

High-Performance Model Compilation for Complex Behavioral Models

September 20th, 2007

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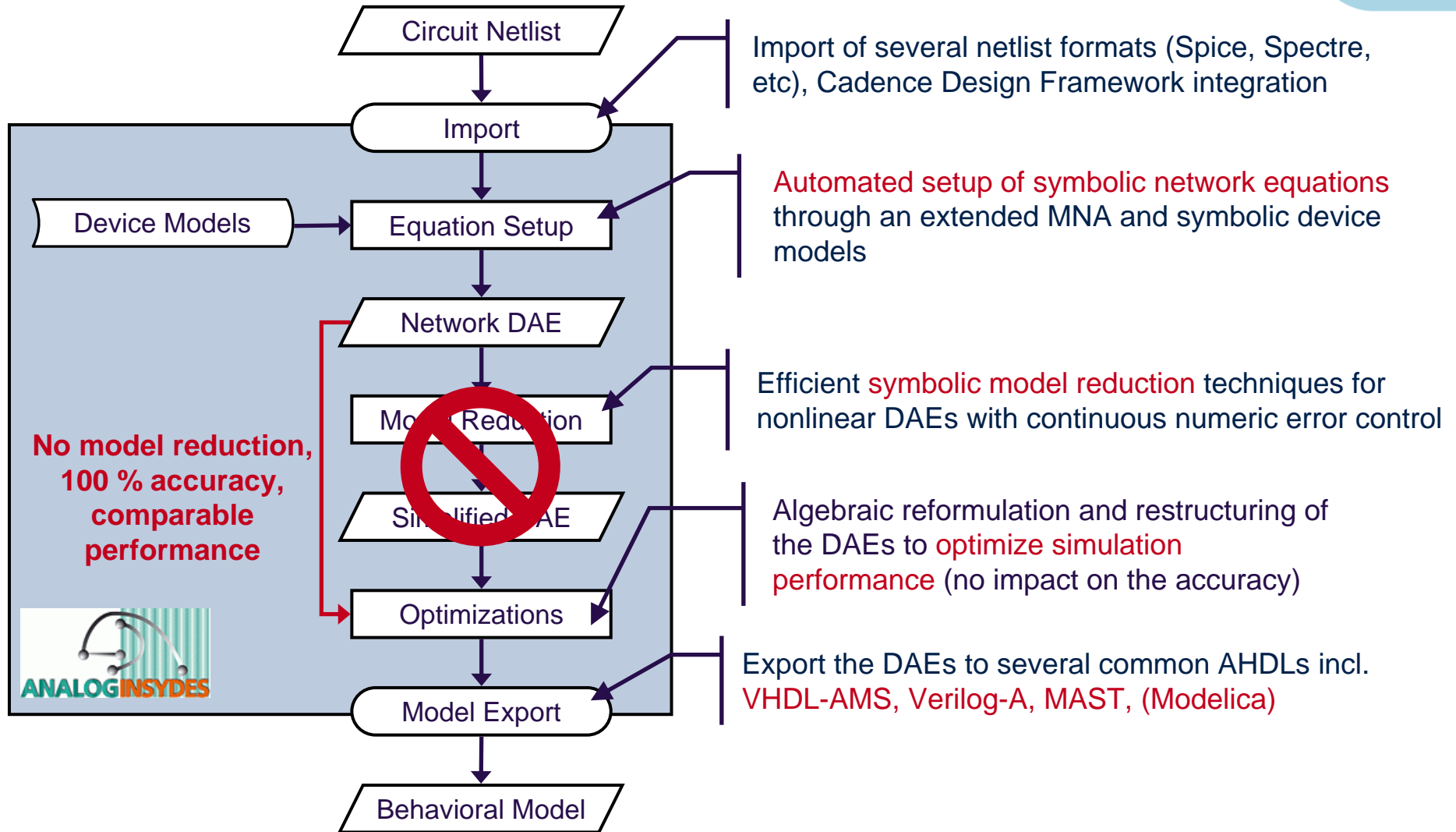
Bottom-Up Modeling Through Symbolic Analysis

Model Compilation

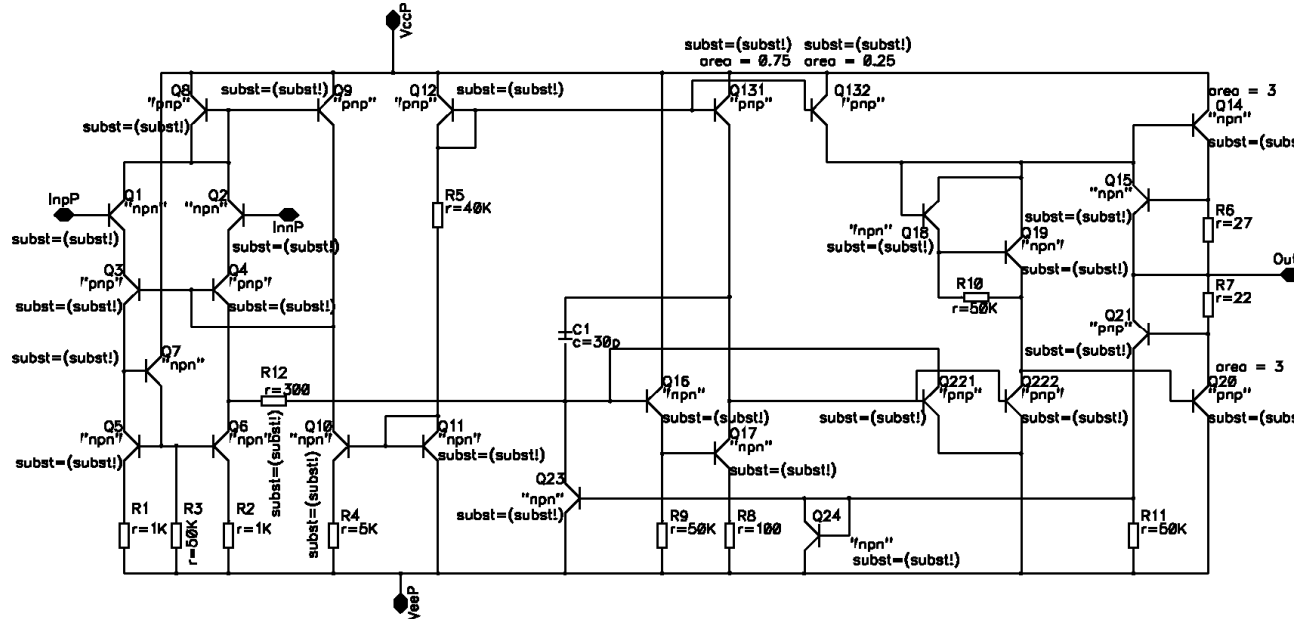
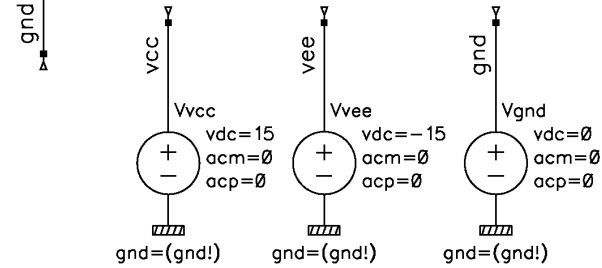
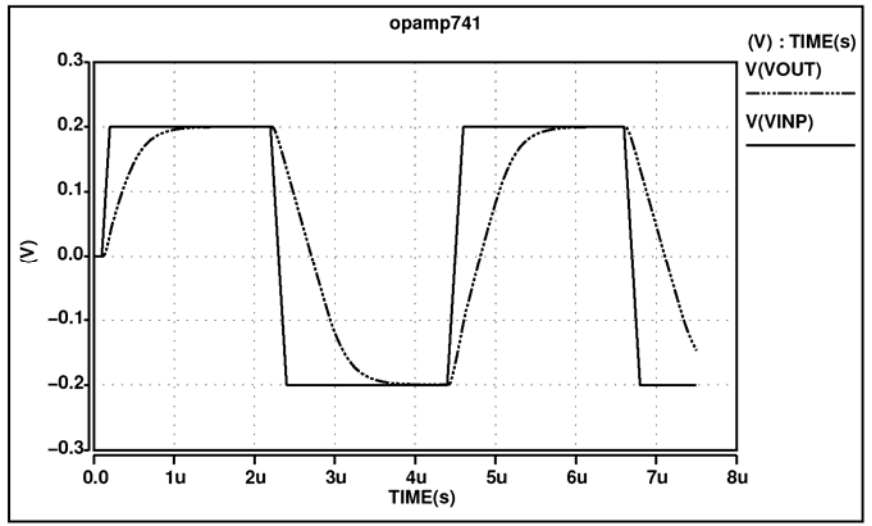
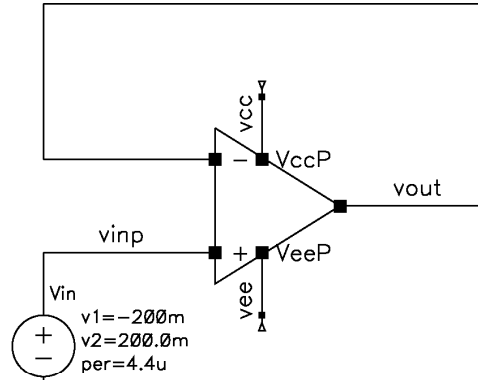
Simulation Results

