Dynamic Partitioned Scheduling of Real-Time DAG Tasks on ARM big.LITTLE Architectures

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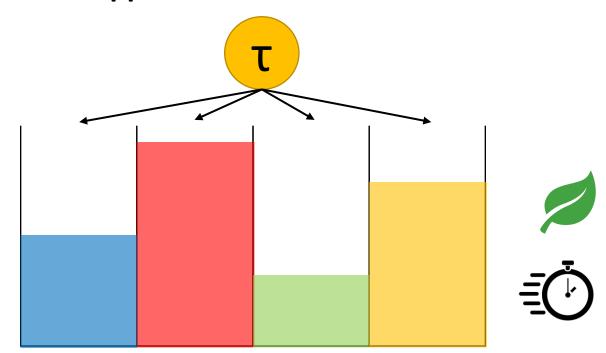
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Dynamic Partitioned Scheduling

- More and more multi-processor devices
 - Android on billion of devices
- More and more interactive application → real-time

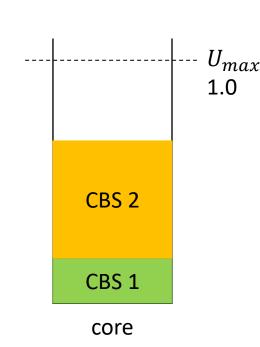




Background

CBS servers

- We make use of resource reservations to enforce temporal isolation among tasks
- A CBS reservation σ_i is associated parameters (Q, P), where Q is the *budget* and P the *reservation period*
- While a reservation is scheduled, the budget is decreased accordingly
- If tasks in the server try to execute for more than Q time units, the server deadline is postponed by P
- Core schedulability condition: $\sum_i U_i = \sum_i \frac{Q_i}{P_i} \leq 1$

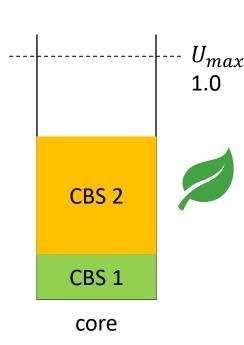




Background

GRUB-PA

- Greed Reclamation of Unused Bandwidth Power Aware
- It is an energy-aware variant of CBS server
 - Implements unused bandwidth reclamation
 - Exploits DVFS capabilities
- Implemented in mainline Linux running SCHED_DEADLINE CBS reservations from version 3.14 (Sept 2017)



State of Art

- Energy-aware scheduling of sequential tasks
 - Linear Programming-based Methods
 - Others consider heuristics and DVFS capabilities
 - Also explored in a previous work of ours (deeply explained later)
- Thermal-aware scheduling of sequential tasks
 - ILP methods
 - Minimizing peak temperature
 - Pattern-based approaches (go idle often to reduce temperature)
 - Feedback-loop-based approaches

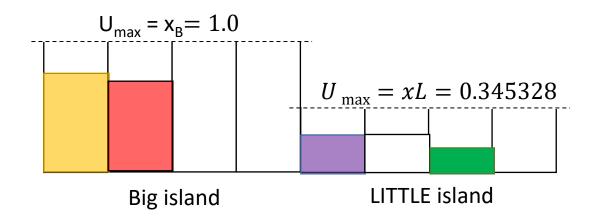
State of Art

- Non-energy-aware DAG scheduling
 - Some works only consider analysis and schedulability issues
 - Partitioning techniques based on Semi-/Federated scheduling
 - Gang scheduling
- Energy-aware DAG scheduling
 - Not much in the literature
 - We base our solution on Guo et al. (deeply explained later)
 - Based on the concept of Speed-Profiles (we don't use)
 - Introduces the Task Decomposition technique that we use



Notation - CPU

ARM big.LITTLE platform

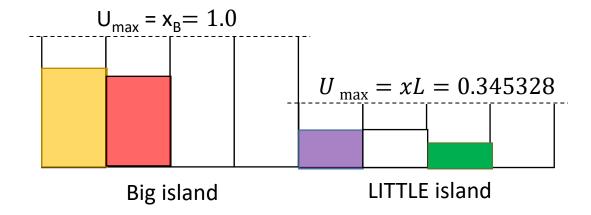


- Power model already implemented in RTSim as in [*]
 - Tries to be a good compromise between representativeness and complexity
 - Power consumption depends on voltage (quadratic) and frequency, and task workload type
 - Tuned on ODROID XU3
 - Does not consider interference due to other tasks, cache and memory



