A Strategy for Parallel Simulation of Declarative Object-Oriented Models of Generalized Physical Networks

Francesco Casella

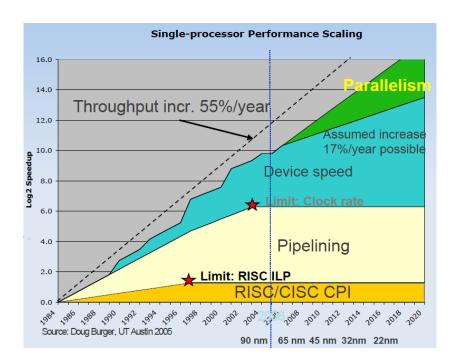
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Introduction and motivation

- Moore's law depends on multicore architectures since 2007
- Declarative O-O modelling is now a mature & established field (1997: Modelica 1.0)

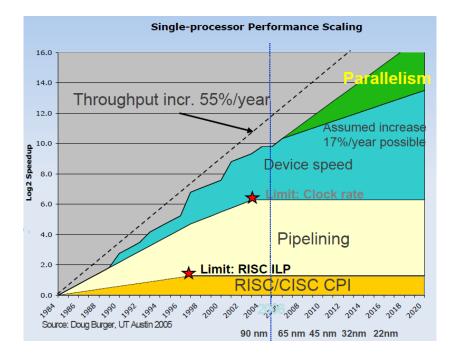


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however

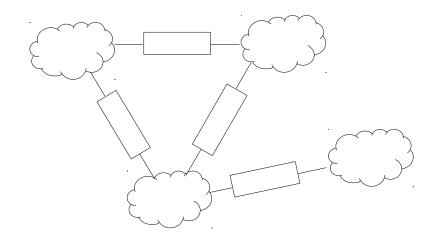
 Simulation code generated by state-of-the-art Modelica tools as of 2012 is still fully sequential!

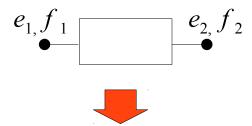


why?

TLM Modelling

- Main idea: physical interactions given by wave propagation phenomena with finite delay
 - pressure / flow waves in hydraulic circuits
 - electromagnetic waves in transmission lines
 - elastic waves in mechanical systems
- Explicitly partion the system with Transmission Line Models

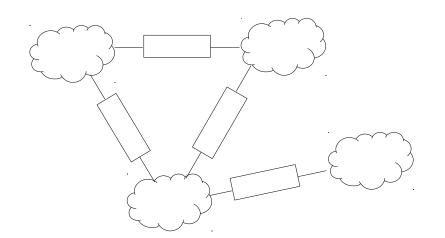


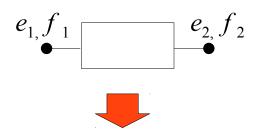


$$\begin{split} e_{1}(t) &= Z f_{1}(t) + e_{2}(t - T_{tl}) + Z f_{2}(t - T_{tl}) \\ e_{2}(t) &= Z f_{2}(t) + e_{1}(t - T_{tl}) + Z f_{1}(t - T_{tl}) \end{split}$$

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Sub-systems can be solved independently (→ in parallel) for T seconds

TLM Modelling

Advantages

Modelling is physically accurate – no approximations

Disadvantages

- Decoupling elements must be inserted manually
- Physical delays T are usually very small
- Maximum step size constrained by T



requires expertise



parallel simulation might be slower than sequential simulation of the coupled model

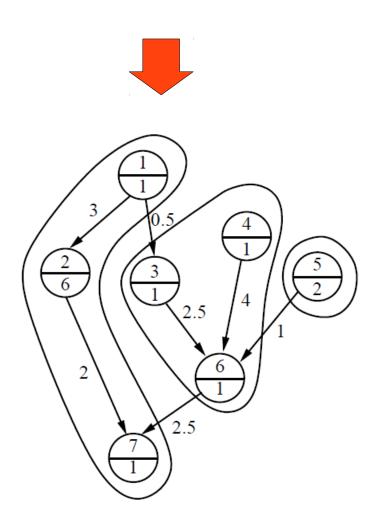
Peter Aronsson's PhD work

$$F(x,\dot{x},v,t)=0$$



$$F(x,\dot{x},v,t)=0 \qquad \dot{x}=f(x,t) \\ v=g(x,t)$$

explicit assignments implicit equations to solve



Peter Aronsson's PhD work

Advantages

- Applies to any equation-based model without manual intervention
- Optimal scheduling possible in principle

