

Modeling and Simulation of Many Core Architectures

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Motivation

"If you can not measure it, you can not improve it."

Lord Kelvin

"Nothing can be more fatal to progress than a too confident reliance on mathematical symbols; for the student is only too apt to take the easier course, and consider the formula not the fact as the physical reality."

Lord Kelvin

Many Core Simulation Group@GT

■ Faculty

- Tom Conte (SCS)
- S. Mukhopadhyay (ECE)
- George Riley (ECE)
- S. Yalamanchili (ECE)

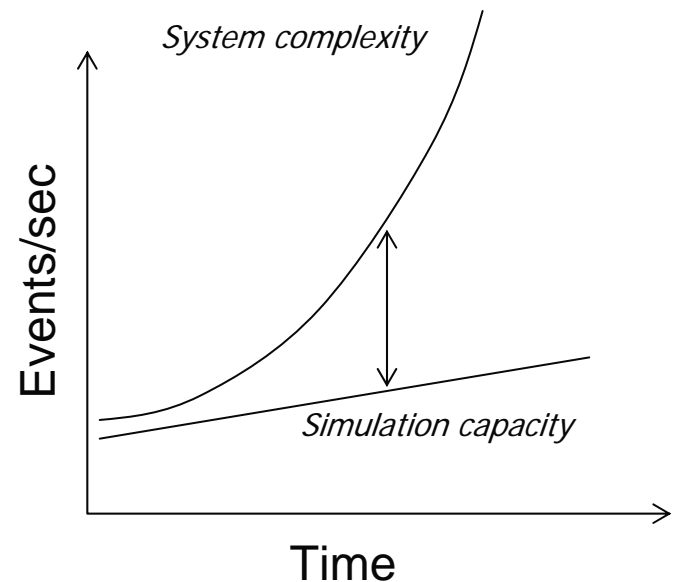
■ Graduate Students

- Paul Bryan
- Gregory Damos
- Brian Hayes
- Chad Kersey
- Minki Lee
- Elizabeth Lynch
- Nikil Sathe

Joint Effort between CERCS and new Center for Manycore Computing

The State of the Practice

- System complexity is outpacing simulation capacity
 - Cannot perform analysis at scale
- Islands of simulators and simulation systems
 - Customized interactions
 - Little leverage of individual investments
- The problem will get worse faster



Simulation Wall!

Prioritized Major Challenges*

1. Cost of building a validated useful simulator
 - Composable
 - New methodologies for building simulators
2. Accuracy
 - Need for calibrated models
 - Methodologies for constructing calibrated models
3. Performance
 - Parallelism, multiscale, and hardware acceleration
4. Power and thermal models
5. Ease of use: Productivity and Management Tools
 - Visualization, deployment, debugging, etc.
 - Documentation & deployability

*From Outbrief: *Performance Prediction and Simulation for Exascale Interconnection Networks*, Interconnect Workshop, DoE Institute for Advanced Architectures, July 2008

Some things to Keep in Mind

■ Research

- Requirements are changing, unknown, or speculative
- Modeling what does not exist at the Exascale
- Confidence levels

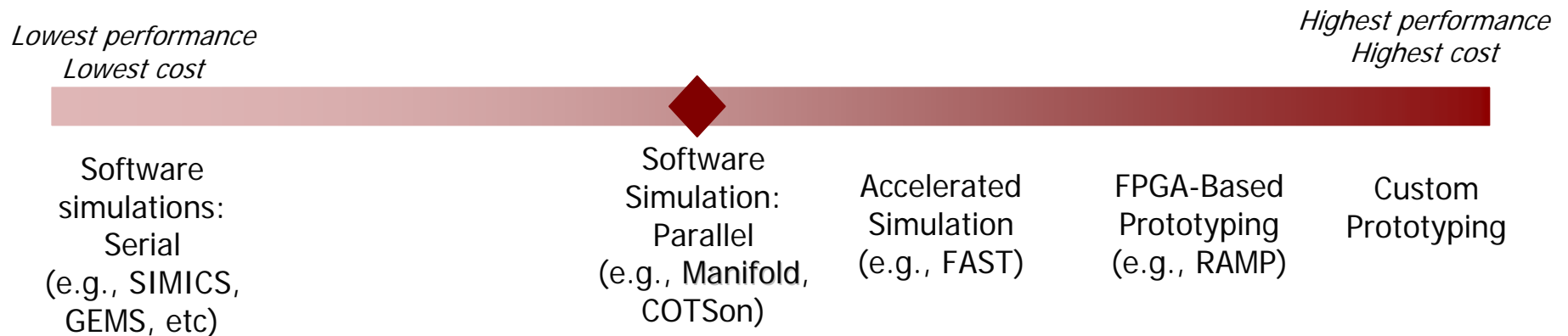
■ Education

- Lack of discipline-oriented courses
- Need more rigor in education for architecture/system modeling and simulation
- Knowledge of the third kind

Key Challenges

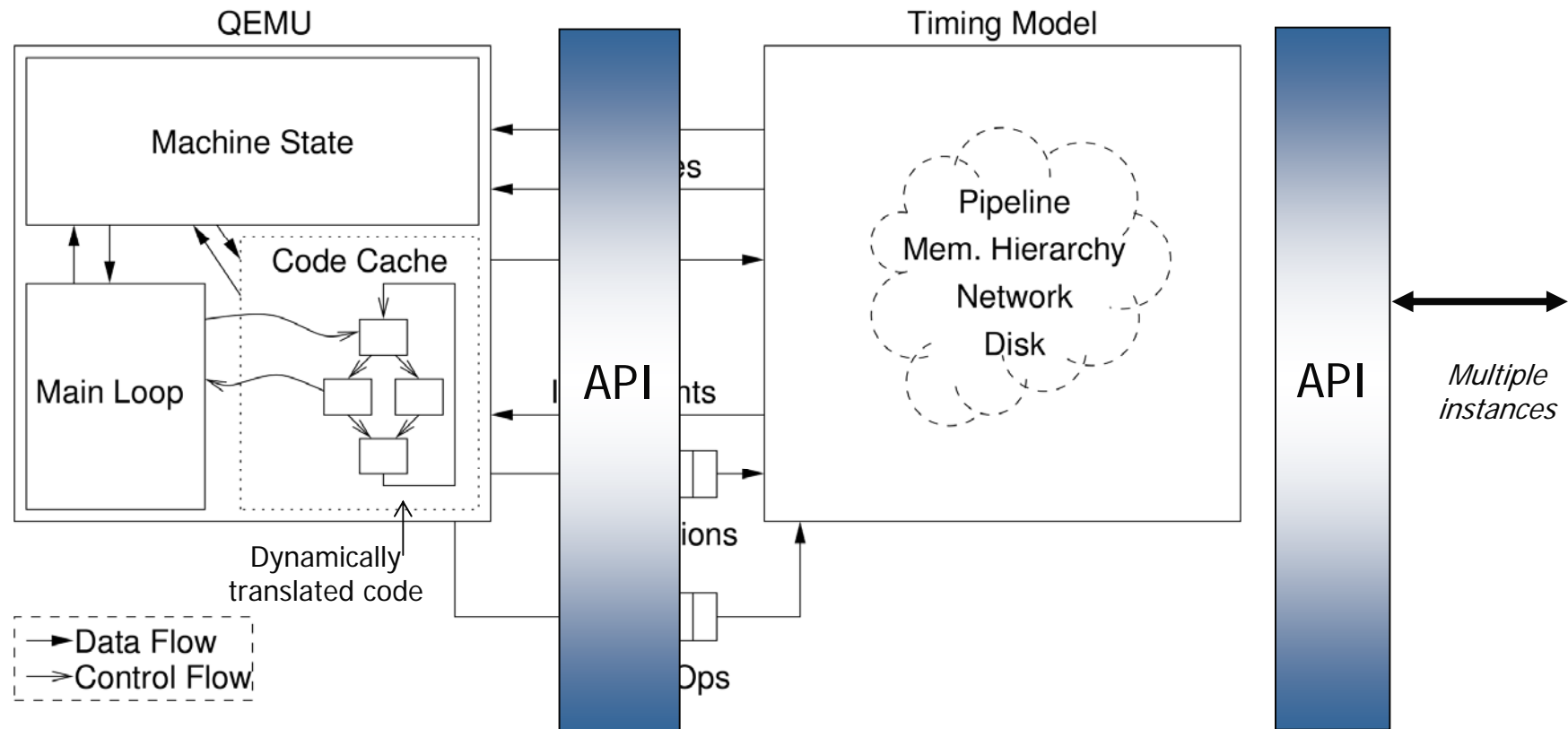
- Managing complexity (items 3 and 5)
 - Parallelism, sampling, acceleration
- Multi-model simulations (items 2 and 4)
 - Power and thermal challenges
 - Feedback between thermal and discrete event simulation
- Productivity (items 1 and 5)
 - Cost of simulator construction
 - Ease of use

