

# Parametric model order reduction of electrical networks with semiconductors

Schloß Ringberg: Workshop on MOR of nonlinear systems

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## Outline

PDAE-model

Finite Element Method

Simulation results

Construction of the reduced model

Residual based parameter sampling

Combined reduction using PABTEC and POD, joint work with A. Steinbrecher  
& Tatjana Stykel

Next steps

# Coupled circuit and semiconductor models [M. Günther '01, C. Tischendorf '03]

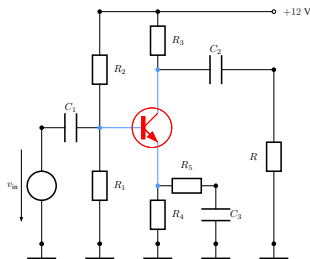
Kirchhoff's' laws (no semiconductors) read

$$A_j = 0, \quad v = A^T e$$

$A$ : (reduced) incidence matrix.

Voltage-current relations of components:

$$j_C = \frac{dq_C}{dt}(v_C, t), \quad j_R = g(v_R, t), \quad v_L = \frac{d\phi_L}{dt}(j_L, t)$$



Modified Nodal Analysis: join all equations to DAE system

$$A_C \frac{dq_C}{dt} (A_C^T e(t), t) + A_R g (A_R^T e(t), t) + A_L j_L(t) + A_V j_V(t) = -A_I i_s(t),$$

$$\frac{d\phi_L}{dt} (j_L(t), t) - A_L^T e(t) = 0,$$

$$A_V^T e(t) = v_s(t).$$

## Coupled circuit and semiconductor models [M. Günther '01, C. Tischendorf '03]

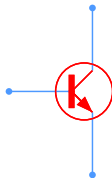
How can semiconductors be introduced?

- ▶ replace semiconductor by a (possibly nonlinear) electrical network,
- ▶ stamp semiconductor network into surrounding network,
- ▶ apply Modified Nodal Analysis.
- ▶ Here: use PDE model for semiconductors  $\rightarrow$  DD equations.

# Coupled circuit and semiconductor models [M. Günther '01, C. Tischendorf '03]

PDE-model (drift-diffusion equations) for semiconductors

$$\begin{aligned}
 \operatorname{div}(\varepsilon \nabla \psi) &= q(n - p - C), \\
 -q \partial_t n + \operatorname{div} J_n &= qR(n, p), \\
 q \partial_t p + \operatorname{div} J_p &= -qR(n, p), \\
 J_n &= \mu_n q (-U_T \nabla n - n \nabla \psi), \\
 J_p &= \mu_p q (-U_T \nabla p - p \nabla \psi),
 \end{aligned}$$



on  $\Omega \times [0, T]$  with  $\Omega \subset \mathbb{R}^d$  ( $d = 1, 2, 3$ ).

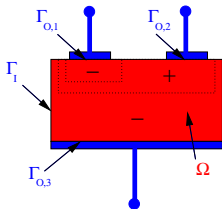
Dirichlet boundary constraints at  $\Gamma_{O,k}$ :

$$\psi(t, x) = \text{next slide}, \quad n(t, x) = \tilde{n}(x), \quad p(t, x) = \tilde{p}(x)$$

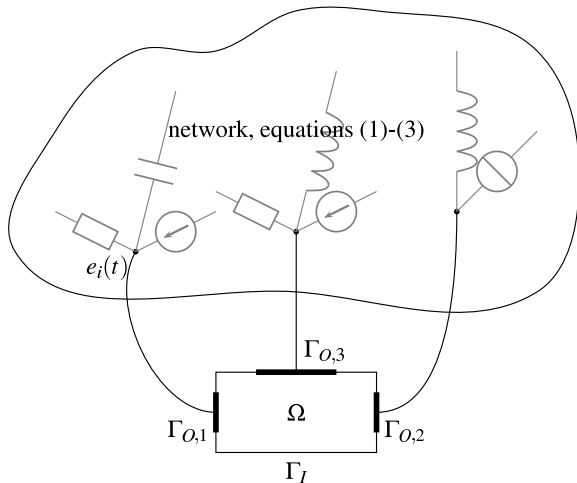
and Neumann boundary constraints at  $\Gamma_I$ :

$$\nabla \psi(t, x) \cdot \nu(x) = J_n \cdot \nu(x) = J_p(t, x) \cdot \nu(x) = 0$$

or mixed boundary conditions at MI contacts (MOSFETs).



