



The Linux Real-Time Task Model Extractor

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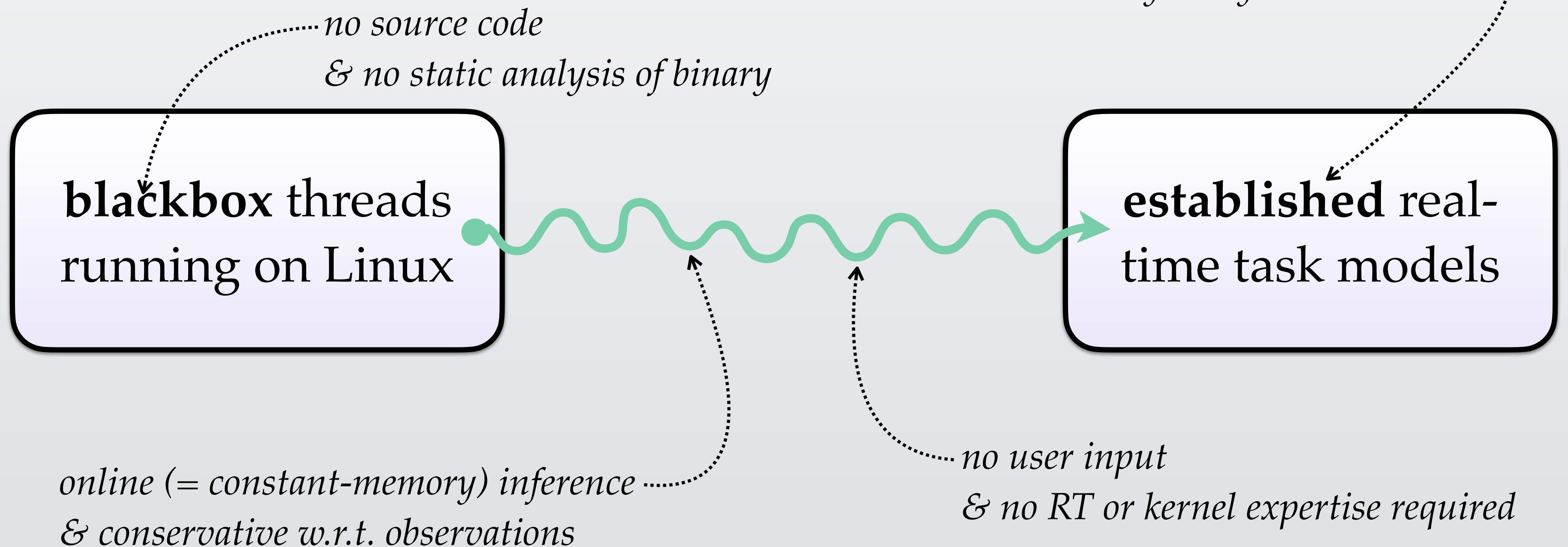
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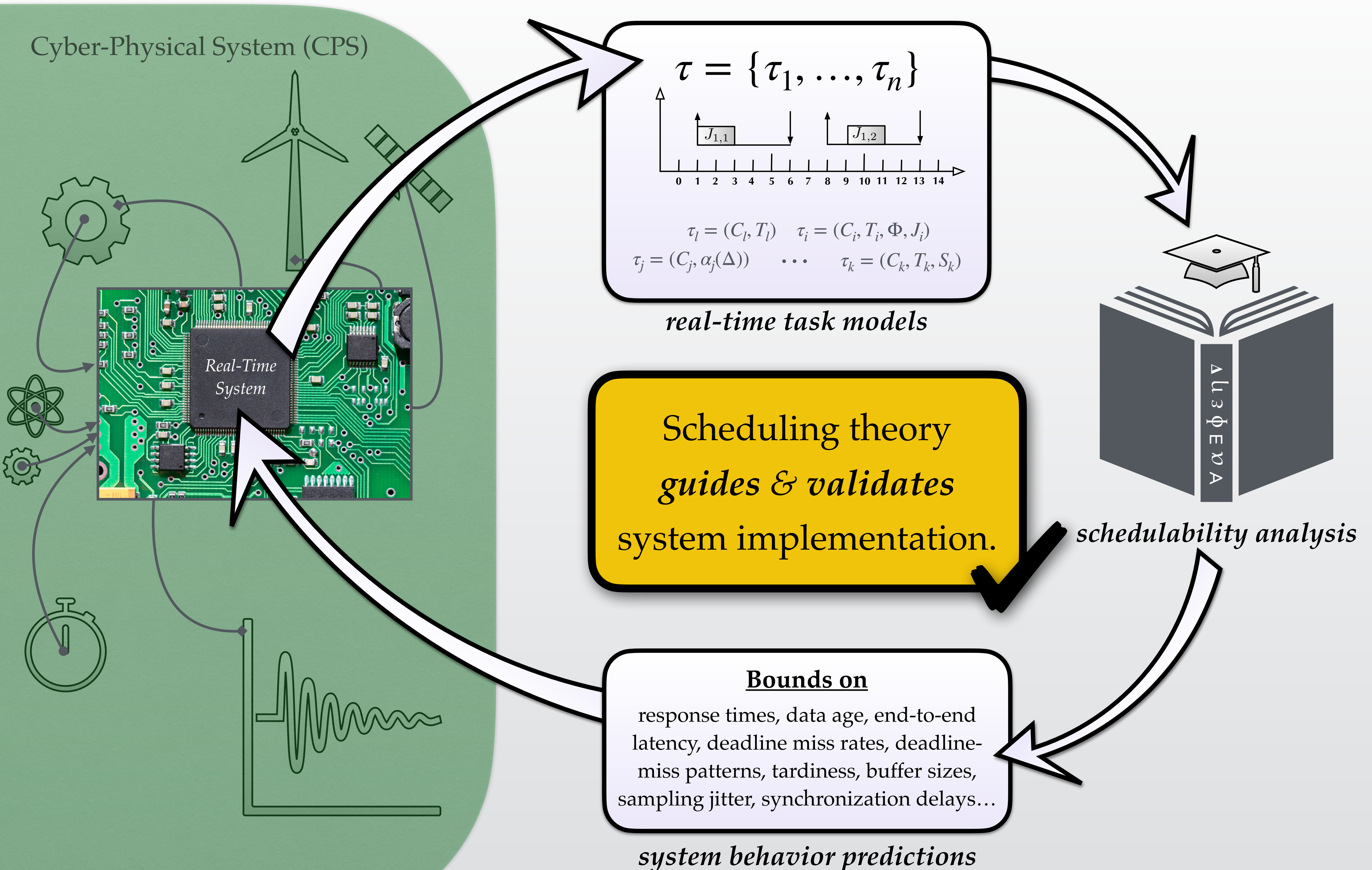
LiME in a nutshell:



