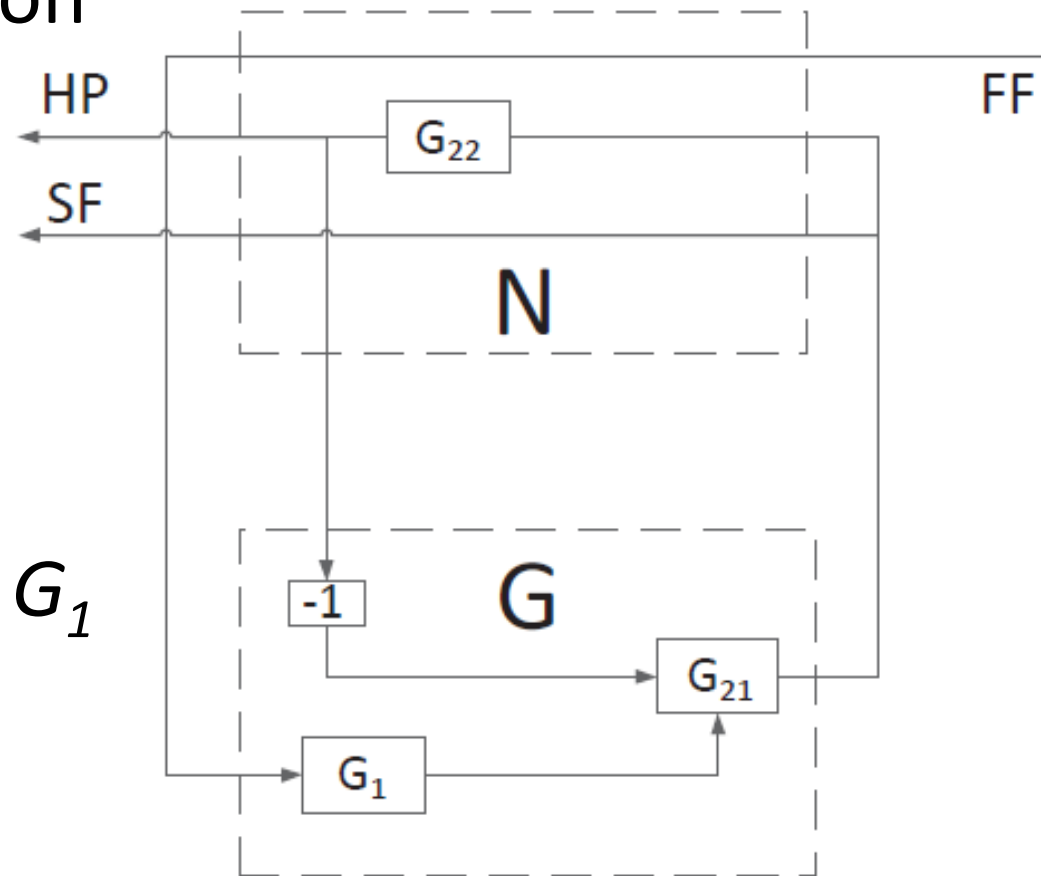
$$w = \begin{pmatrix} FF \\ SD \end{pmatrix}, \quad z = \begin{pmatrix} HP \\ SF \end{pmatrix}$$

- Retain physical interpretation of subsystems and interconnection signals
- Header has 1 state, boilers have 2 states each

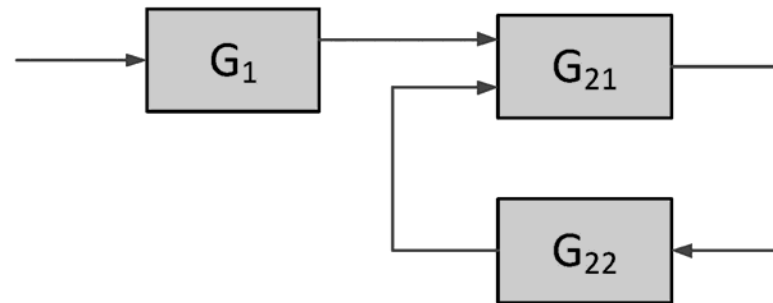


# Model $\mathcal{F}_l(N, G)$

- $G$  – Parallel connection of boilers
- $N$  – Header
- Series connection of  $G_1$  with the negative feedback loop of  $G_{21}$  and  $G_{22}$



