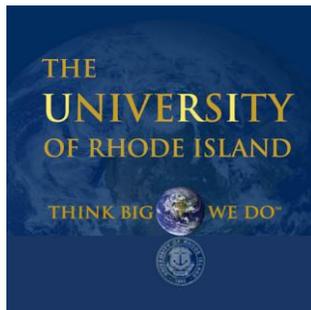




ACM Symposium  
on Cloud Computing

# Achieving Low Latency in Public Edges by Hiding Workloads Mutual Interference

Weiwei Jia, Jiyuan Zhang, Jianchen Shan, Jing Li, Xiaoning Ding



The University of Rhode Island



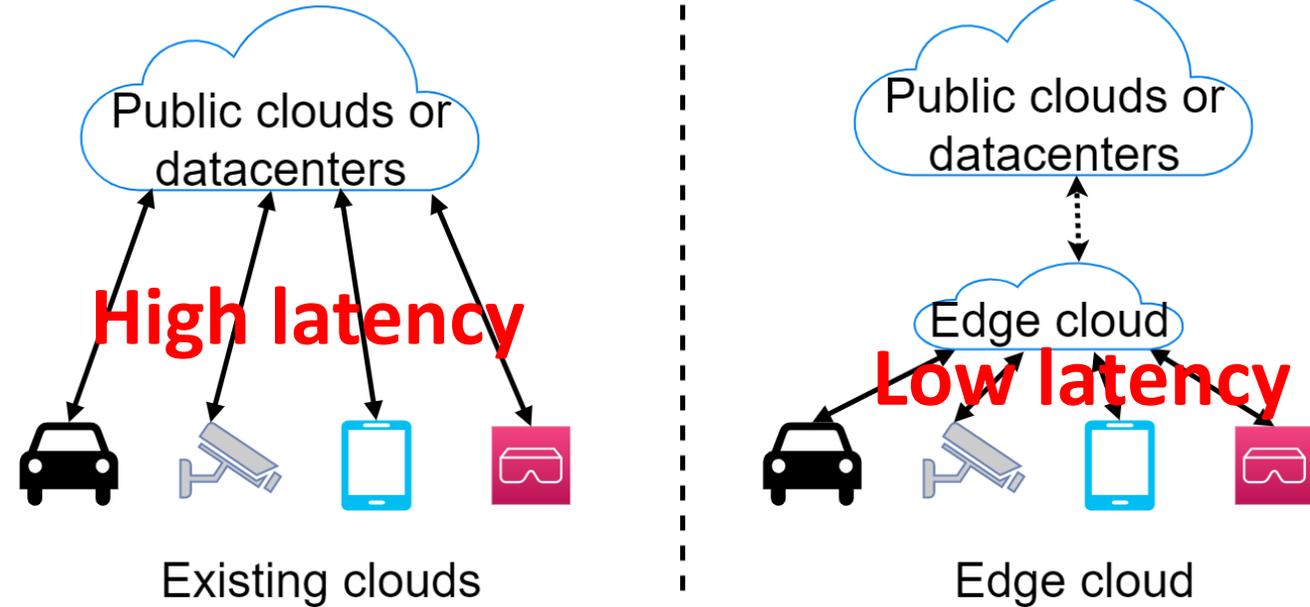
New Jersey Institute of Technology



Hofstra University

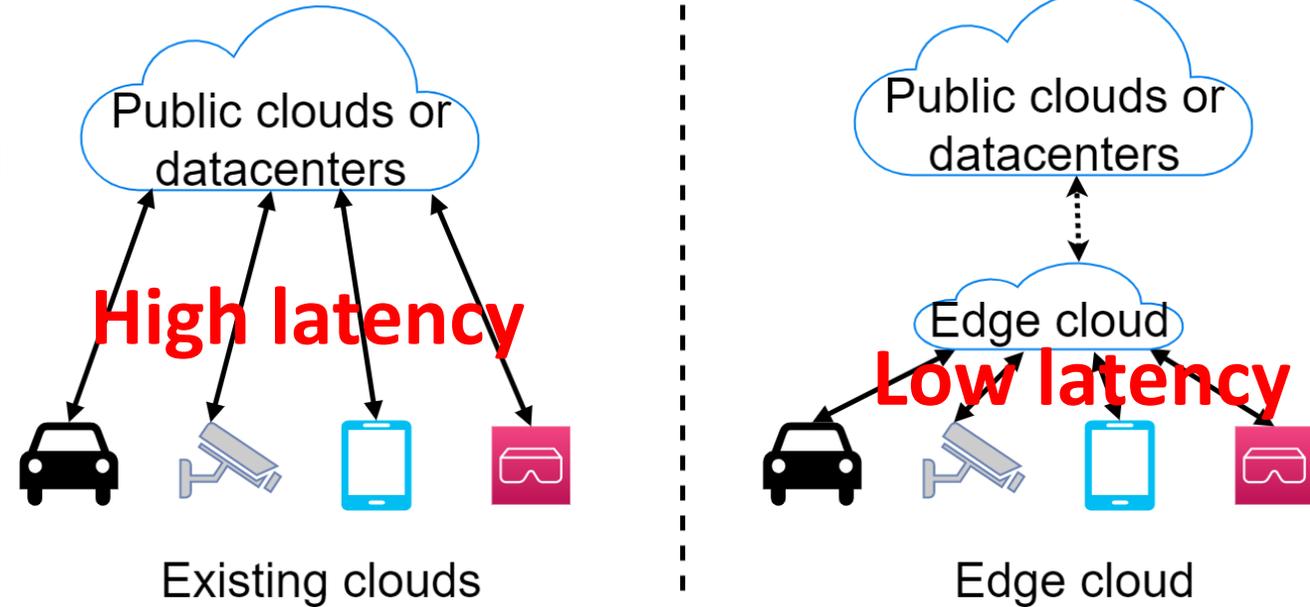
# Background: edge cloud and its applications

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  - Low latency.
  - E.g., AWS local zones.



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- Edge cloud is dominated by **latency sensitive apps** that require **low latency** and **high resource demanding**.
  - E.g., microseconds for autonomous vehicles/robots and AR/VR.
  - Edge applications usually generate large amount of data for computation.
    - Resource-constrained end devices cannot afford the computation.

# Resource sharing increases edge app mutual interference and latency

- Multiple applications from different users share the same edge server.
  - Execution of an app contends resources and interferes with execution of other apps.
    - Interference may cause significant performance penalties if not well controlled.

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- Resource contention and apps **mutual interference are unavoidable in edge cloud!**
  - **Limited number of edge servers** host large number of apps.
    - Hard to effectively distribute and separate interfering apps across different servers.
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  - **Limited number of edge servers** host large number of apps.
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  - Resource **usage patterns** of edge applications may change dynamically.
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- Problem: how to efficiently schedule latency sensitive apps to **reduce their mutual interference and latencies** in edge cloud.
  - This problem is under-studied in edge cloud scenarios.

# Existing cloud solutions are not effective for edge cloud

- Resource over-provisioning (Mengwei Xu et. al.[IMC '21]).
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  - Hard to predict which apps may interfere with each other due to app execution dynamics.
- Resource partitioning (Heracles[ISCA '15]) can barely help due to app dynamic resource usage.
  - Infrequent adjustment of resource partitions may not help.
    - Resources are not adapted to resource demand of the workload.
  - Frequent adjustment may lead to high overhead.

# Outline

- Problem: how to schedule latency sensitive apps to reduce their mutual interference and latencies in edge cloud.
- ✓ DASEC: dynamic asymmetric scheduling for edge computing.
  - Basic idea: move the interference off the tasks on the critical paths of the workloads.
  - Key issues and solutions.
- Evaluation.
  - DASEC has been implemented in Linux/KVM, Linux CFS, and Google user-level scheduler (i.e., ghost [SOSP '21]).
  - Compared to vanilla Linux/KVM, DASEC reduces mean latency and 99<sup>th</sup> tail latency by 46% and 52%, respectively.

# Basic idea: move interference off the tasks on the critical paths of the workloads

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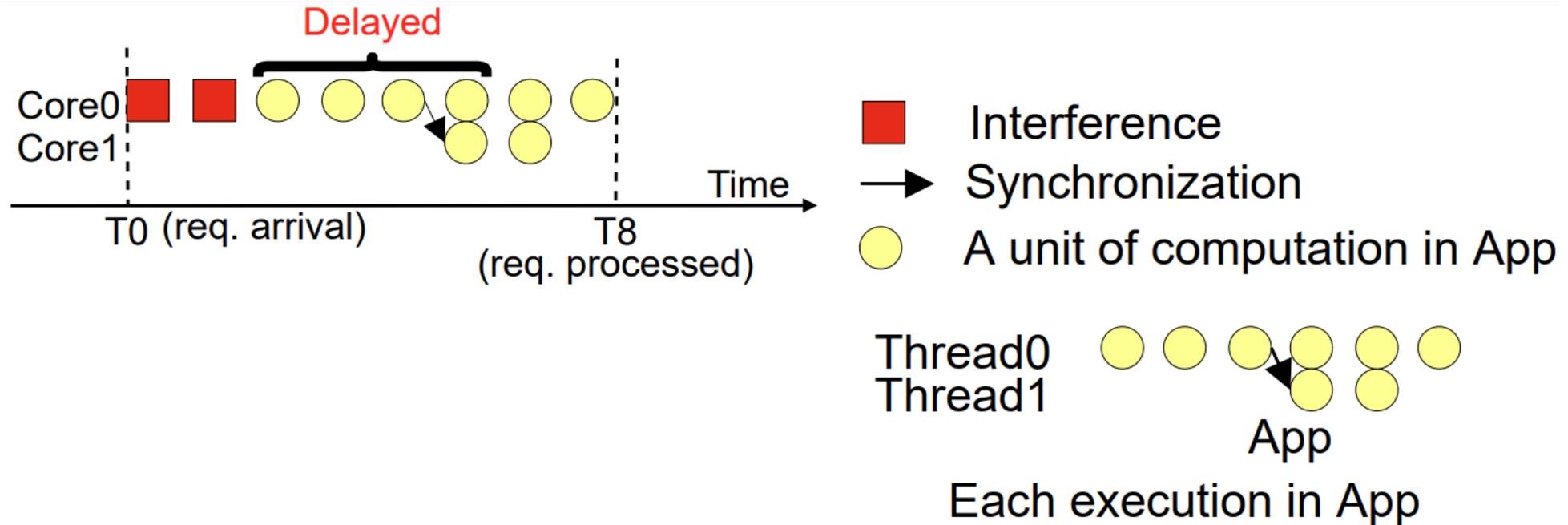
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  - CPU resources, as the most important resource type, have the largest impact on perf.

