



MAX PLANCK INSTITUTE  
FOR SOFTWARE SYSTEMS

# An Exact and Sustainable Analysis of Non-Preemptive Scheduling

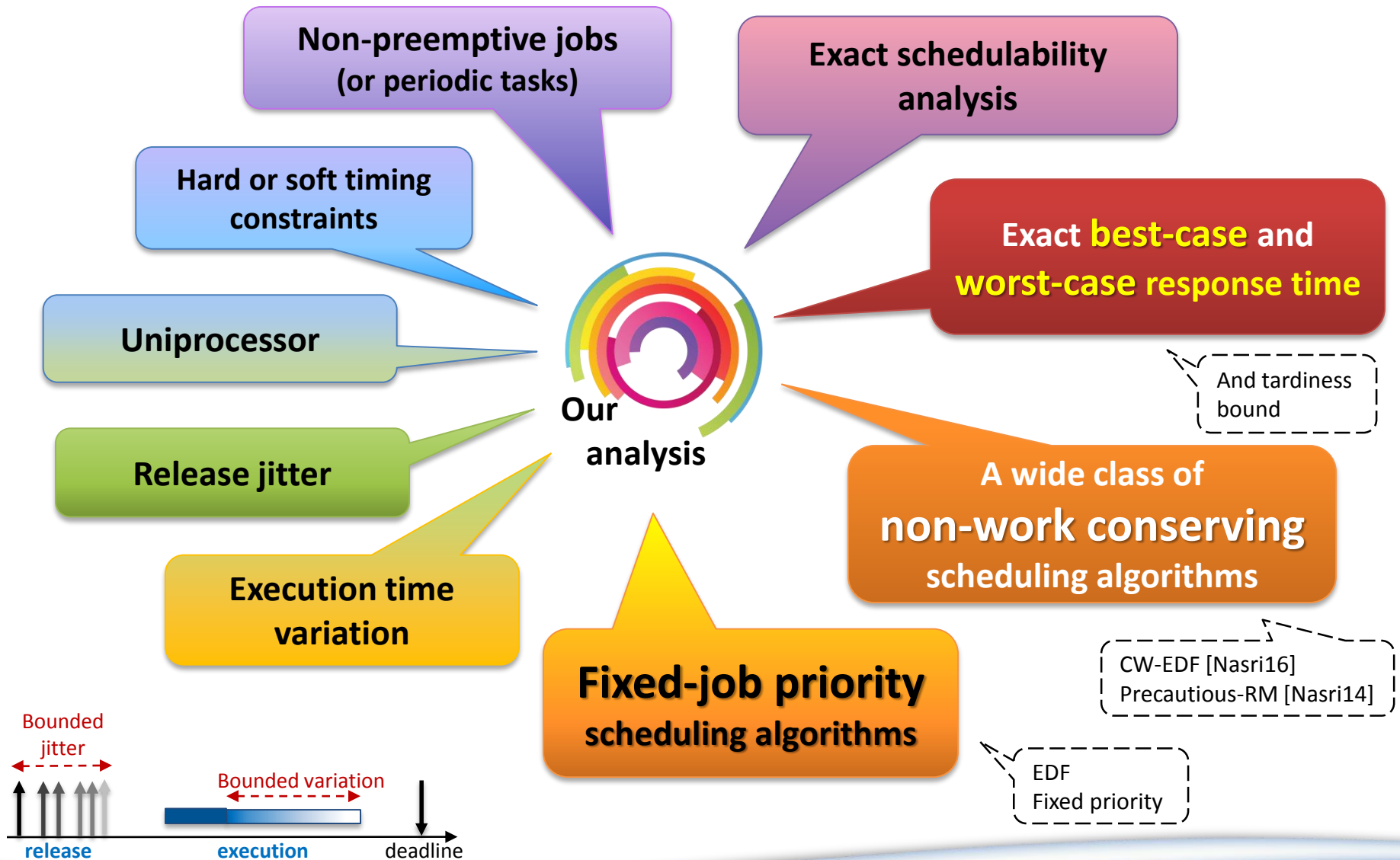
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Max Planck Institute for Software Systems (MPI-SWS)  
Germany

# Our work in a nutshell

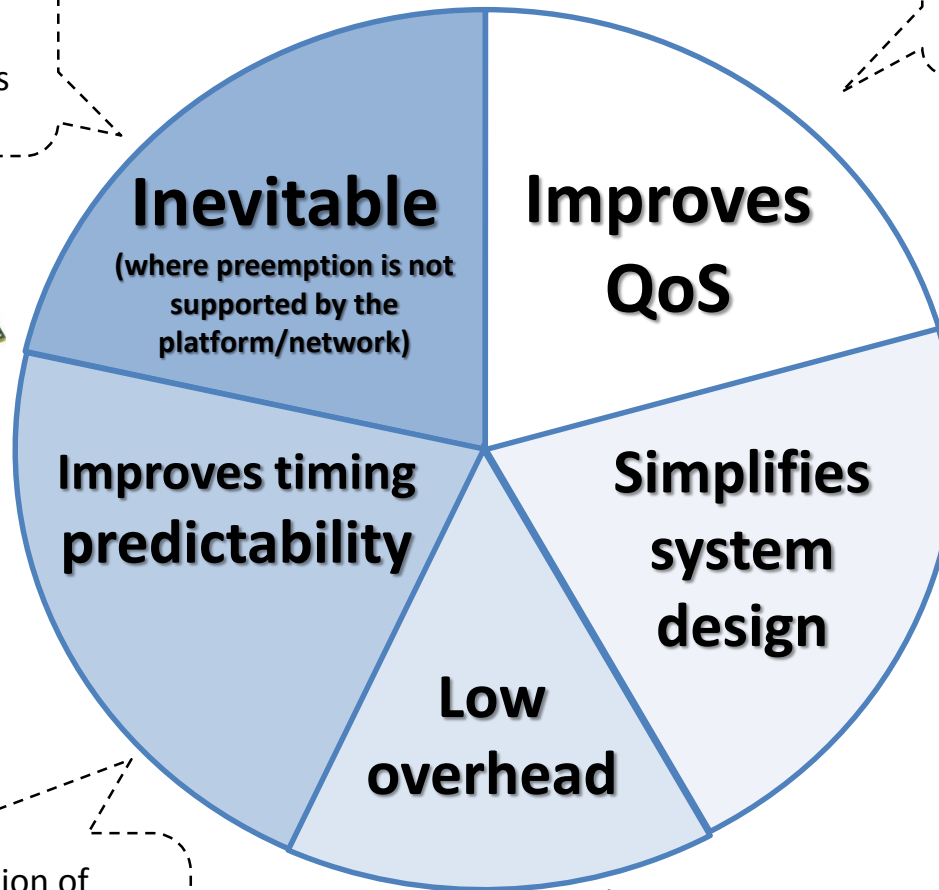
"An exact and sustainable schedulability analysis for non-preemptive scheduling"



# Why non-preemptive scheduling?

## Examples

- GPU device
- Hardware accelerators
- CAN bus



- **Control systems** are sensitive to I/O delay and preemptions

- Simpler resource management policies
- Grants exclusive resource access

- A more accurate estimation of **worst-case execution-time (WCET)**
- More predictable cache

- Reduces context switches
- Avoids intra-task **cache-related preemption delays (CRPD)**

# Why do we need a new analysis?

Most of NPS policies are **not sustainable**  
(w.r.t. execution time variation, release jitter, etc.)

Simulation-based schedulability tests cannot be used



Schedulability analyses for sporadic tasks  
[Jeffay91, Tindel94, Davis07]

Pessimistic for periodic tasks

Not applicable to arbitrary job sets

Existing analyses are not enough



Existing schedulability analyses based on model checking, timed automata, abstraction refinements, etc.

