



University of
BRISTOL



SC22
Dallas, hpc
TX accelerates.

Evaluating ISO C++ Parallel Algorithms on Heterogeneous HPC Systems

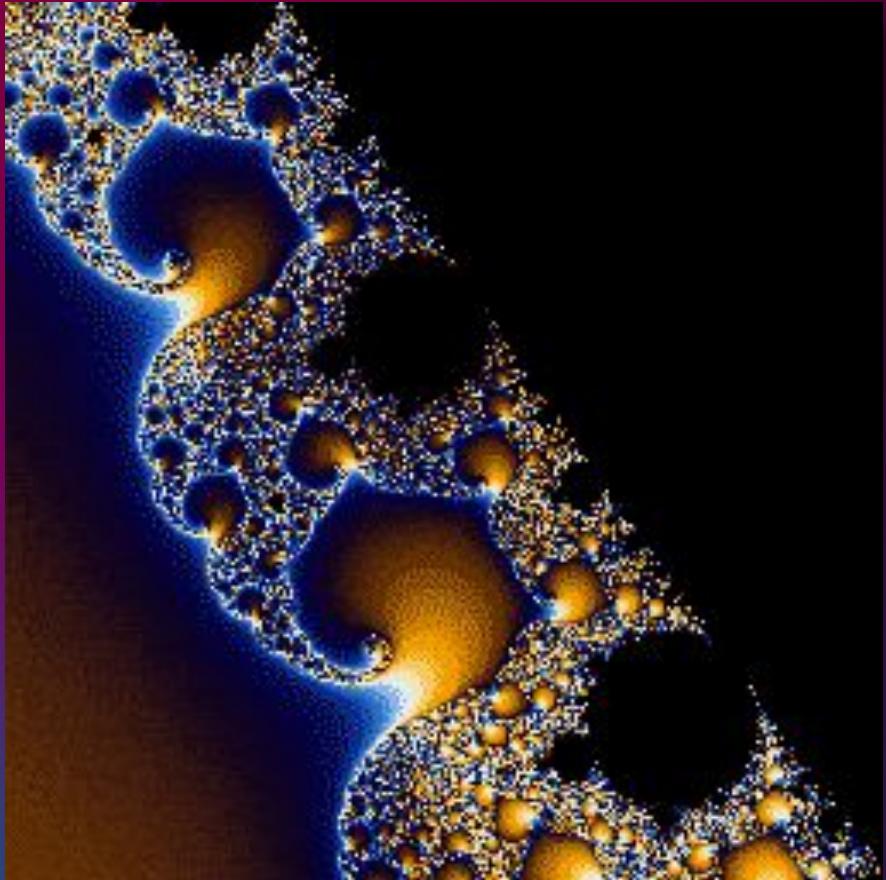
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<http://uob-hpc.github.io/>

Motivating example: Mandelbrot



Animated mandelbrot GIF in < 60 lines of ISO C++ 17*
(*excludes GIF library, shim header)

- Code compiles as-is on Clang, GCC, DPC++, NVC++
 - Runs multithreaded on CPUs
 - Runs on NVIDIA and Intel GPUs

<https://github.com/UoB-HPC/stdpar-mandelbrot>

Motivating example: Mandelbrot

```
struct Colour { uint8_t r, g, b, _ = 0xFF; /* mix ... */ }

template <typename N>
std::pair<std::complex<N>, int> mandelbrot(std::complex<N> c, int imax) {
    std::complex<N> z; int i = 0;
    while (std::abs(z) <= 4 && i < imax) { z = z * z + c; i += 1; }
    return {z, i};
}

Colour *buffer = /* allocate W*H*Colour */
shim::ranged<size_t> r(0, width * height);
std::for_each(shim::par_unseq, r.begin(), r.end(), [=, buffer](auto i) {
    std::array<Colour, 16> Colours{Colour{66, 30, 15}, {25, 7, 26}, {9, 1, 47}, {4, 4, 73}};
    const auto &[t, iter] = mandelbrot(std::complex<double>{X, Y}, 360);
    if (iter < 360) { // smoothed escape time colouring
        auto logZn = std::log(std::abs(t)) / 2;
        auto nu = std::log(logZn / std::log(2)) / std::log(2);
        auto c1 = Colours[size_t(std::floor(iter - nu)) % Colours.size()];
        auto c2 = Colours[size_t(std::floor(iter + 1 - nu)) % Colours.size()];
        buffer[i] = c2.mix(c1, std::fmod(double(iter) + 1 - nu, 1));
    } else buffer[i] = {0, 0, 0};
});
```

Mandelbrot kernel excerpt

Modern C++ constructs

- General control structure
 - Function calls
 - while/for/if/else
- Structured bindings
- Lambda captures
- Pointers (if captured by value)
- POD types
- Stack allocation

<https://github.com/UoB-HPC/stdpar-mandelbrot>

Overview

1. Background: ISO C++
2. C++ STL numeric/algorithm API
3. Index vs. Data traversal
4. C++17 PSTD Implementations
5. Experiments
 - a. Benchmark Platforms
 - b. Benchmark Mini-apps
 - c. Evaluation Results
6. Conclusion

Background: ISO C++

- Statically typed, unmanaged, low cost abstractions
- Rich standard library
- Committee driven: WG21
 - 3 year release cadence since C++11:
 - *C++14, C++17, C++20, C++23, ...*
- Multiple compiler implementations
 - GCC from FSF
 - Clang from multiple vendors
 - DPC++/ICPX from Intel
 - AOCC from AMD
 - NVHPC from Nvidia
 - MSVC from Microsoft



Introduction: C++ STL numeric/algorithm API (Serial)

```
template<class InputIt,
         class UnaryF>
UnaryF for_each(InputIt inFirst, InputIt inLast, UnaryF f) {
    for (; inFirst != inLast; ++inFirst) f(*inFirst);
    return f;
}

template<class InputIt,
         class OutputIt,
         class UnaryF>
OutputIt transform(InputIt inFirst, InputIt inLast, OutputIt outFirst, UnaryF f) {
    while (inFirst != inLast) *outFirst++ = f(*inFirst++);
    return outFirst;
}

template<class InputIt,
         class T,
         class CombineF,
         class MapF>
T transform_reduce(InputIt inFirst, InputIt inLast, T t, CombineF reduce, MapF transform) {
    T acc = t;
    while (inFirst != inLast) acc = reduce(acc, transform(*inFirst++));
    return acc;
}
```

