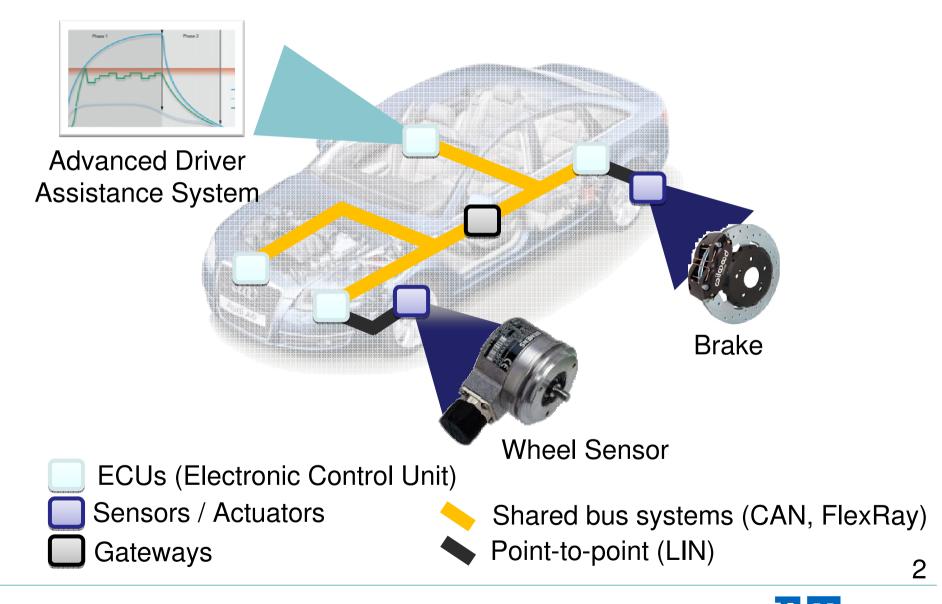
Model-based Design of Distributed Automotive Systems

TiMoBD - Time Analysis and Model-Based Design, from Functional Models to Distributed Deployments **Martin Lukasiewycz**, Samarjit Chakraborty, Michael Glass, Juergen Teich *martin.lukasiewycz@tum-create.edu.sg*



Heterogeneous Network – Motivation

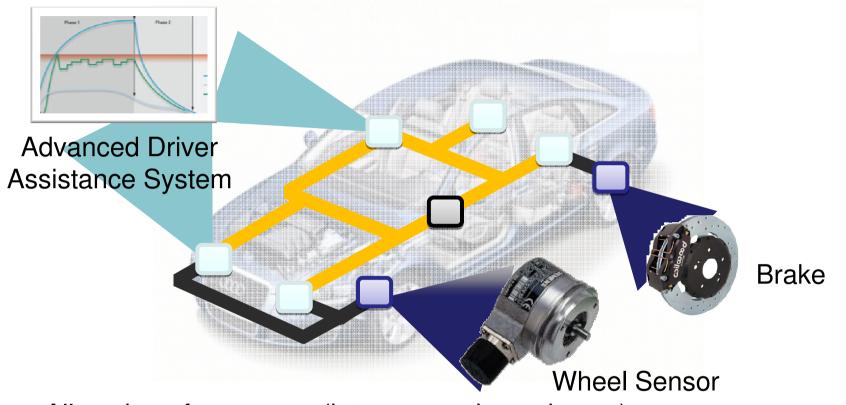


Outline

- Exploration Model
- Optimization of Constrained Combinatorial Problems
- Model Encoding
- Timing Aspects
- Case Studies