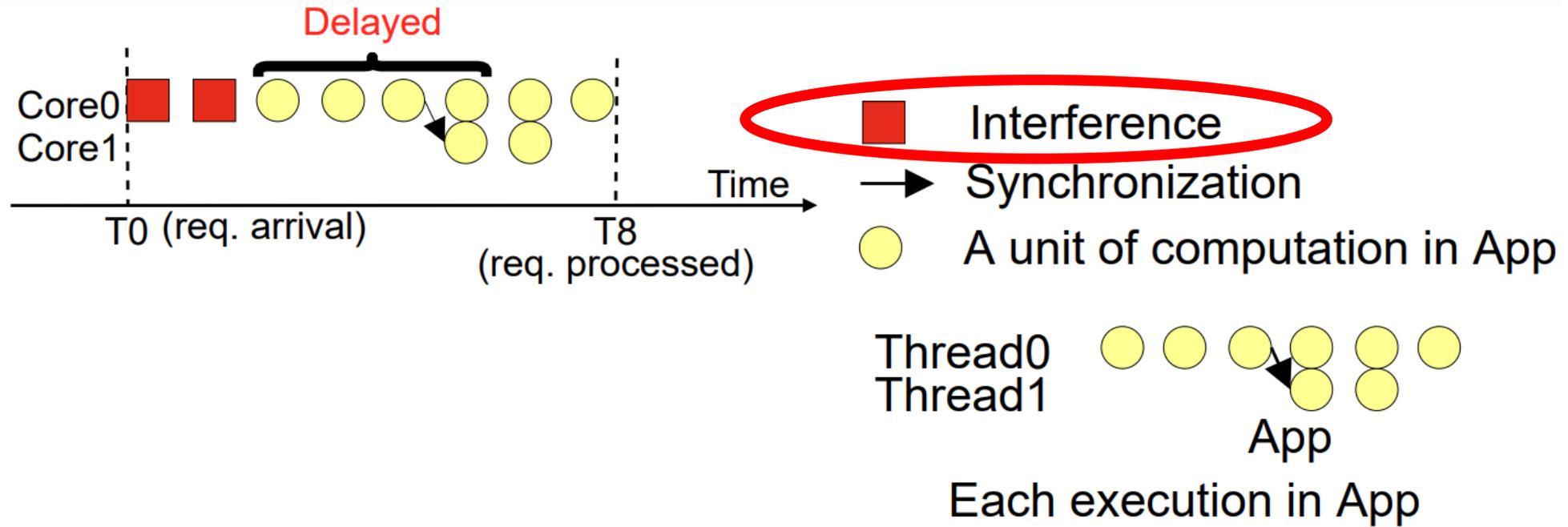
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- This work focuses on the interference caused by sharing CPU cores.
  - CPU resources, as the most important resource type, have the largest impact on perf.
- Tasks in workloads **interfere with each other in three ways.**
  - Tasks on app's critical path are **delayed**, **interrupted**, or **lack CPU share**.

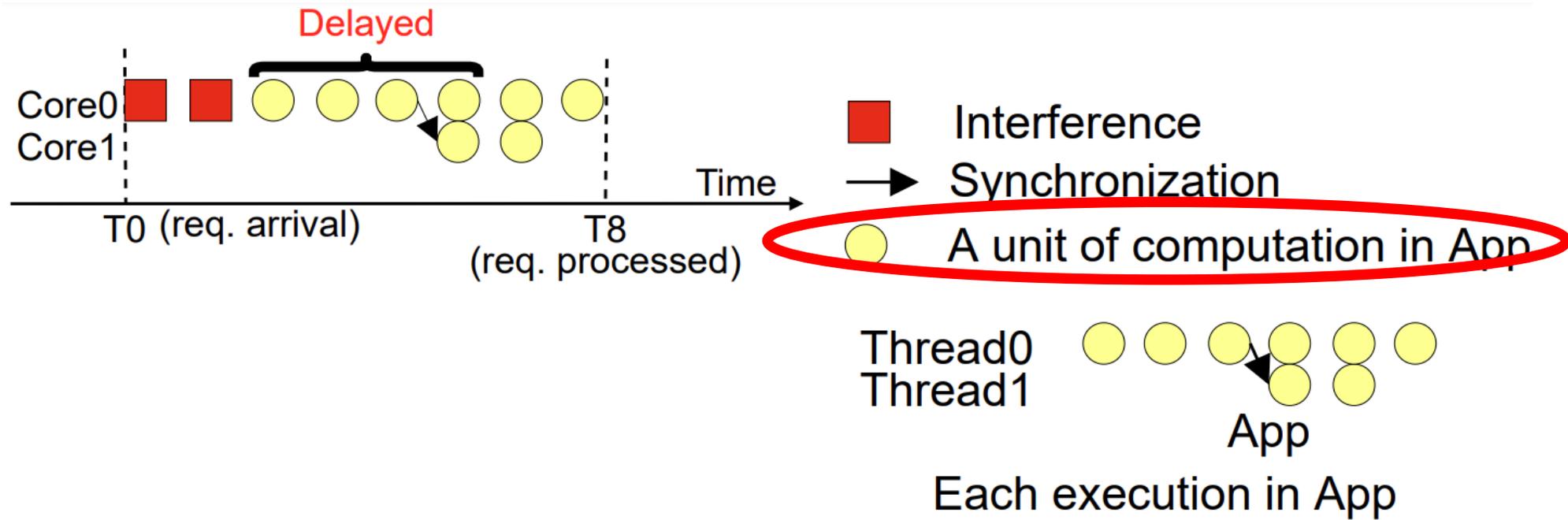
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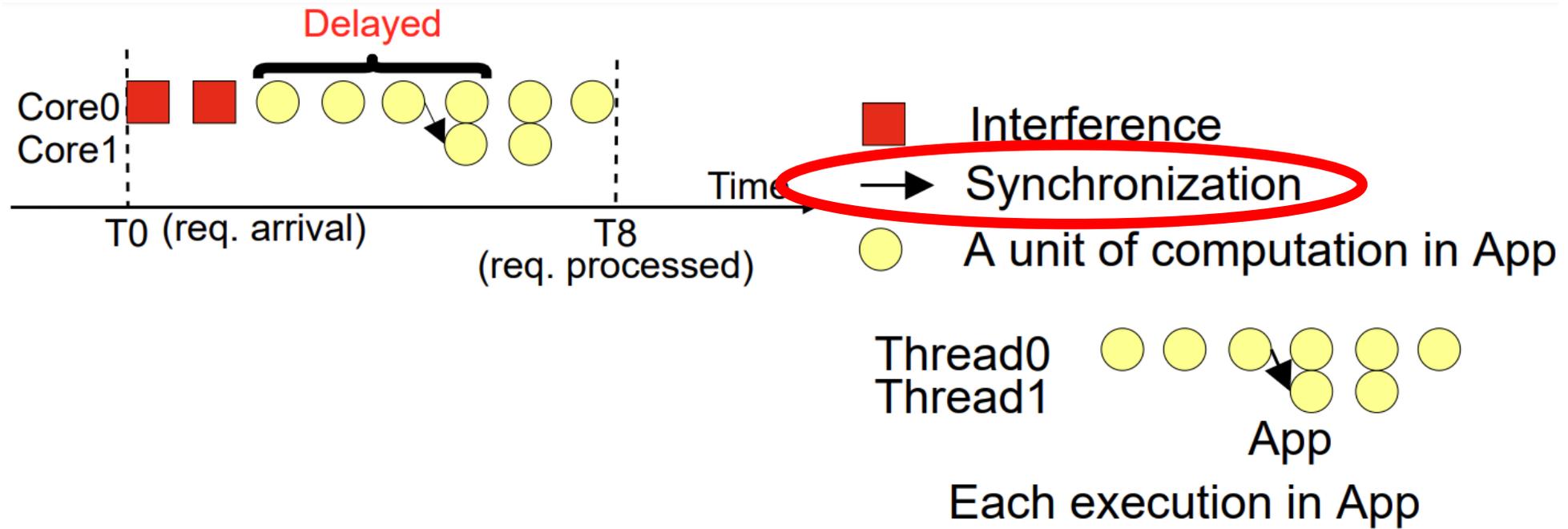
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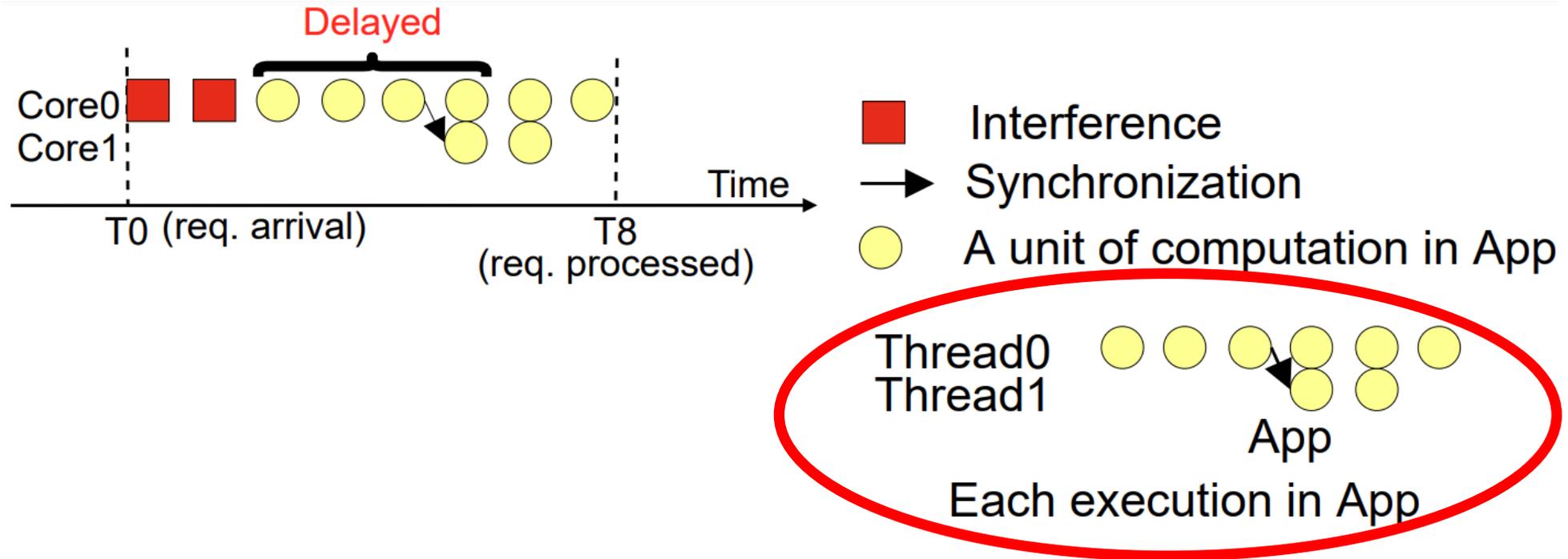
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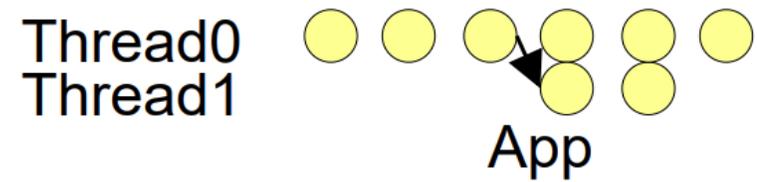
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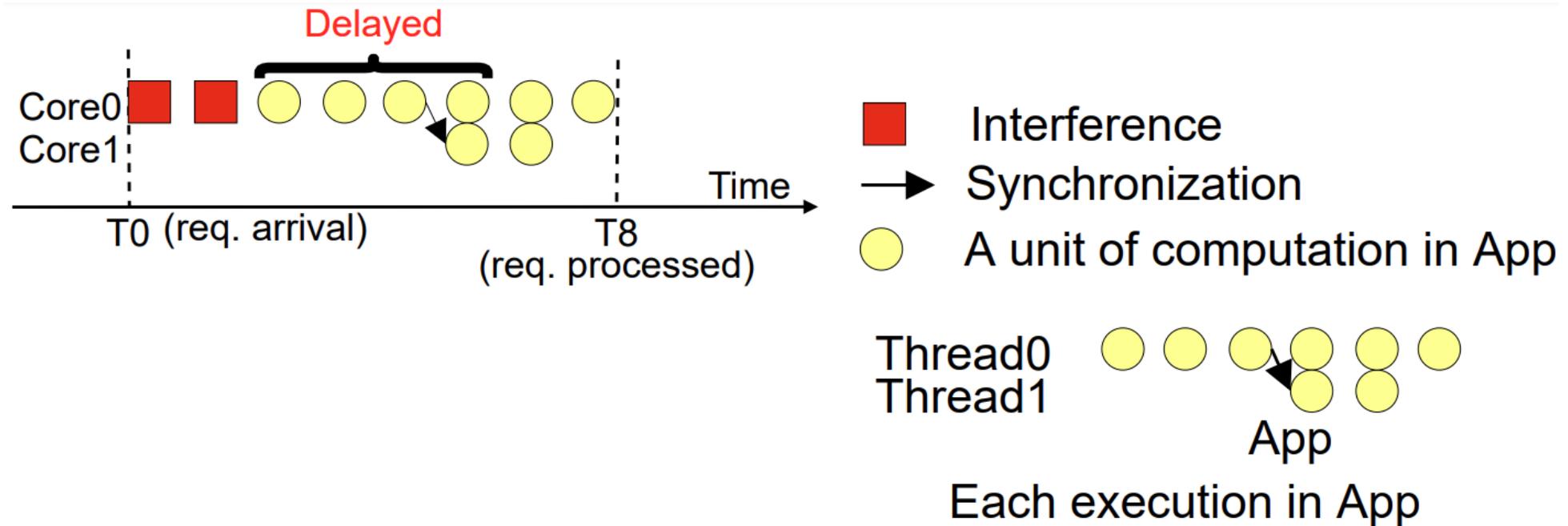


