



MAX PLANCK INSTITUTE
FOR SOFTWARE SYSTEMS



Sant'Anna
Scuola Universitaria Superiore Pisa

Shape-Aware Analysis of End-to-End Latency Under LET

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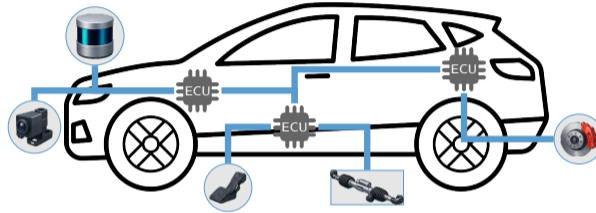
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May 12, 2026 @RTAS 26

This result is part of a project (PropRT) that has received funding from the European Research Council (ERC) under the European Union's Horizon 2020 research and innovation programme (grant agreement No. 865170). This work has been funded by the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) – Project No. 569077889 (PEACH). This work was partially funded by the Swedish Research Council (VR) under the project nr. 2023-04773 and by Sweden's Innovation Agency via the NFFP8 project 2024-01267: PARTI. This work has been partially supported by the project SERICS (PE00000014) under the MUR National Recovery and Resilience Plan funded by the European Union – NextGenerationEU.

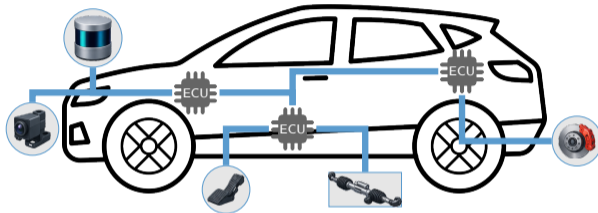
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- Automotive Systems: Tasks work together to perform functionality



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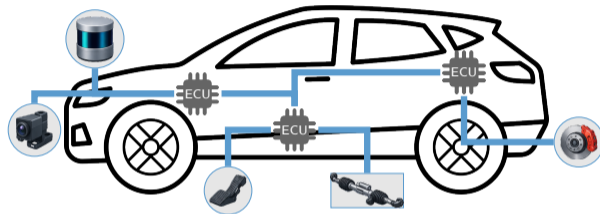
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- Want: Analyze timing behavior of full cause-effect chain

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- For many metrics, we provide *first analysis*

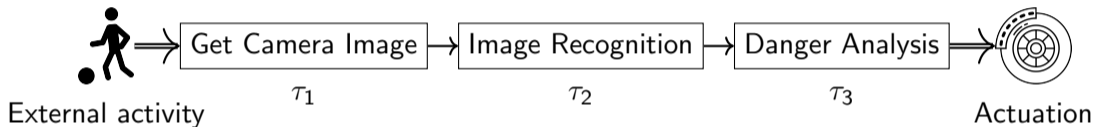
End-to-End Latency of Cause-Effect Chains

Shape-Aware Analysis of Reaction Time

Evaluation

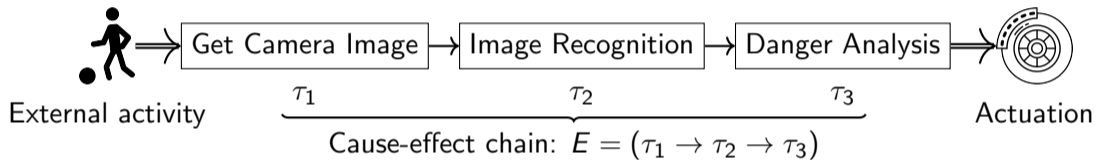
Cause-Effect Chains

- Functionality is described by a sequence of tasks



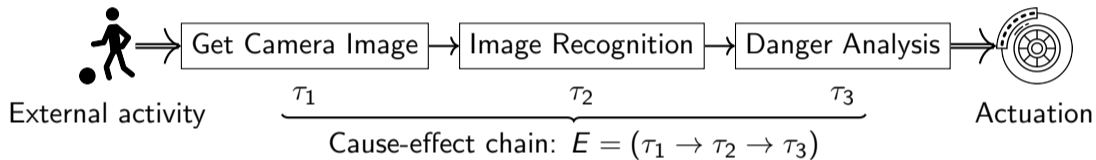
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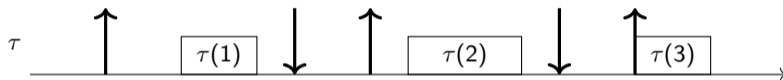


Cause-Effect Chain

Sequence $E = (\tau_1 \rightarrow \tau_2 \rightarrow \dots \rightarrow \tau_n)$ of tasks τ_i

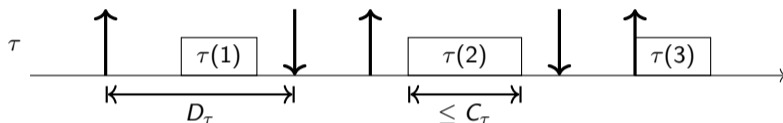
(Real-Time) Task Model

\mathbb{T} task set, $\tau \in \mathbb{T}$ task, releases jobs $\tau(i)$



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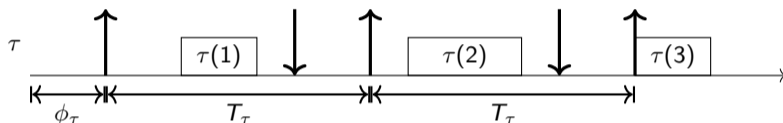
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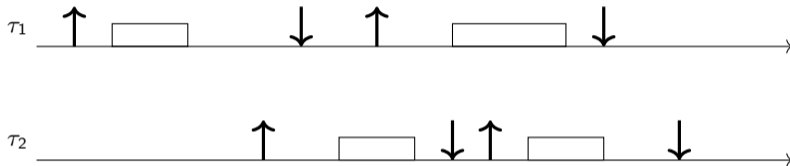
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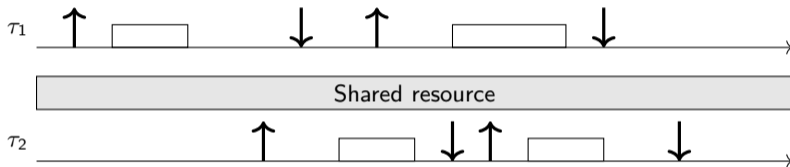
Communication Under Logical Execution Time (LET)