# Existence of Structured Gramians (Combination of Cases 1. and 2.)



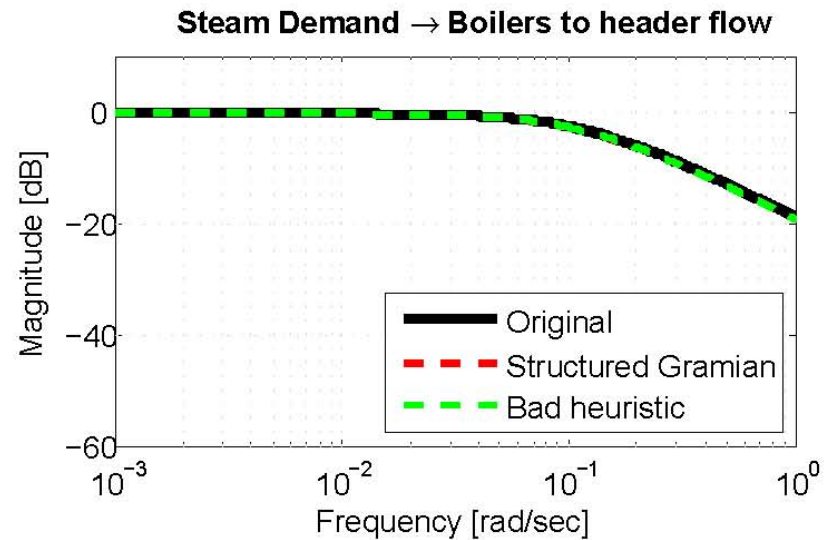
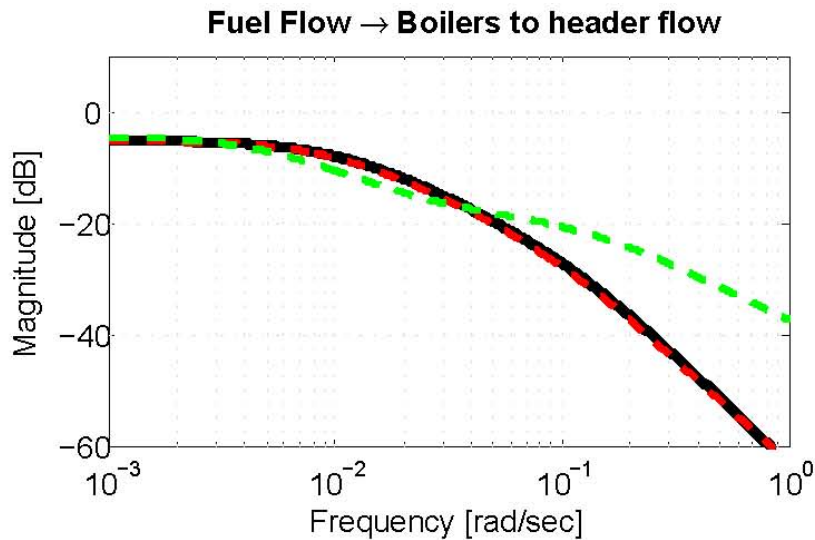
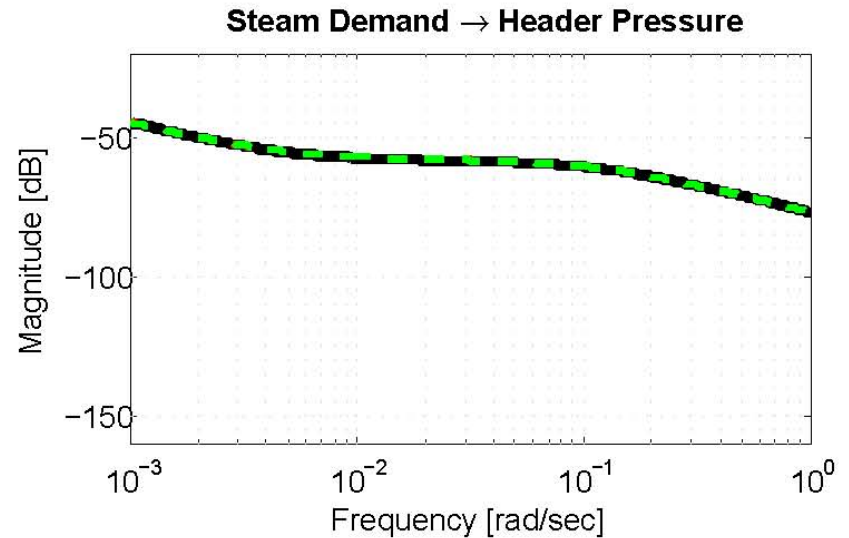
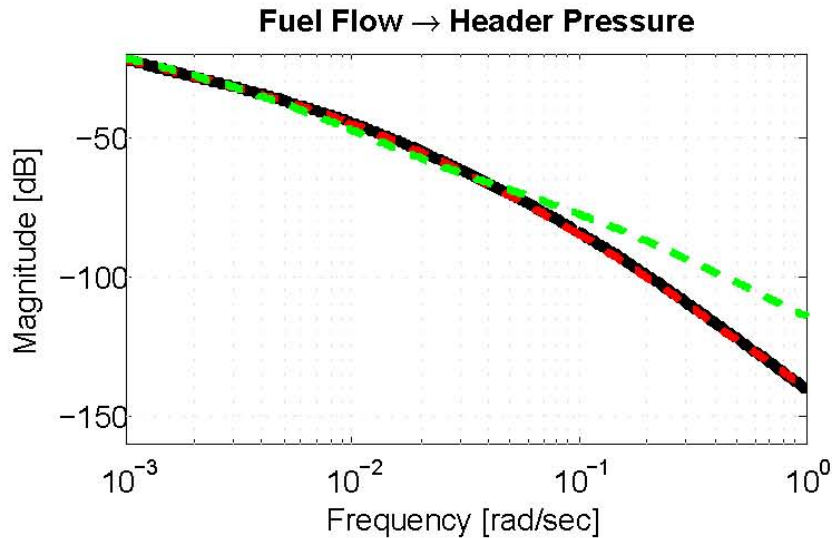
A stable LTI system of dimension  $n_1$  connected in series with two strictly positive real systems of order  $n_{21}$  and  $n_{22}$  connected with negative feedback admits a structured controllability Gramian  $P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}$  and a structured