Notation - CPU

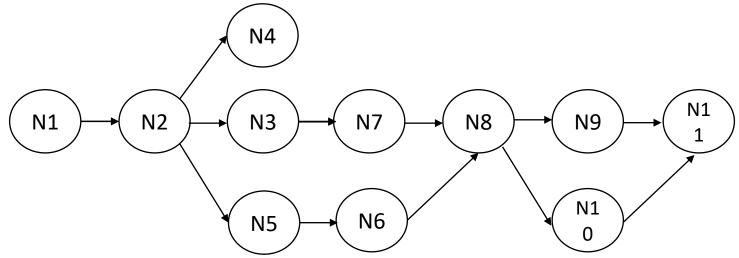
ARM big.LITTLE platform



- Power model
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Notation - task

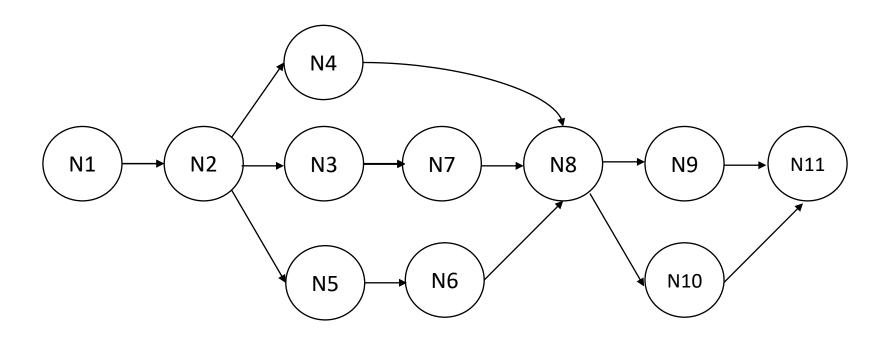
- Set of n **soft** real-time DAG task: τ_1 ... τ_n
- Implicit deadline (DAG task period = deadline)
- au_i contains a set of nodes $au_i = \left\{N_i^j\right\}_{j=1}^{n_i}$ with associated nominal WCET C_i^j and period T_i
- Open and dynamic system user opens and closes apps





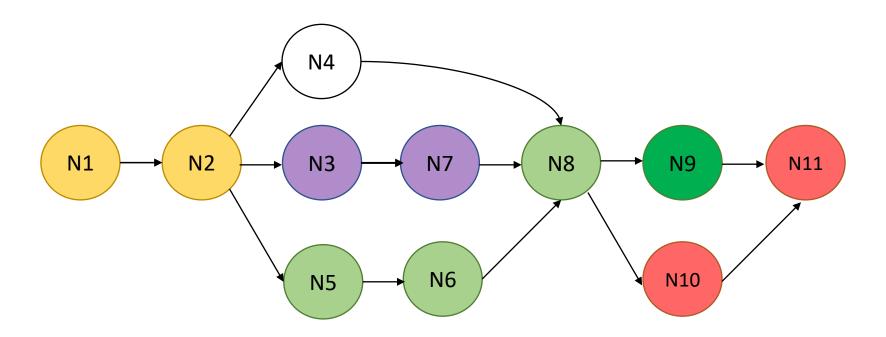
Problem

Given a DAG task τ_i :

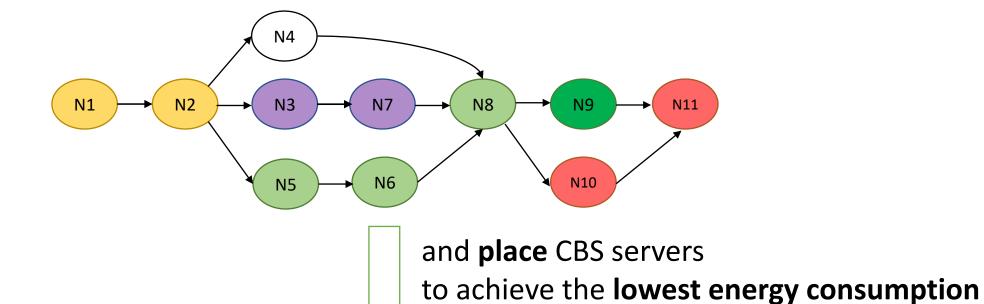


Problem, Novelty

How to split the DAG nodes into groups:



Novelty



& guarantee deadlines

 $U_{max} = x_B = 1.0$ $U_{\text{max}} = xL = 0.345328$ LITTLE island

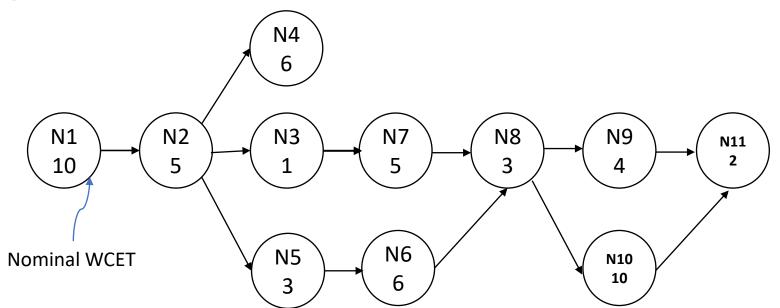
Big island

Our solution



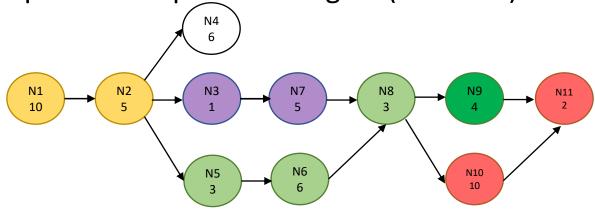
Approach

- Given a DAG task, we need:
 - A way to split it into groups of CBS servers
 - A way to **partition into the CPUs** the CBS servers
 - **Energy**-efficiently
 - Respecting the **soft deadlines**



Approach – DAG to { CBS server }

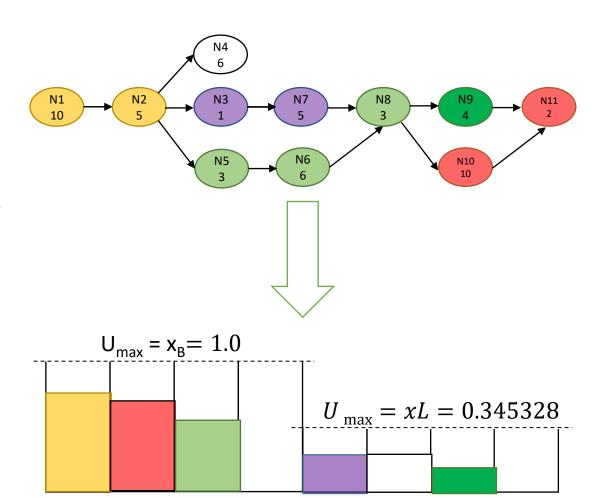
- How to split DAG task τ_1 into groups of CBS servers?
- DAG Task Decomposition Technique as in [*]
 - Guo et al. is based on "Speed-Profile"
 - Guo et al. is not about utilizations (we are)
- Optimized for our use-case
 - Decrease utilization of CBS servers
 - Changes some phases to optimize the goal (see later)





Approach — { CBS server } to CPUs

- How to place CBS servers onto the CPUs?
- Dynamic partitioning as in [*]
 - Previous work was about sequential tasks
 - Extended to DAG tasks
- Transparently assigns CBS parameters
- Decides core and frequency for each CBS server
- Provides the heuristically best core giving min (additional) power consumption