Summary

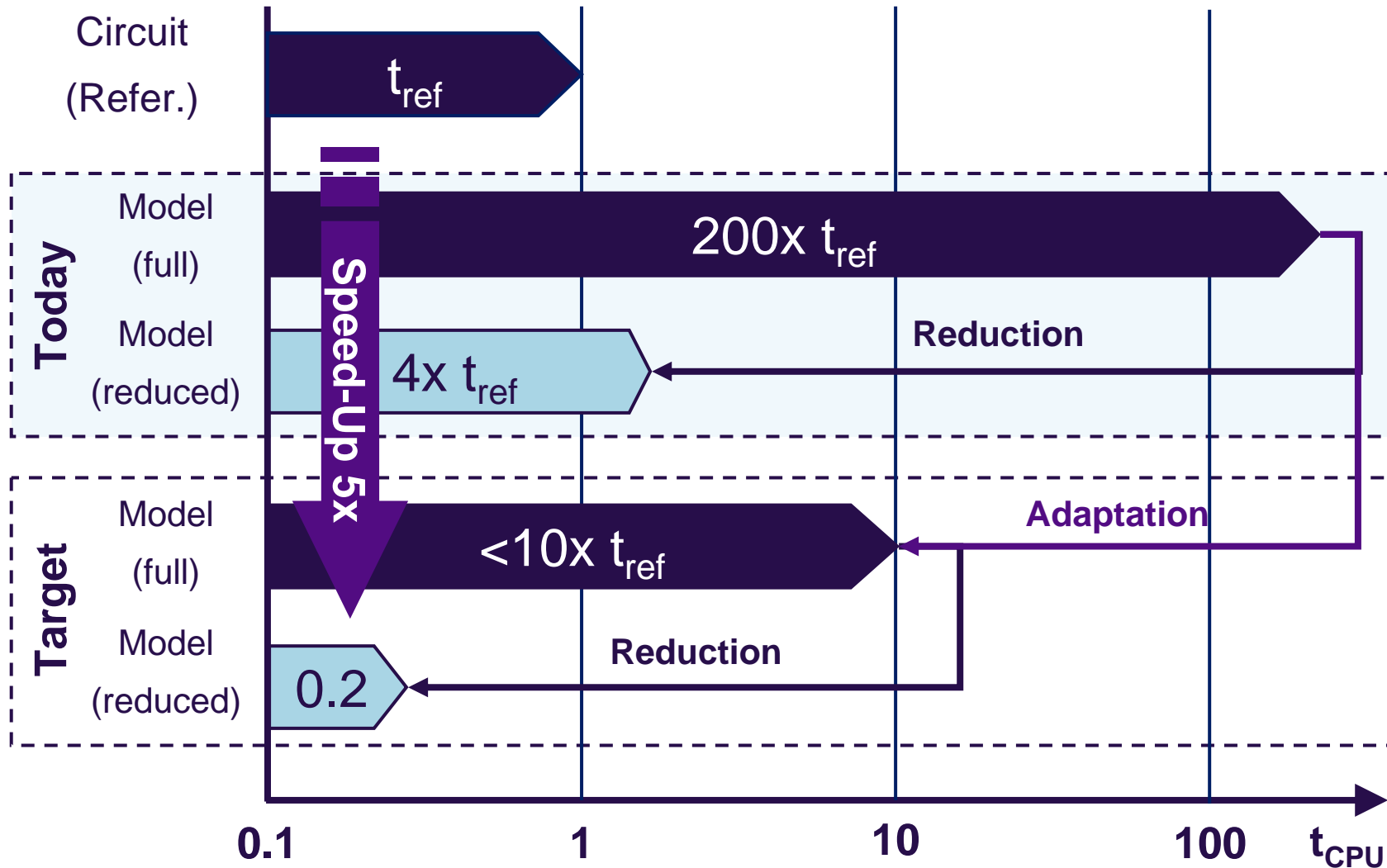
Bottom-Up Modeling Flow



Example μ A741



Motivation



Reasons for Performance Problems



- „Intelligence“ of built-in device models missing
- Equation formulation not optimized for numerical methods
- Model compilers do not efficiently handle such complex behavioral models
- Restrictions by using common AHDLs
 - procedural statements in VHDL-AMS
 - simultaneous equation sets in Verilog-A
 - no optimization of Jacobian matrix possible



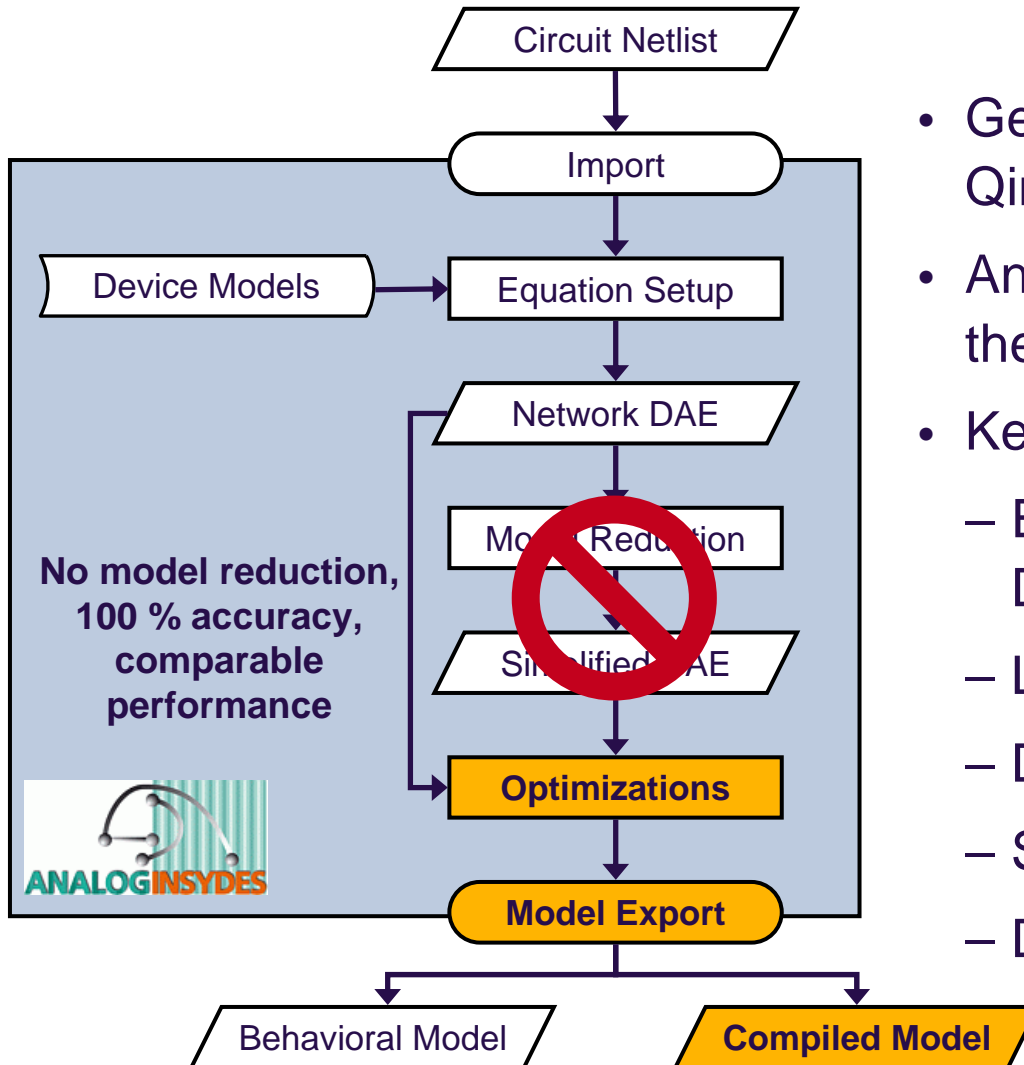
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Model Compilation with Analog Insydes



- Generation of compiled models for Qimonda's inhouse-simulator Titan
- Analog Insydes is used as platform for the model compilation
- Key features:
 - Efficient processing of simultaneous DAE
 - Local solving of sequential equations
 - Data locality (cache)
 - Sparse handling and data structures
 - Direct interface to the simulator kernel



Example – Sequential Network Equations

Sequential

$$vd[t] = -\frac{RS I\$AC[t]}{AREA} + V\$IN[t] - V\$OUT[t]$$

$$id[t] = AREA \left(e^{-\frac{0.00333167q(BV/vd[t])}{k}} IBV + \left(-1 + e^{\frac{0.00181069qvd[t]}{k}} \right) IS \right) + GMIN vd[t]$$

$$cj[t] = AREA CJO \text{ If } \left[vd[t] < 0.25, \frac{1}{(1.-2. vd[t])^{0.333}}, 2.51926 (0.3335 + 0.666 vd[t]) \right]$$

Simultaneous

GND ● $I\$VGND[t] + \frac{V\$GND[t]-V\$OUT[t]}{R0} + C0 (V\$GND'[t] - V\$OUT'[t]) = 0$

A ● $I\$AC[t] + I\$VIN[t] = 0$

K ● $-I\$AC[t] + \frac{-V\$GND[t]+V\$OUT[t]}{R0} + \frac{V\$OUT[t]-V\$OUT2[t]}{RL} + C0 (-V\$GND'[t] + V\$OUT'[t]) = 0$

