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AVT006: Heterogeneous Programming: Distributed Data Structures, Algorithms, and Views in C++

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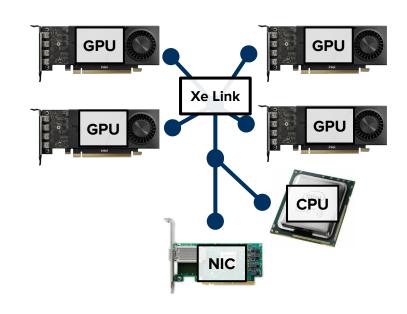
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Problem: writing parallel programs is hard

Multi-GPU, multi-CPU systems require partitioning data

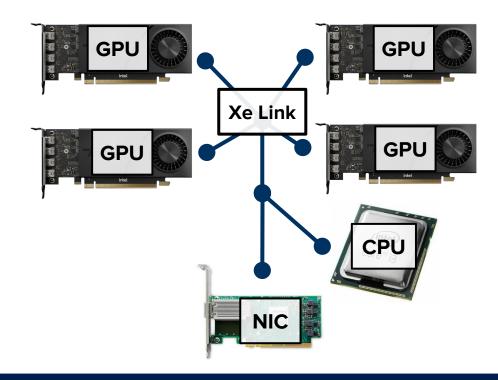
- Users must **manually split up data** amongst GPUs / nodes

 High-level mechanisms for data distribution / execution necessary.



Multi-GPU Systems

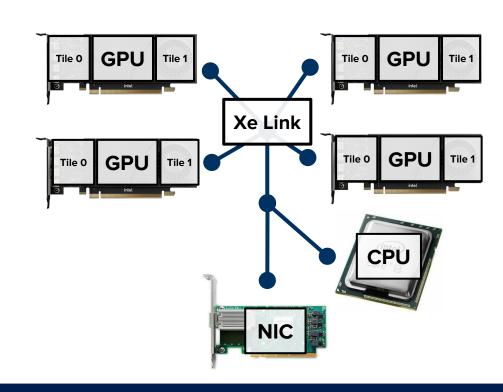
- NUMA regions:
 - 4+ GPUs
 - 2+ CPUs



Multi-GPU Systems

- NUMA regions:
 - 4+ GPUs
 - 2+ CPUs

- Systems becoming more hierarchical: even more memory domains
- Software needed to reduce complexity



Project Goals

- Offer high-level, standard C++
 distributed data structures
- Support distributed algorithms
- Achieve high performance for both multi-GPU, NUMA, and multi-node execution



Agenda

- 1. C++ standard parallelism
- 2. Distributed data structures
- 3. Code Demo
- 4. Performance
- 5. Complex data structures

C++ Standard Parallelism

C++ Parallelism

Data structures

- Organize data

Views

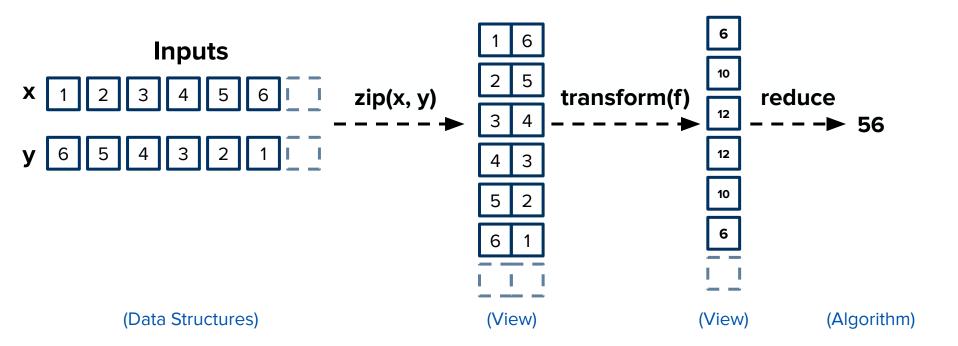
- Provide modified views of data

Algorithms

Operate on and modify data

```
using namespace std;
using namespace std::ranges;
using namespace std::execution;
float dot product(vector<float>& x,
                  vector<float>& y) {
  auto z = views::zip(x, y)
           views::transform([](auto element) {
             auto [a, b] = element;
             return a * b;
           });
 return reduce(par_unseq, z, 0, std::plus());
```