Big island



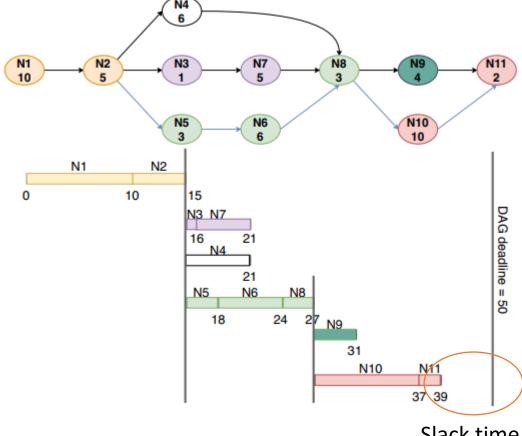
LITTLE island

Approach – decrease CBS servers' utilization

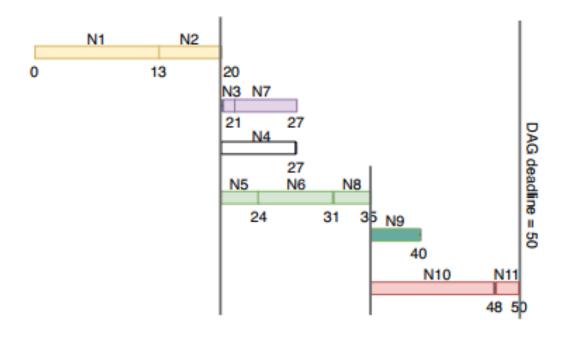
- Smaller CBS utilization -> smaller core utilization -> smaller frequency
 - -> less energy consumption
- Revisited and optimized version of DAG Decomposition Technique (original by Guo et al.)
- Performed transparently at DAG task first arrival
- Many phases:

Phase 1 – DAG Task Decomposition

- In this phase, nodes are divided into groups, which are assigned initial Arrival and Finishing Time
 - DAG task has deadline 50
 - E.g., N3 and N7 (violet) belong to one CBS server
 - N3 first job (WCET = 1) begins at t=15 and finishes at t=16
 - N7 first job (WCET = 5) begins at t=16 and finishes at t=21



Phase 2 – Slack allocation



- Slack time is used to relax nodes finishing times and thus decrease CBS utilizations
- Allocated uniformly among nodes. E.g. (L_i is DAG critical path length):

$$\delta \equiv \frac{T_i}{L_i} = \frac{50}{39} = 1.28$$

$$f_i^{N11} = \lceil 39 \times 1.28 \rceil = 50$$

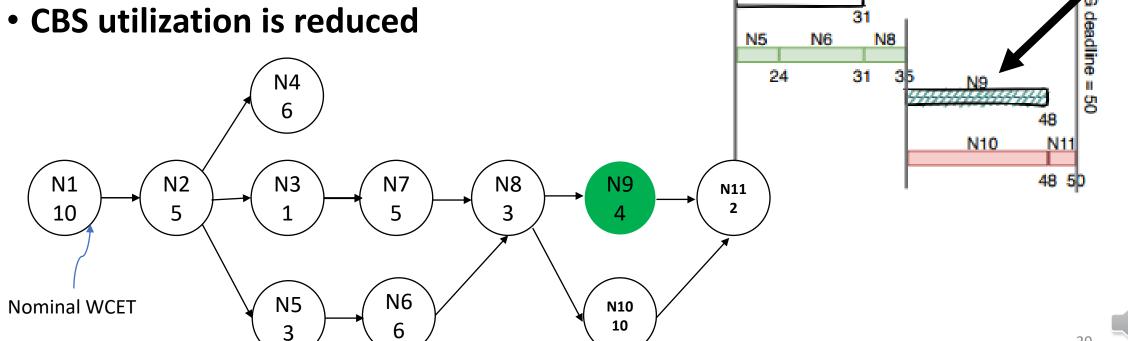
$$f_i^{N10} = \lceil 37 \times 1.28 \rceil = 48$$

$$f_i^{N9} = \lceil 31 \times 1.28 \rceil = 40$$



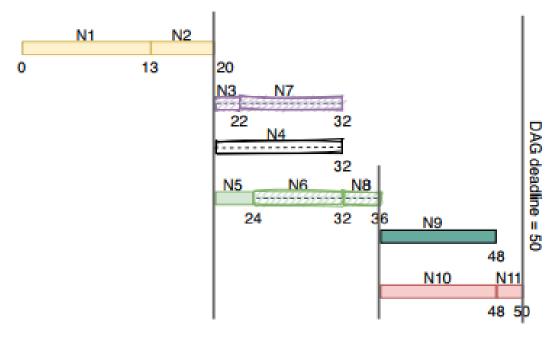
Phase 3 – Segment Extension

- N7 had no reason to finish within the former $f_i^{N9} = 40$
- $\bullet \Rightarrow f_i^{N9} = a_i^{N11} = 48$



N2

Phase 4 – Relaxing finishing times (ours)