Disadvantages

- Accurate estimation of computation and communication delays crucial
- Potentially bad performance if delays not correct
- Task merging and clustering algorithms very involved
- Possibly too complex for large systems



never found way into production tools

Goals of this work

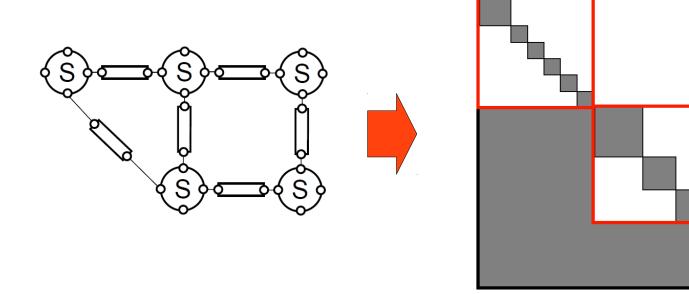


Identify a (large) class of O-O models that have a special structure of the incidence matrix

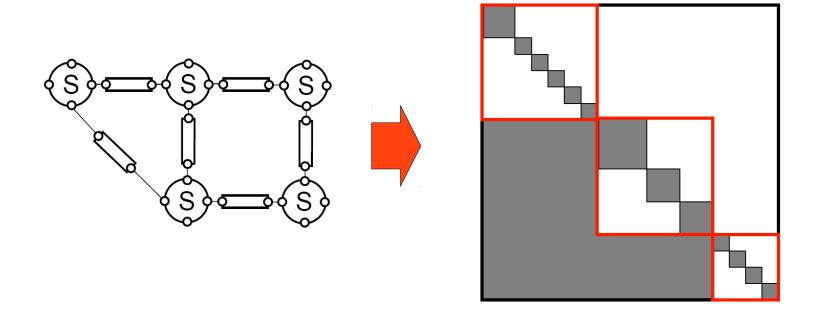


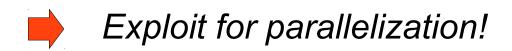
Propose an algorithm which is simple to implement, but provides near optimal performance for that class of models

Main Idea



Main Idea





Proposed algorithm

- 1. Build E-V digraph
- 2. Find a complete matching (possibly using Pantelides / Dummy Derivatives for higher index systems)
- 3. Replace non-matching edges with E→V arc; collapse V-nodes with matching E-nodes (a directed graph is obtained)
- 4. Run Tarjan's algorithm and identify strong components (systems of equations to be solved simultaneously)

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- 5. Collapse each strong component into a single macro-node
- 6. Let i = 1
- 7. Search for all sinks and collect them in set S_i (they correspond to equations that can be solved independently)
- 8. Delete all nodes in set S_i and all associated arcs from the graph
- 9. If there are nodes left, increase *i* by 1 and goto 6.

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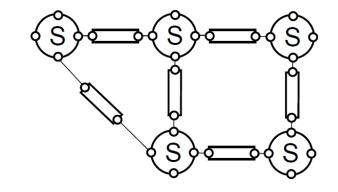


At termination, all equations are collected in the sets S_i Equations in S_i can be solved independently, then equations in S_i , etc.

