



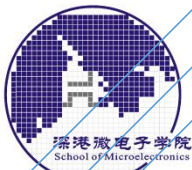
EI-MOR: A Hybrid Exponential Integrator and Model Order Reduction Approach for Transient Power/Ground Network Analysis

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Outline

- Background
- Relations between EI and MOR
- Hybrid EI-MOR method
- Result
- Conclusion

Background

- ❑ **Circuits are bigger**

- ❑ Millions of unknowns for post-layout sim
- ❑ Even larger for power/ground (P/G) network simulation

- ❑ **SPICE is slow**

- ❑ Huge sparse matrix LU, many time points
- ❑ Weeks of verification time

- ❑ **Acceleration techniques exist**

Mainstream circuit simulation acceleration methods



Mode Order Reduction(MOR)

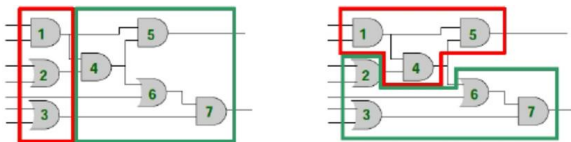
- frequency-domain
- moment-matching
- maintain the I/O relationship

$$e^A x \approx V_m e^{H_m} e$$

Exponential Integrator(EI)

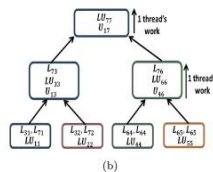
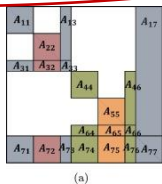
- time-domain
- accuracy, large step sizes, high parallelism
- General circuit simulation scenarios

