



Mapping and Scheduling Automotive Applications on ADAS Platforms using Metaheuristics

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Autonomous vehicles/Assisted driving

Safety – less accidents

Reduced Emissions due to lower fuel consumptions

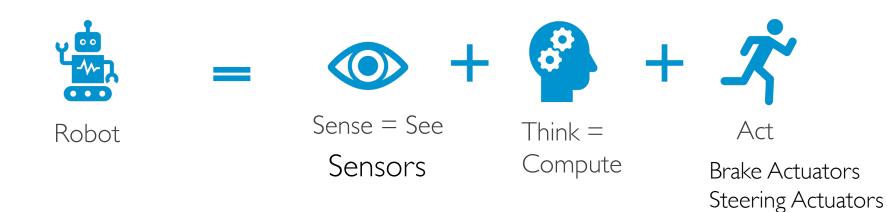
Better Health due to less stress through traffic jams

Increased highway capacity

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An autonomous vehicle can be understood as a robot



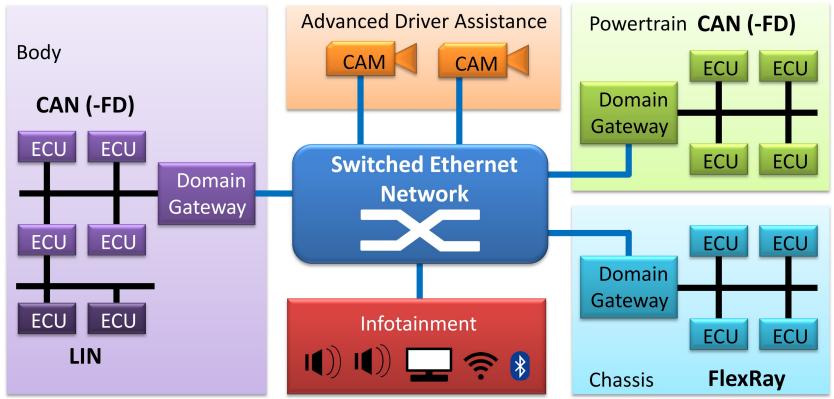


Electric Drivetrain

Modern autonomous vehicles



Traditional approach based on distributed ECUs and separated domains, interconnected through different technologies (ETH, CAN, FlexRay)



Source: Ernst et al. - Ethernet as Future Automotive Communication Backbone

Modern autonomous vehicles – issues



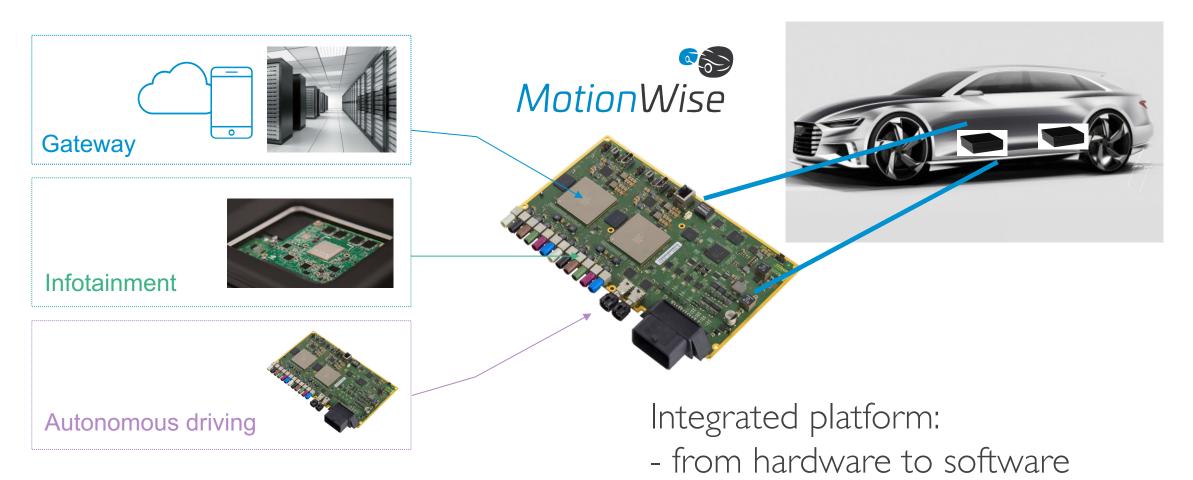
- Rapid growth of software functionality and the necessary compute performance cannot be addressed with current electronics architecture and ECUs
- Too many ECU's with too little processing power and memory
- Limitation of the domain concept (development cost, replication of basic software functions, sources of failure, maintenance cost)
- Fail-operational requirement for level 4 autonomous driving:
 - The domain concept is not sustainable for L4/L5 autonomous driving.
 - Autonomous driving functions require the integration of cross-domain information and functions.



