

# Composing multiple StarPU applications over heterogeneous machines: a supervised approach

#### Andra Hugo

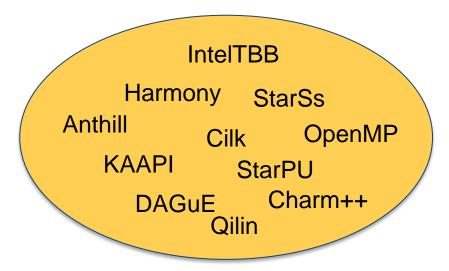
With Abdou Guermouche, Pierre-André Wacrenier, Raymond Namyst Inria, LaBRI, University of Bordeaux

# RUNTIME INRIA Group INRIA Bordeaux Sud-Ouest

#### The increasing role of runtime systems

Code reusability

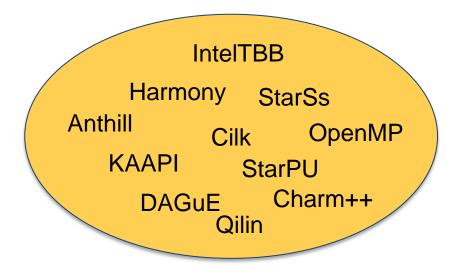
- Many HPC applications rely on specific parallel libraries
  - Linear algebra, FFT, Stencils
- Efficient implementations sitting on top of dynamic runtime systems
  - To deal with hybrid, multicore complex hardware
    - E.g. MKL/OpenMP, MAGMA/StarPU
  - To avoid reinventing the wheel!
- Some application may benefit from relying on multiple libraries
  - Potentially using different underlying runtime systems...



#### The increasing role of runtime systems

Code reusability

- Many HPC applications rely on specific parallel libraries
  - Linear algebra, FFT, Stencils
- Efficient implementations sitting on top of dynamic runtime systems
  - To deal with hybrid, multicore complex hardware
    - E.g. MKL/OpenMP, MAGMA/StarPU
  - To avoid reinventing the wheel!
- Some application may benefit from relying on multiple libraries
  - Potentially using different underlying runtime systems...



Are parallel libraries ready to run simultaneously over the same hardware resources

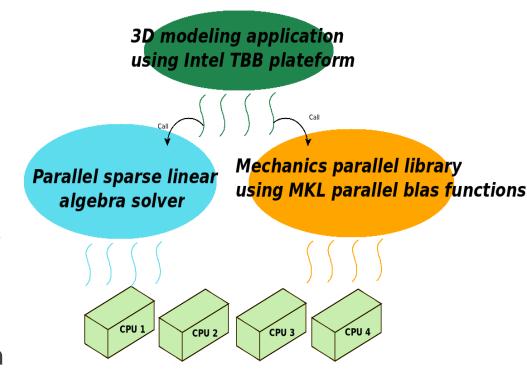


(nria-

## Struggle for resources

Interferences between parallel libraries

- Parallel libraries typically allocate and bind one thread per core
  - Bypass the OS scheduler
  - Control cache utilization
- Each library is unaware of the resource management of the other ones
  - => resource oversubscription
- Optimizations (cache affinity, memory reuse, etc.) are strongly affected