Dot Product Algorithm



Standard C++ Parallelism

Data structures

- Organize data

Views

- Lightweight, modified views of data

Algorithms

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Standard C++ Parallelism

- **Extensible:** with extensions, can automatically run on GPU

```
using namespace std;
using namespace std::ranges;
using namespace std::execution;
```

All depends on ranges,
 concept for iterating over data float dot_product(vector<float>& x,

Standard C++ Parallelism

- Extensible: with extensions, can automatically run on GPU
- All depends on ranges, concept for iterating over data

```
using namespace std;
using namespace std::ranges;
using namespace std::execution;
using namespace oneapi;
float dot product(device vector<float>& x,
                  device vector<float>& y) {
 auto z = views::zip(x, y)
           views::transform([](auto element) {
             auto [a, b] = element;
             return a * b;
           });
 auto dpl policy = ...;
 return dpl::reduce(dpl policy, z, 0, std::plus());
```

Ranges

C++ 20 introduced ranges

A range is a collection of values

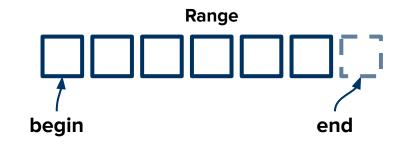
Range concepts provide a standard way to iterate over values

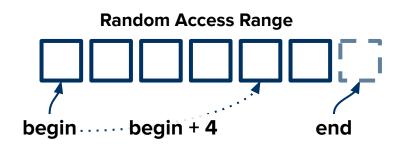


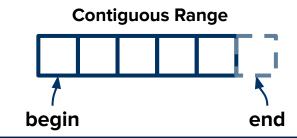
```
// Iteration
for (auto&& value : range) {
  printf("%d\n", value);
// Algorithms
auto r = std::ranges::reduce(range);
auto r = std::ranges::partial sum(range);
// Views
auto add_two = [](auto v) { return v + 2; };
auto view =
     std::ranges::transform view(range, add two);
```

Ranges API

- Have a begin() and end()
- Have a size() (usually)
- Random access: can access any element at random in constant time
- Contiguous: represents a contiguous block of memory



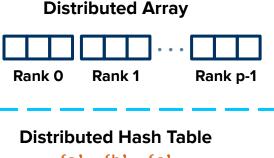


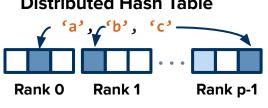


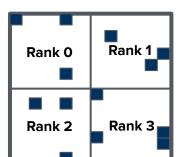
Distributed data structures **split up** data across multiple **segments**

Segments may be stored in different memory regions

We need a unified API for accessing these distributed data structures!







Distributed Matrix

Data is typically **partitioned** amongst processors into **segments**

Segments are remotely accessible, and are located on a single rank

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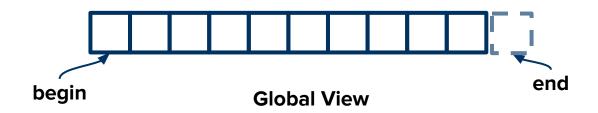


- 1. R is a standard range
- 2. R has segments()

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R needs two things to be a distributed range:

- 1. R is a standard range
- 2. R has segments()



Segmented View
(segments(r))

Segments (Remote Range)

Each of the segments in a distributed range is a remote range

A remote range is a **standard range**

—Plus it has a rank



Segments (Remote Range)

Each of the segments in a distributed range is a remote range

A remote range is a **standard range**

—Plus it has a rank

Algorithms can be implemented hierarchically.