Introduction: C++ STL numeric/algorithm API (C++17)

```
template<class Policy, class InputIt,
         class UnaryF>
UnaryF for_each(Policy&& p, InputIt inFirst, InputIt inLast, UnaryF f) {

    // Offload implementation, selected with the p argument
}

template<class Policy, class InputIt,
         class OutputIt,
         class UnaryF>
OutputIt transform(Policy&& p, InputIt inFirst, InputIt inLast, OutputIt outFirst, UnaryF f) {

    // Offload implementation, selected with the p argument
}

template<class Policy, class InputIt,
         class T,
         class CombineF,
         class MapF>
T transform_reduce(Policy&& p, InputIt inFirst, InputIt inLast, T t, CombineF reduce, MapF transform) {

    // Offload implementation, selected with the p argument
}
```

- std::execution::seq - Ordered sequential execution
- std::execution::unseq - Unordered sequential execution
- std::execution::par - Parallel ordered execution
- std::execution::par_unseq - Parallel unordered execution

Introduction: Data vs. Index Centric Traversal

Data Centric Traversal

```
auto exec = std::execution::par_unseq;

std::vector<T> xs = /*...*/;

std::for_each(exec, xs.begin(), xs.end(), [](T &x) {
    /* ... */
});

std::vector<T> ys(xs.size());

std::transform(exec, xs.begin(), xs.end(), ys.begin(), [](T &x) {
    return /* new value, witnessing x */
});
```

Index Centric Traversal (Naive)

```
auto exec = std::execution::par_unseq;

std::vector<T> xs = /*...*/;
// generate indices
std::vector<int> idxs(xs.size());
std::iota(idxs.begin(), idxs.end(), 0);
// idxs == {0,1,...xs.size()}
std::for_each(exec, idxs.begin(), idxs.end(), [](int i) {
    // i == 0 .. xs.size() - 1
});
```

Introduction: Counting Iterator

```
template <typename N> struct range {
    struct iterator {
        friend class range;
        using difference_type = typename std::make_signed_t<N>;
        using iterator_category = std::random_access_iterator_tag;
        using value_type = N;
        using reference = N;
        using pointer = const N*;
        iterator &operator ++() { ++i_; return *this; }
        iterator operator ++(int) { iterator copy(*this); ++i_; return copy;
    }
    // operator implementation for [], *, +, -, --, +=, -=, ==, >=, <=, >, < omitted
protected: explicit iterator(N start) : i_(start) {}
private: N i_;
};
iterator begin() const { return begin_; }
iterator end() const { return end_; }
range(N begin, N end) : begin_(begin), end_(end) {}
private: iterator begin_, end_;
};
```

C++17



```
range<int> r(0, N);
std::for_each(r.begin(), r.end(), [](int i) {
    // i == 0 .. N
});
```

C++17

```
auto r = std::views::iota(0).begin();
std::for_each_n(r, N, [](int i) {
    // i == 0 .. N
});
```

C++20

Introduction: ISO C++ PSTL Implementations



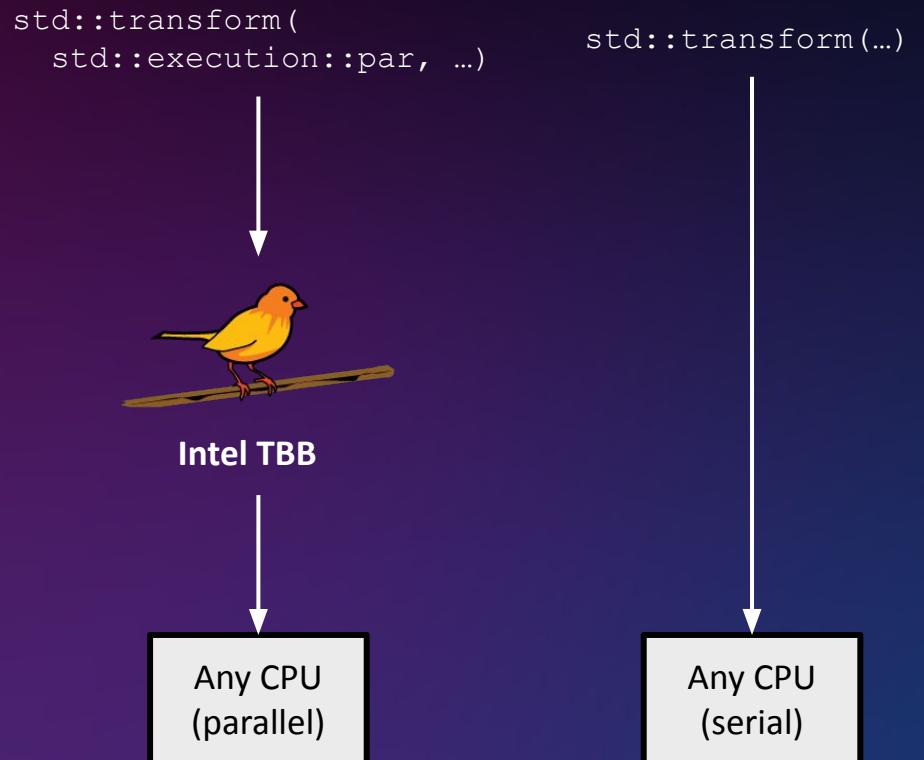
*Deprecated

C++17 Implementation: libstdc++ PSTL



- **Library**, GNU's C++ standard library impl.
 - Ships with GCC
- FOSS, GPLv3
- CPU only
 - C++17 parallel execution policy via Intel TBB
 - Implementation contributed by Intel