**Focus of
this work**

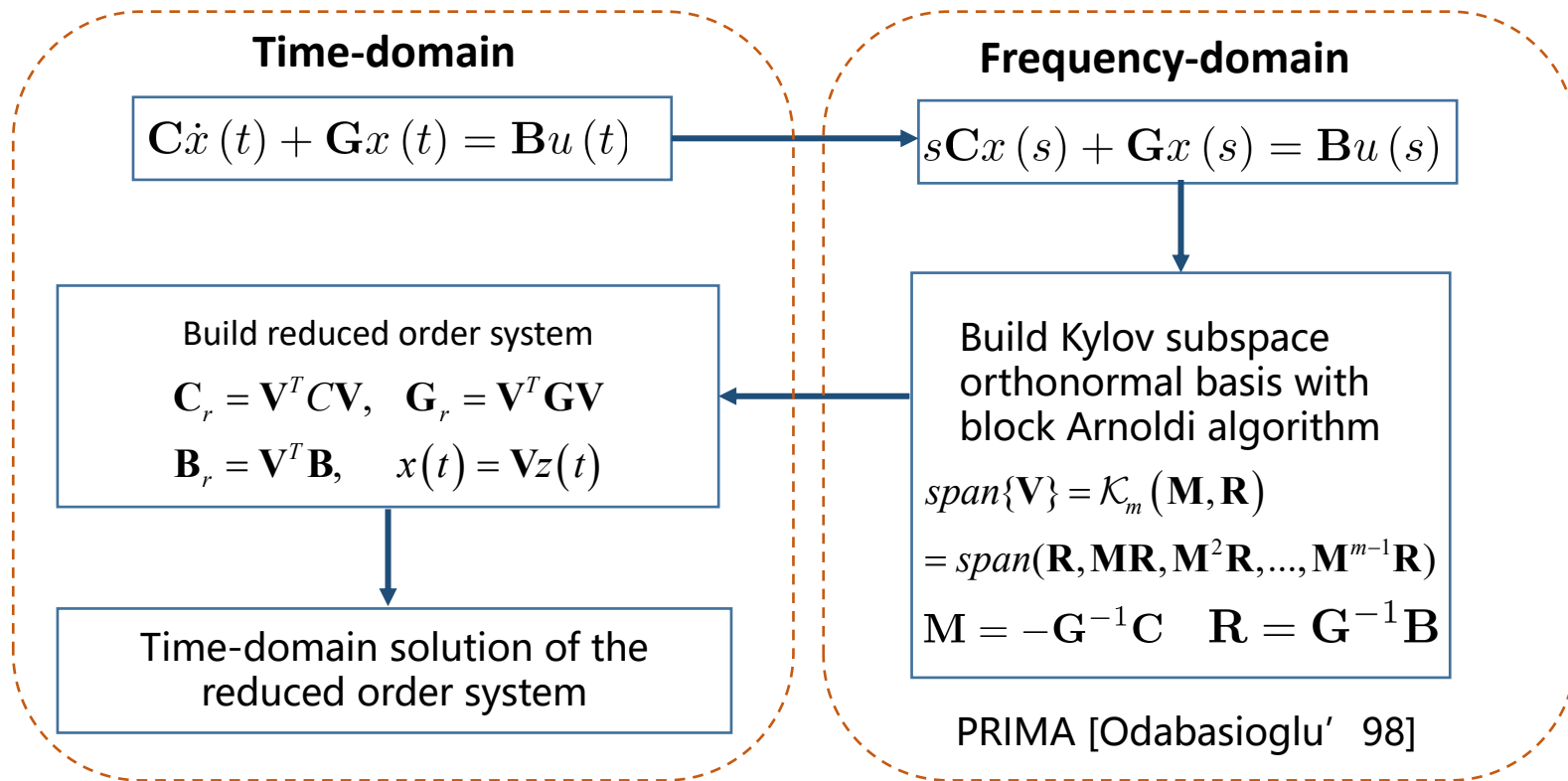


Parallel computing

- graph theory
- circuit/matrix division
- parallel hardware (multi-threaded/multi-core/GPU) acceleration



Krylov Subspace based Model Order Reduction



Exponential Integrator based on Rational Krylov Subspace

$$\mathbf{C}\dot{x}(t) + \mathbf{G}x(t) = \mathbf{B}u(t)$$

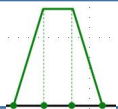
Convert to ODE

$$\dot{x}(t) = \mathbf{A}x(t) + \mathbf{C}^{-1}\mathbf{B}u(t)$$

Analytical solution

$$x_{n+1} = e^{\mathbf{A}h}x_n + \int_0^h e^{\mathbf{A}(h-\tau)}\mathbf{C}^{-1}\mathbf{B}u(\tau) d\tau$$

Input piece-wise-linear
(pwl) approximation



MEXP

$$x_{n+1} = \exp\left(\begin{bmatrix} \mathbf{A}\mathbf{C}^{-1}\mathbf{B}u_n \\ 0 \end{bmatrix} h\right) \begin{bmatrix} x_n \\ 1 \end{bmatrix}$$

Get $\mathbf{V}_m, \mathbf{H}_m$
with Arnoldi algorithm
 $\text{span}\{\mathbf{V}_m\} = \mathcal{K}_m\left((I - \gamma\mathbf{A})^{-1}, \mathbf{R}\right)$
 $\mathbf{H}_m = \mathbf{V}_m^T \mathbf{A} \mathbf{V}_m$ γ : shift parameter

$$e^{\mathbf{A}h}v \approx \beta \mathbf{V}_m e^{(I - \mathbf{H}_m^{-1})h/\gamma} e_1$$

Posteriori residual
estimate

$$res = \beta h_{m+1,m}(\mathbf{C}/\gamma + \mathbf{G})v_{m+1}e_m^T \mathbf{H}_m^{-1} e^{\mathcal{H}_m h} e_1$$

All in time-domain

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Why EI and MOR

- ❑ **Same goal**

- ❑ Accelerate transient circuit simulation

- ❑ **Similar approximation method**

- ❑ Krylov subspace approximation

- ❑ **Separate research topics**

- ❑ MOR: work mostly in the frequency-domain, classical
 - ❑ EI: pure time-domain method, younger (since 2010)

Contribution 1: The relation between MOR and EI

Equivalence between EI and MOR

MOR

$$s\mathbf{C}x(s) + \mathbf{G}x(s) = \mathbf{B}u(s)$$

Generate \mathbf{V}_M matching m
moments around s_0
Obtain the reduced system

$$\mathbf{C}_r \dot{\mathbf{z}}(t) + \mathbf{G}_r \mathbf{z}(t) = \mathbf{B}_r u(t)$$