Spectrum of Solutions



- Simple Premise: Use parallel machines to simulate/emulate parallel machines
- Leverage mature point tools via standardized API for common services
 - Event management, time management, synchronization
- Cull the design space prior to committing to hardware prototyping or hardware acceleration strategies

Managing Complexity

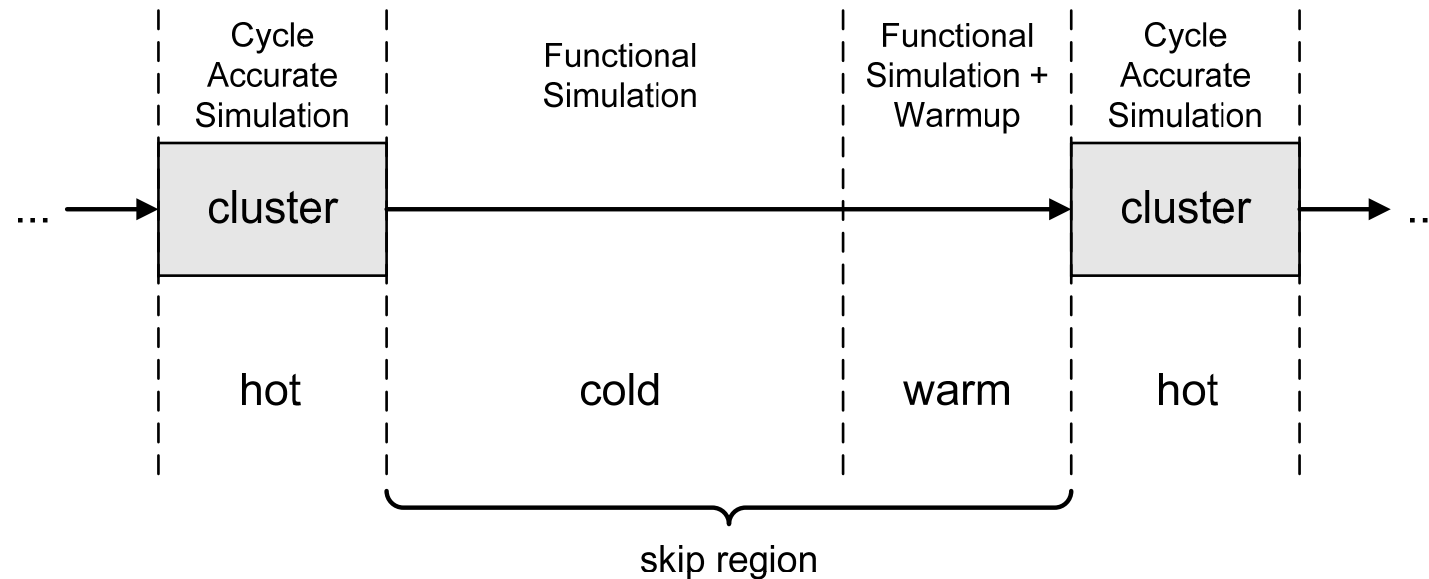


- Coupling timing models and functional models
- Timing models can be 4-5 orders of magnitude slower than real time

Solution Techniques

- Statistical Techniques
- Parallel Simulation
- Acceleration
 - FPGAs and more recently GPUs
- Regression and analytic models for design space exploration
 - For example, work of Lee & Brooks@Harvard

Cluster Sampling for Processor Simulation



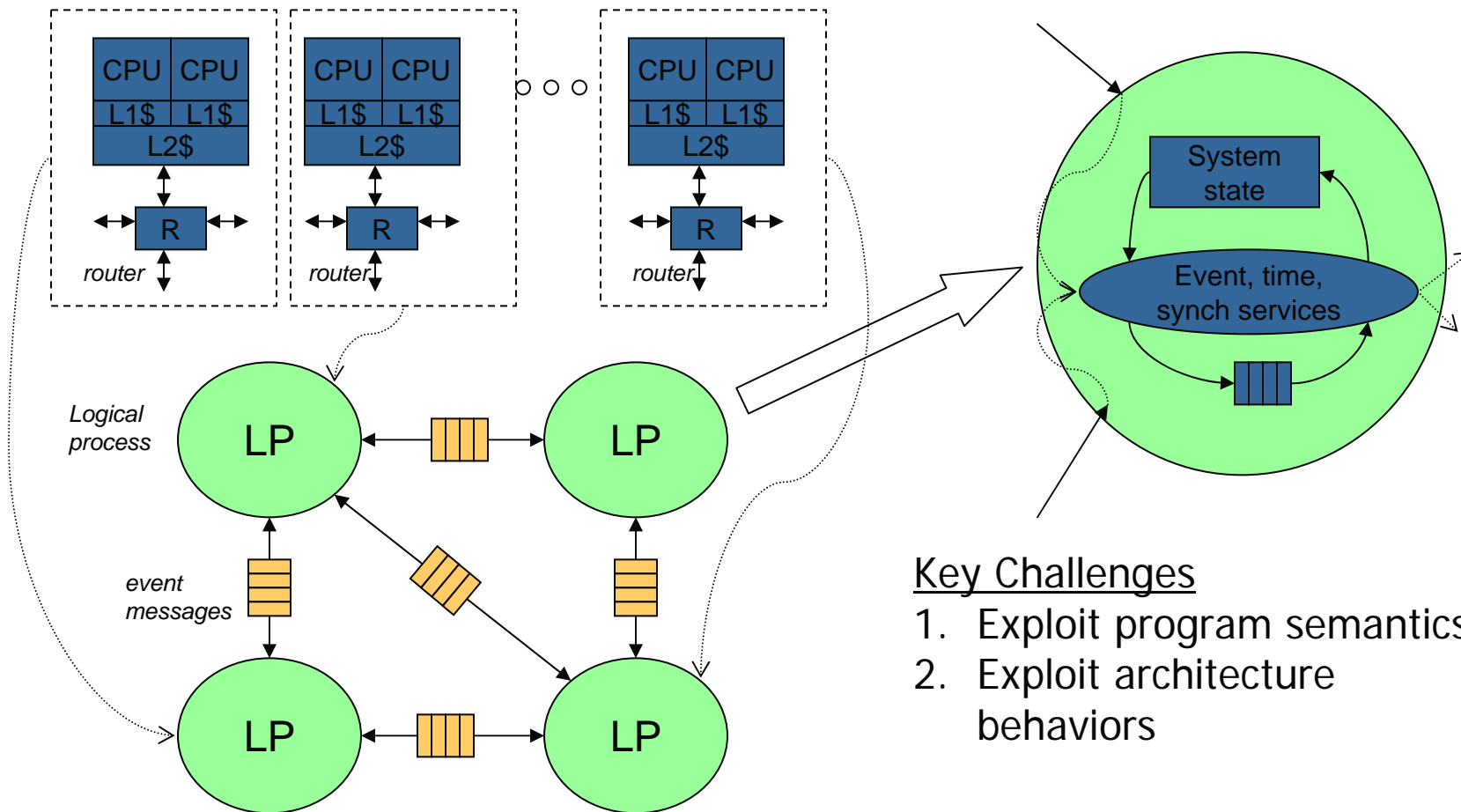
- Hot: simulator state is known
 - Measurements are accurate
- Warm: transition between cold to warm
 - Used to get the simulator to a known state
- Cold: simulator state is unknown
 - Measurements would not be accurate

