

# RESEARCH QUESTION

How might programmers write highly parallel programs in a mainstream language like C++?

## **LOOKING AHEAD**

Agency: a lower-level framework for parallel execution

```
void saxpy(size_t n, float a, const float* x, const float* y, float* z)
{
  using namespace agency;

  bulk_invoke(par(n), [=] (parallel_agent &self)
  {
    int i = self.index();
    z[i] = a * x[i] + y[i];
  });
}
```



## **BEGIN AT THE BEGINNING**

Independent loop iterations represent latent parallelism

```
void saxpy(int n, float a, float *x, float *y)
{
    // Sequential code with latent parallelism
    for(int i=0; i<n; ++i)
    {
        y[i] = a*x[i] + y[i];
    }
}</pre>
```



## PARALLEL LOOPS IN C++17

Library implementation of parallel constructs

```
void saxpy(int n, float a, float *x, float *y)
 auto I = interval(0, n);
  std::for each(std::execution::par, I.begin(), I.end(), [=](int i)
     y[i] = a*x[i] + y[i];
```



## PARALLEL LOOPS

## Increasingly common in standard languages

OpenMP	<pre>#pragma omp parallel for for(int i=a; i<b; ++i)="" pre="" {="" }<=""></b;></pre>	
OpenACC	<pre>#pragma acc loop for(int i=a; i<b; ++i)="" pre="" {="" }<=""></b;></pre>	
Fortran 2008	DO CONCURRENT (I=1:N) END DO	
C++17	<pre>std::for_each(std::execution::par, begin, end, [](int i) { });</pre>	



### STANDARD TEMPLATE LIBRARY

Higher-level library built around algorithms

for\_each(begin, end, function);



Operator

Data



**Function** 

A named pattern of computation and communication.

One or more collections to operate on.

Caller-provided function object injected in pattern.



## C++17 PARALLEL STL

Algorithms + Execution Policies

for\_each(par, begin, end, function);



**Execution Policy** 

Specify *how* operation may execute.

C++ Extensions for Parallelism

Draft technical specification: <a href="http://wq21.link/n4507">http://wq21.link/n4507</a>



## **EXECUTION POLICIES**

#### Specify how algorithms may execute

POLICY NAME	MEANING
seq	Sequential execution alone is permitted.
par	Parallel execution is permitted.
par_unseq	Vectorized parallel execution is permitted.

parallel :: provided function objects can be executed in any order on one or more threads. vectorized :: provided function objects can be also be interleaved when on one thread.



## PARALLEL ALGORITHMS

Many useful patterns beyond loops

Parallelizable algorithms in STL

New additions for parallelism

for\_each

reduce

transform

exclusive scan

copy if

inclusive scan

sort

transform\_reduce

set intersection

transform inclusive scan

etc.

transform exclusive scan



## LARGE VOCABULARY OF ALGORITHMS

How might parallel programmers implement these algorithms?

## AGENCY FRAMEWORK

Developing a lower-level framework for parallel execution

A C++ template library for abstracting execution

Central primitives are execution agents

Portable, with efficient mapping to underlying platform

Extensibility via executors

## AGENCY FRAMEWORK

#### Developing a lower-level framework for parallel execution

```
void saxpy(size_t n, float a, const float* x, const float* y, float* z)
{
  using namespace agency;

  bulk_invoke(par(n), [=](parallel_agent &self)
  {
    int i = self.index();
    z[i] = a * x[i] + y[i];
  });
}
```

## AGENCY FRAMEWORK

#### Developing a lower-level framework for parallel execution

```
void saxpy(size_t n, float a, const float* x, const float* y, float* z)
{
  using namespace agency;

auto executor = get_desired_executor();

bulk_invoke(par(n).on(executor), [=](parallel_agent &self)
  {
   int i = self.index();
   z[i] = a * x[i] + y[i];
  });
}
```



## **AGENCY'S EXECUTION POLICIES**

Specify how control structures execute work

POLICY NAME	MEANING
seq	Sequential execution alone is permitted.
par	Parallel execution is permitted.
par_unseq	Vectorized parallel execution is permitted.
unseq	Vectorized execution is permitted in the current thread.
con	Concurrent forward progress is mandatory.