Advancing one step
using the **EI method**

$$\mathbf{z}_1 = e^{\mathbf{A}_r h} \mathbf{z}_0 + h \phi_1(\mathbf{A}_r h) \mathbf{C}_r^{-1} \mathbf{B}_r u_0$$

Recover the original
solution

$$\mathbf{x}_1^M = \mathbf{V}_M \mathbf{z}_1$$

EI

$$\mathbf{C} \dot{\mathbf{x}}(t) + \mathbf{G} \mathbf{x}(t) = \mathbf{B} u(t)$$

Analytical solution with
PWL approximation

$$\mathbf{x}_1 = e^{\mathbf{A}h} \mathbf{x}_0 + h \phi_1(\mathbf{A}h) \mathbf{C}^{-1} \mathbf{B} u_0$$

Rational Krylov subspace
with shift parameter
 $\gamma = 1/s_0$ to get \mathbf{V}_E and \mathbf{H}_E

$$\mathbf{x}_1^E = e^{\mathbf{A}h} \mathbf{v} \approx \|\mathbf{v}\|_2 \mathbf{V}_E e^{\mathbf{H}_E h} \mathbf{e}_1$$

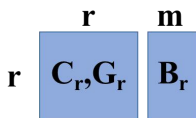
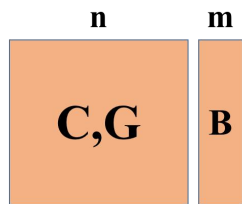
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Zero initial condition \mathbf{x}_0 , a single-column input

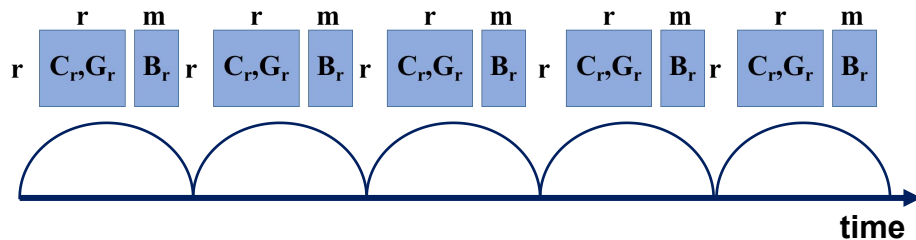
Illustration of EI and MOR Equivalence

Frequency-domain reduction
(multiple inputs)

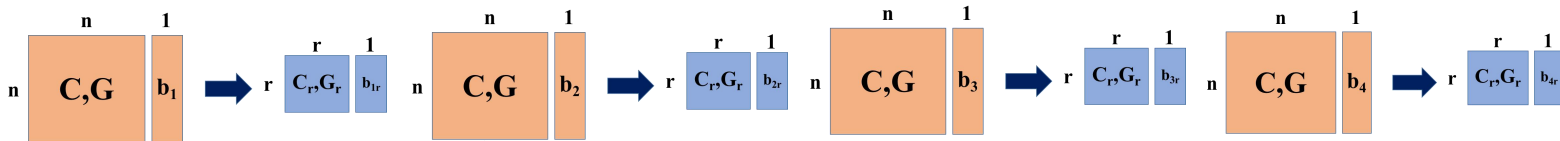
MOR
workflow



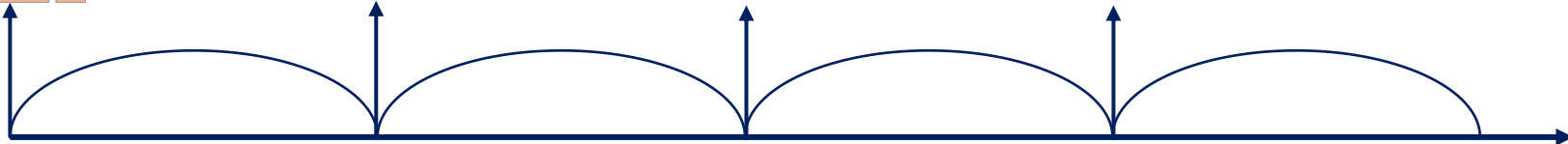
Time-domain simulation (using EI)



Simple MOR
(1 input only)



EI
workflow



Direct time-domain simulation (one MOR per time step)