Further reduces CBS utilization

- Distribute the time window of each CBS to its nodes, proportionally to their WCET
- E.g., for CBS $S = \{ N3 = 1, N7 = 5 \}$:

$$f_i^{N3} = \lceil a_i^{N3} + C_i^{N3} \times \mu \rceil = \lceil 20 + 1 \times 1.83 \rceil = 22$$
$$f_i^{N7} = \lceil 22 + 5 \times 1.83 \rceil = 32$$

$$\mu \equiv \frac{f_S^{N_S^n} - a_S^{N_S^1}}{\sum_{N_S^j} C_S^j} = \frac{31 - 20}{1 + 5} = 1.83$$



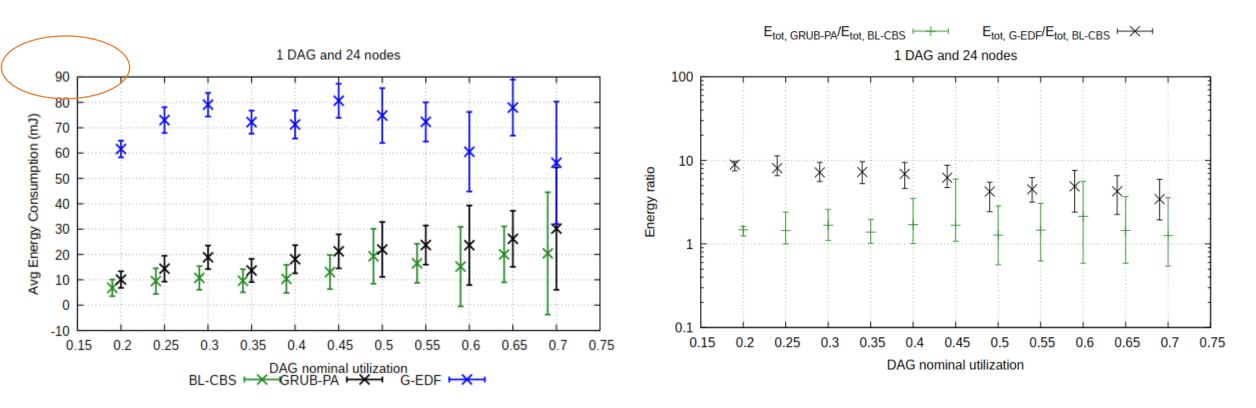
Evaluation

- Compare BL-CBS (i.e., this technique) with GRUB-PA
- Random tasksets of {1,2,3} DAG tasks and {24,12,8} nodes respectively; Different DAG utilizations { 0.2, 0.25, ..., 0.7 }
- Execution time of node instances is set to be uniformly distributed between 0.1 ms and the node nominal WCET => more dynamic environment
- For a given CBS server $\sigma_{i,k}$ containing a set of nodes $\{N_i^J\}$:

$$\begin{cases} Q_{i,k} = \sum_{N_i^j \in \sigma_{i,k}} C_i^j \\ P_{i,k} = \sum_{N_i^j \in \sigma_{i,k}} (f_i^j - \alpha_i^j) \end{cases}$$

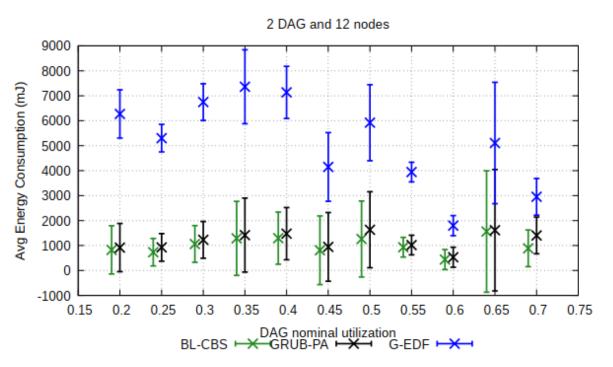


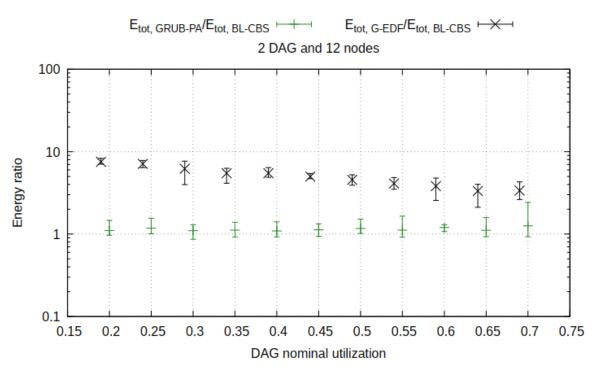
Evaluation (Avg energy)