Challenge: Extensions to multithreaded/parallel codes

Courtesy: Paul Bryan & Tom Conte

Coarse Grain Parallel Simulation

Example Modeled System



Key Challenges

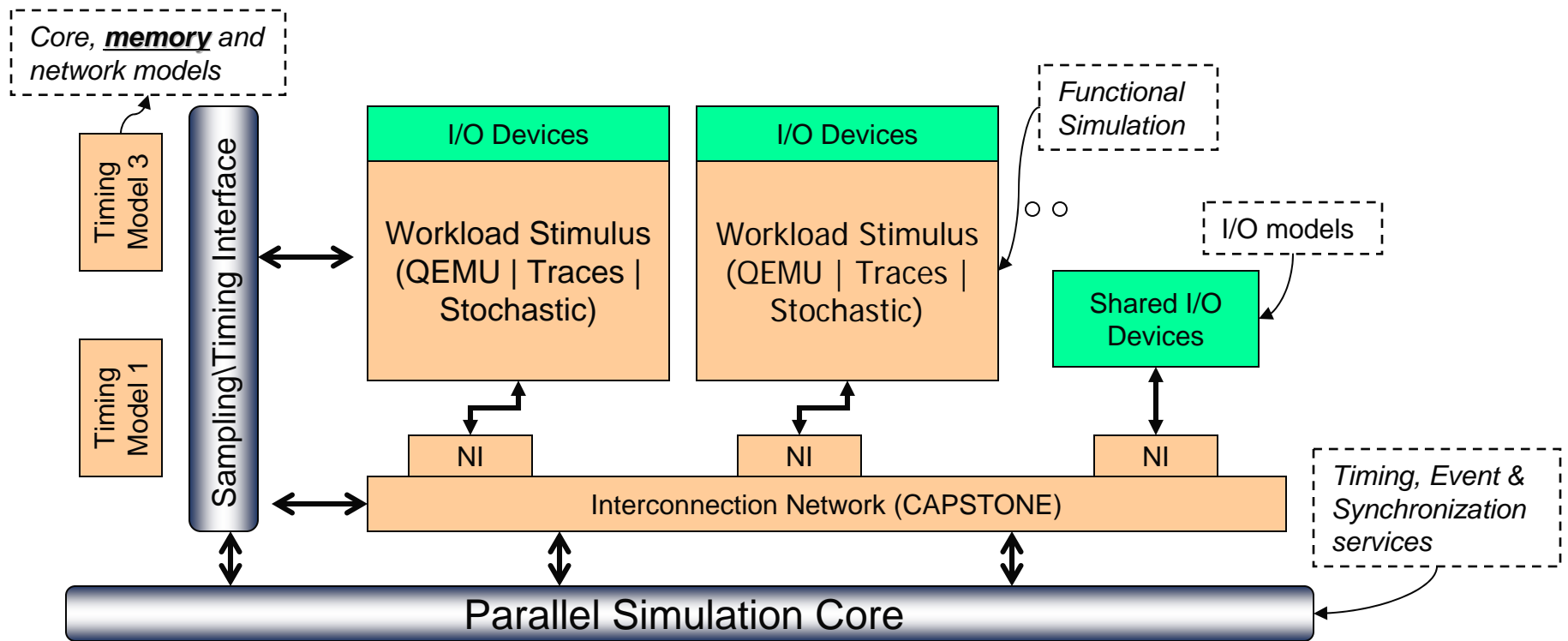
1. Exploit program semantics
2. Exploit architecture behaviors

1. D. Burger and D. Wood, "Accuracy vs. Performance in Parallel Simulation of Interconnection Networks", *ICPP* 1995
2. A. Falcon P. Faraboschi D. Ortega, "An Adaptive Synchronization Technique for Parallel Simulation of Networked Clusters," *ISPASS*, 2008
3. Parallel SST, SNL

Some Simple Goals

- Get to 2-3 orders of magnitude slower than real time for timing simulations
 - Use Petaflop machines to simulate Exascale Machines?
- Consider hardware support for global virtual time
 - Lynch & Riley (ongoing work)
 - Support in the NICs?
 - Hardware barrier synchronization support from the 90's
 - Hardware, fine-grained all-to-all support

Manifold: Overview



Sponsor: Sandia National Laboratories

What Can Be Done?

- Provide an infrastructure to integrate mature point tools
 - Standardized API for time, event, synchronization, and management services
 - Support both time stepped and discrete event simulation
 - Central role for rigorous statistical methods
 - Near Term integration of QEMU with HP Labs COTSon
- Scale-up: Track Moore's Law for Simulation Capacity?
 - Double simulation capacity every 12-18 months
 - High level composition of detailed models: on-chip and off-chip
 - Enable migration of models across hardware and software simulation platforms