B ● $I\$VOUT[t] + \frac{-V\$OUT[t]+V\$OUT2[t]}{RL} = 0$

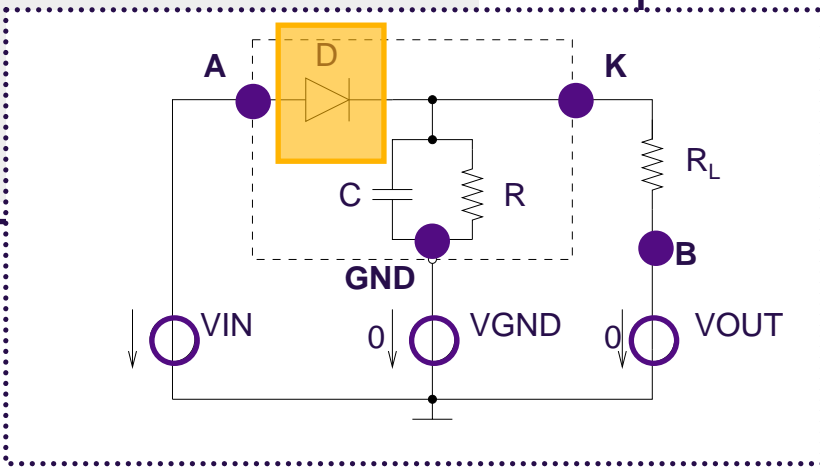
● $I\$AC[t] = id[t] + 1.15 \times 10^{-8} id'[t] + cj[t] vd'[t]$

VIN ○ $V\$IN[t] = VIN$

VOUT ○ $V\$OUT2[t] = 0$

VGND ○ $V\$GND[t] = 0$

- GND** ●
- A** ●
- K** ●
- B** ●
-
- VIN** ○
- VOUT** ○
- VGND** ○





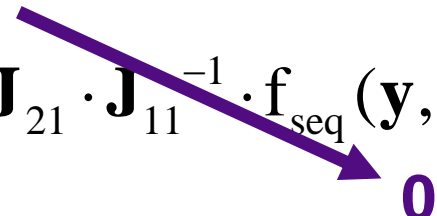
Newton's Method for Sequential DAEs

$$\begin{bmatrix} \partial f_{\text{seq}} / \partial \mathbf{y} & \partial f_{\text{seq}} / \partial \mathbf{x} \\ \partial f_{\text{sim}} / \partial \mathbf{y} & \partial f_{\text{sim}} / \partial \mathbf{x} \end{bmatrix} \cdot \begin{bmatrix} \Delta \mathbf{y} \\ \Delta \mathbf{x} \end{bmatrix} = \begin{bmatrix} \mathbf{J}_{11} & \mathbf{J}_{12} \\ \mathbf{J}_{21} & \mathbf{J}_{22} \end{bmatrix} \cdot \begin{bmatrix} \Delta \mathbf{y} \\ \Delta \mathbf{x} \end{bmatrix} = - \begin{bmatrix} f_{\text{seq}}(\mathbf{y}, \mathbf{x}) \\ f_{\text{sim}}(\mathbf{y}, \mathbf{x}) \end{bmatrix}$$

$$\Delta \mathbf{y} = \mathbf{J}_{11}^{-1} \cdot \left(f_{\text{seq}}(\mathbf{y}, \mathbf{x}) + \mathbf{J}_{12} \Delta \mathbf{x} \right)$$

$$\left(\mathbf{J}_{22} - \mathbf{J}_{21} \cdot \mathbf{J}_{11}^{-1} \cdot \mathbf{J}_{12} \right) \cdot \Delta \mathbf{x} = -f_{\text{sim}}(\mathbf{y}, \mathbf{x}) + \mathbf{J}_{21} \cdot \mathbf{J}_{11}^{-1} \cdot f_{\text{seq}}(\mathbf{y}, \mathbf{x})$$

$$\mathbf{J}_{22}^* \cdot \Delta \mathbf{x} = -f_{\text{sim}}(\mathbf{y}, \mathbf{x})$$



0

1. Calculate y_{n+1} from x_n (sequential equations)

2. Update \mathbf{J} with y_{n+1} and x_n

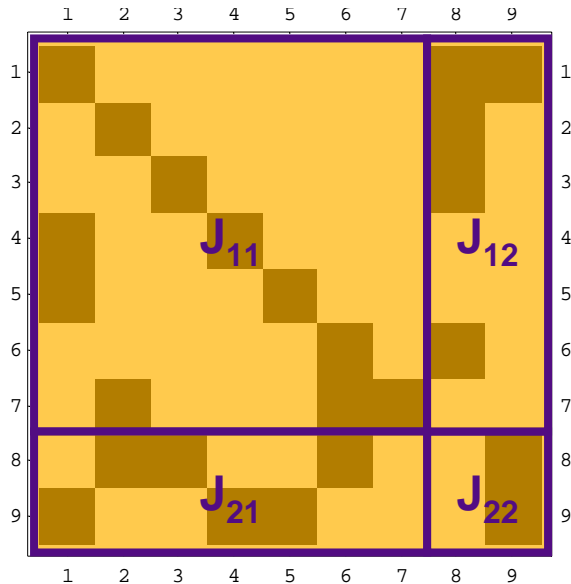
3. Calculate Schur complement

$$\mathbf{J}_{22}^* = \mathbf{J}_{22} - \mathbf{J}_{21} \cdot \mathbf{J}_{11}^{-1} \cdot \mathbf{J}_{12}$$

4. Solve

$$\mathbf{J}_{22}^* \cdot \Delta \mathbf{x} = -f_{\text{sim}}(\mathbf{y}, \mathbf{x})$$

Methods to Calculate Schur Complement



Schur Complement

$$\tilde{\mathbf{J}}_{22} = \mathbf{J}_{22} - \mathbf{J}_{21} \cdot \mathbf{J}_{11}^{-1} \cdot \mathbf{J}_{12}$$

1) Symbolic Schur complement

- extremely complex expressions, impossible

2) Numerical calculation

- matrix inversion and multiplications necessary

3) Semi-symbolic Schur complement with roll-out

- symbolically calculate the Schur complement with references to the evaluated Jacobian matrix
- generate an efficient transformation process by elimination of the submatrix \mathbf{J}_{21}

Example Schur Complement



Structural Matrix

	1	0	0	J[1,4]	J[1,5]
$-J[2,1]$	0	1	0	J[2,4]	J[2,5]
	0	0	1	J[3,4]	J[3,5]
	0	0	0	J[4,4]	J[4,5]
	0	0	0	J[5,4]	J[5,5]

Transformation Process

$$J[2,4] = J[2,1] * J[1,4]$$

$$J[2,5] = J[2,1] * J[1,5]$$

$$J[2,1] = 0$$

$$J[3,4] = J[3,1] * J[1,4]$$

$$J[3,5] = J[3,5] - J[3,1] * J[1,5]$$

$$J[3,1] = 0$$

$$J[4,4] = J[4,4] - J[1,4] * J[4,1]$$

$$J[4,5] = J[4,5] - J[1,5] * J[4,1]$$

$$J[4,4] = J[4,4] - J[2,4] * J[4,2]$$

$$J[4,5] = J[4,5] - J[2,5] * J[4,2]$$

...

Optimization Strategies



- **Identification of Sequential Equations (BLT Transf.)**

Automated recognition of sequential equations from general DAE systems to reduce the number of the simult. equations

- **Common Subexpression Elimination (CSE)**

CSE on DAE level to reduce redundancy within the function evaluation and improve data locality

- **Loop Invariant Expressions / Preloading**

Preevaluation of constant and parametric expressions within the initialization phase, recognition of parametric subexpressions

- **Copy / Constant Propagation**

Reduce memory accesses by propagation of copied or constant values into subsequent expressions

Optimized DAE System



Example - Optimized DAE System

$$vd[t] = VIN - \frac{RS I\$AC[t]}{AREA} - V\$OUT[t]$$

$$SeqExpr2[t] = -\frac{V\$OUT[t]}{R0}$$

$$SeqExpr3[t] = \frac{V\$OUT[t]}{RL}$$

$$id[t] = -AREA \left(e^{-\frac{0.00333167 q (BV+vd[t])}{k}} IBV - \left(-1 + e^{\frac{0.00181069 q vd[t]}{k}} \right) IS \right) + GMIN vd[t]$$

$$cj[t] = AREA CJO \text{ If } \left[vd[t] < 0.25, \frac{1}{(1.-2. vd[t])^{0.333}}, 2.51926 (0.3335 + 0.666 vd[t]) \right]$$

$$SeqExpr1[t] = -C0 V\$OUT'[t]$$

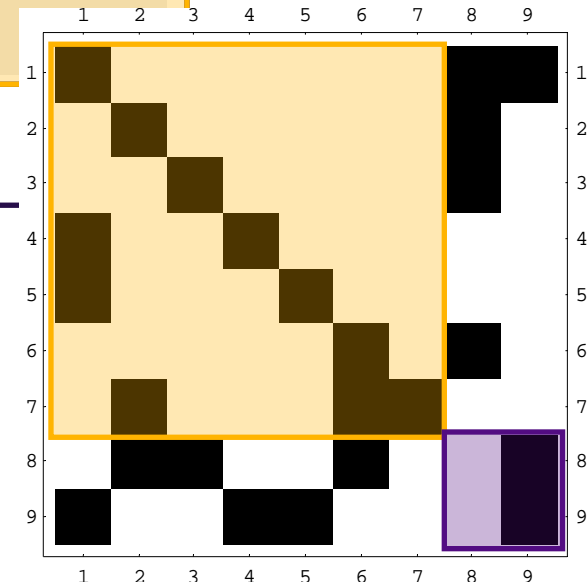
$$I\$VGND[t] = -SeqExpr1[t] - SeqExpr2[t]$$

$$-I\$AC[t] - SeqExpr1[t] - SeqExpr2[t] + SeqExpr3[t] = 0$$

$$I\$AC[t] = id[t] + 1.15 \times 10^{-8} id'[t] + cj[t] vd'[t]$$

Sequential

Simultaneous



- Dimension reduced from 8x8 to 2x2
- No redundancy within DAEs (3 common subexpressions eliminated)