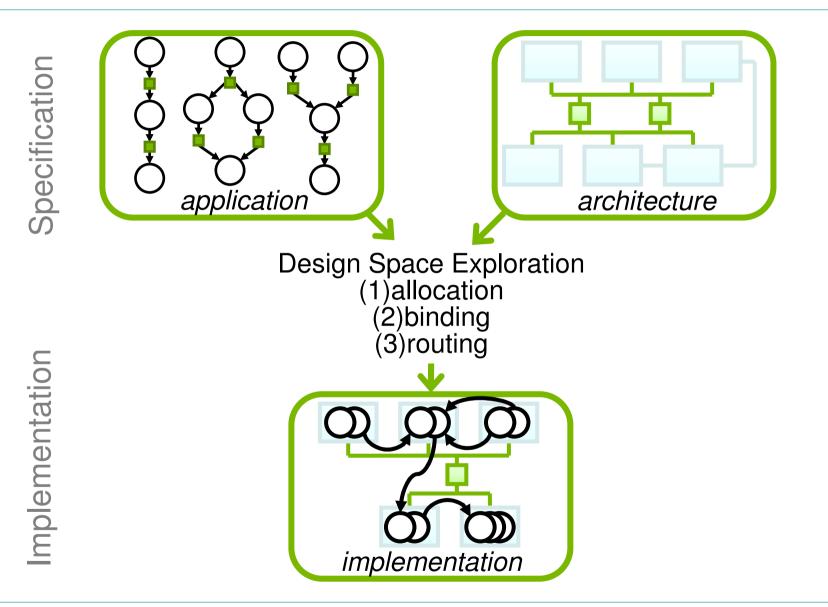
Motivation – Design Space Exploration



- Allocation of resources (interconnection to buses)
- Distribution of tasks
- Routing of messages
- Parameters (priorities, scheduling policies)

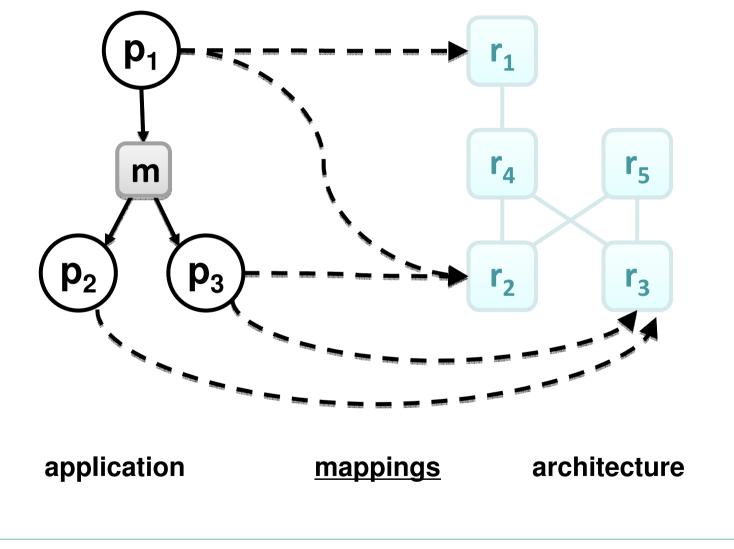


Model (Y-Chart Approach)