The Problem Being Solved

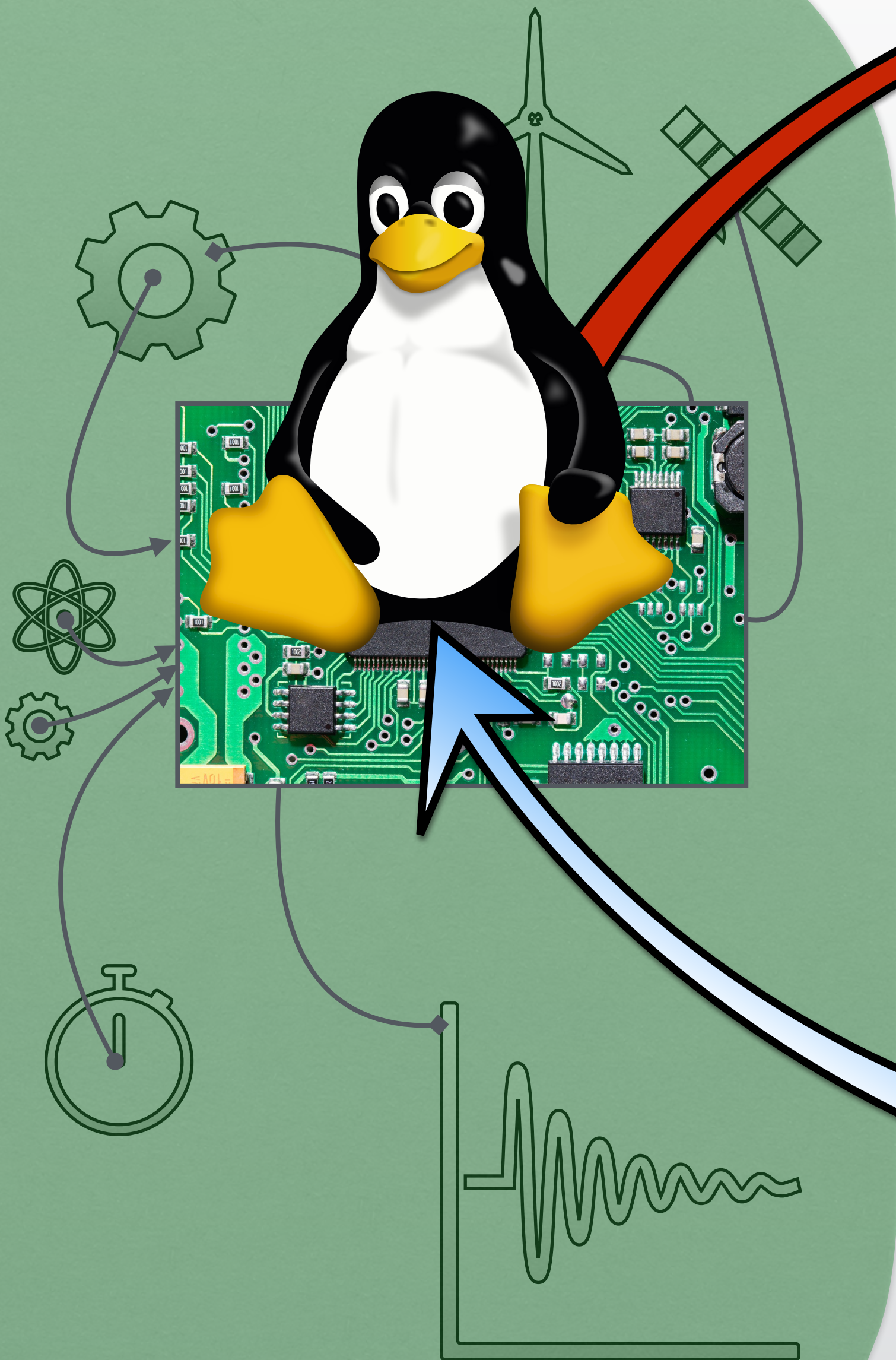


The *System-Model-Analysis-Prediction* Cycle

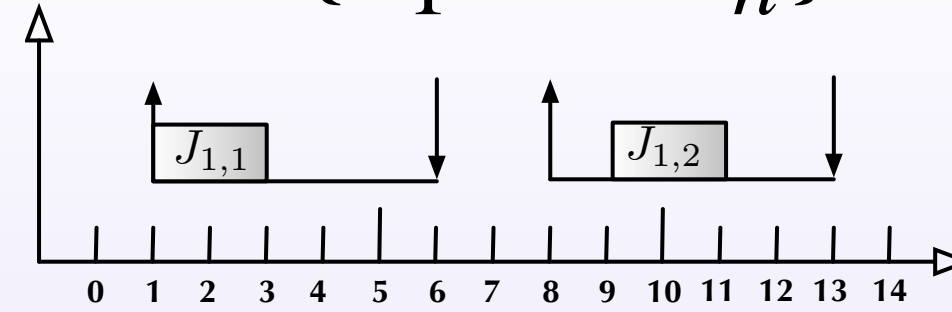


Linux has Become a Serious RT Platform

Cyber-Physical System (CPS)



$$\tau = \{\tau_1, \dots, \tau_n\}$$



$$\begin{aligned} \tau_l &= (C_l, T_l) & \tau_i &= (C_i, T_i, \Phi, J_i) \\ \tau_j &= (C_j, \alpha_j(\Delta)) & \dots & \tau_k &= (C_k, T_k, S_k) \end{aligned}$$

real-time task models

*But what if we're
using Linux...?*

schedulability analysis

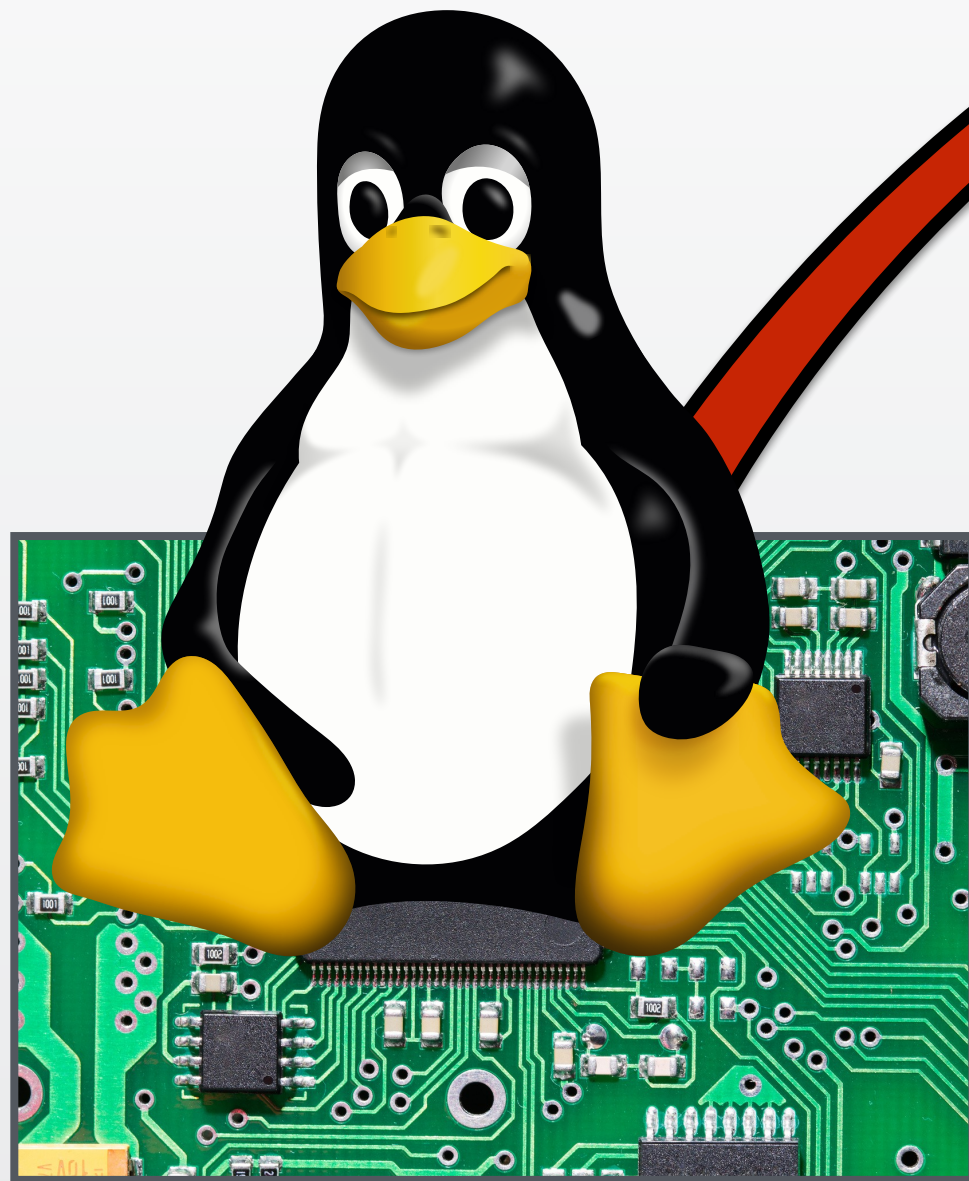
Bounds on

response times, data age, end-to-end
latency, deadline miss rates, deadline-
miss patterns, tardiness, buffer sizes,
sampling jitter, synchronization delays...