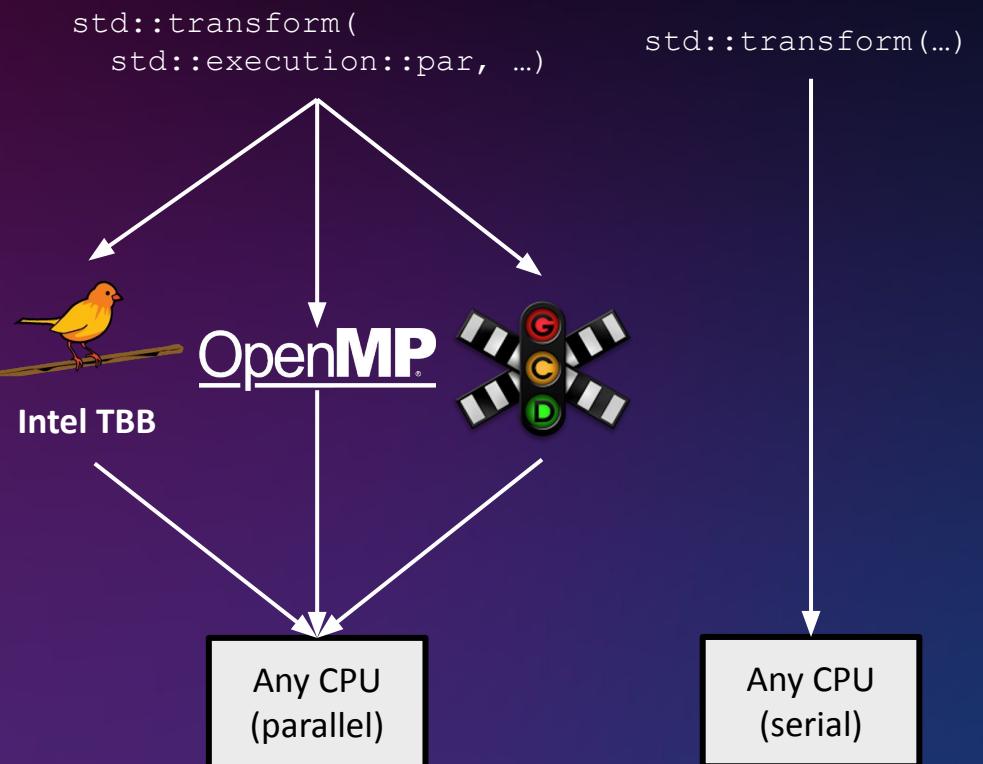
Intel Threading Building Block (TBB/oneTBB)
➤ General purpose concurrency primitive library



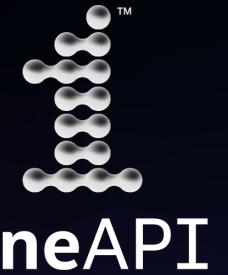
C++17 Implementation: LLVM PSTL



- **Library**, part of the LLVM project
- Planned integration with libc++, LLVM's C++ std library
 - Less complete than libstdc++
 - Not frequently shipped with Clang
- FOSS, UIUC or Apache 2.0 w/ LLVM exception
- CPU only, multiple C++17 parallel execution policy backends
 - Intel TBB (Intel contribution)
 - OpenMP taskloops
 - macOS GCD
 - OpenMP not supported in AppleClang



