

An Adaptive Algorithm for Fully Automated Extraction of Passive Parameterized Macromodels

Alessandro Zanco, Elisa Fevola, Stefano Grivet-Talocia, Tommaso Bradde, Marco De Stefano

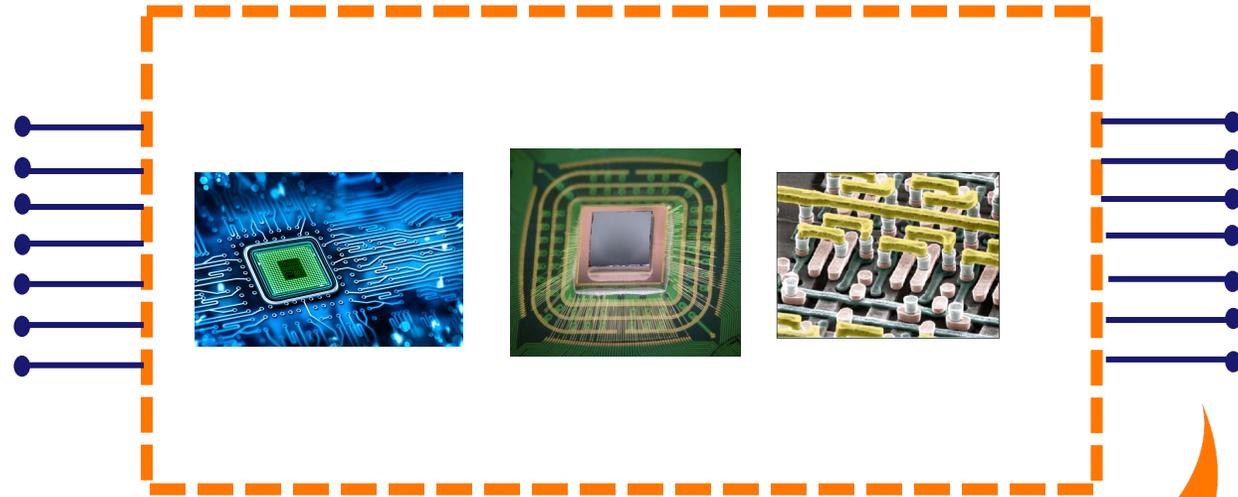
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**POLITECNICO
DI TORINO**

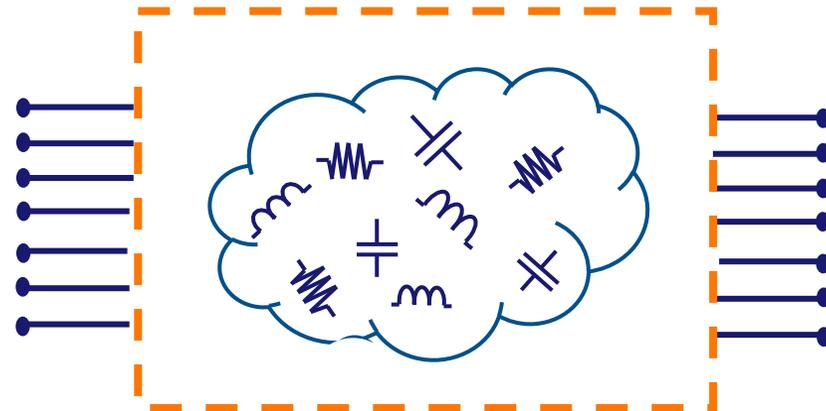
Background: Macromodeling



Complex EM
structure
Field solutions = hours

Advantages:

- Compact
- Accurate
- Reduced-order
- SPICE-compatible

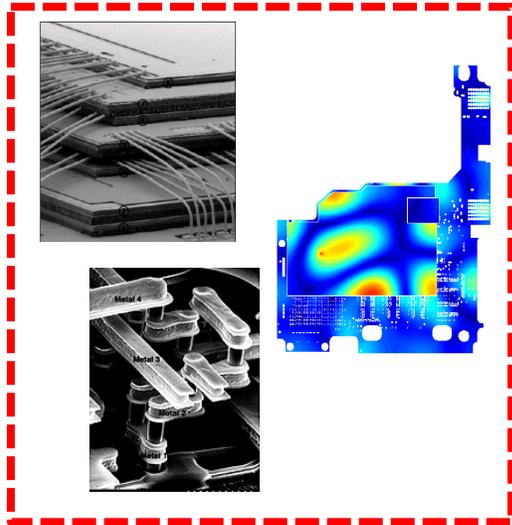


Macromodel

Model solution = seconds

- Reused many times
- Ideal for *optimization*

Background: Data-driven macromodeling flow



Extraction
EM simulation
→
(Measurement)

$$\check{H}_k = \check{H}(j\omega_k) \\ k = 1, \dots, K$$



Rational
Fitting

Passivity
Enforcement



State-space
realization

$$H(s) = \frac{N(s)}{D(s)}$$



OUT

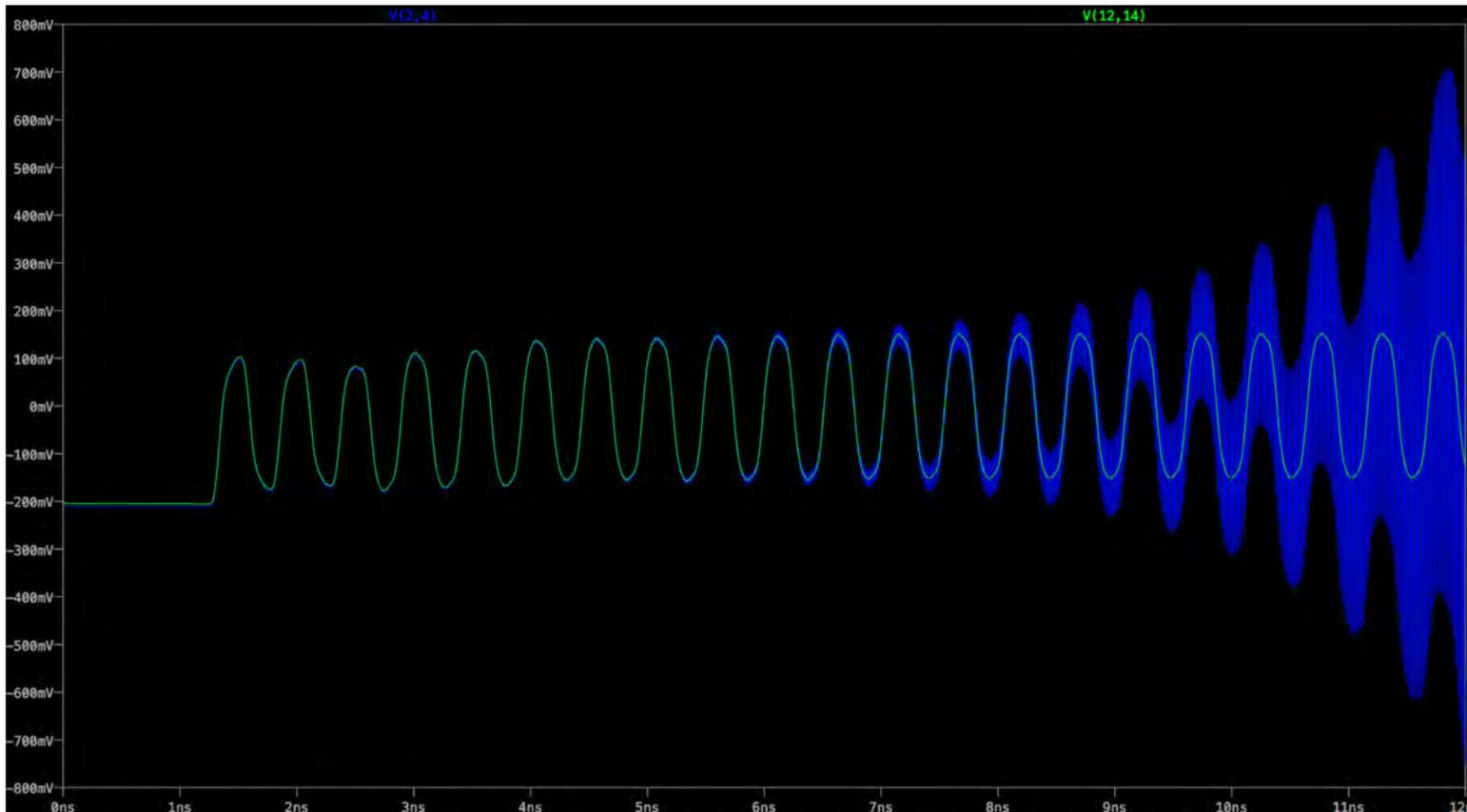


SPICE netlist

$$\dot{w} = Aw + Bu \\ y = Cw + Du$$

Background: Macromodeling

Why a *passive* model?



A passive model is ***guaranteed stable***

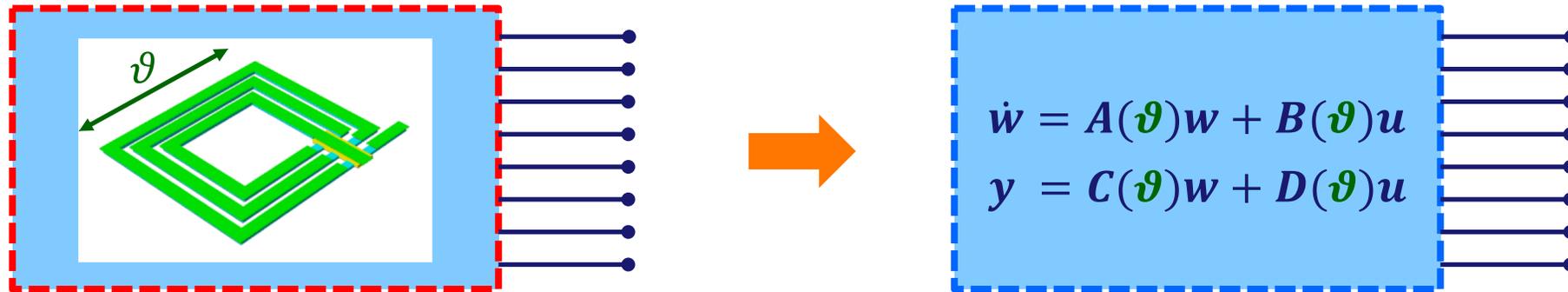
Background: Parameterized Macromodeling

Pre-layout, structure is not fixed but can be parameterized

Input: sets of S-params from field solver (as few as possible)

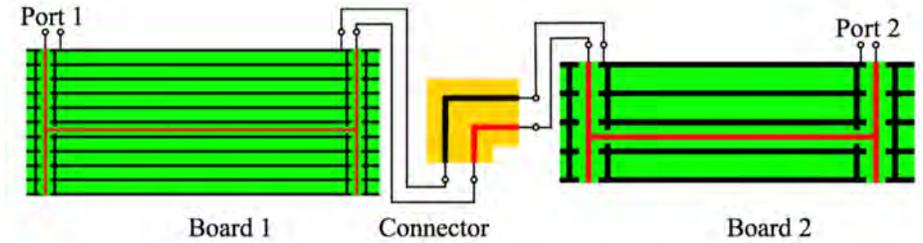
Objective: optimization, what-if, DOE, minimize solver runs

Requirement: parameterized model (SPICE if possible)

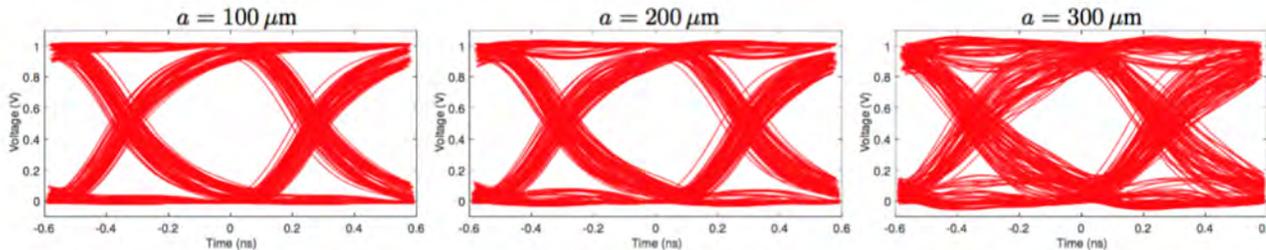
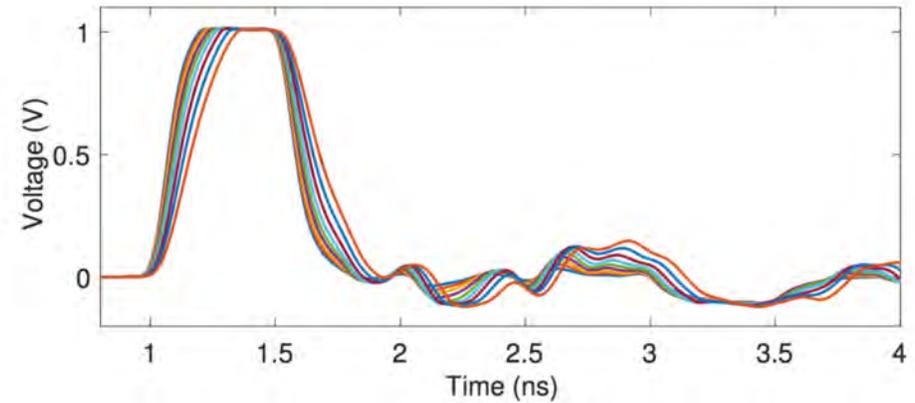
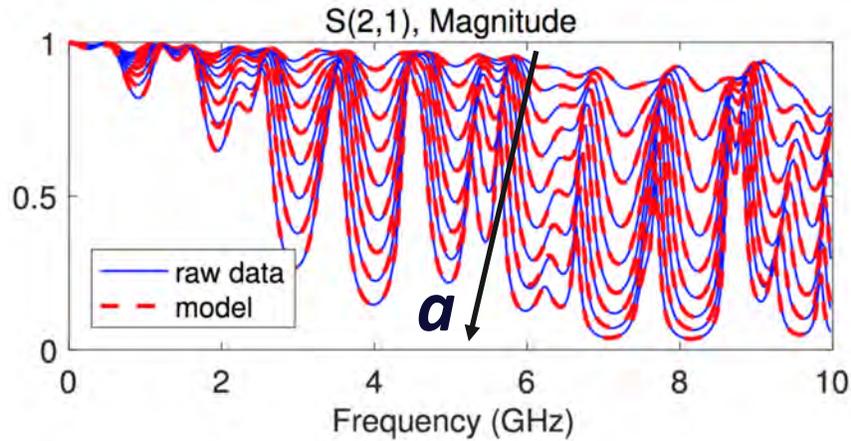