- Communication modeled through shared resource



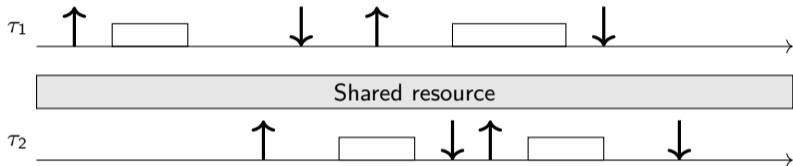
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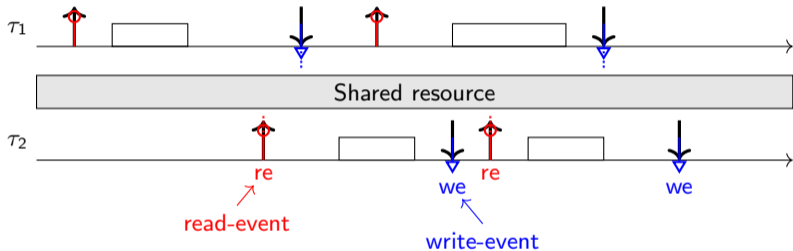
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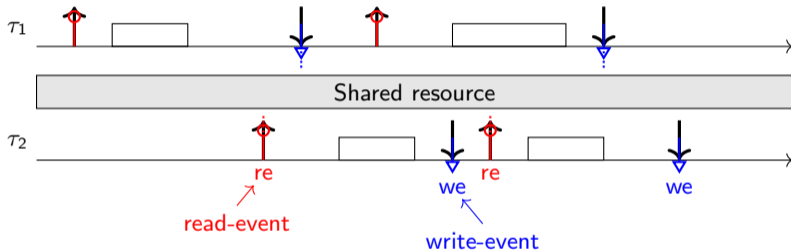
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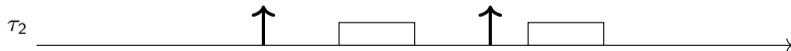


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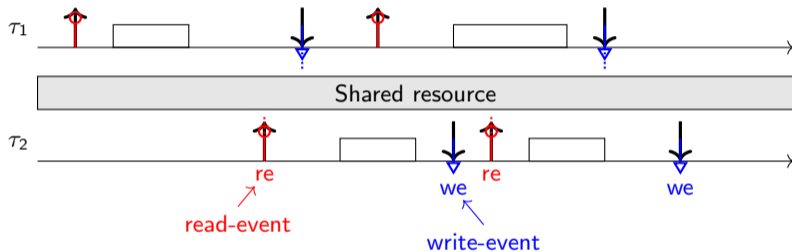


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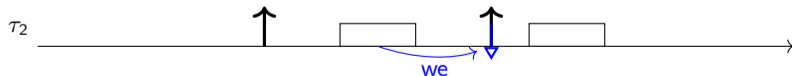


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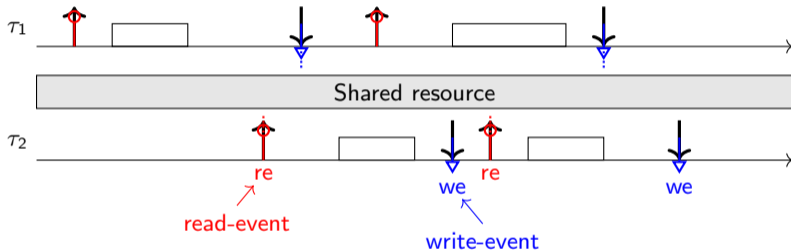


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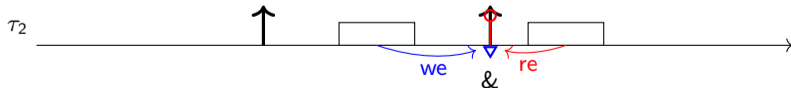


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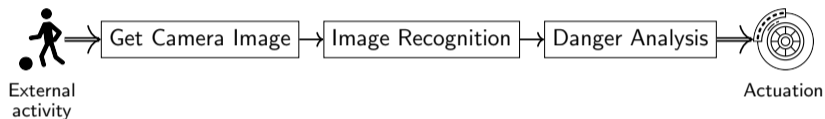
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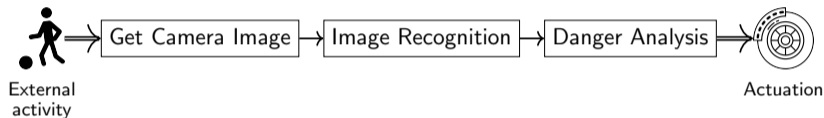
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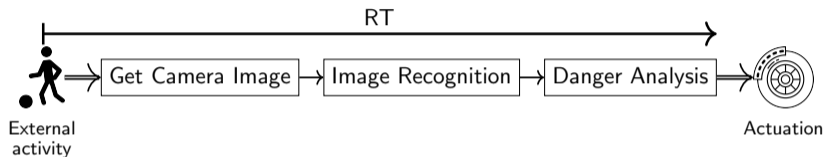
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Reaction Time $RT(z)$

Given an external activity at time z , how long does it take until this external activity is fully processed?

