

A General Method for Multi-Port Active Network Reduction and Realization

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Outline

- Compact modeling via model order reduction
- General hierarchical model order reduction
- Optimization considering magnitude and phase responses
- Experimental results
- Conclusions



Compact Modeling

- **More parasitics for nanometer VLSIs**
 - millions and hundreds of millions of RLC elements
 - Overload simulation engines
 - not all information is useful
- **Compact modeling via model order reduction**
 - Capture port behavior of the circuit block to speed up simulation and synthesis
 - Discard no useful information
 - Black box approach



Existing MOR Algorithms

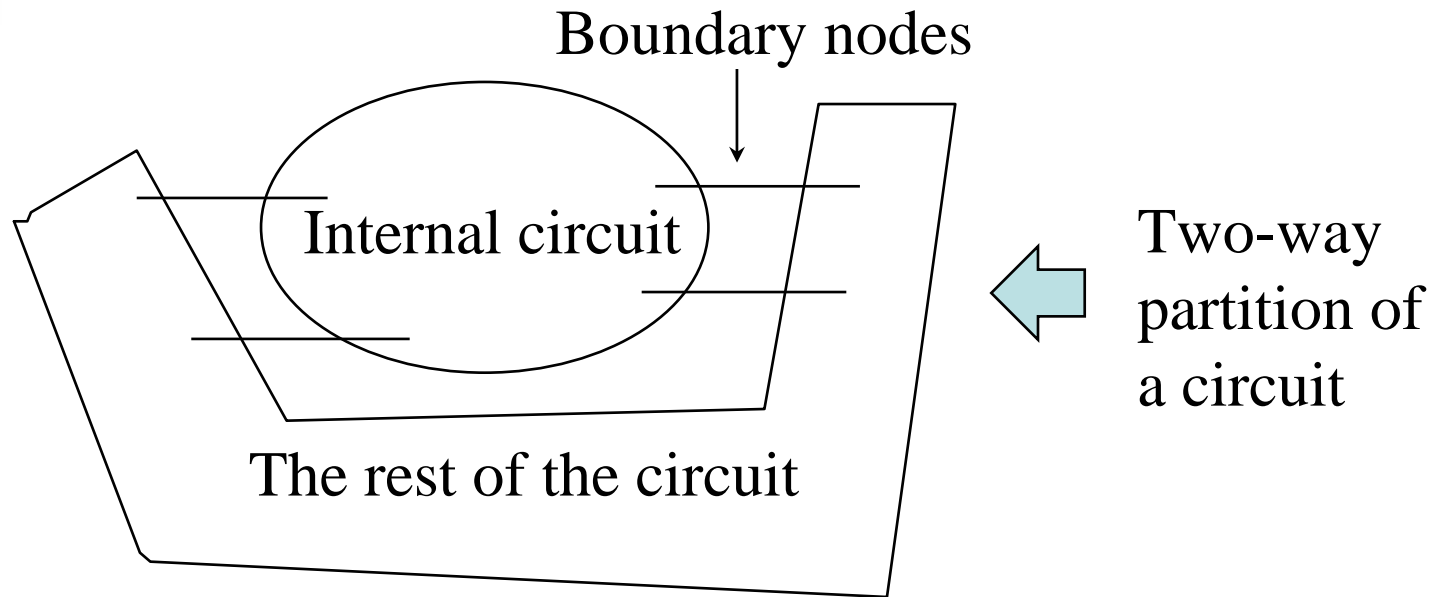
- **Projection-based**
 - AWE, PVL, Matrix PVL, Arnoldi Model, PRIMA
 - Difficulty to deal with mutual inductors (M elements) and their equivalent models (with controlled sources)
- **Node reduction and rational approximation**
 - TICER, DTT, Circuit Crunching, Y-Delta, HMOR
 - Essentially symbolic Gaussian elimination
 - Advantage:
 - applicable to both passive and active circuits
 - No need to solve the whole circuits



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- Model order reduction
- **General hierarchical model order reduction**
 - Hierarchical: block Gaussian elimination
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Partitioned Circuit Matrix



$$\begin{bmatrix} M^{II} & M^{IB} \\ M^{BI} & M^{BB} & M^{BR} \\ & M^{RB} & M^{RR} \end{bmatrix} \begin{bmatrix} x^I \\ x^B \\ x^R \end{bmatrix} = \begin{bmatrix} b^I \\ b^B \\ b^R \end{bmatrix}$$

A light blue arrow points from the equation towards the text "Partitioned circuit equations".