# Coupled circuit and semiconductor models [M. Günther '01, C. Tischendorf '03]



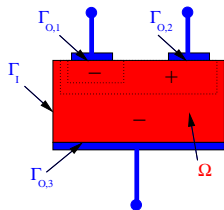
## Coupled circuit and semiconductor models [M. Günther '01, C. Tischendorf '03]

Coupling conditions:

$$j_{S,k}(t) = \int_{\Gamma_{O,k}} (J_n + J_p - \varepsilon \partial_t \nabla \psi) \cdot \nu \, d\sigma,$$

$$\psi(t, x) = \psi_{bi}(x) + (A_S^T e(t))_k$$

for  $(t, x) \in [0, T] \times \Gamma_{O,k}$ ,



and add current  $j_S$  to Kirchhoff's current law:

$$A_C \frac{dq_C}{dt} (A_C^T e, t) + A_R g (A_R^T e, t) + A_L j_L + A_V j_V + A_S j_S = -A_I i_S,$$

$$\frac{d\phi_L}{dt} (j_L, t) - A_L^T e = 0,$$

$$A_V^T e = v_S.$$

Add DD-equations + coupling conditions for each semiconductor.

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## Mixed formulation

The electric field  $E = -\nabla\psi$  plays dominant role in DD-equations.

### Mixed formulation

[Brezzi et al. '05]

Provide additional variable  $g_\psi$  and equation

$$g_\psi = \nabla\psi.$$

Scaled DD equations then read:

$$\begin{aligned} \lambda \operatorname{div} g_\psi &= n - p - C, \\ -\partial_t n + \nu_n \operatorname{div} J_n &= R(n, p), \\ \partial_t p + \nu_p \operatorname{div} J_p &= -R(n, p), \\ g_\psi &= \nabla\psi, \\ J_n &= \nabla n - n g_\psi, \\ J_p &= -\nabla p - p g_\psi. \end{aligned}$$

# Finite Element approximation

## Finite elements

- ▶ piecewise constant ansatz functions for  $\psi$ ,  $n$  and  $p$ .

Basis functions:  $\varphi_i$ ,  $i = 1, \dots, N$ ,  $N = |\mathcal{T}|$ .

- ▶ Raviart-Thomas elements for  $g_\psi$ ,  $J_n$  and  $J_p$ .

Basis functions:  $\phi_j$ ,  $i = 1, \dots, M$ ,  $M = |\mathcal{E}| - |\mathcal{E}_N|$ .

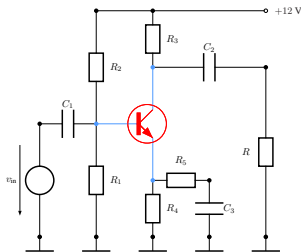
$$RT_0 := \{y : \Omega \rightarrow \mathbb{R}^d : y|_T(x) = a_T + b_T x, a_T \in \mathbb{R}^d, b_T \in \mathbb{R}, \\ [y]_E \cdot \nu_E = 0, \text{ for all inner edges } E\}.$$

Galerkin ansatz:

$$\psi^h(t, x) = \sum_{i=1}^N \psi_i(t) \varphi_i(x), \quad g_\psi^h(t, x) = \sum_{j=1}^M g_{\psi,j}(t) \phi_j(x),$$

and analogously for  $n$ ,  $p$ ,  $J_n$ , and  $J_p$ .

## Full model



$$\begin{aligned}
 A_C \frac{dq_C}{dt} (A_C^T e(t), t) + A_{RG} (A_R^T e(t), t) \\
 + A_{LJ} j_L(t) + A_V j_V(t) + A_S j_S(t) = -A_I i_S(t), \\
 \frac{d\phi_L}{dt} (j_L(t), t) - A_L^T e(t) = 0, \\
 A_V^T e(t) = v_S(t),
 \end{aligned}$$

$$j_S(t) - C_1 j_n(t) - C_2 j_p(t) - C_3 \dot{g}_\psi(t) = 0,$$

$$\begin{pmatrix} 0 \\ -M_L \dot{n}(t) \\ M_L \dot{p}(t) \\ 0 \\ 0 \\ 0 \end{pmatrix} + A_{FEM} \begin{pmatrix} \psi(t) \\ n(t) \\ p(t) \\ g_\psi(t) \\ j_n(t) \\ j_p(t) \end{pmatrix} + \mathcal{F}(n^h, p^h, g_\psi^h) - b(A_S^T e(t)) = 0.$$