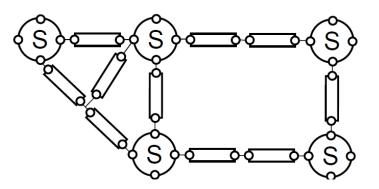
Generalized networks

- Storage components: storage of physical quantities, represented by state variables (known at each time step!)
- Flow components: describe the flow of the quantity, based on the boundary values

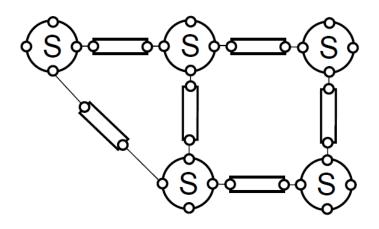
 V-F-V topology: no implicit equations



 V-F-F-V topology: implicit equations (macro-nodes)



Example 1: Thermal Networks



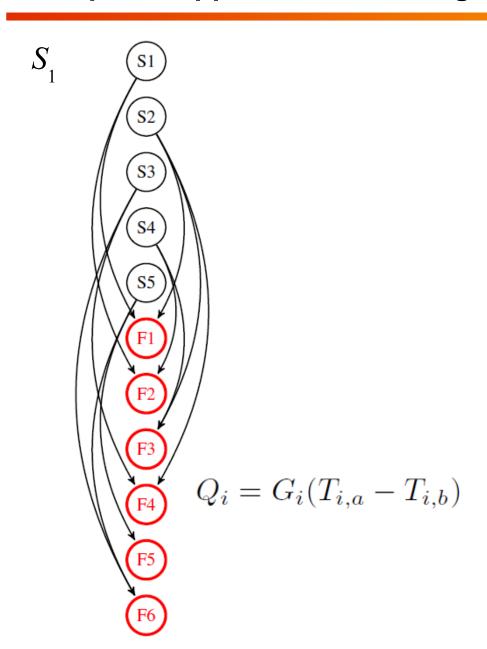
Storage components

$$C(T_i)\frac{dT_i}{dt} = \sum_j Q_{i,j}$$

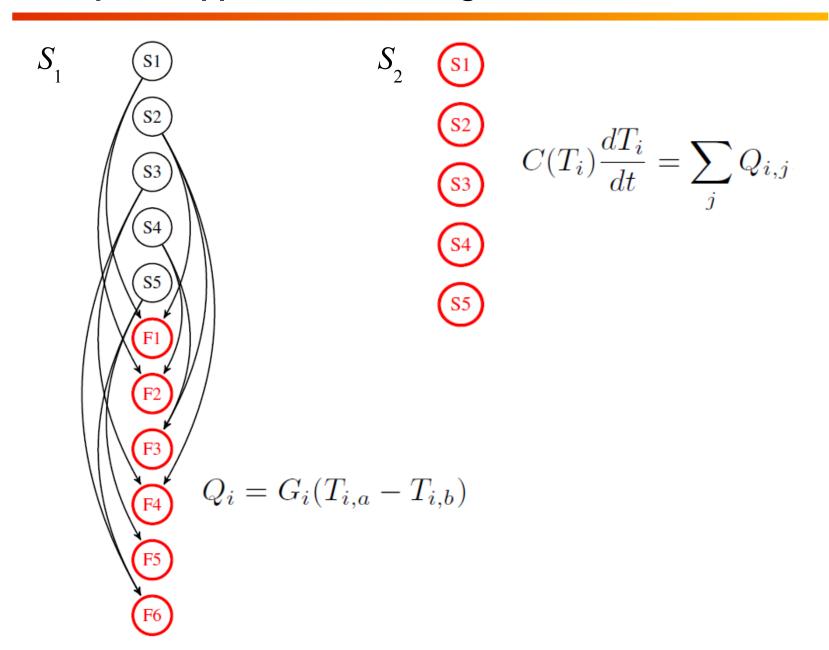
Flow components

$$Q_i = G_i(T_{i,a} - T_{i,b})$$

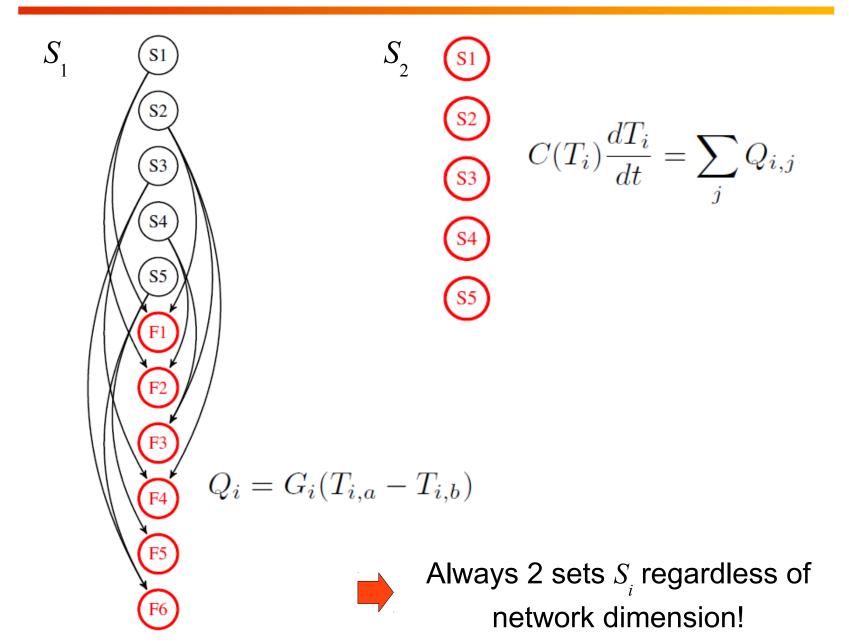