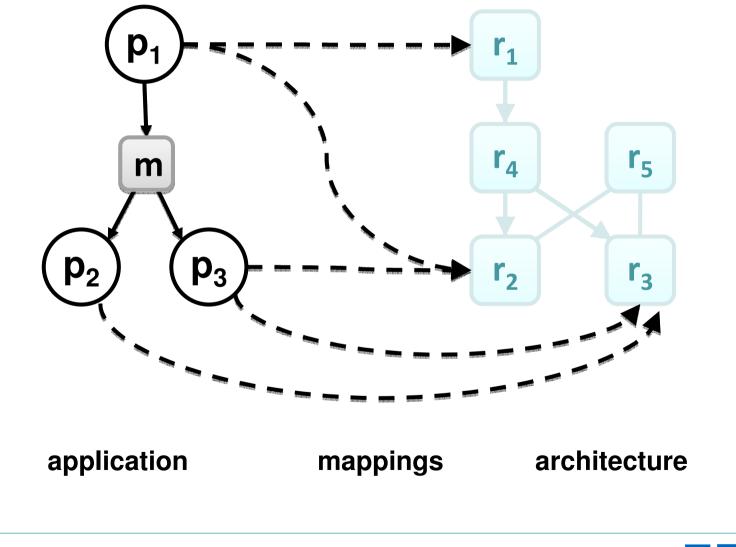
Centre for Electromobility Singapore

Specification



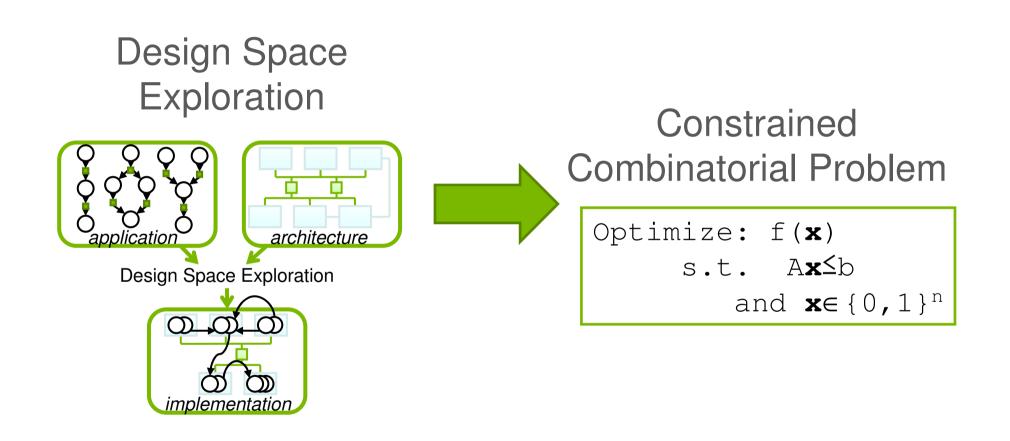


Implementation: Allocation, Binding, Routing





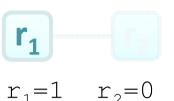
Problem Transformation

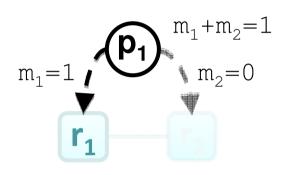




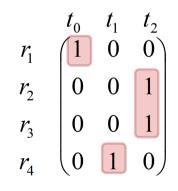
Model Encoding

- Allocation encoding
 - Binary variable for each resource
- Binding encoding
 - Binary variable for each mapping
 - Linear constraints for unique binding and binding to an allocated resource
- Routing encoding
 - Binary variables for each pair of resource and time step
 - Linear constraints fulfill dependencies





 $r_1-m_1 \ge 0$



Constrained Combinatorial Problem

- Definition: Optimize: f(x) s.t. Ax≤b and x∈{0,1}ⁿ
- Objective function \pm is non-linear and multi-objective
- Linear constraints define the feasible solution space
 X_f⊆{0,1}ⁿ

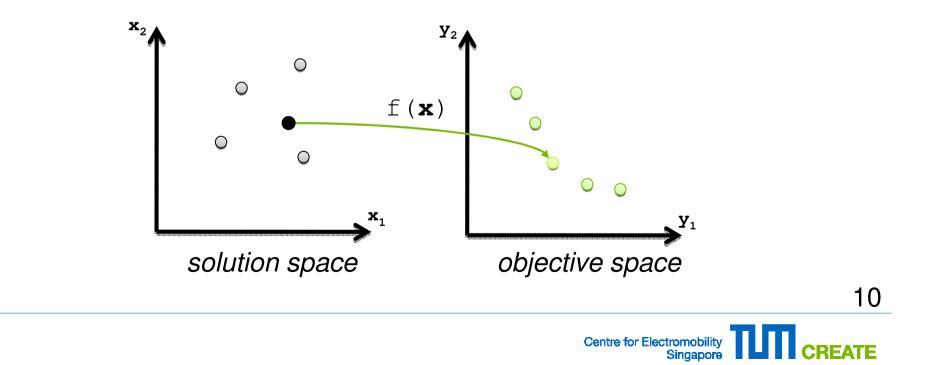
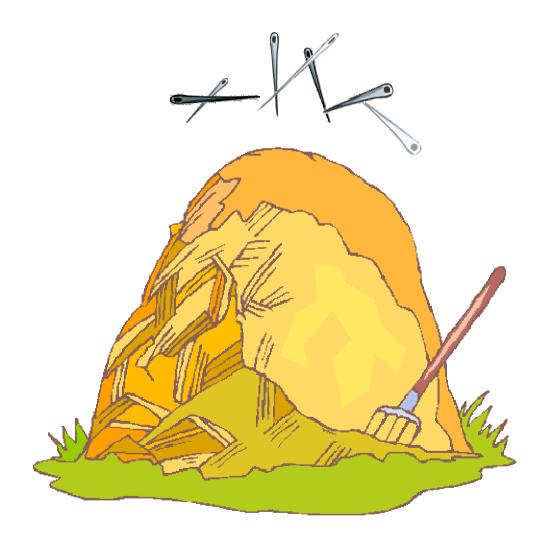