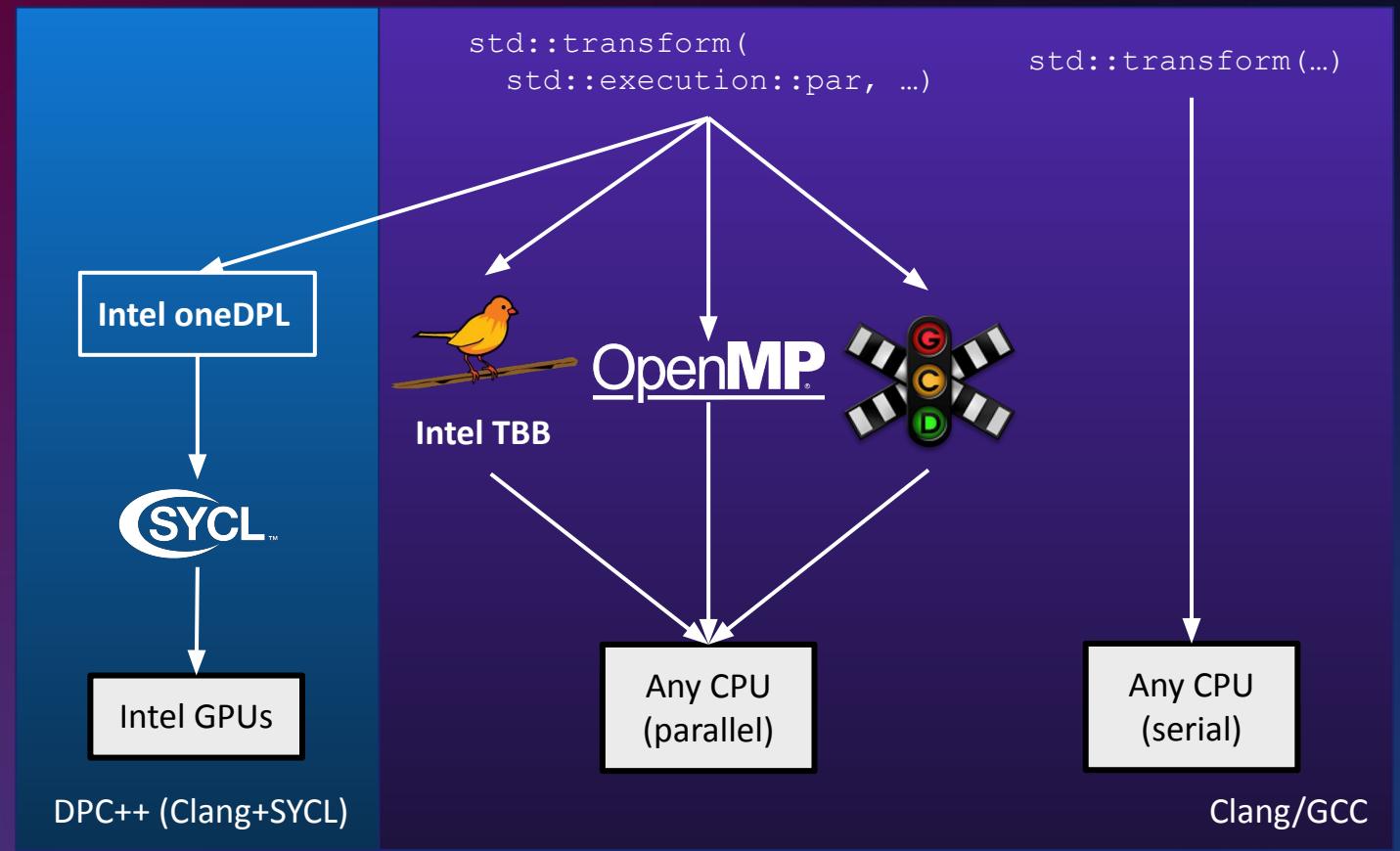
C++17 Implementation: Intel oneDPL



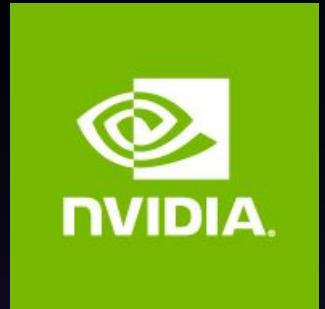
- Fork of the LLVM PSTD library
- FOSS, Apache 2.0 w/ LLVM exception
- CPU support untouched
- GPU offload implemented in **SYCL2020**
 - Kernels must adhere to SYCL2020 constraints
 - Validated on DPC++ compiler

SYCL2020

- Single source accelerator dialect of C++
- Spiritual successor of OpenCL
- 2020 revision adds essential features
 - USM
 - Reduction



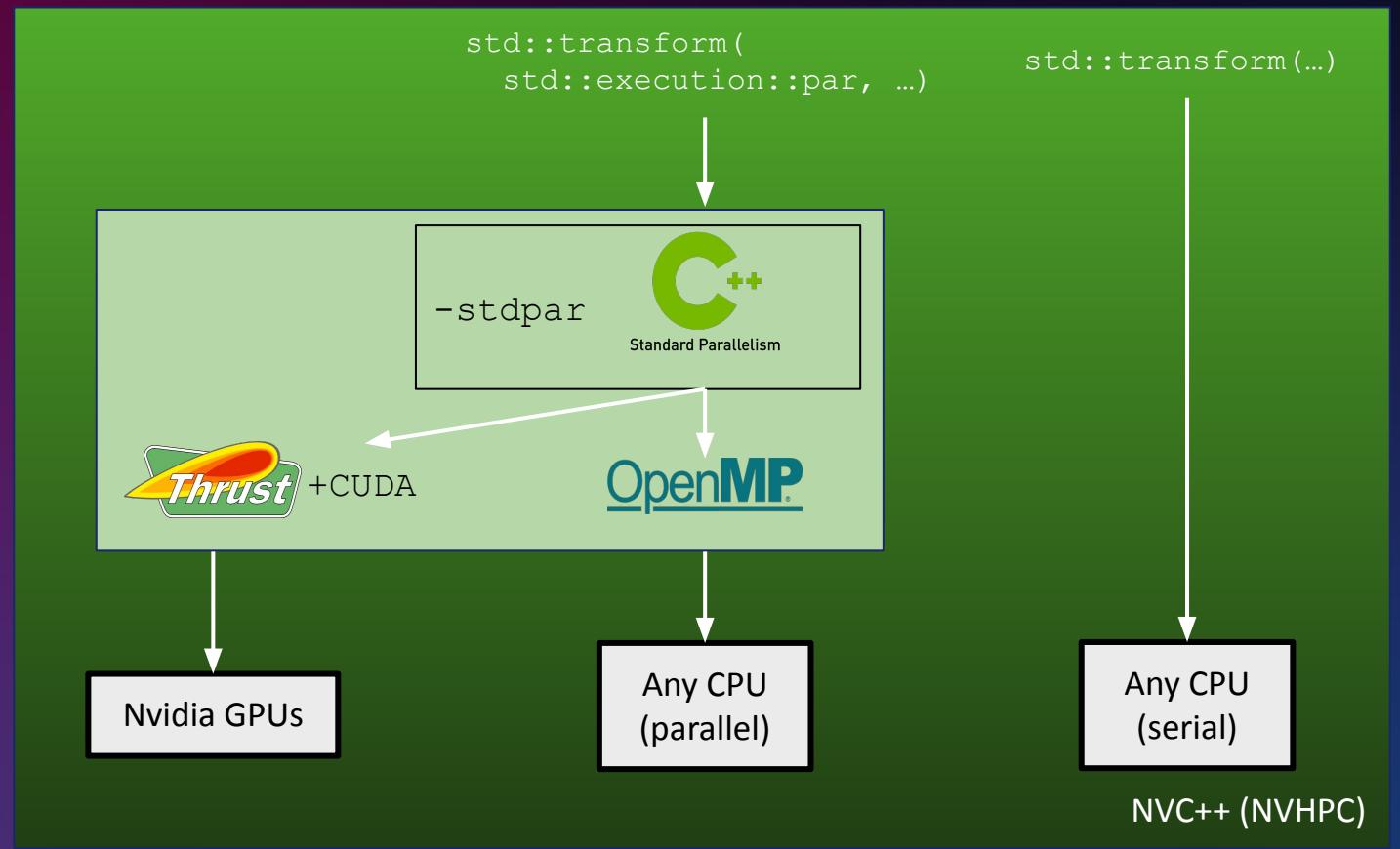
C++17 Implementation: NVIDIA NVHPC



- NVIDIA's heterogeneous **compiler**
- Closed source
- Formerly PGI Compilers
- Fully integrated SDK, compiler handles everything transparently