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**Simulation results**

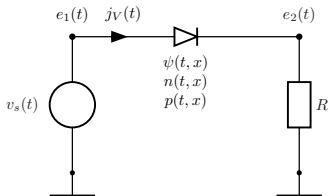
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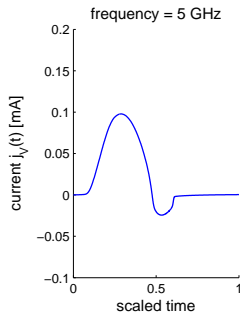
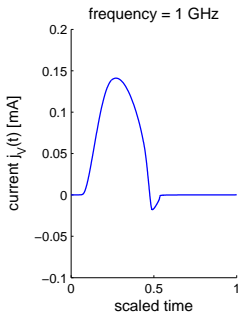
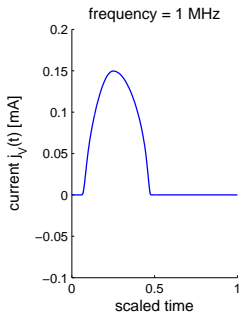
Next steps

## Basic test circuit, simulation results



input voltage:  $v_s(t) = 5[V] \cdot \sin(2\pi f \cdot t)$

similar results obtained by MECS [M. Selva Soto]



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## Snapshot-POD (Proper Orthogonal Decomposition) [L. Sirovich '87]

Full simulation yields snapshots (here:  $y = \psi, n, p, \dots$ )

$$\{y(t_i, \cdot)\}_{i=1, \dots, m} \subset \text{span}\{\varphi_j\}_{j=1, \dots, N}, \quad \text{with} \quad y(t_i, x) = \sum_{j=1}^N \tilde{y}_j(t_i) \varphi_j(x).$$

Gather coefficients in matrix

$$Y := (\tilde{y}(t_1), \dots, \tilde{y}(t_m)) \in \mathbb{R}^{N \times m}.$$

POD in Hilbert space  $X$  as eigenvalue problem:

$$Kv^k = \sigma_k^2 v^k, \quad \text{with} \quad K_{ij} := \langle y(t_i, \cdot), y(t_j, \cdot) \rangle_X.$$

Note that  $K = Y^T M Y$  with  $M_{ij} = \langle \varphi_i, \varphi_j \rangle_X$ . Write POD in terms of SVD:

$$\tilde{U} \Sigma \tilde{V}^T = L^T Y, \quad \text{with} \quad LL^T := M.$$

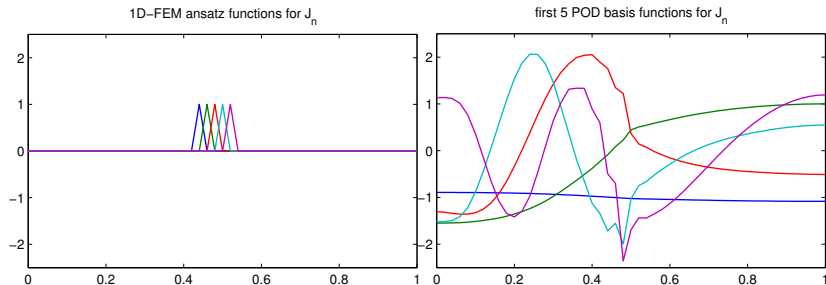
Then, the  $s$ -dimensional POD basis is

$$\left\{ u^i := \sum_{j=1}^N \tilde{u}_j^i \varphi_j(\cdot) \right\}_{i=1, \dots, s}, \quad U := (\tilde{u}^1, \dots, \tilde{u}^s) := L^{-T} \tilde{U}_{(:, 1:s)}.$$