Example 1: Application of the algorithm



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Storage components

$$\begin{bmatrix} e_i & h_i & \rho_i & \frac{\partial \rho_i}{\partial p} & \frac{\partial \rho_i}{\partial T} & \frac{\partial e_i}{\partial p} & \frac{\partial e_i}{\partial T} \end{bmatrix} = f(p_i, T_i)$$

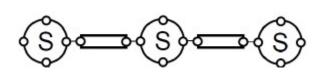
$$M_i = \rho_i V_i$$

$$\frac{dM_i}{dt} = \sum_j w_{i,j}$$

$$\frac{dE_i}{dt} = \sum_j w_{i,j} h_{i,j} + \sum_j Q_{i,j}$$

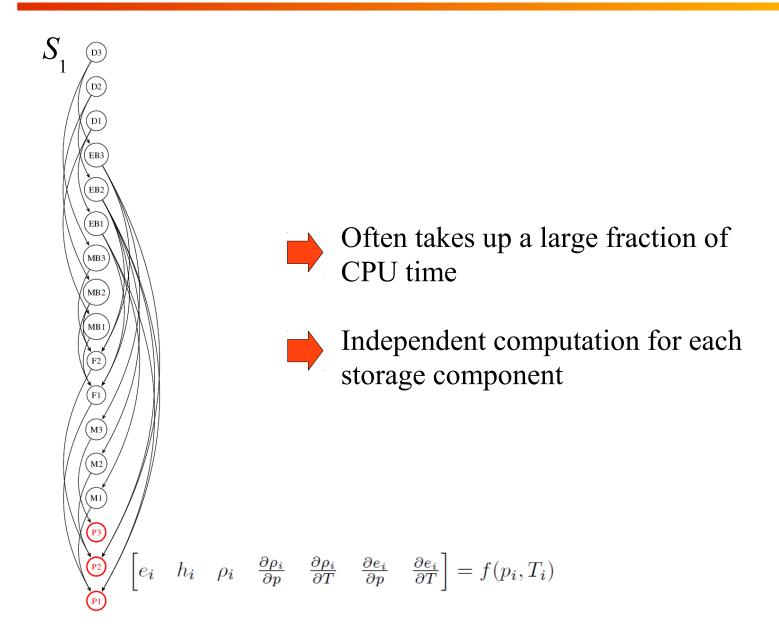
$$\frac{dM_i}{dt} = \frac{\partial \rho_i}{\partial p} \frac{dp_i}{dt} + \frac{\partial \rho_i}{\partial T} \frac{dT_i}{dt}$$

