

Self-Aware CPSs

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oCPS Webinar

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Acknowledgment

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University of Turku, Finland

Outline

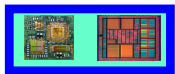
- 1 Motivation
- 2 Architecture for Awareness
- 3 Comprehensive Observation
- 4 Goal Management
- 5 Conclusion

Outline

- 1 Motivation
- 2 Architecture for Awareness
- 3 Comprehensive Observation
- 4 Goal Management
- 5 Conclusion



The Problem

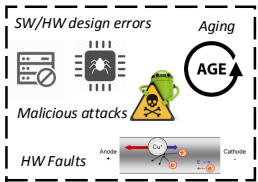


Varying Application and User Demands

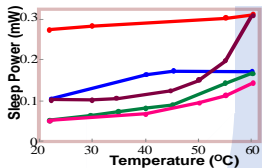
workload phasic behavior
user inputs
varying compute, memory, and communication



Functional Aberrations

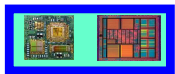


Non-functional Aberrations



The Problem

- Large number of resources

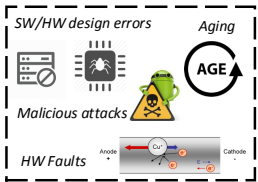


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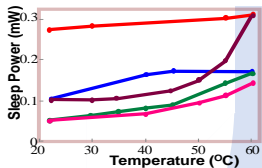
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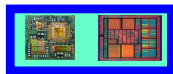


Non-functional Aberrations



The Problem

- Large number of resources
- Many tight constraints

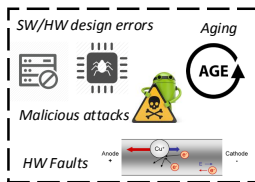


Varying Application and User Demands

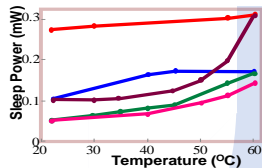
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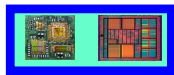


Non-functional Aberrations



The Problem

- Large number of resources
- Many tight constraints
- Varying application demands, both within and between applications;

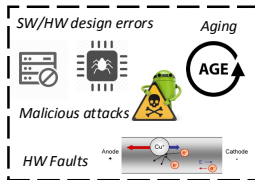


Varying Application and User Demands

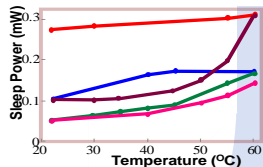
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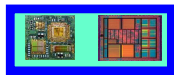


Non-functional Aberrations



The Problem

- Large number of resources
- Many tight constraints
- Varying application demands, both within and between applications;
- Functional Aberrations:
 - Design errors or omissions;
 - Malicious attacks;
 - Aging;
 - Soft errors;
- Non-functional Aberrations:
 - Performance;
 - Power consumption;

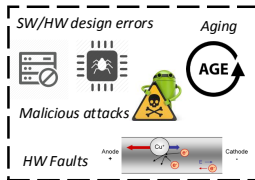


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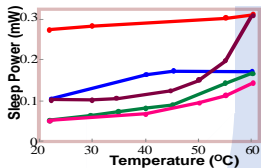
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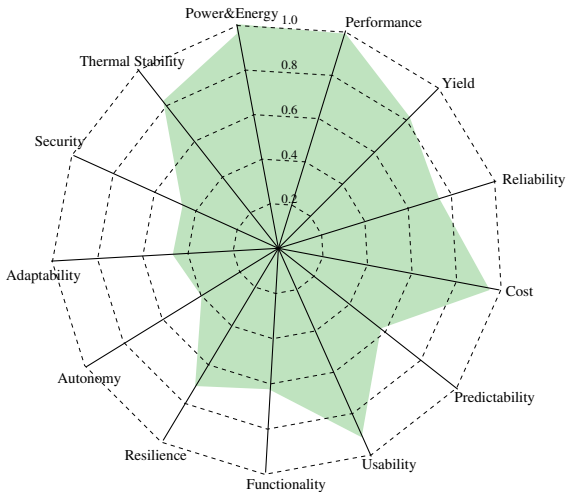
Functional Aberrations



Non-functional Aberrations



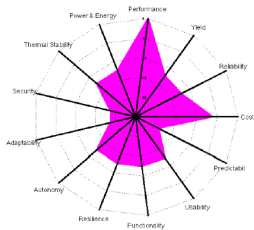
The SoC Radar



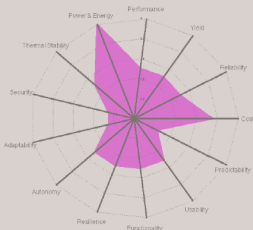
Santanu Sarma et al. "On-Chip Self-Awareness Using Cyberphysical-Systems-On-Chip (GPSoC)". In: *Proceedings of the 12th International Conference on Hardware/Software Codesign and System Synthesis (CODES+ISSS)*. New Delhi, India, Oct. 2014

The SoC Radar

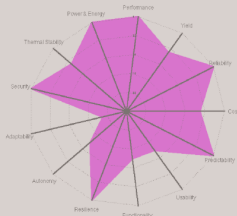
Reality



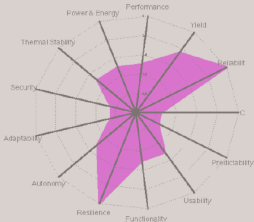
Performance Driven



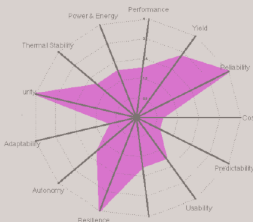
Energy/Power Driven



QoS Combination



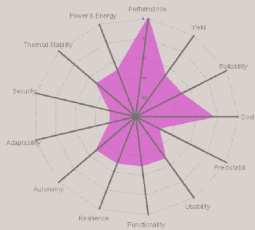
Reliability Driven



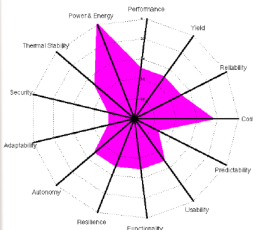
Security Driven

The SoC Radar

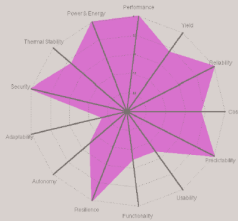
Reality



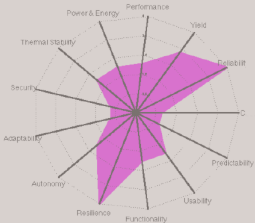
Performance Driven



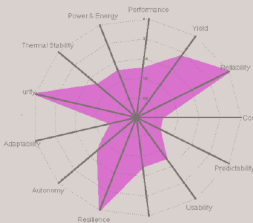
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QoS Combination



Reliability Driven

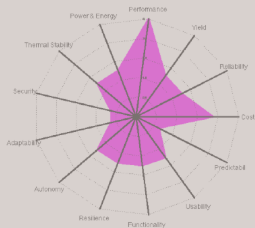


Security Driven

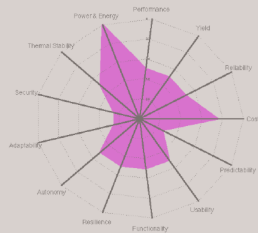


The SoC Radar

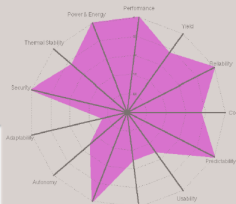
Reality



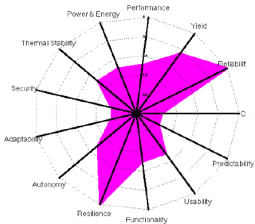
Performance Driven



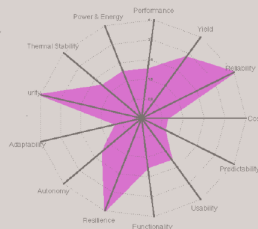
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QoS Combination



Reliability Driven

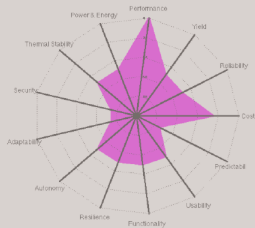


Security Driven

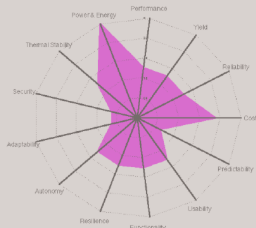


The SoC Radar

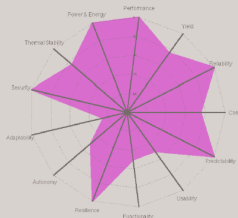
Reality



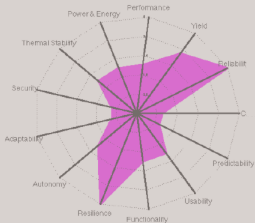
Performance Driven



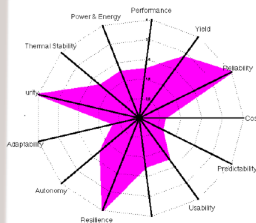
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QoS Combination