## Model Order Reduction

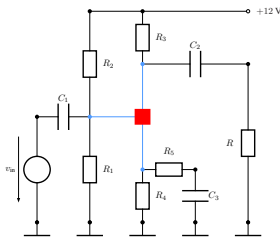
- ▶ Simulate the complete network at one or more reference parameters.
- ▶ Take snapshots of the state of each semiconductor at time points  $t_i$ .
- ▶ Perform POD **component wise** on  $\psi$ ,  $n$ ,  $p$ ,  $g_\psi$ ,  $J_n$  and  $J_p$ .
- ▶ Use the POD basis functions as (non local) FEM ansatz functions:

$$\psi^{POD}(t, x) = \sum_{i=1}^s \gamma_{\psi,i}(t) u_{\psi}^i(x)$$





## Reduced model



$$\begin{aligned}
 A_C \frac{dq_C}{dt} (A_C^T e(t), t) + A_{Rg} (A_R^T e(t), t) \\
 + A_L j_L(t) + A_V j_V(t) + A_S j_S(t) &= -A_I i_S(t), \\
 \frac{d\phi_L}{dt} (j_L(t), t) - A_L^T e(t) &= 0, \\
 A_V^T e(t) &= v_S(t),
 \end{aligned}$$

$$j_S(t) - C_1 U_{J_n} \gamma_{J_n}(t) - C_2 U_{J_p} \gamma_{J_p}(t) - C_3 U_{g_\psi} \dot{\gamma}_{g_\psi}(t) = 0,$$

$$\begin{pmatrix} 0 \\ -\dot{\gamma}_n(t) \\ \dot{\gamma}_p(t) \\ 0 \\ 0 \\ 0 \end{pmatrix} + A_{POD} \begin{pmatrix} \gamma_\psi(t) \\ \gamma_n(t) \\ \gamma_p(t) \\ \gamma_{g_\psi}(t) \\ \gamma_{J_n}(t) \\ \gamma_{J_p}(t) \end{pmatrix} + U^T \mathcal{F}(n^{POD}, p^{POD}, g_\psi^{POD}) - U^T b(A_S^T e(t)) = 0.$$

# Computational complexity

Computational complexity of reduced model still depends on  $n_{FEM}$ :

$$U^T \mathcal{F}(n^{POD}, p^{POD}, g_{\psi}^{POD}) = \underbrace{U^T}_{n_{POD} \times n_{FEM}} \underbrace{F}_{n_{FEM}} \left( \underbrace{U_n}_{n_{FEM} \times n_{POD}}, \gamma_n, U_p \gamma_p, U_{g_{\psi}} \gamma_{g_{\psi}} \right).$$

With matrix-matrix multiplications in Jacobian computation:

$$\underbrace{U^T}_{n_{POD} \times n_{FEM}, \text{ block-dense}} \underbrace{F'(\dots)}_{n_{FEM} \times n_{FEM}, \text{ sparse}} \underbrace{U}_{n_{FEM} \times n_{POD}, \text{ block-dense}}.$$

# Discrete Empirical Interpolation Md. (DEIM) [S. Chaturantabut, D. Sorensen '09]

## DEIM

- ▶ Do POD on snapshots  $\{F(n(t_i), p(t_i), g_{\psi}(t_i))\}$ , obtain basis  $W \in \mathbb{R}^{n_{FEM} \times n_{DEIM}}$  (block diagonal matrix).

- ▶ Ansatz

$$F(U_n \gamma_n(t), U_p \gamma_p(t), U_{g_{\psi}} \gamma_{g_{\psi}}(t)) \approx Wc(t)$$

is overdetermined.

- ▶ Select  $n_{DEIM}$  “useful” rows:

$$P^T F(\dots) \approx P^T Wc(t).$$

- ▶ If  $P^T W$  is regular:

$$F(\dots) \approx Wc(t) = W(P^T W)^{-1} P^T F(\dots)$$

The regularity of  $P^T W$  can be guaranteed, see [CS09].  
 Again we apply the method **component-wise**.