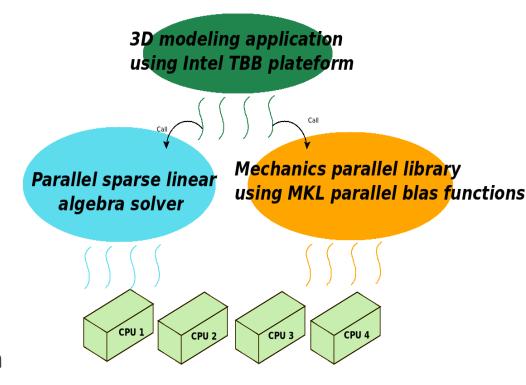


## Struggle for resources

 Parallel libraries typically allocate and bind one thread per core

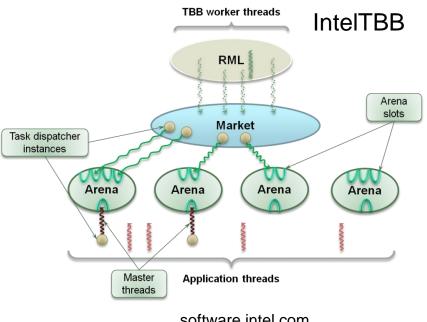
- Bypass the OS scheduler
- Control cache utilization
- Each library is unaware of the resource management of the other ones
  - => resource oversubscription
- Optimizations (cache affinity, memory reuse, etc.) are strongly affected

Interferences between parallel libraries



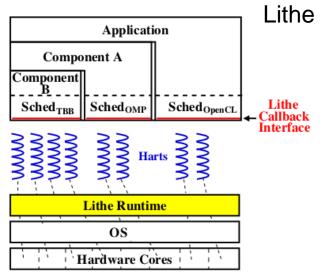
=> Composability problem

## **Composability problem**



software.intel.com

#### How to deal with it?



Phd Thesis: Cooperative Hierarchical Resource Management for Efficient

Composition of Parallel Software, Heidi Pan, 2010

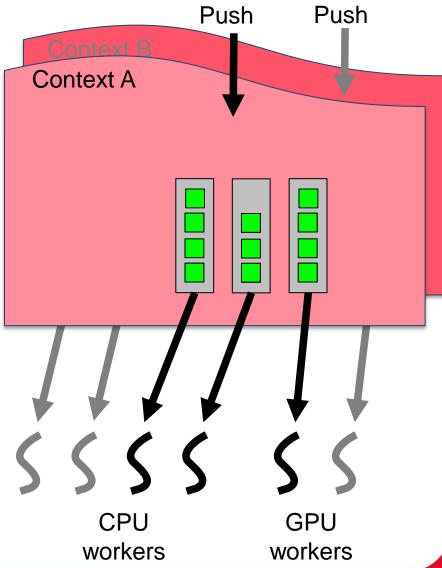
- Advanced environments allow partitioning hardware resources
  - IntelTBB:
    - The pool of workers are split in arenas
  - Lithe
    - Resource sharing management interface
    - Harts are transferred between parallel libraries
- Main challenge: Automatically adjusting the amount of resources allocated to each library



## **Scheduling Contexts**

- Isolate concurrent parallel codes
- Similar to lightweight virtual machines
- Run on top of subsets of available
   PUs
- Minimize interferences
- Enforce data locality
- Contexts may expand and shrink
  - Hypervised approach
  - Maximize overall throughput
  - Use dynamic feedback both from application and runtime



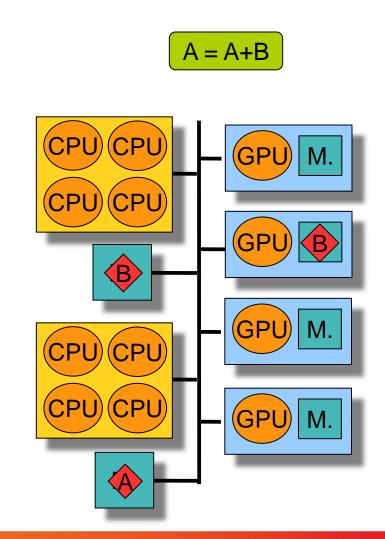




## Using StarPU as an experimental platform

A runtime system for \*PU architectures for studying resource negociation

- The StarPU runtime system
  - Dynamically schedule tasks on all processing units
    - See a pool of heterogeneous processing units
  - Avoid unnecessary data transfers between accelerators
    - Software VSM for heterogeneous machines