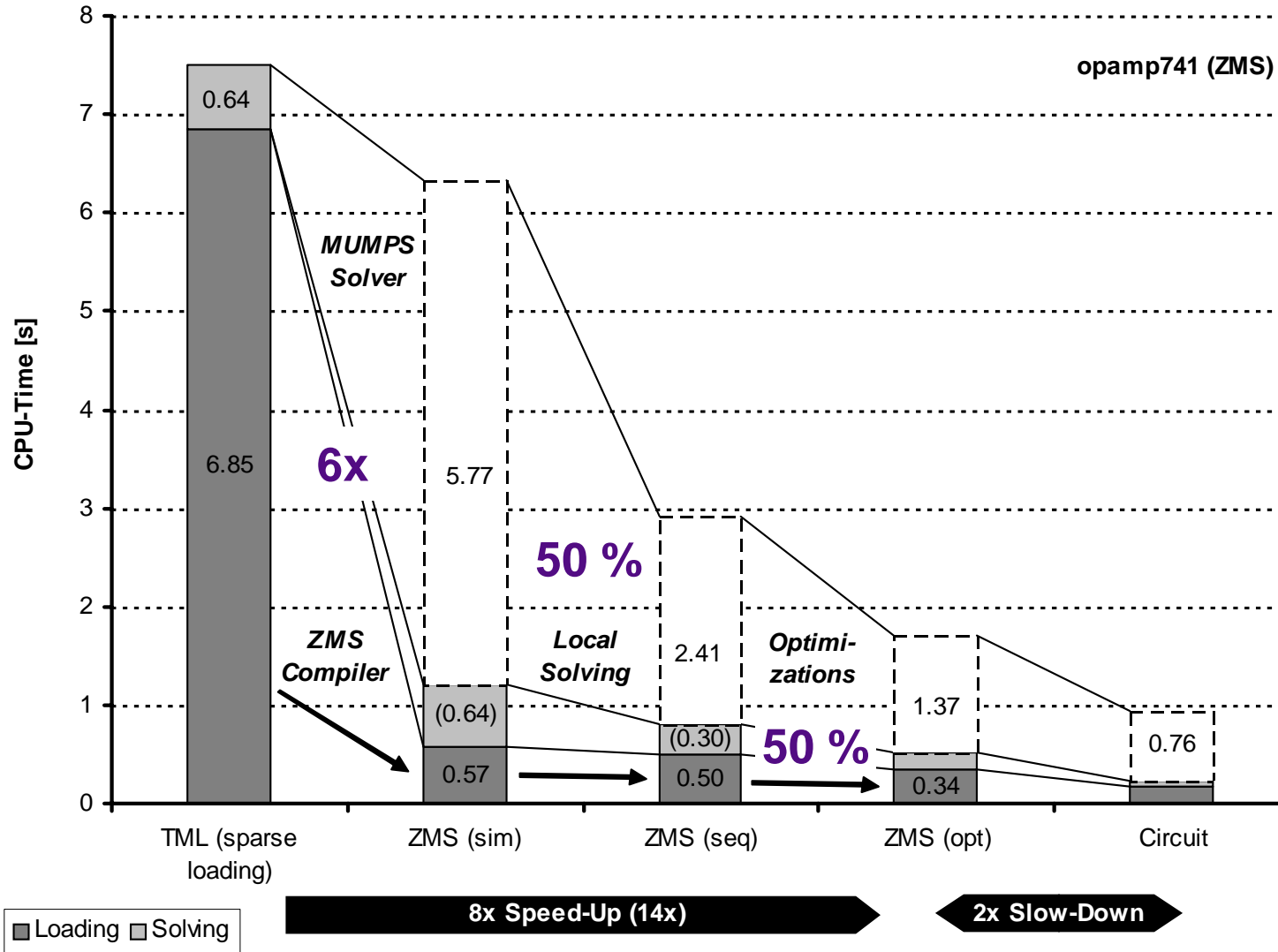
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Summary

Performance Improvement





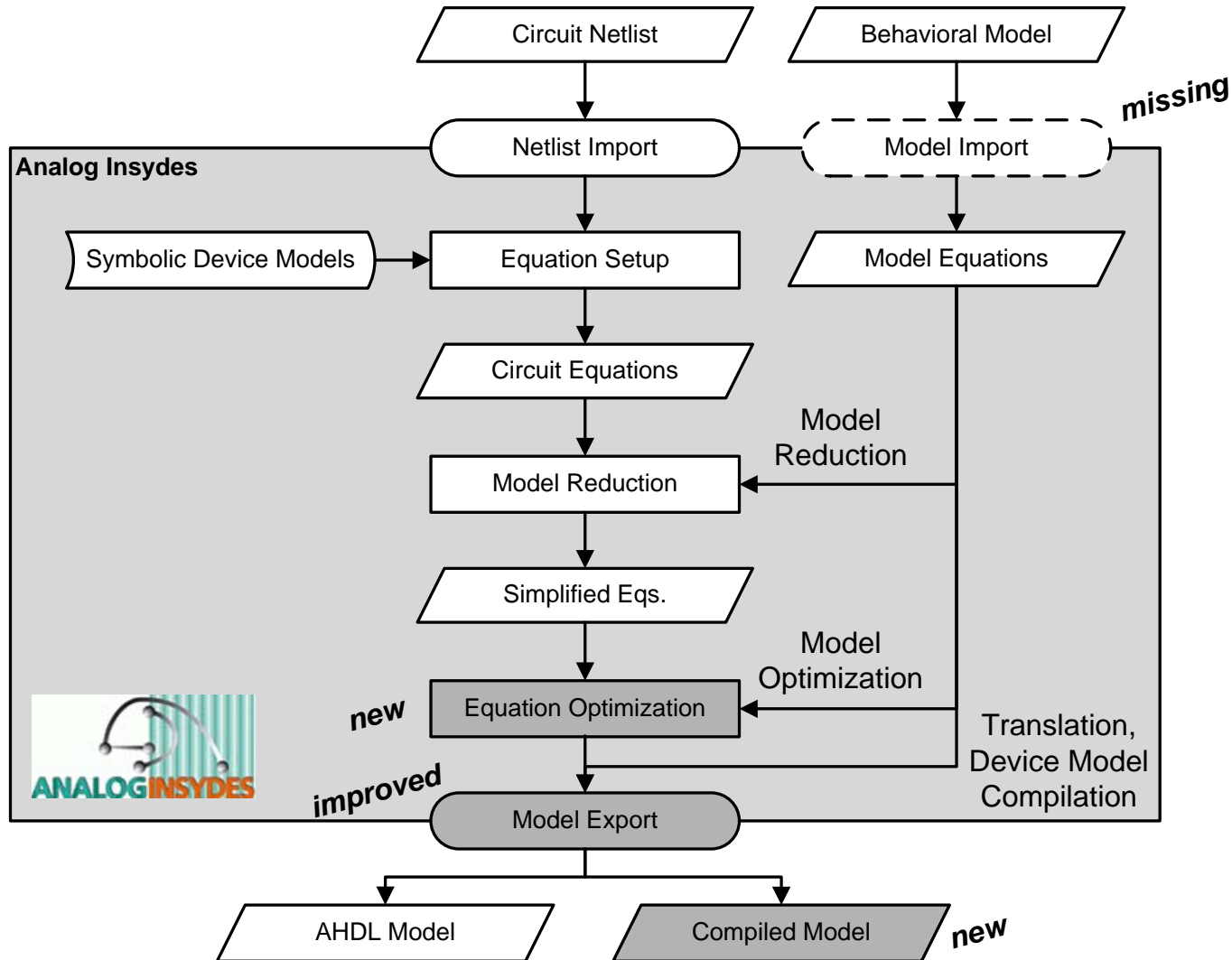
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Outlook