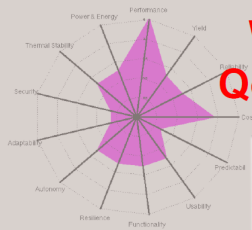
Reliability Driven



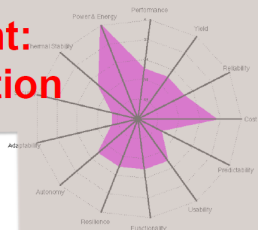
Security Driven

The SoC Radar

**What we want:
QoS Combination**



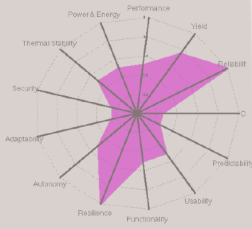
Performance Driven



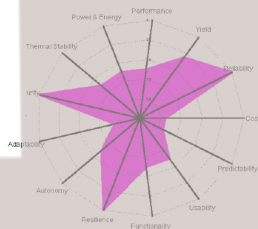
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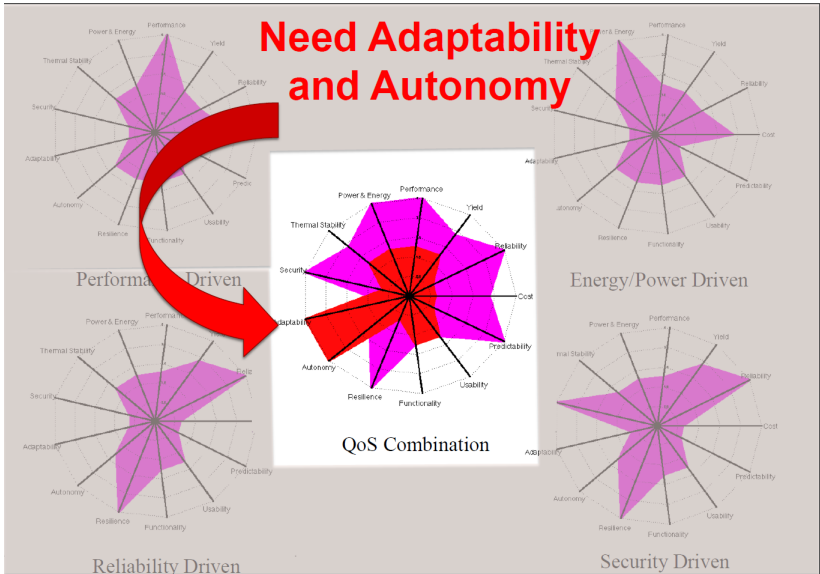
Reliability Driven



Security Driven

The SoC Radar

Need Adaptability and Autonomy



Autonomy and Adaptivity

Autonomy is the ability to operate independently, without external control.

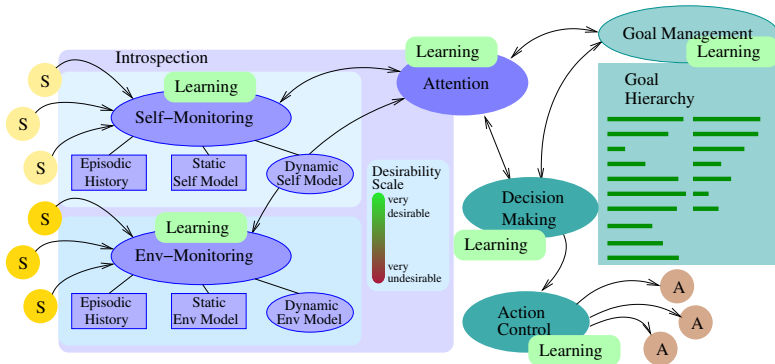
Adaptivity is the ability to effect run-time changes and handle unexpected events.



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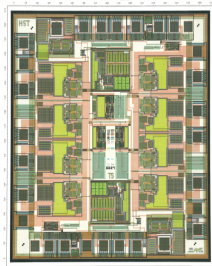
Self-Awareness Architecture



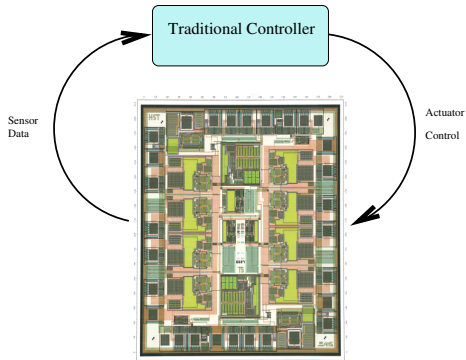
Cyber-Physical SoC



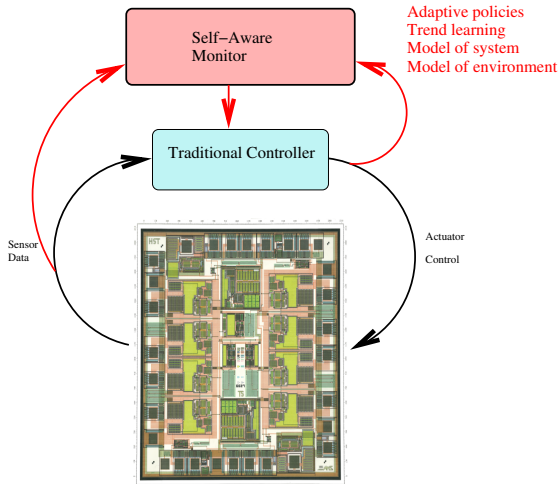
Cyber-Physical SoC



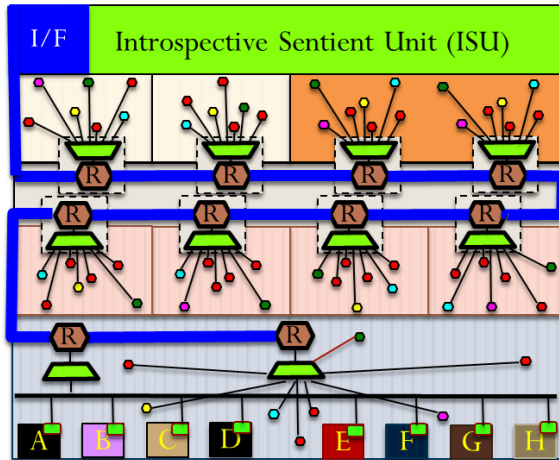
Cyber-Physical SoC



Cyber-Physical SoC

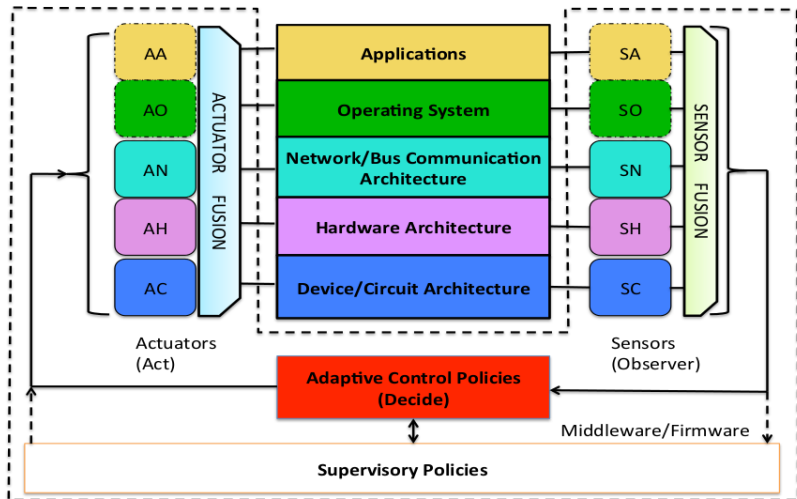


CPSoC - A Sensor Rich SoC Platform



Santanu Sarma et al. "CyberPhysical-System-On-Chip (CPSoC): A Self-Aware MPSoC Paradigm with Cross-Layer Virtual Sensing and Actuation". In: *Proceedings of the Design, Automation and Test in Europe Conference and Exhibition (DATE)*. Grenoble, France, Mar. 2015

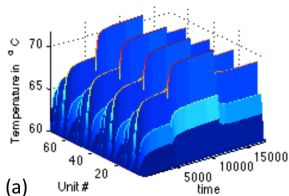
CPSoC - A Sensor Rich SoC Platform



Nikil Dutt, Axel Jantsch, and Santanu Sarma. "Self-Aware Cyber-Physical Systems-on-Chip". In: *Proceedings of the International Conference for Computer Aided Design*. invited. Austin, Texas, USA, Nov. 2015