#### **Overview of StarPU**

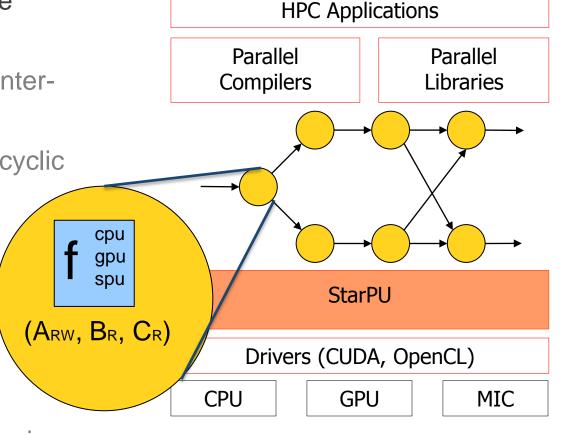
Maximizing PU occupancy, minimizing data transfers

 Accept tasks that may have multiple implementations

 Together with potential interdependencies

Leads to a dynamic acyclic graph of tasks

- Data-flow approach
- Scheduling hints
- Open, general purpose scheduling platform
  - Scheduling policies = plugins



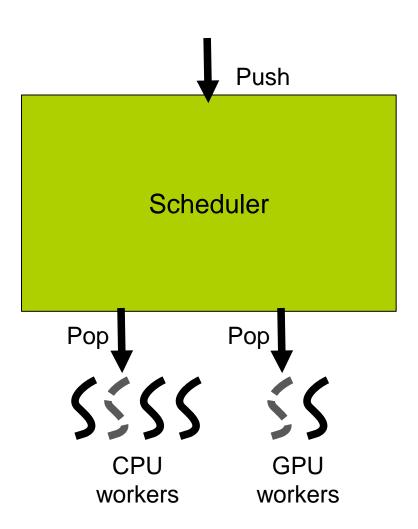


## Tasks scheduling

- When a task is submitted, it first goes into a pool of "frozen tasks" until all dependencies are met
- Then, the task is "pushed" to the scheduler
- Idle processing units actively poll for work ("pop")
- What happens inside the scheduler is... up to you!

How does it work?

- 10

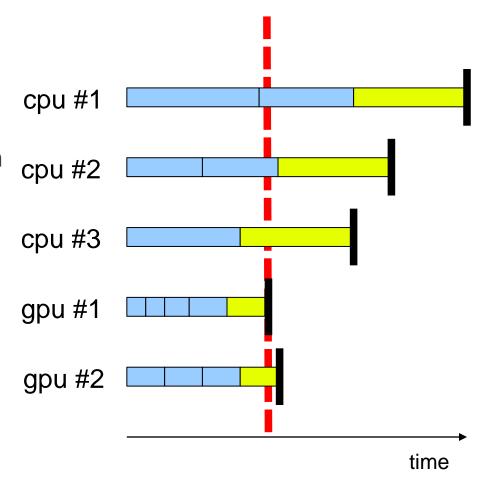




## Dealing with heterogeneous architectures

Performance prediction

- Build-in schedulers:
  - E.g. Minimum Completion
     Time heuristic
- Task completion time estimation
  - History-based performance models
- Data transfer time estimation
  - Sampling based on off-line calibration