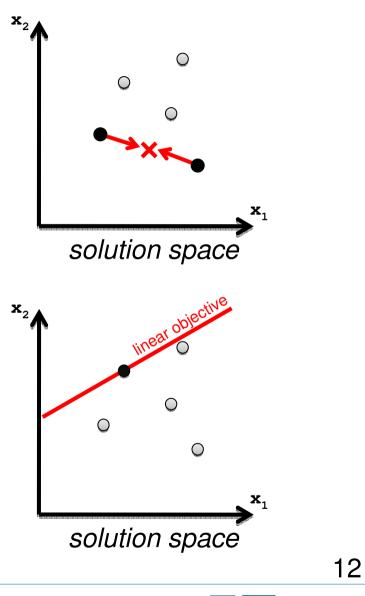
Illustration of Problem Complexity



Common Approaches

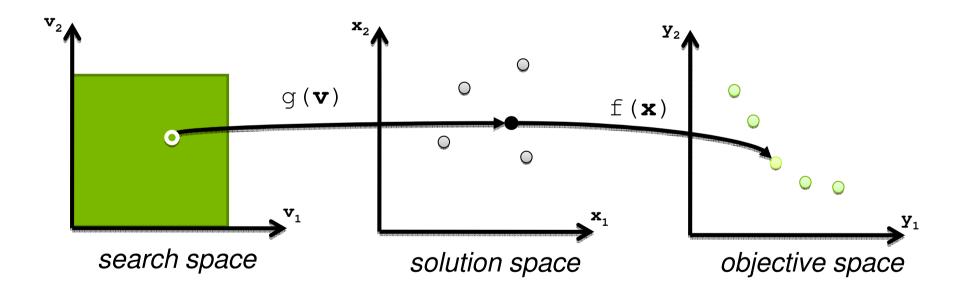
- Evolutionary Algorithms (EA)
 - Suitable for multi-objective optimization problems
 - Restricted-use for constrained optimization problems

- Integer Linear Programming (ILP)
 - Suitable for linearly constrained optimization problems
 - Restricted to single linear objective function



Proposed Approach

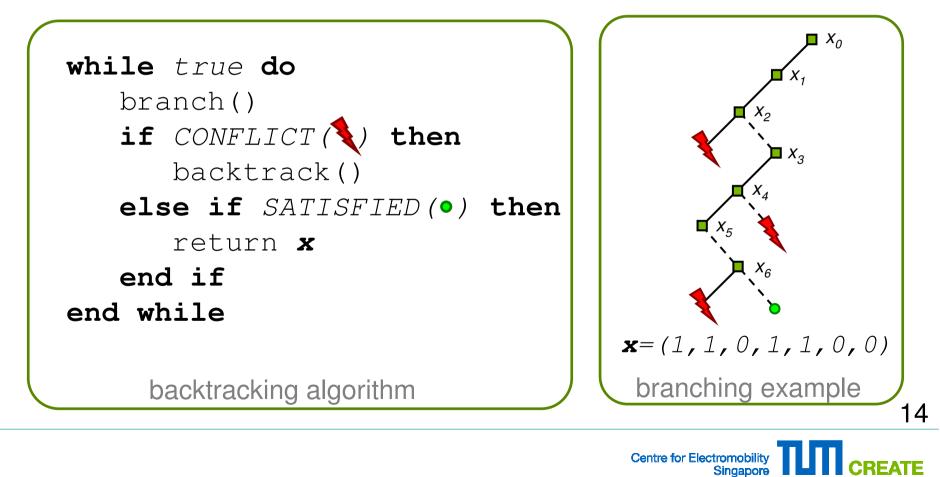
- Decoding Approach
 - Search in the simply constrained search space V
 - Decoding to a feasible solution in the solution space ${\rm X}_{\rm f}$
 - Combination of EA und PB (Pseudo-Boolean) solver