## IMPLEMENTING ALGORITHMS

#### Requires suitable lower level constructs

```
template<class Range>
int sum(Range&& data)
    // organize data into partitions
    auto partitions = make partitioned view(data, par.executor().shape());
      sum each partition
   auto partial sums = bulk invoke(par, [=](auto& self)
        return sum(seq, partitions[self.index()]);
   });
    // reduce the partial sums
   return sum(seq, partial sums);
```



## IMPLEMENTING ALGORITHMS

#### But waiting can be harmful

```
template<class Range>
int sum(Range&& data)
    // organize data into partitions
    auto partitions = make partitioned view(data, par.executor().shape());
       sum each partition
    auto partial sums = bulk invoke(par, [=](auto& self)
        return sum(seq, partitions[self.index()]);
    });
                                                                                      wait
    // reduce the partial sums
    return sum(seq, partial sums);
                                                                                      wait
```

## **ADDING ASYNCHRONY**

#### **Future-based machinery**

```
template<class Range>
future<int> async sum(Range&& data)
   // organize data into tiles
    auto tiles = tile evenly(data, par.executor().unit shape());
    // sum each tile
    auto partial sums = bulk async(par, [=](auto& self)
        return sum(seq, tiles[self.index()]);
    });
    // reduce the partial sums
   return partial sums.then([=](auto& partial sums)
        sum(seq, partial sums);
   });
```

## **ADDING ASYNCHRONY**

#### But phasing can be inefficient

```
template<class Range>
future<int> async sum(Range&& data)
   // organize data into tiles
    auto tiles = tile evenly(data, par.executor().unit shape());
    // sum each tile
    auto partial sums = bulk async(par, [=](auto& self)
        return sum(seq, tiles[self.index()]);
    });
    // reduce the partial sums
   return partial sums.then([=](auto& partial sums)
        sum(seq, partial sums);
   });
```



### ADDING CONCURRENCY

#### Flexible synchronization

```
template<class Range>
future<int> async sum(Range&& data)
   // organize data into tiles
    auto tiles = tile evenly(data, con.executor().unit shape());
   // sum each tile
   return bulk async(con, [=] (auto& self, vector<int>& partials) -> single result<int>
       partials[self.index()] = sum(seq, tiles[self.index()]);
        self.wait();
       // agent 0 computes the final sum
        if(self.index() == 0) return sum(seq, partials);
       else return ignore;
    share<vector<int>>(partitions.size()));
```

# WHERE DOES EXECUTION ACTUALLY HAPPEN?

#### Diverse Control Structures

```
bulk_async(...) for_each(...)
    sort(...) bulk_invoke(...)
your_favorite_control_structure(...)
```

Multiplicative Explosion



Diverse Execution Resources Operating System Threads

SIMD vector units

Thread pool schedulers

GPU runtime

OpenMP runtime

Fibers

Diverse Control Structures

```
bulk_async(...) for_each(...)
    sort(...) bulk_invoke(...)
your_favorite_control_structure(...)
```

Uniform Abstraction

## **Executors**

Diverse Execution Resources Operating System Threads

SIMD vector units

Thread pool schedulers

GPU runtime

OpenMP runtime

Fibers

## **EXECUTOR** FRAMEWORK

#### Abstract platform details of execution

Create execution agents

Manage data they share

Advertise semantics

Mediate dependencies

```
class sample executor
public:
  using execution category = ...;
  using shape type = tuple<size t,size t>;
  template<class T>
  using future = ...;
  template<class T>
  future<T> make ready future(T&& value);
  template<class Function, class Factory1, class Factory2>
  future <...> bulk async execute (Function f,
                                  shape type shape,
                                  Factory1 result factory
                                  Factory2 shared factory);
  . . .
};
```

## PURPOSE OF EXECUTORS

#### Provide control over where/how execution happens

Placement is, by default, at discretion of the system.

```
for_each(par, I.begin(), I.end(), [](int i) { y[i] += a*x[i]; });
```

In some cases, the programmer might want to control placement.

```
auto exec1 = choose_some_executor();
auto exec2 = choose_another_executor();

for_each(par.on(exec1), I.begin(), I.end(), ...);
for_each(par.on(exec2), I.begin(), I.end(), ...);
```

## PURPOSE OF EXECUTORS

Control relationship with calling thread

```
async(launch_flags, function);
```

```
async(executor, function);
```



## **EXECUTOR INTERFACE**

#### Semantic types exposed by executors

TYPE	MEANING
execution_category	Scheduling semantics amongst agents in a task. (sequenced, vector-parallel, parallel, concurrent)
shape_type	Type for indexing bulk launch of agents. (typically n-dimensional integer indices)
future <t></t>	Type for synchronizing asynchronous activities.  (follows interface of std::future)