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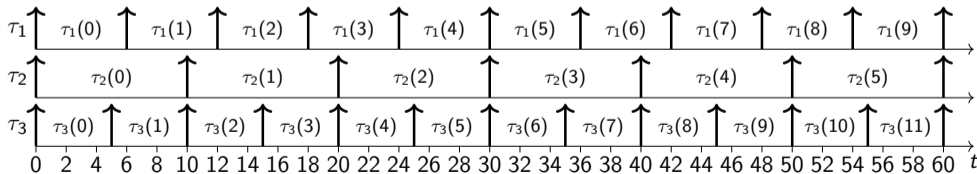
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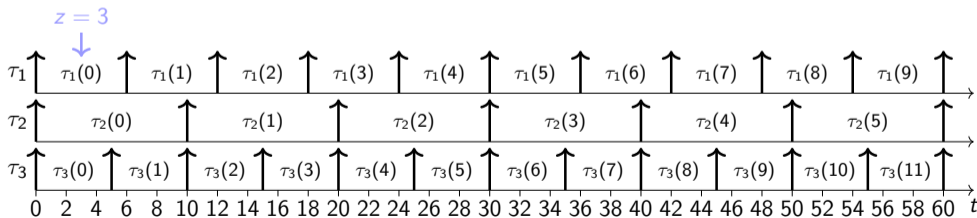


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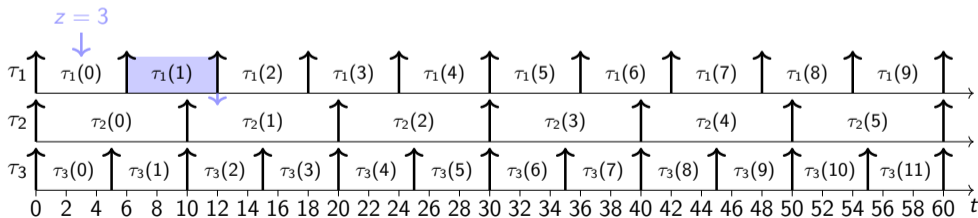


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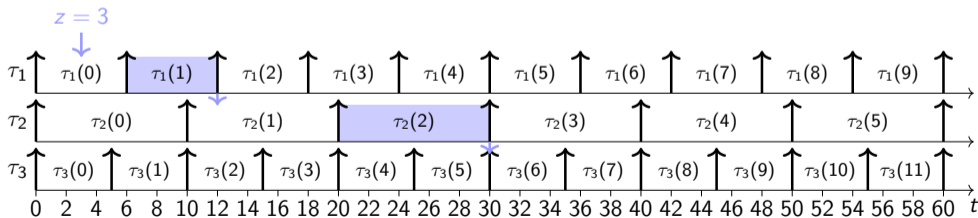


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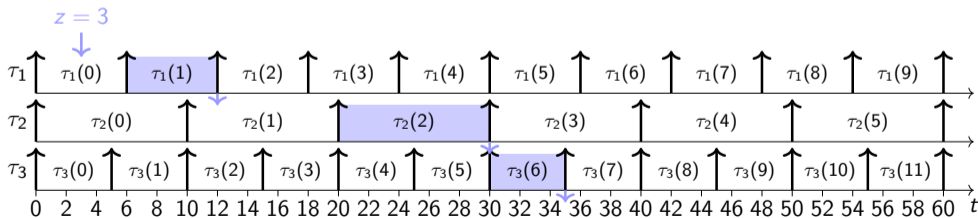


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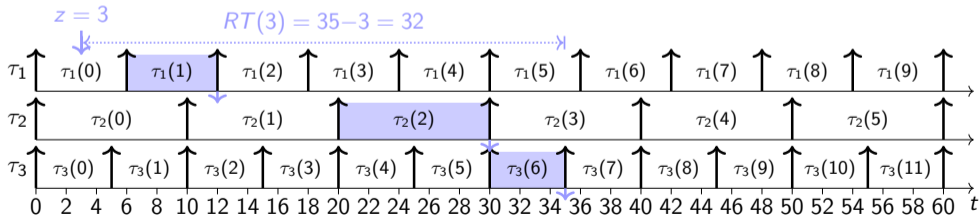


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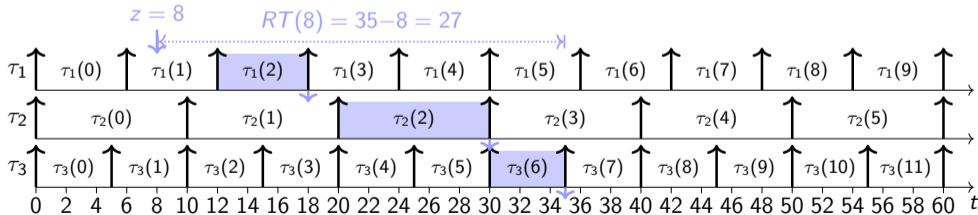


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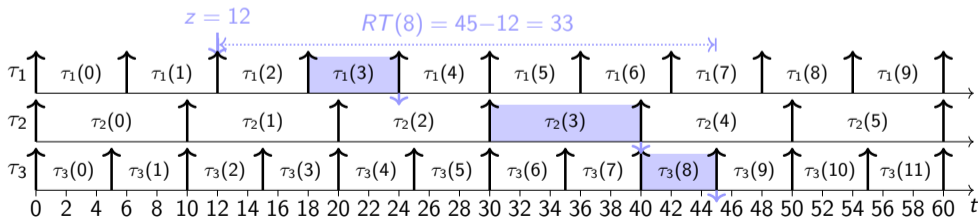