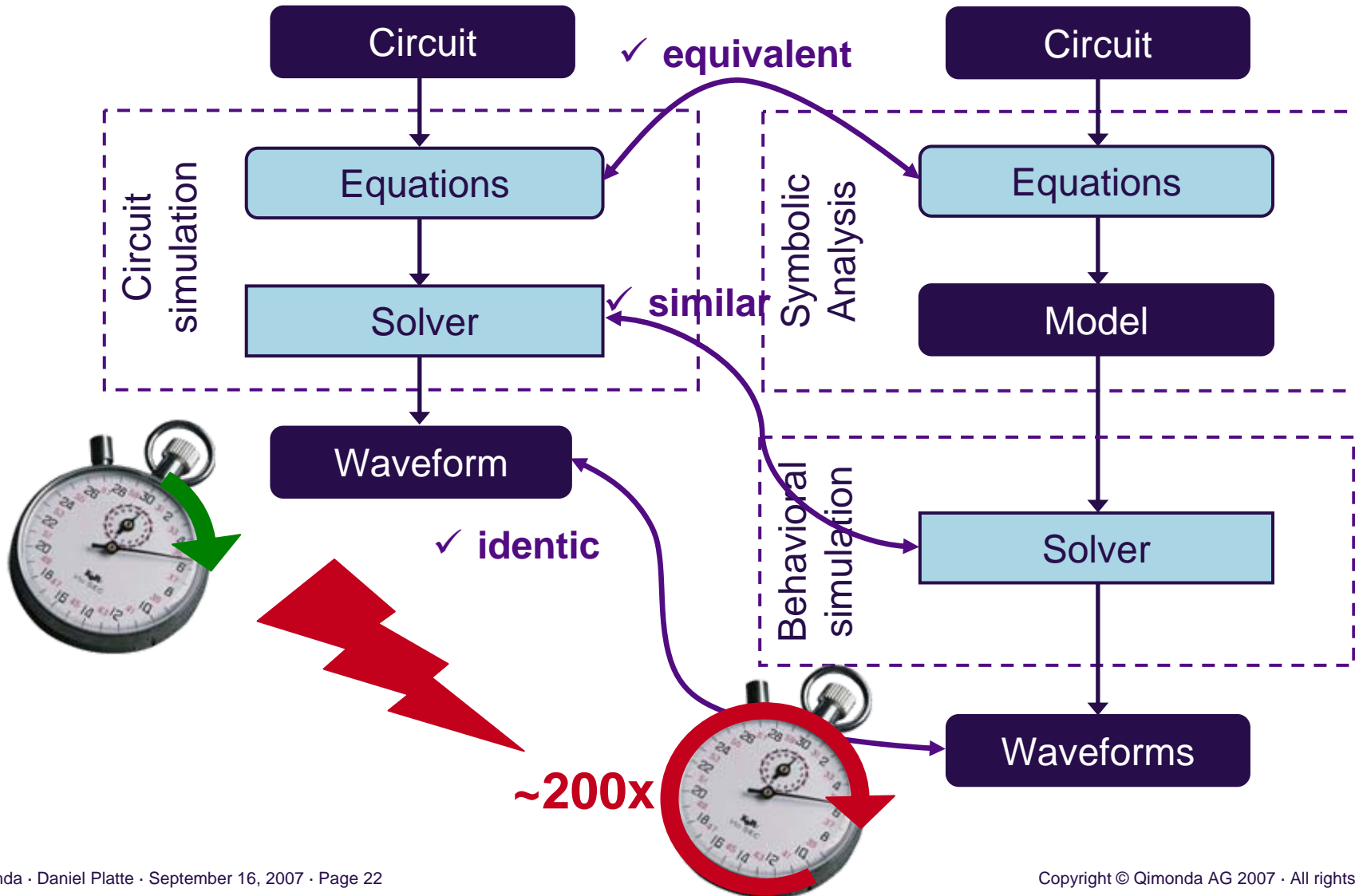
- Model generation based on a **symbolic analysis flow** using the toolbox Analog Insydes presented
- Such complex and high-dimensional AHDL-based models have **insufficient simulation performance**
- **New model compiler** for the Titan simulator developed
- **Efficient processing** of the generated models due to
 - Local solving with semi-symbolic Schur complement
 - Optimization techniques on DAE level
 - Close interaction between model and simulator
- **Performance improvement** by a factor of at least 8

Thank you

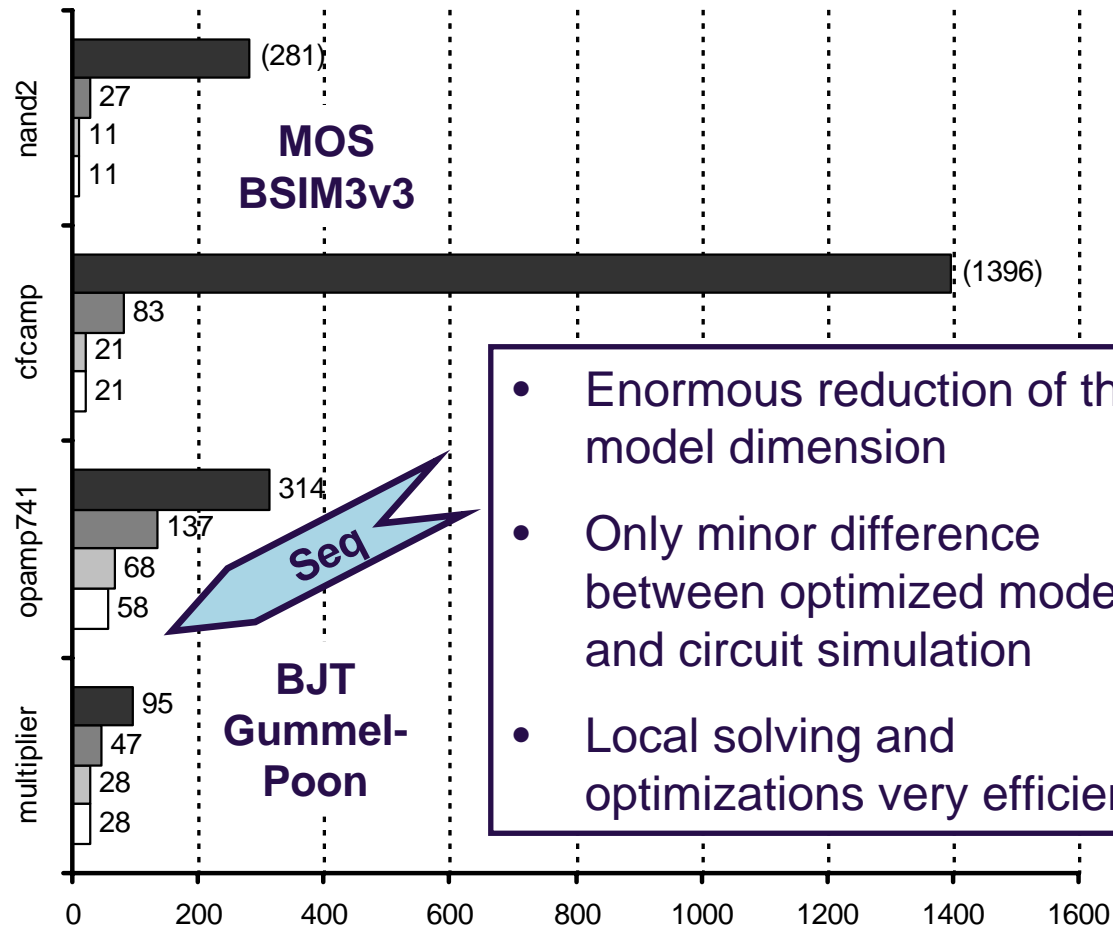
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Performance Problem



Dimension of the Linearized Systems



**MOS
BSIM3v3**

**BJT
Gummel-
Poon**

- Enormous reduction of the model dimension
- Only minor difference between optimized model and circuit simulation
- Local solving and optimizations very efficient