Block Gaussian Elimination

- The suppressed circuit matrix becomes:

$$\begin{bmatrix} M^{BB^*} & M^{BR} \\ M^{RB} & M^{RR} \end{bmatrix} \begin{bmatrix} x^B \\ x^R \end{bmatrix} = \begin{bmatrix} b^{B^*} \\ b^R \end{bmatrix}$$

where

$$M^{BB^*} = M^{BB} - M^{BI} (M^{II})^{-1} M^{IB}$$

$$b^{B^*} = b^B - M^{BI} (M^{II})^{-1} b^I$$

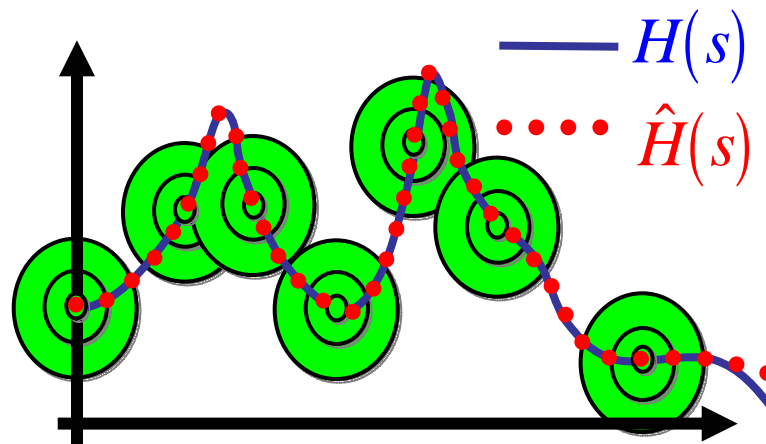
or

$$a_{uv}^{BB^*} = a_{uv}^B - \frac{1}{\det(A^{II})} \sum_{k_1, k_2=1}^m a_{uk_1}^{BI} \Delta_{k_2 k_1}^{II} a_{k_2 v}^{IB}, \quad u, v = 1, \dots, l$$

$$b_u^{B^*} = b_u^B - \frac{1}{\det(A^{II})} \sum_{k_1, k_2=1}^m a_{uk_1}^{BI} \Delta_{k_2 k_1}^{II} b_{k_2}^{IB}, \quad u, v = 1, \dots, l$$

Hierarchical S-Domain Reduction (HMOR) (ICCAD'03)

- Basic idea
 - ✓ Compute s -polynomials for determinants and cofactors via hierarchical symbolic analysis techniques
 - ✓ Keep the exact or only limited order for each poly
- Multi-point frequency expansion for wideband accuracy