Equivalence: Corollary

- Main Theorem also holds for inhomogeneous initial condition $u_0 \neq 0$
 - Decompose original system into two sub-systems to MOR process
- EI using $(N + 1)$ -dimensional V matches the first moment of the exact solution
 - Another perspective explaining the superiority of EI

Comparisons of EI and MOR in circuit simulation

MOR: “once-and-for-all”

- Reduction **once in frequency-domain**, use ROM **repeatedly in time-domain**
- No knowledge of actual inputs, so require accuracy for **any input** during **whole span**
- Cost is high especially for **many-input** and **wide-spectrum** systems
- Difficult to choose expansion points and moments
- **Difficult** to handle nonlinear circuits
- ROMs are **re-usable**

EI: “on-the-fly”

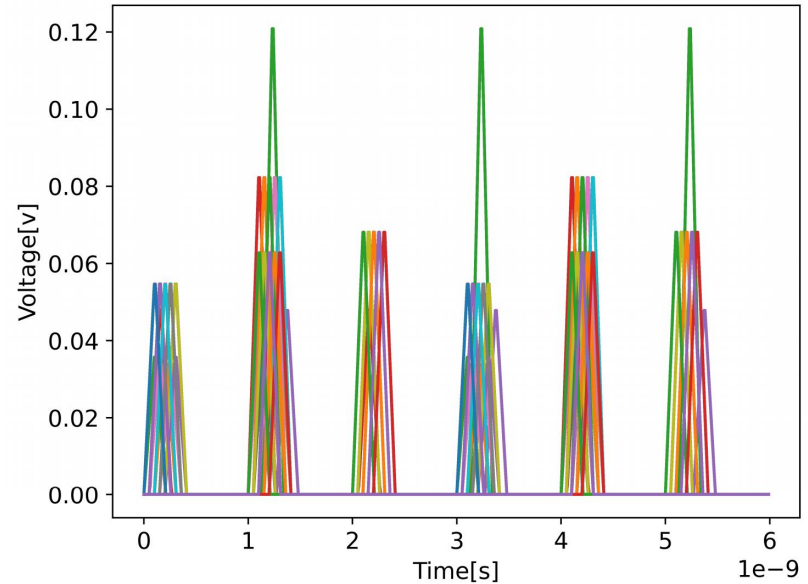
- One MOR per time step, pure **time-domain**
- Full knowledge of actual inputs, require accuracy for **one input** within **one step**
- **Independent of** the number of inputs
- A single expansion point usually suffices. Easy to determine moments
- **Easy** to deal with nonlinear circuits
- ROMs **not re-usable**

Outline

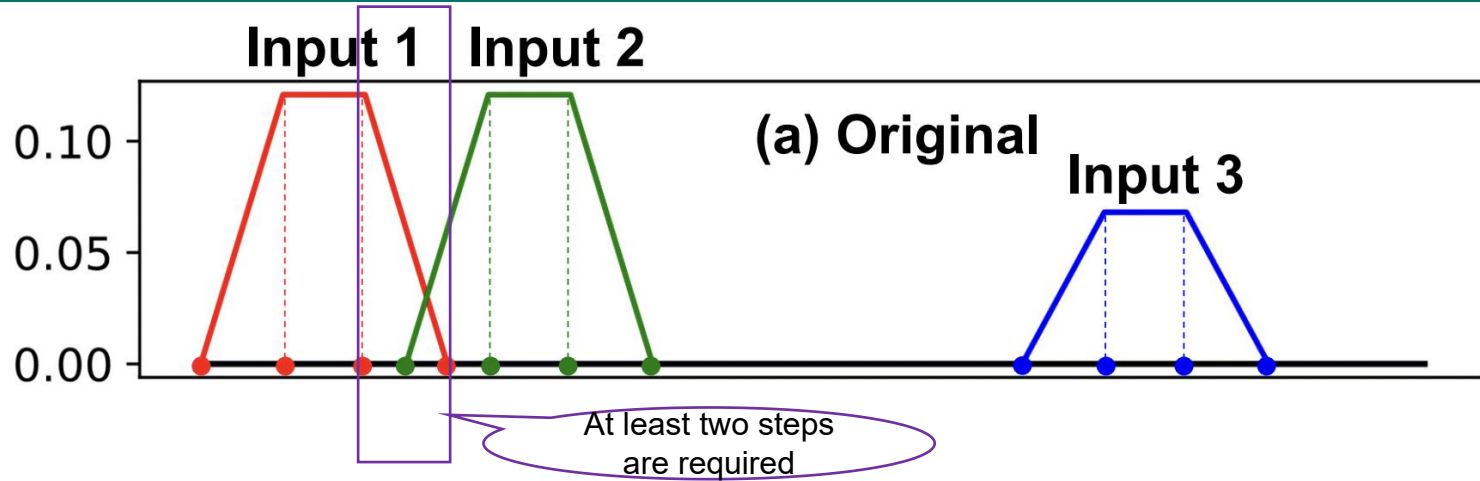
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Hybrid EI-MOR method for P/G Analysis