[Sun97, Baker07, Guan07, Bonifaci10, Burmyakov15, Stigge15]

Not very scalable

Existing analyses are not efficient



Many non-work-conserving scheduling algorithms do NOT have a schedulability analysis yet

No solution yet



# What do we want?

An **efficient, exact, general**  
**schedulability analysis**

THAT includes

a wide class of scheduling algorithms and task models





# Agenda

- ▶ **Main idea:**

Searching all possible schedules **efficiently** and **accurately**

- ▶ Constructing the search graph
- ▶ Evaluation
- ▶ Conclusion

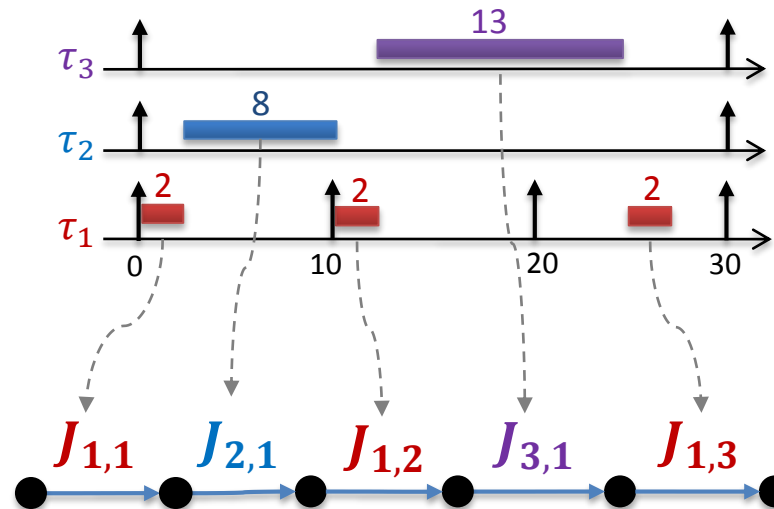
# Basic scenario: no runtime variation in the workload

Task	Period	Execution time
$\tau_3$	30	13
$\tau_2$	30	8
$\tau_1$	10	2

Non-preemptive fixed-priority scheduling

One schedule

One job ordering



schedulable

Both existing tests for sporadic tasks **reject** this task set [Jeffay91, Davis07]

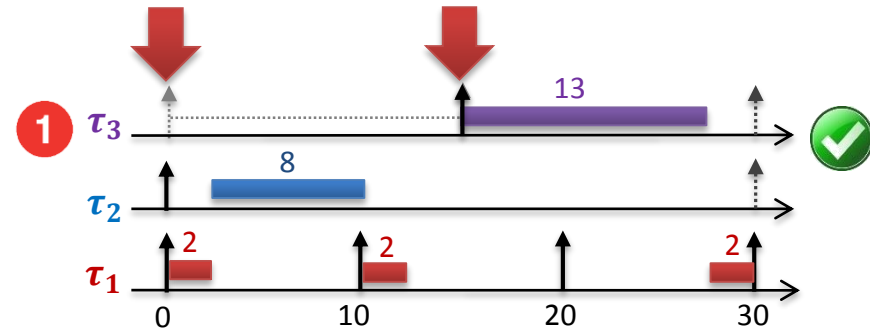
Values are integer.

Scheduling algorithm: Non-preemptive fixed-priority (NP-FP)

A schedule is an assignment of execution intervals to the jobs.

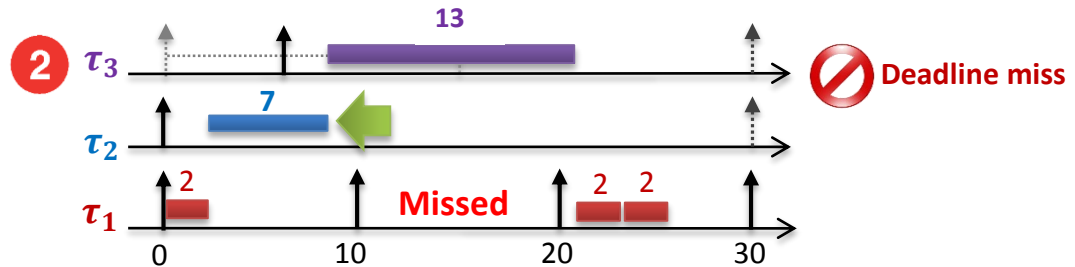
# Scenario: execution time variation and release jitter

Task	Period	Execution time		Release jitter
		Min	Max	
$\tau_3$	30	[3, 13]		15
$\tau_2$	30	[7, 8]		0
$\tau_1$	10	[1, 2]		0

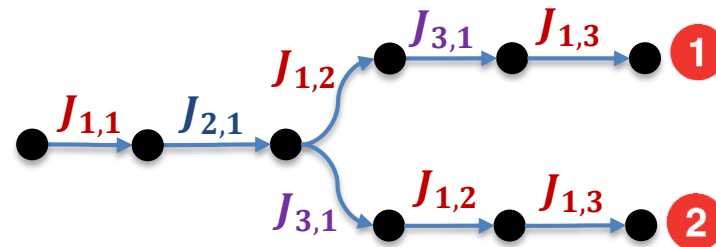


**More than 100**  
different schedules

**Not schedulable**



**Only two**  
different job orderings



A graph of **job orderings**

Values are integer.

Scheduling algorithm: NP-FP

A schedule is an assignment of execution intervals to the tasks.



# Challenges

Due to scheduling anomalies



For an exact analysis, we need to consider  
**all possible execution scenarios**

## Observation

There are fewer permissible  
**job orderings** than **schedules**

## Research question

Is there a way to use  
**job-ordering abstraction**  
to analyze schedulability?

How to **abstract**  
schedules in a graph  
of job orderings?

How to **efficiently**  
find all  
job orderings?

How to identify  
**timing violations** in  
the resulting graph?

# Abstracting schedules in a graph of job orderings

## Requirement

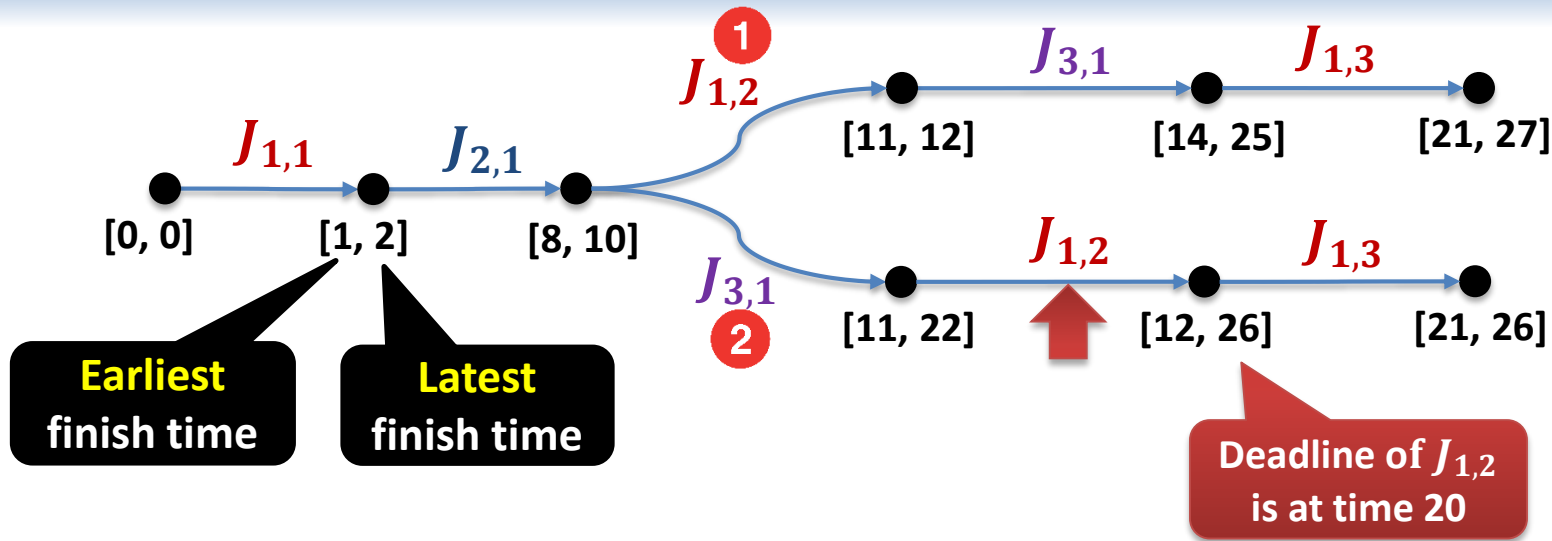
Knowing when a job misses its deadline

## Solution

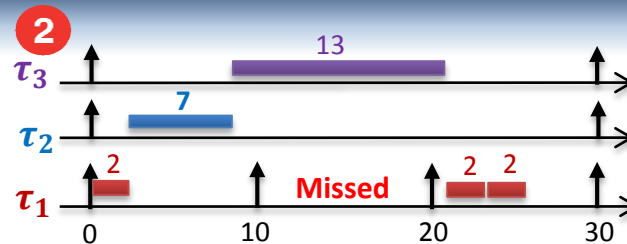
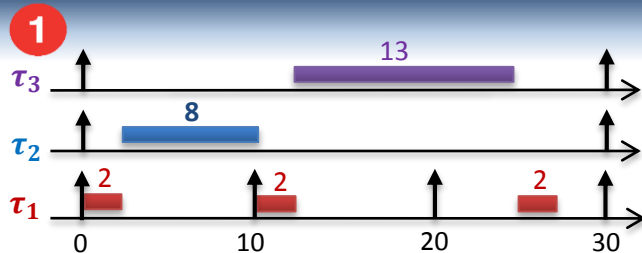
Encode the **earliest** and **latest finish time** of a job