Major Challenges

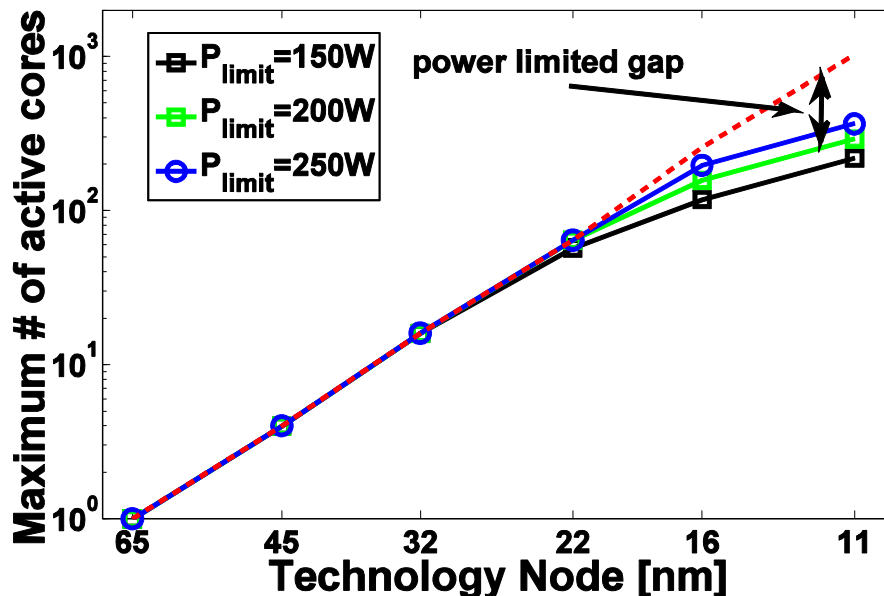
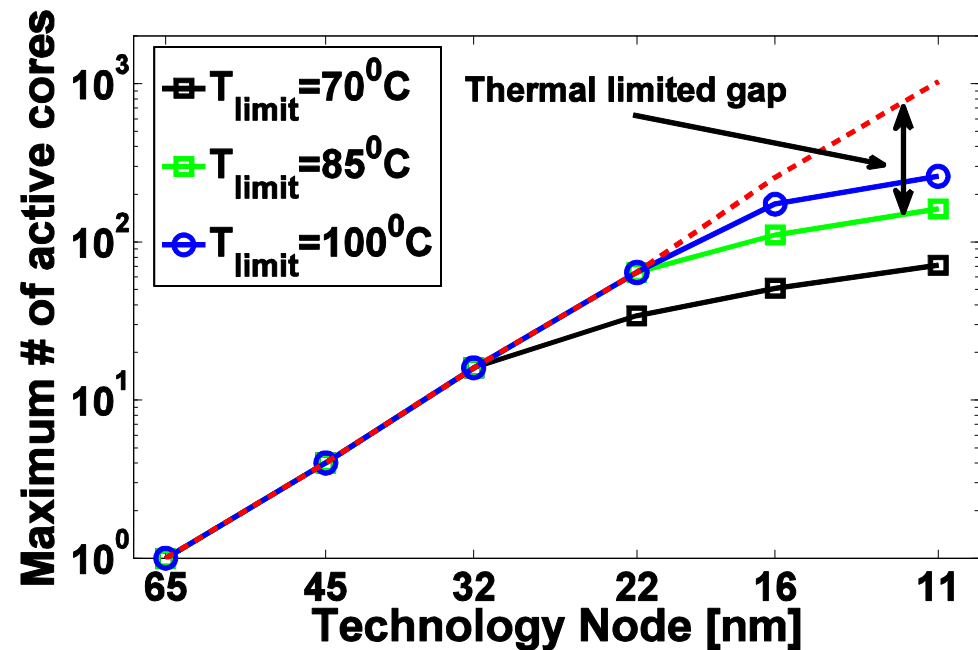
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Power and Thermal Modeling

- Seek a fundamental understanding of energy and thermal challenges at Exascale
- Develop, model, and assess (new) architectural principles for energy management
 - Architectural techniques for energy management
 - Need to couple physics of heat management with detailed architecture simulation

Thermal and Power Scaling Limits

Temperature Limited Performance

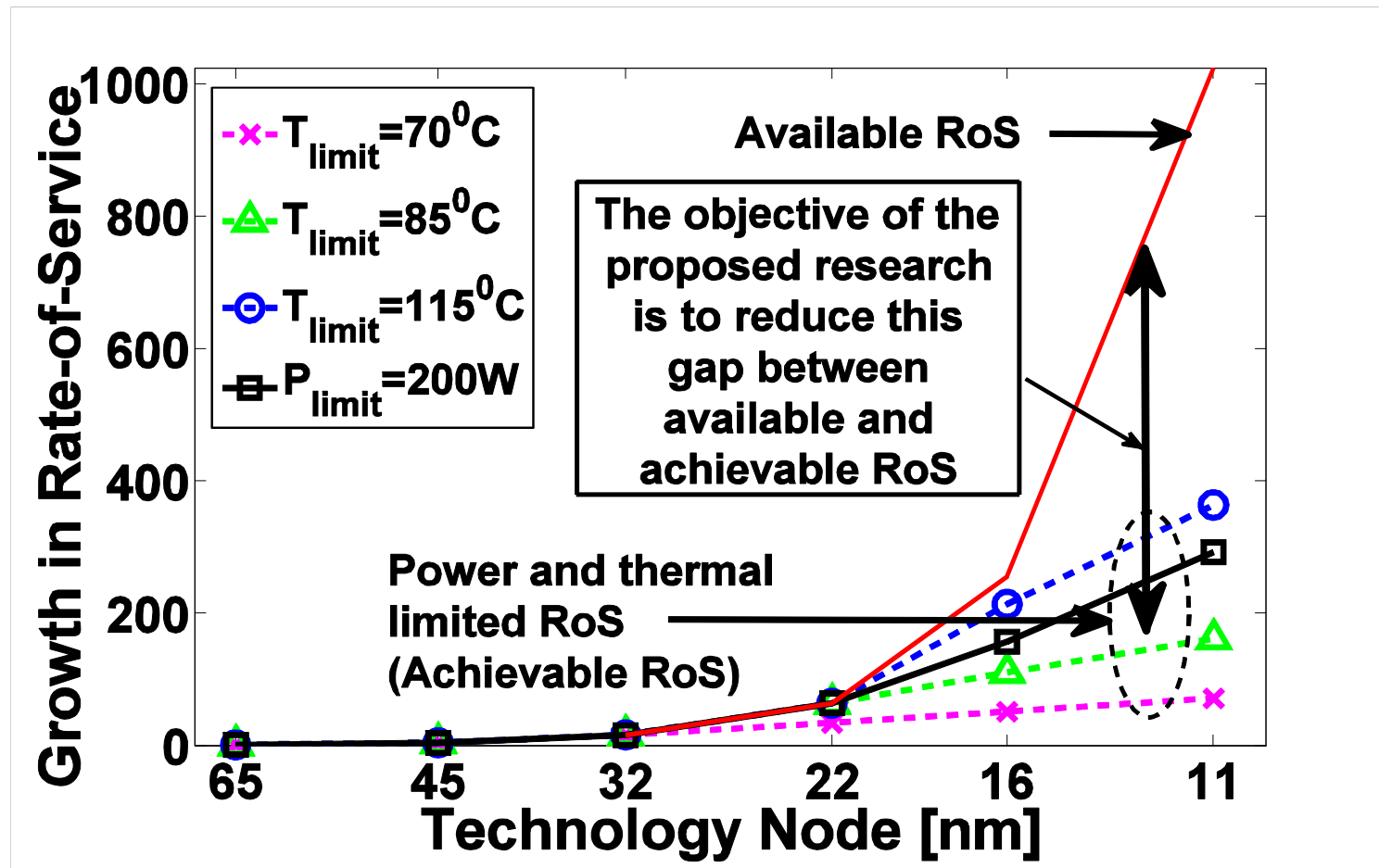


Power Limited Performance



- Hot Spot (UVa)
- IntSim (J. Meindl's group@GT)