From distributed, separate ECUs \gg In-car Computing Platform

ICCP - Integrated platform





- from distributed to centralized

RazorMotion

Processing Resources:

Ix Renesas RH850P/IH-C (ASIL D MCU with lockstep cores @ 240MHz)

2x Renesas R-Car H3 (ASIL B SoC with 4x Cortex A57, 4x Cortex A53, 1x Cortex R7, 1x IMP-X5, 1x IMG GX6650 GPU)

Video Interfaces:

12 x camera inputs (GMSL) incl. remote supply (PoC) 2 x display outputs (FPD-Link III)

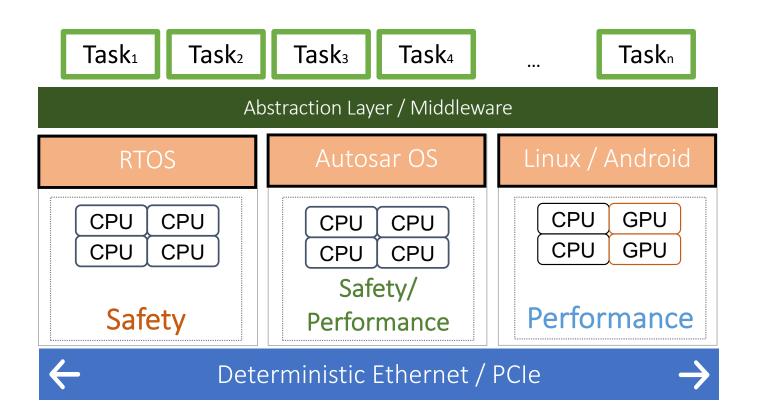
Communication Interfaces:

 $4 \times OABR | 100BASE-T| | 2 \times FlexRay (A/B channel) - wakeup capable 2 \times HS-CAN - wakeup capable 4 \times CAN-FD 2 \times LIN I/O Interfaces 2 \times analog/digital inputs 2 \times high side outputs 1 \times sensor supply output (5V)$



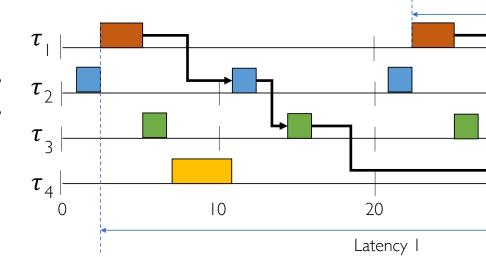






Different multi-core CPUs:

• process the information arriving from a



Heterogeneous multi-core multi-SoC platform featuring a variety of CPUs and GPUs running at different speeds, which are interconnected through either a deterministic Ethernet backbone (TSN) or through PCIe

Integrated platform scheduling problem

T[**[**ech

Periodic hard real-time tasks with (WCET, Period) definition

- Are pre-assigned to CPUs (WCET is already scaled to speed)
- Can be pre-assigned to core, if not assigned, assignment will be part of the allocation problem
- Can have deadline, activation, jitter constraints
- Preemption is allowed, migration is not allowed

Result of scheduling is a static table which determines the exact timely behavior of tasks

Different dimensions to the allocation problem:

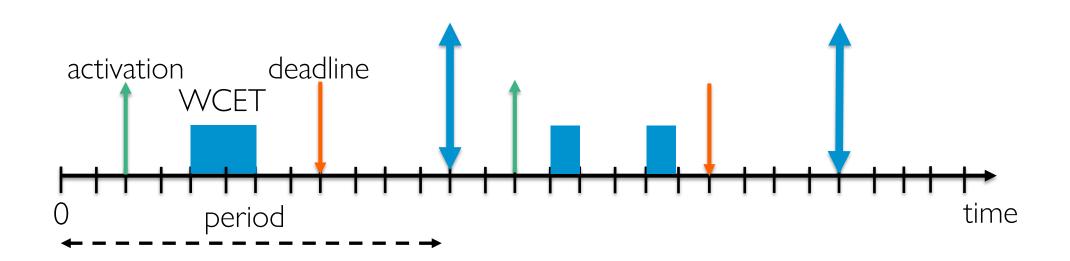
- Assignment of tasks to cores/CPUs
- Scheduling of tasks
- Real-time requirements are met end-to-end