App is ready to run at T0



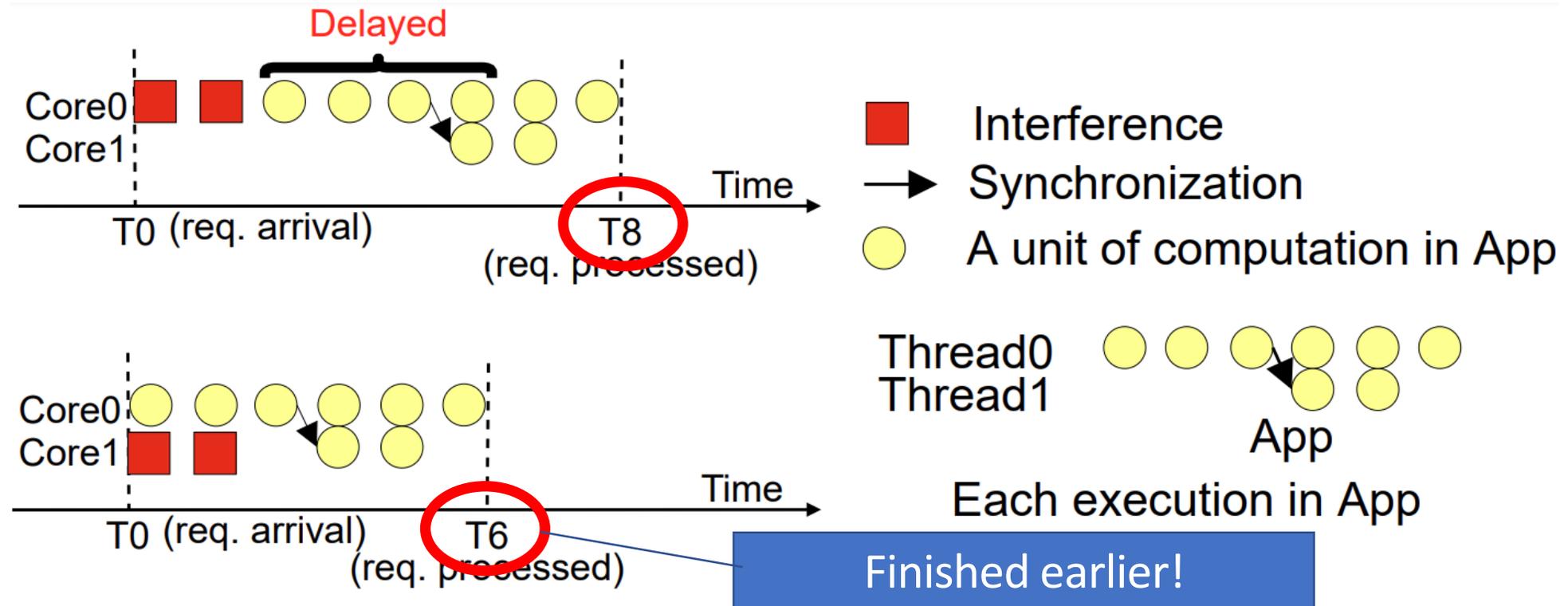
Each execution in App

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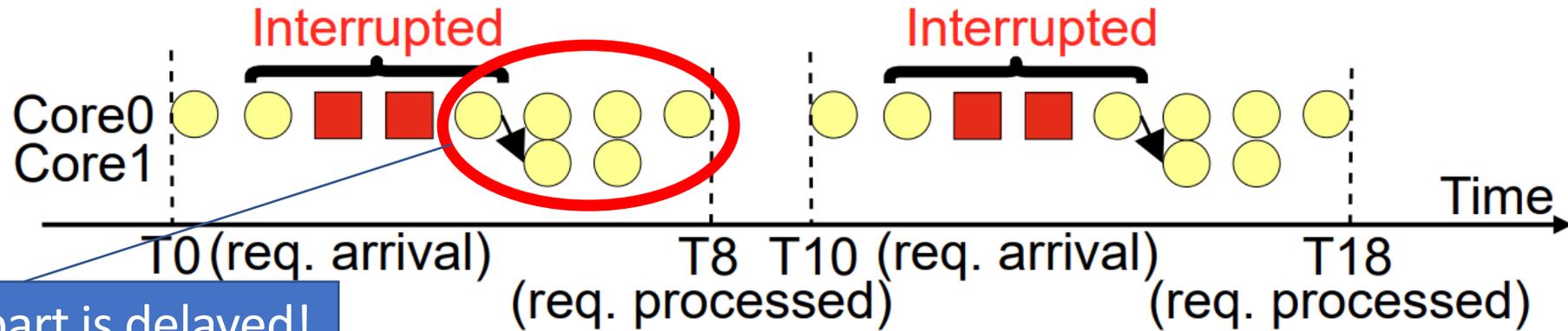
- Detection: at the end of previous time period, a thread/process was in “ready” or “running” status and its timeslice was not used up.

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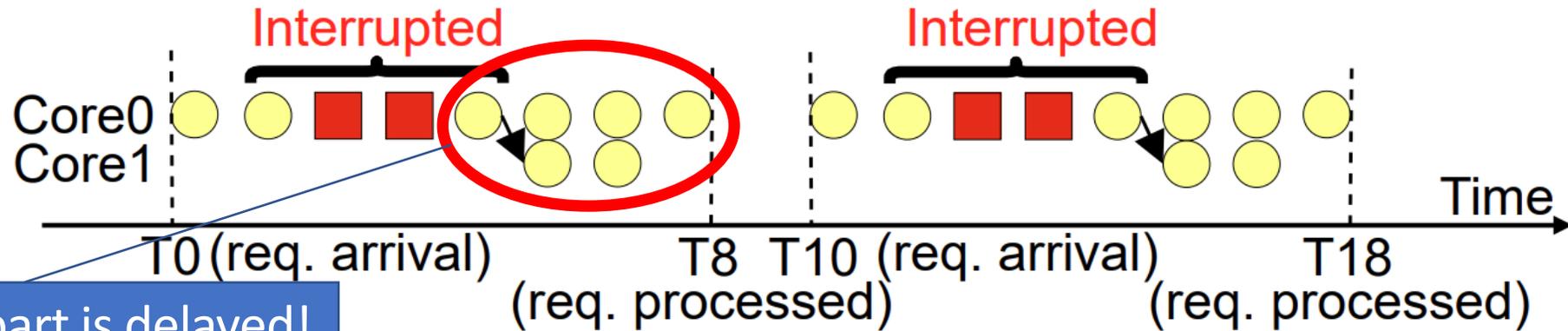
- Detection: at the end of previous time period, a thread/process was in “ready” or “running” status and its timeslice was not used up.
- Solution: reduce rescheduling latency of the thread/process to let it start early.