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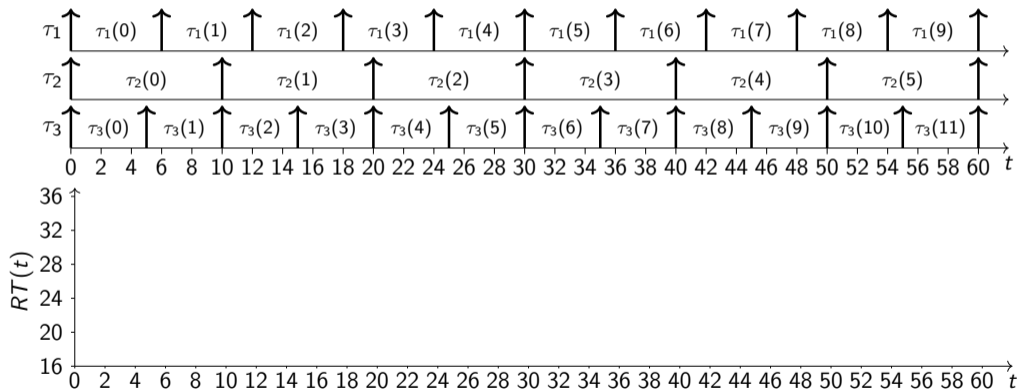
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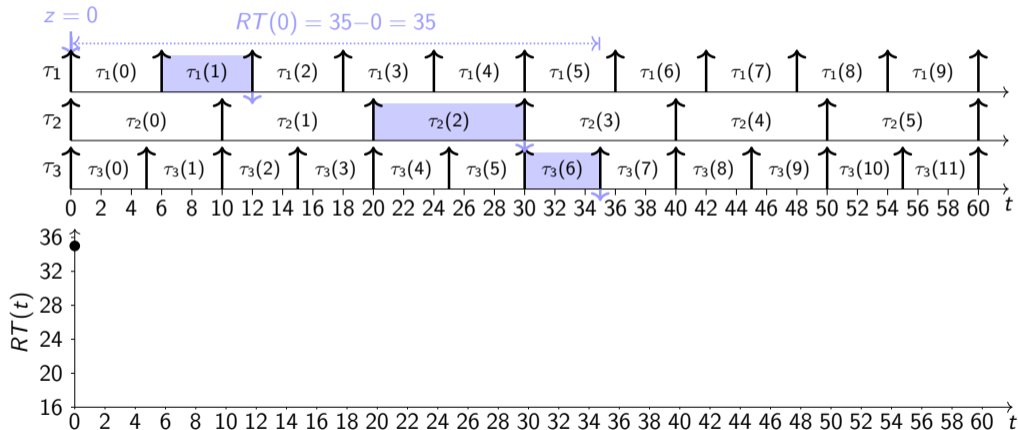
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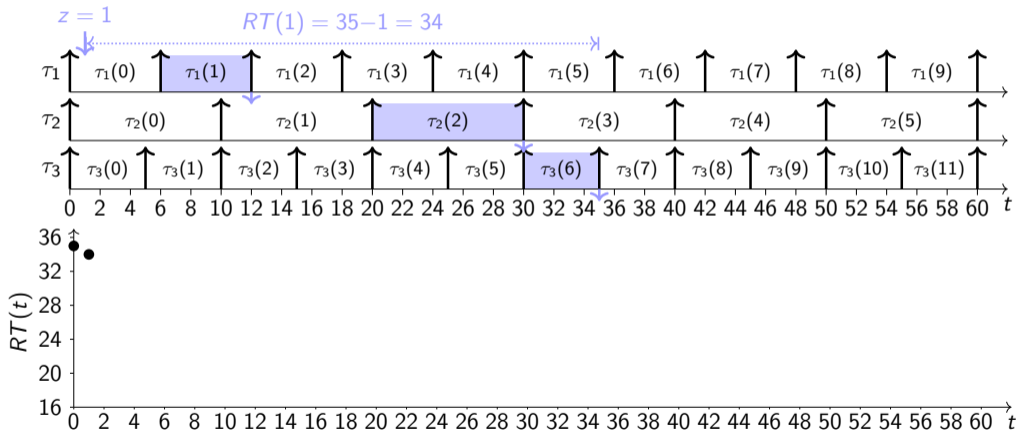
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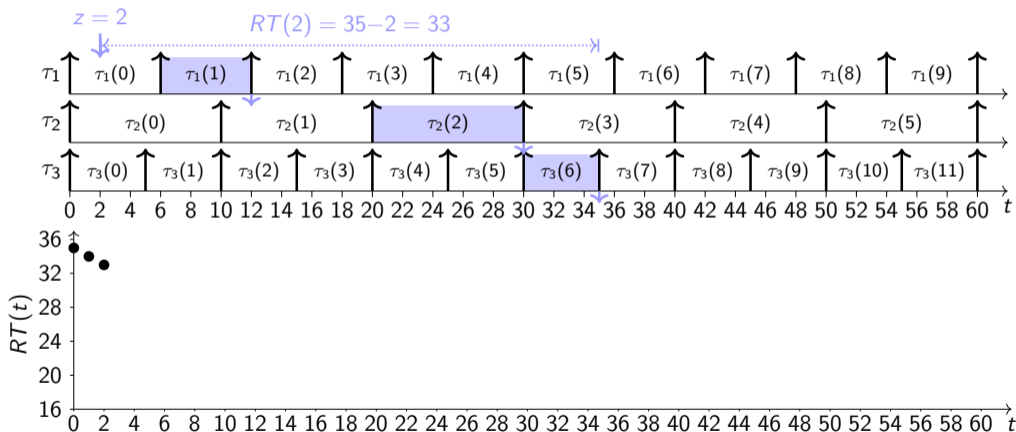
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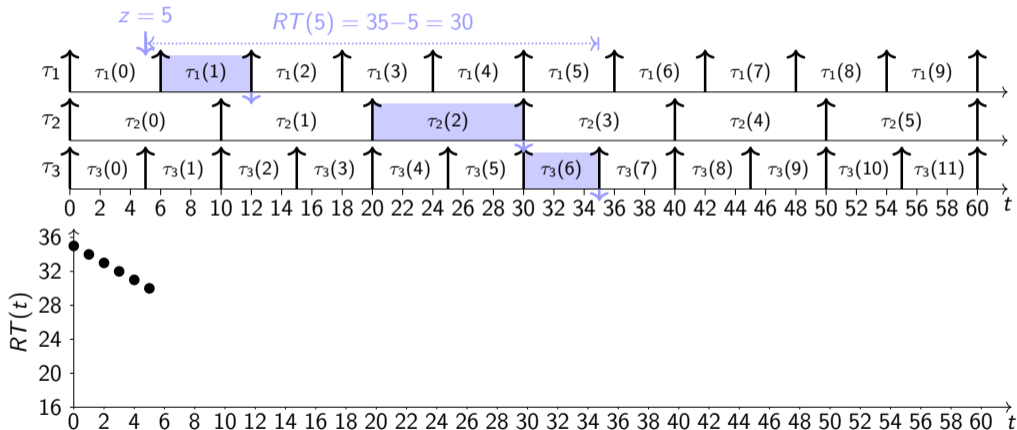
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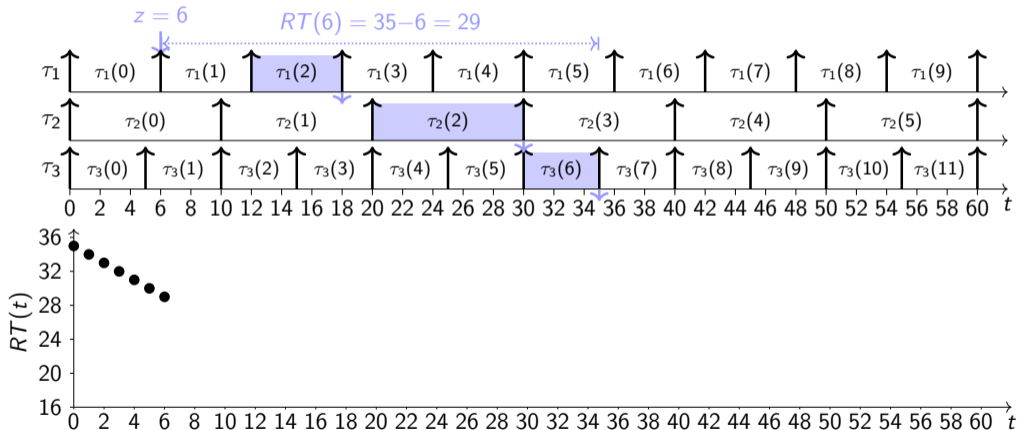