## **EXECUTOR INTERFACE**

#### Three core constructs for launching work

Single-agent

```
result sync_execute(Function f);
future<result> async_execute(Function f);
future<result> then_execute(Function f, Future& predecessor);
```

Multi-agent

## SIMPLE EXAMPLE

#### An executor targeting OpenMP

```
struct omp executor
 using execution category = parallel execution tag;
 template<class Function, class R, class S>
 auto bulk sync execute (Function f, size t n, R result f, S shared f)
   auto result = result f();
   auto shared arg = shared f();
   #pragma omp parallel for
   for(size t i = 0; i < n; ++i)
     f(i, result, shared arg);
   return result;
};
```

## **EXECUTOR COLLECTIONS**

Present a collection of executors as a single logical executor

```
my executor ex0, ex1, ex2;
executor array<my executor> exec array{ex0, ex1, ex2};
exec array.bulk sync execute([](auto idx)
  // work on task (i, j)
},
{3, 2} // create nested agent groups
);
```

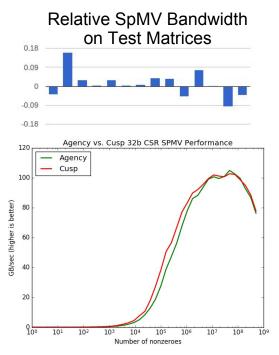
## **EXECUTOR COLLECTIONS**

#### Transparent multi-GPU

```
// create an executor collecting all the GPUs in the system
cuda::multidevice executor exec(cuda::all devices());
// create some arrays spanning all the GPUs in the system
cuda::multidevice array<float> x = ...
cuda::multidevice array<float> y = ...
cuda::multidevice array<float> z = ...
// execute only on GPU 0
saxpy(par.on(exec[0]), x, y, z);
// execute across all GPUs
saxpy(par.on(exec), x, y, z);
```

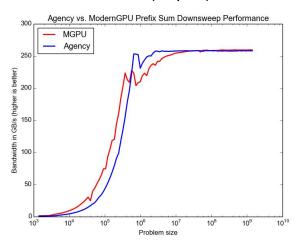
## **PERFORMANCE**

## **ABSTRACTIONS WITH MINIMAL OVERHEAD**



SpMV performance vs. hand-written CUDA GeForce GTX 1070 (Pascal)

Prefix sum performance vs. hand-written CUDA Titan X (Kepler)



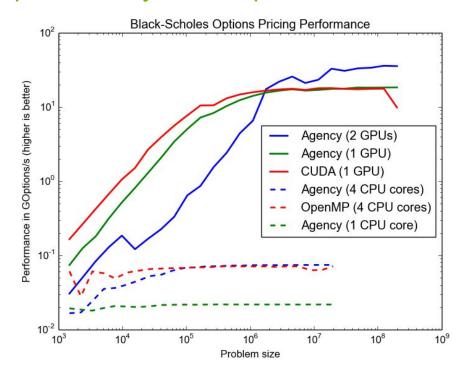
## TARGET-INDEPENDENT CONSTRUCTS

Performance portability for *simple* memory access patterns

Write a portable parallel program

Choose execution target via executors

GPU execution with Unified Memory support provided by Pascal architecture



## SINGLE-NODE HPGMG

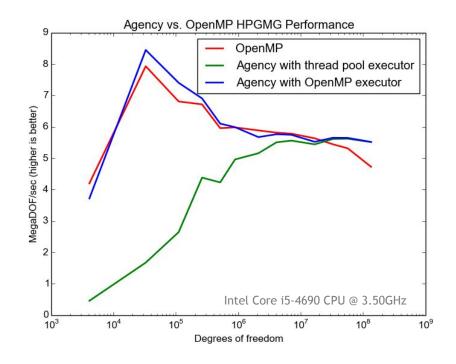
#### Agency CPU implementations vs. OpenMP-based original

Control behavior via choice of executor:

- using OpenMP runtime
- using pool of C++11 std::thread

Quite similar to OpenMP original when using OpenMP runtime

Reduction algorithm appears to account for better performance beyond 10<sup>7</sup> DoFs



## **SUMMARY**

C++17 will deliver standard, high-level parallel algorithms

Agency provides efficient lower-level abstractions for parallel algorithms

Executors are low-level components for abstracting platform details

## **AVAILABLE ONLINE**

#### Open source

On Github:

agency-library.github.io

Try with the latest CUDA:

developer.nvidia.com/cuda-toolkit

Quick start:

github.com/agency-library/agency/wiki/Quick-Start-Guide