Parameterized Macromodeling: an example

Multiboard PCB link
parameter: via radius a

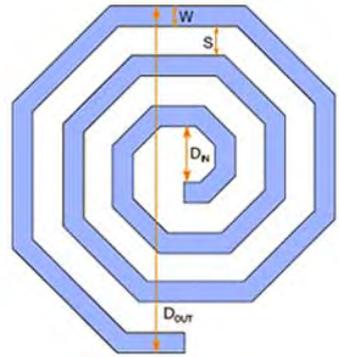


[Courtesy: J. B. Preibisch and C. Schuster, Technische Universität Hamburg-Harburg, Hamburg, Germany]

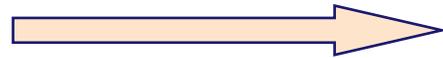


SPICE transient simulations
with NL terminations

Parameterized Macromodeling Flow



Parametric sweep



Simulation or
Measurement

Multiple S-parameters

$$\check{H}_{k;m} = \check{H}(j\omega_k; \vartheta_m)$$

$$k = 1, \dots, K; m = 1, \dots, M$$

IN
.snp



Multivariate
Rational
Fitting

Stability,
Passivity
Enforcement



$$H(s; \vartheta) = \frac{N(s; \vartheta)}{D(s; \vartheta)}$$

Parameterized model

State-space
realization



Circuit
synthesis

OUT



SPICE netlist

$$\dot{w} = A(\vartheta)w + B(\vartheta)u$$

$$y = C(\vartheta)w + D(\vartheta)u$$

Fitting data with a parameterized model

$$\mathbf{H}(s_k; \boldsymbol{\vartheta}_m) \approx \check{\mathbf{H}}(s_k; \boldsymbol{\vartheta}_m)$$

↑
Model

↑
Fitting



Frequency responses
from field solver

Frequency sweep

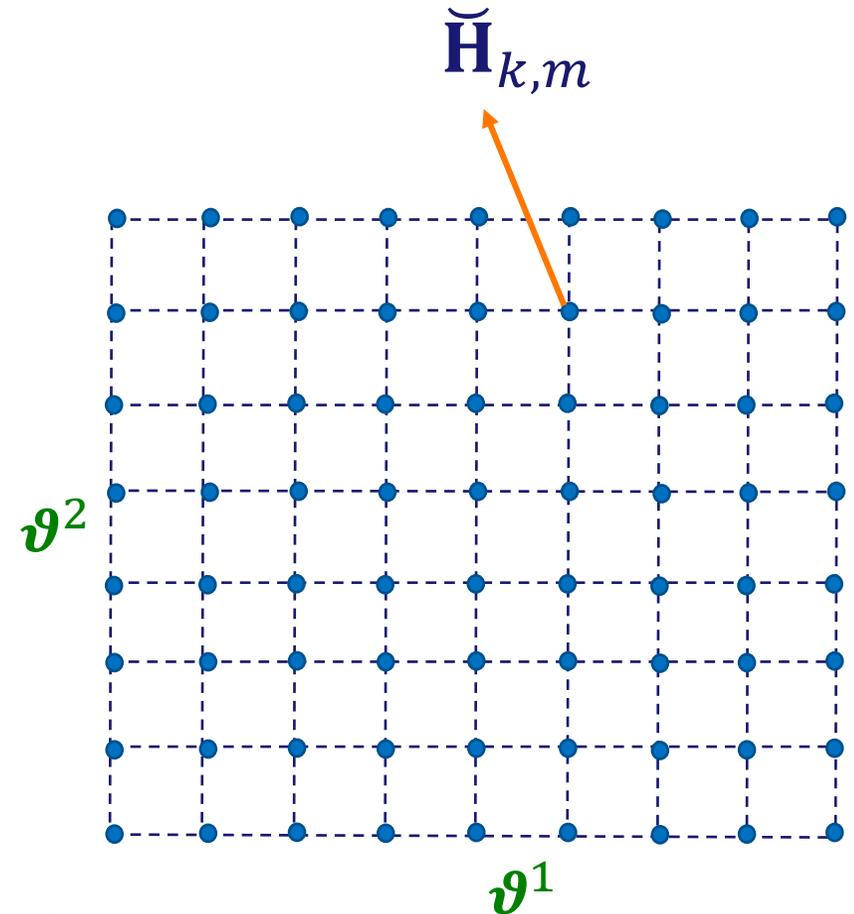
$$s_k = j 2\pi f_k: k = 1, \dots, K$$

Parameter sweep

$$\boldsymbol{\vartheta}_m: m = 1, \dots, M$$

PROBLEMS

- Large number of simulations
- Expensive in terms of time and resources
- Not feasible for large number of parameters



D. Deschrijver, T. Dhaene and D. De Zutter, "Robust Parametric Macromodeling Using Multivariate Orthonormal Vector Fitting," in IEEE Trans. MTT, vol. 56, no. 7, pp. 1661-1667, July 2008

P. Triverio, S. Grivet-Talocia and M. S. Nakhla, "A Parameterized Macromodeling Strategy With Uniform Stability Test," in IEEE Transactions on Advanced Packaging, vol. 32, no. 1, pp. 205-215, Feb. 2009

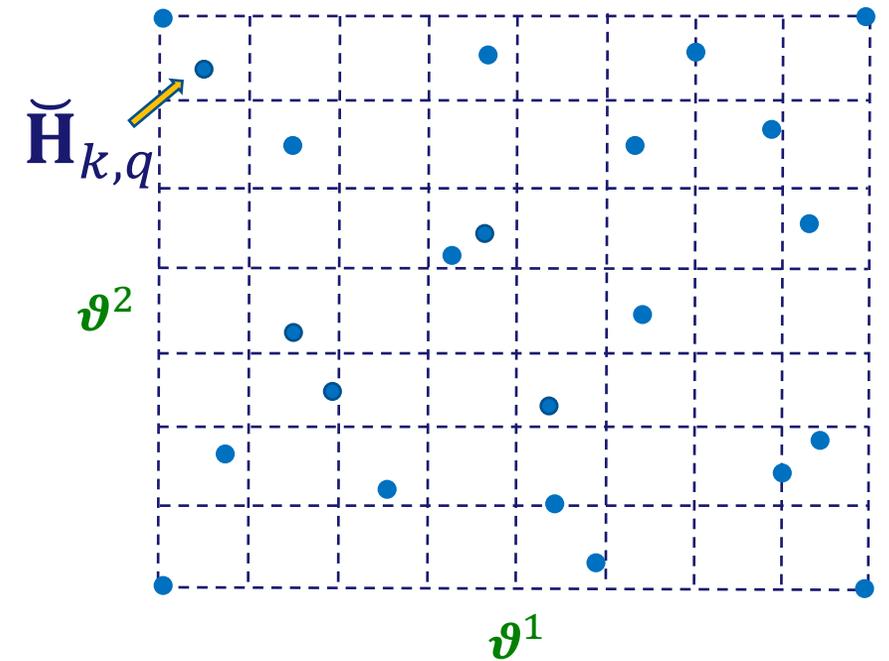
Proposed Approach

Adaptive sampling scheme for point selection



GOALS:

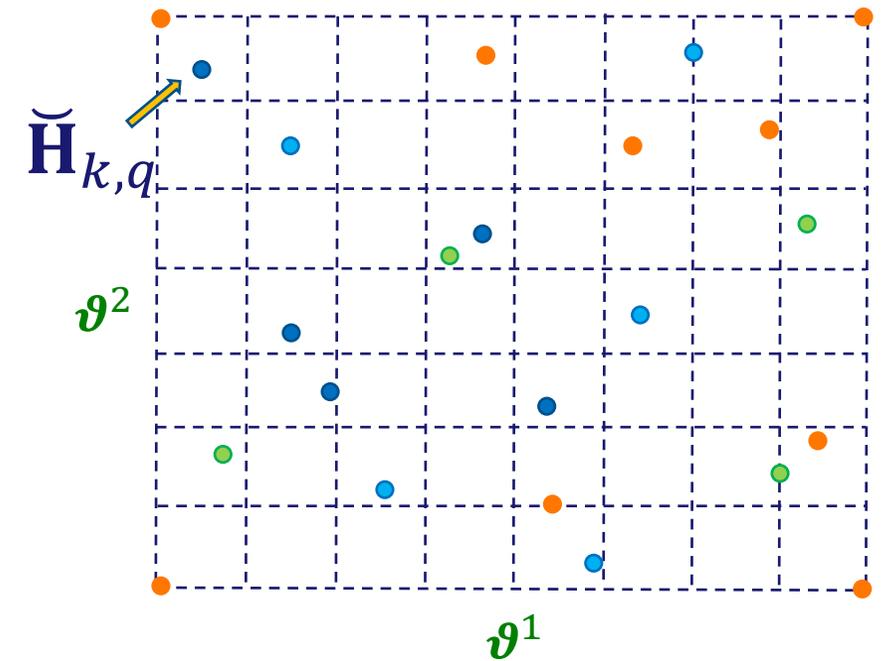
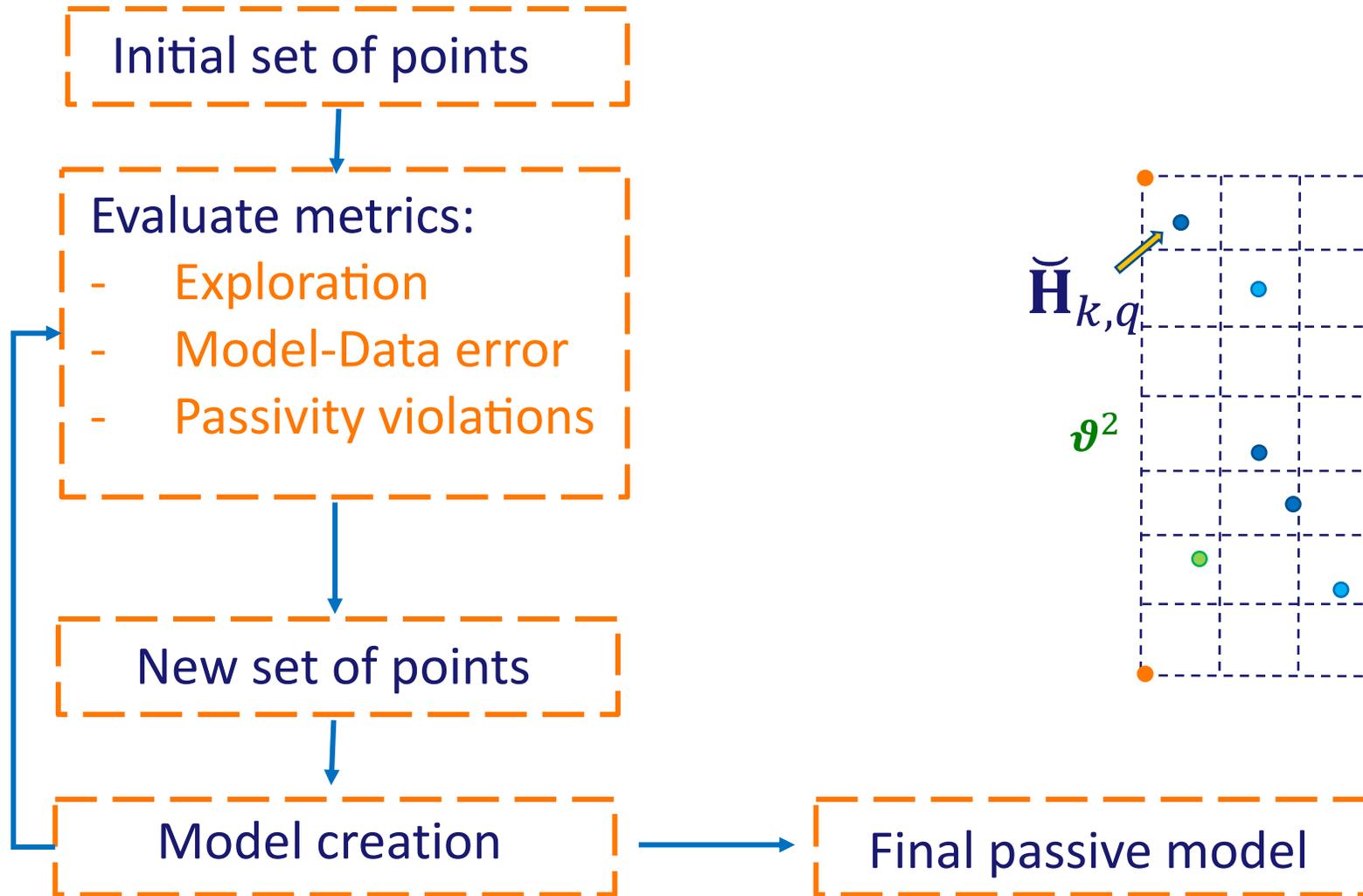
- Minimal number Q of responses
- Optimal parameter configurations ϑ_q
- Final stable and passive model $\mathbf{H}(s; \vartheta)$



MODEL-DRIVEN

D. Deschrijver, K. Crombeq, H.M. Nguyen, T. Dhaene, " Adaptive Sampling Algorithm for Macromodeling of Parameterized S-Parameter Responses," in IEEE Trans. MTT, vol. 59, no. 1, pp. 39-45, 2011