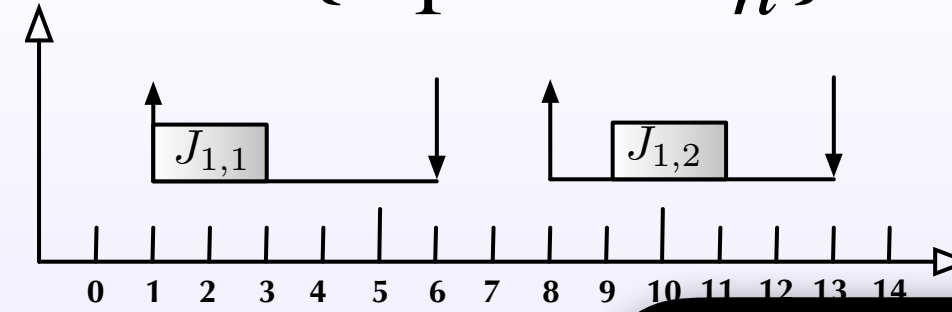
system behavior predictions



The “Dual Expertise” Barrier



$$\tau = \{\tau_1, \dots, \tau_n\}$$



$$\tau_l = (C_l, T_l) \quad \tau_i = (C_i, \alpha_i(\Delta))$$

real-time task

This Paper

Can we *fully automate*
Linux Model Extraction
and bridge this gap
once and for all?



State of the Art

→ *No tool support whatsoever!*

Doing it the hard way:

1. Become familiar with Linux's low-level tracing facilities (*eBPF*).
2. Become familiar with Linux system call semantics *below libc* level.
 - *Dozens of relevant syscalls and flags...*
3. Become familiar with applicable RT theory.
 - *In particular, identify suitable task models that fit Linux thread behavior!*
4. Build and **test** your own in-house tooling.
 - *Or a mess of buggy shell scripts...*



LiME in Action



Demo: Tracing *cyclictest*

```
$ sudo lime-rtw extract -o demo -- cyclictest -p fifo -D 10
```

```
defaulting realtime priority to 2  
# /dev/cpu_dma_latency set to 0us  
policy: fifo: loadavg: 0.08 0.02 0.00 1/214 567452
```

```
T: 0 (567452) P: 2 I:1000 C: 9991 Min: 18 Act: 54 Avg: 52 Max: 75
```

```
Results saved in demo.
```

Normal output of
cyclictest.

Online extraction of task models.
(Offline extraction also supported.)

JSON files containing
task model parameters.

Let's have a look with LiME's viewer...

lime-rtw view /tmp/demo

Tasks

Task ID	Policy	Prio	Comm	Comma
>> 494809-0	FIFO	2	cyclictest	cycli

Separators

>> clock_nanosleep (CLOCK_MONOTONIC, absolute) [9999 occurrences] [P]
suspension [9999 occurrences] [P]

Detected Tasks
Here, we launched only one thread
(→ one task).

Inferred Period
1000000 ns = 1 ms

Observed Max. Release Jitter
25610 ns ≈ 25.61 μs

Job Separator
By default, cyclictest uses
clock_nanosleep to enact periodic activations.

Separator Details

Separator: clock_nanosleep (CLOCK_MONOTONIC, absolute)

Count: 9999

Arrival Models:

Model #1: arrival_curve
Points: 31 (see tables for details)

Model #2: sporadic
Minimum Inter-arrival Time: 1000000 ns

Model #3: periodic
Period: 1000000 ns
Jitter: 25610 ns
Offset: 4226685616377033 ns
On: arrival

Help

Navigation: Left/Right - Switch panels | Up/Down - Select item | Enter - View details | q - Quit

Demo: Tracing *ping*

Can trace arbitrary (and non-RT) tasks.



```
$ sudo lime-rtw extract --best-effort -o demo-ping  
-- ping -c 1000 -i 0.1 contact.mpi-sws.org
```

```
PING contact.mpi-sws.org (139.19.171.199) 56(84) bytes of data.  
64 bytes from contact.mpi-sws.org (139.19.171.199): icmp_seq=1 ttl=59 time=1.27 ms  
64 bytes from swscontact.mpi-sws.org (139.19.171.199): icmp_seq=2 ttl=59 time=1.30 ms  
64 bytes from swscontact3.mpi-sws.org (139.19.171.199): icmp_seq=3 ttl=59 time=1.30 ms  
64 bytes from swscontact3.mpi-sws.org (139.19.171.199): icmp_seq=4 ttl=59 time=1.32 ms  
[...]  
64 bytes from swscontact3.mpi-sws.org (139.19.171.199): icmp_seq=1000 ttl=59 time=1.28 ms  
  
--- contact.mpi-sws.org ping statistics ---  
1000 packets transmitted, 1000 received, 0% packet loss, time 100508ms  
rtt min/avg/max/mdev = 1.248/1.309/2.106/0.058 ms
```

Results saved in demo-ping.


```
lime-rtw view /tmp/demo-ping
```

The image shows a terminal window with a task scheduler interface. The top panel, titled "Tasks", displays a table of tasks. The second task, with ID "494691-0", is selected. To the right, the "Separators" panel shows details for the "suspension" separator, including its count (3001) and arrival models. A yellow callout box points to the "suspension" separator in the table, stating that each thread wake-up is a new activation. A green callout box points to the "arrival_curve" model in the details, noting it is suitable for complex activation behavior. The bottom panel shows navigation instructions.

Task ID	Policy	Prio	Comm	Comm
>> 494691-0	CFS	-	ping	ping

Suspension Separator
Each thread wake-up is a new activation.

Release Curve
Suitable for complex activation behavior.

Separator Details
Separator: suspension
Count: 3001
Arrival Models:
Model #1: arrival_curve
Points: 31 (see tables for details)
Model #2: sporadic
Minimum Inter-arrival Time: 1100041 ns
Model #3: periodic
Period: 33520000 ns
Jitter: 118511893 ns
Offset: 4226024900838459 ns
On: release

Help
Navigation: Left/Right - Switch panels | Up/Down - Select item | Enter - View details | q - Quit

lime-rtw view /tmp/demo-ping

Detailed View: suspension (Scroll: Up/Down)

Separator: suspension Count: 3001

Details

Summary:

WCET range: 6236790 - 17705610 ns

★ Arrival Curve Table:

Shows the minimum and maximum length of intervals containing N job releases

Number of jobs	Minimum interval	Maximum interval
----------------	------------------	------------------

2	1100041	98350299
3	2471339	99767058
4	10642695	101734014
5	19561043	199219047
6	102438821	200622492
7	107209759	202527597
8	116128107	299959667
9	202985815	301397149
10	207862863	303244402
11	216781211	400678491
12	303536494	402182860
13	308739186	403955856
14	317657534	501379167

Minimum Separation ("delta min")

Shortest interval in which N jobs arrive.