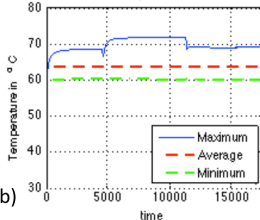
Thermal-Aware Performance

Thermal Profile with MPSoC



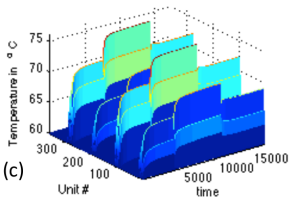
(a)

Maximum Temperature overtime



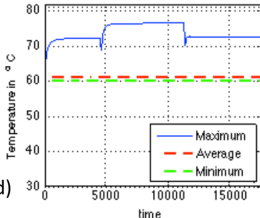
(b)

Thermal Profile with CPSoC



(c)

Maximum Temperature overtime



(d)

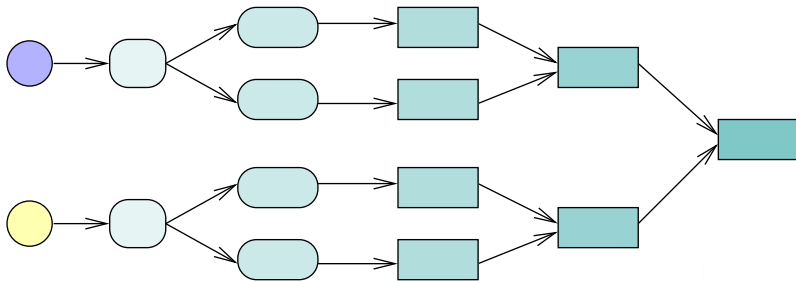
Throughput improvement by 70%-300% for same power and temperature.

Benefit is due to accurate and fine-grain measurement and tight tracking.

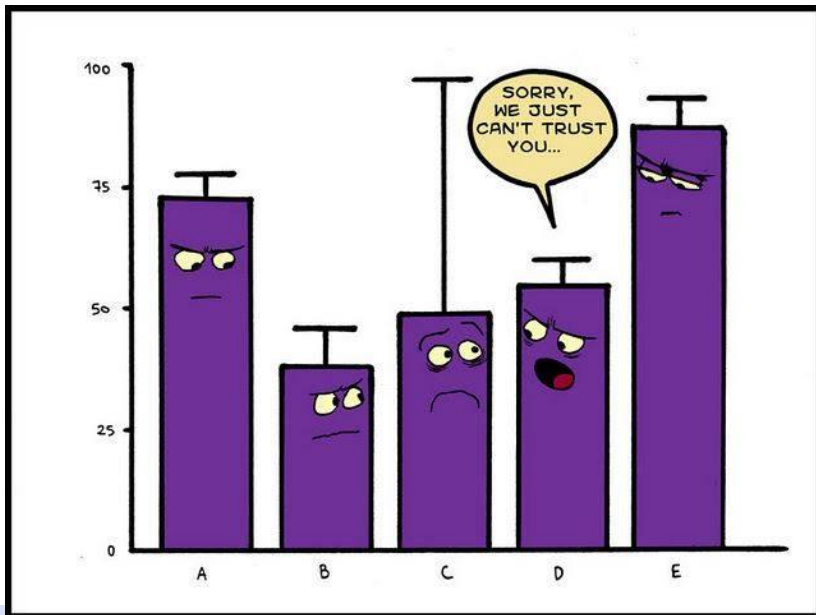
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Observation Pipeline



Data and Meta-Data



Data and Meta-Data



Data and Meta-Data

Accuracy Systematic errors, a measure of statistical bias.



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Precision Random errors, a measure of statistical variability.



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Data Reliability The extent to which a measuring procedure yields the same results on repeated trials.



Data and Meta-Data

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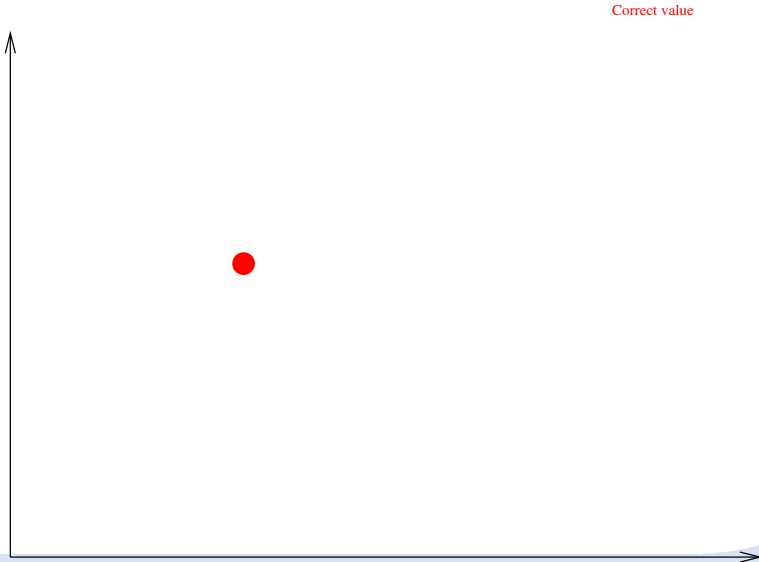
Precision Random errors, a measure of statistical variability.

Data Reliability The extent to which a measuring procedure yields the same results on repeated trials.

Relevance The quality of being important for the matter at hand.