Feasibility vs. Capacity Gap



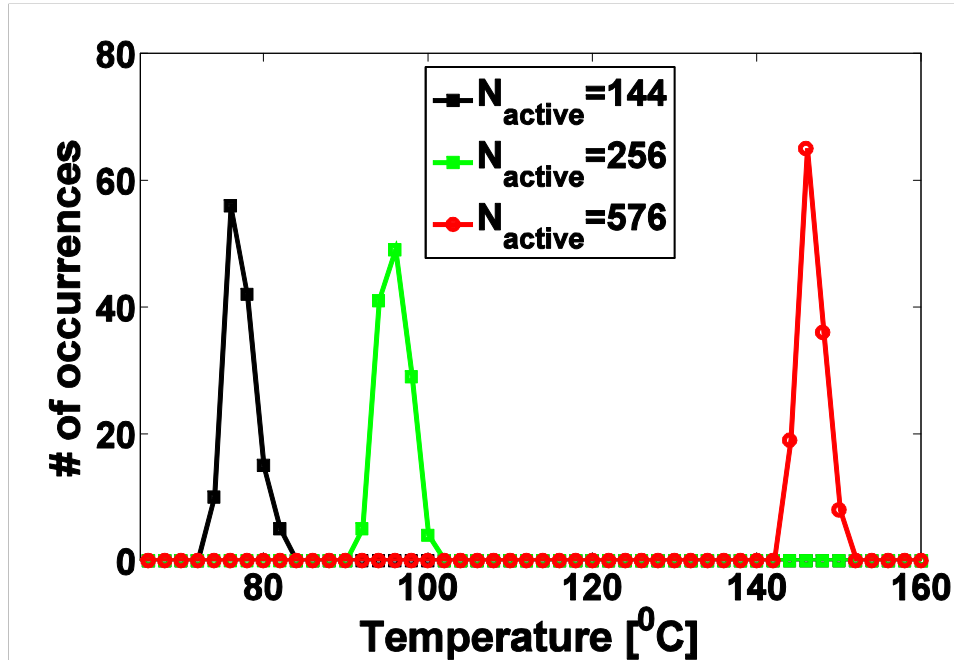
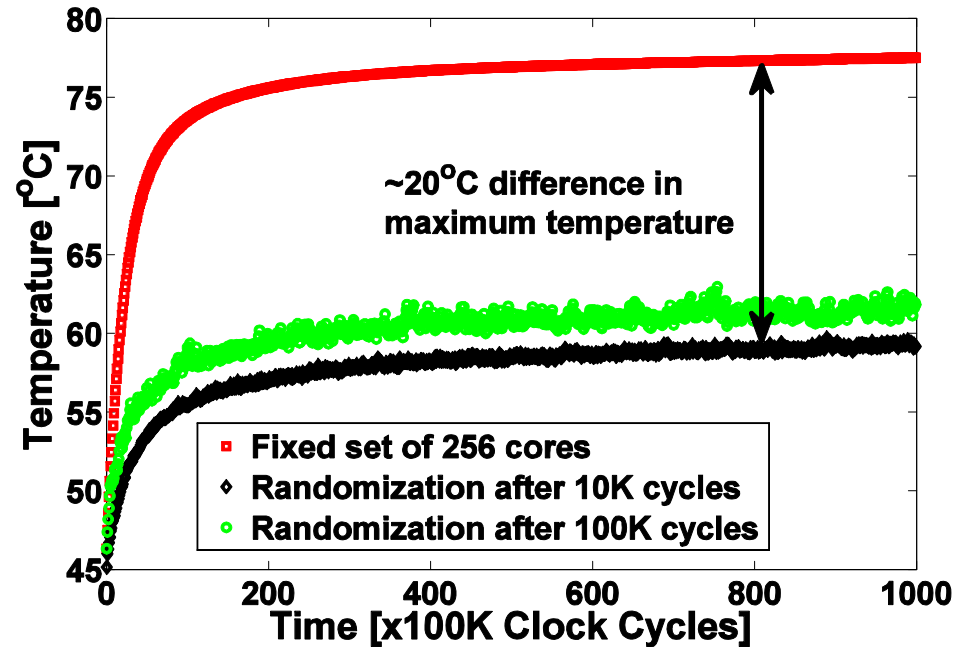
Mukhopadhyay and Yalamanchili

Scaling Principles

- Dynamic core scaling
 - Analog of traditional voltage-frequency scaling
- Spatial scaling
 - Metrics for thermal proximity and thermal compactness for heat management
 - Exploit the physics!
- What do you manage architecturally?
 - Gradients vs. peak temperature

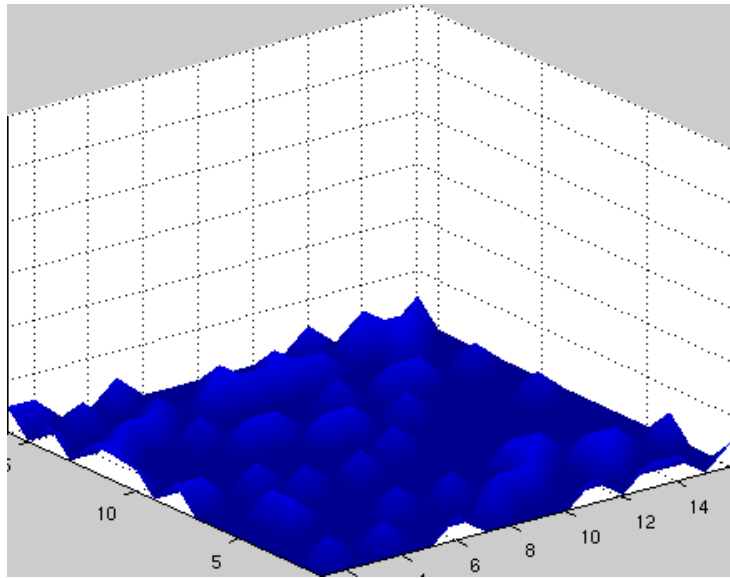
Effects of Spatiotemporal Scaling

Randomized migration



Impact of spatial scaling

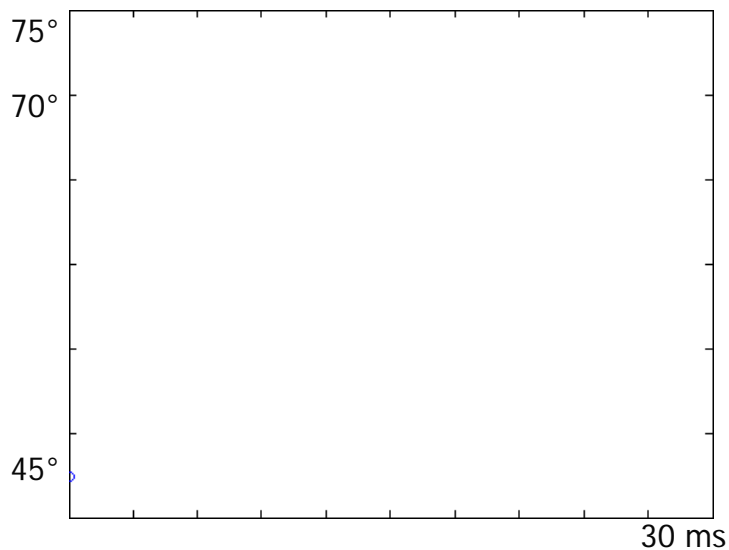
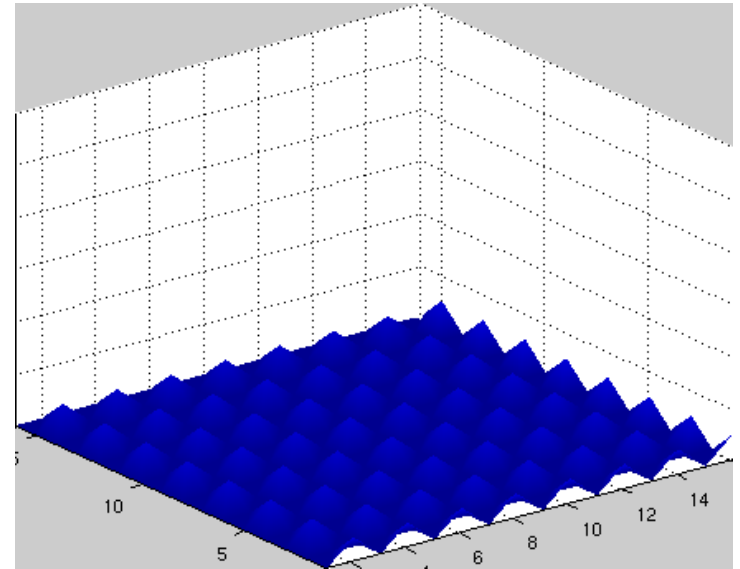
Random Pattern of On-Tiles