## Verification of schedulability

Check if the **latest finish time** is not larger than the deadline



Each path shows a job ordering

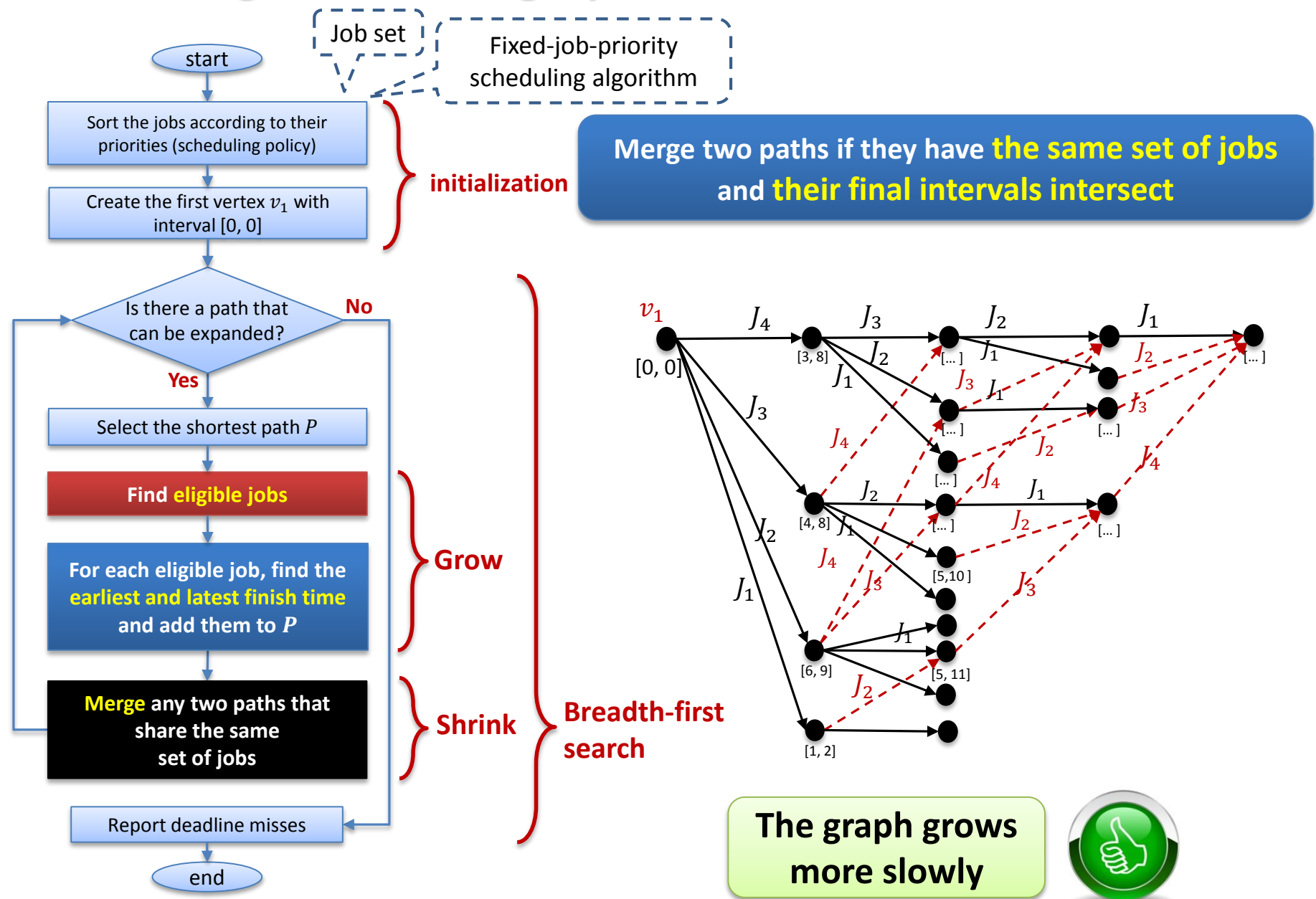


Task	Period	Execution time		Jitter
		Min	Max	
$\tau_3$	10	[3, 13]		15
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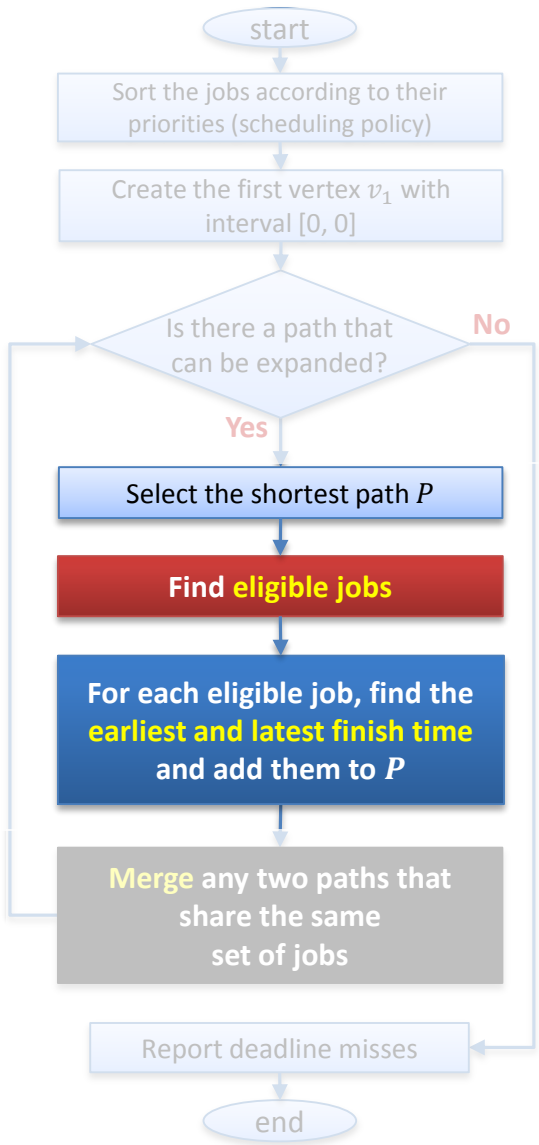
# Agenda

- ▶ Main idea: Searching all possible execution scenarios efficiently and accurately
- ▶ **Constructing the search graph**
- ▶ Evaluation
- ▶ Conclusion