# Interference #2: tasks on app's critical path are interrupted



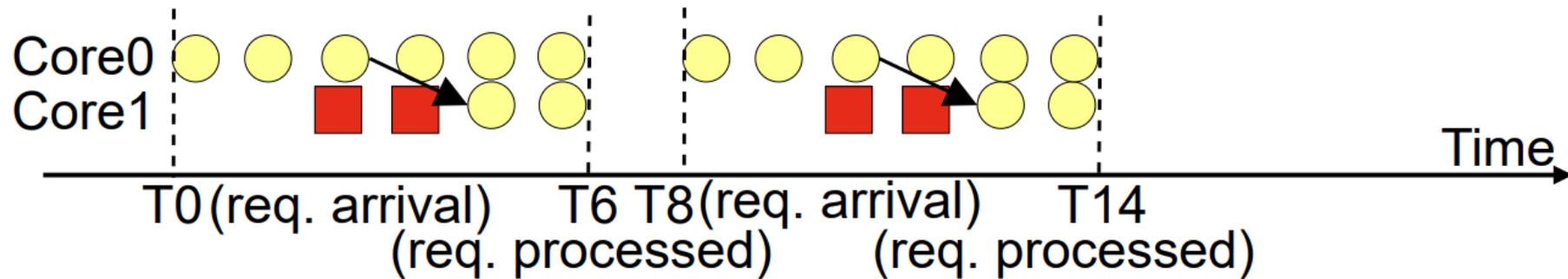
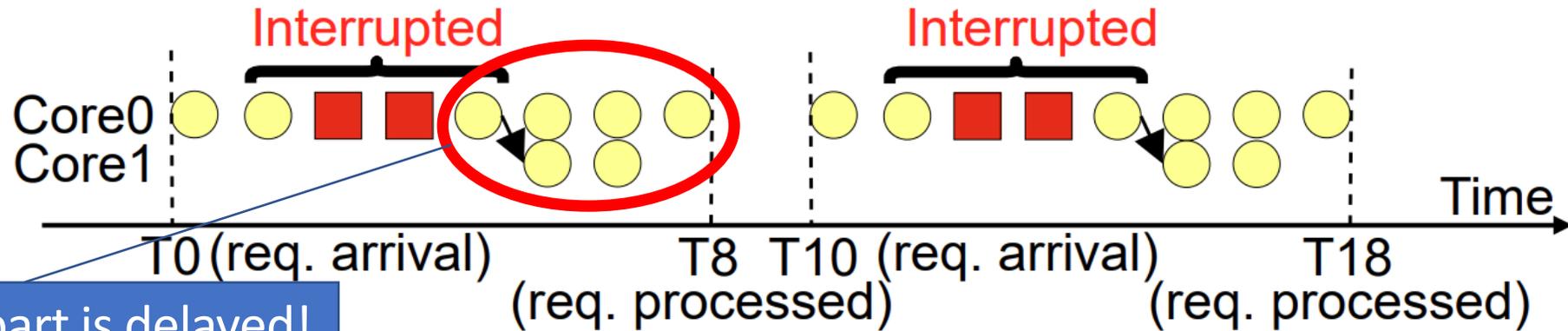
Unfinished part is delayed!

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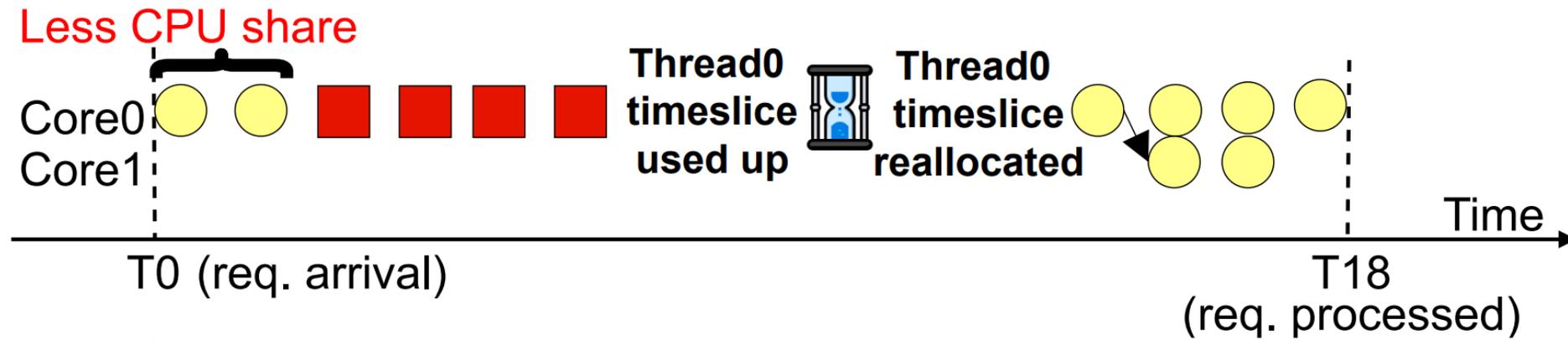
- Detection: threads/processes with low rescheduling latencies are scheduled on the same core or their total timeslices exceed core's capacity.

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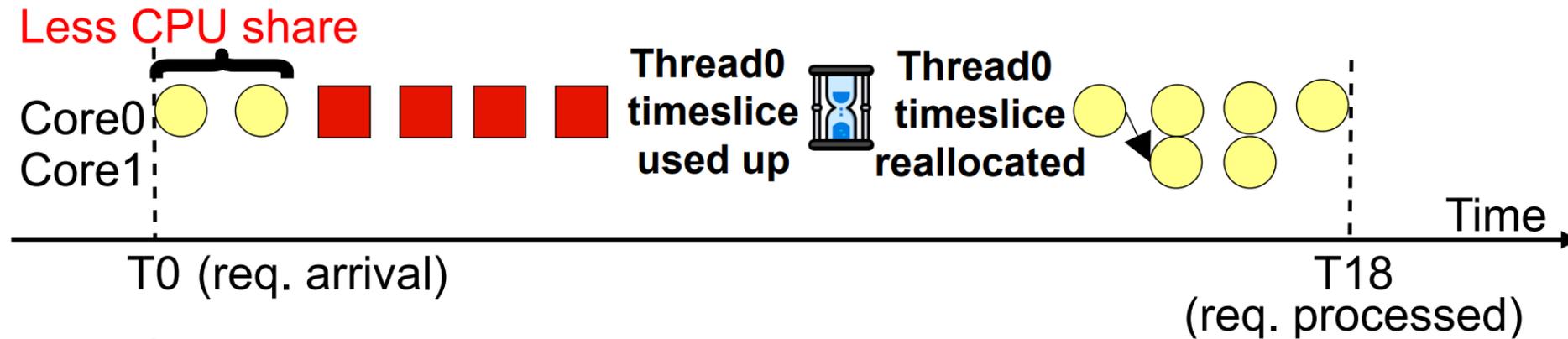


- Detection: threads/processes with low rescheduling latencies are scheduled on the same core or their total timeslices exceed core's capacity.
- Solution: adjust the layout of threads/processes on cores in a conservative way.