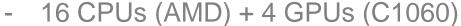


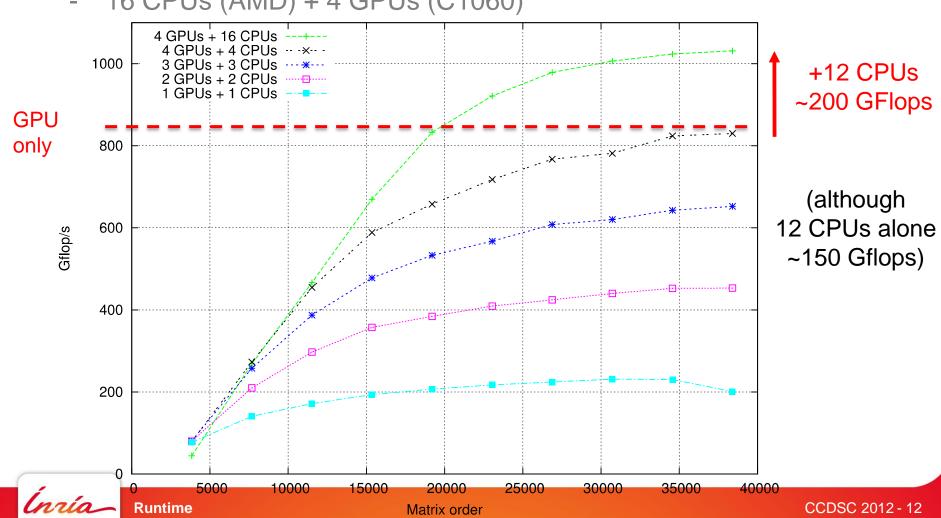


#### MAGMA with StarPU

#### With University of Tennessee & INRIA HiePACS

QR decomposition

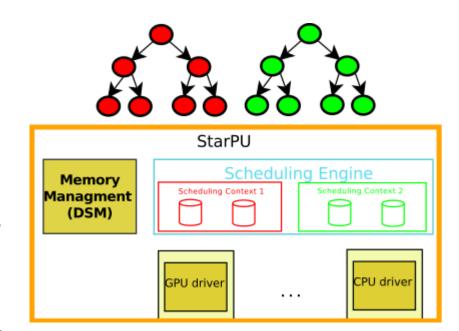




## Scheduling Contexts in StarPU

#### Extension of StarPU

- Heterogeneous machines
- Virtualization of resources
- Each context features its own scheduler
- Scalability workaround
- Contexts may share processing units
  - Avoid underutilized resources
- Allocation of resources
  - The programmer entirely defines it
  - The programmer specifies an interval and leaves the contexts negotiate
  - The hypervisor computes it



⇒ They try to be considerate with each other's needs

(nría-

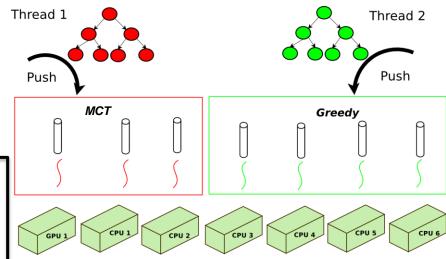
## Scheduling contexts in StarPU

Easily use contexts in your application

```
int resources1[3] = {CPU_1, CPU_2, GPU_1};
int resources2[4] = {CPU_3, CPU_4, CPU_5,
CPU_6};
```

/\* define the scheduling policy and the table of resource ids \*/

```
sched_ctx1 =
starpu_create_sched_ctx("mct",resources1,3);
sched_ctx2 =
starpu_create_sched_ctx("greedy",resources2,4);
```

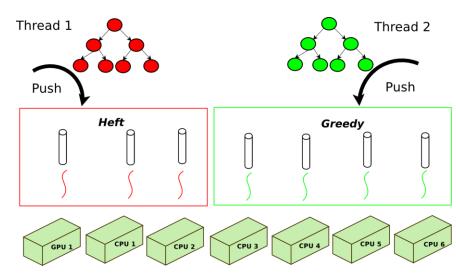


## Scheduling contexts in StarPU

#### Easily use contexts in your application

```
sched_ctx1 =
starpu_create_sched_ctx("heft",resources1,3);
```

```
sched_ctx2 =
starpu_create_sched_ctx("greedy",resources2,4);
```



#### // thread 1:

/\* define the context associated to kernel 1 \*/
starpu\_set\_sched\_ctx(sched\_ctx1);

/\* submit the set of tasks of the parallel kernel 1\*/

for( i = 0; i < ntasks1; i++)
 starpu\_task\_submit(tasks1[i]);</pre>

#### // thread 2:

/\* define the context associated to kernel 2 \*/
starpu\_set\_sched\_ctx(sched\_ctx2);