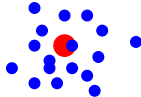
Accuracy and Precision



Accuracy and Precision

Correct value

High accuracy, high precision

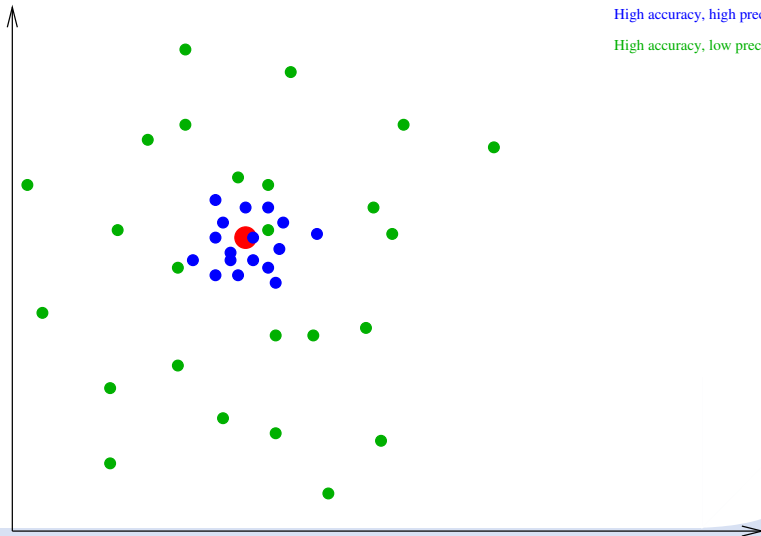


Accuracy and Precision

Correct value

High accuracy, high precision

High accuracy, low precision



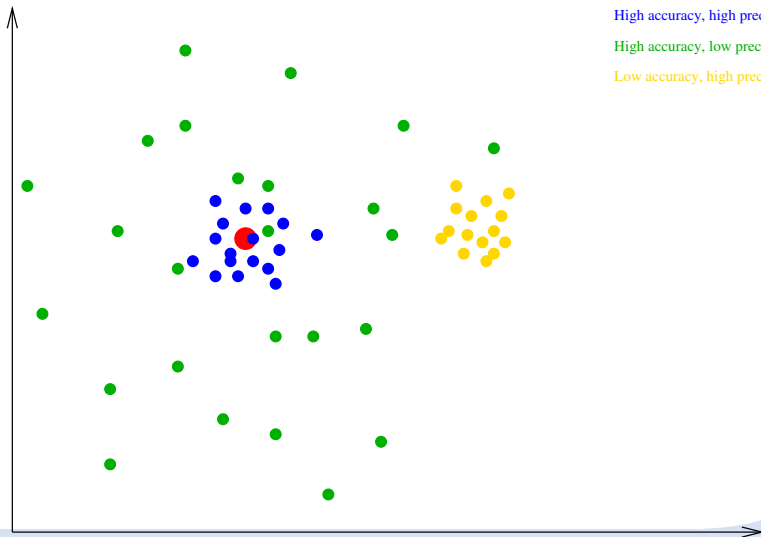
Accuracy and Precision

Correct value

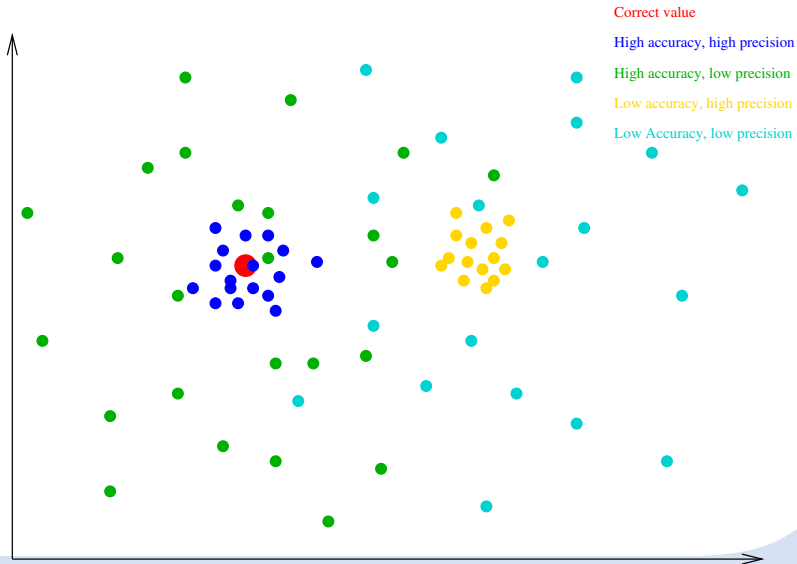
High accuracy, high precision

High accuracy, low precision

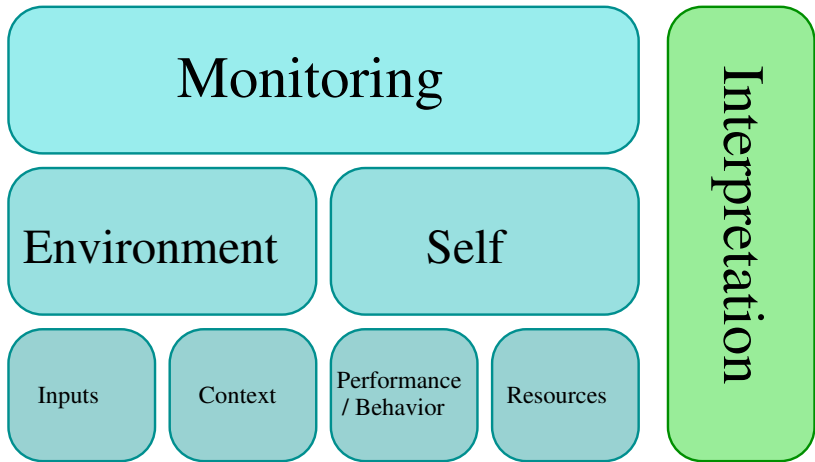
Low accuracy, high precision



Accuracy and Precision

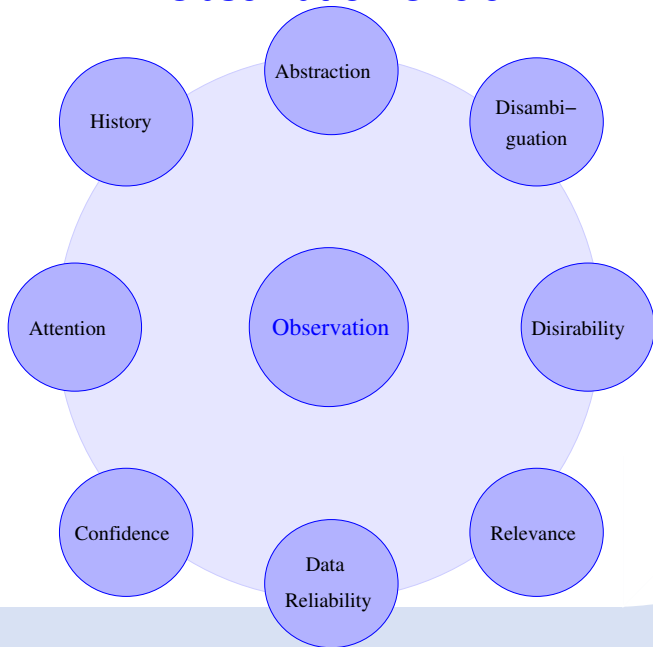


Comprehensive Observation



Nima TaheriNejad, Axel Jantsch, and David Pollreisz. "Comprehensive Observation and its Role in Self-Awareness - An Emotion Recognition System Example". In: *Proceedings of the Federated Conference on Computer Science and Information Systems*. Gdansk, Poland, Sept. 2016

Observation Circle



Early Warning Score

Score	3	2	1	0	1	2	3
Heart rate ¹	<40	40–51	51–60	60–100	100–110	110–129	>129
Systolic BP ²	<70	70–81	81–101	101–149	149–169	169–179	>179
Breath rate ³		<9		9–14	14–20	20–29	>29
SPO ₂ (%)	<85	85–90	90–95	>95			
Body temp. ⁴	<28	28–32	32–35	35–38		38–39.5	>39.5

¹beats per minute, ²mmHg, ³breaths per minute, ⁴ °C

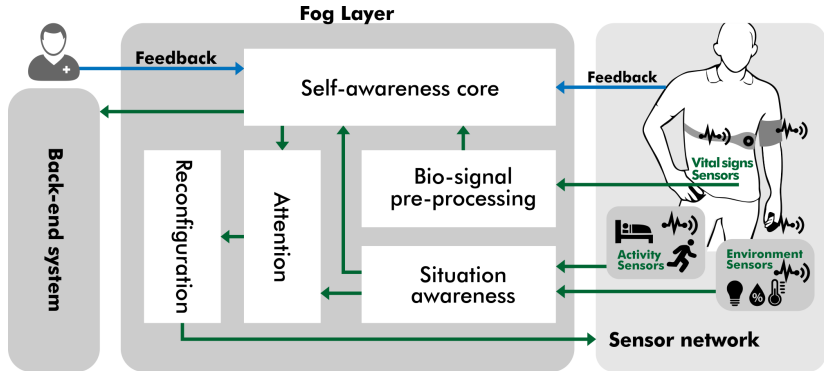


EWS Improvement

- Data reliability:
 - Values in reasonable scope
 - Changes in reasonable scope
 - Consistency between sensors
- Situation awareness
- Power efficiency



Enhanced Early Warning Score



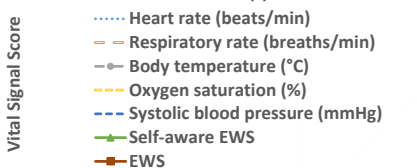
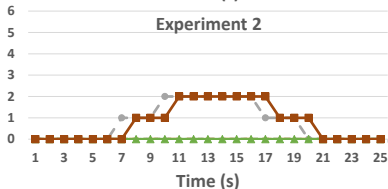
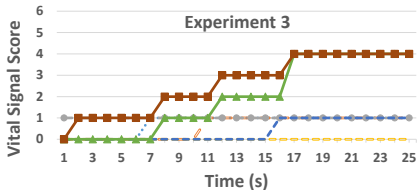
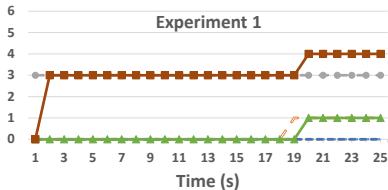
Arman Anzanpour et al. "Self-Awareness in Remote Health Monitoring Systems using Wearable Electronics". In: *Proceedings of Design and Test Europe Conference (DATE)*. Lausanne, Switzerland, Mar. 2017

Enhanced Early Warning Score - Data Reliability

- 1 Check on the reliability of sensed values
- 2 Check on the reliability of value changes
- 3 Check on consistency between sensor data

Enhanced Early Warning Score - Data Reliability

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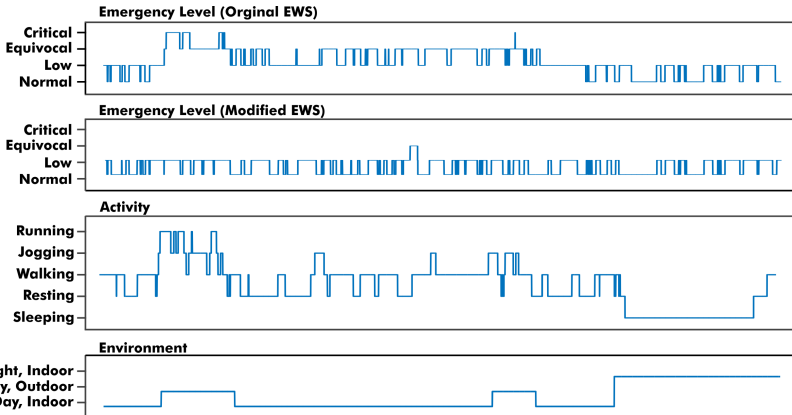


Enhanced Early Warning Score - Situation Awareness

- 1 Consider the activity mode of person
- 2 Consider time of day
- 3 Consider location

Enhanced Early Warning Score - Situation Awareness

- 1 Consider the activity mode of person
- 2 Consider time of day
- 3 Consider location