Adaptive Point Selection



Exploration

Identify regions with **fewer** data points

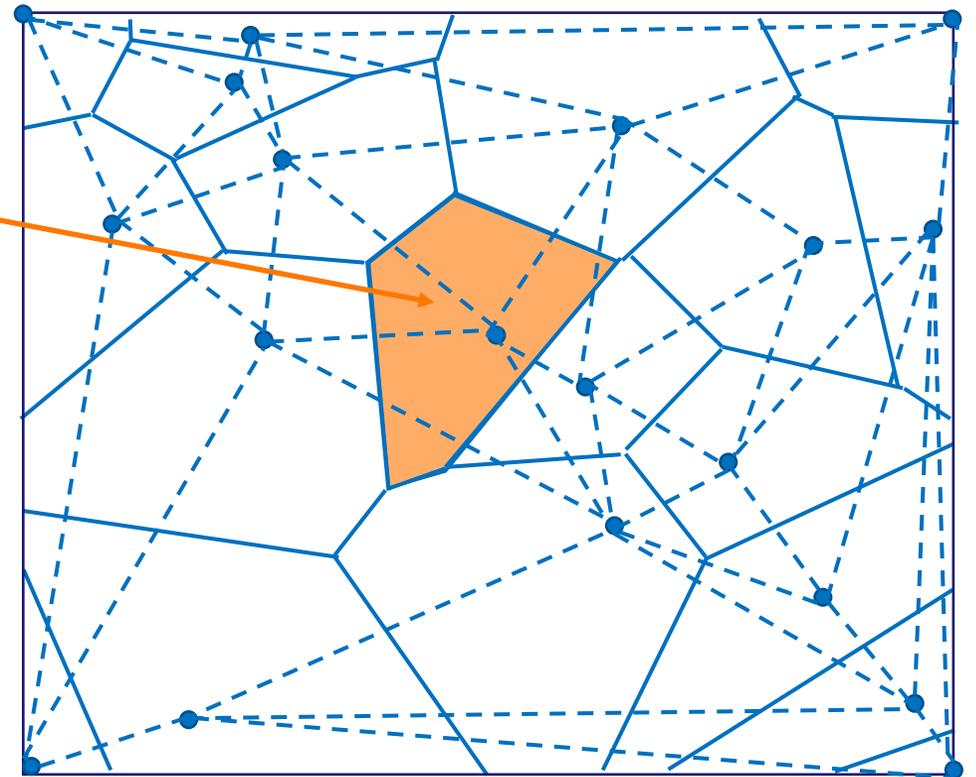


Volume of Voronoi cells $V(C_q)$

**EXPLORATION
METRIC**

$$\Lambda_1(\boldsymbol{\vartheta}_q) = \frac{V(C_q)}{\sum_{q=1}^Q V(C_q)}$$

Voronoi Tessellation



D. Deschrijver, K. Crombeq, H.M. Nguyen, T. Dhaene, " Adaptive Sampling Algorithm for Macromodeling of Parameterized S-Parameter Responses," in IEEE Trans. MTT, vol. 59, no. 1, pp. 39-45, 2011

Model-Data error

Identify regions where model is **less accurate**

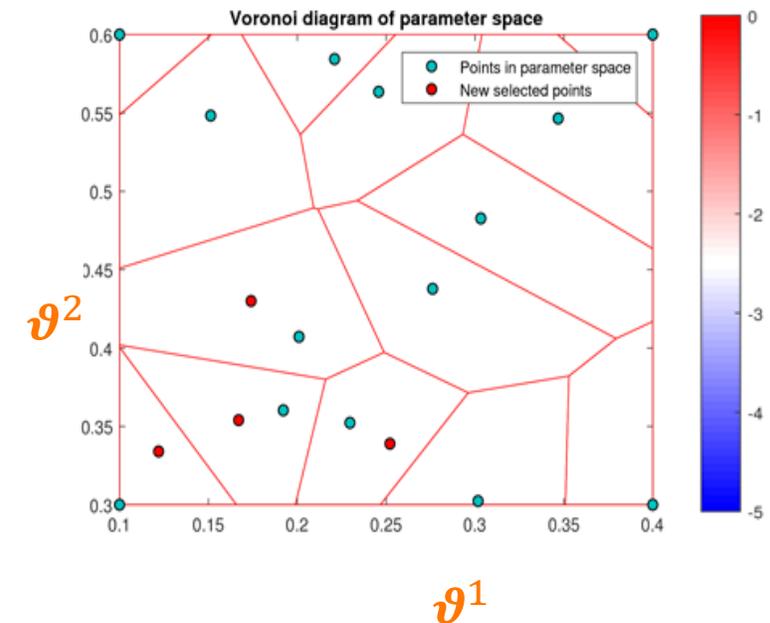
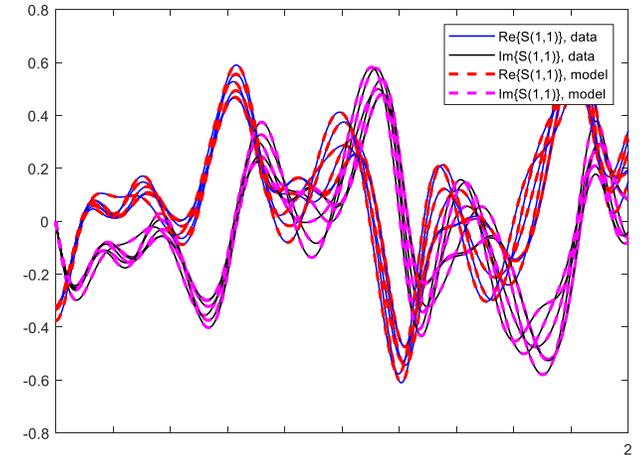


Error between data and intermediate model:

$$R^2(\boldsymbol{\vartheta}_q) = \frac{\sum_k \|\mathbf{H}(s_k, \boldsymbol{\vartheta}_q) - \check{\mathbf{H}}(s_k, \boldsymbol{\vartheta}_q)\|^2}{\sum_k \|\check{\mathbf{H}}(s_k, \boldsymbol{\vartheta}_q)\|^2}$$

**MODEL ERROR
METRIC**

$$\Lambda_2(\boldsymbol{\vartheta}_q) = \frac{R(\boldsymbol{\vartheta}_q)}{\sum_{q=1}^Q R(\boldsymbol{\vartheta}_q)}$$



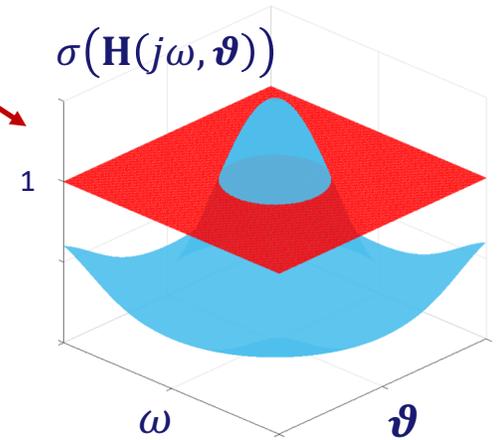
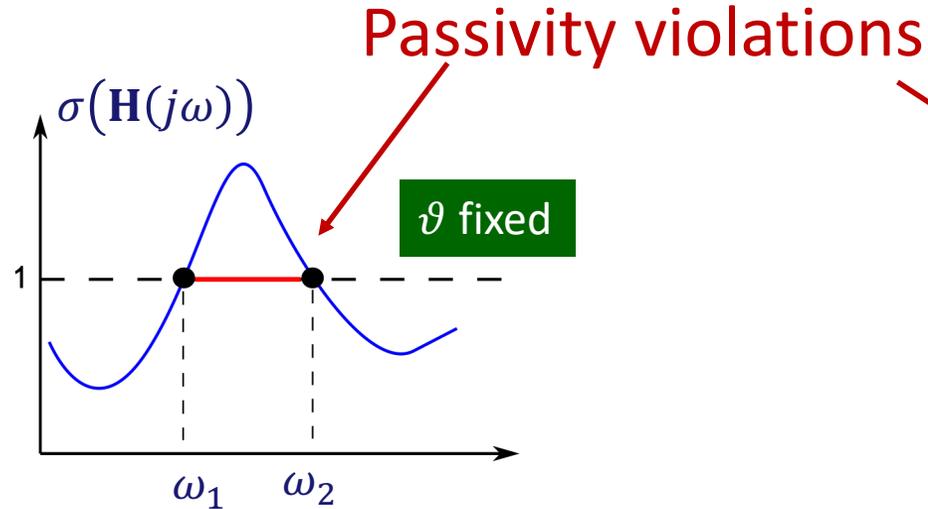
Passivity Violations

Passivity condition - Scattering

$$\sigma_{\max}\{\mathbf{H}(j\omega; \vartheta)\} \leq 1 \quad \forall \omega, \vartheta$$



No energy gain



How to detect violations?

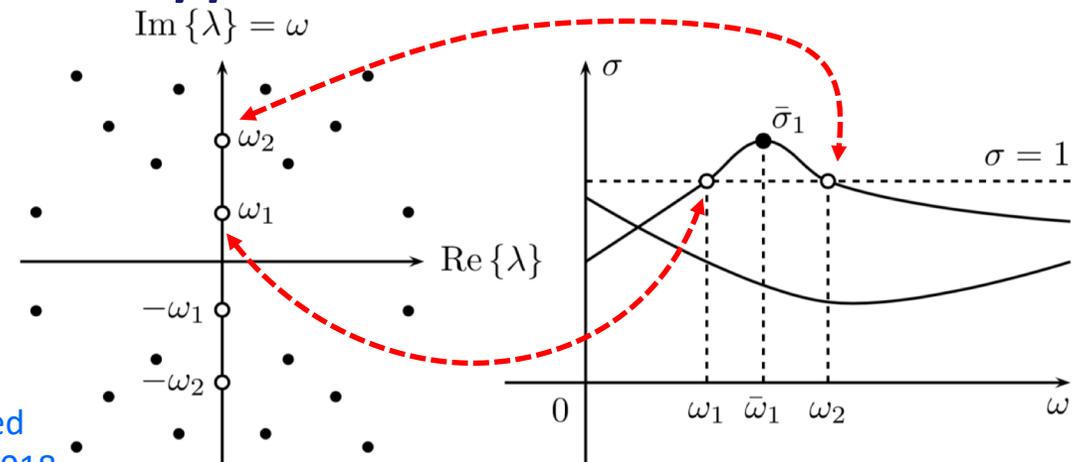


Hamiltonian Based approach

$$\mathbf{H}(s) \rightarrow \begin{cases} \mathbf{E} \dot{x} = \mathbf{A} x + \mathbf{B} u \\ y = \mathbf{C} x \end{cases}$$

ϑ fixed

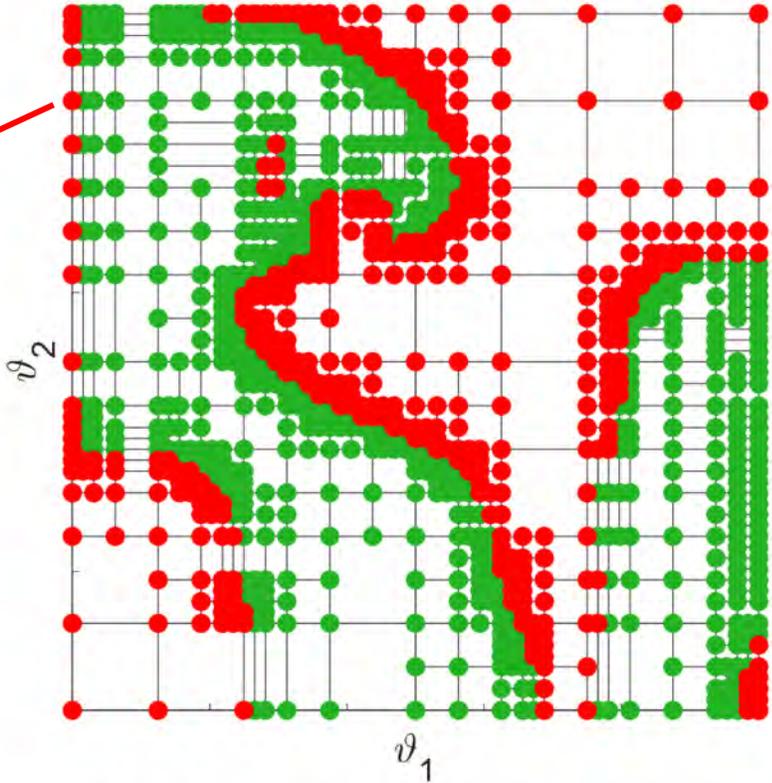
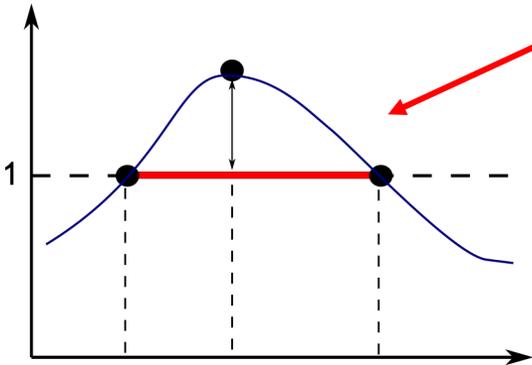
$$\mathcal{M} = \begin{pmatrix} \mathbf{A}^T & \mathbf{B}\mathbf{B}^T \\ -\mathbf{C}^T\mathbf{C} & -\mathbf{A}^T \end{pmatrix}, \quad \mathcal{K} = \begin{pmatrix} \mathbf{E} & \mathbf{0} \\ \mathbf{0} & \mathbf{E}^T \end{pmatrix}$$



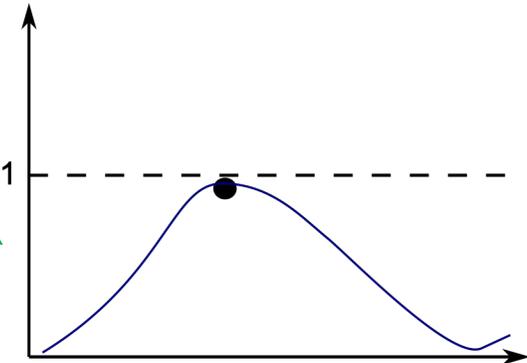
A. Zanco, S. Grivet-Talocia, T. Bradde, M. De Stefano, "Enforcing passivity of parameterized LTI macromodels via Hamiltonian-driven multivariate adaptive sampling", in IEEE TCAD, 2018

Passivity check

non-passive point



passive point



Finding all Passivity Violations of parameterized models

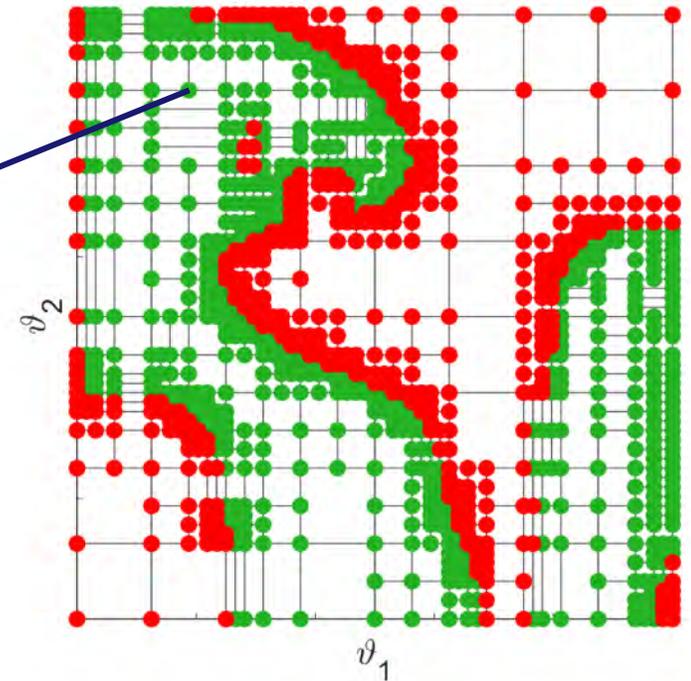
Why a passivity metric?