→ *enables response-time analysis (RTA)*

Help

Detail View Mode Press ESC to return to the main view.

How Does LiME Work?



The LiME Pipeline

Blackbox Threads

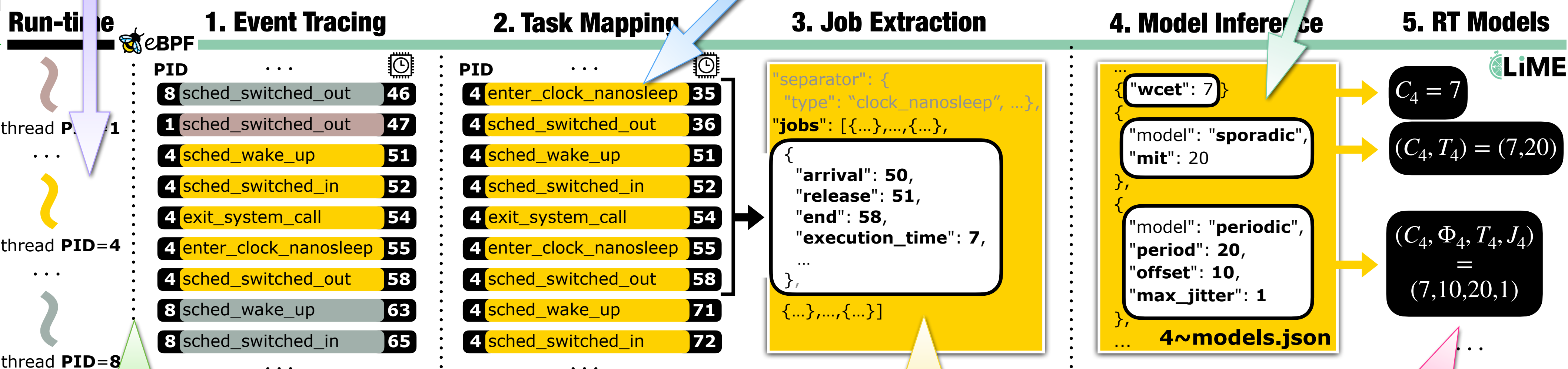
any code using *any* system calls in *any* order, without access to source code or static analysis of binary

Split Event Stream by Task

One thread can correspond to *multiple* tasks over time (see paper for details).

Online Model Inference

Infer task parameters from stream of jobs with *bounded memory*.



Kernel Instrumentation

installs *65 eBPF probes* in scheduler, selected system calls, etc. monitoring *all threads*

Identify Individual Task Activations (= Jobs)

RT task models assume repeatedly activated tasks:
task = stream of jobs

Output

Conservative models:
each model must reflect *all* observations.

Key Idea: Job Separators

Target real-time task models: “*a task is a sequence of **jobs***”

But Linux is **jobless**: each thread is an *arbitrary* sequence of system calls!

Fundamental Challenge

Where does one job end and the next start?

(1) LiME Insight

On Linux, a job (= one task activation) *always* starts with a *potentially blocking* system call.

*A real-time task **must wait** for next **event** or **passage of time** → blocking system calls.*

(2) LiME Heuristic

Recurrent real-time tasks
≈ “*doing the same thing over and over*”

→ *typically every job starts with the **same** system call, so LiME assumes this.*

Job Separator Examples

Time-driven:

```
struct timespec next = now();
while (true) {
    /* periodic activation */
    clock_nanosleep(&next);
    do_something();
    timespec_add(&next, PERIOD);
}
```

→ job separator: *clock_nanosleep*

Event-driven:

```
int fd = open_udp_socket(PORT);
while (true) {
    /* sporadic activation */
    int nbytes = read(fd, &buf);
    if (nbytes <= 0) break;
    do_something(buf, nbytes);
}
```

→ job separator: *read(fd)*

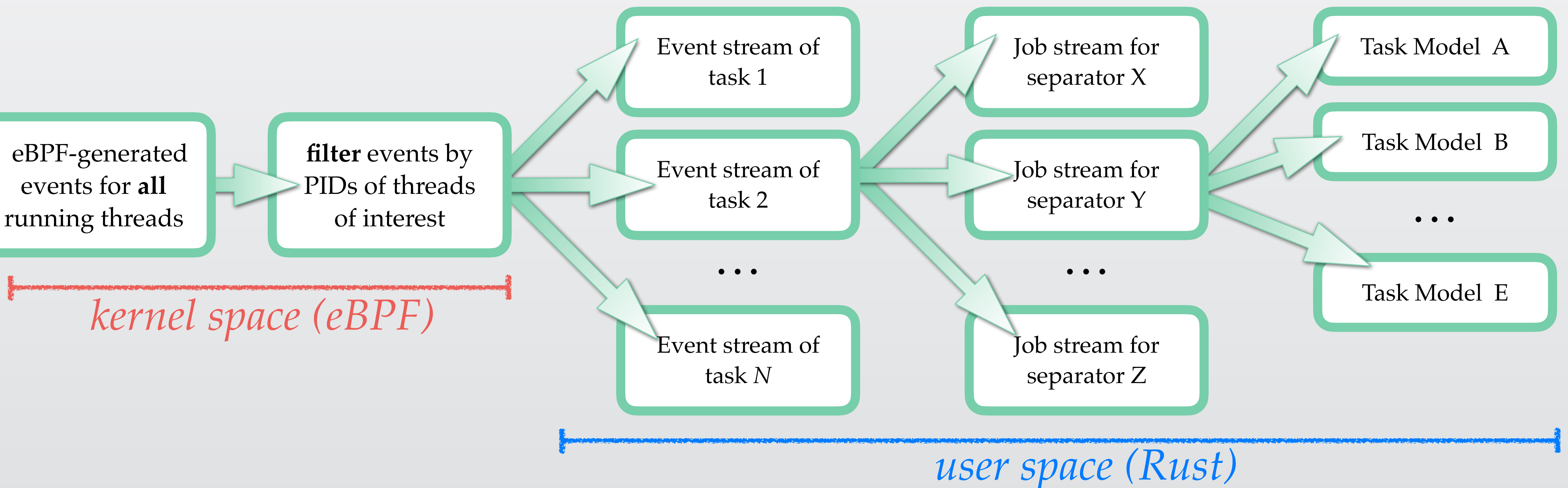
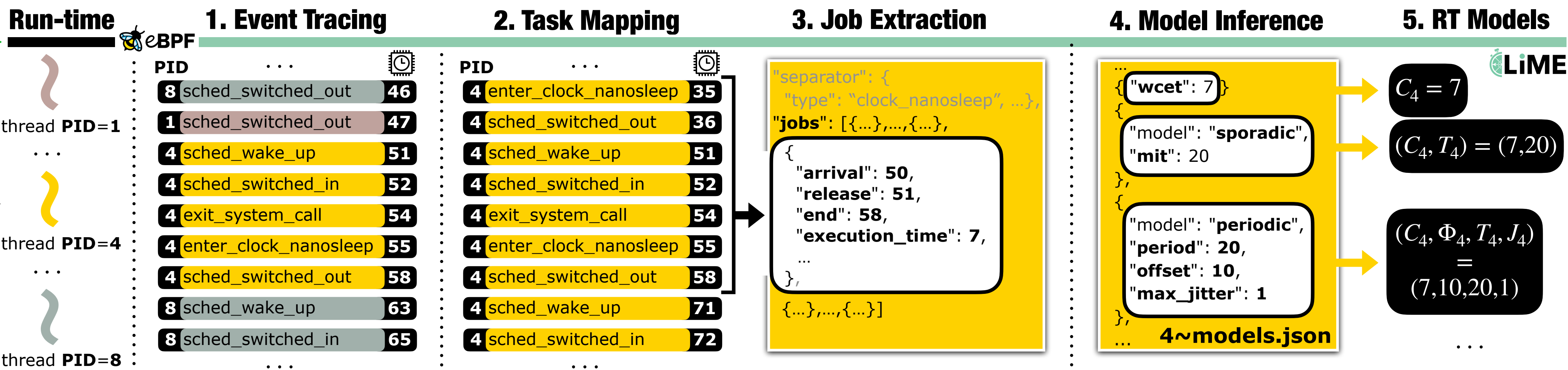
In both examples:

- one job = one loop iteration
- each job starts with *potentially* blocking system call
- all jobs of one task start with the *same* system call

Inherent ambiguity:

- task may execute many different system calls in *do_something()*
- There are potentially *many* job separator *candidates*.