NVIDIA Thrust + CUDA

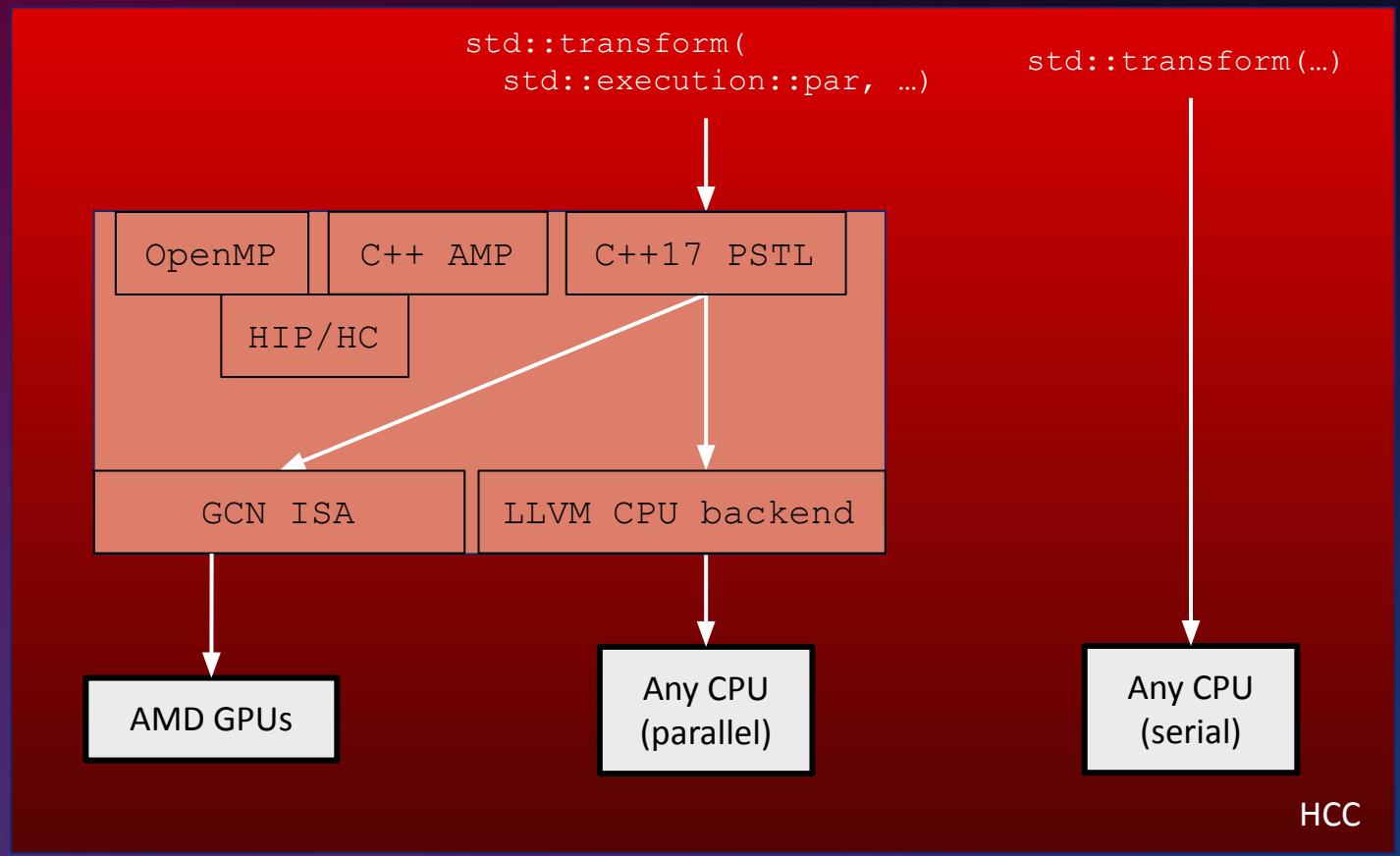
- CUDA: NVIDIA's accelerator C++ dialect
- Thrust: Productivity abstraction layer that implements STL
 - Multiple backends, including CUDA