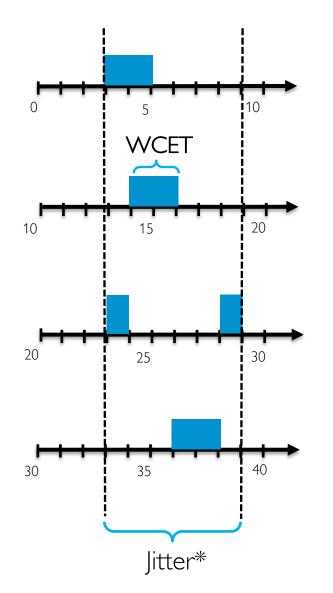
Real-time requirements – activation, deadline, period





Real-time requirements – jitter





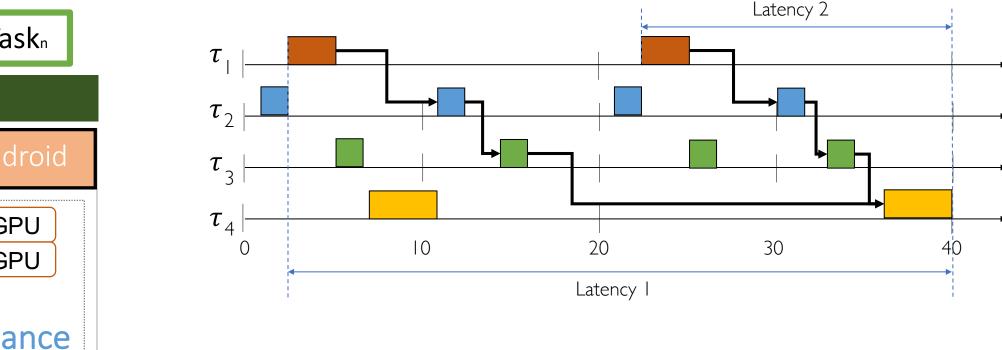
 $Jitter = Jitter^* - WCET$

Real-time requirements – dependency chains



Characteristic of automotive software – Cause-effect chains:

- provide additional timing and dependency requirements on the execution of tasks
- can span across multiple activation patterns
- include multiple tasks, even the same task multiple times
- have priorities and end-to-end latencies
- include communication latencies



SA algorithm

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Simulated Annealing (SA)-based metaheuristic approach which uses an EDF-based heuristic to solve the task scheduling problem.

The scheduling heuristic allows task preemption by simulating an Earliest Deadline First (EDF) scheduling policy parameterized by task offsets and local deadlines decided by SA

```
Algorithm 1 SimulatedAnnealing(\mathcal{A}, \Gamma, s_0, t_s, cr, i)
1: t \leftarrow t_s
 2: s \leftarrow ScheduleSynthesis(\mathcal{A}, \Gamma, s_0)
 3: s^* \leftarrow s
 4: while timeleft do
          while t > 1.0 do
 5:
               for k \leftarrow 1 to i do
 6:
                    s' \leftarrow \text{GenerateNeighbor}(\mathcal{A}, \Gamma, s)
 7:
                    if Cost(s') < Cost(s) then
 8:
                         s \leftarrow s'
 9:
                         if Cost(s') < Cost(s^*) then
10:
                              s^* \leftarrow s'
11:
                         end if
12:
                    else if exp(\frac{Cost(s)-Cost(s')}{t}) > random[0,1] then
13:
                         s \leftarrow s'
14:
                    end if
15:
                    t \leftarrow t \cdot (1 - cr)
16:
               end for
17:
          end while
18:
19: end while
20: return s^*
```

EDF simulation

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EDF is an optimal online scheduling algorithm which at each time instant prioritizes the task with the earliest deadline

We can use it to generate a static schedule table – simulate EDF until 2*Hyperperiod + max_offset

Schedulability test: $\forall t_1 \in \Phi^{\sigma}, \forall t_2 \in \Delta^{\sigma}, t_1 < t_2$:

$$\sum_{\tau_i \in \Gamma^{\sigma}} C_i \times \left(\left\lfloor \frac{t_2 - \phi_i - D_i}{T_i} \right\rfloor - \left\lceil \frac{t_1 - \phi_i}{T_i} \right\rceil + 1 \right)_0 \le t_2 - t_1,$$

where

$$\begin{split} \Phi^{\sigma} &\stackrel{def}{=} \{a_{i,j} = \phi_i + j \times T_i | \tau_i \in \Gamma^{\sigma}, j \ge 0, a_{i,j} \le \lambda^{\sigma} \}, \\ \Delta^{\sigma} &\stackrel{def}{=} \{d_{i,j} = a_{i,j} + D_i | \tau_i \in \Gamma^{\sigma}, j \ge 0, d_{i,j} \le \lambda^{\sigma} \}, \\ \lambda^{\sigma} &= max(\{\phi_i | \tau_i \in \Gamma^{\sigma}\}) + 2 \times lcm(\{T_i | \tau_i \in \Gamma^{\sigma}\}). \end{split}$$