Dataflow from Kernel to Model Extraction



LiME design choice: expose job-separator ambiguity to user.

Supported Task Models



Supported Task Models

Arrival Models

sporadic

(Mok, 1983)

arrival curves

(Thiele *et al.*, 2000; Richter, 2005)

periodic + jitter & offset

(Liu & Layland, 1973; Leung & Merrill, 1980; Audsley *et al.*, 1993)

WCET

(Liu & Layland, 1973)

WCET(*n*)

(Quinton *et al.*, 2012)

Execution Time

(*measurement-based*)

Self-Suspension

dynamic

(Ming, 1994)

generalized segmented

(von der Brüggen *et al.*, 2017)

Why Curve-Based Models?

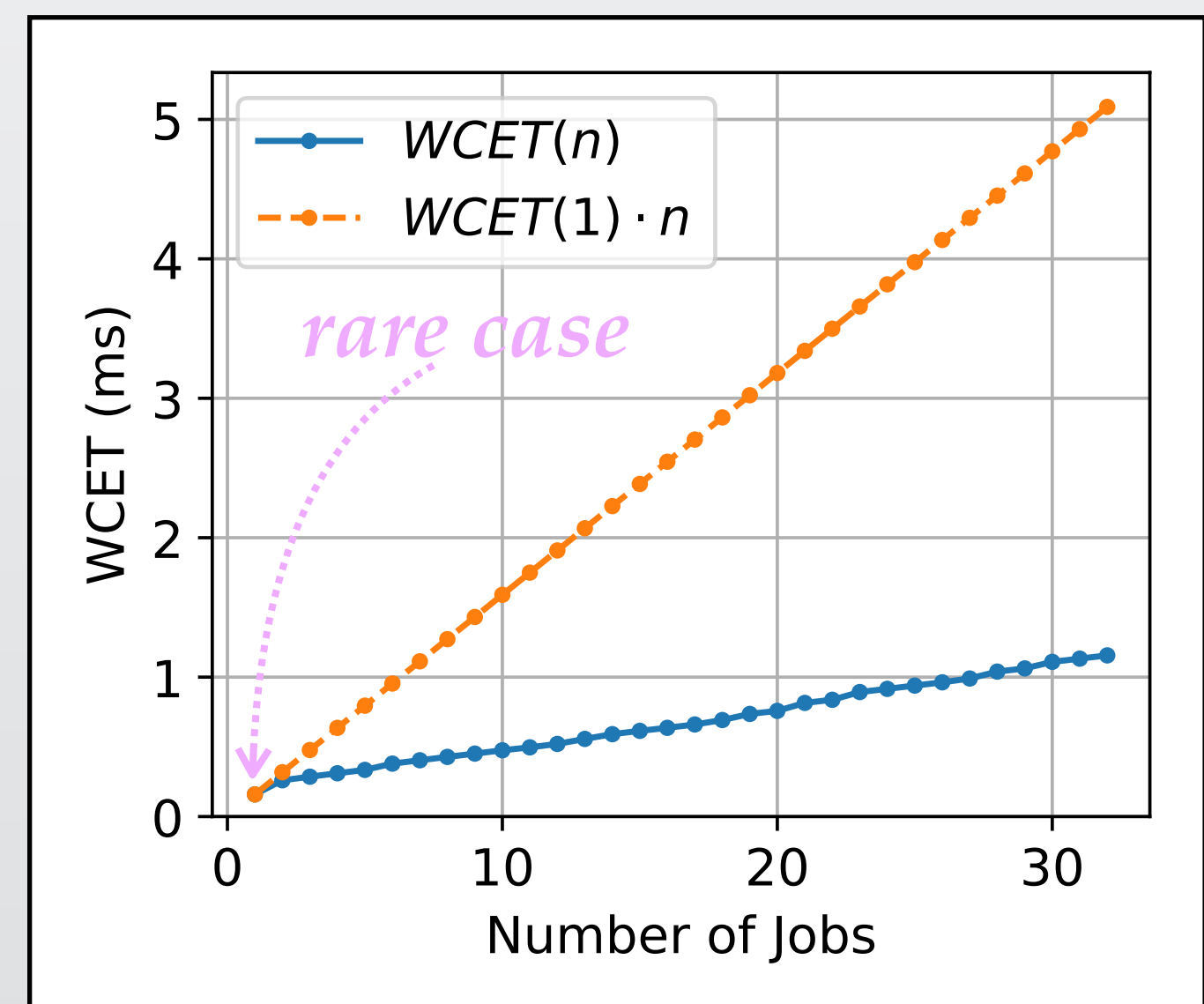
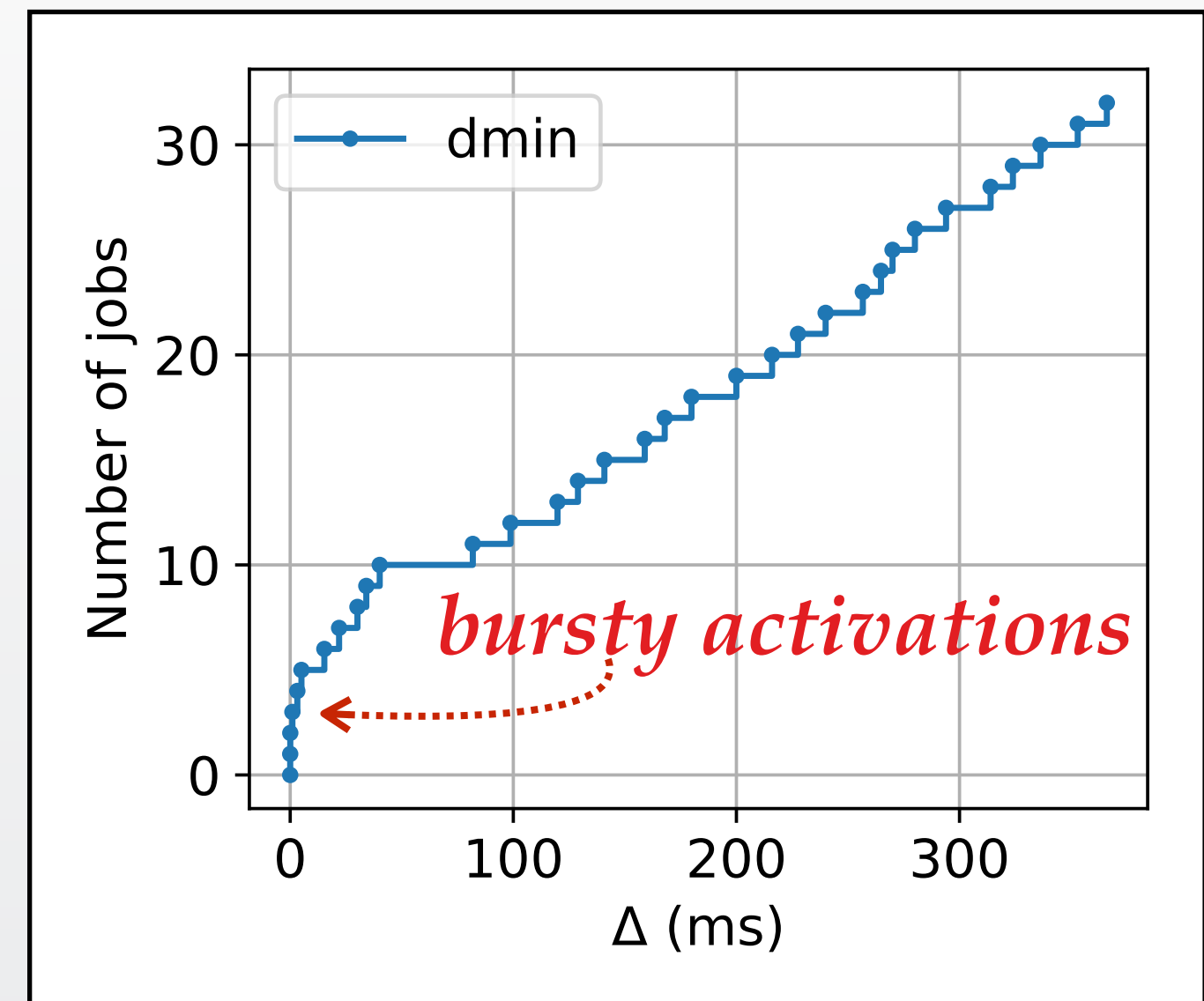
Why not stick with “tried and true” scalar model parameters?