# Discrete Empirical Interpolation Md. (DEIM) [S. Chaturantabut, D. Sorensen '09]

Reduced model

$$U^T F(U_n \gamma_n, U_p \gamma_p, U_{g_\psi} \gamma_{g_\psi})$$

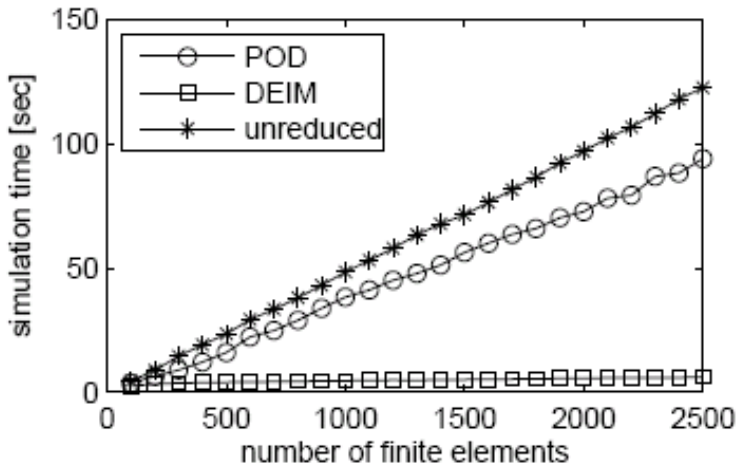
with DEIM:

$$\underbrace{(U^T W (P^T W)^{-1})}_{n_{POD} \times n_{DEIM}, \text{ block-dense}} \underbrace{P^T F(U_n \gamma_n, U_p \gamma_p, U_{g_\psi} \gamma_{g_\psi})}_{\substack{n_{DEIM} \\ n_{FEM}}}$$

Results for 1D-diode:

	$n_{FEM}$	FEM	$n_{POD}$	ROM	$n_{DEIM}$	ROM + DEIM
	3003	3.15 sec.	220	3.52 sec.	187	1.93 sec.
	15009	23.5 sec.	229	19.9 sec.	198	4.04 sec.
	48015	82.3 sec.	229	74.2 sec.	199	9.87 sec.
order		$\approx n_{FEM}^{1.18}$		$\approx n_{FEM}^{1.10}$		$\approx n_{FEM}^{0.578}$

# Discrete Empirical Interpolation Md. (DEIM) [S. Chaturantabut, D. Sorensen '09]



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**Residual based parameter sampling**

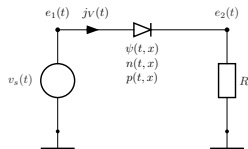
Combined reduction using PABTEC and POD, joint work with A. Steinbrecher & Tatjana Stykel

Next steps

## Problem setting

### MOR test problem

Basic circuit with **frequency  $f$**  of the voltage source  $v_s(t) = 5[V] \cdot \sin(2\pi f \cdot t)$   
 as **model parameter**.

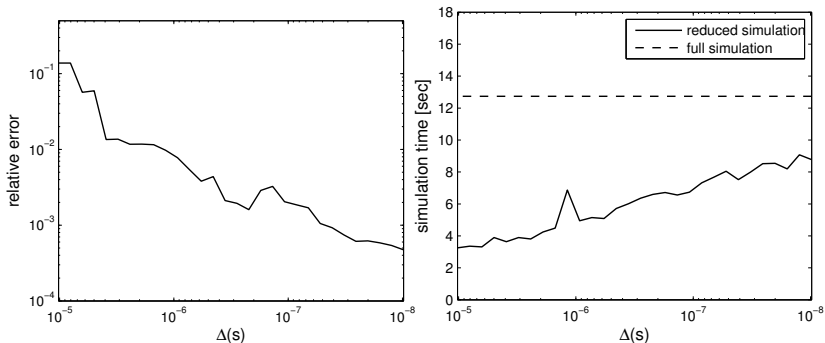


### Lack of information

Select number of snapshots so that  $\Delta(s) = \sqrt{\frac{\sum_{i=s+1}^m \sigma_i^2}{\sum_{i=1}^m \sigma_i^2}} \approx \text{tol}$ .

## Reduced model at a fixed frequency

First test: Compare reduced and unreduced system at a fixed frequency.



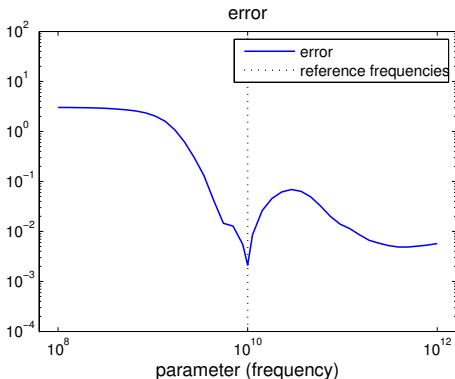


## Reduced model over parameter space

Construction of reduced model requires snapshots from full simulations at reference parameters.