C++17 Implementation: AMD HCC

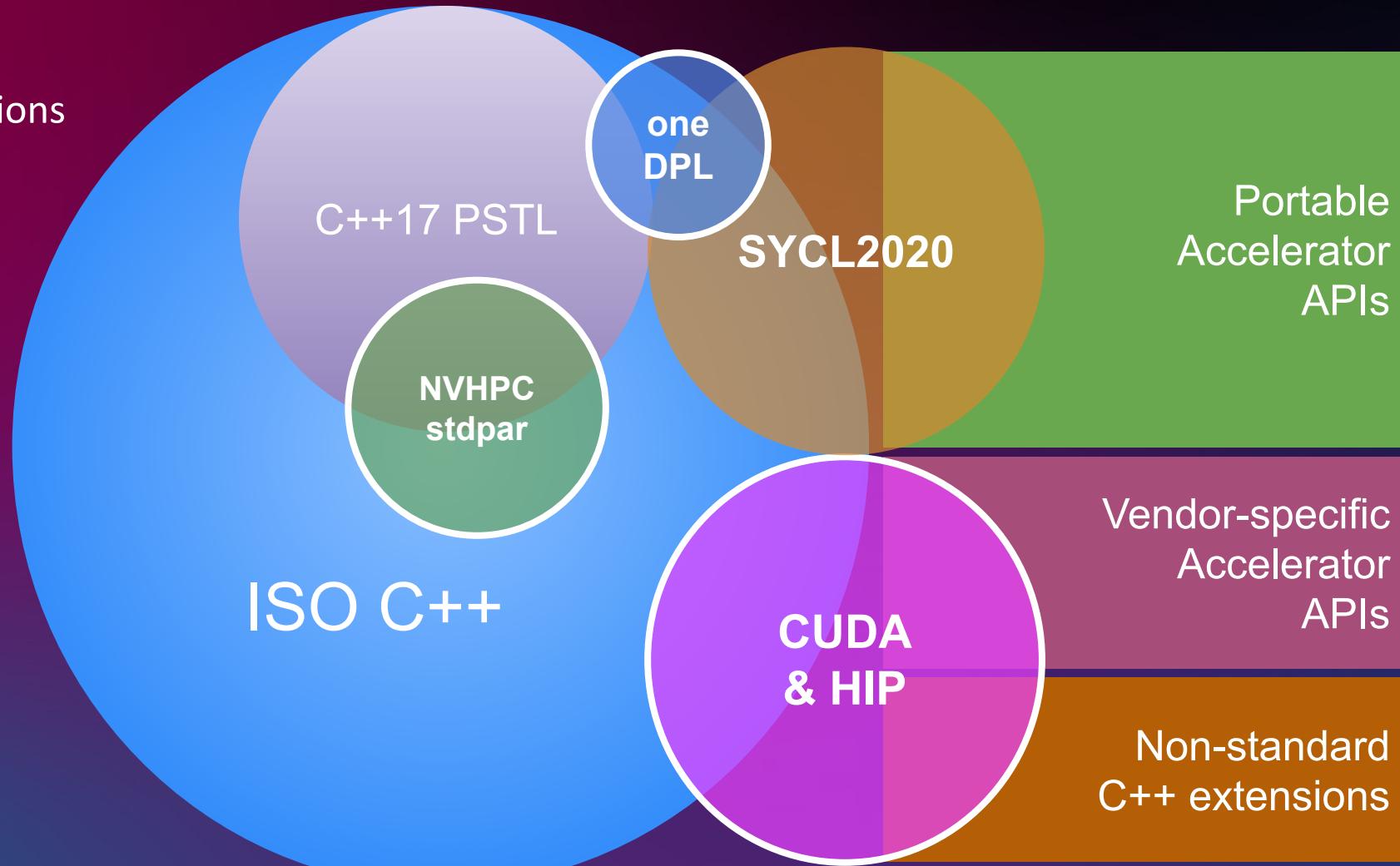


- AMD's heterogeneous **compiler**
- FOSS, UIUC License
- **Deprecated**, unmaintained since 2019
 - Focus shifted to HIP and feature parity with NVIDIA's software stack
- Fully integrated SDK, compiler handles everything transparently



Expressing Parallelism

- Implementations
- Standards
- Feature sets



Shim Header

```
#ifdef USE_ONEDPL // oneDPL C++17 PSL
#include <oneapi/dpl/algorithm>
#include <oneapi/dpl/execution>
#if ONEDPL_USE_DPCPP_BACKEND
    #include <CL/sycl.hpp>
namespace shim {
const auto par_unseq = oneapi::dpl::execution::device_policy<>{oneapi::dpl::execution::make_device_policy(cl::sycl::default_selector{})};
template <typename T> T *alloc_raw(size_t size) { return sycl::malloc_shared<T>(size, par_unseq.queue()); }
template <typename T> void deallocate_raw(T *ptr) { sycl::free(ptr, par_unseq.queue()); }
} // namespace shim
#else
namespace shim {
static constexpr auto par_unseq = dpl::execution::par_unseq;
}
#define USE_STD_PTR_ALLOC_DEALLOC
#endif
#else // Normal C++17 PSL
#include <algorithm>
#include <execution>
namespace shim {
static constexpr auto par_unseq = std::execution::par_unseq;
}
#define USE_STD_PTR_ALLOC_DEALLOC
#endif
#ifndef USE_STD_PTR_ALLOC_DEALLOC
namespace shim {
template <typename T> T *alloc_raw(size_t size) { return reinterpret_cast<T *>(std::malloc(sizeof(T) * size)); }
template <typename T> void deallocate_raw(T *ptr) { std::free(ptr); }
} // namespace shim
#endif
```



Experiments

Port representative HPC mini-apps to C++17 and compare performance

- Existing mini-app must already have implementation in multiple models
- Characteristics
 - Memory-bandwidth bound
 - Compute bound
 - Complex, multi-kernel application
- Comprehensive platform
 - x86 and AArch64 CPUs
 - All supported GPUs

Experiments

AArch64 and x86 CPUs

- Kokkos (OpenMP backend; Clang+GCC+NVHPC)
- OpenMP (Clang+GCC+NVHPC)
- C++17 (NVHPC)
- C++17 (TBB backend; Clang+GCC)
- C++17 (oneDPL OpenMP backend; Clang+GCC)

Nvidia GPUs

- Kokkos (CUDA backend, NVHPC)
- OpenMP Target (NVHPC)
- CUDA (NVHPC)
- C++17 (NVHPC)

Intel GPUs

- Kokkos (SYCL backend, DPC++)
- OpenMP Target (ICPX)
- SYCL2020 (DPC++)
- C++17 (oneDPL SYCL; DPC++)

Vendor	Name	Architecture	Abbreviation	Device Type	Total NUMA nodes
Intel	Xeon Gold 6338	x86, Ice Lake	Xeon	HPC CPU (32C*2)	2 (1 per socket)
AMD	EPYC 7713	x86, Zen3 (Milan)	EPYC	HPC CPU (64C*2)	8 (4 per socket)
AWS	Graviton 2	AArch64, Neoverse N1	Graviton2	HPC CPU (64C*1)	1
AWS	Graviton 3	AArch64, Neoverse V1	Graviton3	HPC CPU (64C*1)	1
NVIDIA	Tesla A100 (SXM 40GB)	Ampere	A100	HPC GPU	N/A
NVIDIA	Tesla V100 (PCIe 16GB)	Volta	V100	HPC GPU	N/A
Intel	UHD P630 (Xeon E2176G)	Gen9.5	UHD	Server iGPU	N/A
Intel	IrisPro 580 (i7 6670HQ)	Gen9	IrisPro	Consumer iGPU	N/A



Mini-app: BabelStream



- Memory-bandwidth bound
- Source code available on GitHub
 - <https://github.com/UoB-HPC/BabelStream>
- Port of the McCalpin STREAM benchmark to paradigms
 - Abstraction libraries:
 - Kokkos, RAJA
 - C/C++ dialects
 - SYCL, OpenCL, CUDA, HIP
 - C/C++ directives
 - OpenMP, OpenMP target, OpenACC
 - Libraries
 - Intel TBB, NVIDIA Thrust,
 - Languages
 - ISO C++17 (data&index), Rust, Julia, Scala, Java