Example: Linux kernel threads

- ➔ *migration* threads (one per core)
- ➔ assist Linux process scheduler
- ➔ execute at **max. RT priority**
- ➔ their CPU use must be accounted for

How much CPU use in 300 ms?

WCET	Arrivals	Bound
scalar	scalar	> 1500 ms
curve	scalar	> 340 ms
scalar	curve	≈ 4.3 ms
curve	curve	≈ 0.99 ms



LiME: The Linux Real-Time Task Model Extractor

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Abstract—We present LiME, a novel dynamic real-time task model extractor. LiME observes the temporal behavior of Linux real-time threads and automatically maps the observed activity to established real-time task models: sporadic and periodic tasks, upper and lower arrival curves, cumulative execution-time curves, and two self-suspension models (dynamic and segmented). LiME runs on unmodified Linux kernels and requires neither knowledge of real-time theory nor familiarity with Linux internals to be used effectively. An extensive evaluation shows LiME to achieve very high inference accuracy—in particular 100% accuracy for common automotive periods—with low kernel overhead, low latency impact, and low processor utilization (at best-effort priority).

I. INTRODUCTION

The recent acceptance of the PREEMPT_RT patch into the mainline Linux kernel, after 20 years of development [58], has made official what has long been true: Linux is a major platform for hosting modern, complex real-time workloads across various industries. Notable examples feature demanding applications such as *automotive systems* [52], *autonomous vehicles* [30, 31], *unmanned aerial vehicles* (UAVs) [19, 27, 35], *spacecraft* [37], in particular crewed rockets [59], NASA's Mars helicopter *Ingenuity* [57], and tens of thousands of Linux systems deployed in orbit as part of *Starlink* constellations [54]. As noted recently by Erik Vallow, a representative of the RTOS vendor *LYNX Software Technologies*, in a retrospective on the evolving role of Linux in the aerospace and defense industries [56]: “Linux has become a formidable contender in safety-critical systems due to advancements in real-time capabilities and reliability. [...] Linux is increasingly capable of meeting the demands of real-time applications on its own, reducing the need for a separate RTOS in some cases.” In addition, the availability of drivers for high-performance GPUs is increasingly a factor favoring the consolidation of AI-enabled or otherwise GPU-accelerated real-time workloads on Linux. However, while the popularity of Linux as a versatile and feature-rich RTOS has soared in the real-time systems industry, there is a growing disconnect with the analytical foundations studied in the scientific literature on real-time systems. Rooted in abstract system models and high-level mathematical descriptions of workloads, state-of-the-art methods for establishing temporal guarantees are far removed from the engineering realities of a low-level embedded Linux environment. It stands to reason, then, that only a diminishingly small fraction of the many real-time workloads deployed on Linux

over the past decade have been modeled and formally evaluated using published schedulability analyses. A major contributor to this disconnect is the lack of tool support: without automated system introspection tools, engineers interested in formally characterizing the timing properties of a real-time workload running on Linux require *dual expertise* in both Linux implementation details, particularly the kernel's low-level tracing facilities and system-call interface, and state-of-the-art temporal modeling and analysis techniques. This, along with the associated *manual* effort that would be required (and not just once, but repeatedly as systems evolve to meet changing requirements), presents a formidable barrier to the widespread adoption of state-of-the-art real-time analysis techniques in a Linux context.

Could the schedulability analysis of Linux workloads be automated and thus made *easily* accessible to non-experts? Motivated by this question, we address the first, fundamental problem that precedes any practical analysis: the *automated* modeling of real-time tasks deployed on *unmodified* Linux kernels.

While applications developed using higher-level *model-first* approaches [9, 24, 44], or using programming languages with explicit timing semantics [7, 10, 11, 15, 26, 28, 42, 43], can certainly be *compiled down* to Linux binaries and executed efficiently (*i.e.*, model-driven engineering), the converse is far from obvious: *Is it possible to extract high-level temporal models of running Linux threads suitable for schedulability analysis simply by observing their low-level runtime behavior?* Is it possible to do this for *black-box* threads (*i.e.*, without access to source code)? Fully *automatically*, without user guidance, annotations, or specifications of intended timing? And can it be done *in situ* on a target embedded platform without unduly perturbing the timing of the real-time threads?

This paper. We show that the answer to each of these questions is ‘yes’ by presenting LiME, the **Linux** real-time task **Model** Extractor (Sec. III). LiME is a dynamic introspection tool that maps sequences of low-level thread-kernel interactions (Sec. II-A), observed via the kernel's *eBPF* tracing facility, to task models from the real-time scheduling literature (Sec. II-B).