← Spatial gradient:
10.5°@0.75mm
Temporal gradient:
2.5°@100Kcycles

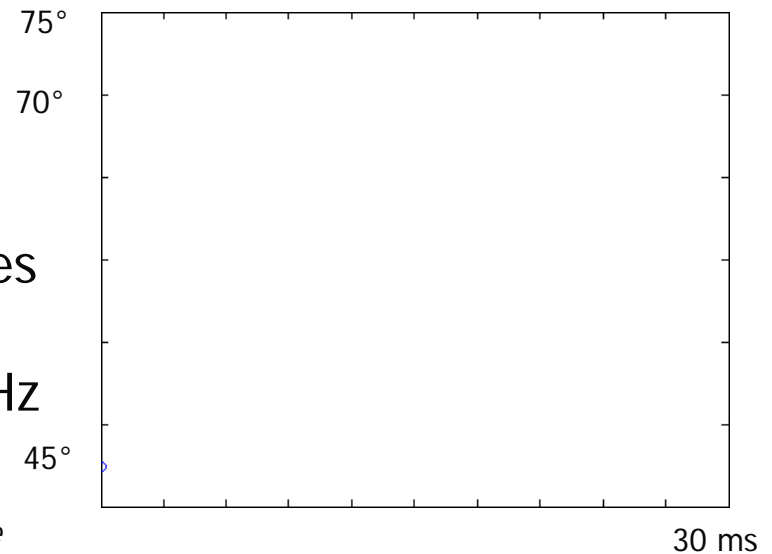
Spatial gradient:
2.5°@0.75mm →
Temporal gradient:
1.99°@100Kcycles

Cyclic Pattern of On-Tiles



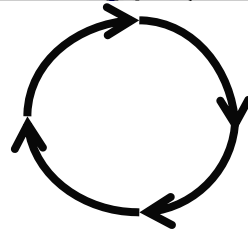
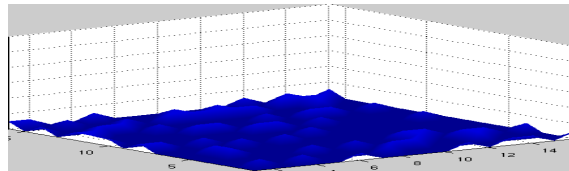
- 64 on-tiles
- 256 total tiles
- 100K cycles interval@3GHz

Courtesy: Nikil Sathe



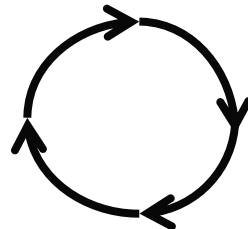
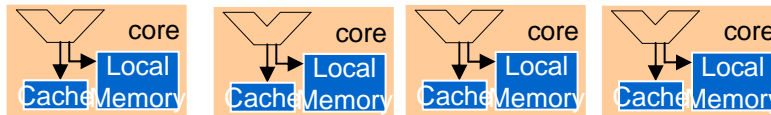
The Need for Feedback

Thermal profile



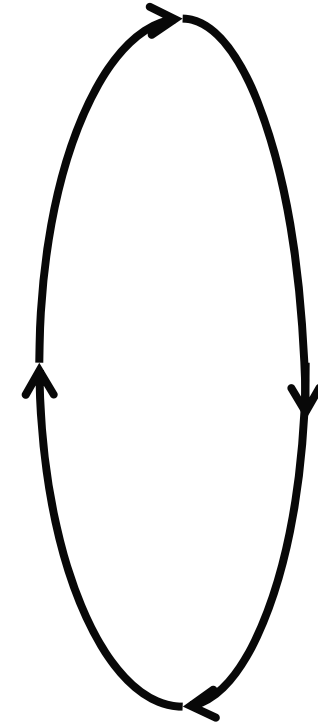
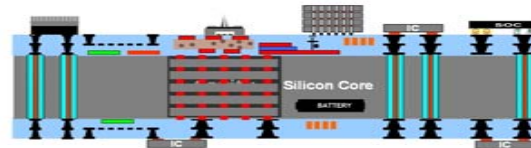
Co-exploration of thermal management/architecture management

Spatiotemporal migration



Co-design power distribution/architecture management

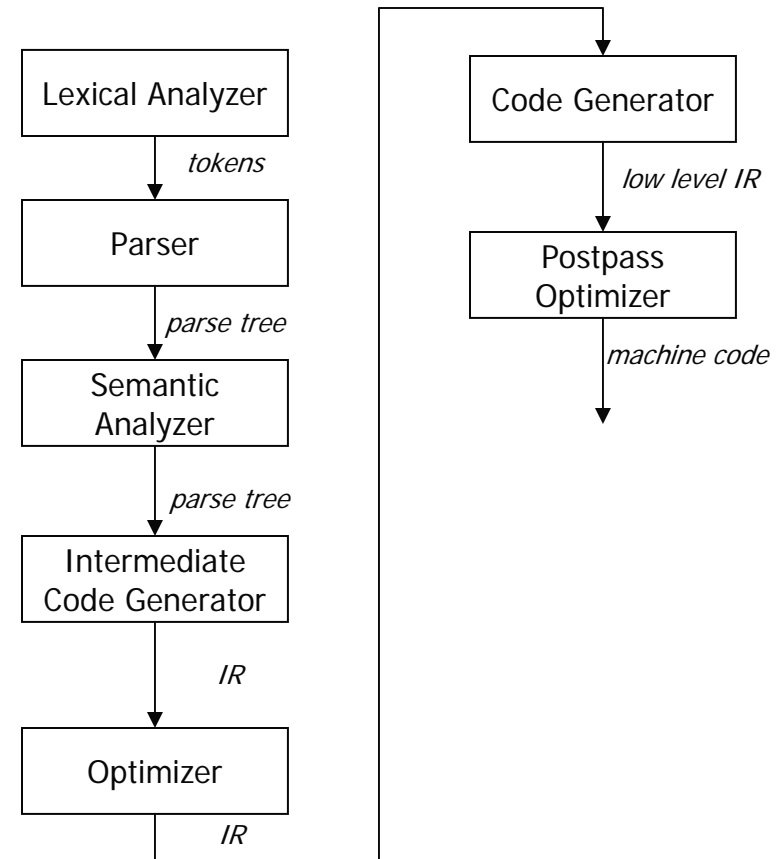
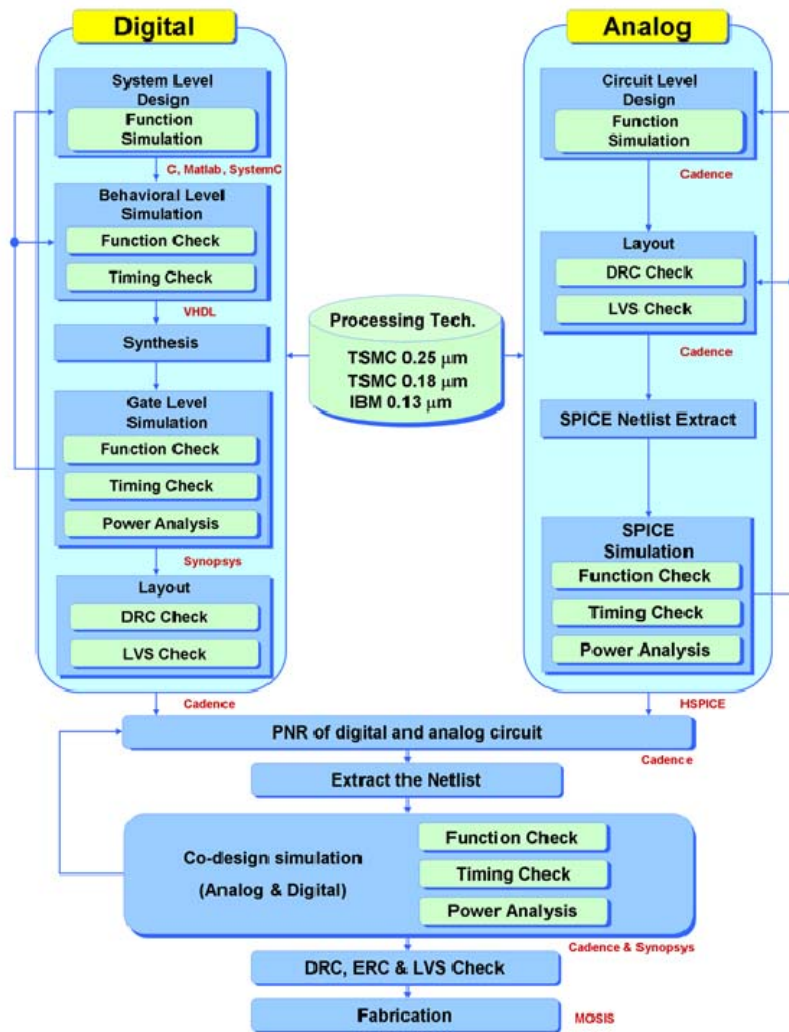
Power distribution network