Properties of HMOR

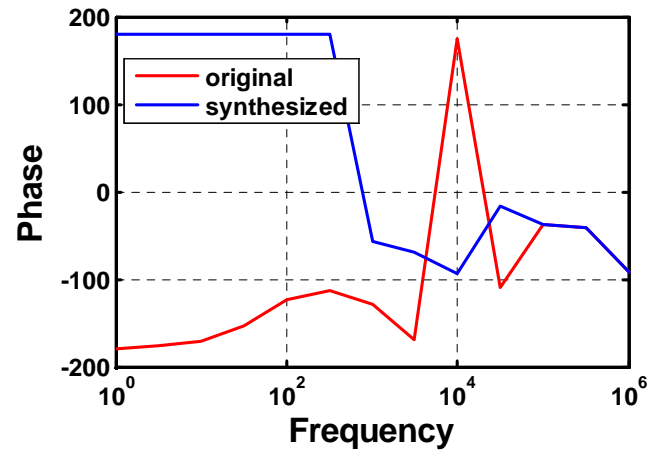
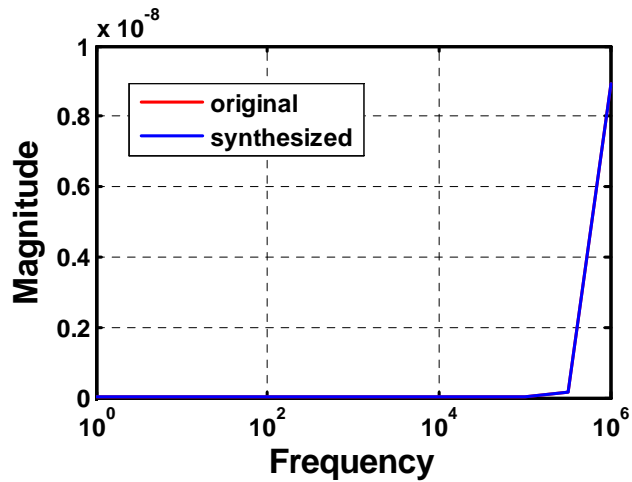
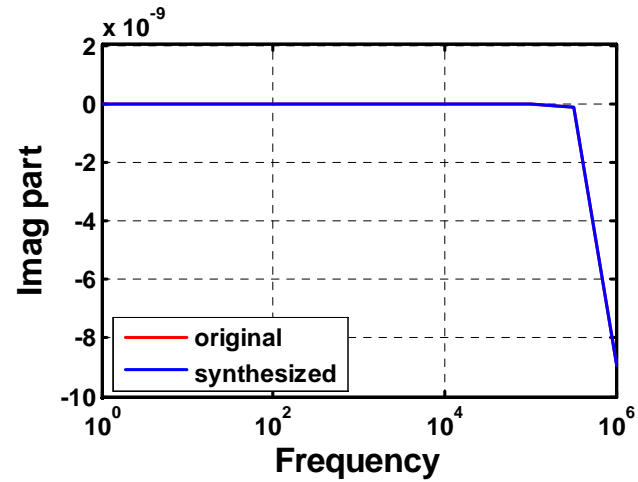
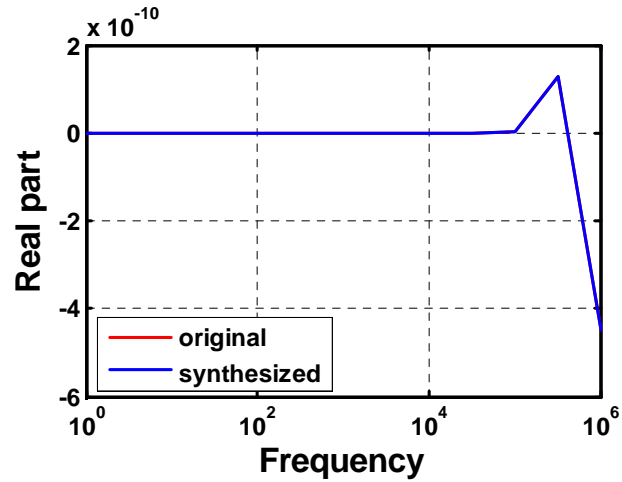
- **Theorem:** HMOR method dose not change the reciprocal property of a linear system.
 - A reciprocal network is one in which the power losses are the same between any two ports regardless of direction of propagation.
 - If a system is reciprocal, the reduced system by HMOR is reciprocal.
 - If a system is not reciprocal, the reduced system by HMOR is not reciprocal.



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 - General: any linear passive or active circuits
- **Optimization considering magnitude and phase responses**
 - Motivation for considering both magnitude and phase responses
 - Constrained least square based optimization
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Motivation





Optimization Problem

- Admittance matrix of the reduced order model

$$\hat{\mathbf{Y}}(s) = \begin{bmatrix} \hat{Y}_{1,1}(s) & \cdots & \hat{Y}_{1,n}(s) \\ \vdots & \ddots & \vdots \\ \hat{Y}_{n,1}(s) & \cdots & \hat{Y}_{n,n}(s) \end{bmatrix}$$

- Optimization problem (only considering magnitude)

$$\min \left(\sum_{k=1}^T \|\hat{Y}_{p,q}(s_k) - \tilde{Y}_{p,q}(s_k)\|_2^2 \right)$$

Where $\tilde{Y}_{p,q}(s_k)$ is the exact values of the admittance at the entry (p,q) at the k -th frequency point



Optimization Problem (cont'd)

- Basic Idea:

$$\hat{Y}(s) = s\hat{Y}_\infty + \hat{Y}_0 + \sum_{m=1}^M \frac{rr_m}{s - pr_m} + \sum_{n=1}^N \left(\frac{rc_n}{s - pc_n} + \frac{rc_n^*}{s - pc_n^*} \right)$$

- Find a set of residues such that the errors are mixed in terms of both magnitude and phases.