Sim: All model equations simultaneous

Seq: Sequential equations defined within device models

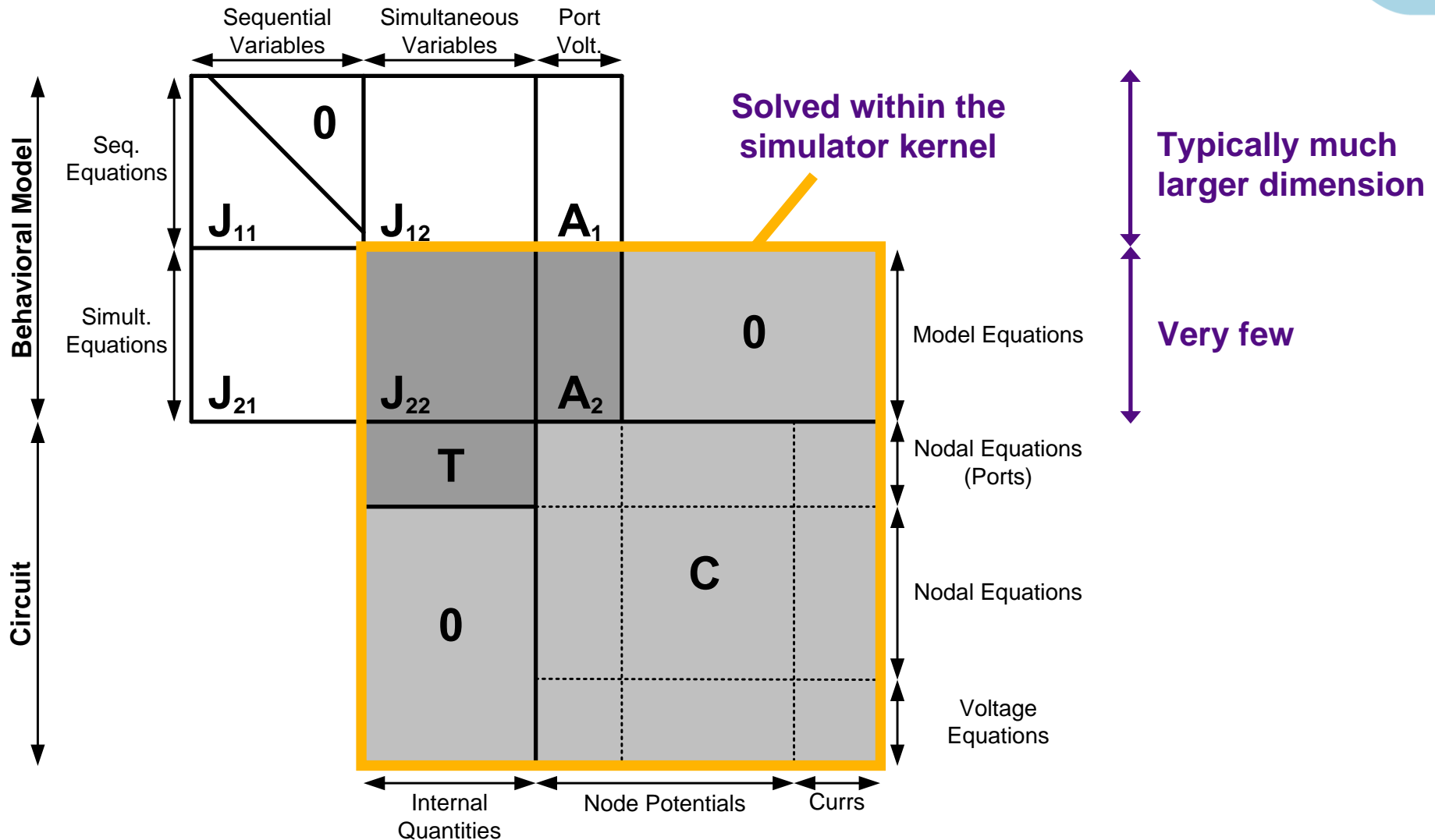
Blt: With identification of sequential equations

Cir: Dimension of the MNA equations in circuit simulation

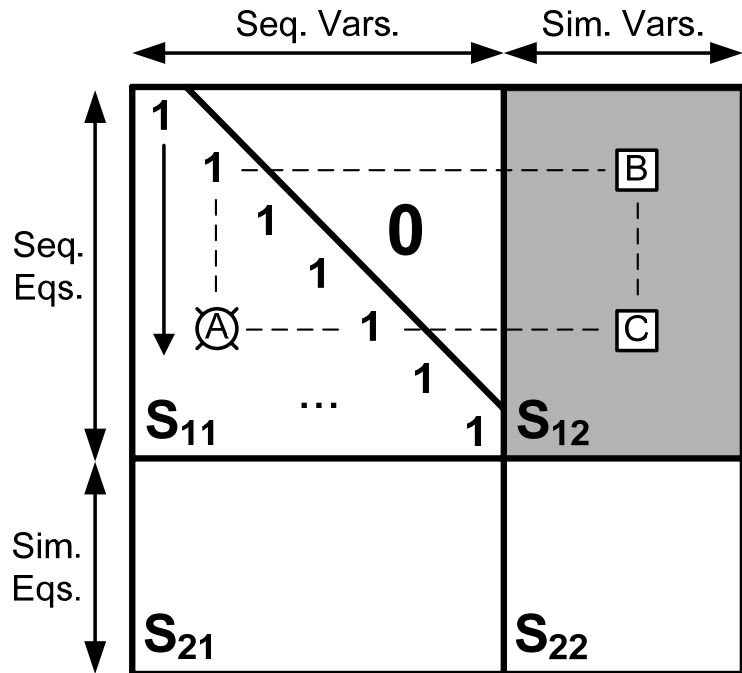
□ cir □ blt □ seq ■ sim

Dim [1]

Jacobian Matrix of Sequential DAEs

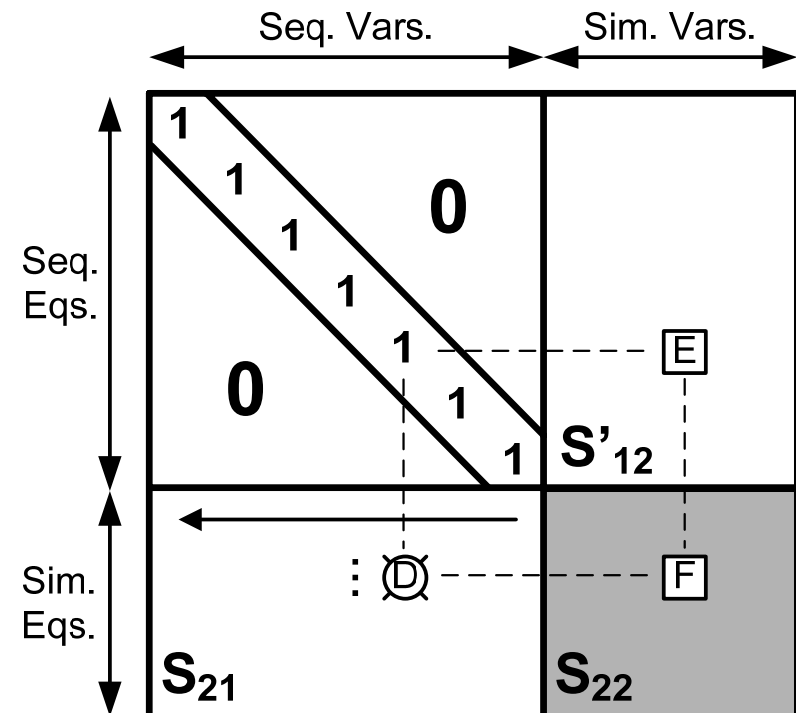


Semi-Symbolic Schur Complement



- 1) Transform S_{11} to unity matrix
 → changes within S_{12} , fill-ins

- 2) Eliminate S_{21}
 → update of S_{22} , fill-ins





DAE System with Sequential Structure

A system of differential algebraic equations (DAE)

$$\mathbf{f}(\mathbf{y}, \mathbf{y}', \mathbf{x}, \mathbf{x}', t) = 0$$

with

$$\mathbf{y} \in \mathbb{R}^m$$

sequential variables

$$\mathbf{x} \in \mathbb{R}^n$$

simultaneous variables

$$\mathbf{f} = \left\{ \mathbf{f}_{\text{seq}}, \mathbf{f}_{\text{sim}} \right\}$$

$$\mathbf{f}_{\text{seq}} : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \longrightarrow \mathbb{R}^m$$

sequential equations

$$\mathbf{f}_{\text{sim}} : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \longrightarrow \mathbb{R}^n$$

simultaneous equations

is of sequential structure, if

$$\mathbf{y}_i = \mathbf{f}_{\text{seq},i}(\mathbf{y}_1 \cdots \mathbf{y}_{i-1}, \mathbf{y}'_1 \cdots \mathbf{y}'_{i-1}, \mathbf{x}, \mathbf{x}', t) \quad \text{for} \quad \mathbf{i} = 1 \dots \mathbf{n}$$



Modeling DAE Systems with Seq. Structure



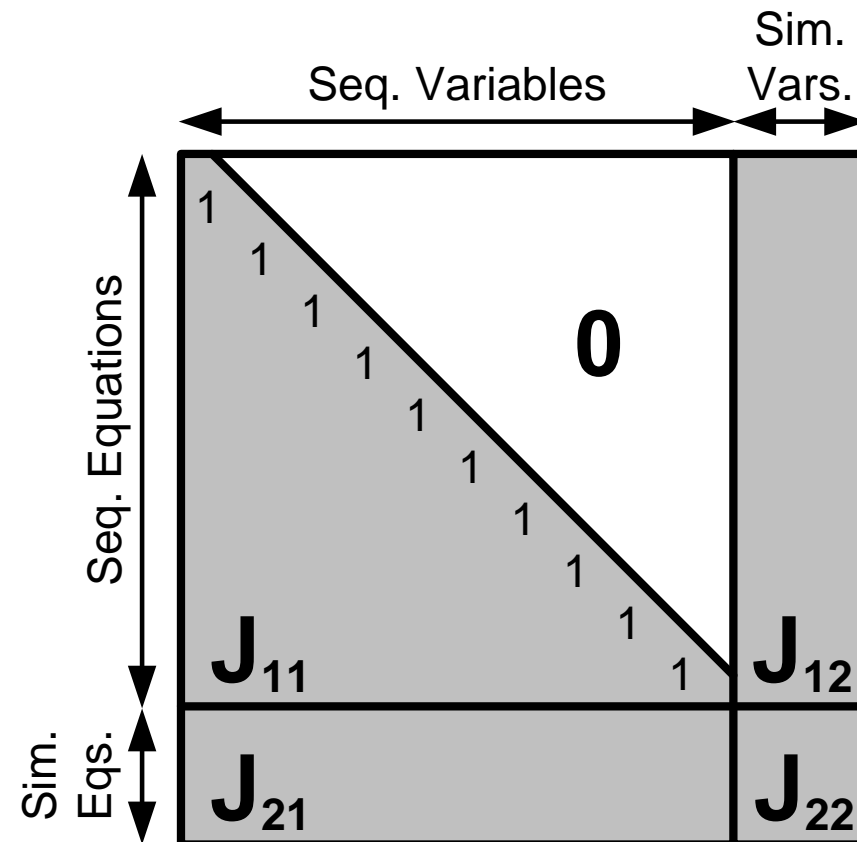
	Verilog-A
Sequential Equations	<pre>// Analog procedural // assignment real vd; vd = -RS*X(I_A))/AREA- X(V_K)+V(A);</pre>
Simultaneous Equations	<pre>// (Indirect) branch // contribution DAEVar V_G; X(V_G) <+ X(V_G)+id-0.115E- 7*X(id_d1)- X(I_A)+cj*X(vd_d1);</pre>