# Interference #3: tasks on app's critical path lack CPU share

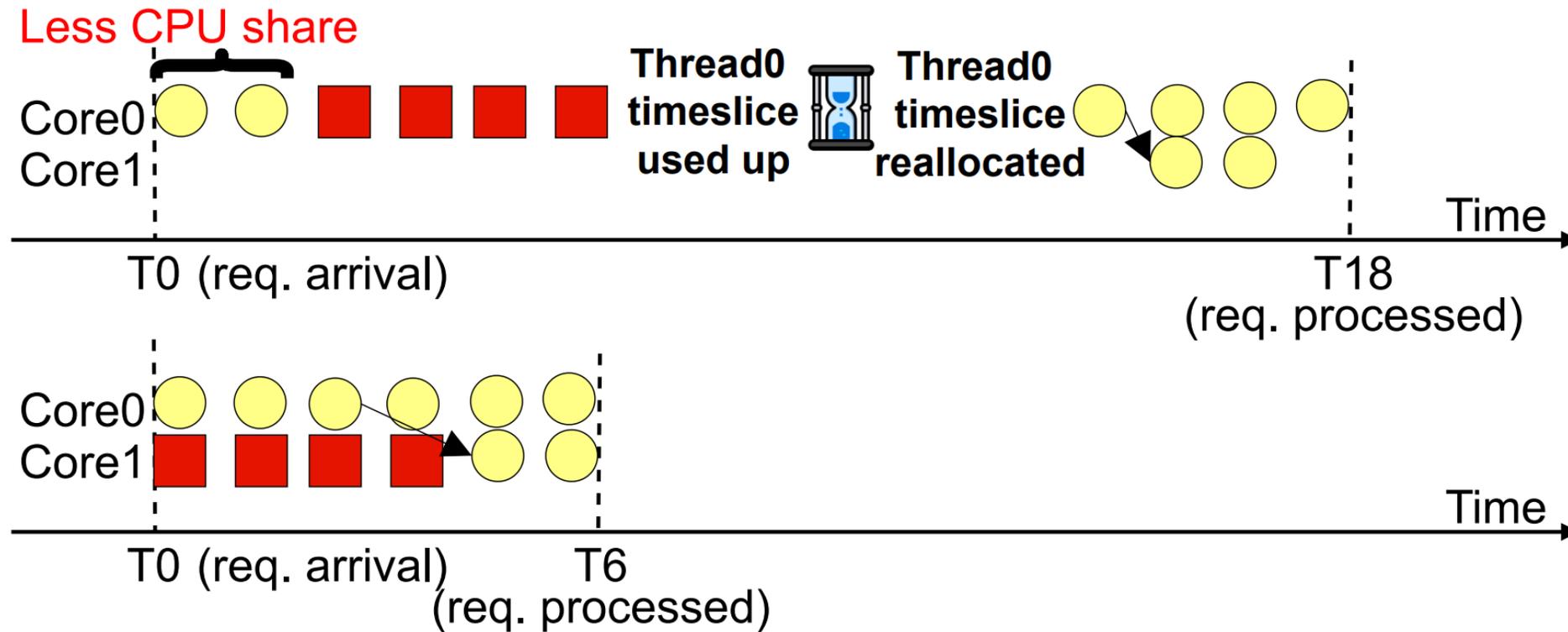


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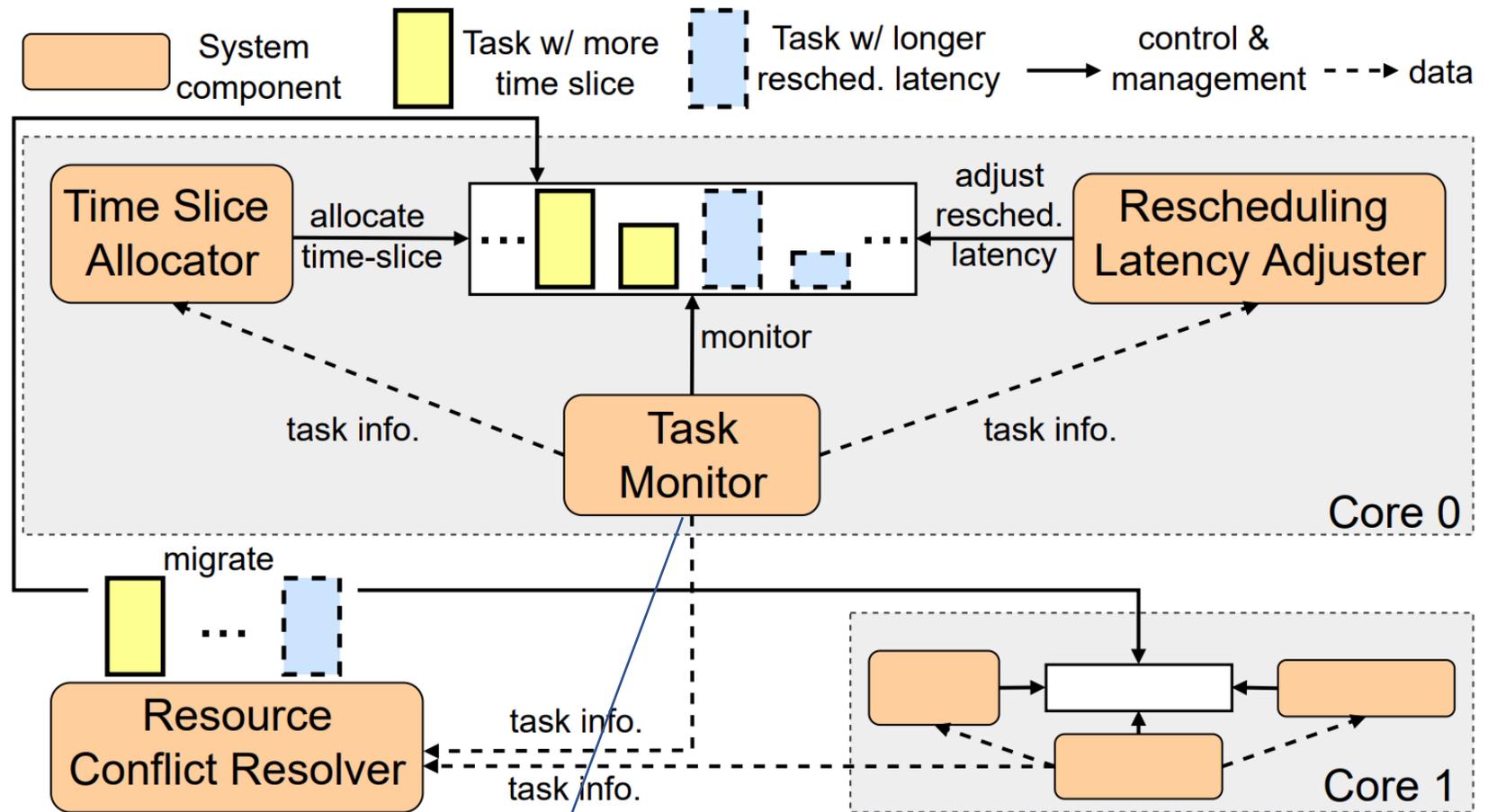
- Detection: threads/processes were preempted due to depletion of timeslice in previous time period.

# Interference #3: tasks on app's critical path lack CPU share



- Detection: threads/processes were preempted due to depletion of timeslice in previous time period.
- Solution: keep total timeslice of the app fixed and allocate more time share to the threads/processes on app's critical path.

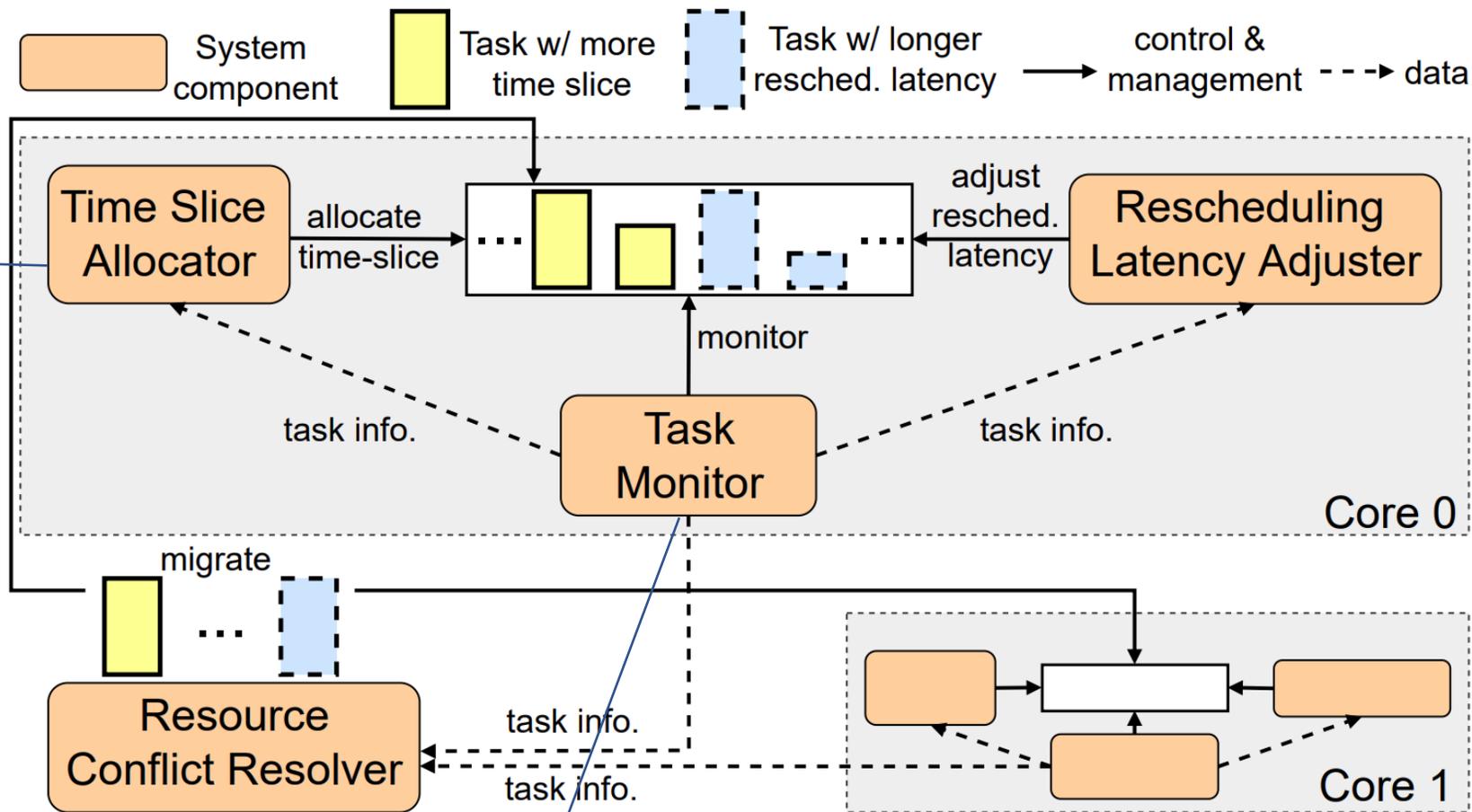
# DASEC implementation



Periodically monitors tasks' states and events, e.g., remaining timeslice. (Implemented in Linux Proc FS)

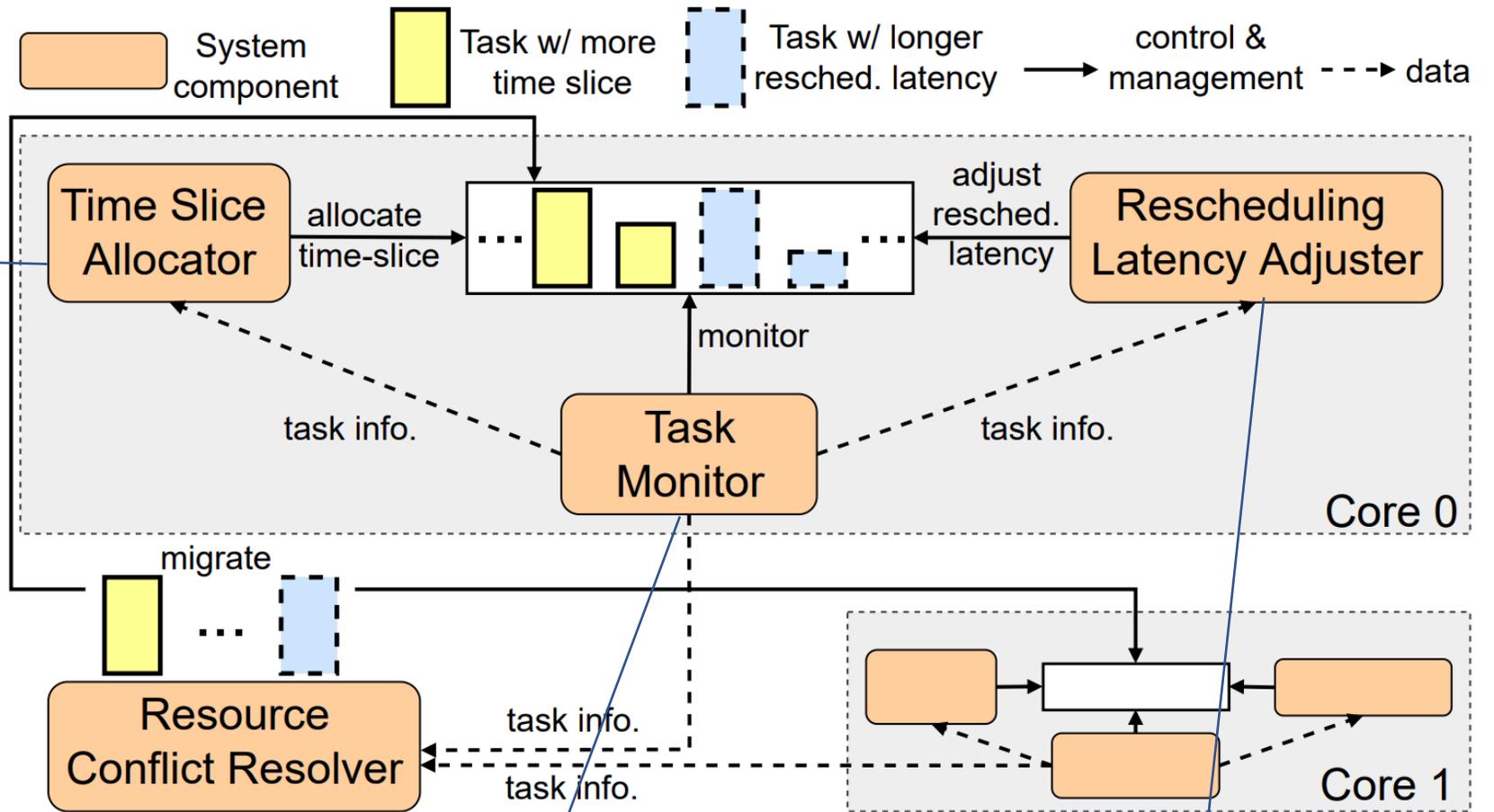
# DASEC implementation

Assign timeslice to each task for the upcoming time period through changing tasks' weights. (Implemented in Linux CFS)