Is the model valid over a large parameter space?

reference parameter:  $P_1 := \{f_1\} := \{10^{10}[\text{Hz}]\}$   
parameter space  $\mathcal{P} = [10^8, 10^{12}]$



## Reduced model over parameter space - sampling

### Goal

Find new sampling parameter  $f_{k+1}$  (reference frequency) without simulating the full, unreduced system. Set  $P_{k+1} := P_k \cup \{f_{k+1}\}$ .

- ▶ We do not consider the PDE discretization error.
- ▶ Rigorous upper bound for the error not available

$$\|\mathcal{E}(f; P_k)\| = \|y^h(f) - y^{POD}(f; P_k)\| \leq ?(s)$$

where  $y^h := (\psi^h, n^h, p^h, g_\psi^h, J_n^h, J_p^h)^\top$ ,  $y^{POD} := (\psi^{POD}, n^{POD}, \dots)^\top$ .

- ▶ Rigorous RB methods, Greedy algorithm [see e.g. A. Patera, G. Rozza '07]: a-posteriori error estimates required.
- ▶ Linear ODEs [see e.g. B. Haasdonk, M. Ohlberger '09]: build difference between residual and unreduced equation to derive an ODE for the error.

## Residual based sampling

Define residual  $\mathcal{R}(z^{POD}(f; P_k))$ : insert  $z^{POD}(f; P_k)$  into unreduced equation,

$$\mathcal{R} := \begin{pmatrix} 0 \\ -M_L \dot{n}^{POD}(t) \\ M_L \dot{p}^{POD}(t) \\ 0 \\ 0 \\ 0 \end{pmatrix} + A_{FEM} \begin{pmatrix} \psi^{POD}(t) \\ n^{POD}(t) \\ p^{POD}(t) \\ g_{\psi}^{POD}(t) \\ J_n^{POD}(t) \\ J_p^{POD}(t) \end{pmatrix} + \mathcal{F}(n^{POD}, p^{POD}, g_{\psi}^{POD}) - b(e^{POD}(t)).$$

**Residual admits different scales.**

Scale with block diagonal matrix-valued function

$$D(f) := \text{diag}(d_{\psi}(f)I, d_n(f)I, d_p(f)I, d_{g_{\psi}}(f)I, d_{J_n}(f)I, d_{J_p}(f)I)$$

and choose  $d_{\psi}(f)$  according to

$$d_{\psi}(f_j) \cdot \|\mathcal{R}_{\psi}(y^{POD}(f_j; P_k))\| = \frac{\|\psi^h(f_j) - \psi^{POD}(f_j; P_k)\|}{\|\psi^h(f_j)\|}, \quad \forall f_j \in P_k.$$

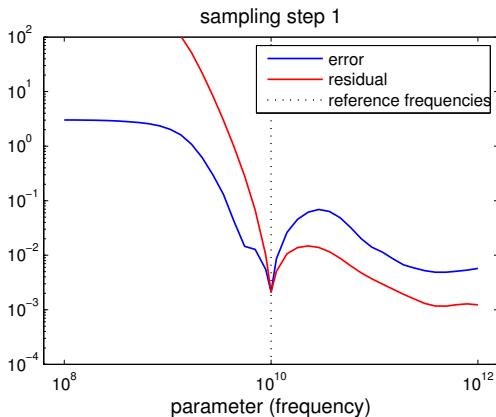
## Residual based sampling

### Algorithm: sampling

1. Select  $f_1 \in \mathcal{P}$ ,  $P_{test} \subset \mathcal{P}$ ,  $tol > 0$ , and set  $k := 1$ ,  $P_1 := \{f_1\}$ .
2. Simulate the unreduced model at  $f_1$  and calculate the reduced model with POD basis functions  $U_1$ .
3. Calculate weight functions  $d_{(\cdot)}(f) > 0$  for all  $f \in P_k$ .
4. Calculate the scaled residual  $\|D(f)\mathcal{R}(z^{POD}(f, P_k))\|$  for all  $f \in P_{test}$ .
5. Check termination conditions, e.g.
  - ▶  $\max_{f \in P_{test}} \|D(f)\mathcal{R}(z^{POD}(f, P_k))\| < tol$ ,
  - ▶ no progress in weighted residual.
6. Calculate  $f_{k+1} := \arg \max_{f \in P_{test}} \|D(f)\mathcal{R}(z^{POD}(f, P_k))\|$ .
7. Simulate the unreduced model at  $f_{k+1}$  and create a new reduced model with POD basis  $U_{k+1}$  using also the already available information at  $f_1, \dots, f_k$ .
8. Set  $P_{k+1} := P_k \cup \{f_{k+1}\}$ ,  $k := k + 1$  and goto 3.