Optimization Problem

Some definitions

$$x = [x_1^r \quad \cdots \quad x_M^r \quad x_1^c \quad \cdots \quad x_{2N}^c \quad Y_0 \quad Y_\infty]^T$$

$$A_k = [a_1^r(s_k) \quad \cdots \quad a_M^r(s_k) \quad a_1^c(s_k) \quad \cdots \quad a_{2N}^c(s_k) \quad 1 \quad s_k]$$

Where
$$a_m^r(s_k) = \frac{1}{s_k - pr_m}$$

$$a_n^c = \frac{1}{s_k - pc_n} + \frac{1}{s_k - pc_n^*}, a_{n+1}^c = \frac{j}{s_k - pc_n} - \frac{j}{s_k - pc_n^*}$$



Minimization for Real and Imaginary Part

Define

$$A_{lin} = \begin{bmatrix} re(A) \\ im(A) \end{bmatrix}, Y_{lin} = \begin{bmatrix} re(\tilde{Y}) \\ im(\tilde{Y}) \end{bmatrix}$$

Then $min(\|A_{lin}x - Y_{lin}\|_2^2)$



Constraints for Sign of Phase

define

$$Y_D = \text{diag}(Y_{lin})$$

$$D_{lin} = Y_D A_{lin}$$

Then

$$D_{lin}x \geq 0$$



Constraints for Value of Phase

- Define

$$Y_I = \text{diag} \left(\left[\begin{array}{c} \text{im}(\tilde{Y}) \\ \text{re}(\tilde{Y}) \end{array} \right] \right)$$

$$I_{lin} = \begin{bmatrix} 1 & \cdots & 0 & -1 & \cdots & 0 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 1 & 0 & \cdots & -1 \end{bmatrix}$$

$$C_{lin} = I_{lin} Y_I A_{lin}$$

Then $C_{lin} x = 0$



Optimization Problem

- The constrained linear least square optimization problem considering both magnitude and phase

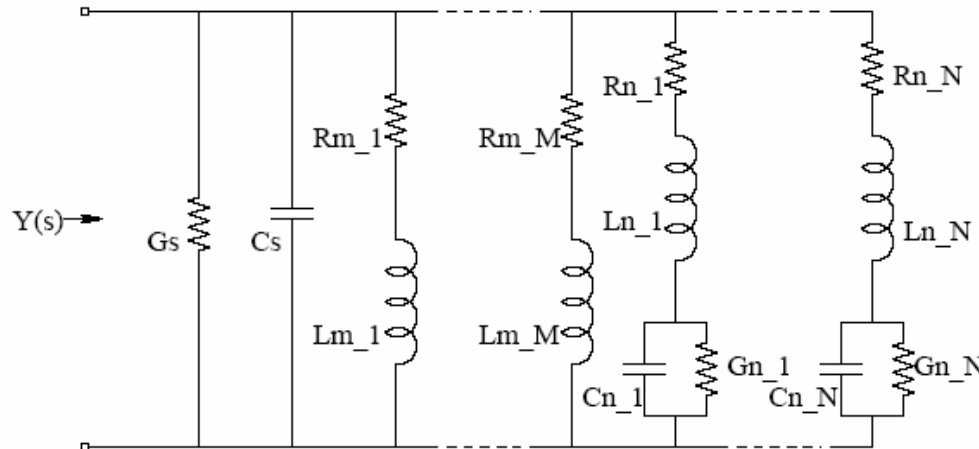
$$\min(\|A_{lin}x - Y_{lin}\|_2^2) \quad \text{subject to} \quad \begin{array}{l} D_{lin}x \geq 0 \\ C_{lin}x = 0 \end{array}$$

Realization (one-port)

- Foster's canonical form for 1x1 admittance $Y(s)$

$$Y(s) = sY_\infty + Y_0 + \sum_{m=1}^M \frac{x_m^r}{s - pr_m} + \sum_{n=1}^N \left(\frac{x_n^c + x_{n+1}^c j}{s - pc_n} + \frac{x_n^c - x_{n+1}^c j}{s - pc_n^*} \right)$$