Enhanced Early Warning Score - Power Efficiency

1 Prioritize different situations



Enhanced Early Warning Score - Power Efficiency

- 1 Prioritize different situations
- 2 Distinguish different modes of urgency

Emergency Level:	Score:0 Normal				Score:1-3 Low				Score:4-6 Medium				Score>6 High			
	Indoor		Outdoor		Indoor		Outdoor		Indoor		Outdoor		Indoor		Outdoor	
	Day	Night	Day	Night	Day	Night	Day	Night	Day	Night	Day	Night	Day	Night	Day	Night
Sleeping	E	E	E	E	C	D	D	D	B	C	C	C	A	A	B	B
Resting	D	D	D	D	C	C	C	C	B	B	B	B	A	A	B	B
Walking	C	C	C	C	B	C	C	C	B	B	B	B	A	A	A	B
Jogging	C	C	C	C	B	B	B	C	B	B	B	B	A	A	A	B
Running	C	C	C	C	B	B	B	B	B	B	B	B	A	A	A	A

Enhanced Early Warning Score - Power Efficiency

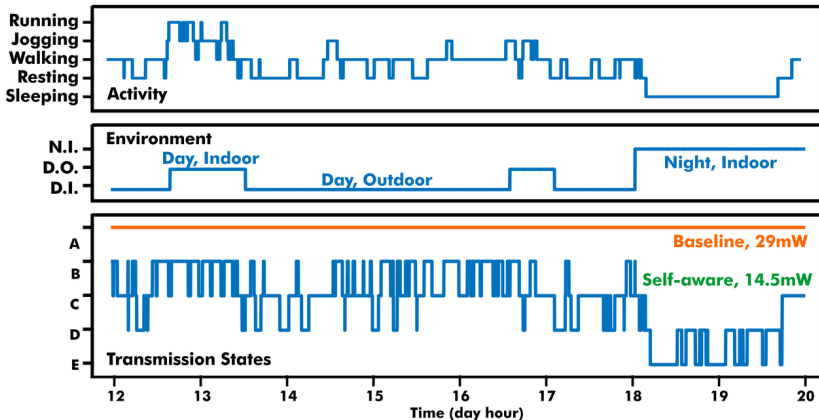
- 1 Prioritize different situations
- 2 Distinguish different modes of urgency
- 3 Define sensing activity for each mode

State	Respiration Rate Activity	Blood Pressure	Heart Rate, SpO2, and Body Temp.	Transmission Power Consumption
A	Continuous	Every hour in day Disabled in night	Every sec.	29 mW
B	2 min continuous 8 min OFF	Every hour in day Disabled in night	Every sec.	26.8 mW
C	2 min continuous 3 min OFF	Every 3 hours in day Disabled in night	Every min.	12.5 mW
D	2 min continuous 8 min OFF	Every 3 hours in day Disabled in night	Every min.	7 mW
E	2 min continuous 18 min OFF	Disabled	Every min.	4.3 mW



Enhanced Early Warning Score - Power Efficiency

Over a day half the energy can be saved.



Enhanced Early Warning Score Summary

- Considering data reliability improves quality of observation;
- Considering situation improves quality of observation;
- Collecting needed data only improves efficiency.



Attention Based Temperature Measurement



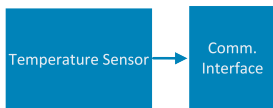
Attention Based Temperature Measurement

- How many temperature measurements are required in an MPSoC?
- It varies over several orders of magnitude depending on activity and current temperature.

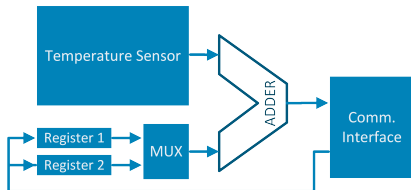


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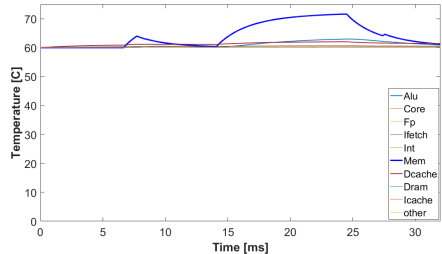
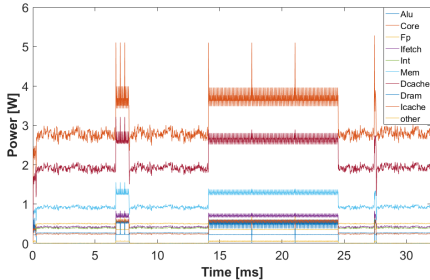
Conventional Architecture



Proposed Architecture

Nima TaheriNejad, M. Ali Shami, and Sai Manoj P. D. "Self-aware sensing and attention-based data collection in Multi-Processor System-on-Chips". In: *15th IEEE International New Circuits and Systems Conference (NEWCAS)*. June 2017, pp. 81–84

Attention Based Temperature Measurement



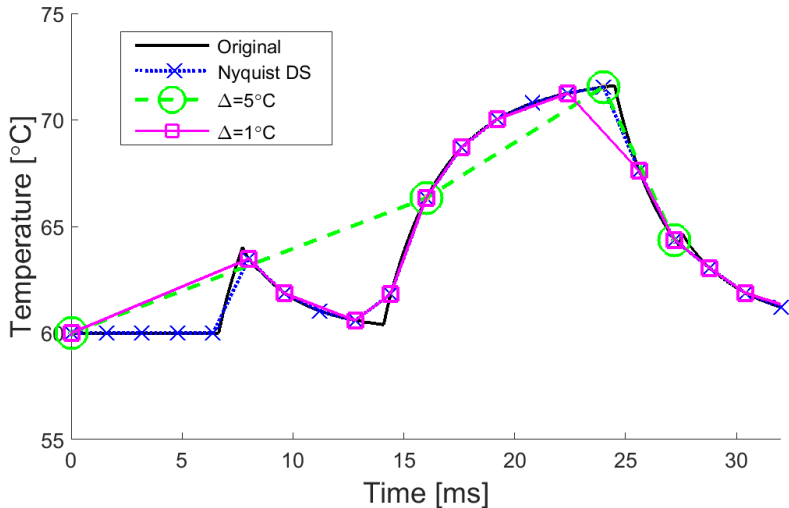
Intel Nehalem processor, running Barnes from SPLASH-2 Benchmarks, using Snipersim and Hotspot.

Attention Based Temperature Measurement

- When only differences $> \Delta = 1, 2, 5^{\circ}\text{C}$ are reported, 7 out of 10 sensors send only 1 value in this experiment.
- Reduction of temperature reports for Memory, ALU and D-Cache:

Unit	$\Delta = 1$	Imp.	$\Delta = 2$	Imp.	$\Delta = 5$	Imp.
Memory	13	35%	9	55%	4	80%
ALU	4	80%	2	90%	1	95%
D-Cache	2	90%	2	90%	1	95%
All others	1	95%	1	95%	1	95%

Attention Based Temperature Measurement



Attention Based Temperature Measurement

- Rate of temperature reporting can be significantly reduced and fine tuned;



Attention Based Temperature Measurement

- Rate of temperature reporting can be significantly reduced and fine tuned;
- Can depend on
 - relative difference,
 - absolute difference,
 - absolute value,
 - system level mode;

Attention Based Temperature Measurement

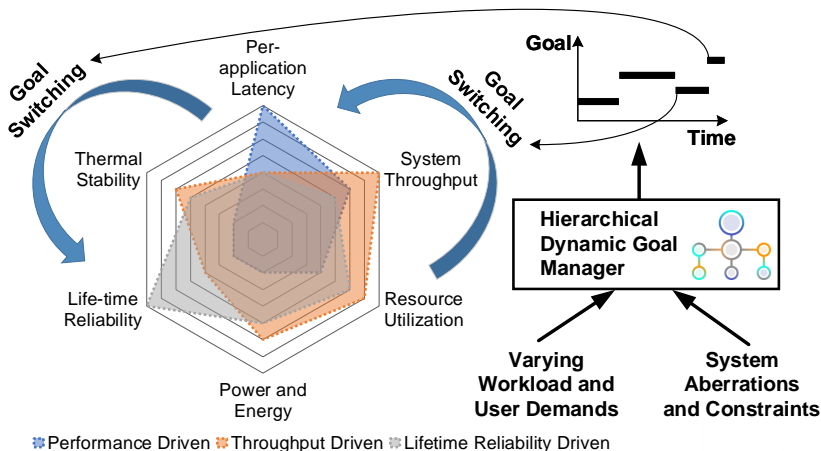
- Rate of temperature reporting can be significantly reduced and fine tuned;
- Can depend on
 - relative difference,
 - absolute difference,
 - absolute value,
 - system level mode;
- Potential benefits:
 - reduced processing,
 - reduced communication,
 - reduced measurements.