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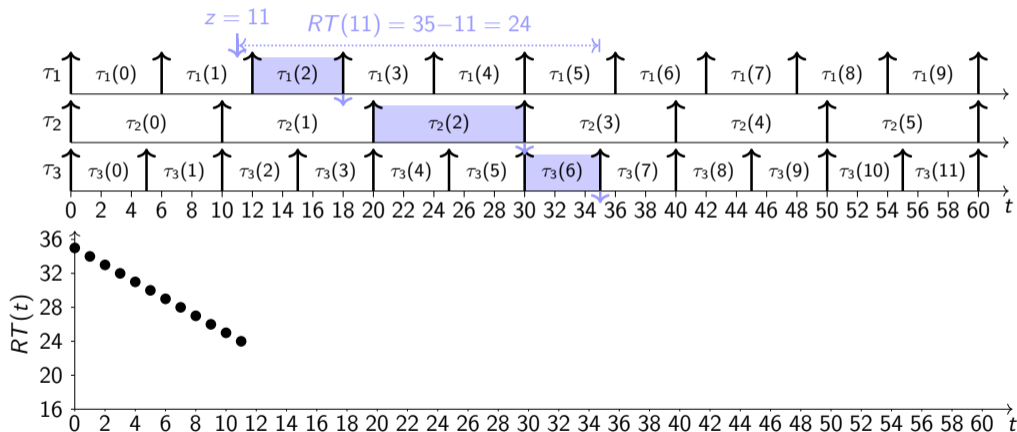
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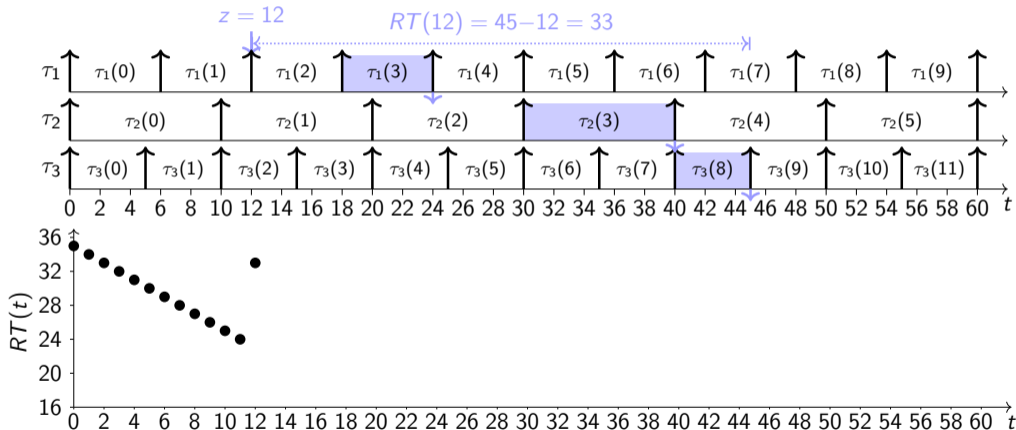
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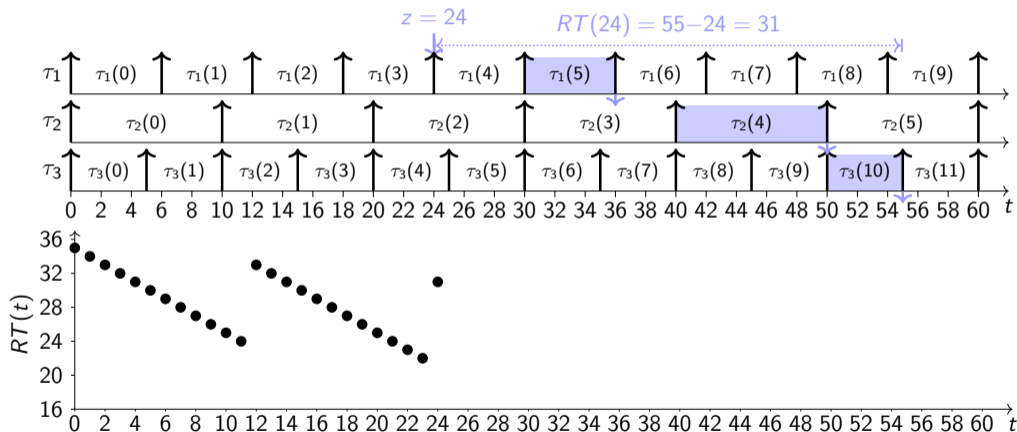
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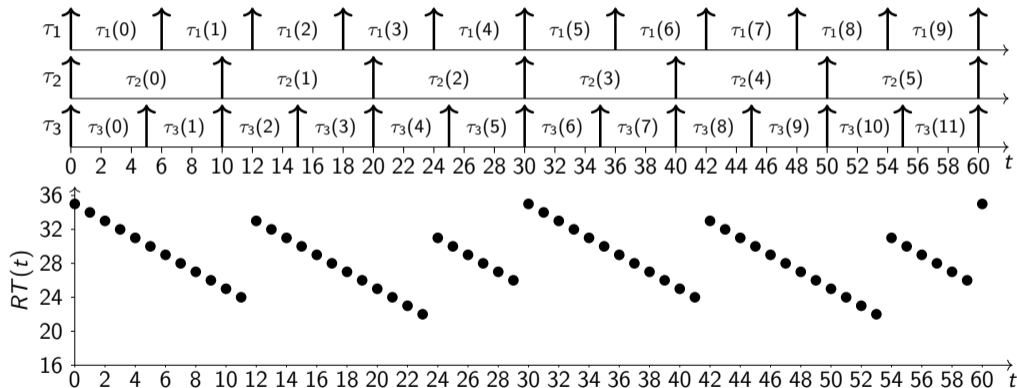
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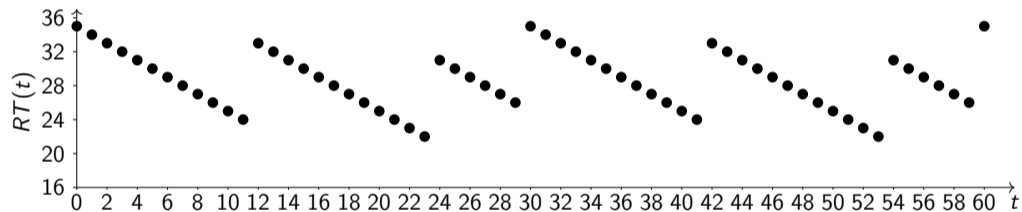
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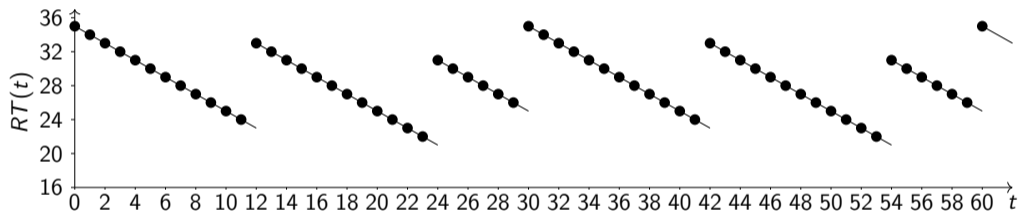
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Observations