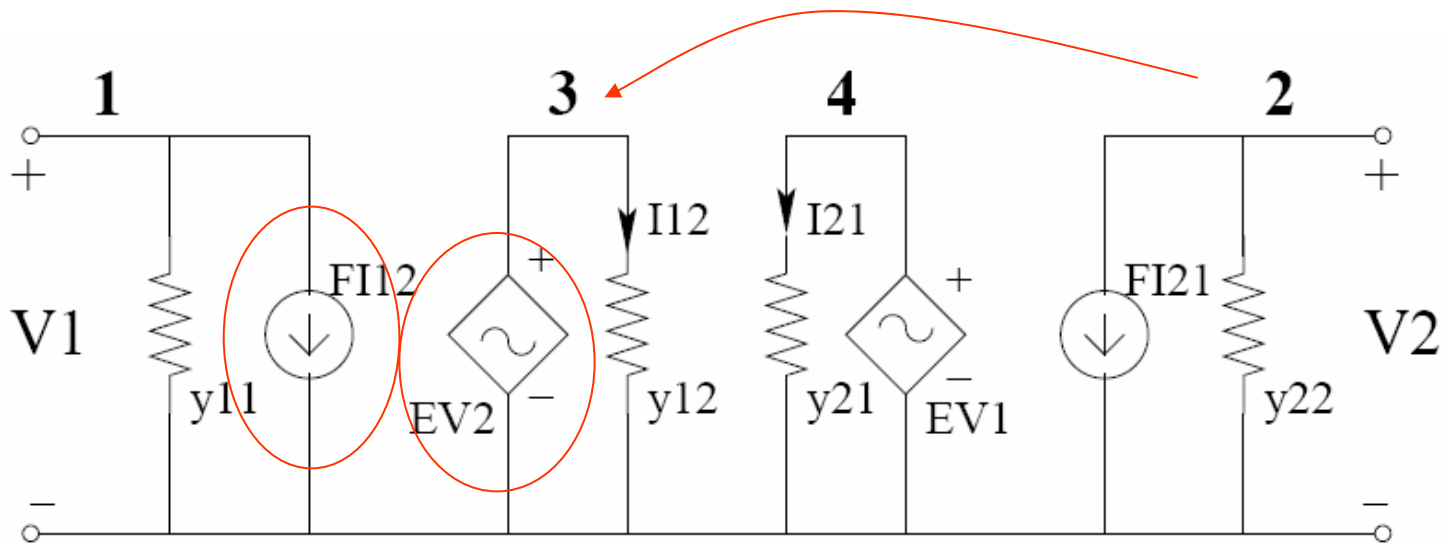
- Realization of Foster's canonical form



Multi-port non-reciprocal Realization

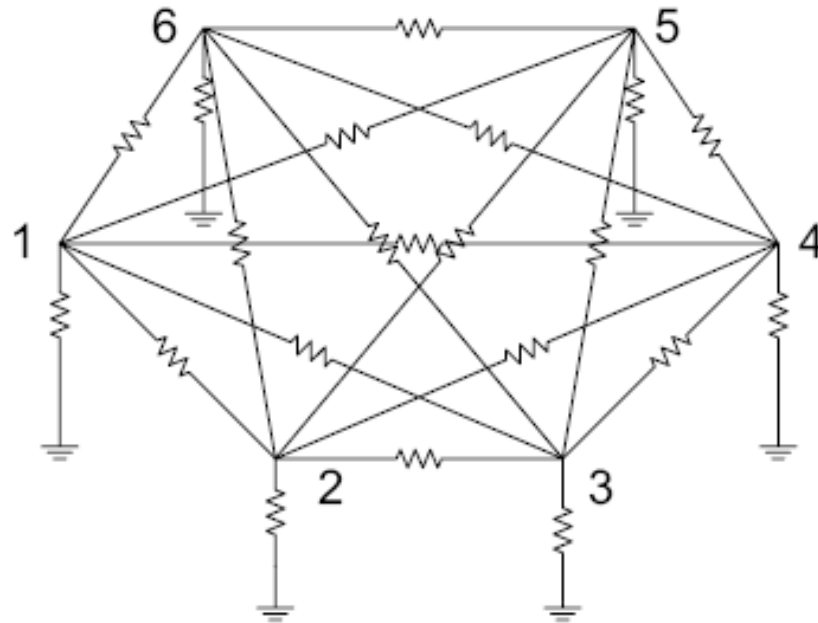
- 2 port

$$Y_{2 \times 2}(s) = \begin{bmatrix} y_{11}(s) & y_{12}(s) \\ y_{21}(s) & y_{22}(s) \end{bmatrix},$$