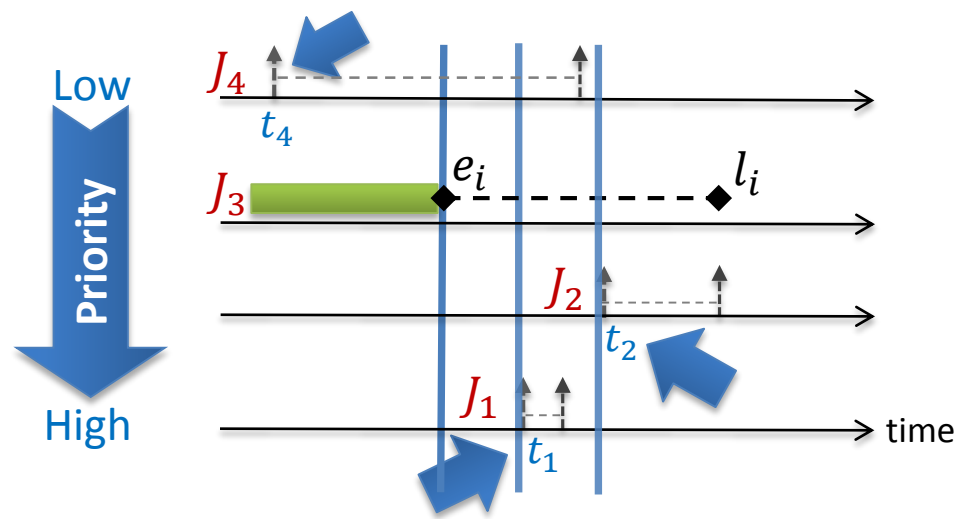
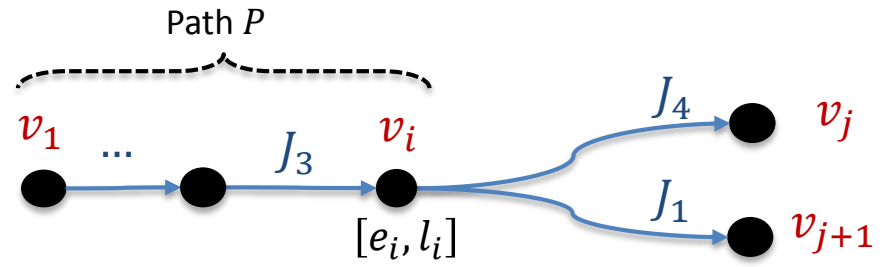
# Constructing the search graph



# Growing the graph



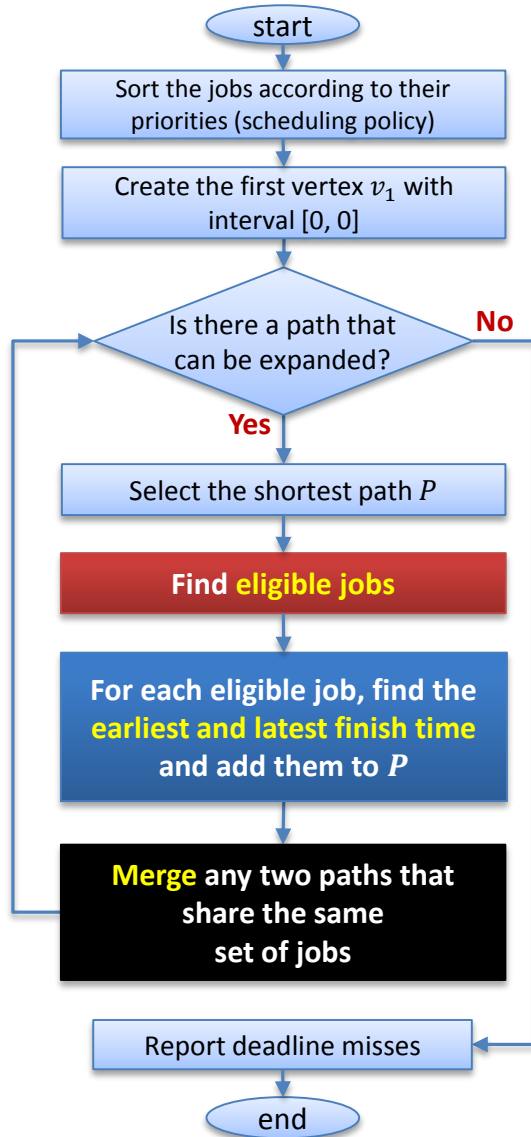
An **eligible job** for path  $P$  is a job that can be scheduled after  $P$  in **at least one execution scenario**



$e_i$  = the earliest finish time of path  $P$   
 $l_i$  = the latest finish time of path  $P$



# Requirements of an exact analysis



**“Eligibility conditions”** are necessary and sufficient

The **“final interval”** of each is exact:  
For any time  $t$  in the interval, there must be an execution scenario that ends at  $t$

Final intervals remain “exact” after **merging process**

In our work, we have proved these properties for

- Fixed-job-priority scheduling algorithms
- Tasks with release jitter and execution time variation
- Hard and soft timing constraints
- Work-conserving and non-work-conserving scheduling algorithms



# How to apply the analysis to a new system or algorithm?

**1**

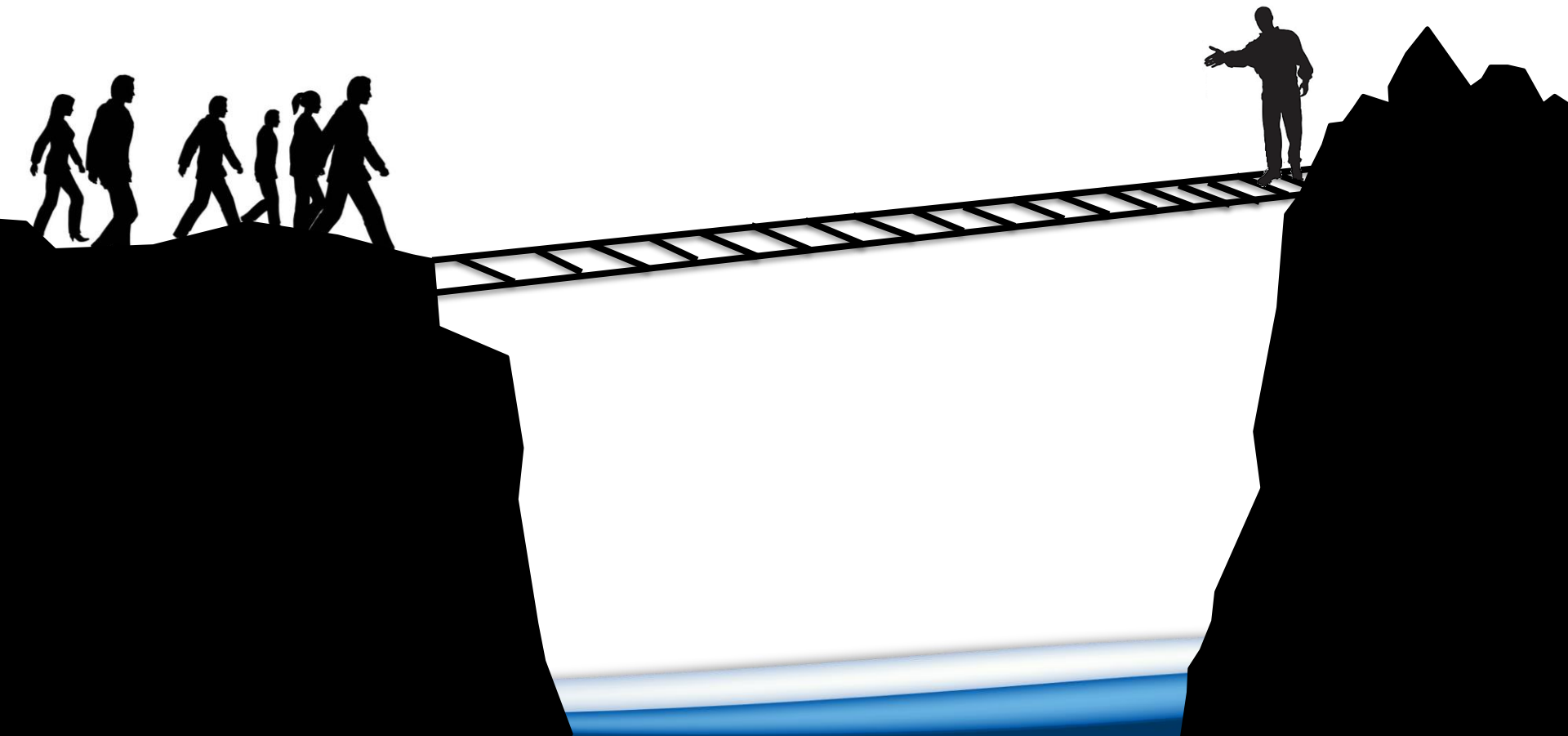
Define eligibility  
conditions

**2**

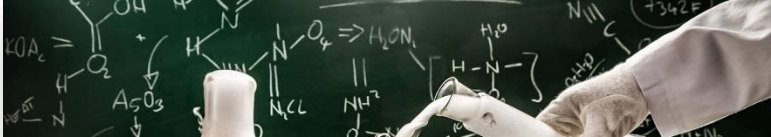
Define how to obtain  
the final intervals