- In-band passivity violations \longrightarrow Bad model accuracy
- ***Predictive capabilities*** of Hamiltonian-based approach

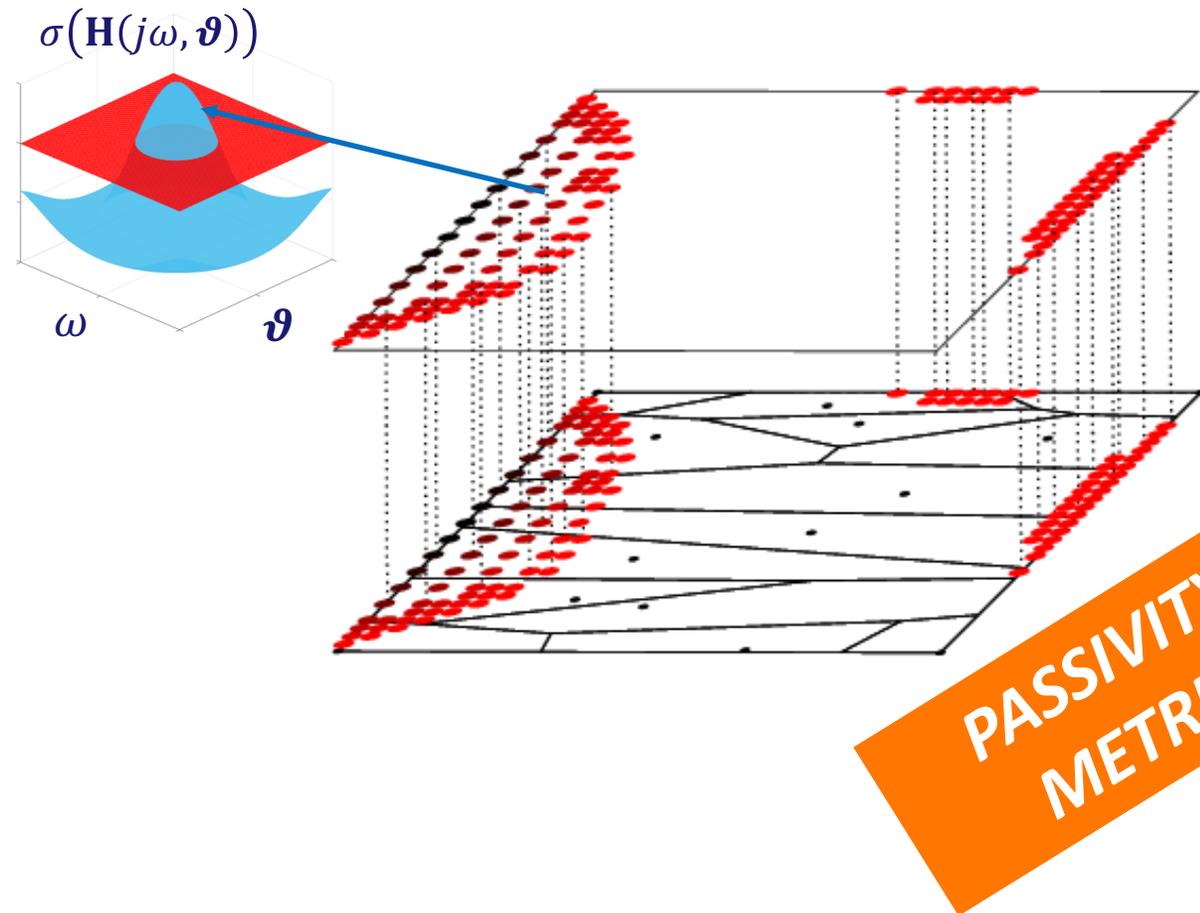


PROXY FOR MODEL-
DATA ERROR

*model responses,
NOT simulation points*



Passivity Violations Metric



- Superimpose in-band passivity violations to Voronoi cells
- Associate to each cell the largest violation

$$S(\boldsymbol{\vartheta}_q) = \max_{\omega \in \Omega} \sigma_{\max}(\mathbf{H}(s; \boldsymbol{\vartheta}))$$

$$\Lambda_3(\boldsymbol{\vartheta}_q) = \frac{S(\boldsymbol{\vartheta}_q)}{\sum_{q=1}^Q S(\boldsymbol{\vartheta}_q)}$$

E. Fevola, A. Zanco, S. Grivet-Talocia, T. Bradde, M. De Stefano, "An Adaptive Sampling Process for Automated Multivariate Modeling Based on Hamiltonian-Based Passivity Metric," to appear in IEEE Trans. CPMT, 2019

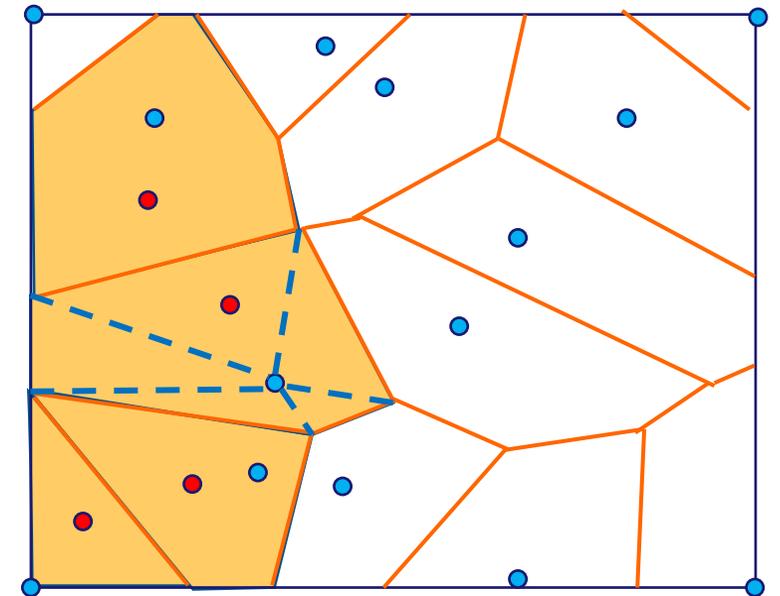
Adaptive Selection of New Points

Final metric

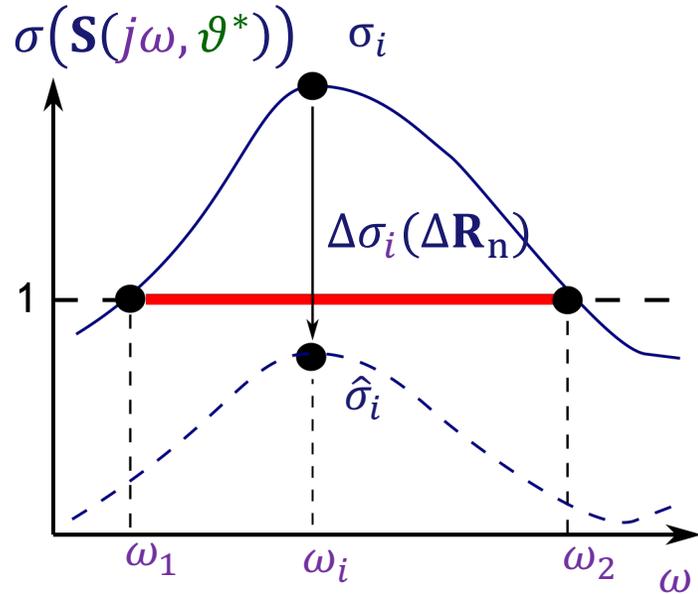
$$\Lambda(\boldsymbol{\vartheta}_q) = w' \Lambda_1(\boldsymbol{\vartheta}_q) + w'' \Lambda_2(\boldsymbol{\vartheta}_q) + w''' \Lambda_3(\boldsymbol{\vartheta}_q)$$

weights to tune metrics

- Rank the **Voronoi cells** on the basis of the final metric
- Take 1/3 of cells with **higher ranking**
- Place the new points inside these cells



Passivity Enforcement



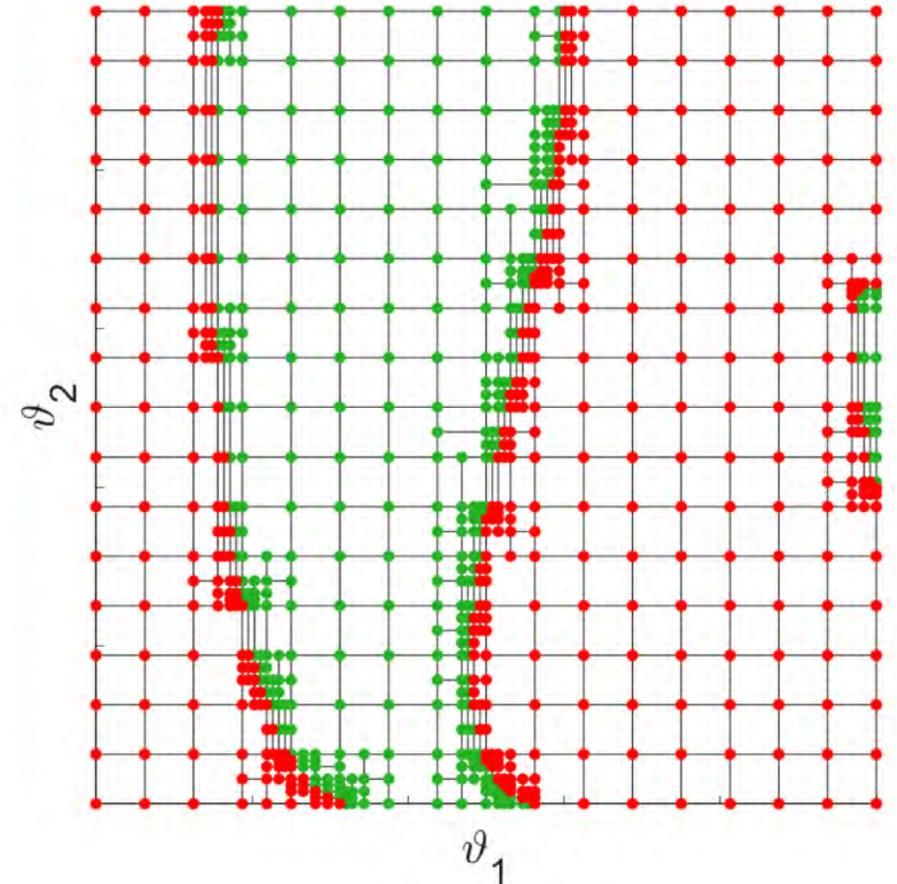
Why do we need an enforcement?



out - of band violations

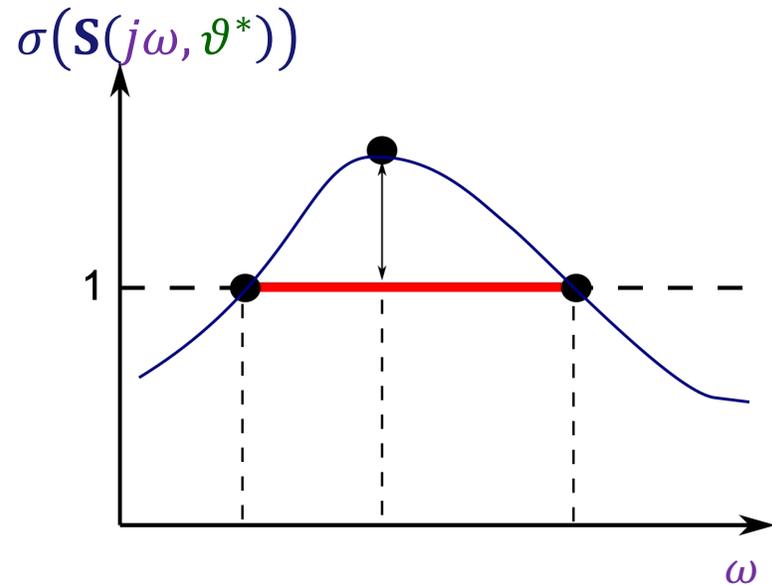
Perturbation of model coefficients (ΔR_n)