observability Gramian  $Q = \begin{pmatrix} Q_1 & 0 \\ 0 & Q_2 \end{pmatrix}$ , where

$P_1, Q_1 \in R^{(n_1+n_{21}) \times (n_1+n_{21})}$  and  $P_2, Q_2 \in R^{n_{22} \times n_{22}}$ .

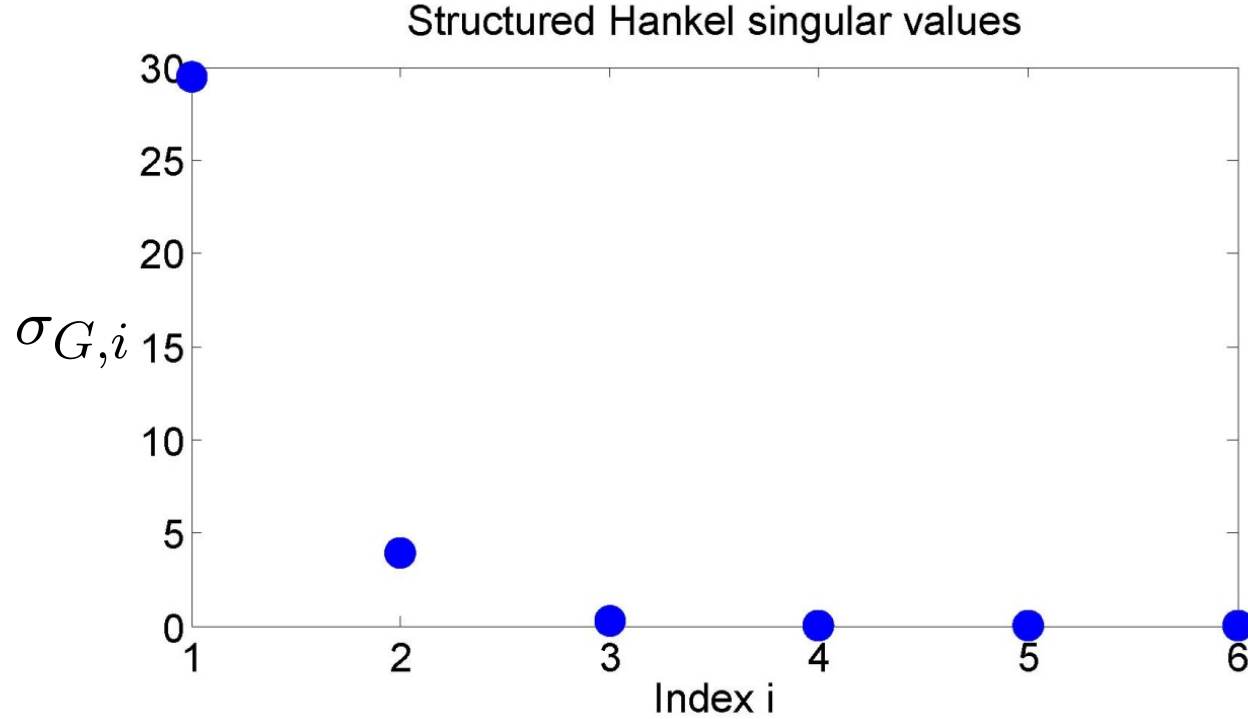