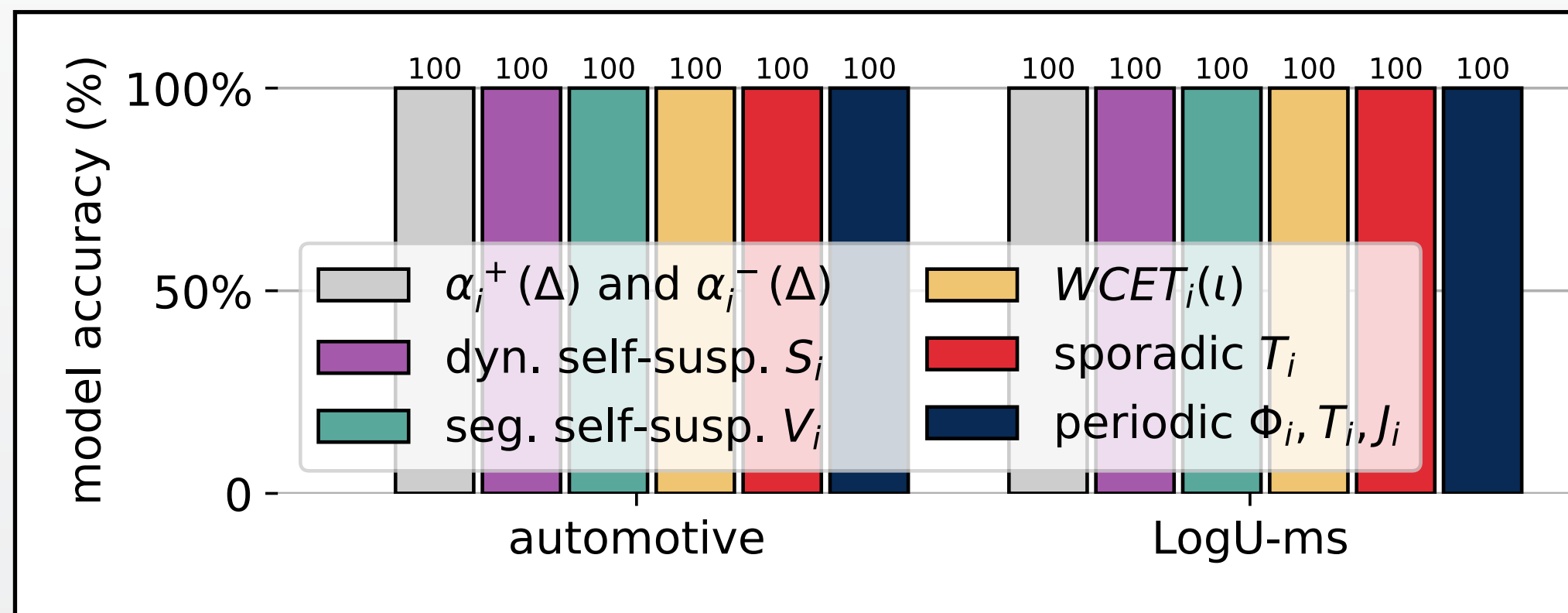
Fig. 1 illustrates the entire pipeline: Given an arbitrary black-box workload (in Fig. 1, a thread implementing periodic activations with `clock_nanosleep`), LiME uses eBPF to observe key scheduling events and system calls throughout the whole system (Inset 1). After reconstructing the timeline for each target thread (Inset 2), LiME identifies job boundaries based on its builtin understanding of Linux system call

*†‡ Authors are listed in alphabetical order. †‡ This work was carried out while affiliated with the Max Planck Institute for Software Systems, Germany.

Evaluation in the Paper



LiME is Accurate and Effective



100% Inference Accuracy

→ *automotive periods*

→ *random periods (ms-granularity)*

The ROSACE Case Study: From Simulink Specification to Multi/Many-Core Execution

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Abstract—This paper presents a complete case study - named ROSACE for Research Open-Source Avionics and Control Engineering - that goes from a baseline flight controller, developed in MATLAB/SIMULINK, to a multi-periodic controller executing on a multi/many-core target. The interactions between control and computer engineers are highlighted during the development steps, in particular by investigating several multi-periodic configurations. We deduced ways to improve the discussion between engineers in order to ease the integration on the target. The

P3 : rise time, that is the time it takes to rise from 10% to 90% of the steady-state value;

P4 : steady-state error, that is the difference between the input and the output for a prescribed test input as $t \rightarrow \infty$.

The time-domain performance properties are illustrated in the figure 1 for a step input. At this stage, these properties are analyzed through SIMULINK simulations.

ROSACE Case Study

→ *longterm* observation

→ detect subtle timing *drift*

→ *validate fix*

Pagetti *et al.*, RTAS 2014

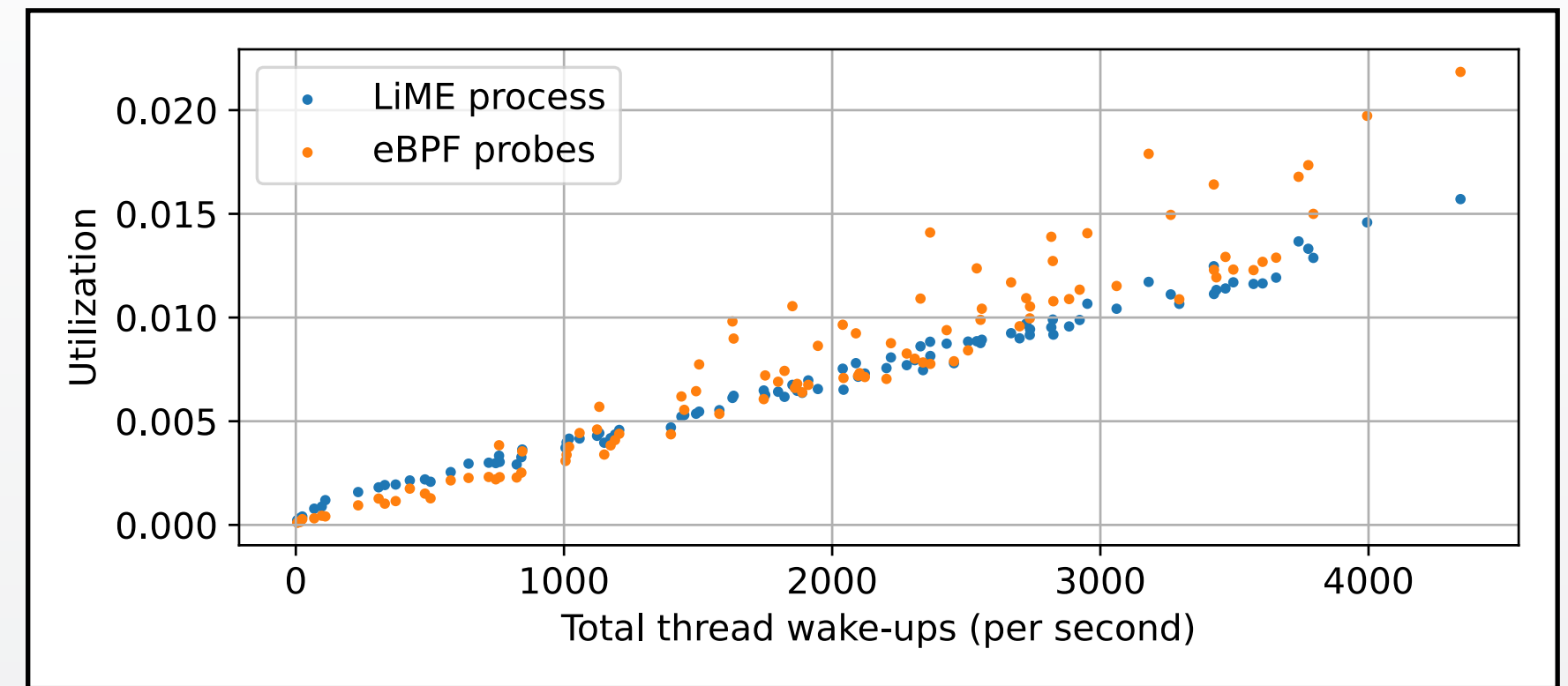
LiME Causes Only Low Overhead

CPU overhead

Kernel (eBPF): $< 2.5\%$ (of 4 cores)