Iterative process



S. Grivet-Talocia, "A Perturbation Scheme for Passivity Verification and Enforcement of Parameterized Macromodels," *IEEE Transactions on Components, Packaging and Manufacturing Technology*, 2017

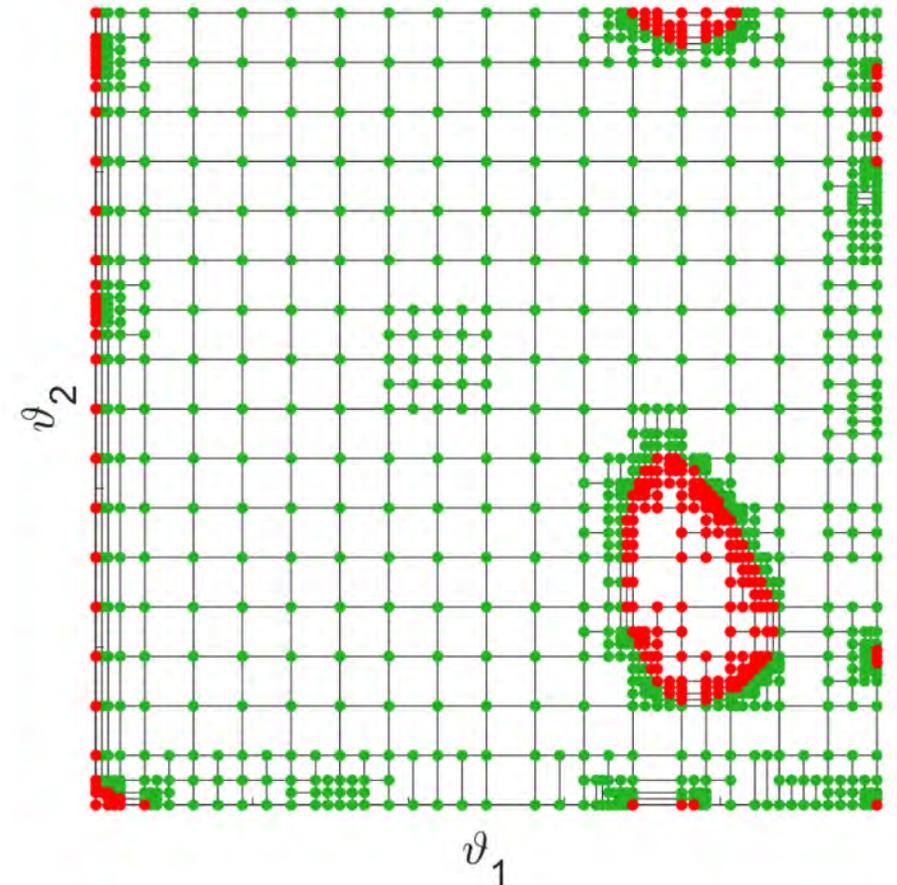
Passivity Enforcement



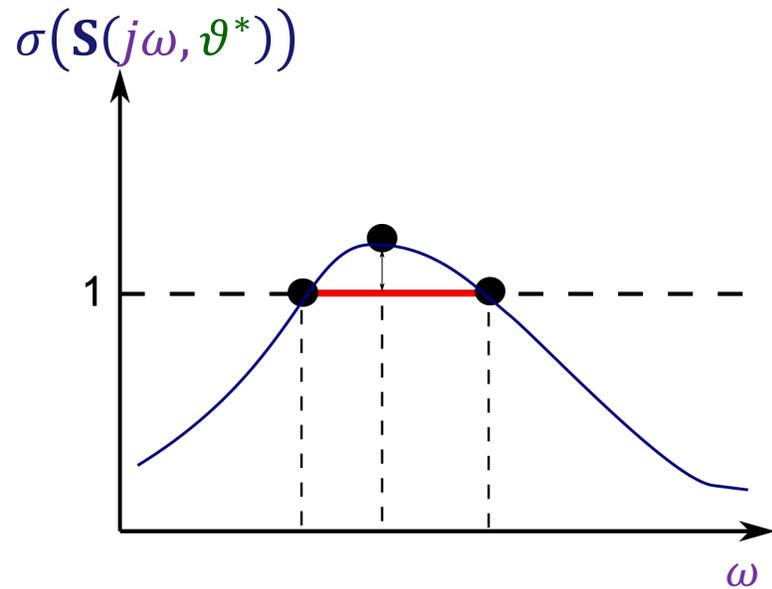
Perturbation of model coefficients (ΔR_n)

S. Grivet-Talocia, "A Perturbation Scheme for Passivity Verification and Enforcement of Parameterized Macromodels," to appear in *IEEE Transactions on Components, Packaging and Manufacturing Technology*, 2017

Iterative process



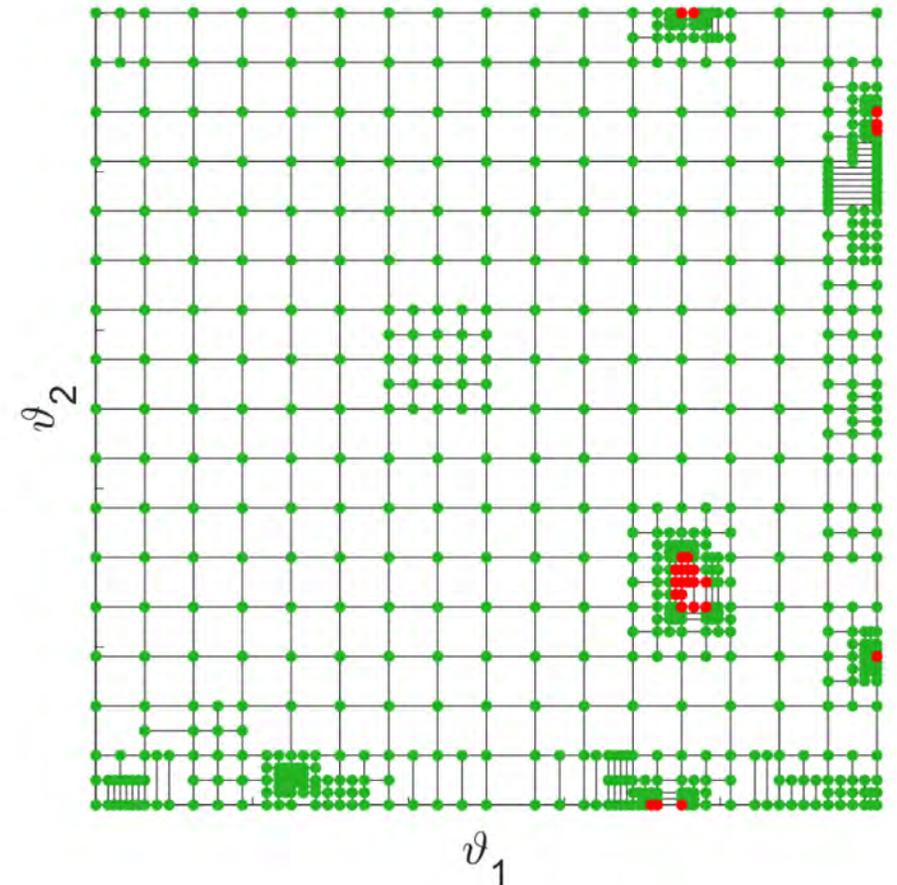
Passivity Enforcement



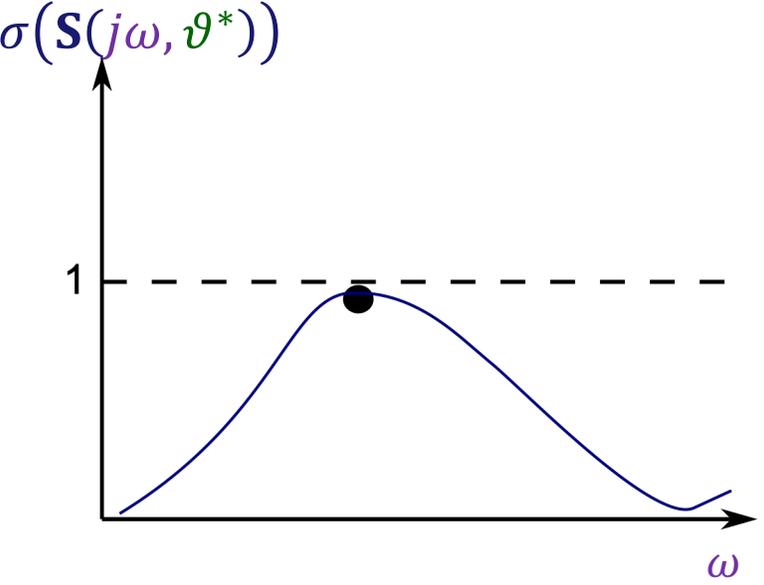
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Iterative process



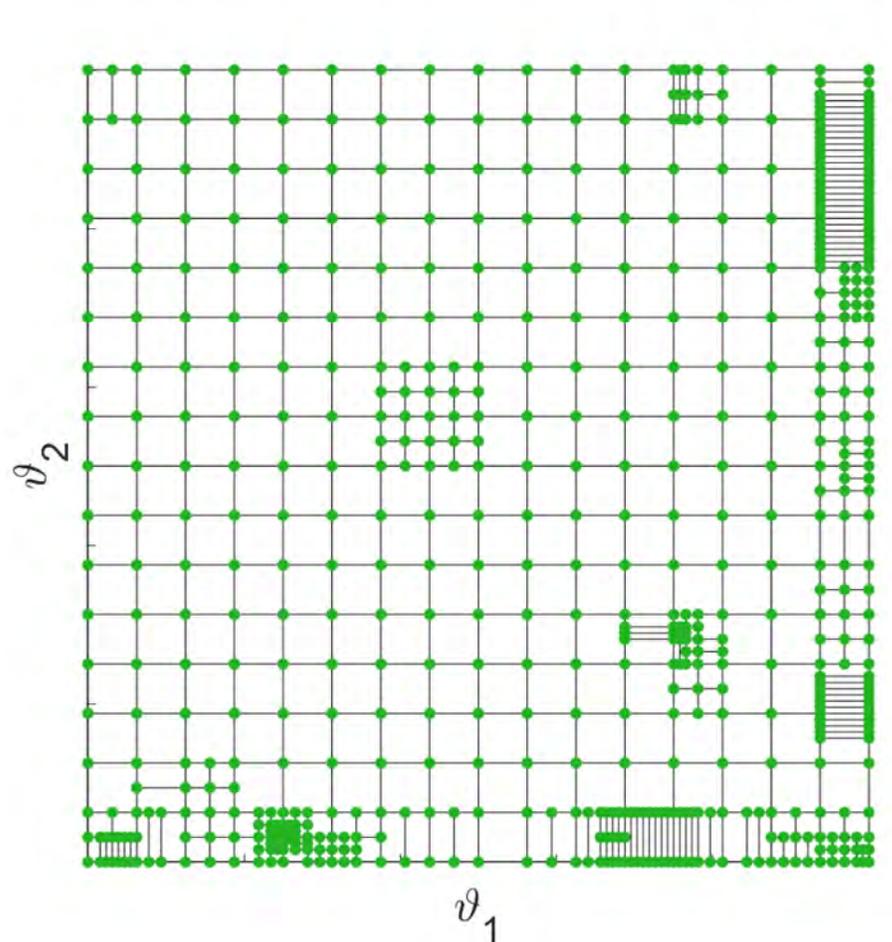
Passivity Enforcement



Perturbation of model coefficients (ΔR_n)

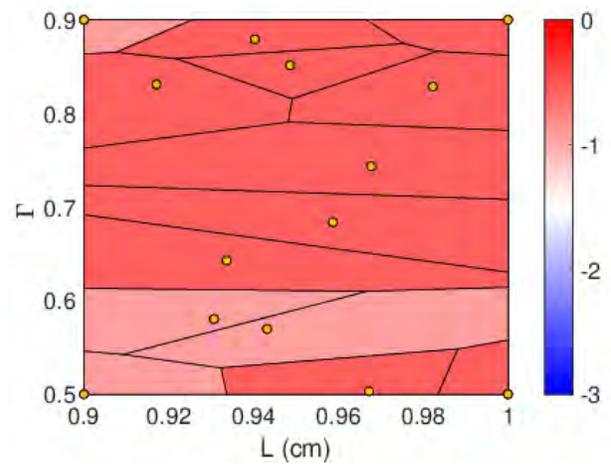
S. Grivet-Talocia, "A Perturbation Scheme for Passivity Verification and Enforcement of Parameterized Macromodels," to appear in *IEEE Transactions on Components, Packaging and Manufacturing Technology*, 2017