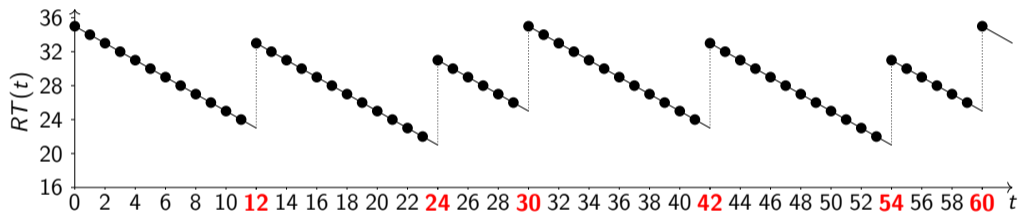


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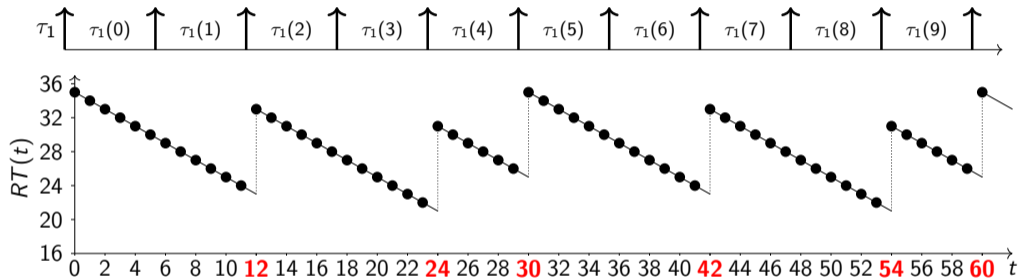
1. RT Linearly decreasing

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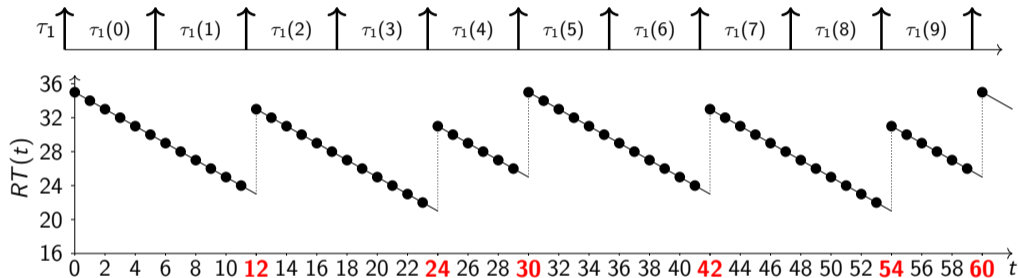
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Formalization of Observations

Lemma 8

RT restricted to the interval $[\text{re}(\tau_1(m)), \text{re}(\tau_1(m+1))]$, $m \in \mathbb{N}$ is linearly decreasing:

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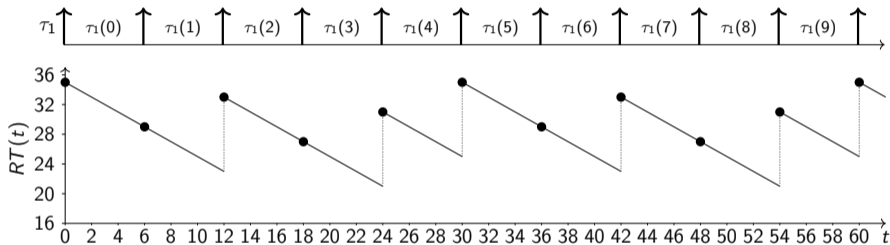
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⇒ Full shape of RT can be determined by evaluating at read-events of τ_1

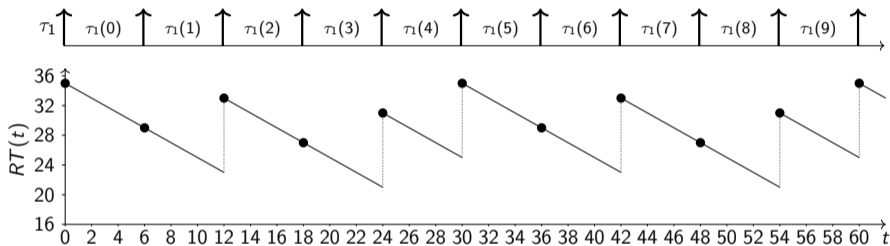
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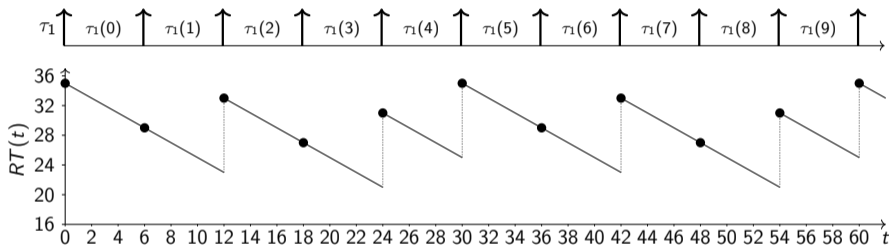


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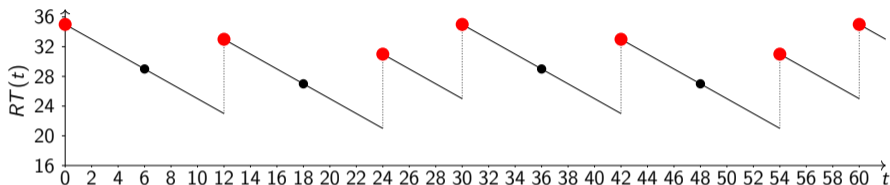


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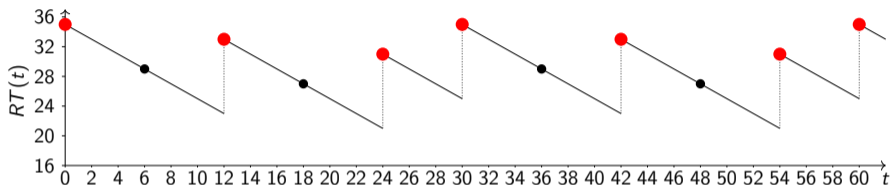
$$\mathcal{A}^1 = \left\{ \begin{array}{l} (0, 35), (6, 29), (12, 33), \\ (18, 27), (24, 31), (30, 35), \\ (36, 29), (42, 33), (48, 27), \\ (54, 31), (60, 35), \dots \end{array} \right\}$$

Minimal Set of Anchor Points



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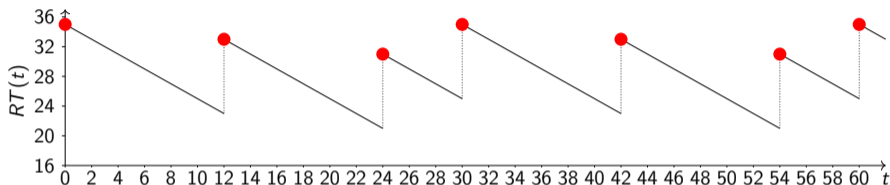
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$$\mathcal{A}^1 = \{(0, 35), (6, 29), (12, 33), (18, 27), (24, 31), (30, 35), (36, 29), (42, 33), (48, 27), (54, 31), (60, 35), \dots\}$$

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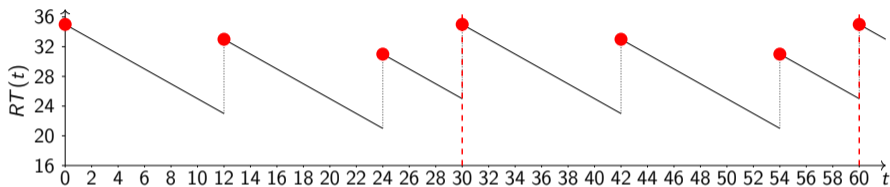


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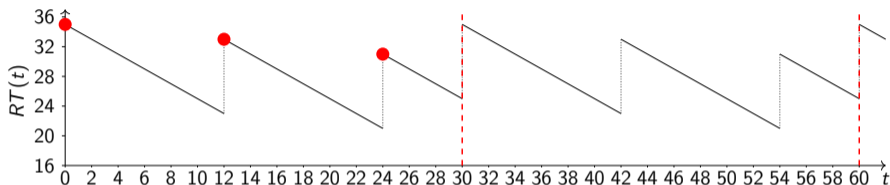


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1. Remove redundant anchor points from \mathcal{A}^1
2. Consider first hyperperiod $[0, H)$ with $H = \text{LCM}(T_1, \dots, T_n)$ ($= 30$ for example)