General Multi-Port Network Realization

- For general $n \times n$ network



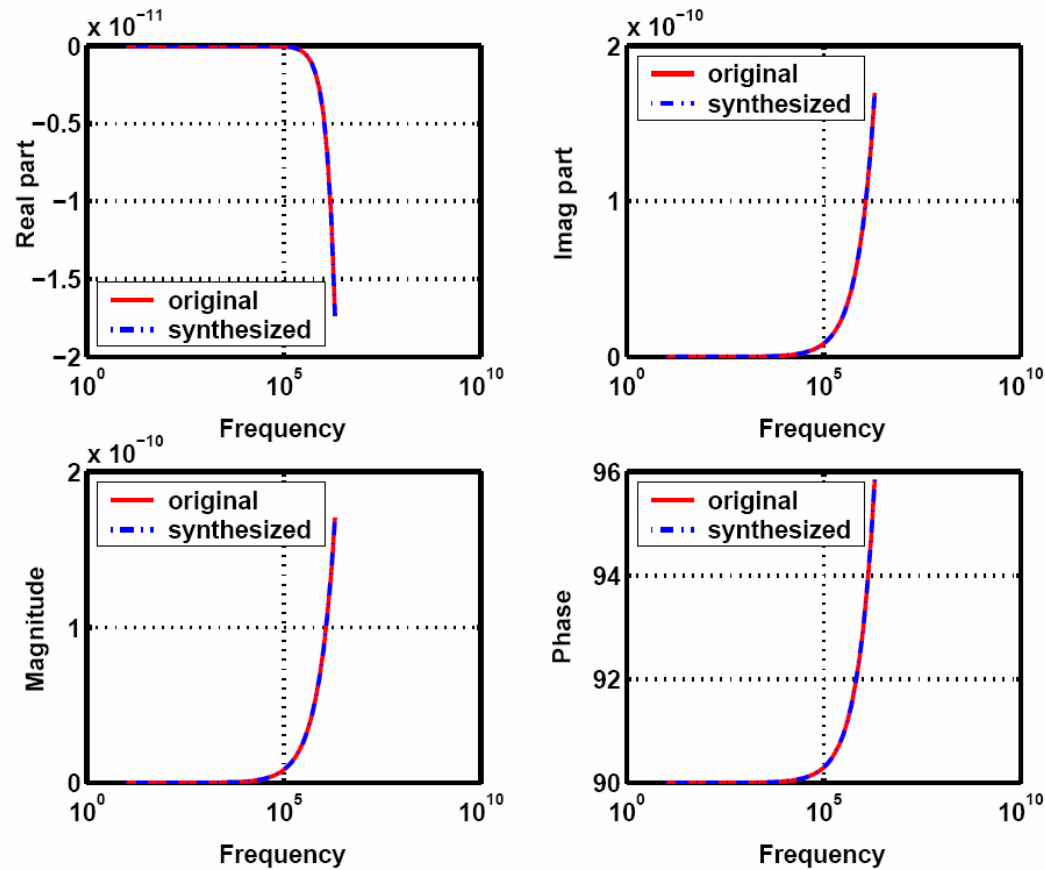


Outline

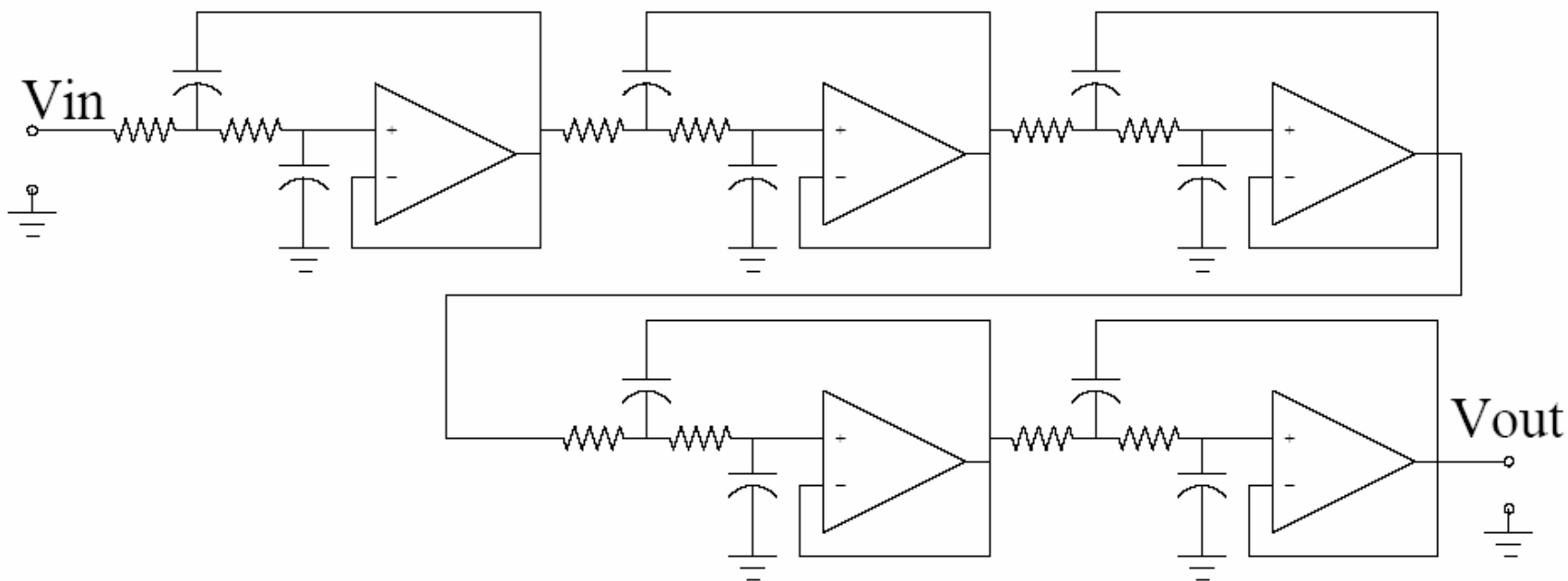
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- Y_{12} frequency responses of the original circuit and the reduced model