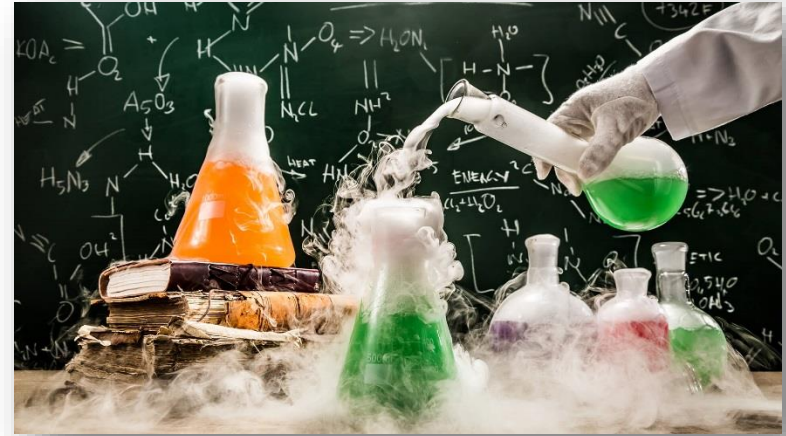
**3**

Prove the aforementioned  
properties



# Agenda

- ▶ Main idea: Searching all possible execution scenarios efficiently and accurately
  - ▶ Constructing the search graph
  - ▶ **Evaluation**
  - ▶ Conclusion
- 
- A photograph of a chemistry experiment. A hand is pouring a green liquid from a graduated cylinder into a round-bottom flask. The flask is placed on a stand over a Bunsen burner, and a large amount of white foam is erupting from the top. In the background, a chalkboard is covered with various chemical equations and diagrams, including
- $\text{H}_2\text{N}_2$
- ,
- $\text{H}_2\text{O}$
- ,
- $\text{H}_2\text{O}_2$
- , and
- $\text{H}_2\text{O}_3$
- .



# Main questions

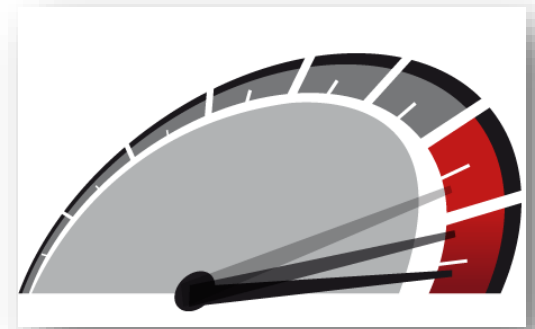


- ▶ Is our analysis **effective**?

- Does it actually improve the accuracy of schedulability analysis?
- What is our **achievement** for **non-work-conserving** scheduling policies?

- ▶ Is our analysis **efficient**?

- How fast is the analysis?



# Evaluation setup

## Automotive benchmark task sets [Kramer15]

## Synthetic task sets

No jitter

Small jitter  
(up to 100 microseconds)

Large jitter  
(up to 20% of the period)

- Variable parameter: **utilization**
- Generate **runnables** according to [Kramer15] until the given utilization is reached
- Pack a random number of runnables together to build a **task**
- Up to 30 tasks per task set

- Variable parameter: **maximum number of jobs in a hyperperiod**
- Periods are from [1, 1000]ms with log-uniform distribution
- Up to 50% runtime variation in the execution time
- 10 tasks per task set

**To evaluate the effectiveness**  
in a realistic setup and different utilization values

**To evaluate the efficiency**  
when there are a large number of jobs

[Kramer15] S. Kramer, D. Ziegenbein, and A. Hamann, "Real world automotive benchmark for free," in WATERS, 2015.  
**Note:** only task sets that pass the necessary schedulability condition of non-preemptive scheduling were considered.