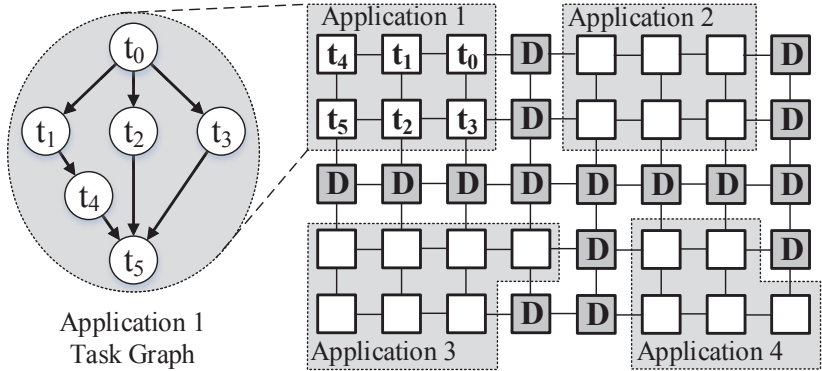
Outline

- 1 Motivation
- 2 Architecture for Awareness
- 3 Comprehensive Observation
- 4 Goal Management**
- 5 Conclusion

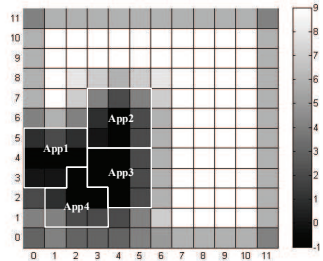
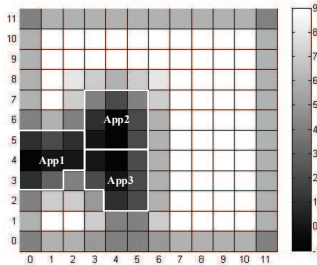
Goals for Dynamic Task Mapping



Dynamic Task Mapping



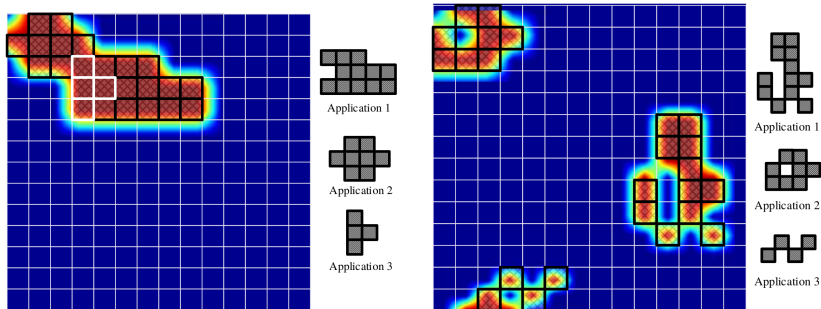
Example 1: Performance Driven Task Mapping



MapPro prefers compact and contiguous regions.

Mohammad-Hashem Haghbayan et al. "MapPro: Proactive Runtime Mapping for Dynamic Workloads by Quantifying Ripple Effect of Applications on Networks-on-Chip". In: *Proceedings of the International Symposium on Networks on Chip*. Vancouver, Canada, Sept. 2015

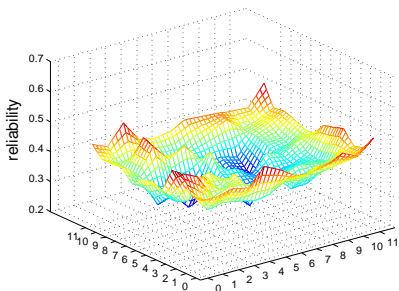
Example 2: Throughput- and Power-Constrained Task Mapping



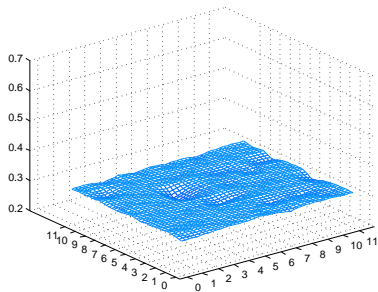
The patterning algorithm disperses mapped cores to maximize the Thermal Safe Power budget.

Anil Kanduri et al. "Dark Silicon Aware Runtime Mapping for Many-core Systems: A Patterning Approach". In: *Proceedings of the International Conference on Computer Design (ICCD)*. New York City, USA, Oct. 2015, pp. 610–617

Example 3: Lifetime-Reliability-Driven Task Mapping



MapPro:
lifetime=5.52 years



Reliability aware mapping:
lifetime=12 years

The plots show the reliability of cores at the end of the system's lifetime.
The end of the system's life is reached when the reliability of one core drops below 30%.

M. H. Haghbayan et al. "A lifetime-aware runtime mapping approach for many-core systems in the dark silicon era".
In: *Design, Automation Test in Europe Conference Exhibition (DATE)*. Mar. 2016, pp. 854–857

Goal Management Levels

- 1 Single objective; Design time;

Goal Management Levels

- 1 Single objective; Design time;
- 2 Multiple objectives; Design time;

Goal Management Levels

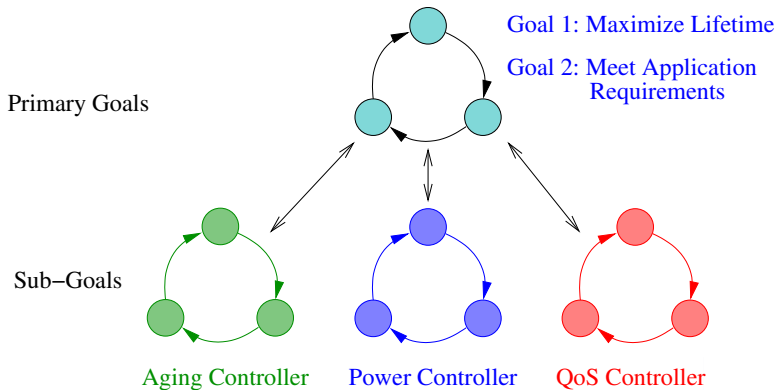
- 1 Single objective; Design time;
- 2 Multiple objectives; Design time;
- 3 Multiple objectives; Run time;

Goal Management Levels

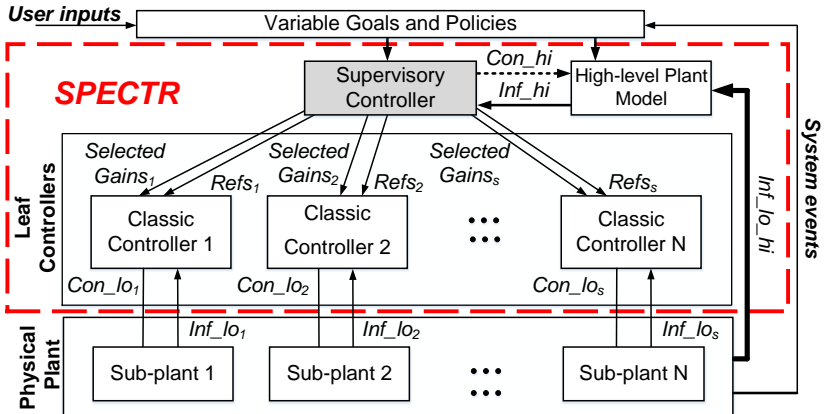
- 1 Single objective; Design time;
- 2 Multiple objectives; Design time;
- 3 Multiple objectives; Run time;
- 4 Multiple, hierarchical objectives; Run time;



Hierarchical Goal Management

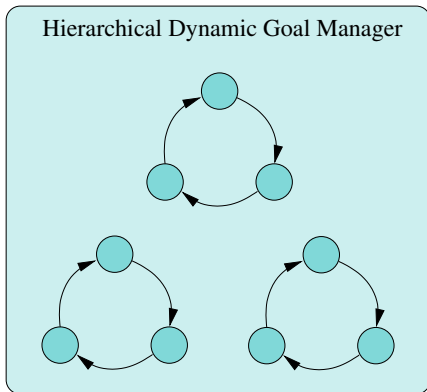


Supervisory Control



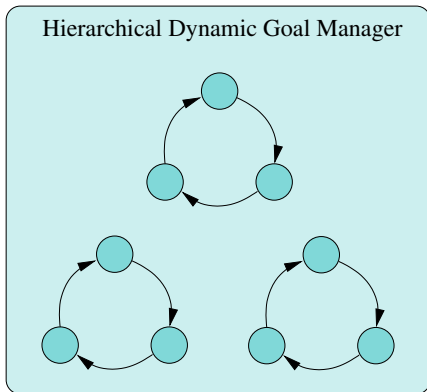
Amir M. Rahmani et al. "SPECTR - Formal Supervisory Control and Coordination for Many-core Systems Resource Management". In: *Proceedings of the 23rd ACM International Conference on Architectural Support for Programming Languages and Operating Systems*. Williamsburg, VA, USA, Mar. 2018; T. R. Mück et al. "Design Methodology for Responsive and Robust MIMO Control of Heterogeneous Multicores". In: *IEEE Transactions on Multi-Scale Computing Systems* PP.99 (2018), pp. 1–1