/\* submit the set of tasks of parallel kernel 2\*/
for( i = 0; i < ntasks2; i++)

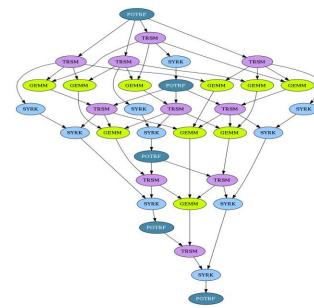
starpu\_task\_submit(tasks2[i]);

(nría

## **Experimental evaluation**

#### Platform and Application

- 9 CPUs (two Intel hexacore processors, 3 cores devoted to execute GPU drivers) + 3 GPUs
- MAGMA Linear Algebra Library
  - Implementation based on StarPU
  - Cholesky Factorization kernel
- Euler3D solver
  - Computational Fluid Dynamic benchmark
  - Rodinia benchmark suite
  - Iterative solver for 3D Euler equations for compressible fluids
  - Implementation based on StarPU



MAGMA – Cholesky Factorization

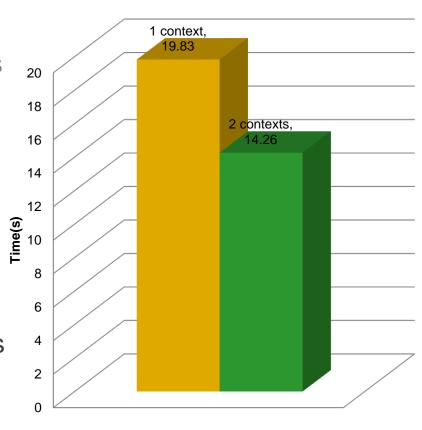
## Composing Magma and the Euler3D solver

Different parallel kernels

#### CFD:

- Domain decomposition parallelization
- Independent tasks per iteration
- Dependencies between iterations
- Strong affinity with GPUs
- 2 sub-domains: 2 GPUs
- Cholesky Factorization:
  - Scalable on both CPUs & GPUs
  - 1GPU & 9 CPUs
  - Large number of tasks
- Contexts enforce locality constraints

## CFD And Cholesky Factorization

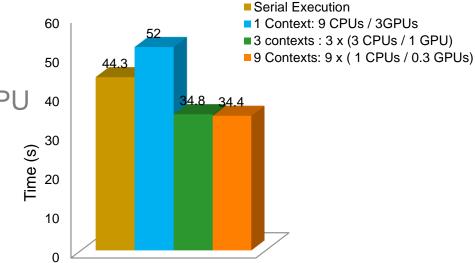




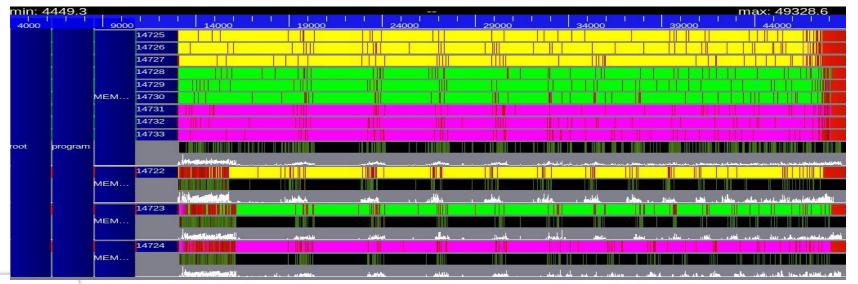
## What about 9 Cholesky factorizations...?