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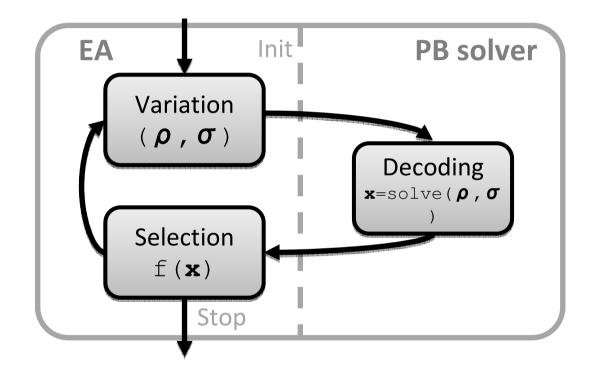
PB Solver

- Backtracking algorithm for ILPs with binary variables
- Branching determined by
 - $\boldsymbol{\rho}$ priority for each variable
 - $\,\sigma\,$ decision phase for each variable



Proposed Approach – Flow

• The search space $V=R^n \times \{0, 1\}^n$ is defined as the set of all branching strategies for the PB solver

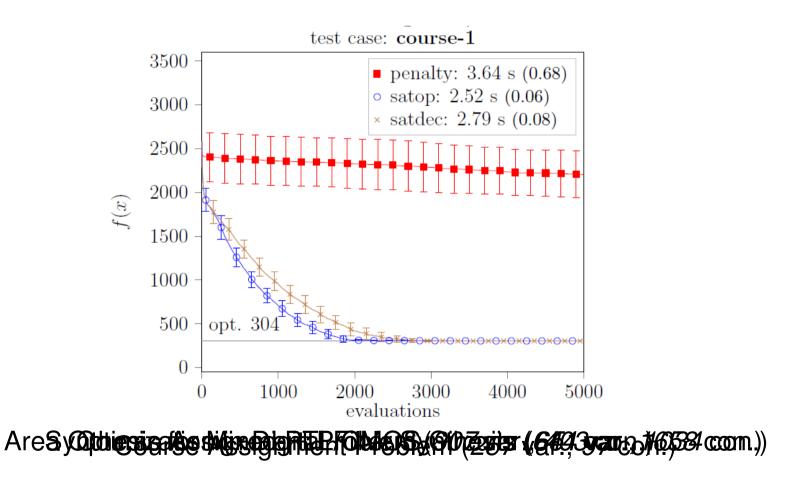


 (ρ, σ) – branching strategy for the PB **x** – feasible solution solver



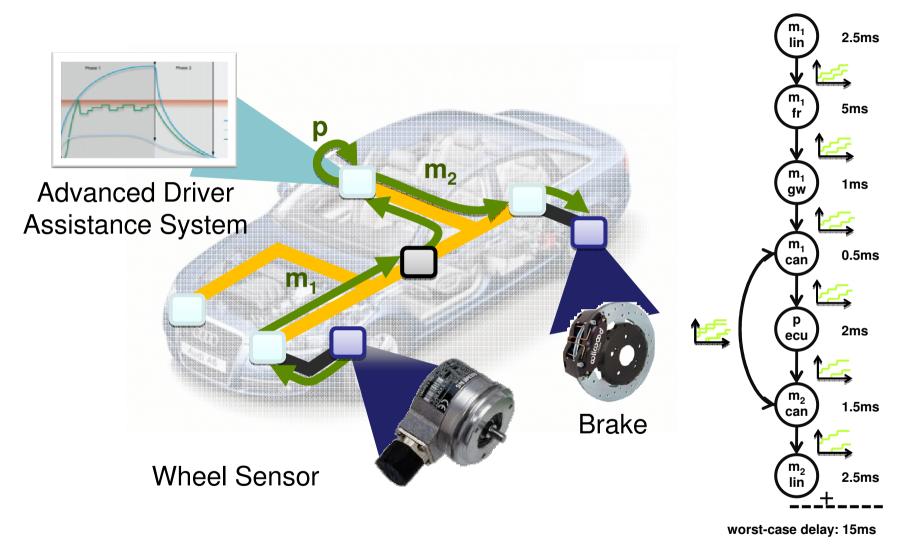
Results

- Comparison using test cases from "PB Evaluation 2009"
 - ILPs with a single objective function and binary variables