Iterative process

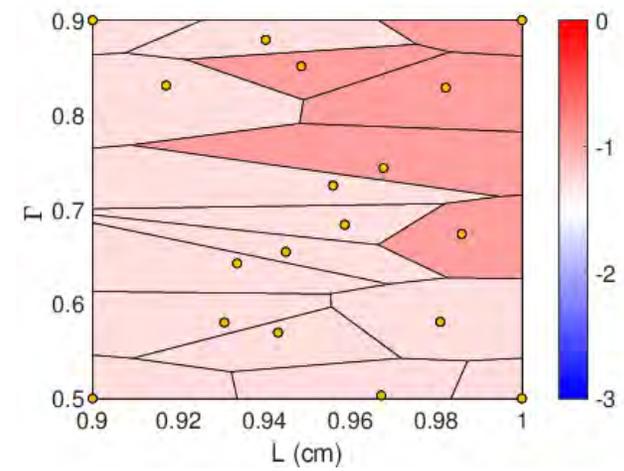


Examples – Transmission Line Network

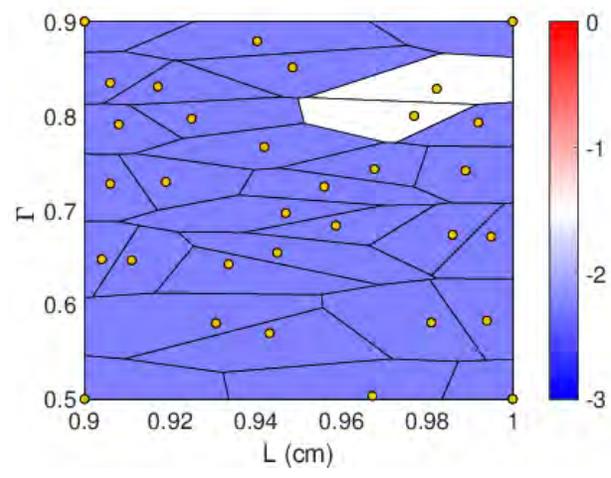
Iteration 0



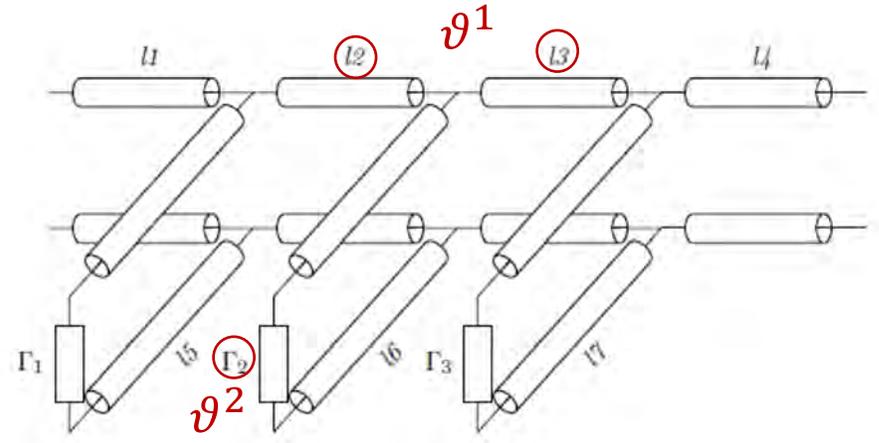
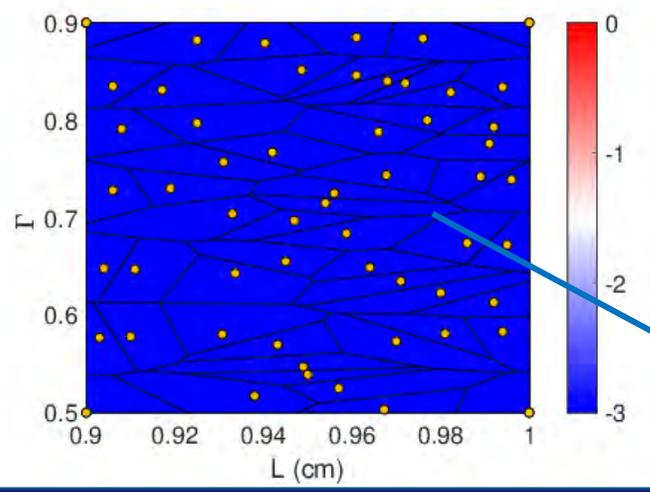
Iteration 2



Iteration 3



Iteration 5



Parameters:

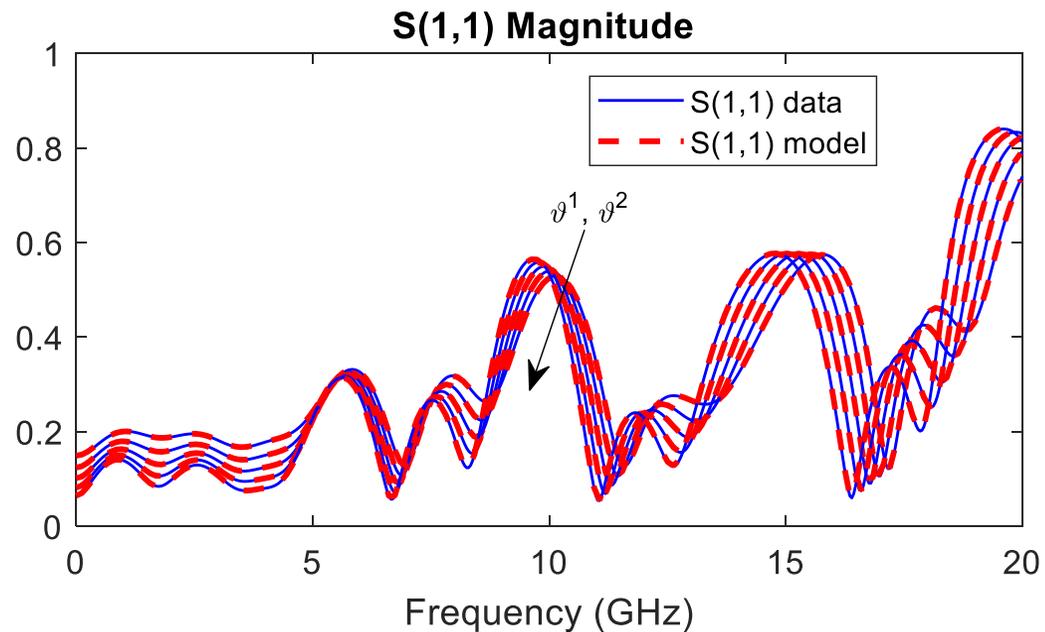
$$\vartheta^1 = L \in [9 - 10] \text{ mm}$$