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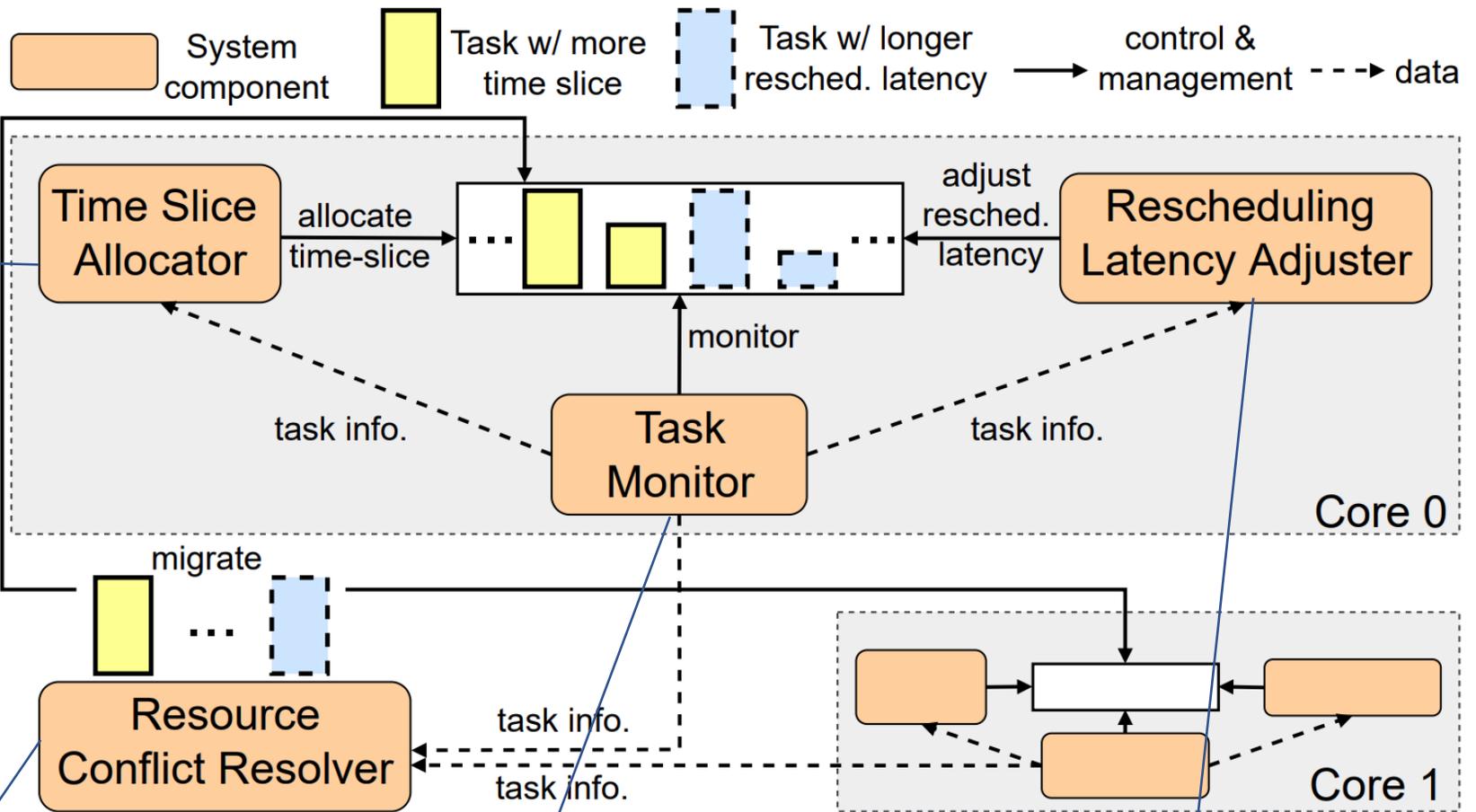


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# DASEC implementation



Assign timeslice to each task for the upcoming time period through changing tasks' weights. (Implemented in Linux CFS)

Resolve time slice conflicts and rescheduling latency conflicts on cores through changing tasks' layout on cores. (Implemented with Linux set affinity interfaces)

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Adjust wakeup latency to change each task's rescheduling latency. (Implemented in Linux CFS)

# Outline

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  - Key issues and solutions.
- ✓ Evaluation.
  - DASEC has been implemented in Linux/KVM, Linux CFS, and Google user-level scheduler (i.e., ghost [SOSP '21]).
  - Compared to vanilla Linux/KVM, DASEC reduces mean latency and 99th tail latency by 46% and 52%, respectively.

# Experimental setup

- HPE ProLiant DL580 Gen10 server with 80 cores, 256GB DRAM, and two 2TB SSDs.
- Both VMs and VMM (Linux QEMU/KVM) use Ubuntu Linux 18.04 with the same Linux 5.3 kernel and software configuration.
- Each VM has 16 vCPUs and 16GB memory.
- Compared with vanilla Linux/KVM, PARTIES ([ASPLOS '19]), and BVT ([SOSP '99] and [EuroSys '14]).
- Test under two settings.
  - Multiple VM instances of the same workload.
  - Multiple VM instances of different workloads.

# Evaluation applications and workloads

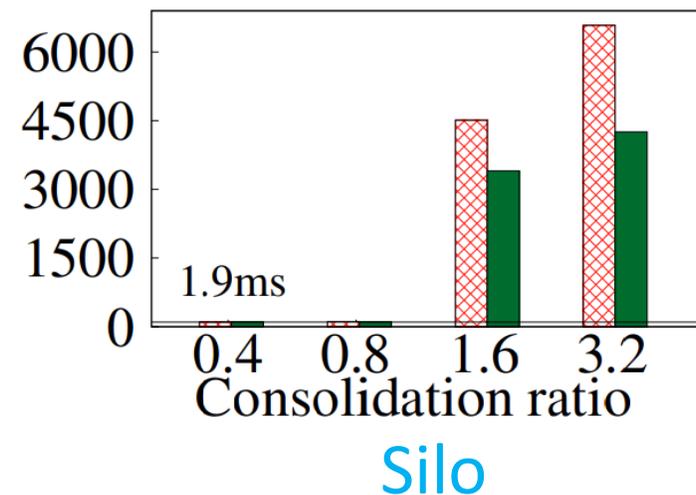
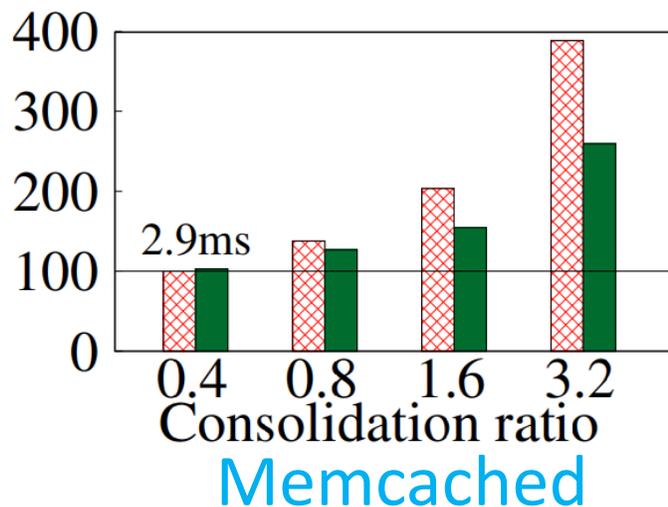
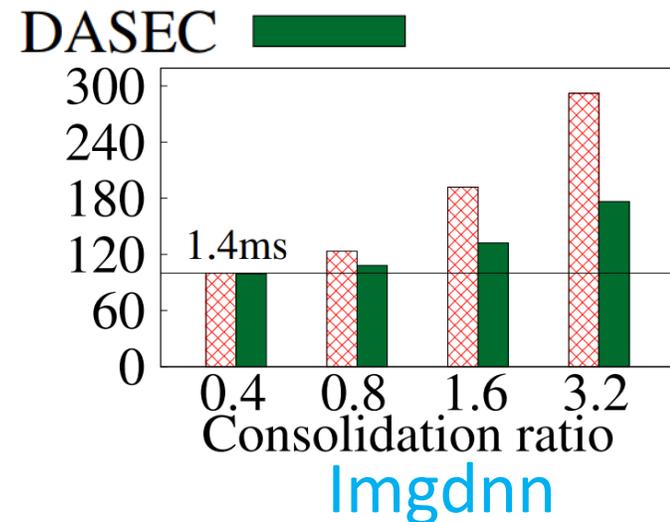
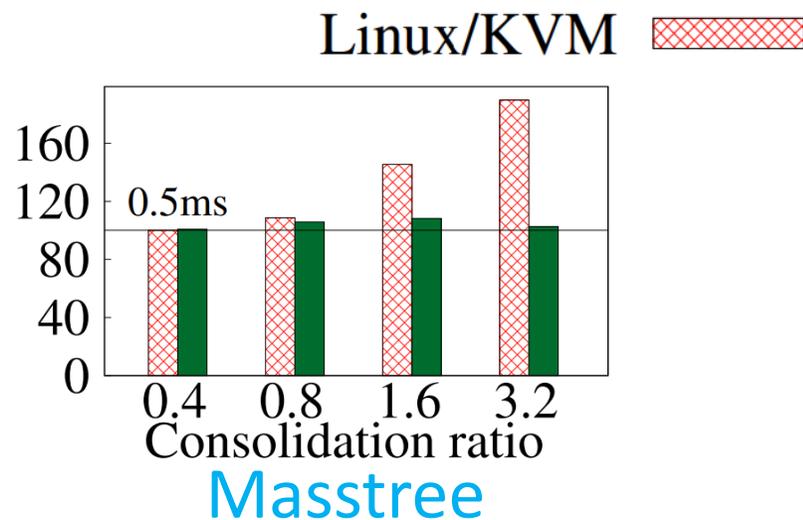
Application	Workload description
Image-classify	Image classification on ImageNet
Action-recognize	Video action recognition
Img-dnn	Handwriting recognition based on OpenCV
Masstree	In memory Key/Value store with 50% GET and 50% SET
Silo	In-memory transactional database with TPCC
Memcached	Serve requests (random keys, 50% SET, 50% GET)

# Evaluation objectives

- What is DASEC's performance?
- How much performance improvement can be achieved with DASEC, compared with PARTIES?
- How effective is each technique in DASEC?
- What is DASEC's applicability and overhead?