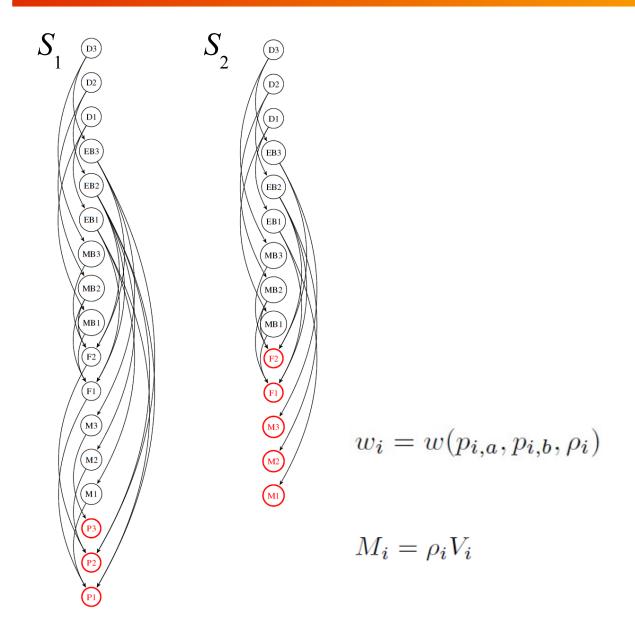
$$\frac{dE_i}{dt} = \left(\frac{\partial e_i}{\partial p} \frac{dp_i}{dt} + \frac{\partial e_i}{\partial T} \frac{dT_i}{dt}\right) M_i + e_i \frac{dM_i}{dt}$$