#### Gain performance from data locality

- Mixing parallel kernels:
  - Unnecessary data transfers
     between Host Memory & GPU
     memory -> blocking waits
  - Memory flushes



- 18

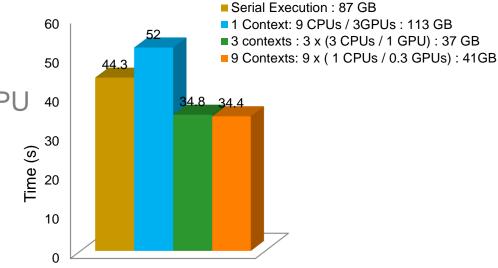


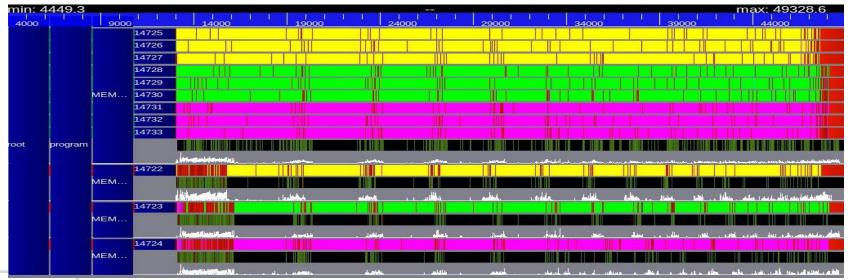


## What about 9 Cholesky factorizations...?

#### Gain performance from data locality

- Mixing parallel kernels:
  - Unnecessary data transfers
     between Host Memory & GPU
     memory -> blocking waits
  - Memory flushes



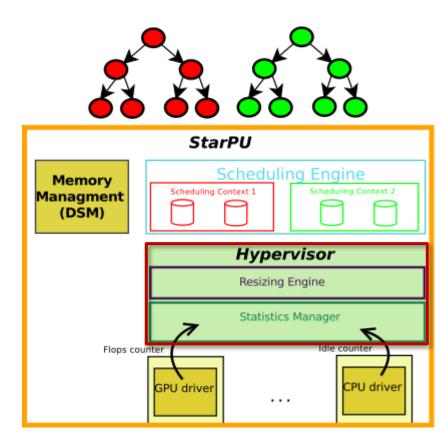




## The Hypervisor

What if static dimensioning doesn't work?

- Dynamically resizes scheduling contexts
- Triggers resizing:
  - At some instances of time
    - feedback from the application
  - When the initial configuration deteriorates the performances
    - feedback from the runtime
    - different metrics: Idle resources,
       Speed of the contexts
- User's constraints for the negotiation of resources
  - Idleness limit
  - Resize interval limitation

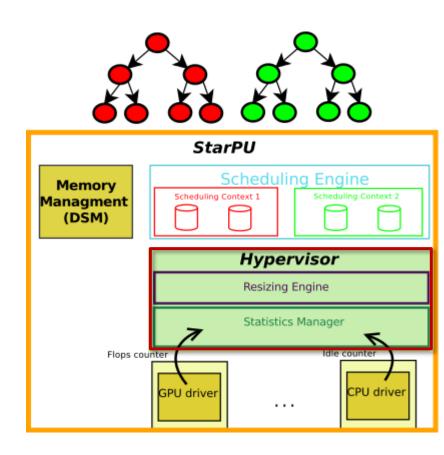




## The Hypervisor

What if static dimensioning doesn't work?

- Stores contexts information and resource performance statistics
- Policies for resizing contexts
  - Minimize the makespan
  - Need the workload of the parallel codes
  - Linear programs to compute the best allocation of resources



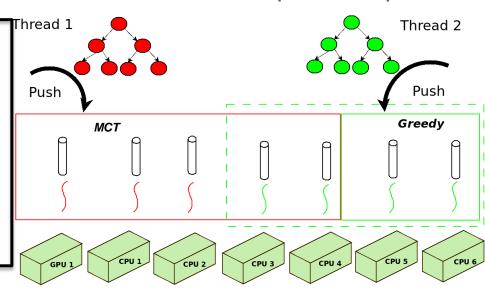


## The Hypervisor in StarPU

#### Simple example

/\* select an existing resizing policy \*/
struct hypervisor\_policy policy;
policy.name = "min\_makespan\_policy";