- In P/G networks nonlinear devices are modeled as a large number of independent current sources
- Typically pulse or PWL waveforms



Input Misalignment Problem for EI



- ❑ PWL assumption in EI \rightarrow every input linear within the same step
- ❑ Input misalignment will induce many breakpoints
- ❑ Usually not an issue for traditional SPICE, but serious for EI

Contribution 2: Inspired by EI-MOR equivalence, develop a hybrid EI-MOR method to address the misaligned input issue

Motivations

	Input #	Time step #
El cost	Independent	Sensitive
MOR cost	Sensitive	Not sensitive

Our idea: a hybrid EI-MOR method

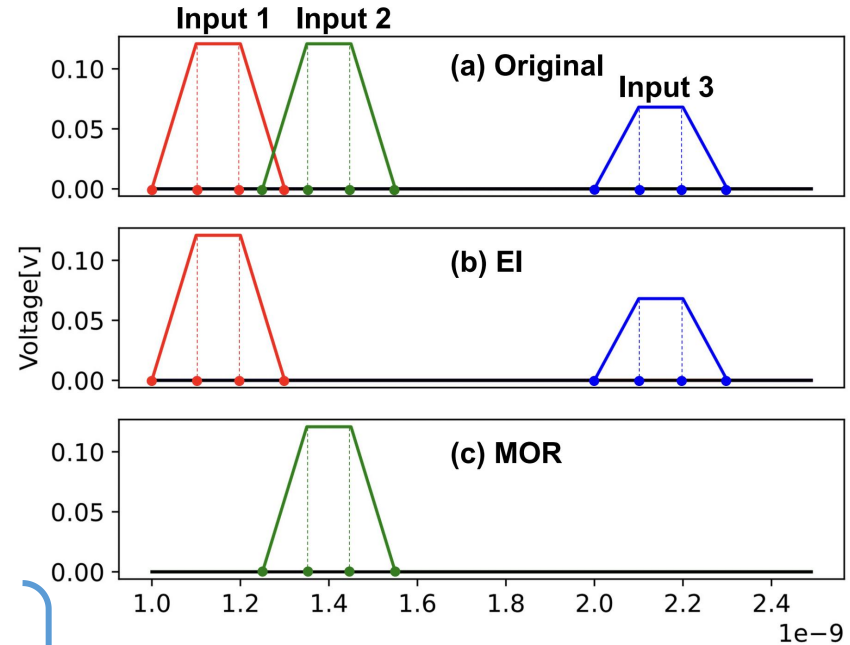
- ❑ EI handles well-aligned inputs (majority)
- ❑ MOR deals with misaligned inputs(minority)

Divide the original inputs

- Divide inputs into two groups
 - A majority group of well-aligned inputs for EI
 - A minority group of misaligned inputs for MOR
- EI inputs handled as usual with all breakpoints
- MOR inputs used in PRIMA to generate a reduced model for later time domain simulation
- The EI solution and MOR solution superimposed to obtain the final solution

$$\begin{pmatrix} 1 & 2 & 3 \\ \text{blue bar} & \text{blue bar} & \text{blue bar} \end{pmatrix} = \begin{pmatrix} 1 & 2 & 3 \\ \text{blue bar} & \text{white bar} & \text{blue bar} \end{pmatrix} + \begin{pmatrix} 1 & 2 & 3 \\ \text{white bar} & \text{blue bar} & \text{white bar} \end{pmatrix}$$

u u_{EI} u_{MOR}



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- Introduction and Background
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- **Results**
- Conclusion