Goal Management Inputs

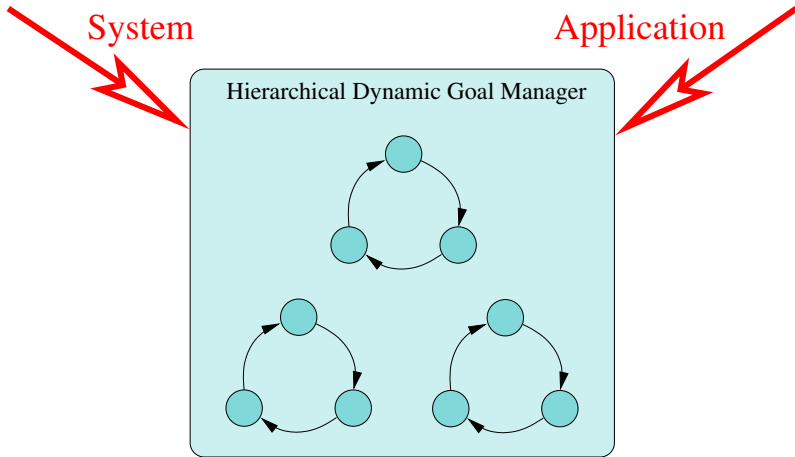


Goal Management Inputs

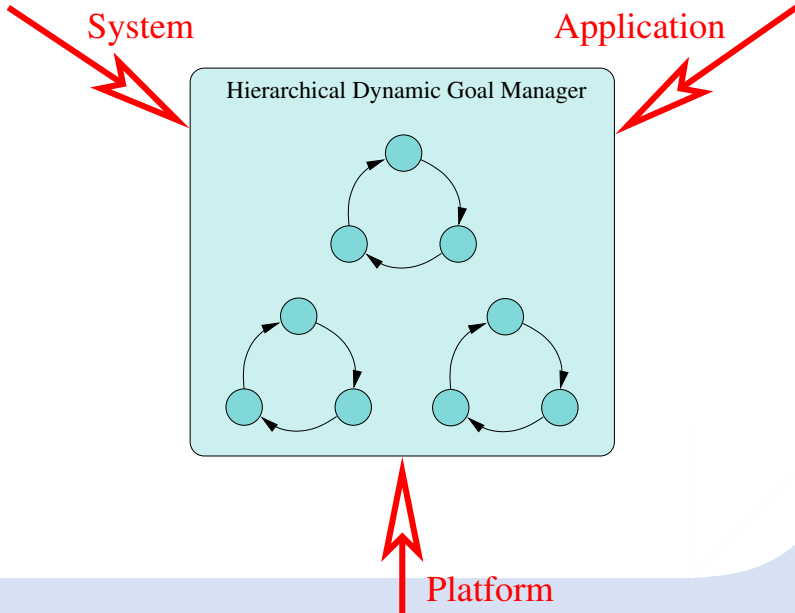
Application



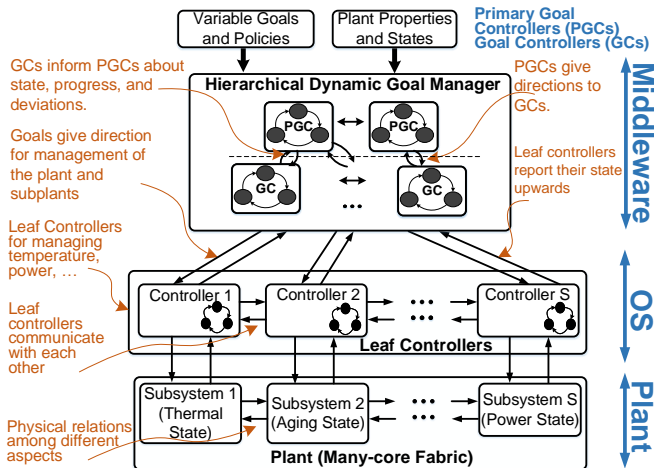
Goal Management Inputs



Goal Management Inputs

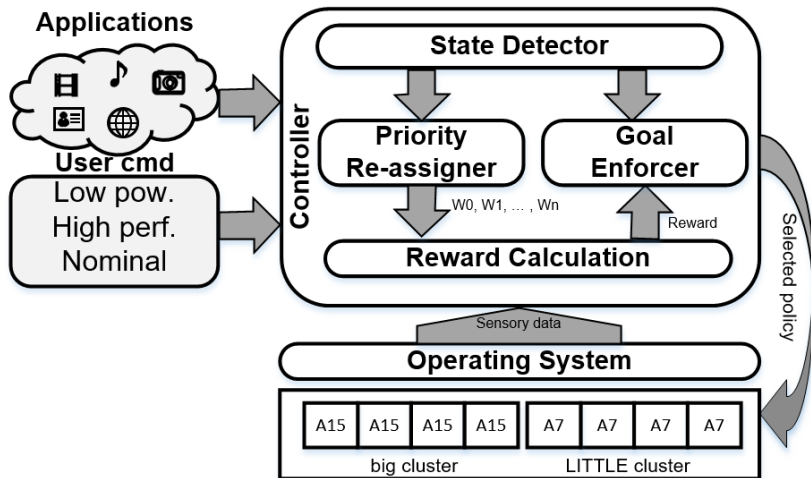


Hierarchical Goal Management



- The system's requirements changes over its lifetime.
- Different objectives are invoked at different time.

Goal Driven Autonomy



State Detection

State vector:

- **Power:** Violation: $TDP < p$
Potential Violation: $0.8 TDP \leq p \leq TDP$
No Violation: $p \leq 0.8 TDP$
- **User Command:** High Performance
Low Power
- **Performance per application:**
[Min run time, Max run time]

Priority Assignment

- Primary goals: thermal safety
- Secondary goals: User experience
- Tertiary goals: Application requirements

Priority Assignment - Urgency

Urgency is the extent of a violation of a parameter:

$$U_{Pow} = \frac{P_{cur}}{P_{ref}}$$

P_{cur} is the instantaneous power consumption;

P_{ref} is the fixed upper bound on power (TDP)

Priority Assignment - Urgency

$$U_{perf} = \frac{perf_{max} - perf_{curr}}{perf_{max} - perf_{ref}}$$

$perf_{max}$ the maximum required application performance;

$perf_{curr}$ the instantaneous measured performance;

$perf_{ref} \begin{cases} \frac{perf_{max} + perf_{min}}{2} & \text{if User Command = High Performance} \\ perf_{min} & \text{if User Command = Low Power} \end{cases}$

Goal Enforcement

- Selects action that most likely will satisfy the highest priority goal;
- Action = Resource allocation policy;
- Initial action is randomly selection;
- Actions are assessed in a reinforcement learning loop;
- Reinforcement learning is based on a reward function.

Reward Calculation

$$\text{Reward} = W_0 \times R_0 + W_1 \times R_1 + W_2 \times R_2 + \dots + W_n \times R_n$$

E.g. with two goals for power and performance:

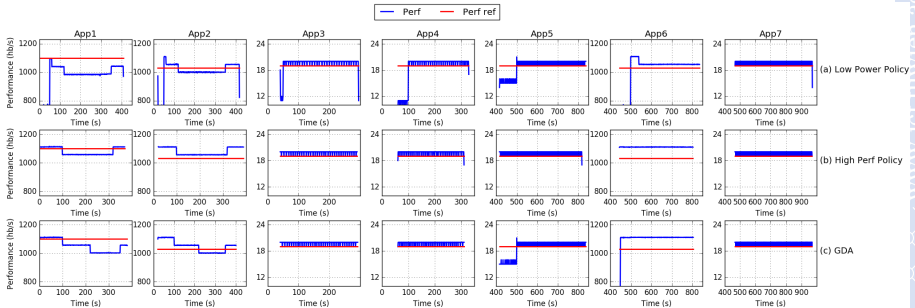
$$\text{Reward} = W_{\text{Power}} \times R_{\text{Power}} + W_{\text{Perf}} \times R_{\text{Perf}}$$

$$R_{\text{Power}} = \frac{P_{\text{ref}} - P_{\text{curr}}}{P_{\text{ref}}}$$

$$R_{\text{Perf}} = \frac{1}{n} \sum_{i=1}^n \frac{\text{Perf}_i - \text{Perf}_{\min}}{\text{Perf}_{\max} - \text{Perf}_{\min}}$$

Perf_i the measured performance of the i_{th} application
 $\text{Perf}_{\min}, \text{Perf}_{\max}$ minimum and maximum required performance
 n the total number of applications running
 W_i assigned by the priority re-assigner.