BL-CBS < GRUB-PA
BL-CBS << G-EDF

Evaluation (Avg energy)



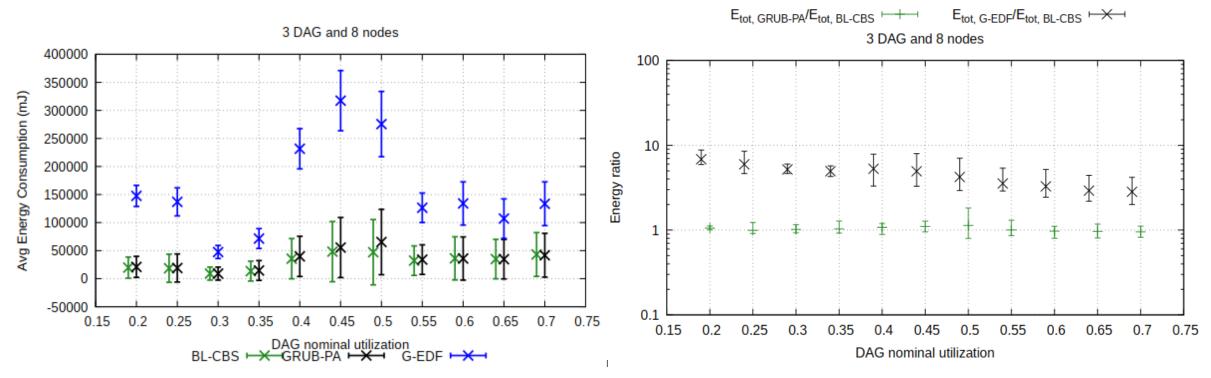


 $9000 \text{ mJ} = 90 \text{mJ} \times 100$

BL-CBS < GRUB-PA
BL-CBS << G-EDF



Evaluation (Avg energy)

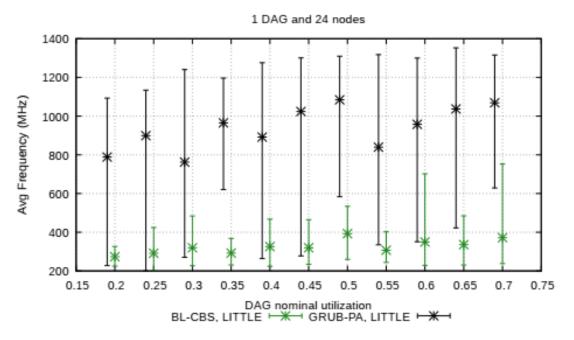


 $400.000 \text{ mJ} = 90 \text{ mJ} \times 4.444$

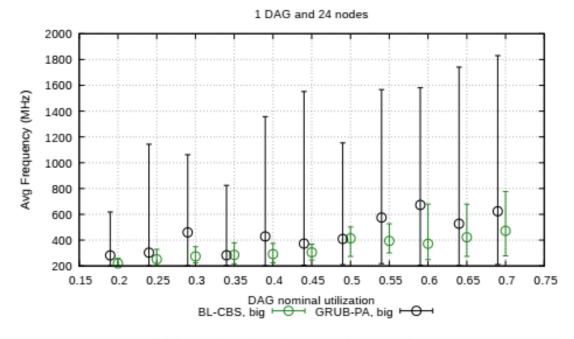
BL-CBS <= GRUB-PA
BL-CBS << G-EDF



Evaluation (avg frequency)