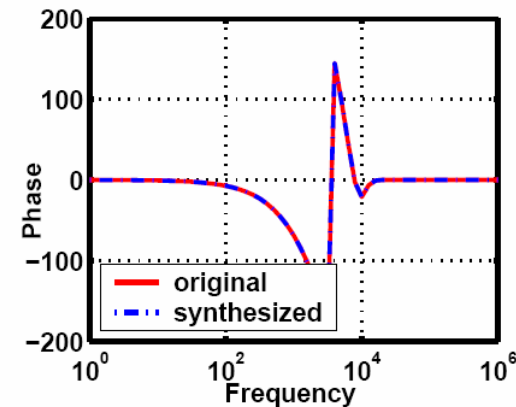
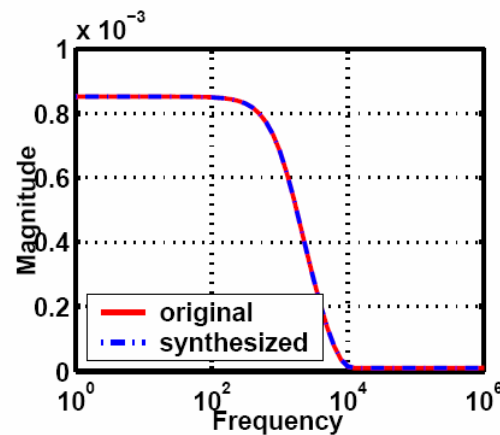
An active Sallen-Key low-pass filter



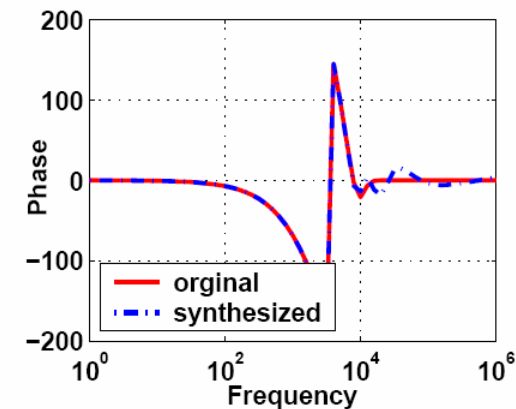
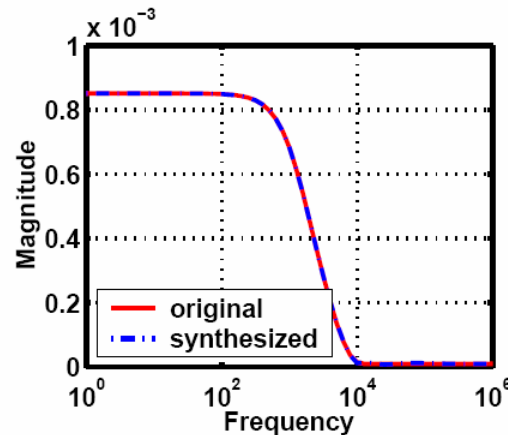
An active Sallen-Key low-pass filter