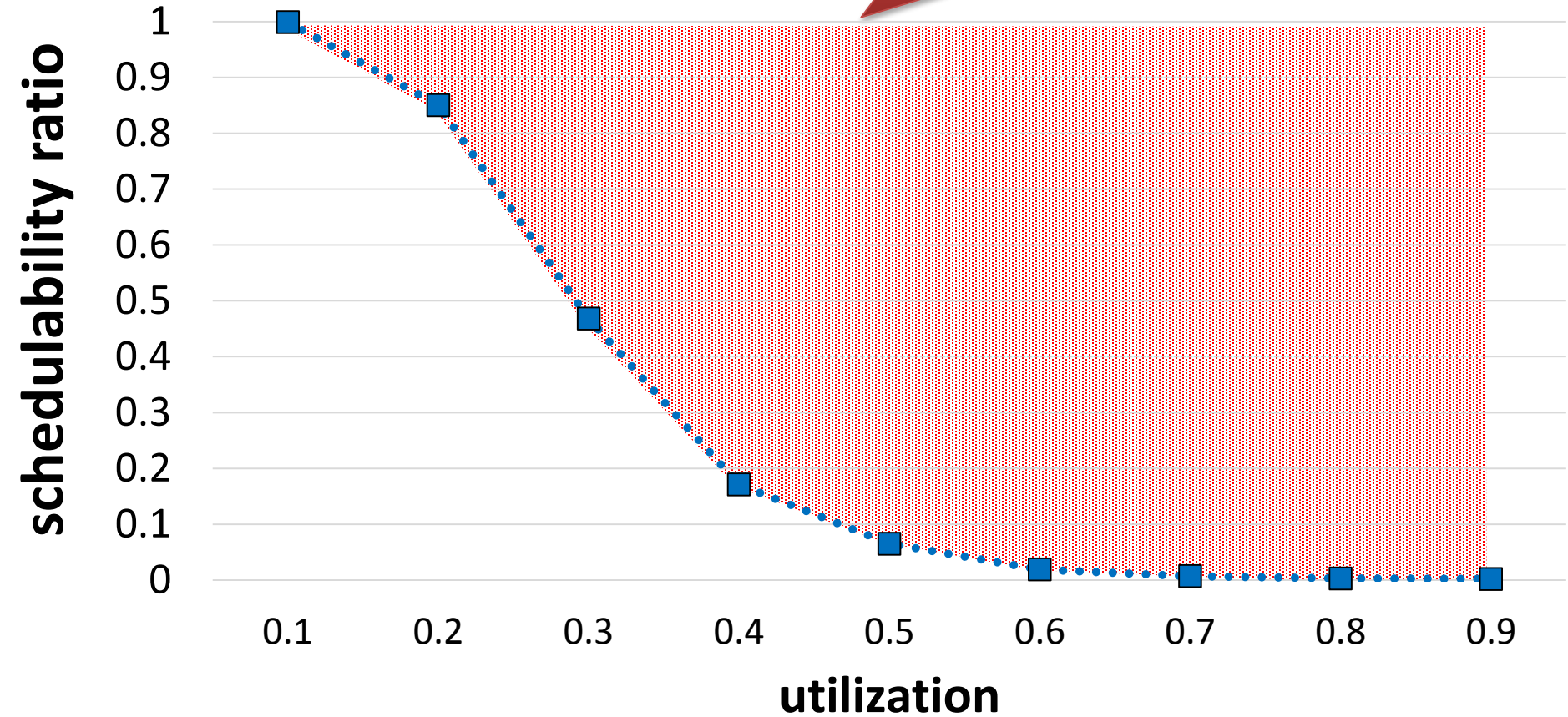


# How effective is our schedulability analysis?

Automotive  
benchmark, no jitter

•■• NP-FP classic test

Many task sets do not pass the test



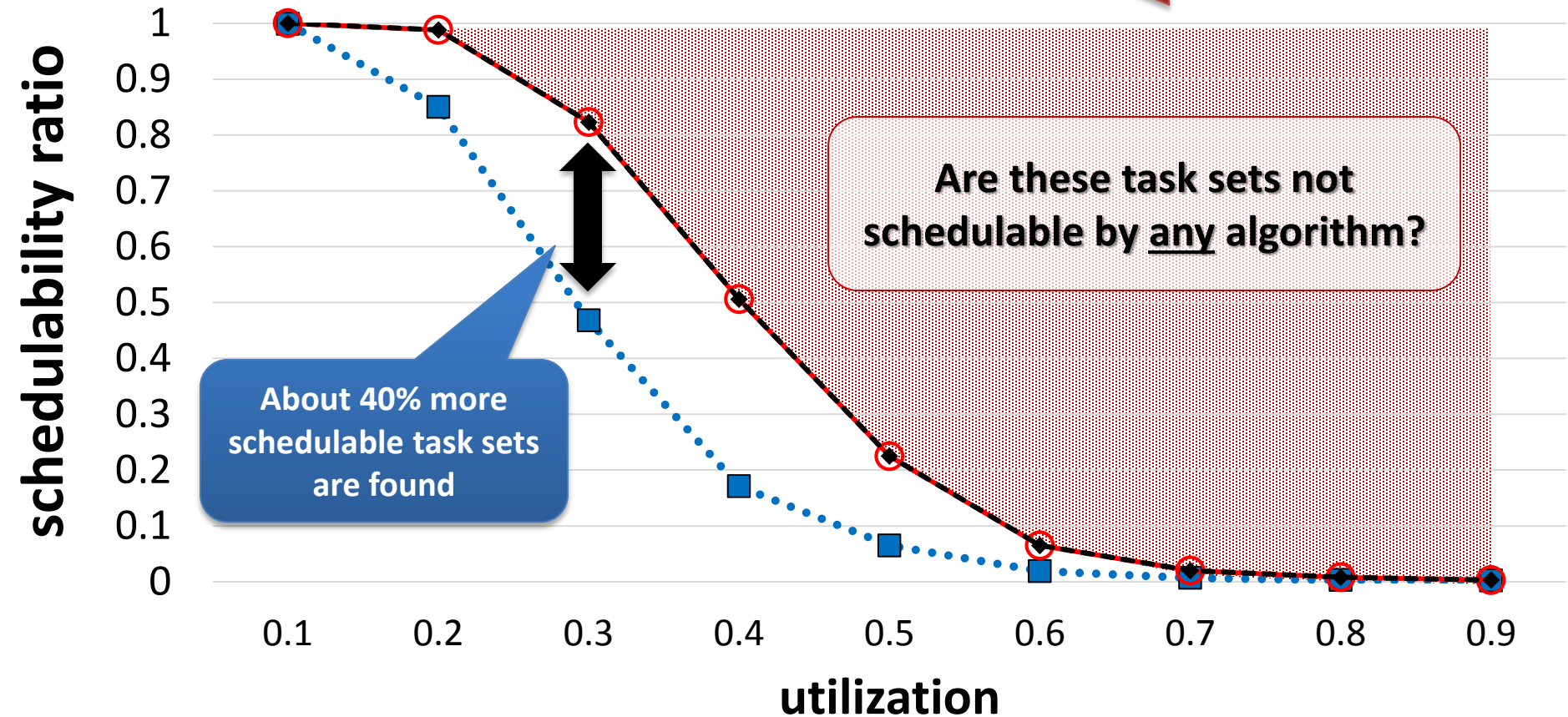
Task sets in this experiment have up to 35 tasks and 3500 jobs

# How effective is our schedulability analysis?

Automotive  
benchmark, no jitter

- NP-FP classic test
- This paper: NP-EDF
- ◆ This paper: NP-FP

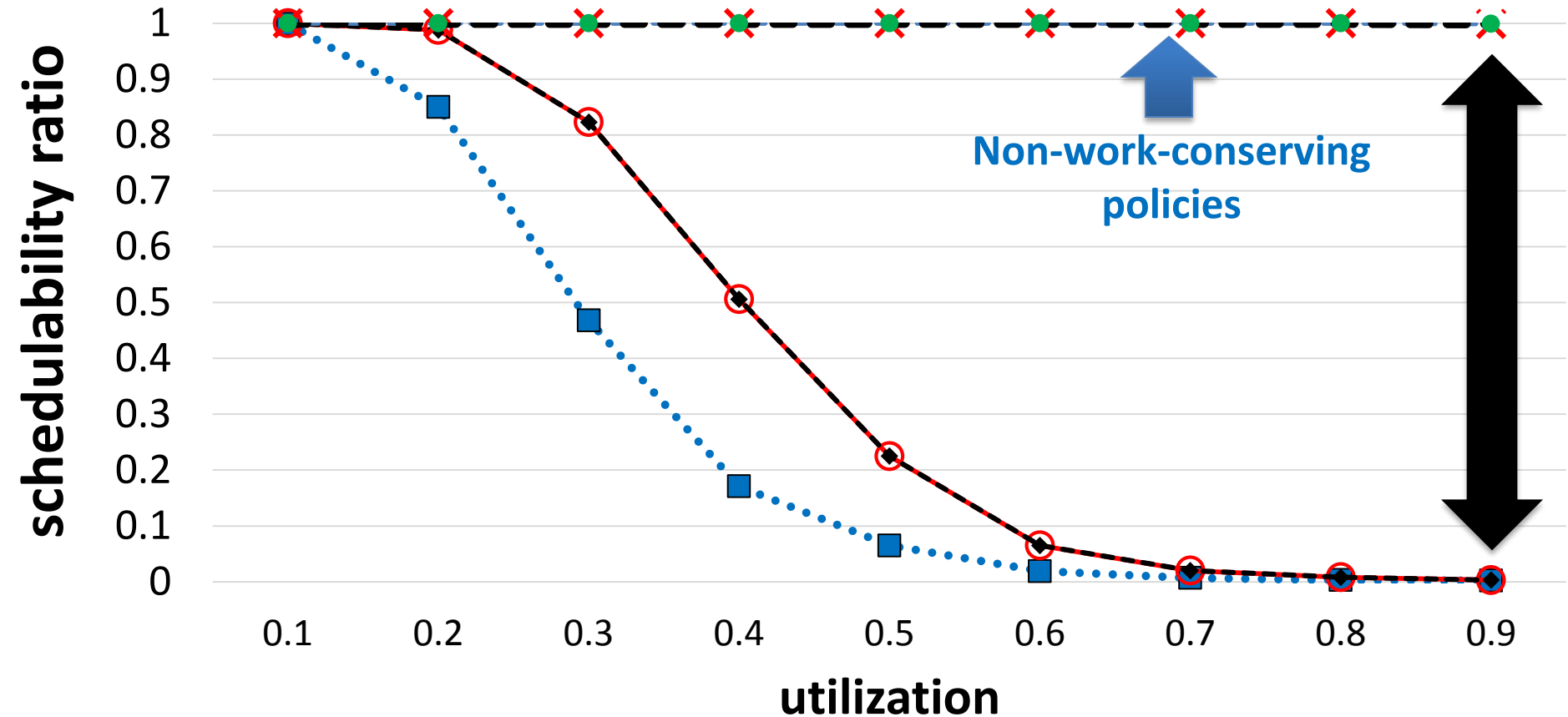
Still, many task sets  
are not schedulable



# How effective is our schedulability analysis?

Automotive  
benchmark, no jitter

- NP-FP classic test
- This paper: NP-EDF
- This paper: NP-FP
- This paper: Precautious-RM
- This paper: CW-EDF+