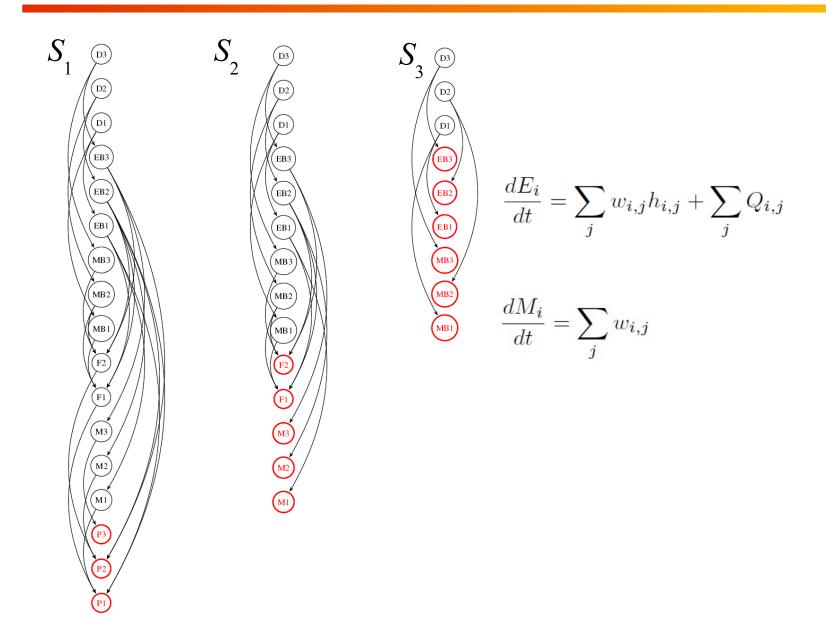


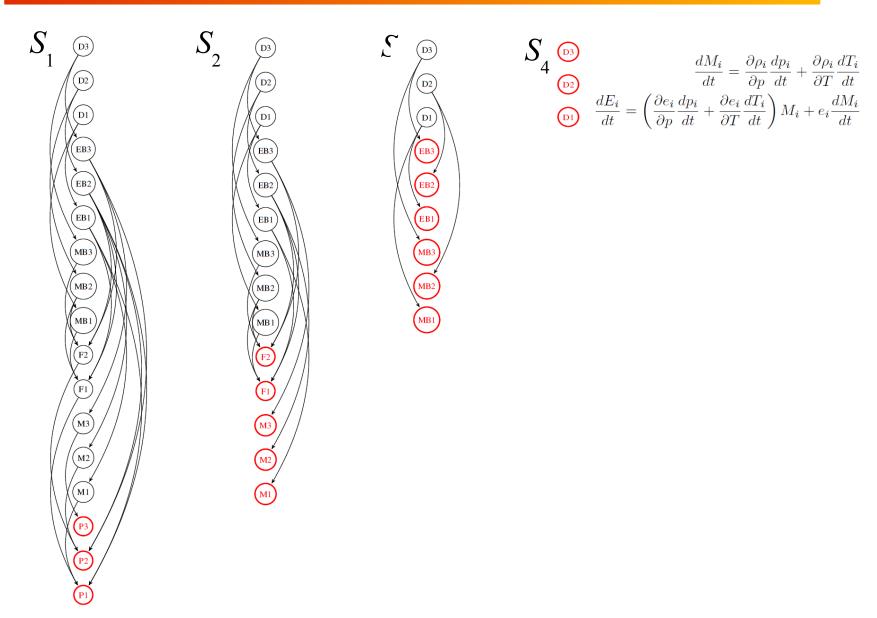
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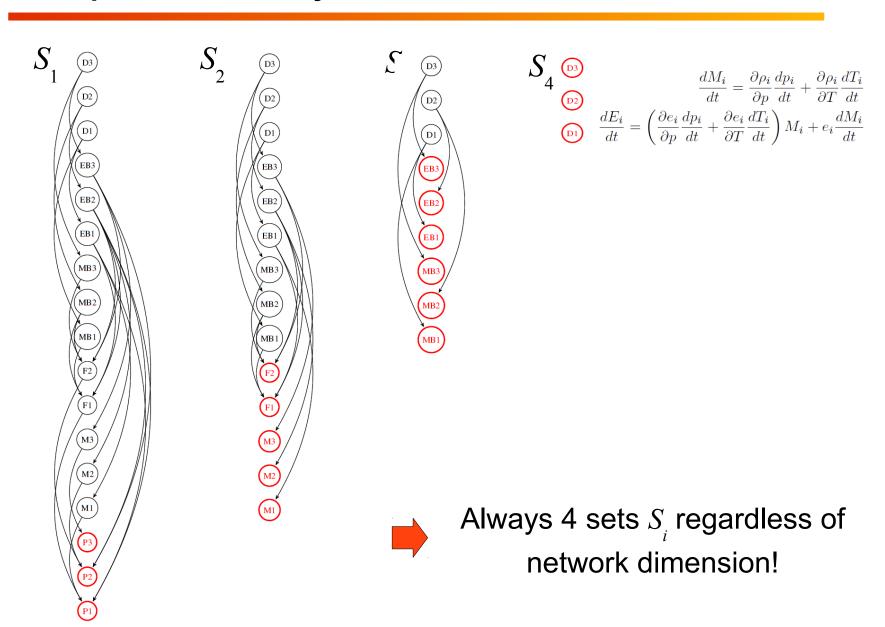
$$Q_i = G_i(T_{i,a} - T_{i,b})$$
$$w_i = w(p_{i,a}, p_{i,b}, \rho_i)$$











Task Scheduling

- Shared memory → negligible communication costs (beware of L1 caching issues!)
- The larger $\#(S_i)$ / N_{cores} is, the less the actual scheduling policy matters

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 - if many more independent tasks than cores, a first-come, first-served scheduling policy will result in all the cores running almost all the time
- Avoid waiting for the slowes one: start slowest tasks earlier
 - minimize chance of N-1 cores waiting for the Nth to complete its task
- Avoid overhead: merge several short tasks in a bigger one
- Only rough estimates of the running time are required

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- Set up threads at the beginning of the simulation, activate them @ each time step

Conclusions

- Exploiting Moore's law from 2008 requires exploiting parallelism
- Strategies for parallel simulation of O-O declarative models exist, not implemented in mainstream tools yet (implementation difficulties, not enough cores to justify the overhead)
- A very simple algorithm has been proposed in this work
- It can provide nearly optimal parallel allocation of resources in generalized network models
- Only very rough estimates of task execution times are needed
- Other parallelization strategies could be implemented together with this one to improve overall performance
 - parallel numerical computation of Jacobians
 - parallel solution of large implicit systems in stiff solvers
 - **–** ...
- Beware of thread switching overhead and L1 caching issues!

Thank you for you kind attention!