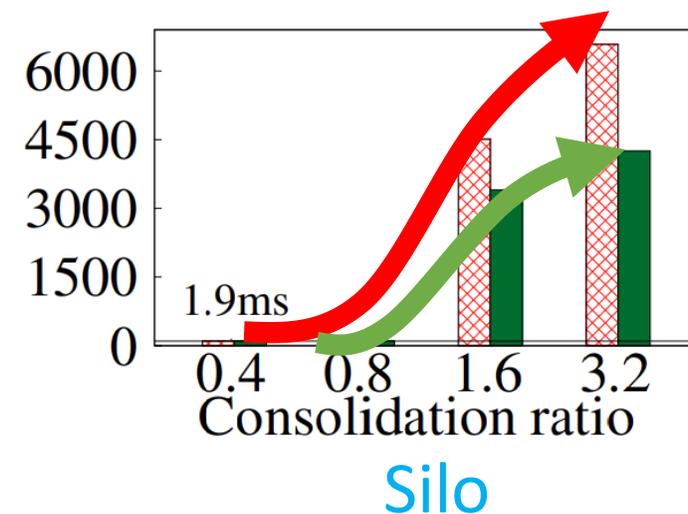
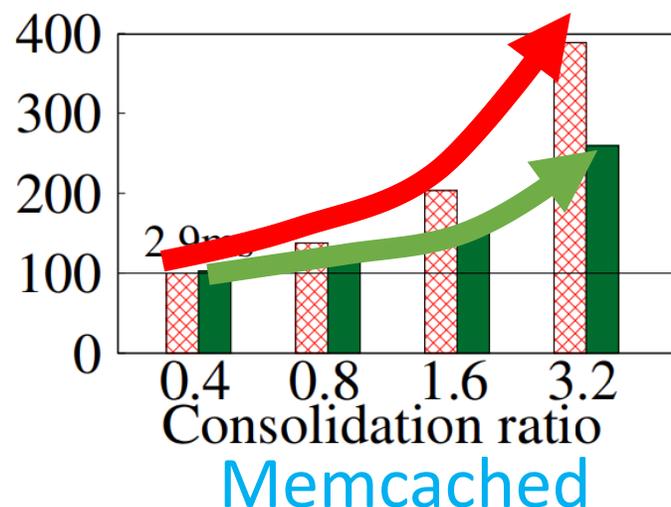
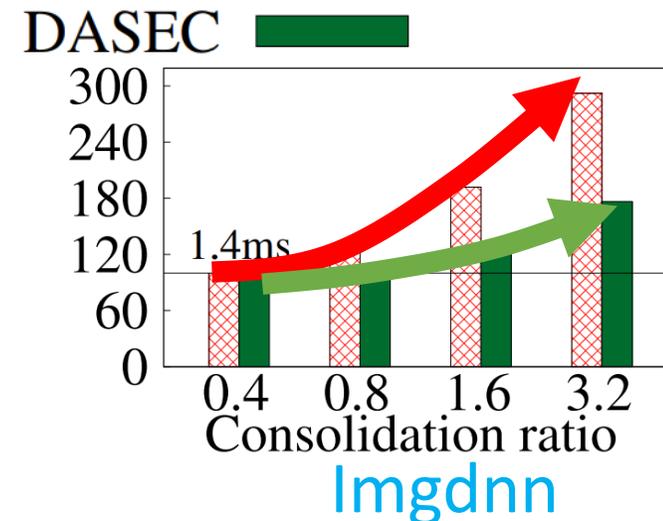
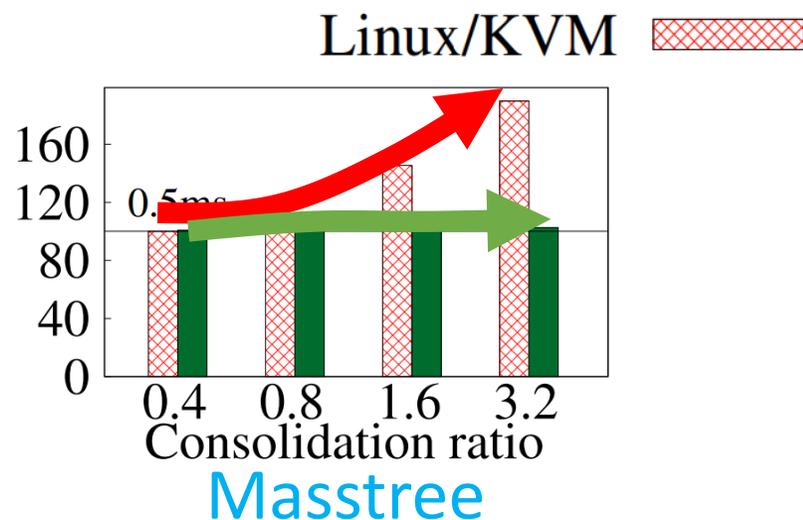
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\*Latencies relative to Linux/KVM when consolidation ratio is 0.4.



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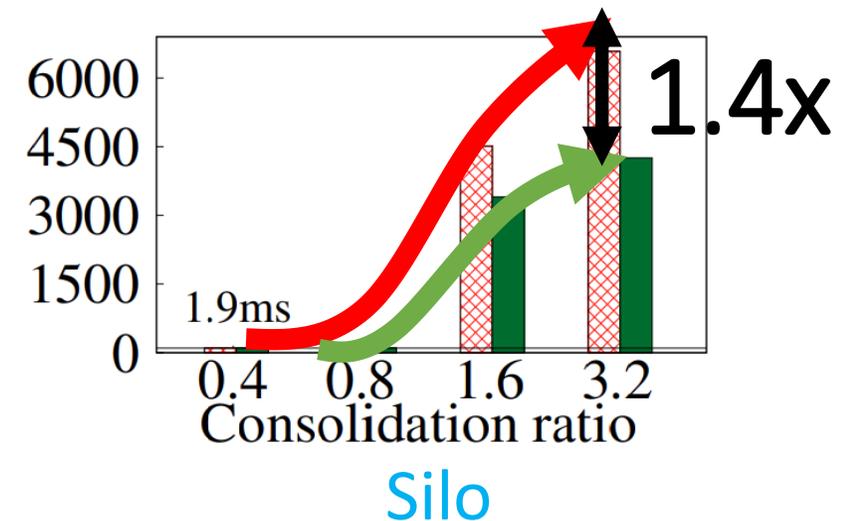
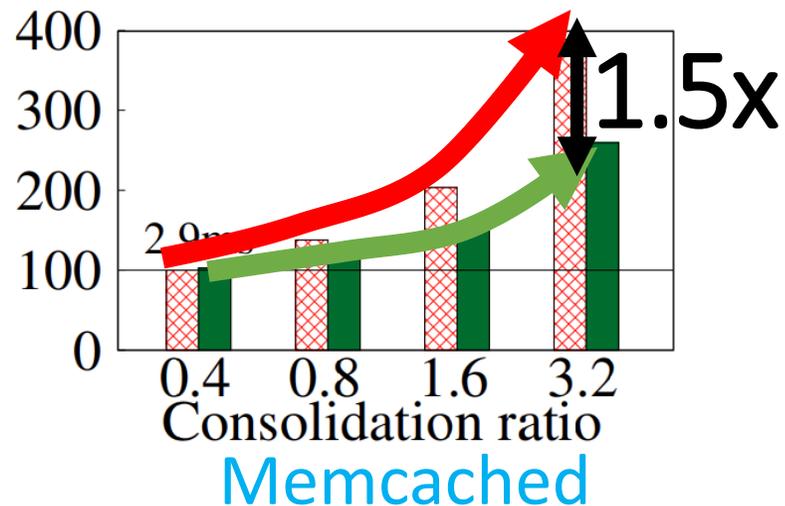
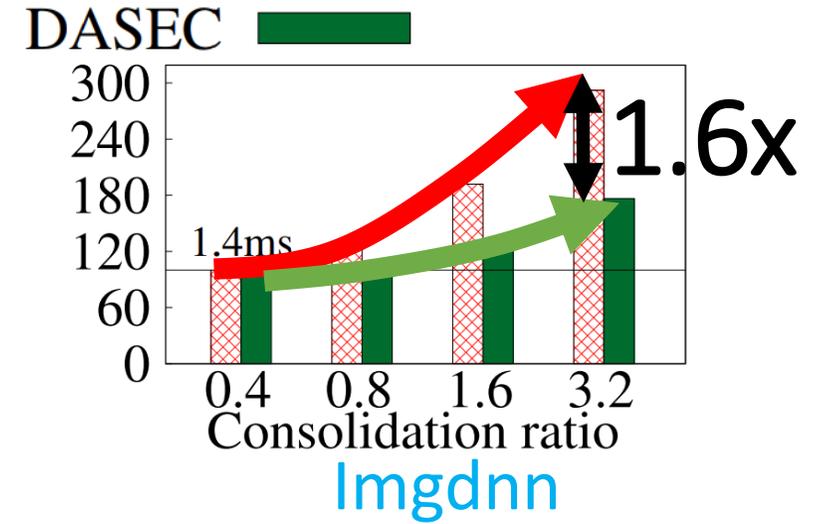
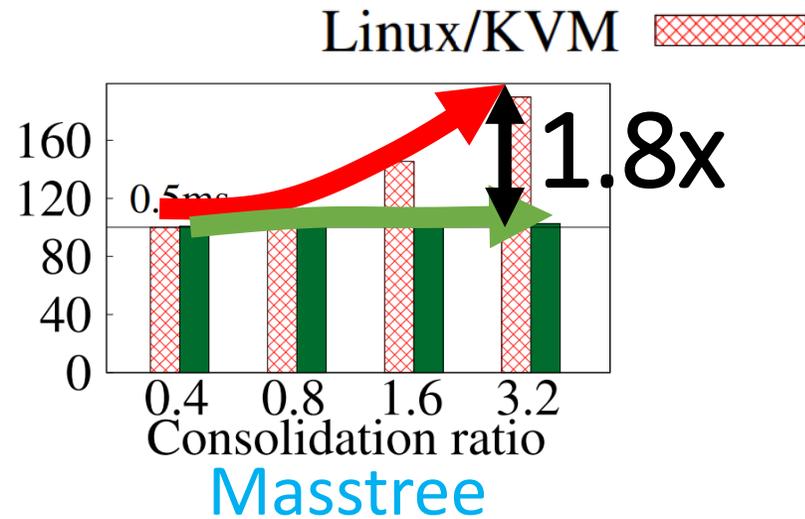
\*Latencies relative to Linux/KVM when consolidation ratio is 0.4.