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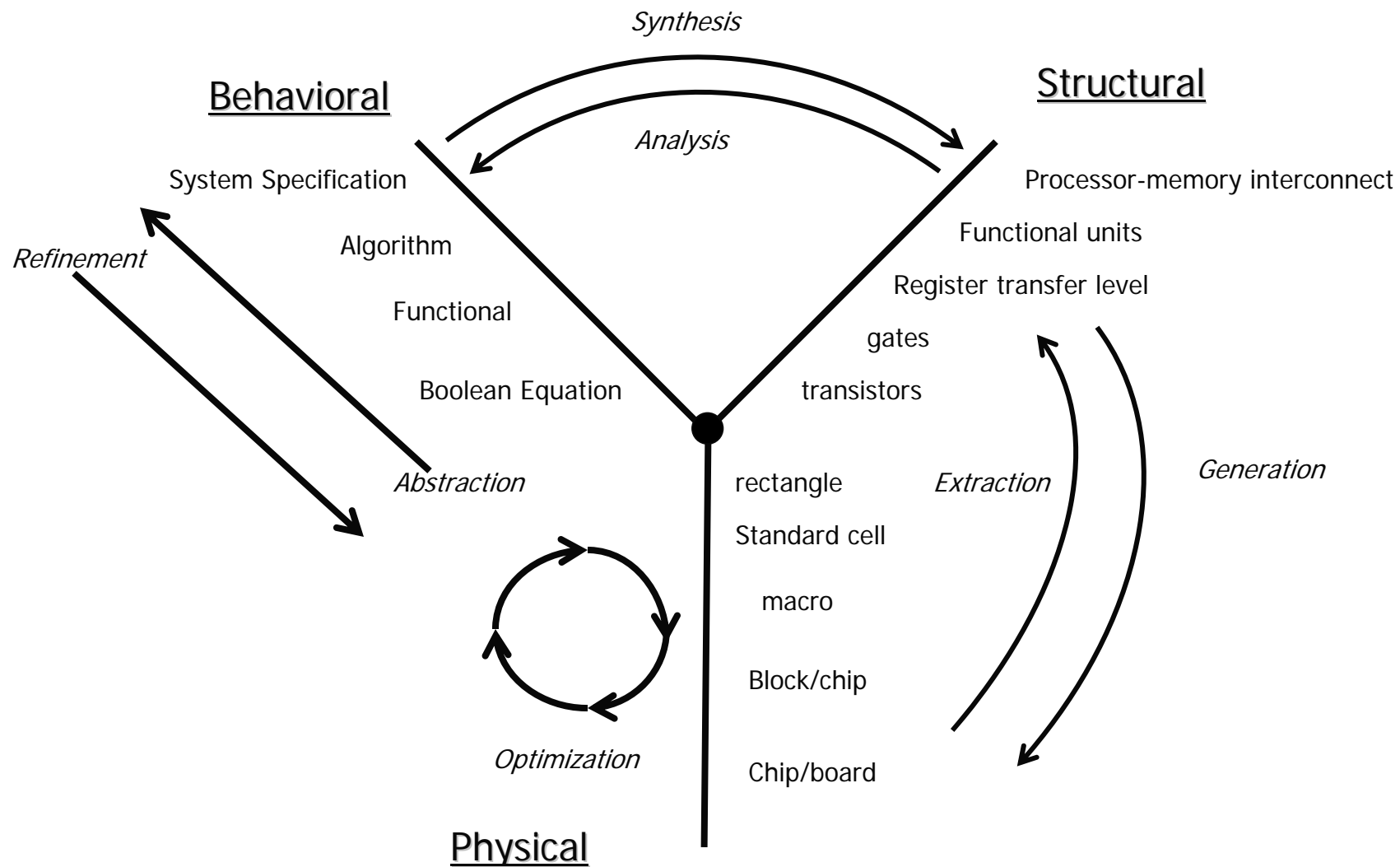
A Thought: Learn from Design Flows