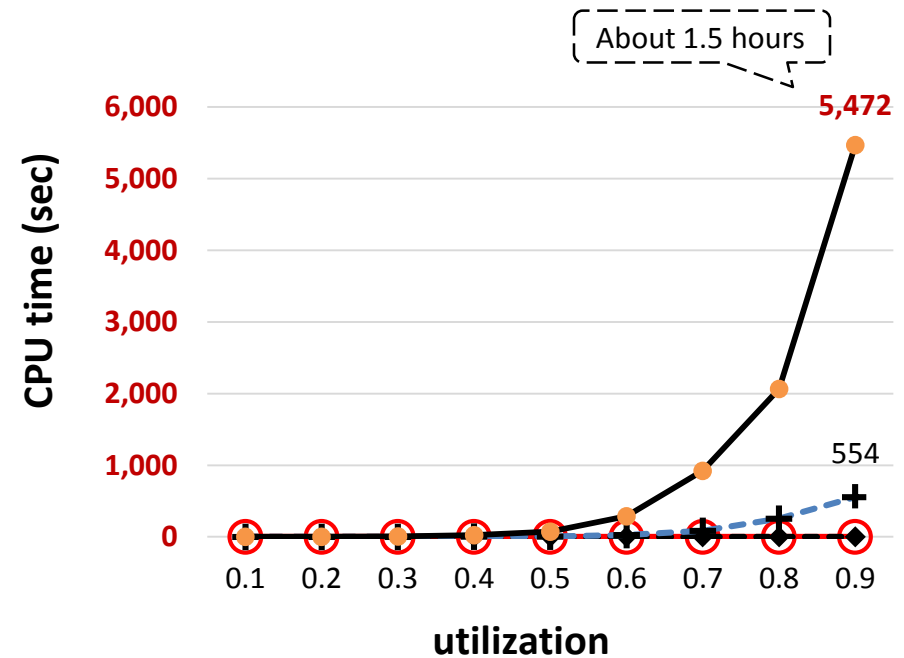
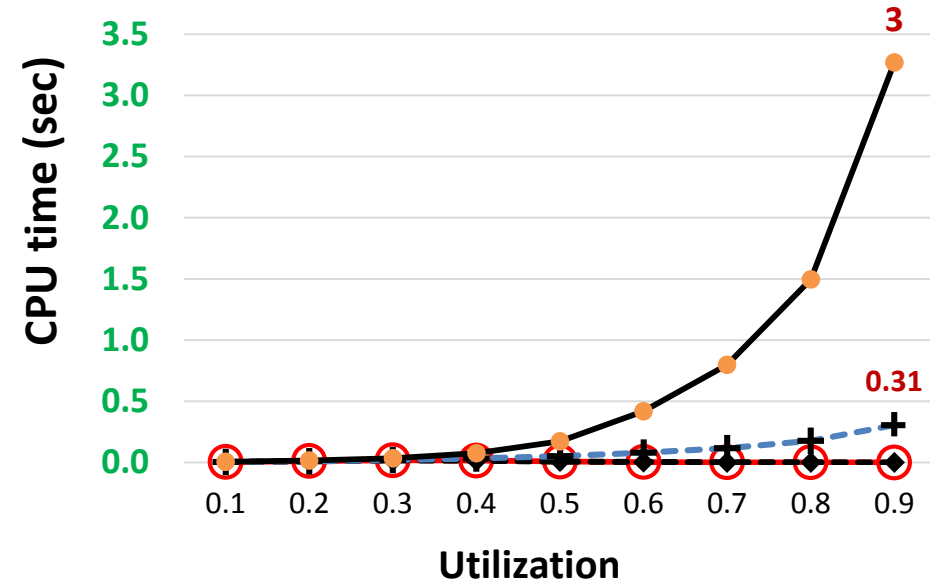
# How efficient is our schedulability analysis?

Automotive benchmark

No jitter

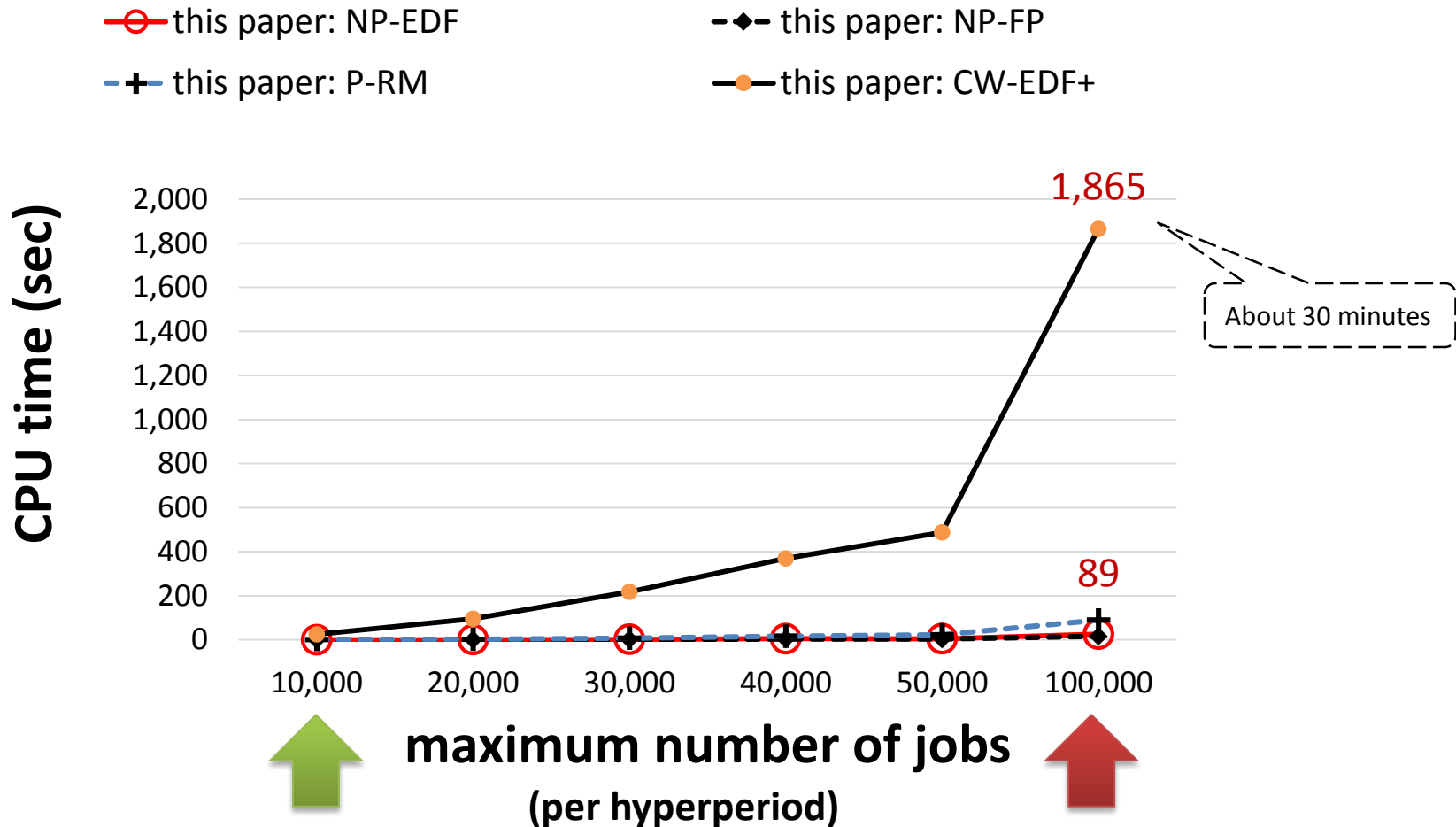
Large jitter

- This paper: NP-EDF
- ◆— This paper: NP-FP
- +— This paper: Precautious-RM
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# How efficient is our schedulability analysis?

