# Online Scheduling of Task Graphs on Hybrid Platforms

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### **Hybrid Platforms**

Many CPUs + few accelerators (GPUs, Xeon Phis, ...)

### Task Graphs (DAGs)

Used in runtime schedulers (StarPU, OmpSs, XKaapi, ...)

### **Online Scheduling**

- Unknown graph
  - tasks not submitted yet
  - depends on results

- Advantages vs offline
  - quicker decisions
  - robust to inaccuracies
- ► Semi-online: partial information, e.g., bottom-levels (≈ critical path)

#### Main challenge: take binary decisions without knowing the future

### Model

- $m \text{ CPUs} \ge k \text{ GPUs}$
- Graph of tasks  $T_i: \{\overline{p_i} = \text{CPU time}; \underline{p_i} = \text{GPU time}\}$
- Online: only available tasks are known



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### Related work

### Existing offline algorithms (NP-Complete)

- Independent tasks:
  - $\frac{4}{3} + \frac{1}{3k}$  approx Expensive PTAS
  - Low-complexity: 2 approx
    - 3.41 approx

DAG: 6 - approx (LP rounding)

[Canon, Marchal, Vivien 2017]

[Bonifaci, Wiese 2012]

[Beaumont, Eyraud-Dubois, Kumar 2017]

[Bleuse, Kedad-Sidhoum, Monna, Mounié, Trystram 2015]

[Kedad-Sidhoum, Monna, Trystram 2015]

#### **Existing online algorithms**

Independent tasks: 4 - competitive

3.85 - competitive

[Imreh 2003]

[Chen, Ye, Zhang 2014]

**DAG:**  $4\sqrt{\frac{m}{k}}$  - compet. ER-LS

[Amarís, Lucarelli, Mommessin, Trystram 2017]

### 1. Lower bounds on online algorithms

• No online algorithm can be  $<\sqrt{m/k}$  - competitive

### 2. Propose improvements of ER-LS

Competitive ratio

Outline

- Average performance
- Validation on simulations

#### Theorem

No online algorithm  $\mathscr{A}$  is  $<\sqrt{m/k}$  - competitive for any m, k.

**Proof (where**  $\tau = \sqrt{m/k} = 3$ ): graph built in  $n\tau$  phases.

Phase 1 -  $k\tau$  independent tasks  $\{\overline{p_i} = \tau ; \underline{p_i} = 1\}$ :  $\mathscr{A}$  needs a time  $\tau$ 



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Graph with k = 2, n = 3



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kτ

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 $\implies$  Makespan obtained by  $\mathscr{A}$ :  $n\tau^2$ 

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# Generalized lower bounds

Recall previous lower bound:  $\sqrt{m/k}$ , for *m* CPUs, *k* GPUs

### **Precomputed information**

- ▶ Bottom-level (≈ remaining critical path) does not help
- All descendants: non-constant LB =  $\Omega((m/k)^{1/4})$

#### **Powerful scheduler**

- Kill + migrate does not help
- Preempt + migrate hardly helps

#### Note: allocation is difficult

- How to choose which tasks to speed-up?
- Fixed allocation: 3 competitiveness

# ER-LS algorithm $(4\sqrt{m/k}$ -competitive, [Amarís et al.])

#### Main concept

m CPUs, k GPUs

- Pick any available task T<sub>i</sub>
- Allocate T<sub>i</sub> to CPUs or GPUs
- Schedule it as soon as possible

# Where to allocate an available task $T_i$ If $T_i$ can be executed on GPU before time $\overline{p_i}$ : $\blacktriangleright$ put $T_i$ on GPU Otherwise: $\blacktriangleright$ if $\frac{\overline{p_i}}{\underline{p_i}} \le \sqrt{\frac{m}{k}}$ : put it on CPU $\blacktriangleright$ else : put it on GPU

# Our proposition: QA (Quick Allocation) algorithm

#### Main concept

m CPUs, k GPUs

- Pick any available task T<sub>i</sub>
- Allocate T<sub>i</sub> to CPUs or GPUs
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#### Where to allocate an available task $T_i$

If  $T_i$  can be executed on GPU before time  $\overline{p_i}$ :

put T; on GPU

Otherwise:

- if  $\frac{\overline{p_i}}{p_i} \le \sqrt{\frac{m}{k}}$ : put it on CPU
- > else : put it on GPU

# Our proposition: QA (Quick Allocation) algorithm

#### Main concept

m CPUs, k GPUs

- Pick any available task T<sub>i</sub>
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### Where to allocate an available task $T_i$

If  $T_i$  can be executed on GPU before time  $\overline{p_i}$ :

put T; on GPU

Otherwise:

• if 
$$\frac{\overline{p_i}}{p_i} \leq \sqrt{\frac{m}{k}}$$
: put it on CPU

> else : put it on GPU

#### Theorem

QA is  $2\sqrt{m/k} + 1$  - competitive. This ratio is (almost) tight.

## What about easy cases?

### Problem with **QA**

m CPUs, k GPUs

- Expect the worse: aim at  $\Theta(\sqrt{m/k})$ -competitiveness
- Sector Poor performance on easy graphs

### Well-known EFT algorithm (Earliest Finish Time)

- Terminate each  $T_i$  as soon as possible;
- Greedy version, works great on non-pathological cases
- ▶ ⓒ Can be really bad:  $\geq (\frac{m}{k} + 2)$  OPT

### Can we have both benefits? MIXEFT

- Run EFT and simulate QA;
  When EFT is λ times worse than QA: switch to QA;
- ► Tunable:  $\lambda = 0 \rightarrow QA$  ;  $\lambda = \infty \rightarrow EFT$
- $(\lambda + 1)(2\sqrt{m/k} + 1)$ -competitive conjectured max $(\lambda, 2\sqrt{m/k} + 1)$
- Same idea as ER-LS but pushed to the extreme

# Simulations

m CPUs, k GPUs

### Heuristics (makespan normalized by offline HEFT's)

- ▶ EFT (= MIXEFT as EFT better than QA here)
- QA (switch at  $\sqrt{m/k}$ )
- ER-LS (= QA + greedy rule: slightly more tasks on GPUs)
- QUICKEST (= QA with switch at 1: more tasks on GPUs)
- **RATIO** (= QA with switch at m/k: more tasks on CPUs)

### **Datasets for** m = 20 **CPUs and** k = 2 **GPUs**

Cholesky 4 types of tasks Synthetic STG set, 300 tasks, random GPU acceleration ( $\mu = \sigma = 15$ ) Ad-hoc one chain & independent tasks

10/14

# Results for Cholesky graphs (lower is better)



# Results for synthetic graphs (lower is better)



# Results for 300-tasks ad-hoc graphs (lower is better)



# Conclusion

#### Summary

- ► No online algo. is <√m/k competitive Additional knowledge or power hardly helps
- QA:  $(2\sqrt{m/k}+1)$  competitive MIXEFT: compromise effectiveness / guarantees
- Extended to multiple types of processors (not in this talk)

#### Perspectives

- Low-cost offline algorithm with constant ratio
- Communication times

[Yesterday's talk by Alix Munier-Kordon]

Parallel tasks