Two knobs to play around with: offset and deadline of each task

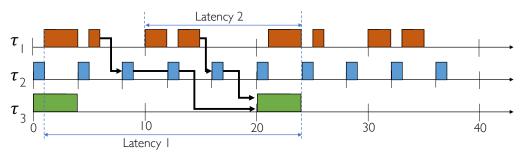
Simulated Annealing + EDF simulation



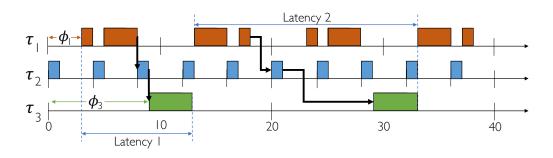
Simple Algorithm:

Generate initial candidate:

- task offsets = 0
- task deadlines = Period
- task to core assignment based on best-fit/first-fit (load balancing)



(a) End-to-end task chain latencies **not** satisfied



od design transformations

- SwapTask, AdjustOffset and AdjustDeadline
- Evaluate a solution based on the cost metric

Generate new candidate through performing

EDF schedulability test/EDF simulator

Simulated Annealing Loop:

$$Cost(s) = \begin{cases} \sum_{\substack{\aleph_i \in \mathcal{L}_{\aleph} \\ |\mathcal{L}_{\aleph}|}} \frac{1}{|\mathcal{L}_{\aleph}|} \cdot w_1 & \text{if } \chi(s) = \text{true} \\ w_1 + \rho_{\aleph} + \rho_D + \rho_J & \text{if } \chi(s) = \text{false} \end{cases}$$

Experimental evaluation



Five test cases, ranging from 100% to 500% in scale, i.e., for ADAS1×100% the application contains 151 tasks and 31 chains using a model of the architecture A test case is a scenario consisting of 30 synthetically generated task sets, with each undergoing 30 trials (900 trials for each algorithm)

Test case	Time	Greedy				SA						GA										
		Chains		Jitter		Sched.	. Chains		Jitter		Sched.	Chains		Jitter		•	Sched.					
		Min	Avg	Max	Min	Avg	Max		Min	Avg	Max	Min	Avg	Max		Min	Avg	Max	Min	Avg	Max	
ADAS1x100%	1 hour	0.97	0.98	1.00	0.58	0.61	0.68	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
ADAS1x200%	2 hours	0.97	0.99	1.00	0.55	0.67	0.75	1.00	0.98	1.00	1.00	0.94	1.00	1.00	1.00	0.98	1.00	1.00	0.71	0.95	1.00	1.00
ADAS1x300%	3 hours	0.97	0.99	1.00	0.52	0.64	0.72	1.00	0.97	0.99	1.00	0.70	0.87	1.00	1.00	0.97	0.99	1.00	0.70	0.88	1.00	1.00
ADAS1x400%	4 hours	0.97	0.97	0.98	0.52	0.64	0.73	1.00	0.97	0.99	1.00	0.69	0.80	0.88	1.00	0.94	0.99	1.00	0.70	0.81	0.92	1.00
ADAS1x500%	5 hours	0.97	0.98	0.98	0.51	0.62	0.70	1.00	0.95	0.98	0.99	0.63	0.78	0.86	1.00	0.95	0.98	1.00	0.64	0.79	0.87	1.00

EVALUATION RESULTS ON SYNTHETIC TEST CASES

EVALUATION RESULTS ON REALISTIC TEST CASES

	Test case Tin	ne	Greedy			SA	
Real-world test-cases with 151 tasks and		Chair	ns Jitter S	Sched.	Chains	Jitter	Sched.
					Min Avg Max	Min Avg May	X
31 chains	ADAS1 3.2	0.81	0.37	1.00	0.97 0.99 1.00	0.95 0.99 1.00) 1.00
ST Chains	ADAS2 6.4	0 0.65	0.21	1.00	0.94 0.99 1.00	0.84 0.99 1.0	1.00
	ADAS3 13.2	20 0.48	0.21	1.00	0.84 0.99 1.00	0.74 0.97 1.0	1.00

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