- Y_{12} Frequency-domain pulse responses

- Considering phase

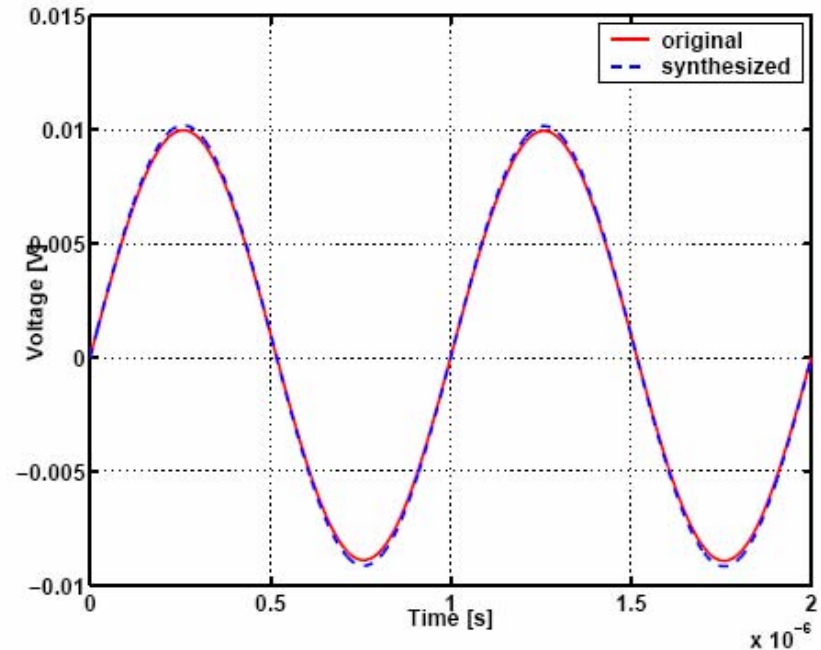
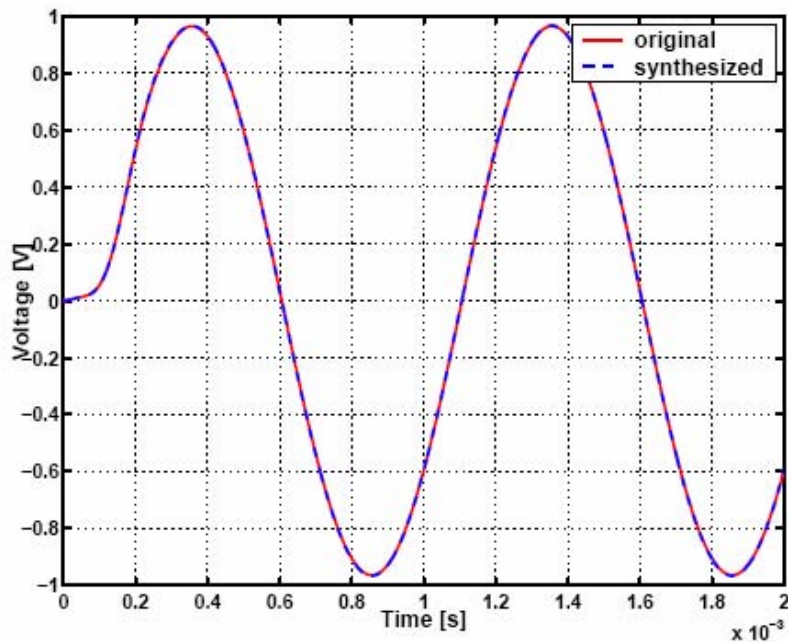


- Without Considering phase



An active Sallen-Key low-pass filter

- Time-domain responses excited by different sources





An active Sallen-Key low-pass filter

- Reduction efficiency:
 - Original model: 636 elements
 - Realized reduced order model: 88 RLC, 2 VCVS, 2CCCS elements
 - Reduction ratio: 85.5%.

Realized circuits have very regular structure, which can be utilized for further reduction or acceleration in time-domain simulation



Conclusions

- A general multi-point hierarchical model order reduction and realization flow for active circuits
- Constrained linear least square optimization technique considering both magnitude and phase responses for any linear active network
- Multi-port non-reciprocal circuit realization
- SPICE-in SPICE-out reduction/realization for maximum model portability and flexibility