Algorithm 1 BabelStream kernels

```
1: procedure COPY( $A[n]$ ,  $C[n]$ ,  $n$ )
2:   for  $i \leftarrow 0$  to  $n$  do  $C[i] \leftarrow A[i]$ 
3: procedure MUL( $A[n]$ ,  $B[n]$ ,  $C[n]$ , scalar,  $n$ )
4:   for  $i \leftarrow 0$  to  $n$  do  $B[i] \leftarrow \text{scalar} * C[i]$ 
5: procedure ADD( $A[n]$ ,  $B[n]$ ,  $C[n]$ ,  $n$ )
6:   for  $i \leftarrow 0$  to  $n$  do  $C[i] \leftarrow A[i] + B[i]$ 
7: procedure TRIAD( $A[n]$ ,  $B[n]$ ,  $C[n]$ , scalar,  $n$ )
8:   for  $i \leftarrow 0$  to  $n$  do  $A[i] \leftarrow B[i] + (\text{scalar} * C[i])$ 
9: procedure DOT( $A[n]$ ,  $B[n]$ , scalar,  $n$ )
10:  for  $i \leftarrow 0$  to  $n$  do  $\text{sum} \leftarrow \text{sum} + (A[i] * B[i])$ 
    return  $\text{sum}$ 
```

Porting: BabelStream

```
// OpenMP
#pragma omp parallel for
for (int i = 0; i < array_size; i++)
    c[i] = a[i] + b[i];

// Data centric
std::transform(exe_policy, a, a + array_size, b, c, [](auto l, auto r) {
    return l + r;
});

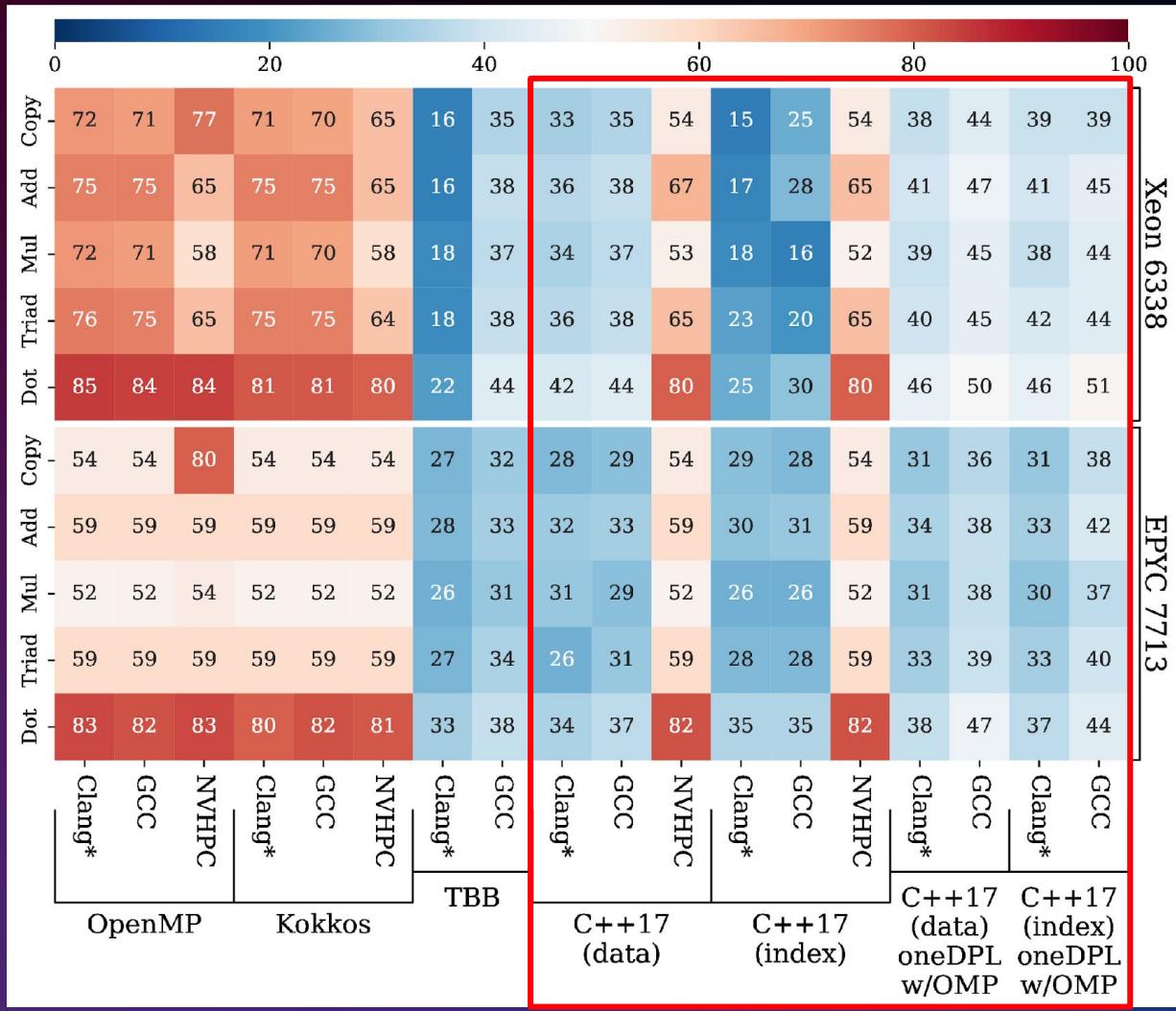
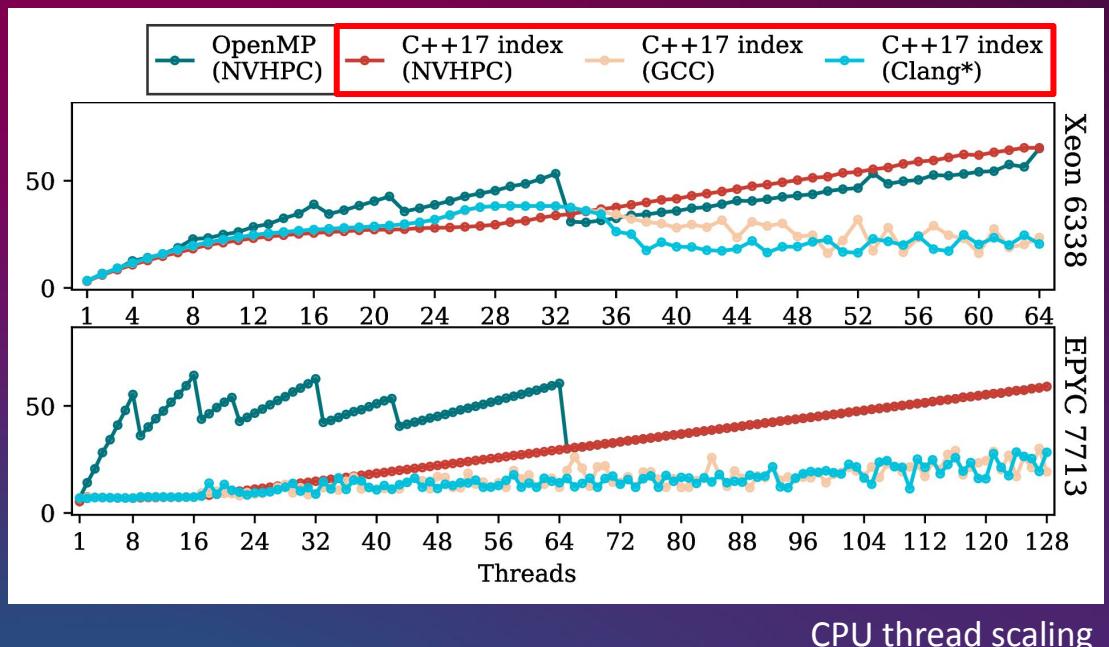
// Index centric
ranged<int> range(0, array_size);
std::transform(exe_policy, range.begin(), range.end(), c, [a, b](int i)
{
    return a[i] + b[i];
});
```