## Numerical example - sampling step 1

Let  $f_1 := 10^{10}[\text{Hz}]$ ,  $P_1 := \{10^{10}[\text{Hz}]\}$ ,  $\mathcal{P} = [10^8, 10^{12}]$ .

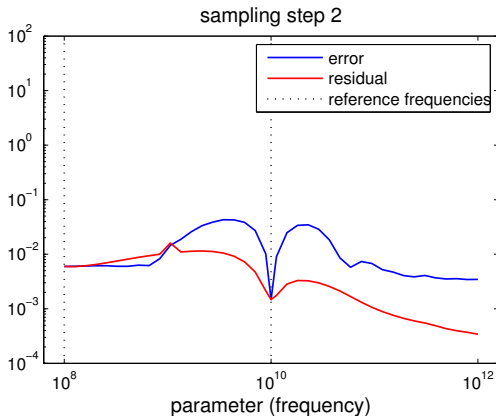


$$f_2 = \arg \max_{f \in P_{\text{test}}} \|D(f)\mathcal{R}(z^{\text{POD}}(f, P_1))\| = 10^8[\text{Hz}]$$

$$P_2 = \{10^8[\text{Hz}], 10^{10}[\text{Hz}]\}$$

## Numerical example - sampling step 2

$$P_2 = \{10^8[\text{Hz}], 10^{10}[\text{Hz}]\}$$

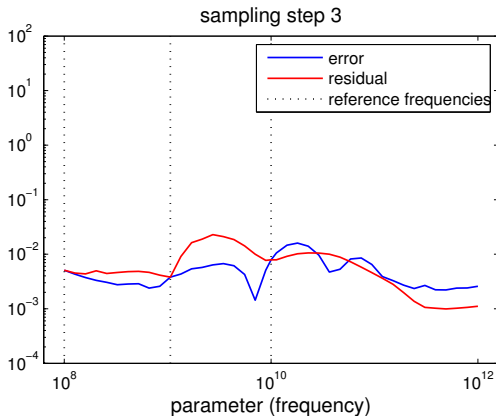


$$f_3 = \arg \max_{f \in P_{\text{test}}} \|D(f)\mathcal{R}(z^{\text{POD}}(f, P_2))\| = 1.0608 \cdot 10^9[\text{Hz}]$$

$$P_3 = \{10^8[\text{Hz}], 1.0608 \cdot 10^9[\text{Hz}], 10^{10}[\text{Hz}]\}$$

## Numerical example - sampling step 3

$$P_3 = \{10^8[\text{Hz}], 1.0608 \cdot 10^9[\text{Hz}], 10^{10}[\text{Hz}]\}$$



Terminate with “no progress in residual”.