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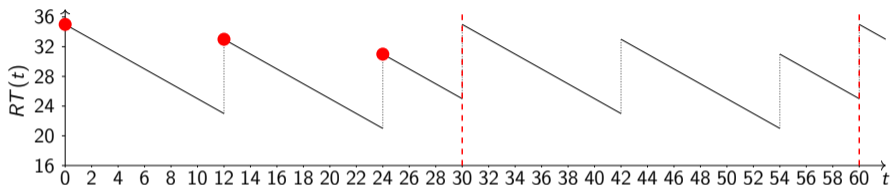


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1. Remove redundant anchor points from \mathcal{A}^1
2. Consider first hyperperiod $[0, H)$ with $H = \text{LCM}(T_1, \dots, T_n)$ ($= 30$ for example)
 $\Rightarrow \mathcal{A}^*|_H$ is sufficient to describe the whole shape of RT

Analysis of Metrics Becomes Simple

Given: Anchor points $\mathcal{A}^*|_H = \{(x_0, y_0), \dots, (x_g, y_g)\}$ over one hyperperiod $[0, H)$

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Analysis of Metrics:

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- Average: $AvRT = \frac{1}{2H} \sum_{i=0}^g (x_{i+1} - x_i) \cdot (y_i + \hat{y}_i)$

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- $Thr = \frac{1}{30} \cdot 3 = 0.1$

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“largest time interval that constraint is exceeded”
- Determine $\mathcal{A}^*|_H$ efficiently using *partitioned job chains*
 \Rightarrow Speedup of $\approx 1000x$ for automotive tasks

Implementation in Python

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- Simple specification of cause-effect chains:

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{"ID": 0, "tasks": [{"phase": 0, "period": 6, "deadline": 6}, {"phase": 0, "period": 10, "deadline": 10}, {"phase": 0, "period": 5, "deadline": 5}]}
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Evaluation of Case Studies

Case Study	Max	Min	Av	Thr	(m,k)	LE
Wat17-C1	50	40	45.0	0.100	(0,10)	2.50
Wat17-C2	212	112	162.0	0.010	(0,10)	10.60
Wat19-C1	908	470	689.0	0.003	(1,10)	45.40
Wat19-C2	855	445	650.0	0.003	(4,10)	42.75
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Wat19-C5	164	86	125.0	0.015	(0,10)	8.20
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RTSS-C1	610	510	560.0	0.010	(0,10)	30.50
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RTSS-C4	410	310	360.0	0.010	(0,10)	20.50
RTSS-C5	320	220	270.0	0.010	(1,10)	16.00
APD	275	225	250.0	0.020	(0,10)	13.75
Bec24	360	240	282.0	0.017	(0,10)	18.00
Gem21-UP	19	13	16.0	0.200	(0,10)	0.95
Gem21-LP	31	21	26.0	0.100	(0,10)	1.55
lye20	360	310	335.0	0.020	(2,10)	18.00
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Observations:

- **Few anchor points:**
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- **Low Runtime:**
Total analysis time: 0.062 seconds
(single core with 2.25 GHz)

Runtime Comparison with SOTA

Two benchmarks:

Runtime Comparison with SOTA

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- WATERS: Periods in $\{1, 2, 5, 10, 20, 50, 100, 200, 1000\}$ with given distribution¹

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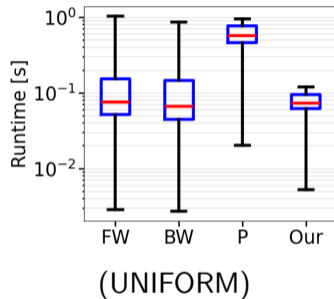
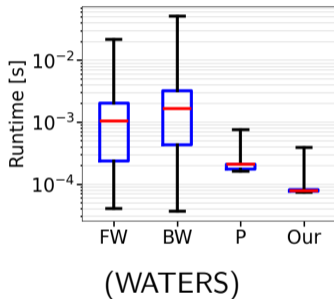
Approaches for Comparison:

Paper	Approach	Metrics
RTNS 2023 (FW)	Forward	MaxRT
RTAS 2023 (BW)	Backward	Reac
ECRTS 2023 (P)	Partitioned	MaxRT, MaxRedRT
This paper (Our)	Partitioned	MaxRT, MinRT, MaxRedRT, Reac, AvRT, Thr, (m,k), LE

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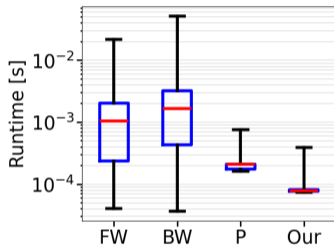
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Plots with $N = 50$ tasks per chain (more in the paper):

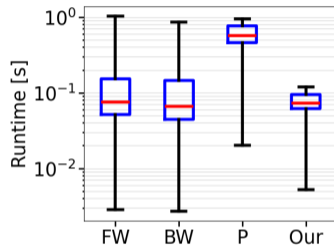


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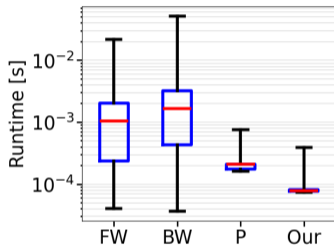


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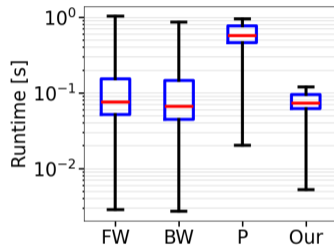
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- Comparable runtime for UNIFORM and improved runtime for WATERS
- **Note:** Major effort (= determining anchor points) only *once!*
⇒ Our analyzes *multiple* metrics in similar time

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Framework²:



²<https://github.com/marioguenzel/Shape-Aware-E2E.git>

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- Demonstrate: Allows simple (and fast!) analysis of multiple metrics

However, this is only the first step!



⇒ *It is easy to integrate further metrics into the framework!*

Future work:

- Further metrics and exploitation of shape for design choices
- Data-age related metrics
- Other communication paradigms such as implicit communication

²<https://github.com/marioguenzel/Shape-Aware-E2E.git>