Distributed Algorithms

- Algorithms use the distributed range concept (segments())
- Written hierarchically using oneDPL algorithms

```
using namespace dr::shp;
using namespace oneapi;
float reduce(auto policy,
             distributed vector<float>& v) {
  float init = 0.0f;
  for (auto&& segment : v.segments()) {
    auto device = devices()[segment.rank()];
    init += dpl::reduce(device, segment);
 return init;
```

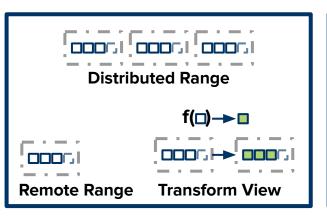
Distributed Views

- Views implement segments() by applying transformation to parents' segments
- Views can be built hierarchically

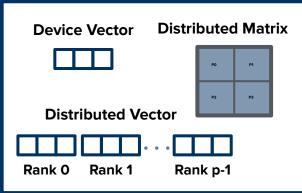
```
template <typename Range,
          typename Fn>
class transform_view {
  auto segments() {
    return base.segments()
            views::transform(
            [](auto&& segment) {
              return segment
                    views::transform(fn);
            });
```

Distributed Ranges Project

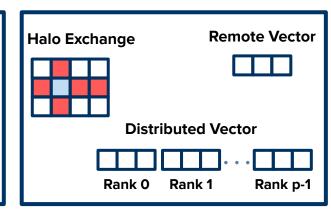
Shared Concepts and Views



GPU Data Structures and Algorithms ("shp")



MPI Data Structures and Algorithms ("mhp")



SYCL Codebase (shp)