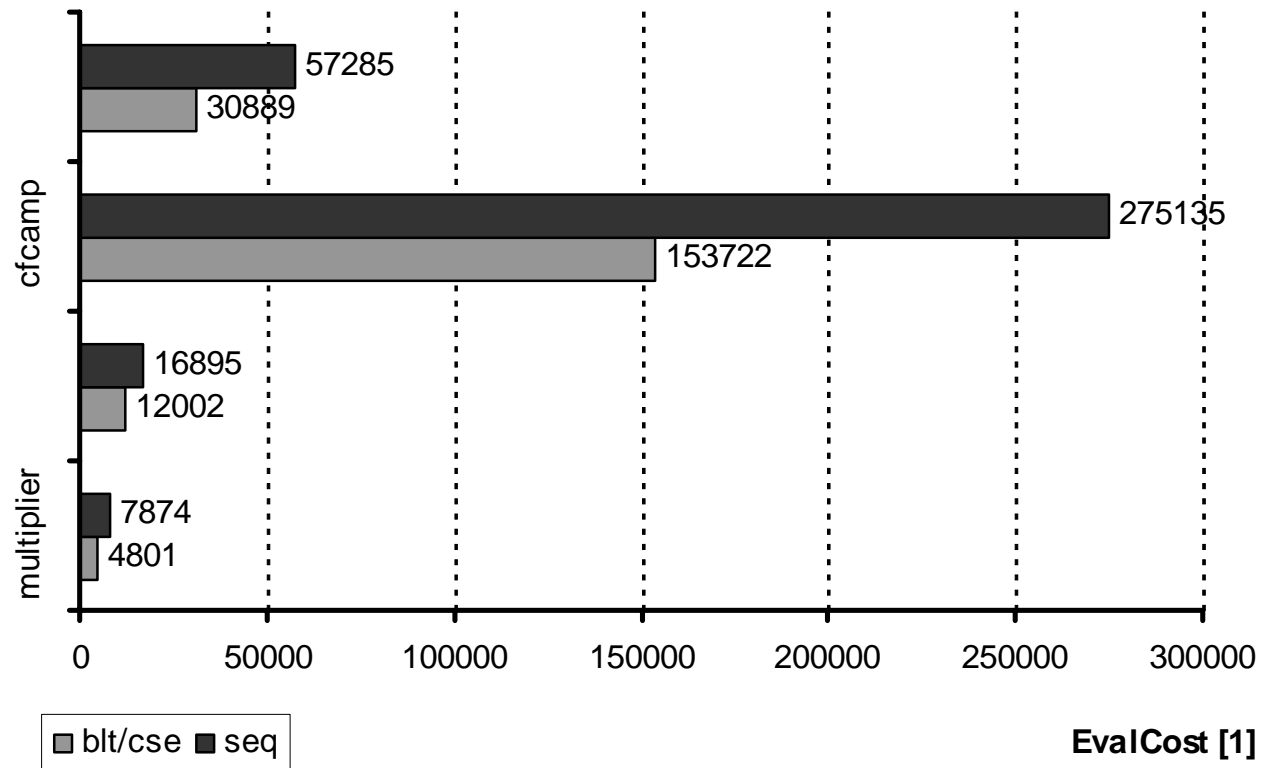
Modeling DAE Systems with Seq. Structure



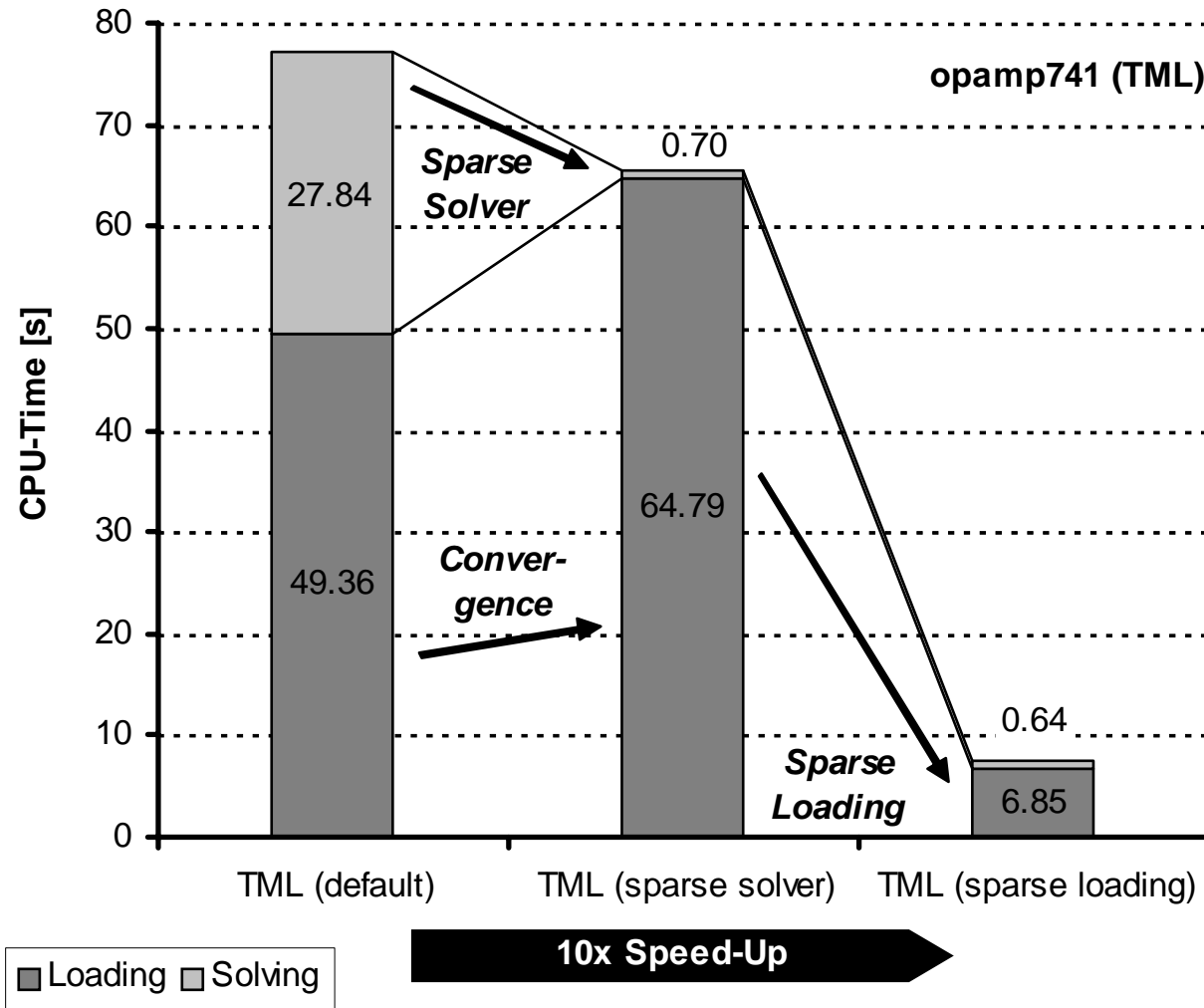
	Verilog-A	VHDL-AMS
Sequential Equations	<pre>// Analog procedural // assignment real vd; vd = -RS*X(I_A)/AREA- X(V_K)+V(A);</pre> 	<pre>-- Simultaneous -- procedural statement vd:=-RS*I_A/AREA-V_K+A; -- or: Function function vd (I_A,...) is ... </pre>
Simultaneous Equations	<pre>// (Indirect) branch // contribution DAEVar V_G; X(V_G) <+ X(V_G)+id- 0.115E-7*X(id_d1)- X(I_A)+cj*X(vd_d1);</pre>	<pre>-- Simple simultaneous -- statement QUANTITY V_G : Voltage; I_A-ID-0.115E-7*ID_D1- CJ*VD_D1 == 0.0 Tolerance "Current";</pre> 

Block Lower-Triangular Matrix





Speed-Up for TML Models



Sequential Equations (Definition)



$$\hookrightarrow \underline{x}_1 = \underline{G}_1(\underline{y})$$

$$\hookrightarrow \underline{x}_2 = \underline{G}_2(\underline{y}, \underline{x}_1)$$

$$\hookrightarrow \underline{x}_3 = \underline{G}_3(\underline{y}, \underline{x}_1, \underline{x}_2)$$

...