# Structured Gramians vs. Bad Heuristic



**Bad heuristic does not capture the global dynamics!**

# Hankel Singular Values and Error Bound



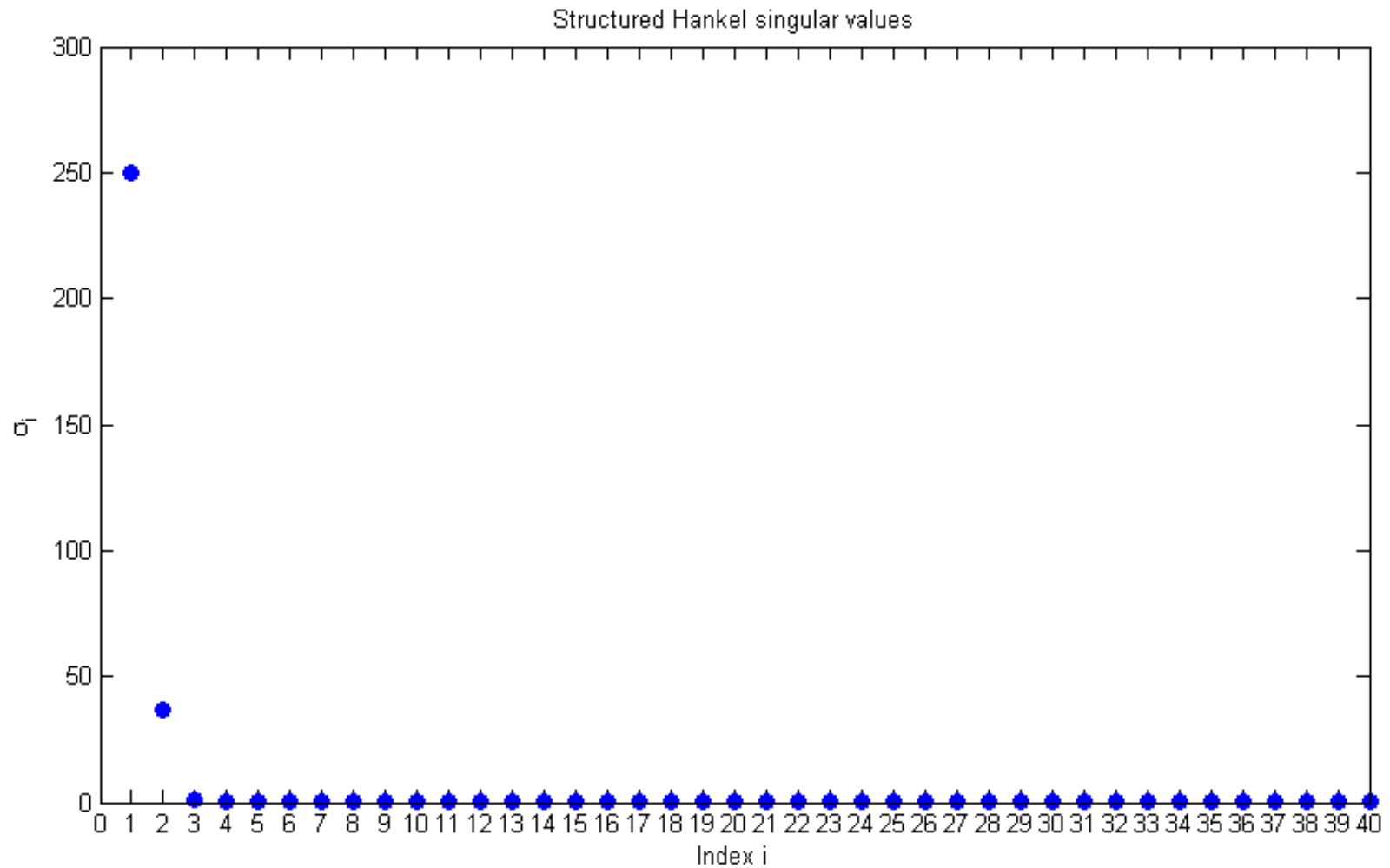
- Upper a priori error bound and a posteriori error:

$$\|\mathcal{F}_l(N, G) - \mathcal{F}_l(N, \hat{G})\|_\infty \leq 2 \sum_{i=3}^6 \sigma_{G,i} = 0.6274$$

$$\|\mathcal{F}_l(N, G) - \mathcal{F}_l(N, \hat{G})\|_\infty = 0.3146$$

- Interpretation: Three parallel boilers have been lumped into one, while preserving global performance

# Hankel Singular Values (20 Boilers)



# Other Uses of Generalized Gramians

- Used to improve error bounds in balanced truncation (enforce multiplicity of Hankel singular values), see Hinrichsen and Pritchard, 1990
- To obtain *optimal*  $H_\infty$ -model approximations, see Kavranoglu and Bettayeb, 1993
- Model reduction of uncertain systems, see Beck, Doyle and Glover, 1996
- ...
- **Computational problems:** Computationally expensive to solve LMIs;  $O(n^5)$ - $O(n^6)$  operations.
- **Idea:** If we know there are structured solutions to LMIs, look for non-optimal solutions by solving many much smaller Lyapunov equations.

# Key Problem: When Can We Ensure Existence of Structured Gramians?

Existence can be proven for the following structures:

- Series/cascade and parallel systems
- Feedback of passive systems
- Port-Hamiltonian systems

What to do when no structured Gramians can be found?

- Use the good heuristic without error bound
- Search for *extended structured Gramians*

# Extended Gramians (in Discrete Time)

Lyapunov inequalities  $P - APA^T - BB^T > 0, \quad P > 0, \quad (2)$

$Q - A^TQA - C^TC > 0, \quad Q > 0, \quad (3)$

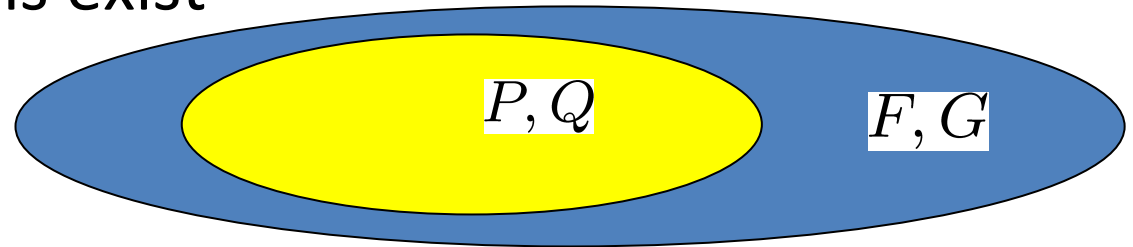
Extended Lyapunov inequalities  $\begin{bmatrix} P & AF & B \\ F^T A^T & F + F^T - P & 0 \\ B^T & 0 & I \end{bmatrix} > 0, \quad (4)$

$\begin{bmatrix} G + G^T - Q & GA & 0 \\ A^T G^T & Q & C^T \\ 0 & C & I \end{bmatrix} > 0, \quad (5)$

*Theorem* The inequalities (2)–(3) have solutions  $P = P^T, Q = Q^T$  if, and only if, the inequalities (4)–(5) have solutions  $P = P^T, Q = Q^T$  and  $F, G$ .

# Extended Balanced Truncation

- Larger feasibility set  $\Rightarrow$  More likely structured extended Gramians exist



- Balance the extended Gramians and truncate  $F=G=\Sigma_e$
- Extended Hankel singular values give error bound:

$$\Sigma_e = \text{diag}\{\sigma_{e,1}, \dots, \sigma_{e,n}\}$$
$$\|z - \hat{z}\|_2 \leq \left( 2 \sum_{i=r+1}^n \sigma_{e,i} \right) \|w\|_2$$

[S., ACC 2008, TAC 2010]

- Example in [S.,2010]: Structured extended balanced truncation works with 22% larger feedback gain.

# Summary

- Avoid bad heuristic. Take surrounding into account
- Good heuristic often works well. Error bounds if structured generalized (or extended) Gramians exist
- Structured generalized Gramians do exist in many cases! (Series/cascade, parallel, passive, port-Hamiltonian systems)
- Method illustrated on boiler-header system