- Data **automatically distributed** amongst **multiple GPUs**
- Distributed algorithms: each
 GPU calls into oneDPL
 algorithms

```
using namespace dr::shp;
float dot product(distributed vector<float>& x,
                  distributed vector<float>& y) {
  auto z = views::zip(x, y)
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SYCL Codebase (shp)

- Data **automatically distributed** amongst **multiple GPUs**
- Distributed algorithms: each
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return reduce(par_unseq, z, 0, std::plus());

return a * b;

using namespace dr::shp;

});

Multi-Node Codebase (mhp)

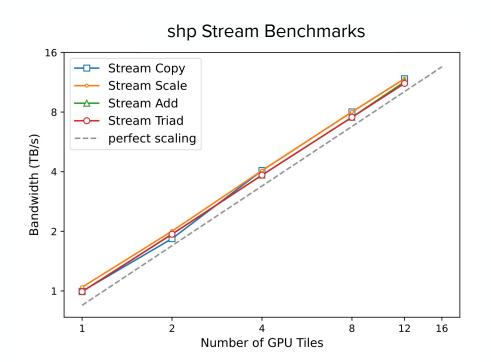
- Multi-process, SPMD program
- Data structures
 automatically distributed on multiple nodes using MPI
- Data structure constructors and algorithms are collective

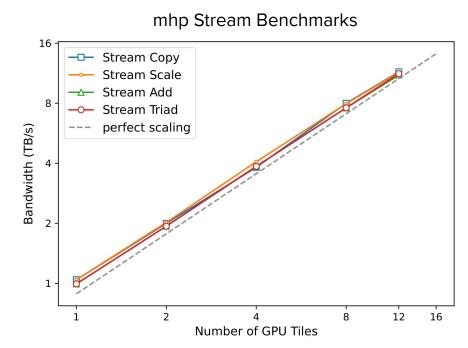
```
using namespace dr::mhp;
float dot product(distributed vector<float>& x,
                  distributed vector<float>& y) {
  auto z = views::zip(x, y)
            views::transform([](auto element) {
             auto [a, b] = element;
             return a * b;
           });
  return reduce(par_unseq, z, 0, std::plus());
```

Data Structure and Algorithms Demo

Performance

Stream Benchmarks

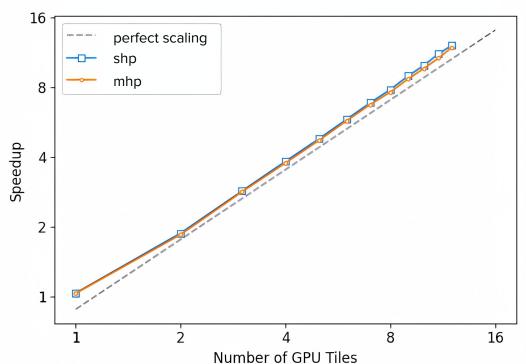






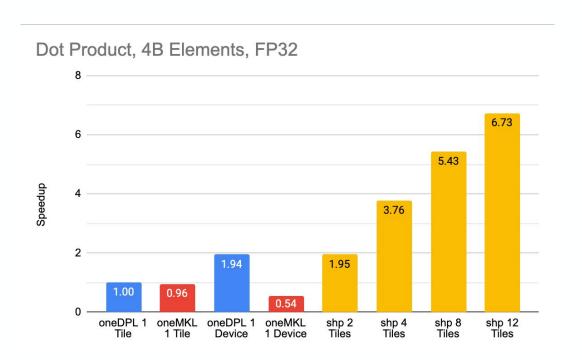
Black Scholes

Black Scholes - 2B Elements / Tile





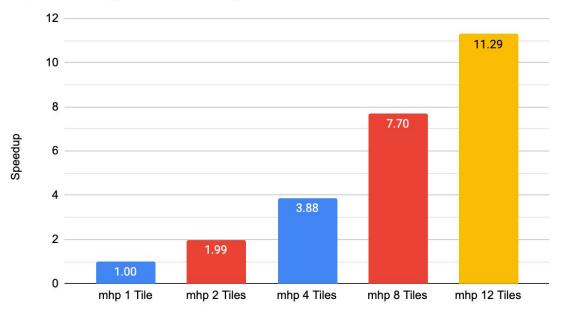
Dot Product - shp





Dot Product - mhp

Dot Product, 4B Elements, FP32

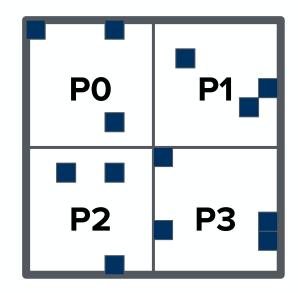




Beyond Standard Data Structures

Beyond Standard Data Structures - Matrices

- Can implement more complex data structures using distributed range abstraction
- Distributed matrix data structure splits up matrix

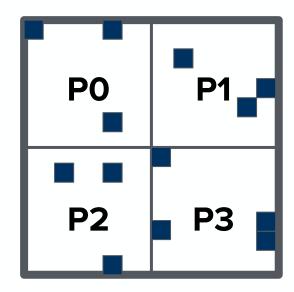




Beyond Standard Data Structures - Matrices

- Each tile is a **remote range** representing the submatrix
- All of these tiles together constitute the matrix

- Tiles can be **sparse** or **dense**





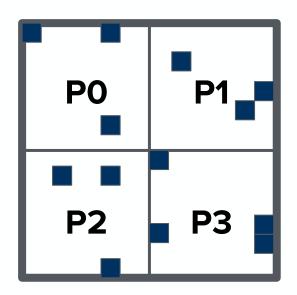
GraphBLAS C++ Matrix Concept

- When iterating through a matrix, observe an unordered sequence of tuples
- This works for all varieties of **sparse matrices**
- Can access other, data structure-specific iteration methods using customization points



Matrices - Can Also Access Tiles Individually

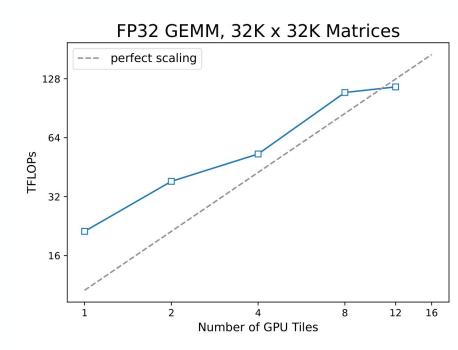
- tile() get remote view of tile
- get_tile() get copy of tile
- get_tile_async() get copy of tile, asynchronously





Matrix Multiply

- Implement an DMA-based,
 multi-GPU matrix multiply
- GPUs copy the tiles they need for the multiply





Call to Action

- Standard C++: Jump in, the water's fine!
- Our work is **open-source**: https://github.com/oneapi-src/distributed-ranges



Questions?



https://github.com/oneapi-src/distributed-ranges

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