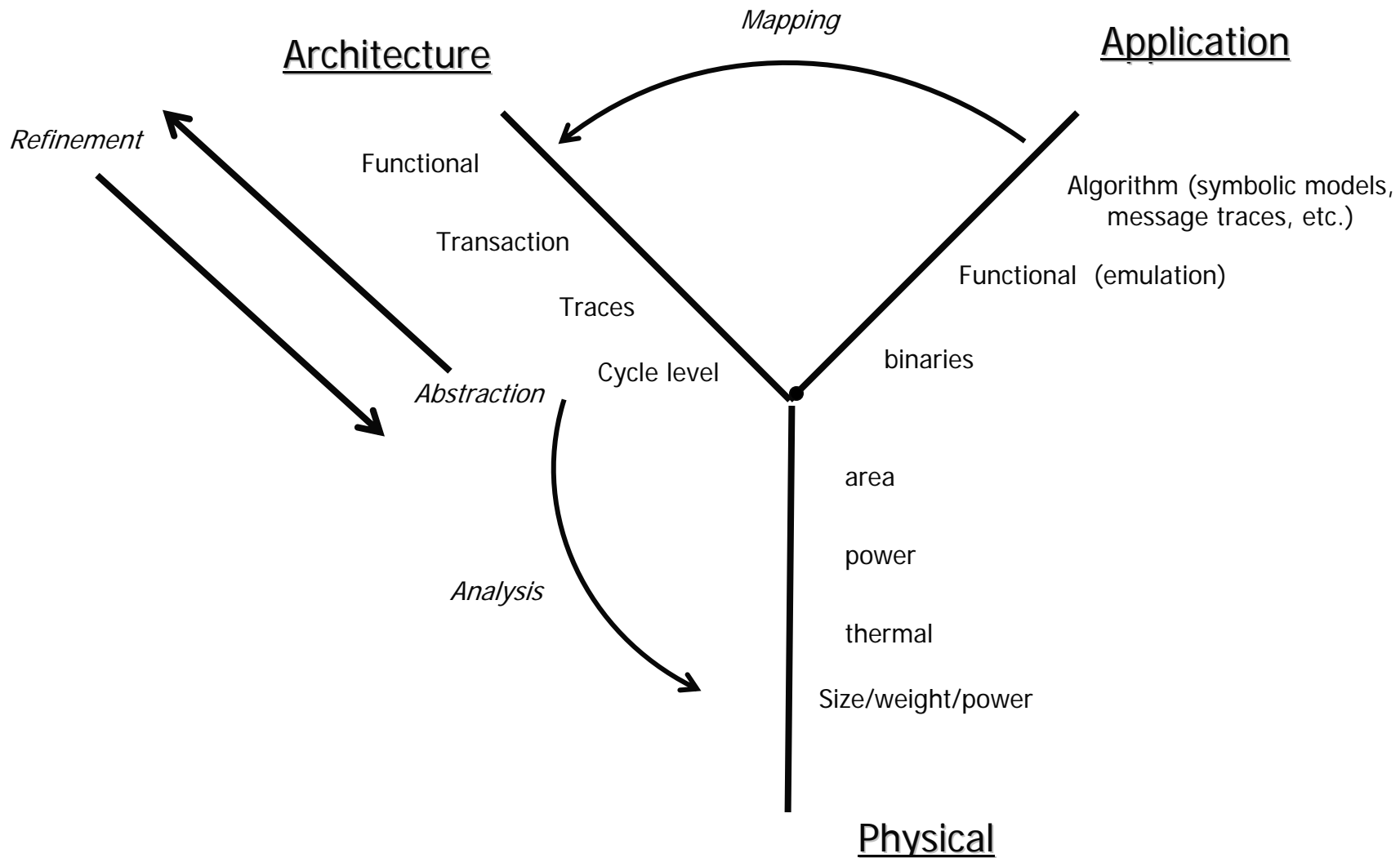
Implications

- Need a hierarchy of representations
 - Accompanied by successive refinement
- Some example simulation flow steps
 - Parsing a system description language
 - Component partitioning & assignment,
 - Design rule (model) check
- Need Structure
 - Who is the customer for these tools?

Example Y-Chart Based Design (Gajski, Kuhn)



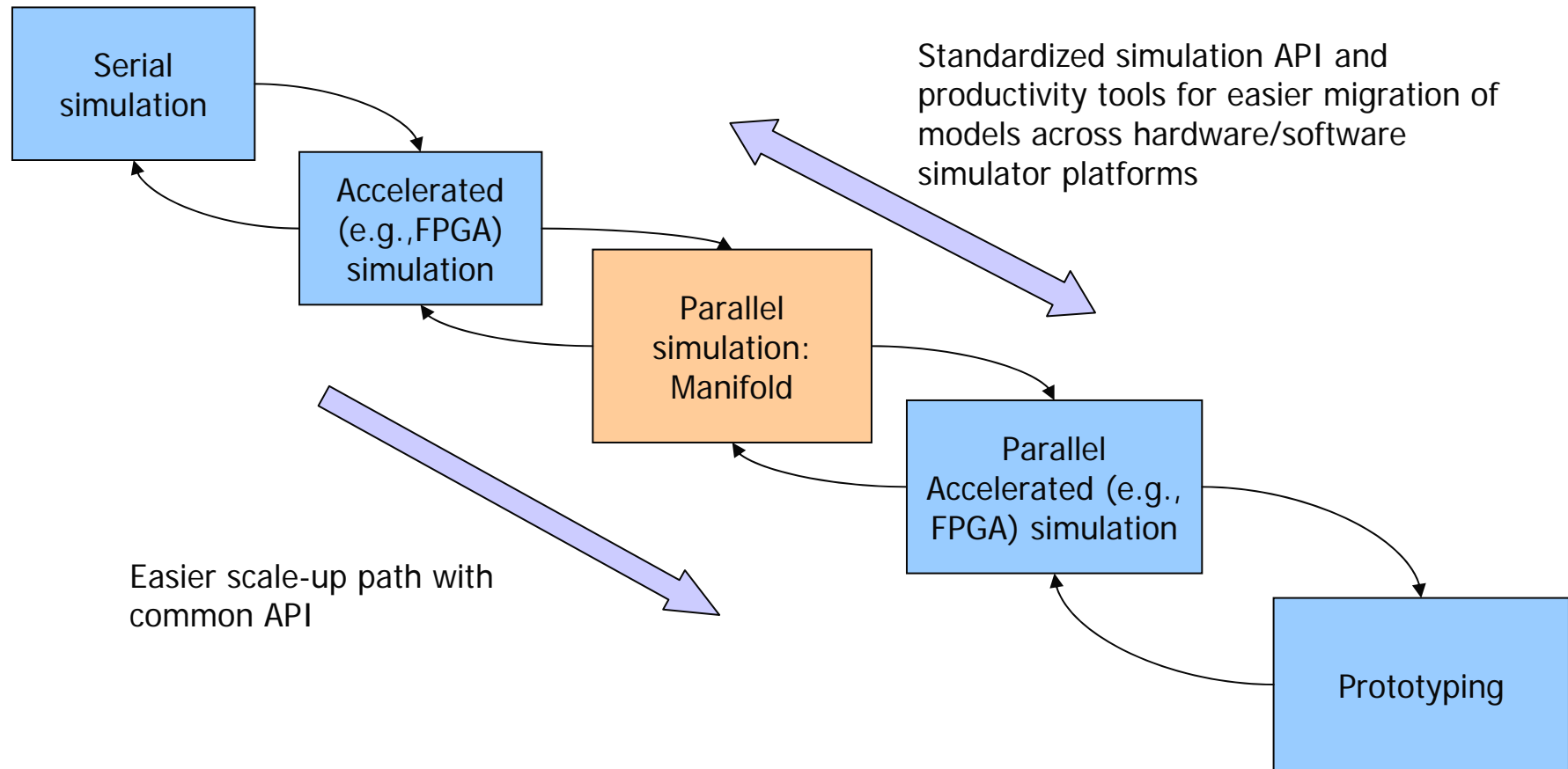
Example Y-Chart Based Design for Simulation



What Can Be Done?

- Create a simulation flow for the construction of models
 - Spiral or waterfall model of construction
 - Use architecture/system description languages
- Are we doomed to build what we can predict?
- Reporting Methodology
 - Challenging!
 - Publication of individual software and a centralized managed code base to support reproducibility

An Ecosystem for Many Core Simulation



From DoE IAA White Paper (SNL, ORNL, GTech, UMD, UT)

Summary

Coupling/Feedback
Across Modeling
Tools

Thank You

Productivity
Tools

Complexity
Management