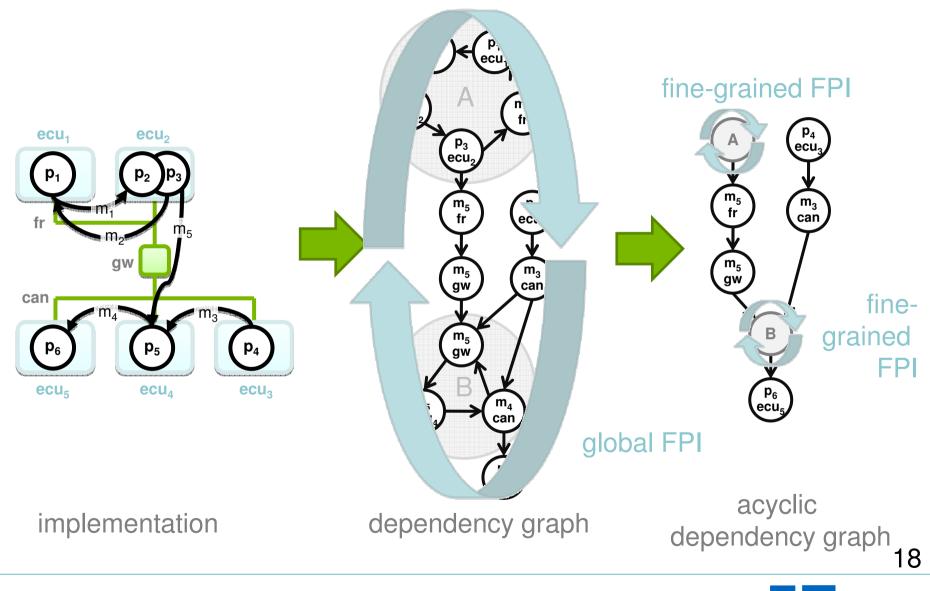
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Compositional Timing Analysis – Motivation