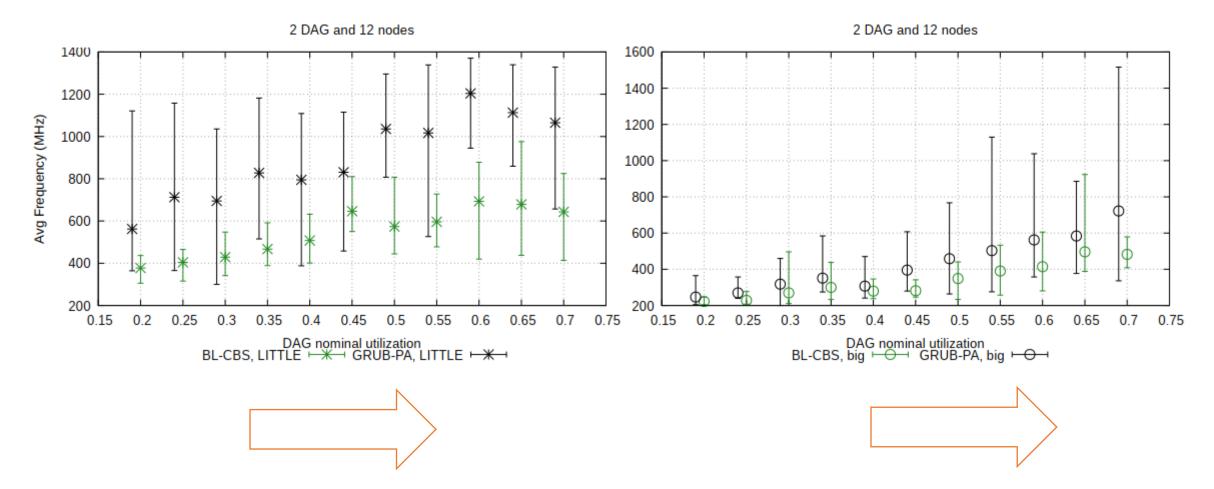
(a) LITTLE island, 1 DAG and 24 nodes



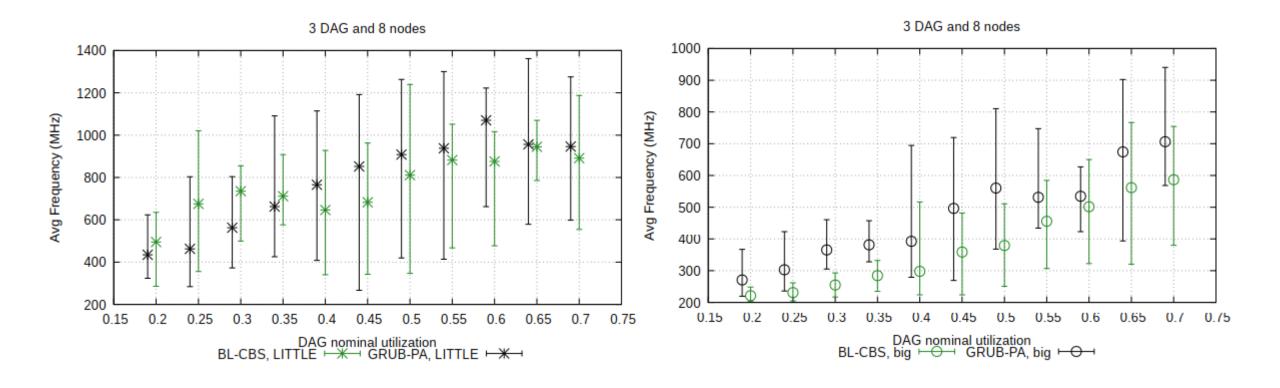
(b) big island, 1 DAG and 24 nodes



Evaluation (avg frequency)



Evaluation (avg frequency)



LITTLE island has higher frequencies than big island

Conclusions

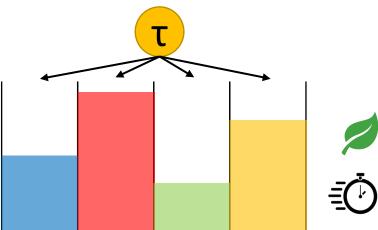
Incorporate DAG placement and splitting strategies

Improvement for energy saving

- **BL-CBS for DAG** tasks
- Support to "open" and dynamic system (Android use-cases)
- Made for **ARM big.LITTLE** architecture

Experimental results

- On random tasksets
- Energy saving in average 10% over all the experiments with respect to the state-of-the-art GRUB-PA





Future works

- Improve the **performance** of the placement algorithm
- Place the servers on the cores also considering
 - The nodes relationships
 - The memory consumption and different workload types
- Consider CPU deep-idle states
- Incorporate Bandwidth reclaiming and feedback mechanisms

Thank you!

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Slide & paper

https://owncloud.retis.santannapisa.it/index.php/s/py1WwfF2aSUjciV