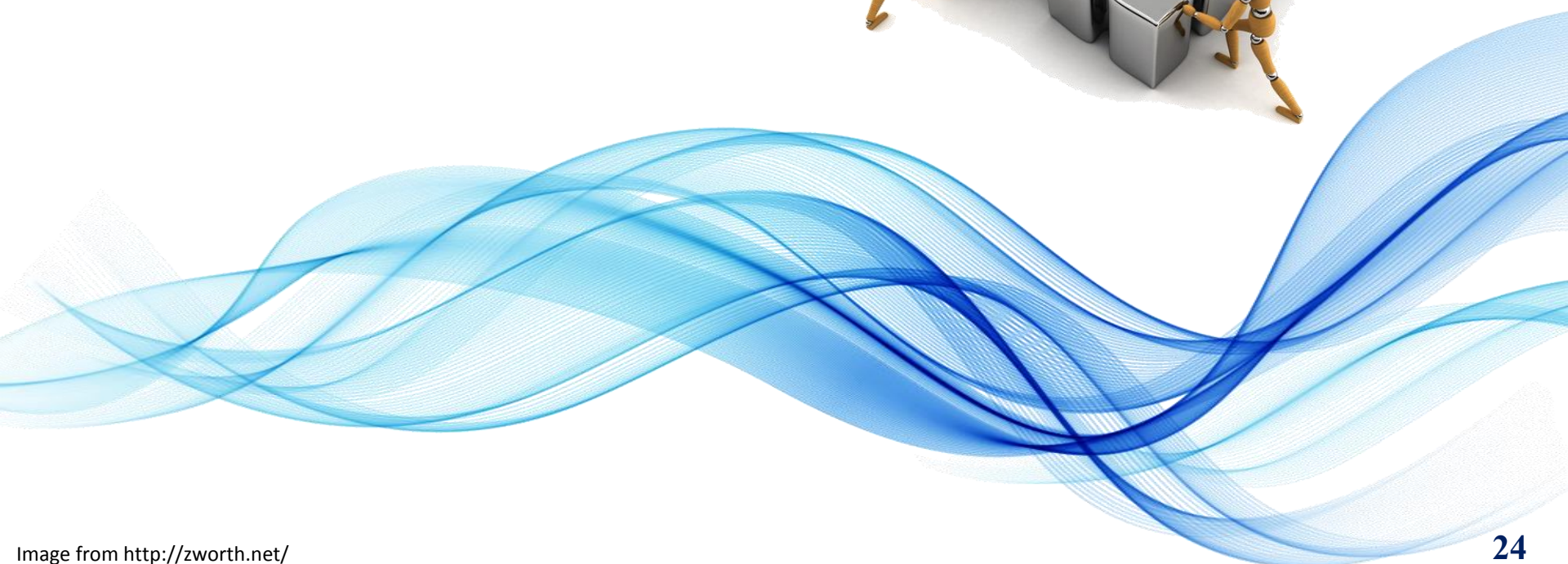
Synthetic tasks  
Small jitter





# Agenda

- ▶ Main idea: Searching all possible execution scenarios efficiently and accurately
- ▶ Constructing the search graph
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- ▶ **Conclusion**



# Conclusion

## Goal



An **efficient**, **exact**, and **general** schedulability analysis for a wide class of scheduling algorithms

## Solution



Constructing a precise **abstraction** of all possible schedules

## Method



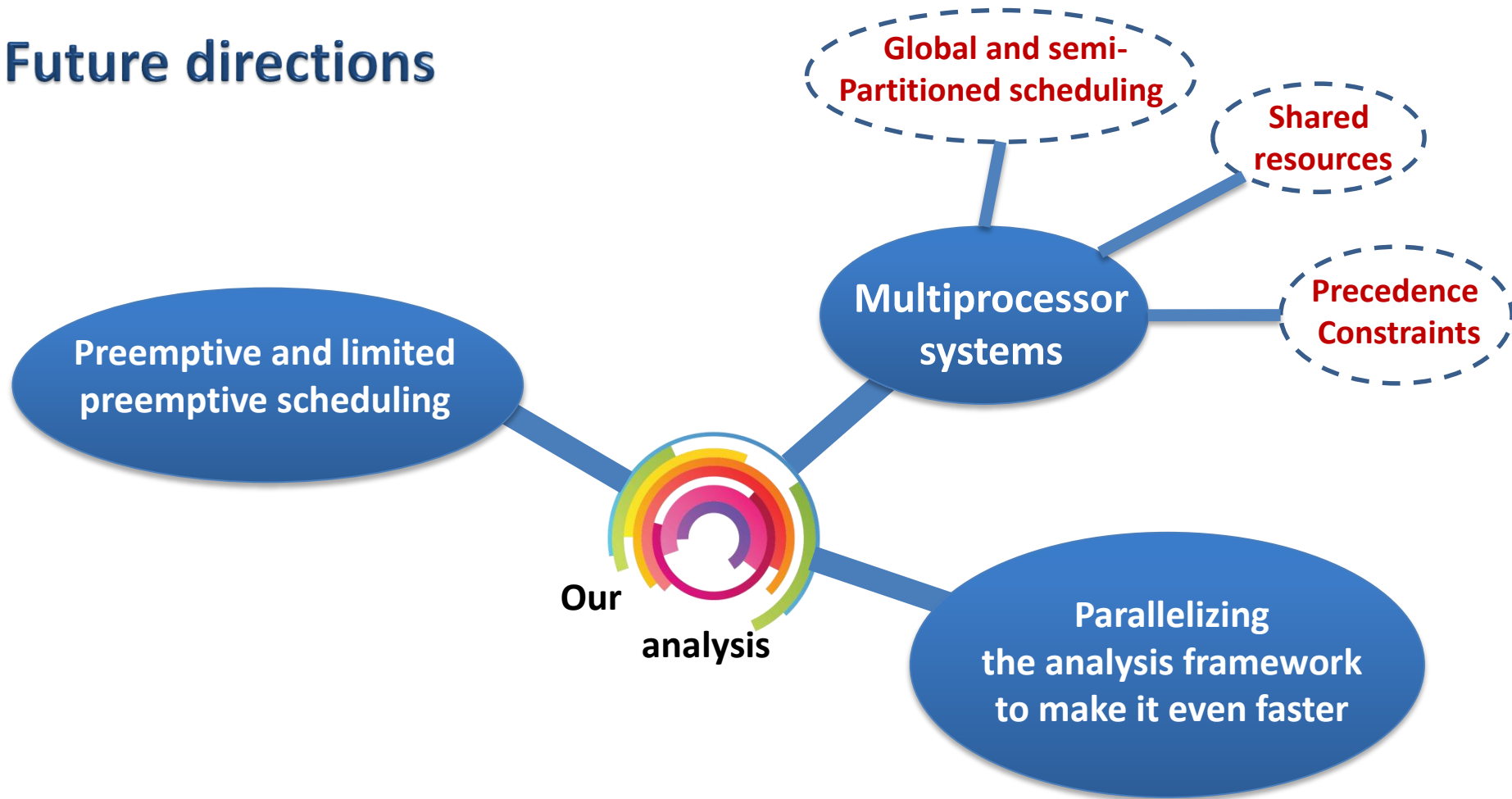
Building a **schedule-abstraction** graph based on job ordering

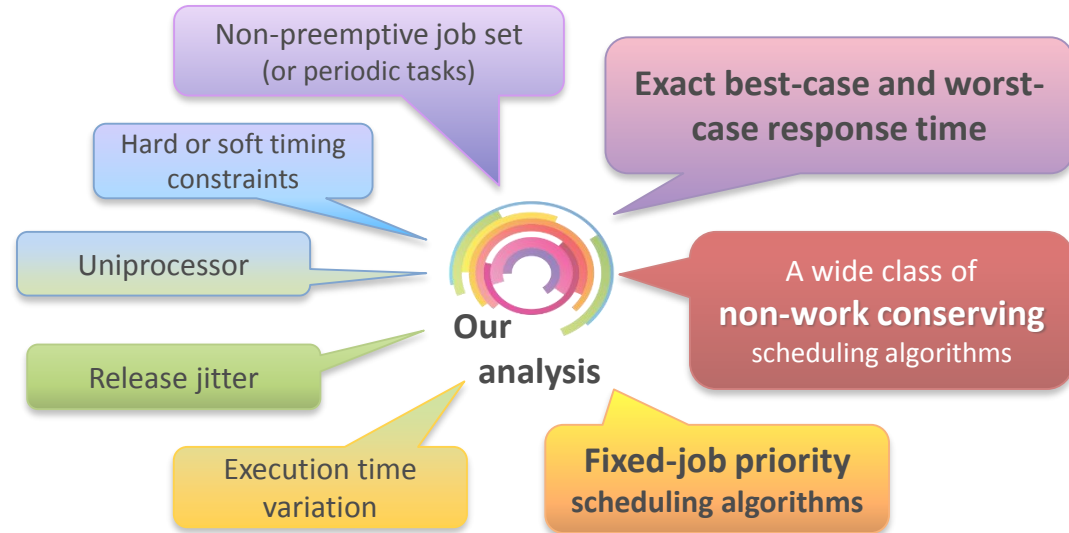
## Key idea



An **efficient merge** technique to defer the state-space explosion

# Future directions





Source code available at

<https://people.mpi-sws.org/~bbb/papers/details/rtss17/index.html>

*Thank you*