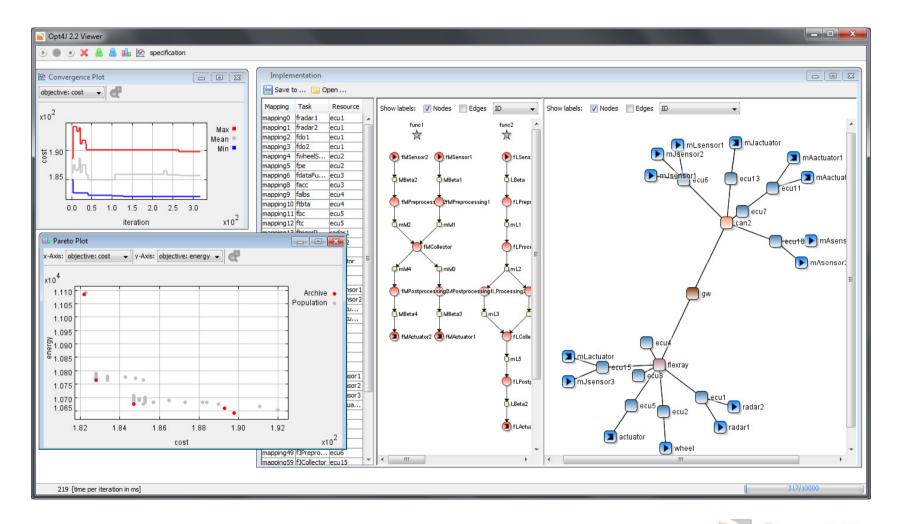


CREATE

Compositional Timing Analysis



Design Space Exploration Framework



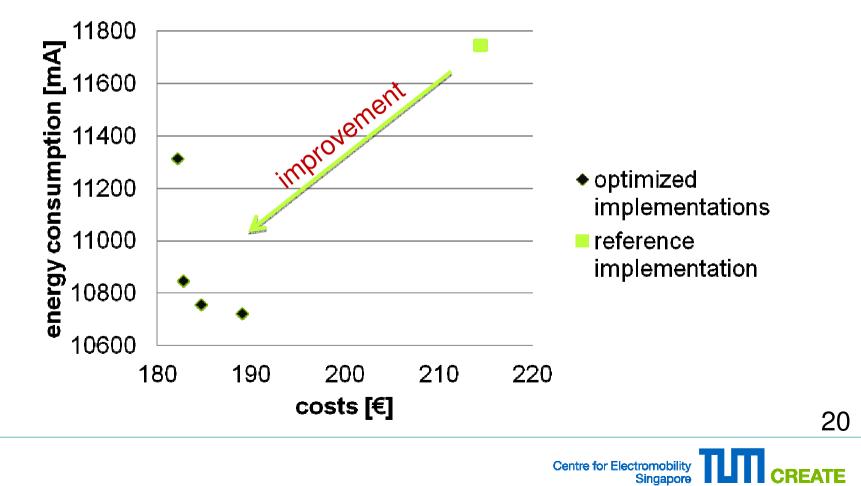
Based on open source optimization framework www.opt4j.org

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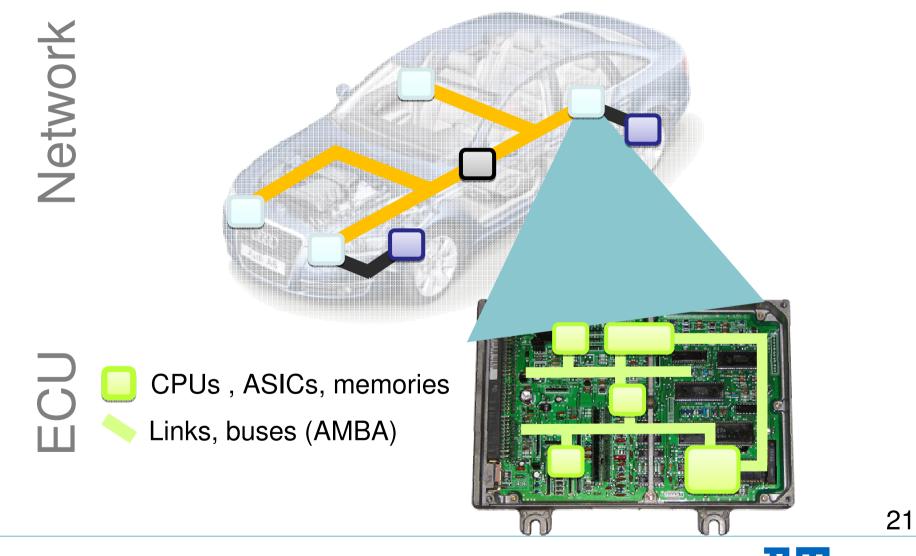
CREATE

Automotive Network Case Study

- Application: 46 tasks, 42 messages
- Architecture: 15 ECUs, 2 CANs, FlexRay, Gateway
- Optimization runtime: 1h (timing analysis)

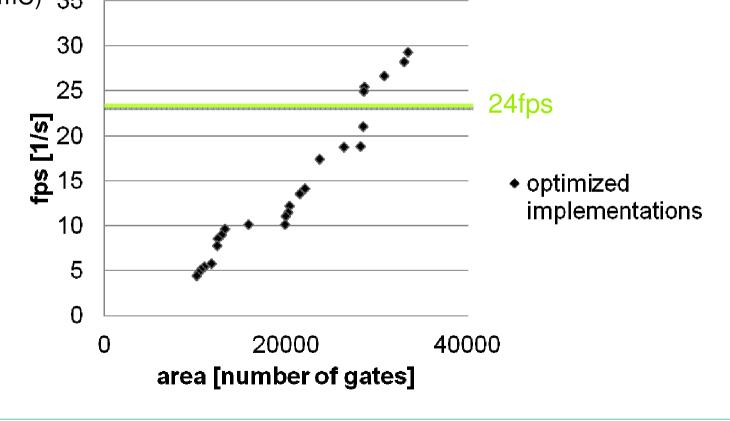


Application Domains



Motion-JPEG Decoder Case Study

- Application: 21 tasks, 56 communication tasks
- Architecture: ARM processors, DSP, buses, gateway
- Optimization runtime: 17h (performance simulation with SystemC) 35



Conclusions

- Design Space Exploration
 - Flexible model
 - Flexible multi-objective optimization
 - Applicable to arbitrary domains, e.g., automotive networks, MPSoCs, NoCs
- Timing Analysis
 - Worst-case analysis
 - Simulations