Experiments



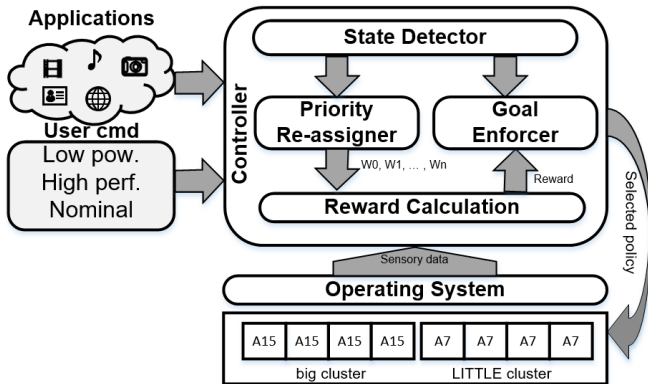
Experiments with a set of microkernel benchmarks;
Hardkernel Odroid XU3 board,
with two clusters (4 big (A15) and 4 little (A7) CPU cores);
Performance in heartbeats/sec.

Comparison

Tech.	Obj	Cmd	Pwr viol.	Perf. viol.	Avg. pwr
LP policy	Power	X	3%	65%	2.99
HP policy	Perf.	X	67%	0%	3.8
GDA	Dynamic	✓	20%	34%	3.2



Goal Driven Autonomy



Elham Shamsa et al. "Goal-Driven Autonomy for Efficient On-chip Resource Management: Transforming Objectives to Goals". In: *Proceedings of the Design and Test Europe Conference (DATE)*. Florence, Italy, Mar. 2019

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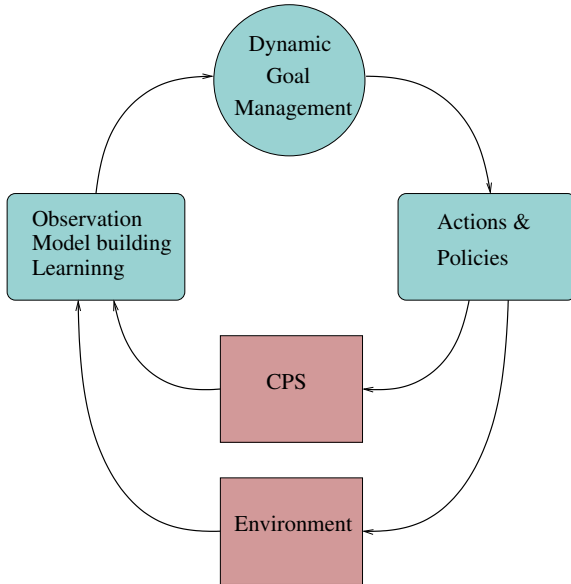
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Self-Aware Control Loop



Let's Get Out



PROACTIVE COMPUTING

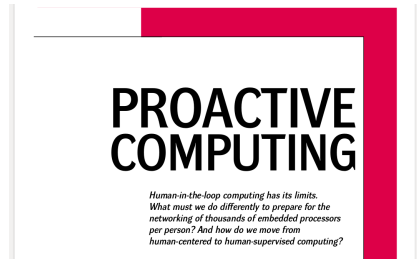
*Human-in-the-loop computing has its limits.
What must we do differently to prepare for the
networking of thousands of embedded processors
per person? And how do we move from
human-centered to human-supervised computing?*

David Tennenhouse. "Proactive Computing". In:
Communications of the ACM 43.5 (May 2000), pp. 43–50



Let's Get Out

- Let's get physical

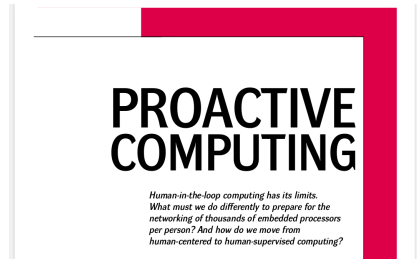


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Let's Get Out

- Let's get physical
- Let's get real



David Tennenhouse. "Proactive Computing". In: *Communications of the ACM* 43.5 (May 2000), pp. 43–50



Let's Get Out

- Let's get physical
- Let's get real
- Let's get out



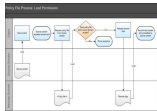
PROACTIVE COMPUTING

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Traditional Design Flow



Requirements specification



Design



Verification

Manufacturing



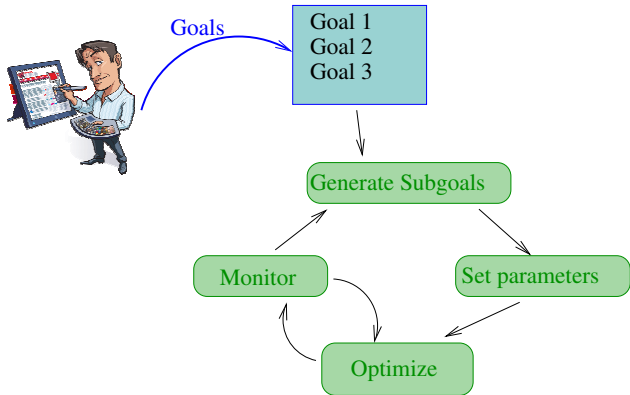
Validation



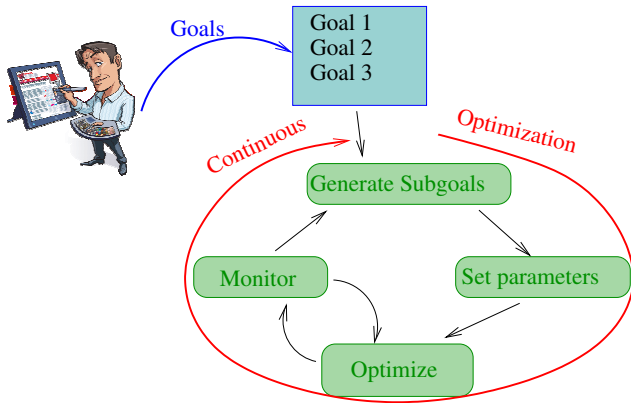
Implementation



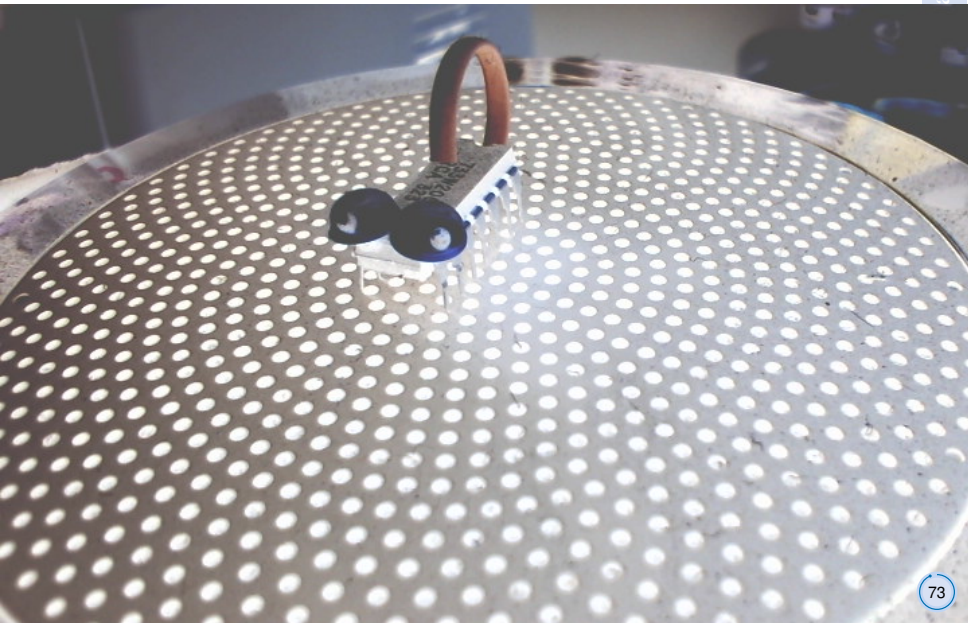
Design of Self-Aware Chips



Design of Self-Aware Chips



Questions ?



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Nikil Dutt, Axel Jantsch, and Santanu Sarma. “Towards Smart Embedded Systems: A Self-Aware System-on-Chip Perspective”. In: *ACM Transactions on Embedded Computing Systems, Special Issue on Innovative Design Methods for Smart Embedded Systems* 15.2 (Feb. 2016). invited, pp. 22–27.



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References V



Santanu Sarma et al. "CyberPhysical-System-On-Chip (CPSoC): A Self-Aware MPSoC Paradigm with Cross-Layer Virtual Sensing and Actuation". In: *Proceedings of the Design, Automation and Test in Europe Conference and Exhibition (DATE)*. Grenoble, France, Mar. 2015.



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References VI



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