$$\vartheta^2 = \Gamma \in [0.5 - 0.9]$$

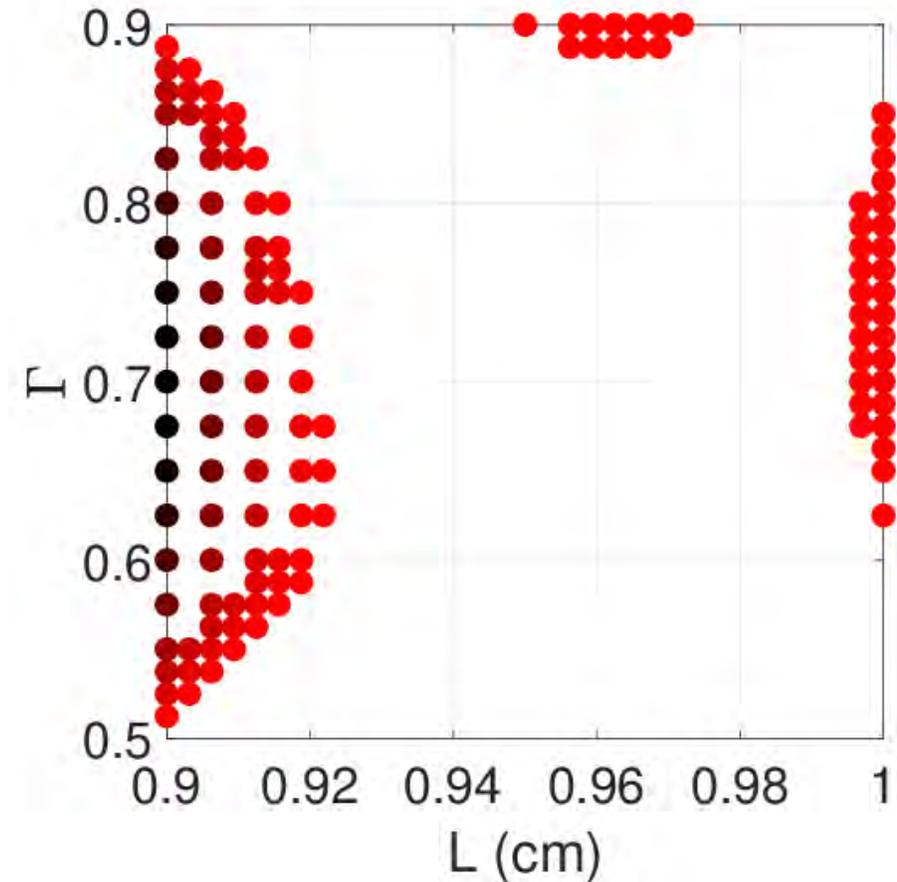
Q = 56 points

Examples – Transmission Line Network

Model vs data responses

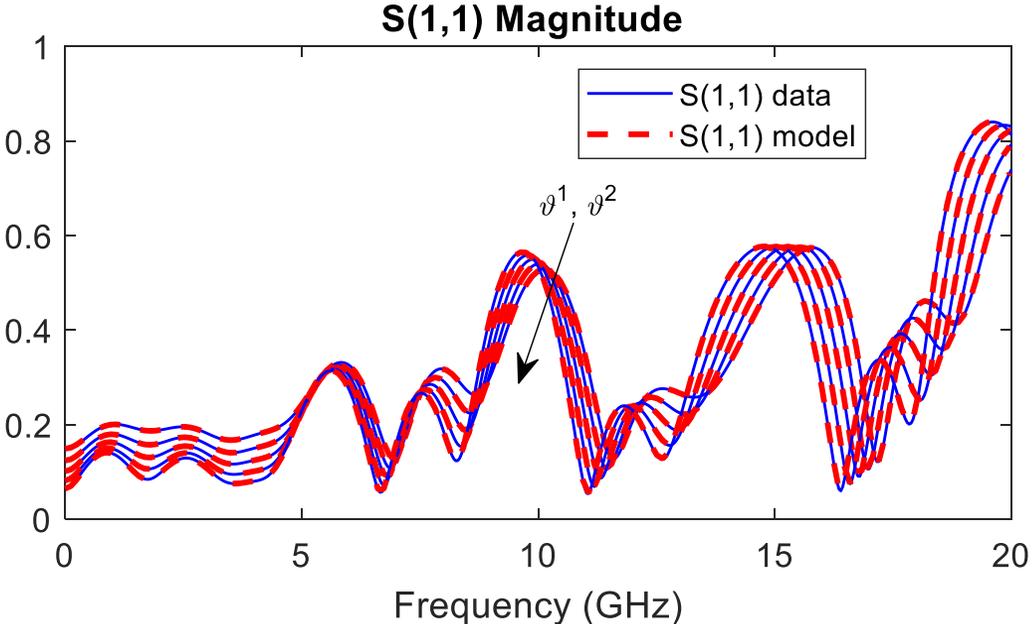


Passivity metric – Iteration 1

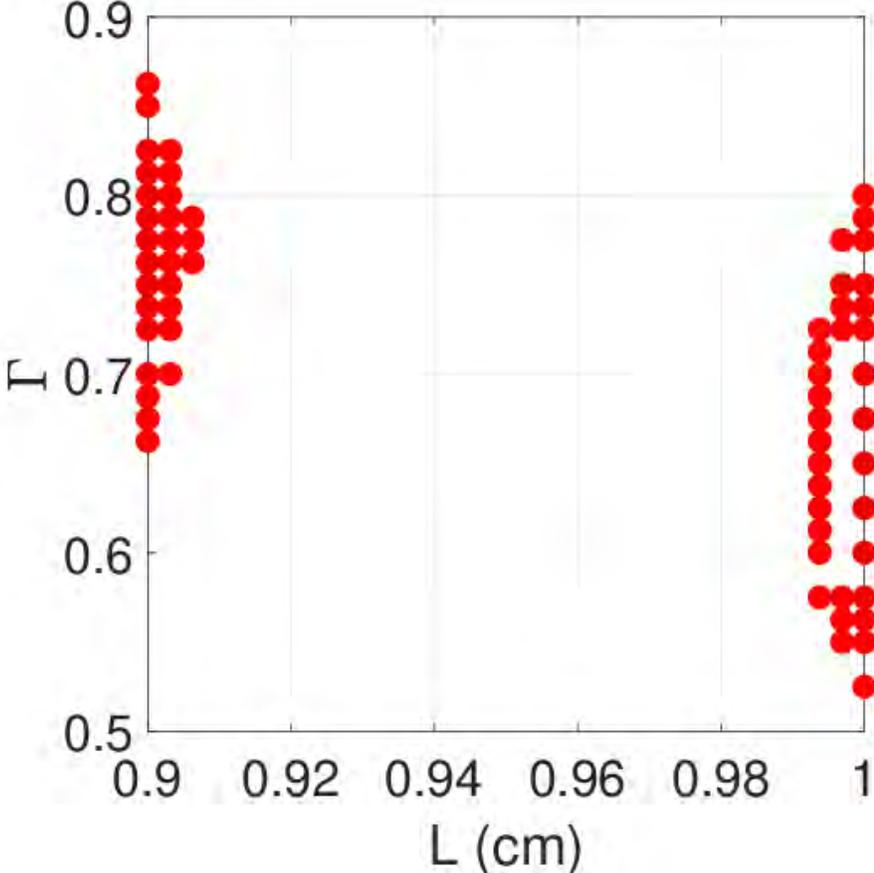


Examples – Transmission Line Network

Model vs data responses

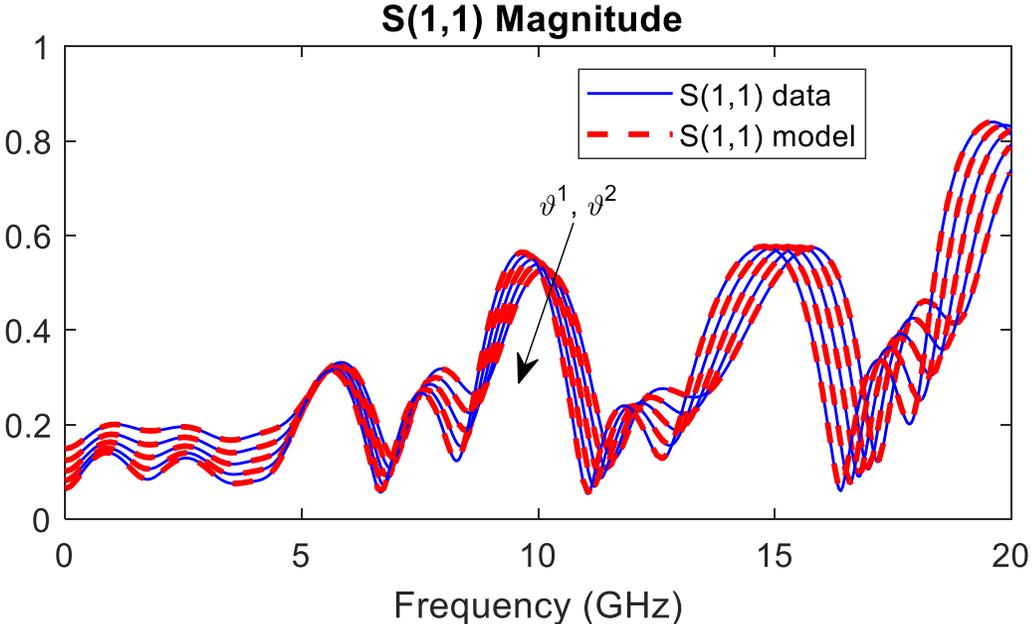


Passivity metric – Iteration 2

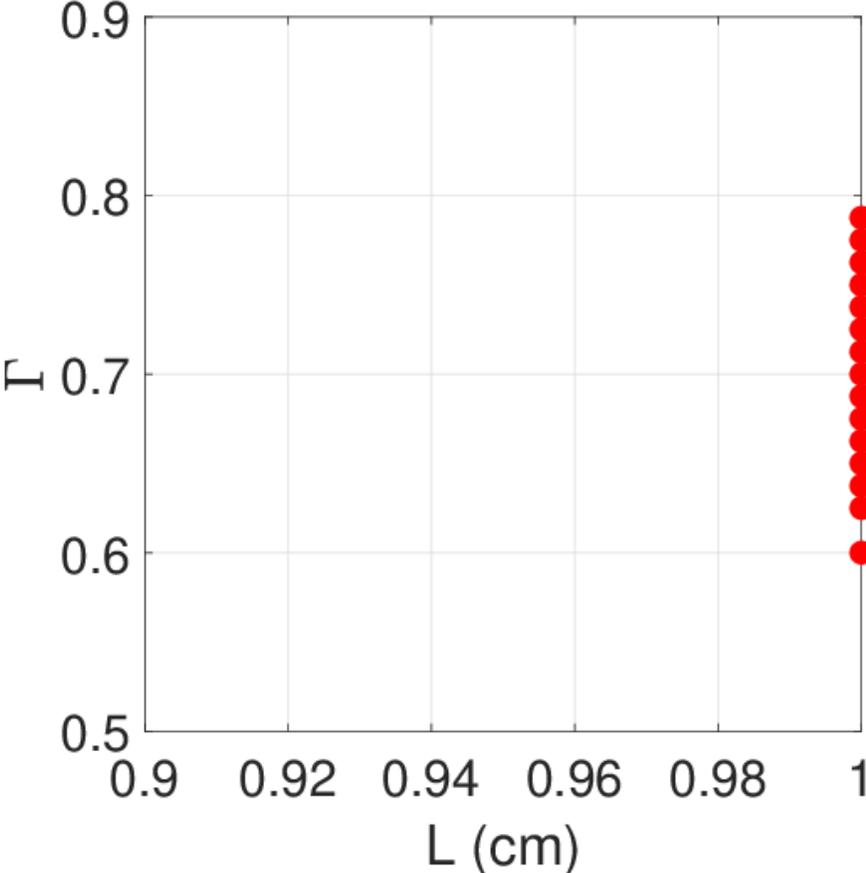


Examples – Transmission Line Network

Model vs data responses

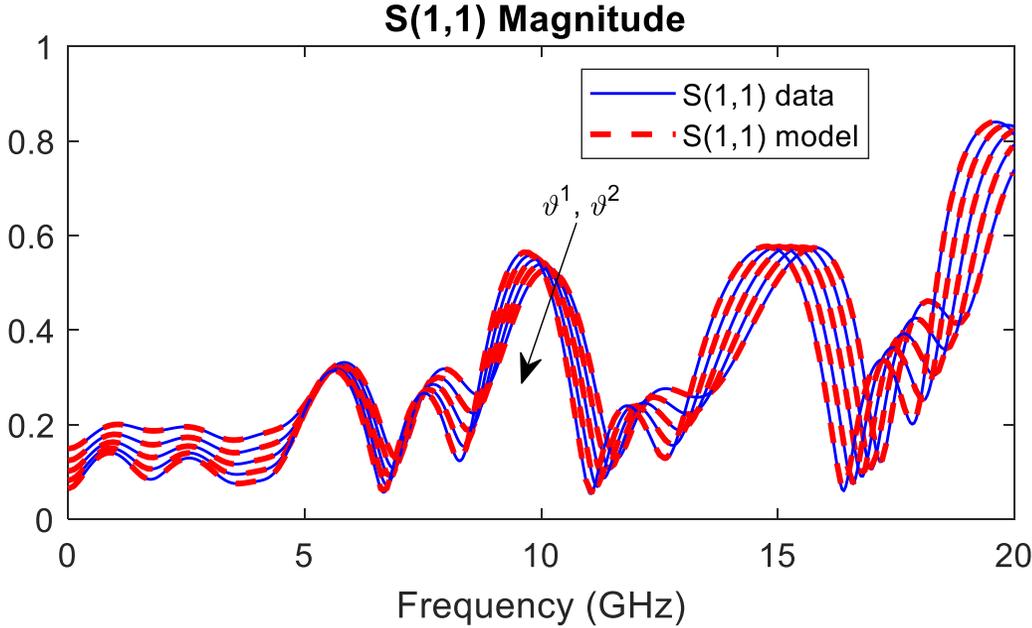


Passivity metric – Iteration 3

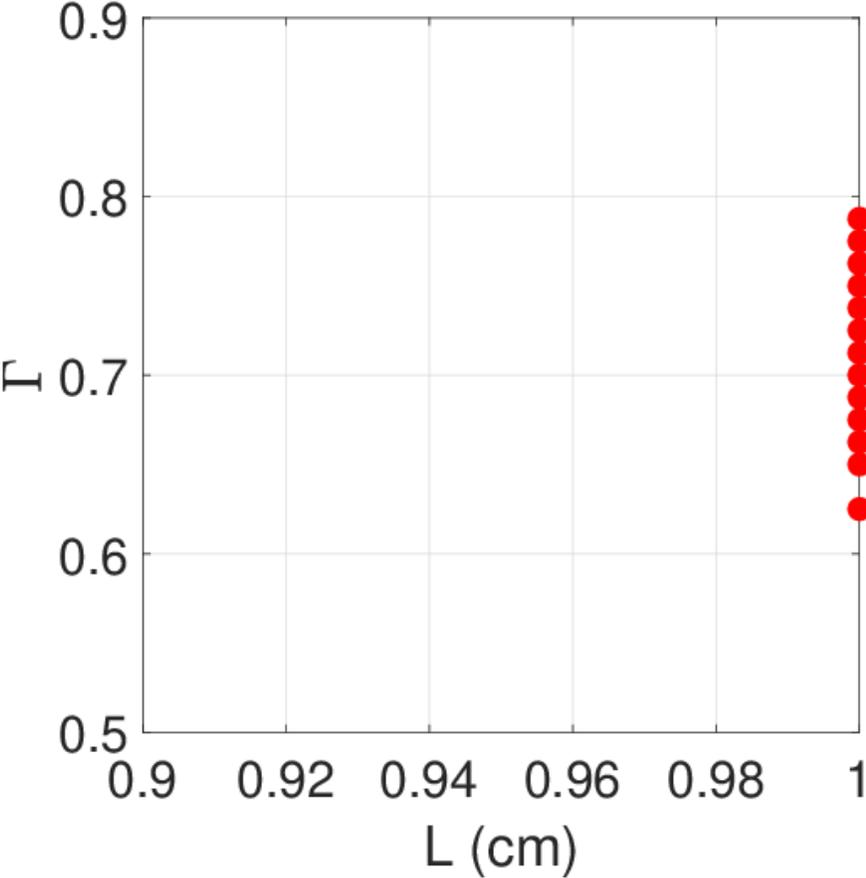


Examples – Transmission Line Network

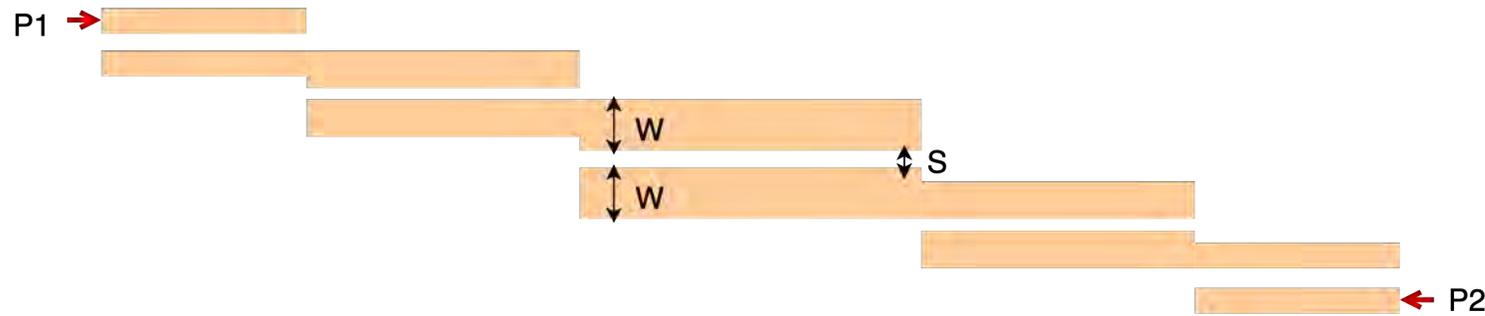
Model vs data responses



Passivity metric – Iteration 4



Examples – Coupled Line Bandpass filter



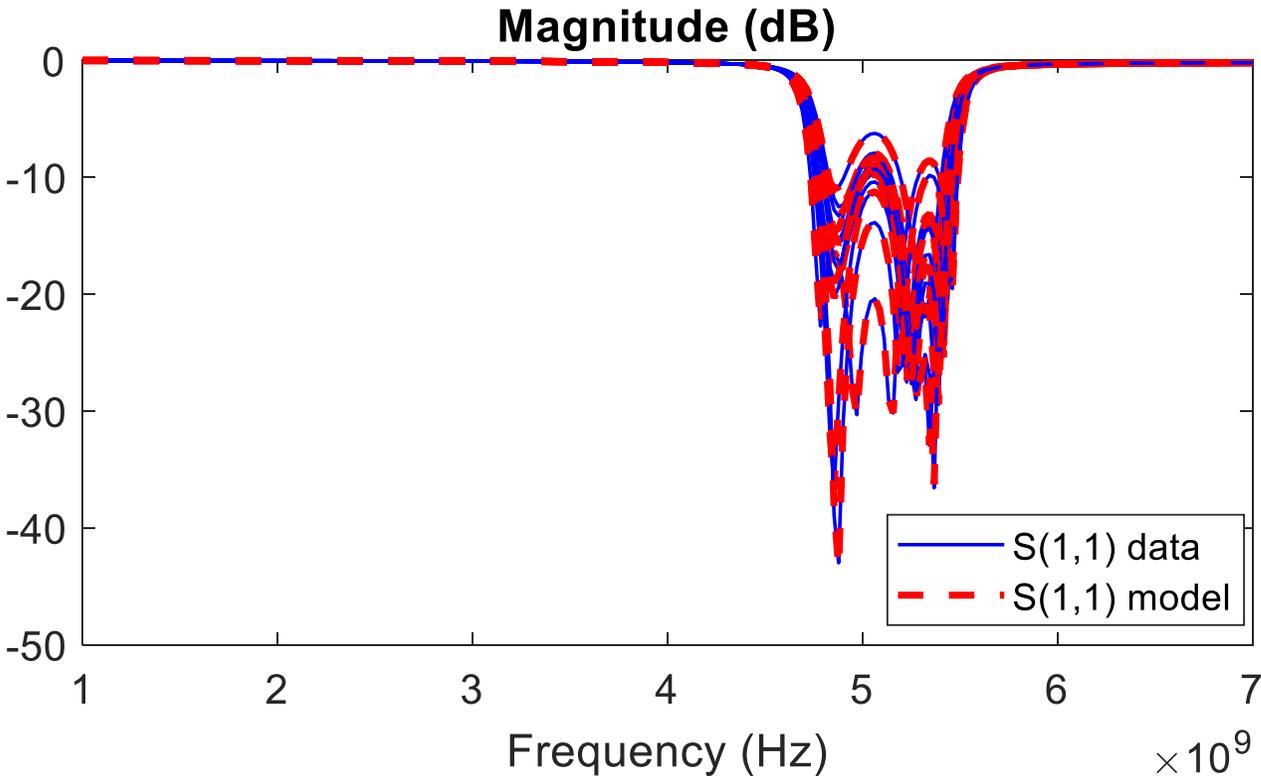
Parameters:

$$\vartheta^1 = W \in [25 - 35] \text{ mil}$$

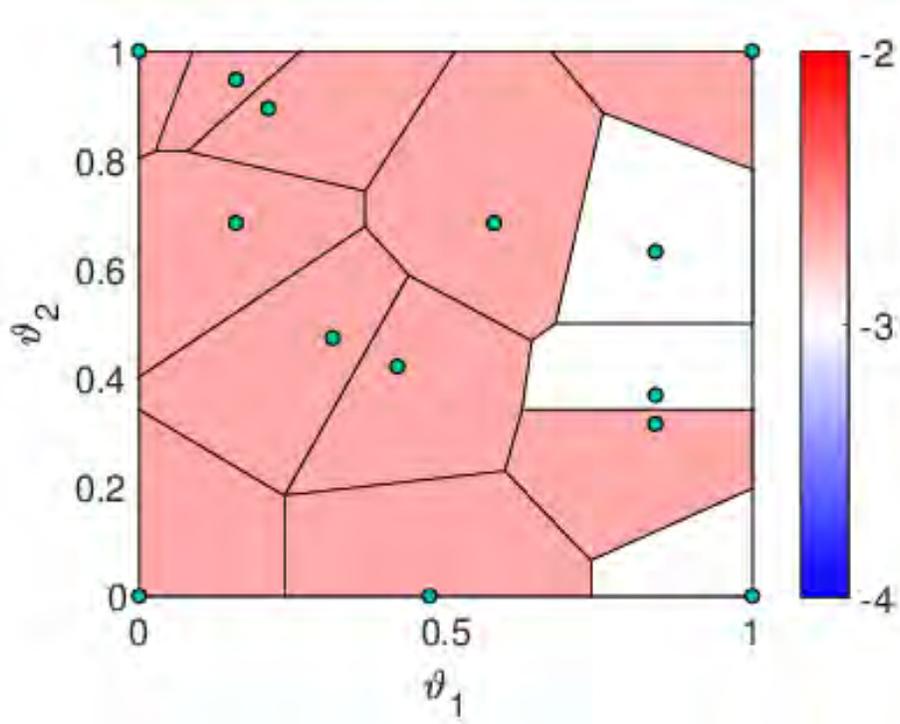
$$\vartheta^2 = S \in [25 - 35] \text{ mil}$$