/\* initialize the hypervisor and
set its resizing policy \*/
sched\_ctx\_hypervisor\_init(policy);



#### Hypervisor

/\* register context 1 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx1);

/\* register context 2 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx2);

/\* define the constraints for the resizing \*/
sched\_ctx\_hypervisor\_ctl(sched\_ctx1,
HYPERVISOR\_MIN\_CPU\_WORKERS, 3,

HYPERVISOR\_MAX\_CPU\_WORKERS, 7, NULL):

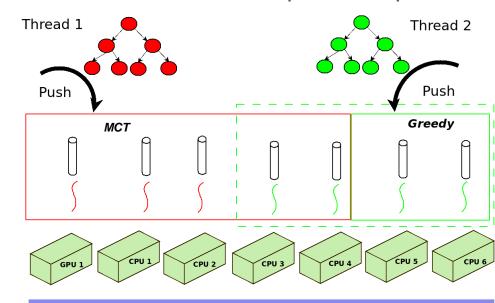
(nría\_

## The Hypervisor in StarPU

/\* select an existing resizing policy \*/
struct hypervisor\_policy policy;
policy.name = "min\_makespan\_policy";

/\* initialize the hypervisor and
set its resizing policy \*/
sched\_ctx\_hypervisor\_init(policy);

#### Simple example



#### Hypervisor

/\* register context 1 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx1);

/\* register context 2 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx2);

/\* define the constraints for the resizing \*/
sched\_ctx\_hypervisor\_ctl(sched\_ctx1,
HYPERVISOR\_MIN\_CPU\_WORKERS, 3,
HYPERVISOR\_MAX\_CPU\_WORKERS, 7,
NULL);

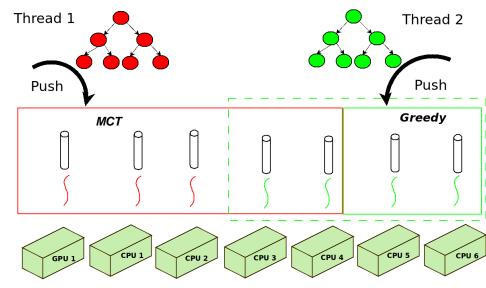
Inria

## The Hypervisor in StarPU

/\* select an existing resizing policy \*/
struct hypervisor\_policy policy;
policy.name = "min\_makespan\_policy";

/\* initialize the hypervisor and
set its resizing policy \*/
sched\_ctx\_hypervisor\_init(policy);

#### Simple example



#### Hypervisor

/\* register context 1 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx1);

/\* register context 2 to the hypervisor \*/
sched\_ctx\_hypervisor\_register\_ctx(sched\_ctx2);

/\* define the constraints for the resizing \*/
sched\_ctx\_hypervisor\_ctl(sched\_ctx1,
HYPERVISOR\_MIN\_CPU\_WORKERS, 3,
HYPERVISOR\_MAX\_CPU\_WORKERS, 7,
NULL);

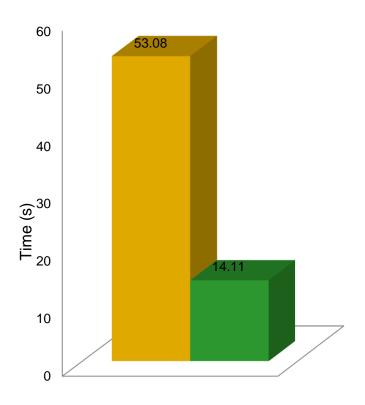


## Dealing with non scalable kernels

- CFD decomposed in 2 sub-domains
- Static distribution:
  - CFD: 3 GPUs
  - Cholesky Factorization: 9 CPUs
- Hypervisor's intervention:
  - CFD: 2GPUs
  - Cholesky Factorization: 1 GPU & 9
     CPUs

#### Idleness-based policies

- Static distribution of resources
- Dynamically adjusted distribution of resources