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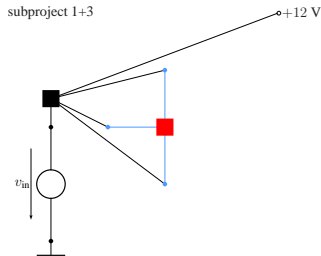
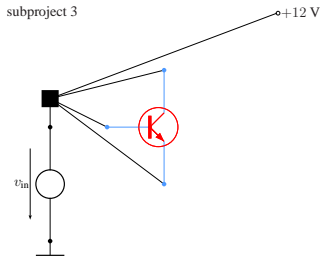
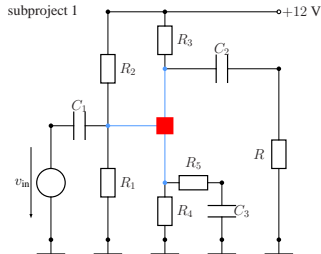
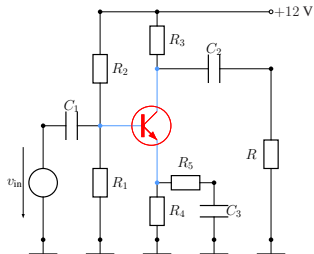
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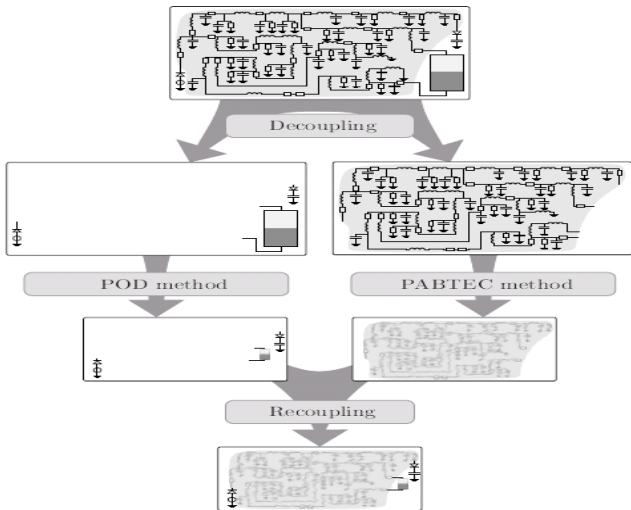
Next steps



# Combination of PABTEC and POD; joint work with [A. Steinbrecher, T. Stykel]



# Combination of PABTEC and POD; Int. J. Numer. Model. 2012



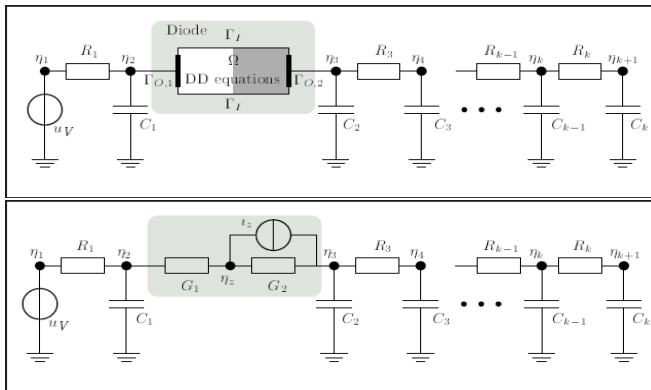
## Substitution of nonlinear components for PABTEC and recoupling

A. Steinbrecher, T. Stykel (Int. J. Circuits Theory Appl., 2012):

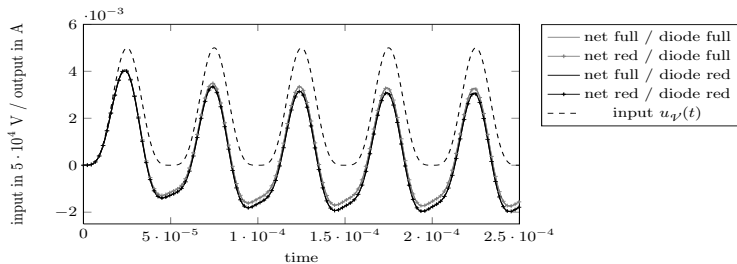
Nonlinear inductor  $\rightarrow$  current source

Nonlinear capacitor  $\rightarrow$  voltage source

Nonlinear resistor  $\rightarrow$  linear circuit with 2 serial resistors and one voltage source parallel to one of the resistors



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network (MNA equations)		diode (DD equations)		coupled system	simul. time	Jacobian evaluations	absolute error	relative error
type	dim.	type	dim.	dim.			$\ y - \hat{y}\ _{L_2}$	$\frac{\ y - \hat{y}\ _{L_2}}{\ y\ _{L_2}}$
unreduced	1503	unreduced	6006	7510	23.37s	20		
reduced	24	unreduced	6006	6031	16.90s	17	$2.165 \cdot 10^{-8}$	$7.335 \cdot 10^{-4}$
unreduced	1503	reduced	105	1609	1.51s	16	$2.952 \cdot 10^{-6}$	$1.000 \cdot 10^{-1}$
reduced	24	reduced	105	130	1.19s	11	$2.954 \cdot 10^{-6}$	$1.000 \cdot 10^{-1}$

## Next steps

- ▶ Include QDD models.
- ▶ Include EM effects.
- ▶ Generalize approach to other equation networks containing simple and complex components.

Thank you for attending!



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