Examples – Coupled Line Bandpass Filter

Model vs data responses

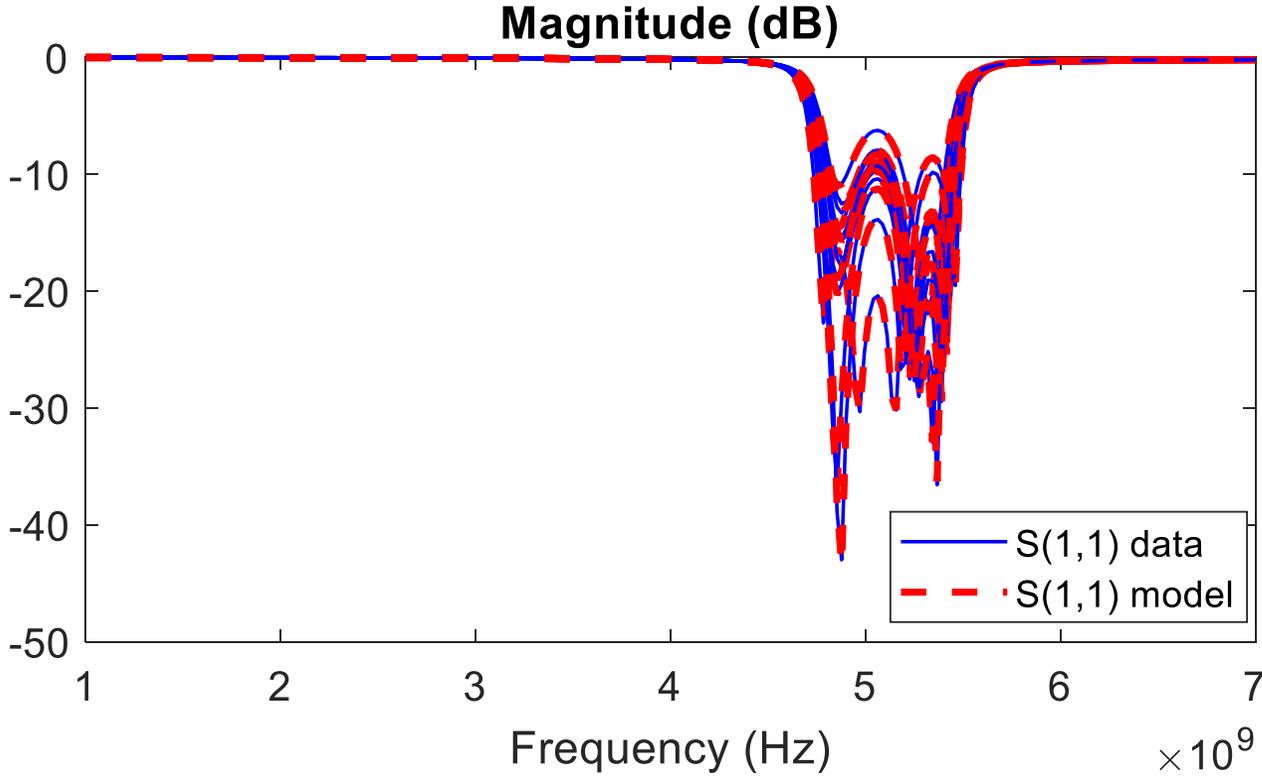


Iteration 0

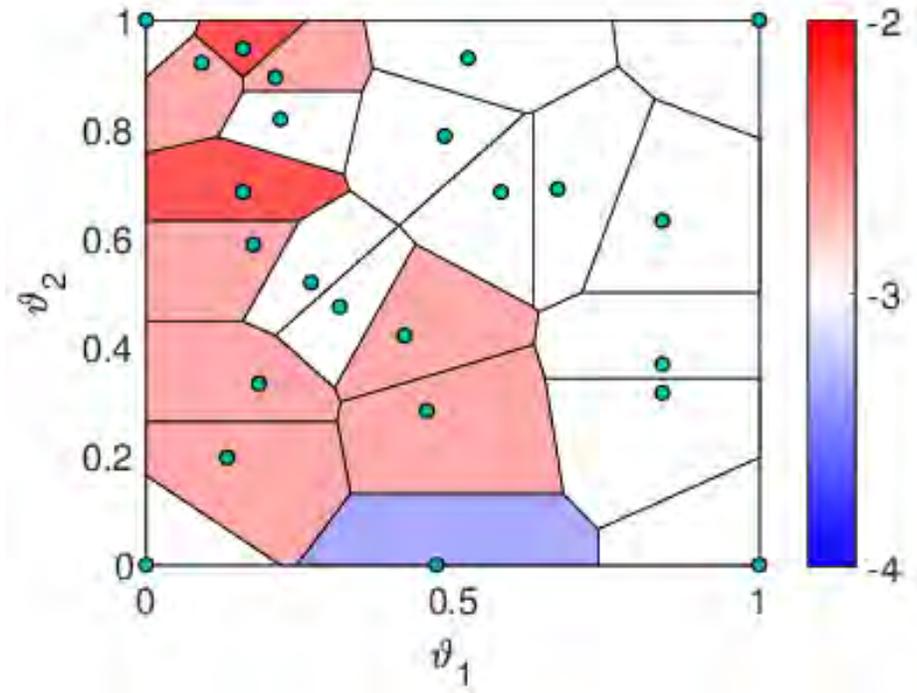


Examples – Coupled Line Bandpass Filter

Model vs data responses

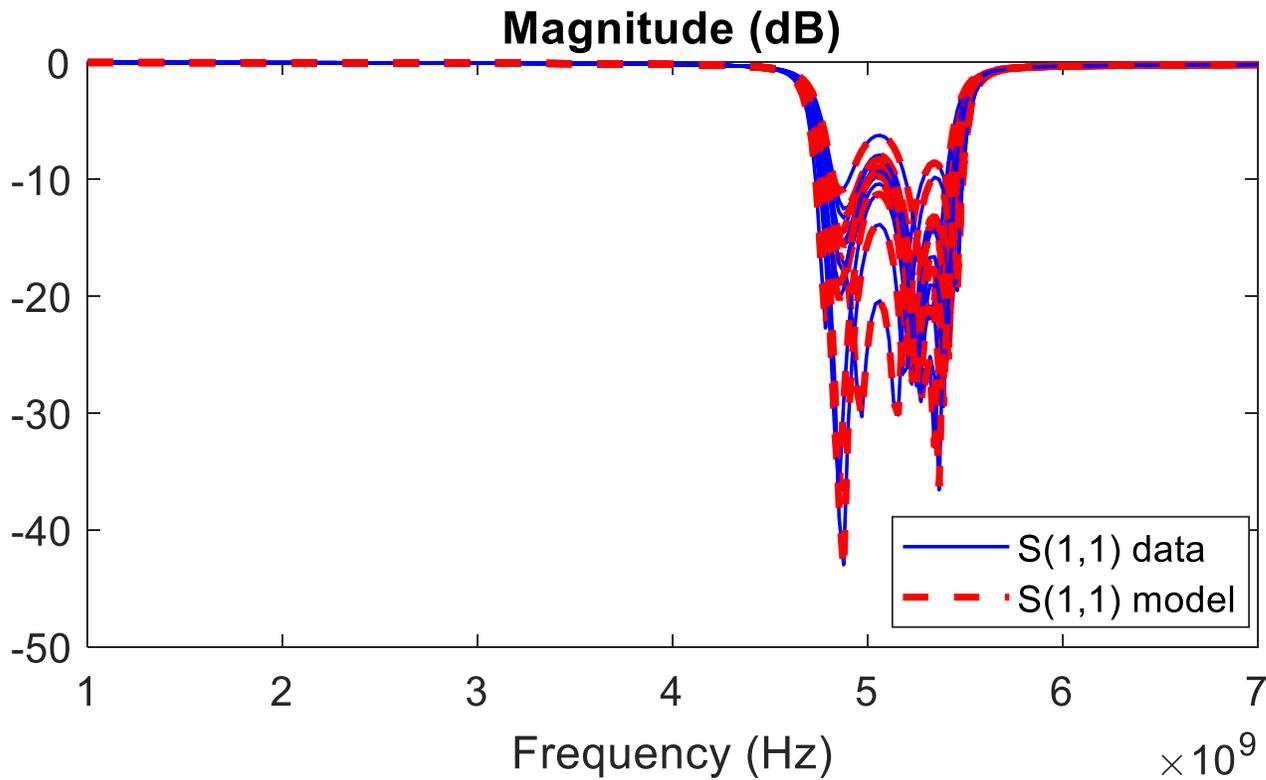


Iteration 2

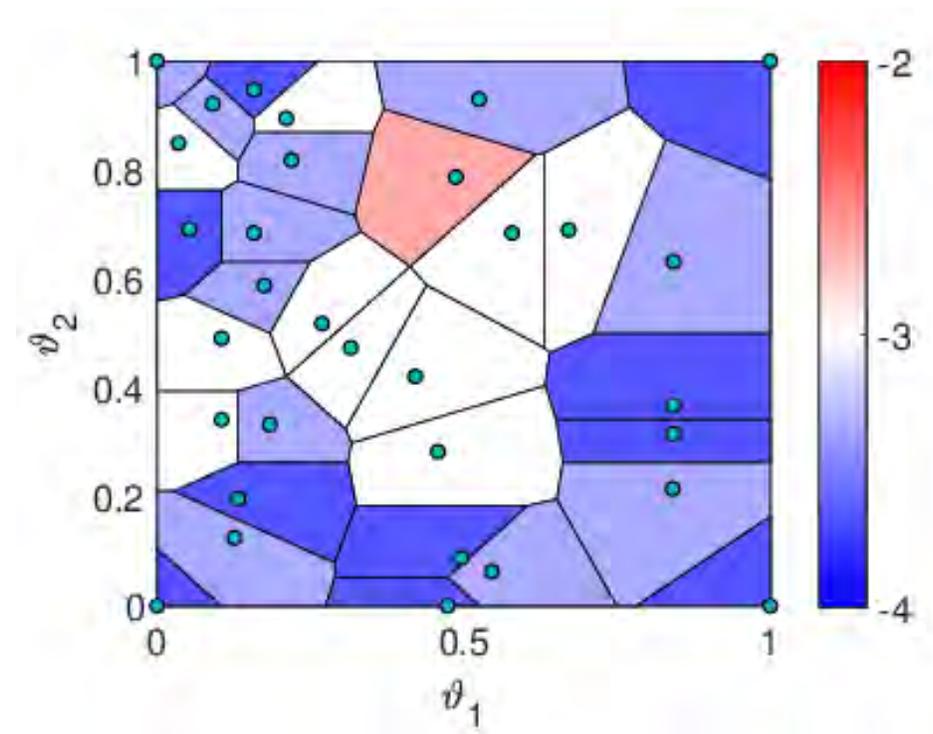


Examples – Coupled Line Bandpass Filter

Model vs data responses



Iteration 3



Examples – Coupled Line Bandpass Filter

Runtime

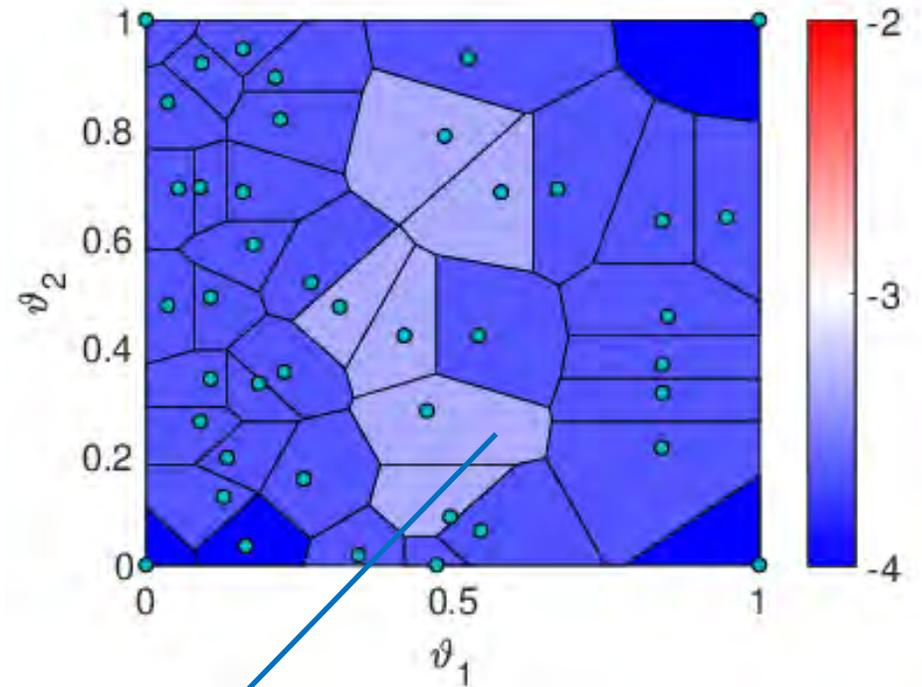
42 field solver responses → **1.2 hours**

Proposed algorithm → **417 s**



Less than 9% of total time

Iteration 4



Q = 42 points

Conclusions

- Fully automated algorithm for the generation of stable and passive parameterized macromodels.
- Identification of a quasi minimal set of simulation points.
- Enforcement of passivity and stability on the final model using perturbation approach.