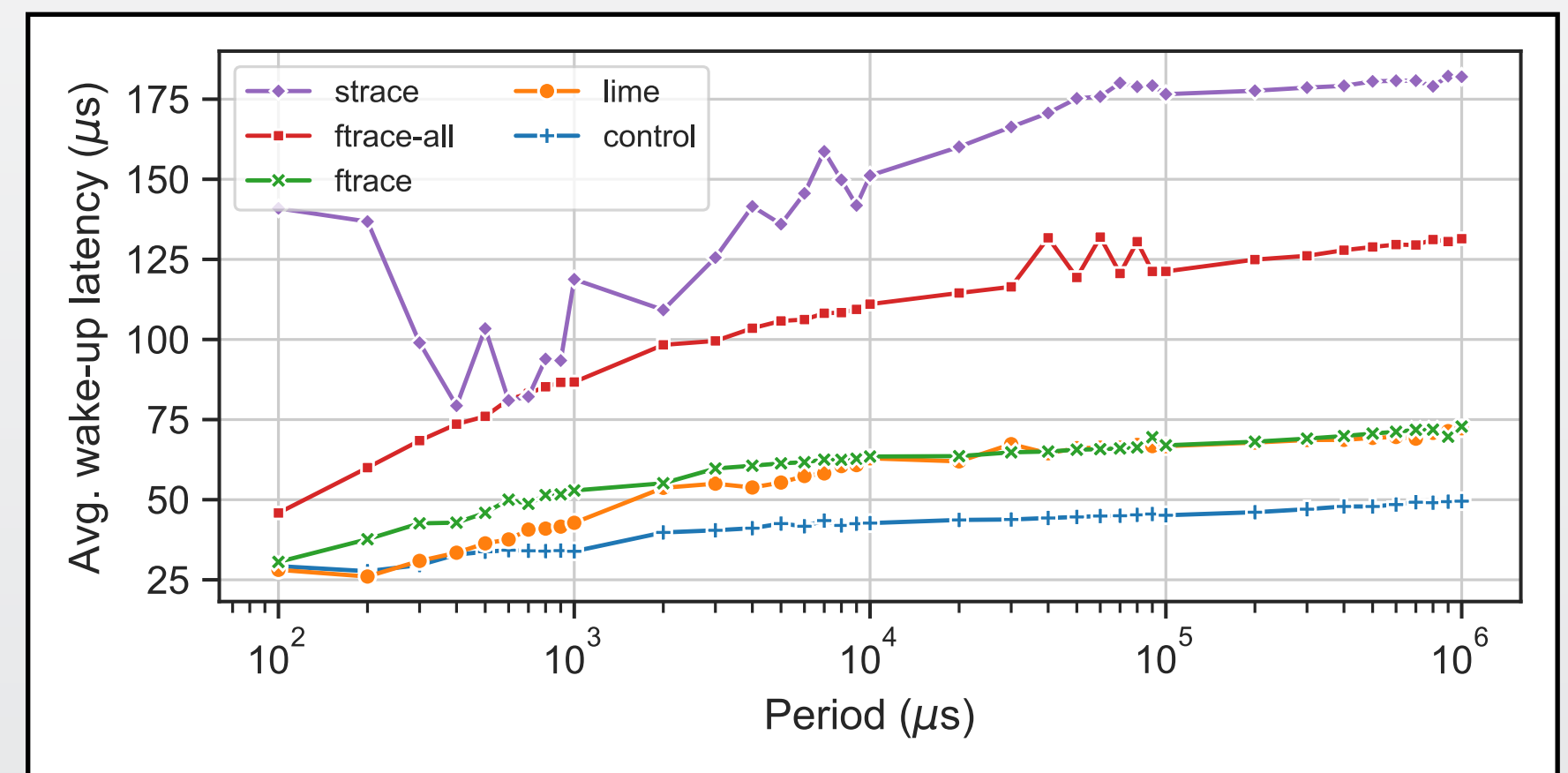
User space: $< 2.0\%$

(on a Raspberry Pi 4 “Model B”)



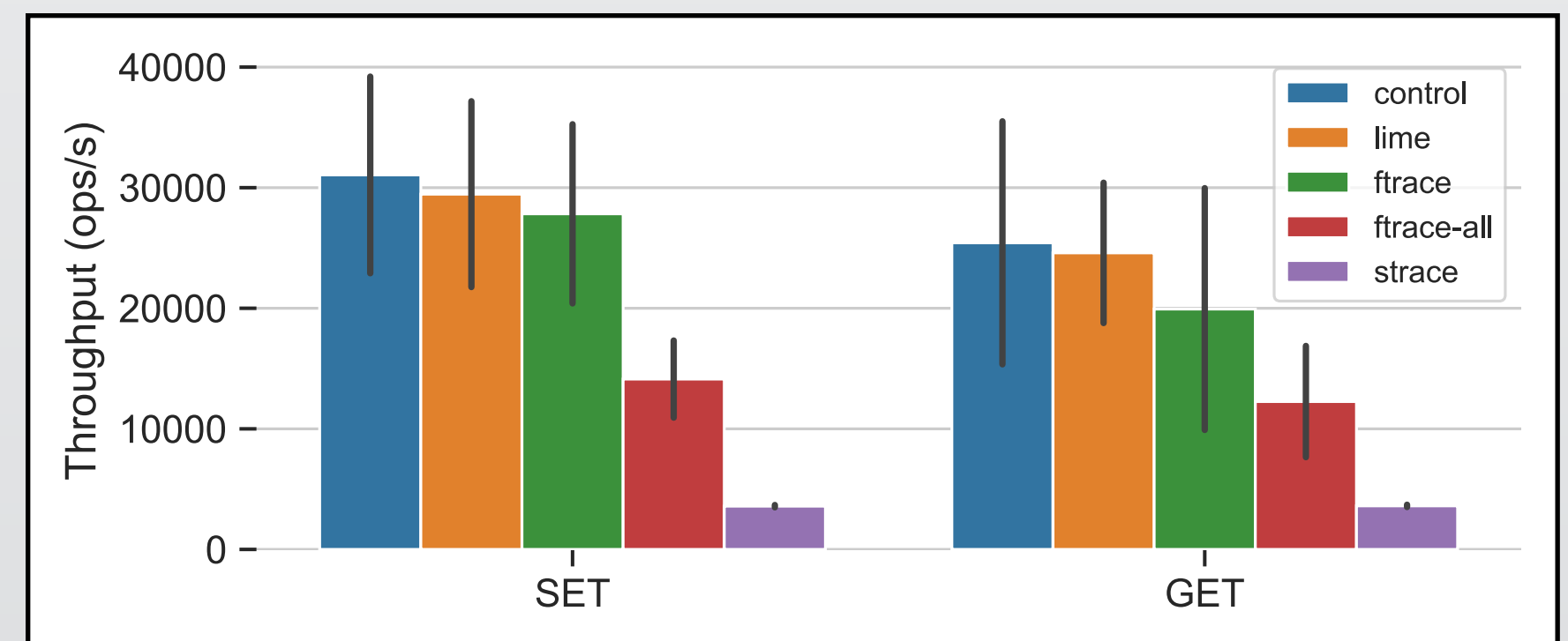
cyclictest latency benchmark

→ *avg. latency impact $\approx 25\mu\text{s}$*



Redis key-value server

→ *minor throughput impact*



Conclusion



What Can LiME Do For You?

1

If you are doing **systems** research:

→ *Integrate schedulability analysis (almost) "for free"*

2

If you are working with a **real system**:

→ *Monitor, understand, debug, validate its timing behavior*

3

If you work on **schedulability analysis**:

→ *Get some real models rather than just random numbers*

→ *Demonstrate that your model assumptions are viable*

4

If you **teach**:

→ *Let students explore RT foundations hands-on*



<https://lime.mpi-sws.org/>

**blackbox threads
running on Linux**



**established real-
time task models**

→ fully automated
→ online inference

→ highly accurate
→ low overheads

Thank you for your attention! Any Questions?



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