$$\hookrightarrow \underline{x}_n = \underline{G}_n(\underline{y}, \underline{x}_1, \dots, \underline{x}_{n-1})$$

n sequential equations

n sequential variables \underline{x}

sequential variables as function of simultaneous variables \underline{y}

$$\underline{x} = \underline{x}(\underline{y})$$

$$\underline{F}(\underline{x}, \underline{y}) = 0$$

m simultaneous equations

m simultaneous variables \underline{y}

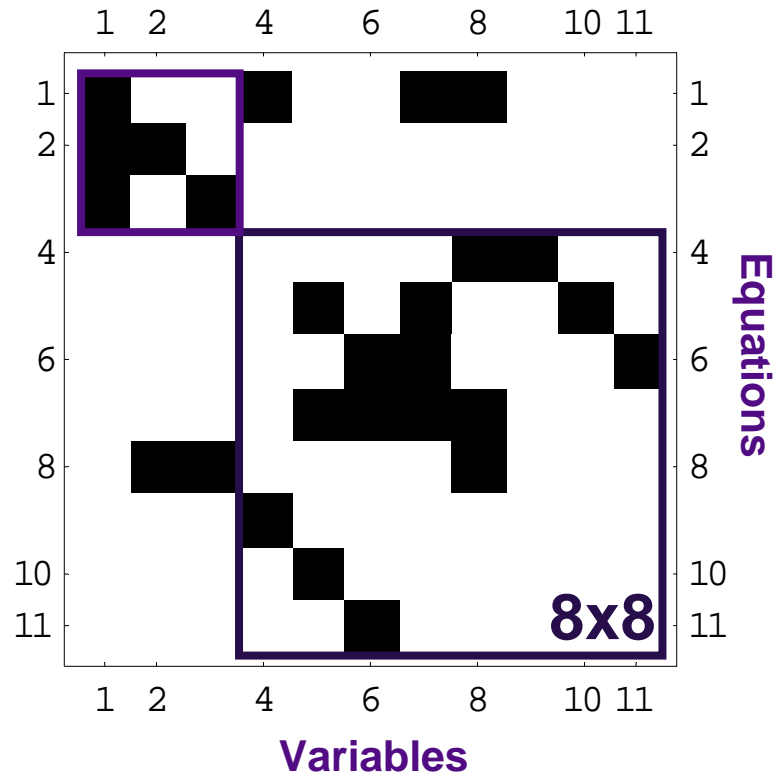
$$\rightarrow \tilde{\underline{F}}(\underline{y}) = \underline{F}(\underline{x}(\underline{y}), \underline{y}) = 0$$

Advantages:

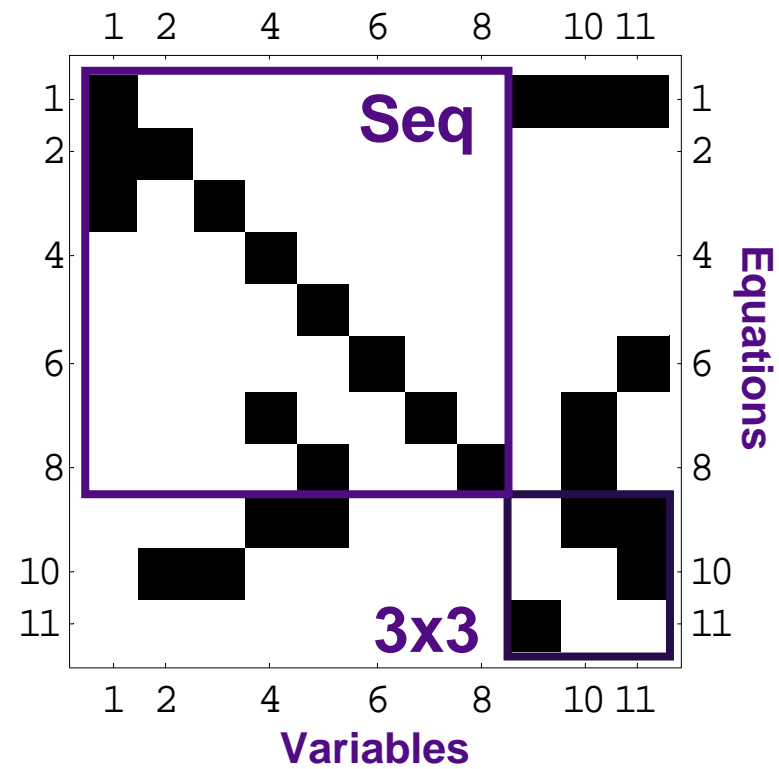
- Reduced dimension of Jacobian matrix
- Iterative solution for less variables

$$(n + m) \times (n + m) \rightarrow m \times m$$

Recognition of Sequential Equations



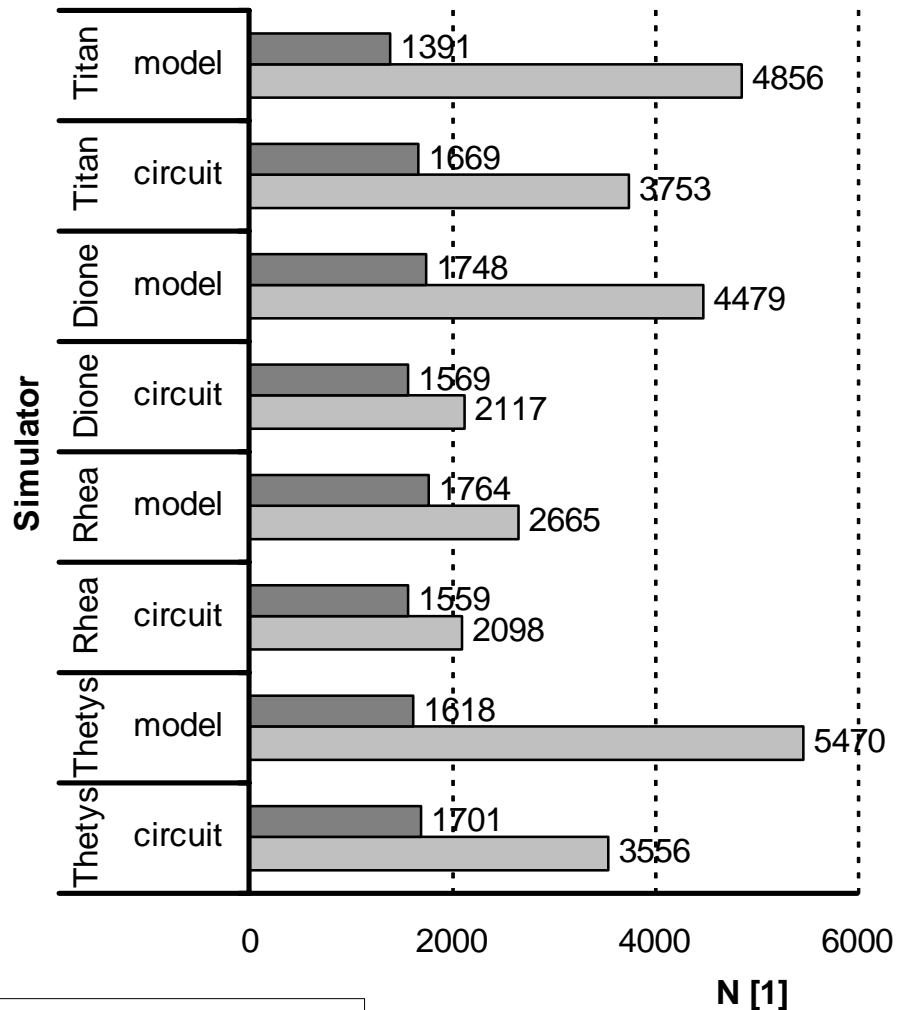
Identification



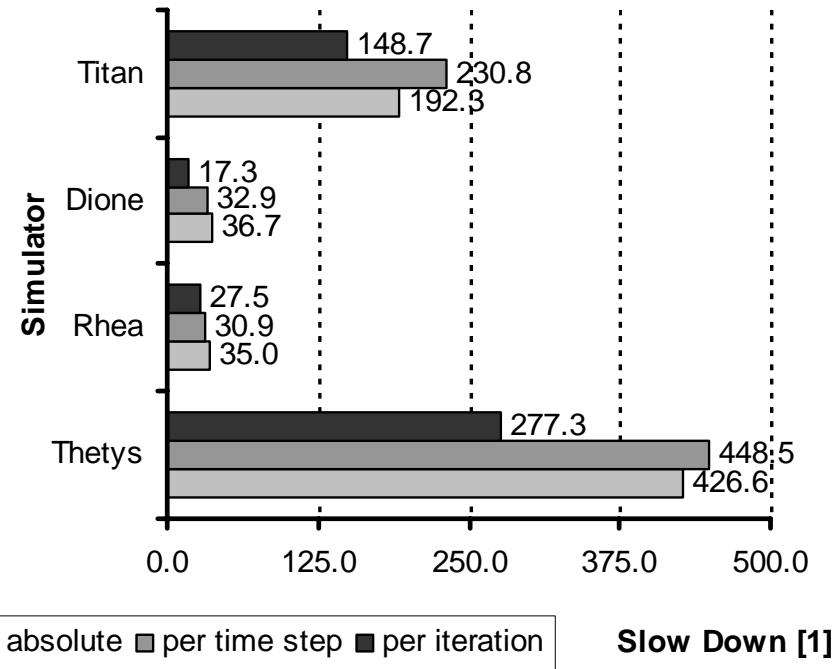
Simulator Comparison for opamp741



opamp741



opamp741



iterations time steps