Result: Verification of EI-MOR Equivalence

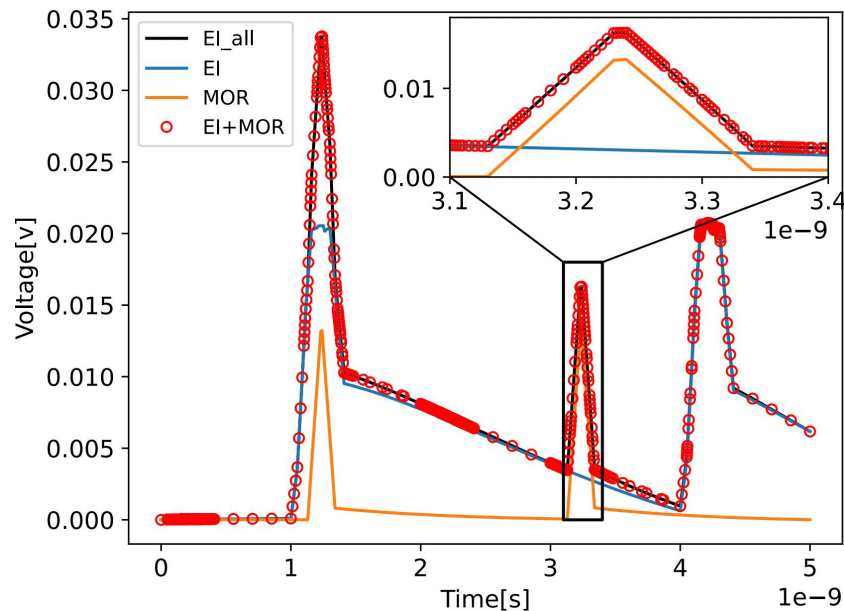
- ❑ Data: ibmpg1t
- ❑ Nonzero initial condition x_0
- ❑ a single-column input
- ❑ 3 different expansion points and 3 different moments matched of each expansion point

**EI solution and MOR solution
are numerically equal**

Expansion Point	Moments	$\frac{\ x^{EI} - x^{MOR}\ }{\ x^{EI}\ }$
1e7	4	2.5418×10^{-14}
	6	8.1615×10^{-14}
	8	1.2139×10^{-13}
1e8	4	2.3310×10^{-14}
	6	5.9894×10^{-14}
	8	1.0884×10^{-13}
1e9	4	4.6395×10^{-14}
	6	7.7582×10^{-13}
	8	6.3915×10^{-12}

Result: Validity of “divide-and-superposition”

- ❑ Test instance: ibmpg1t (5883 current source, 3 for MOR, 5880 for EI)
- ❑ MOR based on PRIMA ($s_0=1e8$, $m=15$)
- ❑ Both EI and MOR inputs have non-trivial contribution to the transient responses
- ❑ Produce the correct total solution when superimposed



Result: Performance of EI-MOR

$i_{MOR} = \arg \max_i (k_0 - k_i) .$ MOR inputs picked based on ability to reduce breakpoints

Case	Node	VSource	ISource
ibmpg1t	39681	14308	5883
ibmpg2t	164238	330	18535
ibmpg3t	1041535	955	114191

Case	Method	ISource	bp_{All}	bp_{EI}	$Arnoldi_{EI}$	$t_{EI}(s)$	m_{MOR}	ROM Size	$t_{MOR}^{Reduce}(s)$	$t_{MOR}^{Sim}(s)$	$t_{total}(s)$
ibmpg1t	EI	5883	517	517	16006	225.39	-	-	-	-	225.39
	EI-MOR	5880(EI)3(MOR)	517	342	12570	177.14	15	45	1.08	17.31	195.52
	EI-MOR	5877(EI)6(MOR)	517	277	8496	122.41	15	90	3.48	28.60	154.48
ibmpg2t	EI	18535	533	533	15840	1173.51	-	-	-	-	1173.51
	EI-MOR	18532(EI)3(MOR)	533	389	11509	865.06	15	45	5.32	86.77	957.16
	EI-MOR	18529(EI)6(MOR)	533	277	7700	634.85	15	90	17.12	84.71	736.68
ibmpg3t	EI	114191	509	509	13882	8420.87	-	-	-	-	8420.87
	EI-MOR	114188(EI)3(MOR)	509	375	10067	6306.15	20	60	93.33	471.59	6871.07
	EI-MOR	114185(EI)6(MOR)	509	277	7318	4733.61	20	120	194.13	672.05	5599.79

