Online Scheduling of Task Graphs on Hybrid Platforms

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Breaking down the title

Hybrid Platforms

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

Task Graphs (DAGs)

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

Online Scheduling

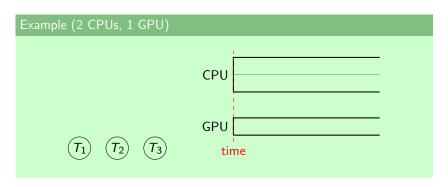
- Unknown graph
 - tasks not submitted yet
 - depends on results

- 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

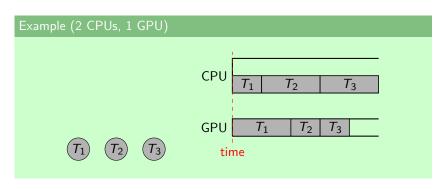
Model

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



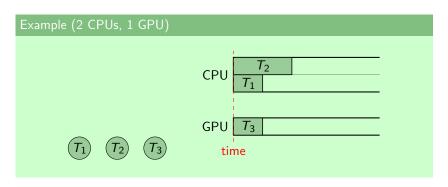
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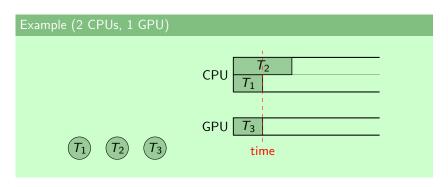
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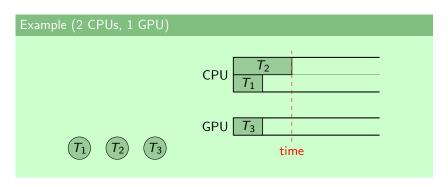
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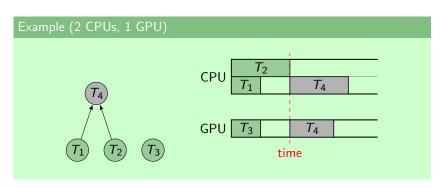
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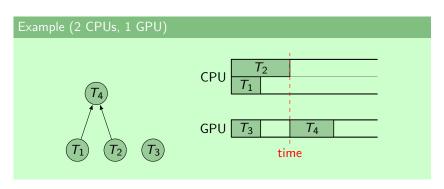
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Existing offline algorithms (NP-Complete)

- ► Independent tasks:
 - $\frac{4}{3} + \frac{1}{3k}$ approx [Bleuse, Kedad-Sidhoum, Monna, Mounié, Trystram 2015] Expensive PTAS [Bonifaci, Wiese 2012]
 - Low-complexity: 2-approx

[Canon, Marchal, Vivien 2017]
[Beaumont, Eyraud-Dubois, Kumar 2017]

► DAG: 6 - approx (LP rounding)

[Kedad-Sidhoum, Monna, Trystram 2015]

Existing online algorithms

► Independent tasks: 4 - competitive

[Imreh 2003]

3.85 - competitive

3.41 - approx

[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
- Average performance
- Validation on simulations

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 τ

Graph with
$$k = 2$$
, $n = 3$



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Phase 2 - same as phase 1, but are successors of the last task

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1

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Phase 3 - same as phase 2, but are successors of the last task

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kτ

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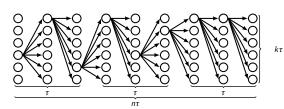
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Phase $x - \dots$

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

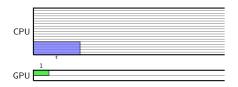


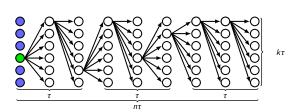
m CPUs, k GPUs

Theorem

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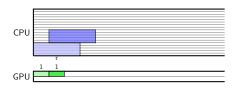


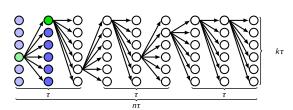
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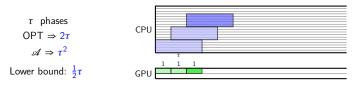
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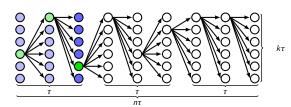




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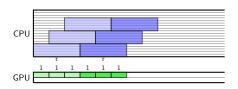


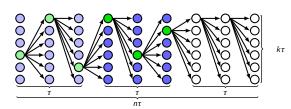


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 2τ phases OPT $\Rightarrow 3\tau$ $\mathscr{A} \Rightarrow 2\tau^2$ Lower bound: $\frac{2}{3}\tau$

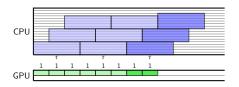


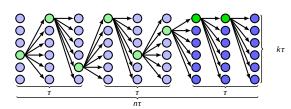


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$$n\tau$$
 phases OPT $\Rightarrow (n+1)\tau$ $\mathscr{A} \Rightarrow n\tau^2$ Lower bound: $\frac{n}{n+1}\tau$





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_i
- ► Allocate *T_i* 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 $\overline{p_i}$:

 \triangleright put T_i 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

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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
- 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
- ▶ ② 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

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**

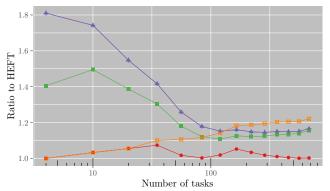
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Cholesky 4 types of tasks
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Synthetic STG set, 300 tasks, random GPU acceleration ($\mu = \sigma = 15$)

Ad-hoc one chain & independent tasks

Results for Cholesky graphs (lower is better)

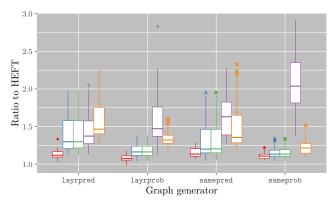
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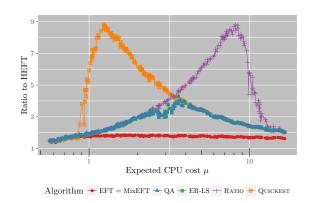
Algorithm
$$lacktriangledow$$
 EFT = MixEFT $lackledow$ QA $lackledow$ ER-LS $+$ RATIO $lackledow$ QUICKEST

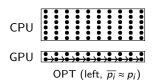
$$\frac{m}{k} = 10$$
 $\sqrt{\frac{m}{k}} \approx 3.3$ $\frac{\text{CPU time}}{\text{GPU time}} \in \{28, 26, 11, \underbrace{2}_{\text{POTRF}}\}$

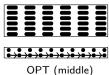
Results for synthetic graphs (lower is better)

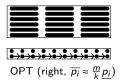


Results for 300-tasks ad-hoc graphs (lower is better)









Summary

- No online algo. is $<\sqrt{m/k}$ 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
- Parallel tasks