- As consolidation ratio increases, Linux/KVM's **mean latency increases much more** compared to DASEC.

# Mean latency (DASEC vs Linux/KVM)

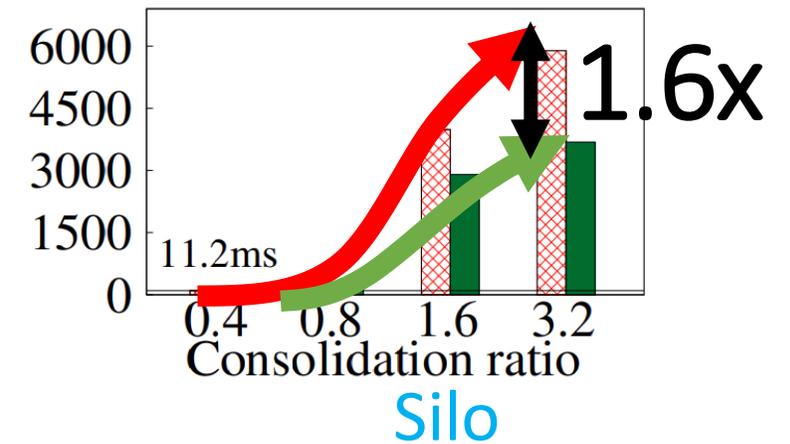
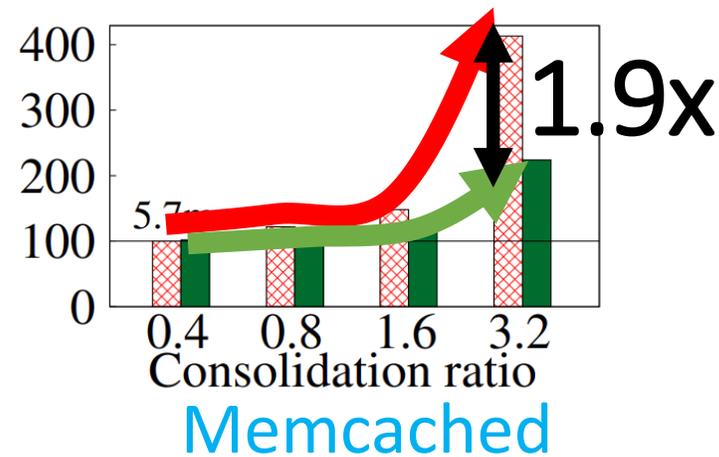
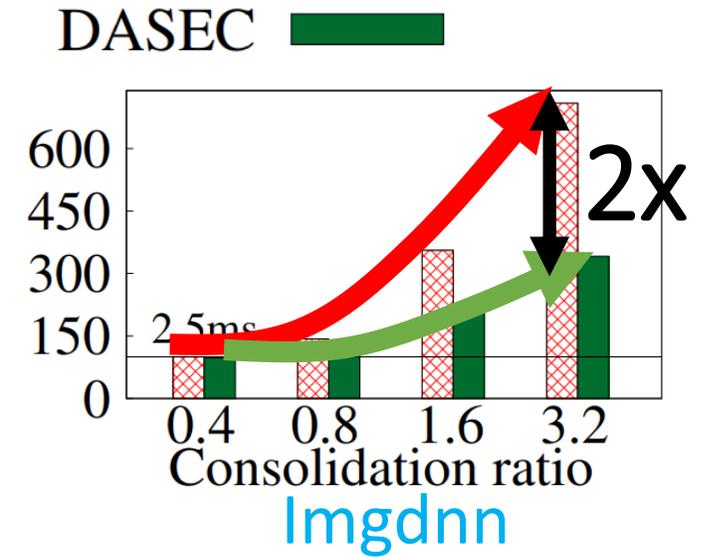
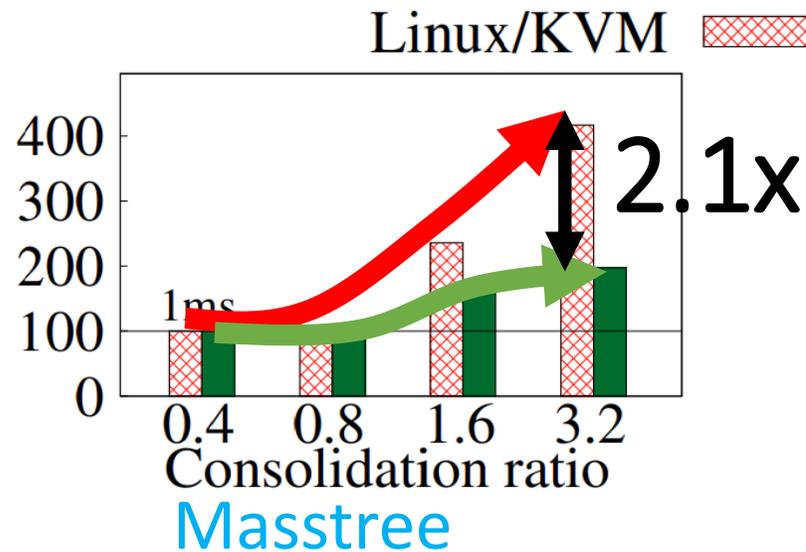
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- Compared to Linux/KVM, DASEC reduces **mean latencies** by **46%** on average.

# 99th tail latency (DASEC vs Linux/KVM)

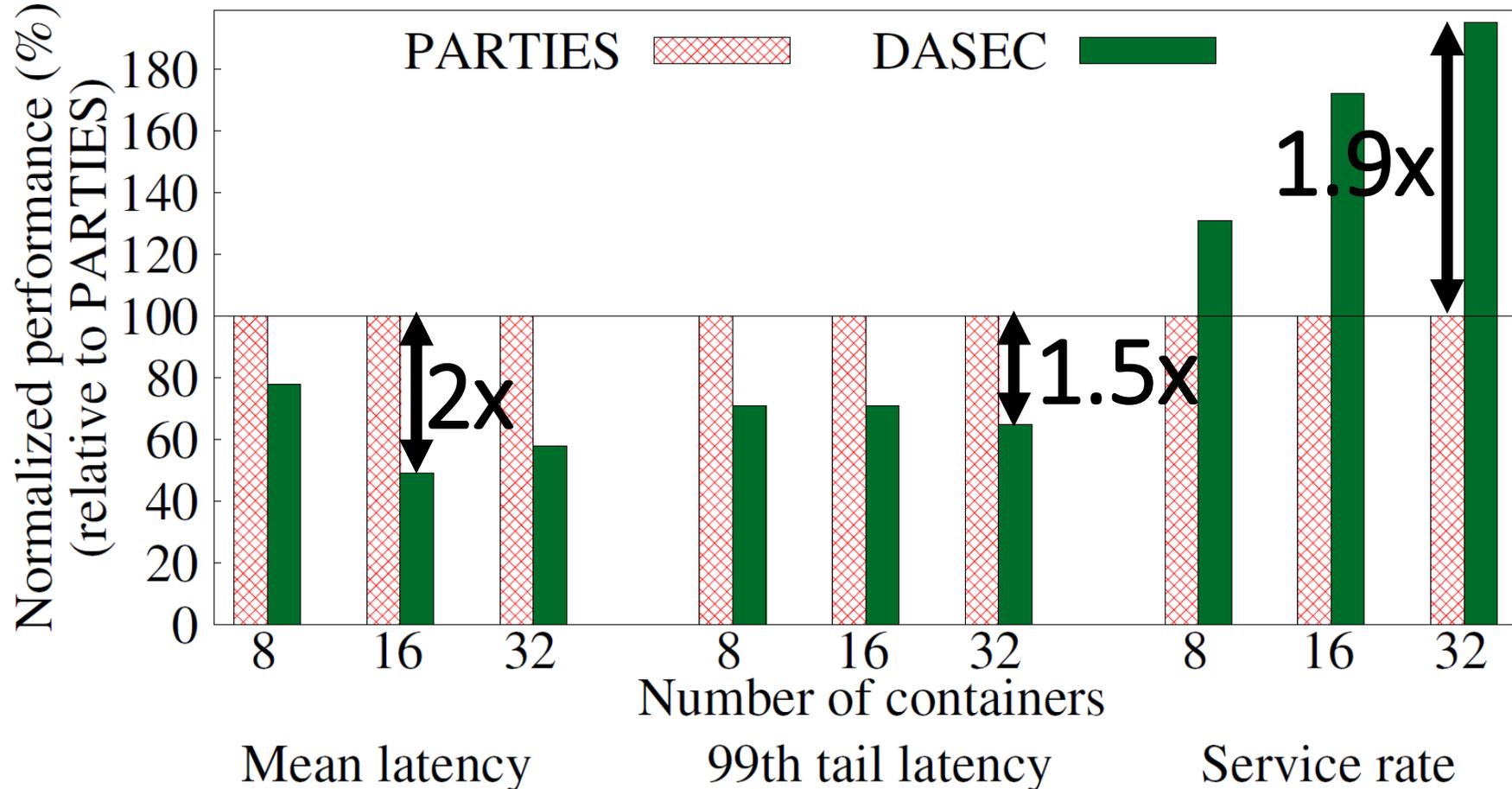
\*99th tail Latencies relative to Linux/KVM when consolidation ratio is 0.4.



- Compared to Linux/KVM, DASEC reduces 99th tail latencies by 52% on average.

# Performance (DASEC vs PARTIES)

\*All containers run the same [Masstree](#) workload.



- Compared to PARTIES, DASEC offers up to **51% lower mean latencies**, **35% lower 99th tail latency**, and **95% more service rate**.

# Conclusions

- How to efficiently schedule latency sensitive applications with **low latency by reducing their mutual interference** in edge cloud.
  - Edge cloud is **resource constrained** and dominated by latency sensitive workloads.
  - Such applications are **resource demanding** and have **dynamic resource usage**.
  - Existing cloud approaches are not effective to reduce latency in edge cloud.

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- DASEC is an efficient solution for reducing workload mutual interference and latency in edge cloud.
  - **Move the interference off the tasks on the critical paths** of the workload.
  - Detections and solutions for workload mutual interference in three ways.

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  - **Move the interference off the tasks on the critical paths** of the workload.
  - Detections and solutions for workload mutual interference in three ways.
- Evaluation shows **DASEC can substantially reduce latency** compared to related systems in edge cloud.

# References

- [1] Xu, Mengwei, Zhe Fu, Xiao Ma, Li Zhang, Yanan Li, Feng Qian, Shangguang Wang, Ke Li, Jingyu Yang, and Xuanzhe Liu. "From cloud to edge: a first look at public edge platforms." In Proceedings of the 21st ACM Internet Measurement Conference, pp. 37-53. 2021.
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Thank you!  
Questions?

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