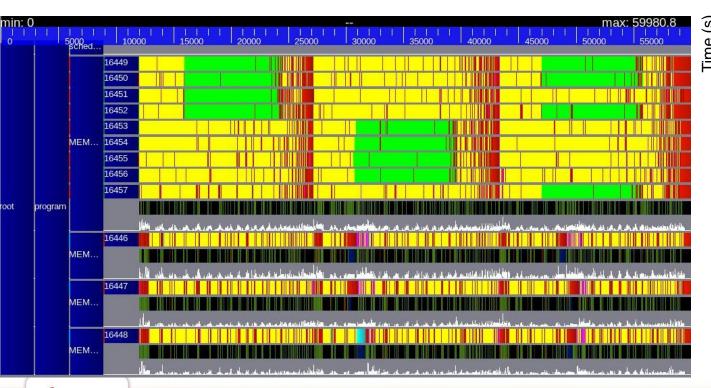


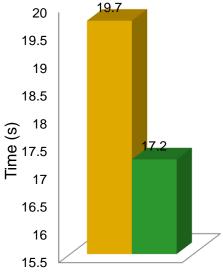


## Feedback of the application

When to resize?

- 2 streams of parallel kernels
- 1 of them pops in from time to time
- The hypervisor: assigns some CPUs to the intruder





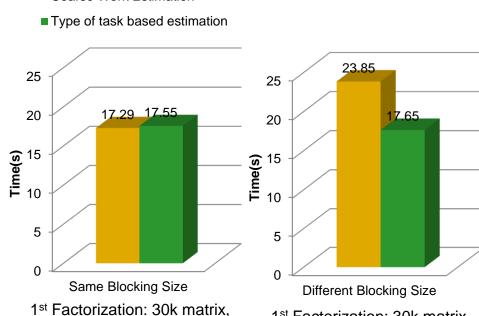
- Overlapping contexts
- Dynamically adjusted distribution of resources

Inria

## Facing irregular applications

#### Speed-based resizing policies

- Composing parallel kernels facing different granularities
- Same workload, different number of tasks
- Coarse work estimation:
  - Flops statistics
  - Same granularity for the concurrent parallel kernels
- Type of task based estimation:
  - Number of tasks statistics
  - Considers the type of tasks



2<sup>nd</sup> Factorization: 15k matrix, 2<sup>nd</sup> Factorization: 15k matrix, 960 block size

192 block size

**Runtime** - 27

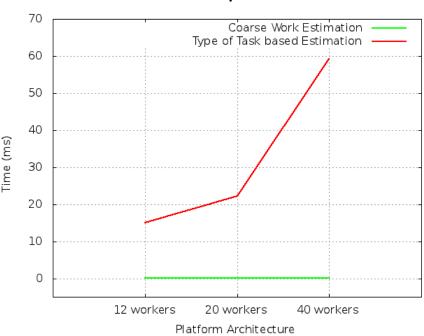
Coarse Work Estimation

## Which one is the best for our application?

Compromise between efficiency and precision?

- It depends ...
- Coarse Work estimation
  - Faster
  - Less accurate
  - Irregular application may require it multiple times
- Type of Task based Estimation
  - Higher complexity
  - More precise
  - Useful if the resizing is required a few times

#### Cost of the policies





#### **Conclusion & Future Work**

- Scheduling Contexts allow you use multiple parallel libraries simultaneously
  - Currently implemented in StarPU runtime system
  - A Hypervisor dynamically shrinks / extends contexts
- Further Work
  - New metrics to guide resizing
  - More intelligent sharing of resources (GPUs)
  - And much more!



#### **Potential collaborations**

- Extend the techniques to other runtime systems
  - Intel TBB / OS/R / Charm++ / OCR / Unistack
- Experiment with code-coupling applications
  - Great expected benefit from running multiple kernels concurrently
- Experiment with different hypervising strategies
- Any collaboration topic related to (task-based) runtime systems for (hybrid) multicore machines is welcome!