Note: EI or EI-MOR does not exclude other techniques like circuit partition, parallel LU and spectral graph sparsifier

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-

Conclusion

❑ Equivalence between EI and MOR

- EI equivalent to one MOR per time step

❑ A hybrid EI-MOR method for P/G networks

- Combine EI and MOR to address the misaligned inputs in EI
- Good performance gain

❑ New insights and opportunities for this classic EDA topic.

Thank you!
Q & A

Equivalence: Lemma

Lemma1: With zero initial condition $x_0 = 0$, a single-column \mathbf{b} matrix and a real expansion point $s_0 = 1/\gamma$, let $\mathbf{V}_M \in \mathcal{R}^{N \times m}$ be an m dimensional orthonormal basis with

MOR

$$\begin{aligned} \text{colspan} \{ \mathbf{V}_M \} &= \mathcal{K}_m(\mathbf{M}, \mathbf{R}) \\ &= \mathcal{K}_m \left(-(s_0 \mathbf{C} + \mathbf{G})^{-1} \mathbf{C}, (s_0 \mathbf{C} + \mathbf{G})^{-1} \mathbf{b} \right) \end{aligned}$$

and $\mathbf{V}_E \in \mathcal{R}^{N \times (m+1)}$ be an $m + 1$ dimensional orthonormal basis with

$$\text{colspan} \{ \mathbf{V}_E \} = \mathcal{K}_m \left(\left(\mathbf{I} - \gamma \tilde{\mathbf{A}} \right)^{-1}, \tilde{x}_n \right)$$

EI

So, we can get the **relationship of the orthonormal basis**

$$\mathbf{V}_E = \begin{bmatrix} \mathbf{o} \mathbf{V}_M \\ 1 \quad \mathbf{o}^T \end{bmatrix} \quad \mathbf{o} \in \mathcal{R}^{N \times 1}$$

Equivalence: Main Theorem

With the assumptions stated in Lemma 3.1, EI, when solved with a $(n + 1)$ -dimensional rational Krylov subspace with n , is equivalent to applying a Krylov subspace based MOR, with the expansion point λ_0 and the subspace dimension n , to the original system, and advancing the reduced system one step in the time domain using again the EI approach.

$$x_1^{MOR} = \mathbf{V}_M z_1 = -h \mathbf{V}_M \varphi_1 \left((s_0 \mathbf{I}_M + \mathbf{H}_M^{-1}) h \right) \mathbf{H}_M^{-1} e_1 u_0$$

$$\begin{aligned} x_1^{EI} &\approx h \mathbf{V}_M \varphi_1 \left((s_0 \mathbf{I}_M + \mathbf{H}_M^{-1}) h \right) \mathbf{V}_M^T \mathbf{C}^{-1} \mathbf{b} u_0 \\ &= h \mathbf{V}_M \varphi_1 \left((s_0 \mathbf{I}_M + \mathbf{H}_M^{-1}) h \right) \mathbf{V}_M^T \mathbf{C}^{-1} \mathbf{K} \mathbf{K}^{-1} \mathbf{b} u_0 \\ &= h \mathbf{V}_M \varphi_1 \left((s_0 \mathbf{I}_M + \mathbf{H}_M^{-1}) h \right) \mathbf{V}_M^T (\mathbf{K} \mathbf{C}^{-1})^{-1} \mathbf{V}_M e_1 u_0 \\ &= -h \mathbf{V}_M \varphi_1 \left((s_0 \mathbf{I}_M + \mathbf{H}_M^{-1}) h \right) \mathbf{H}_M^{-1} e_1 u_0 \end{aligned}$$

$$\mathbf{x}_1^{MOR} = \mathbf{x}_1^{EI}$$