Results: BabelStream x86 CPU

% of peak memory bandwidth per platform; higher is better

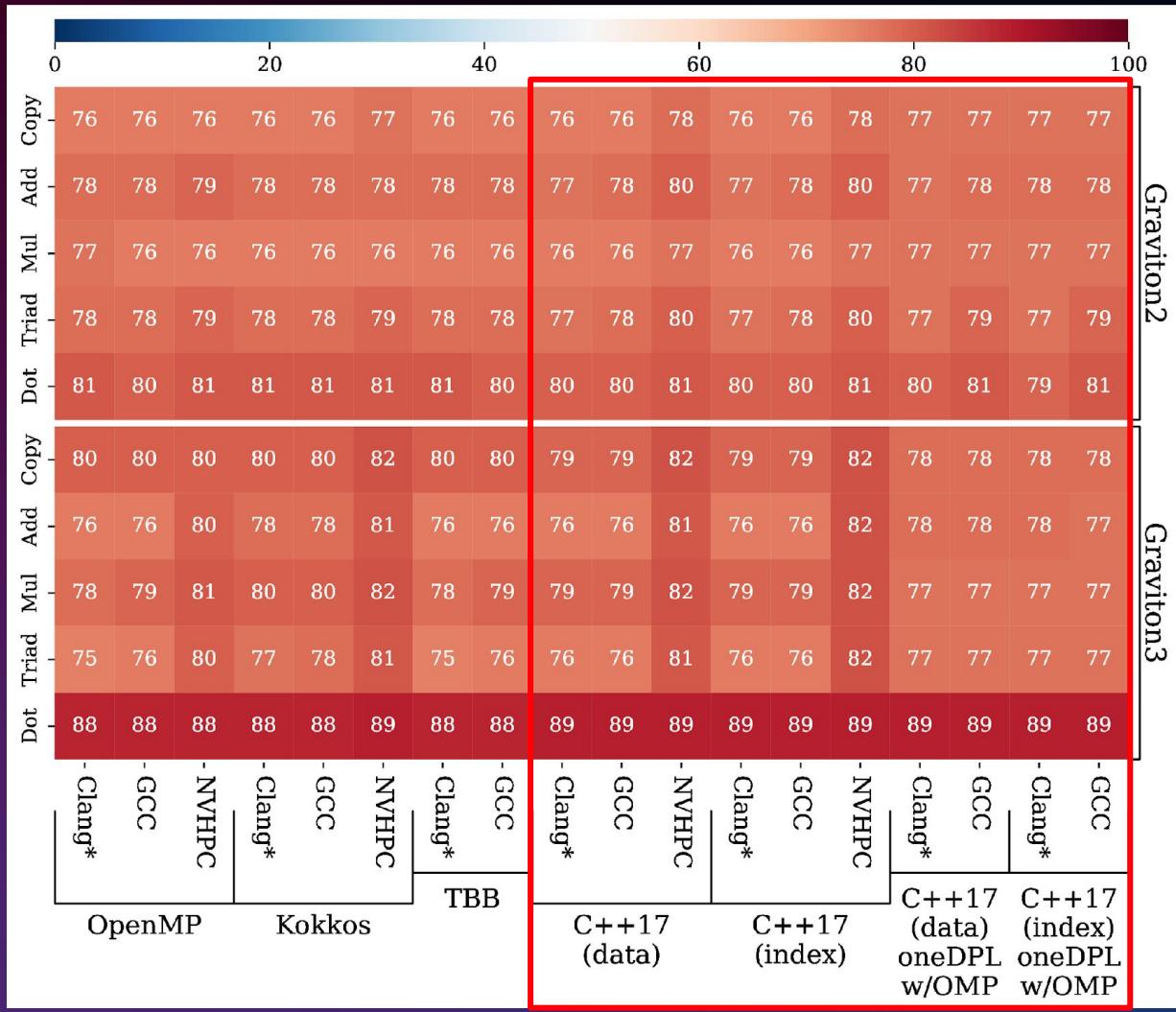
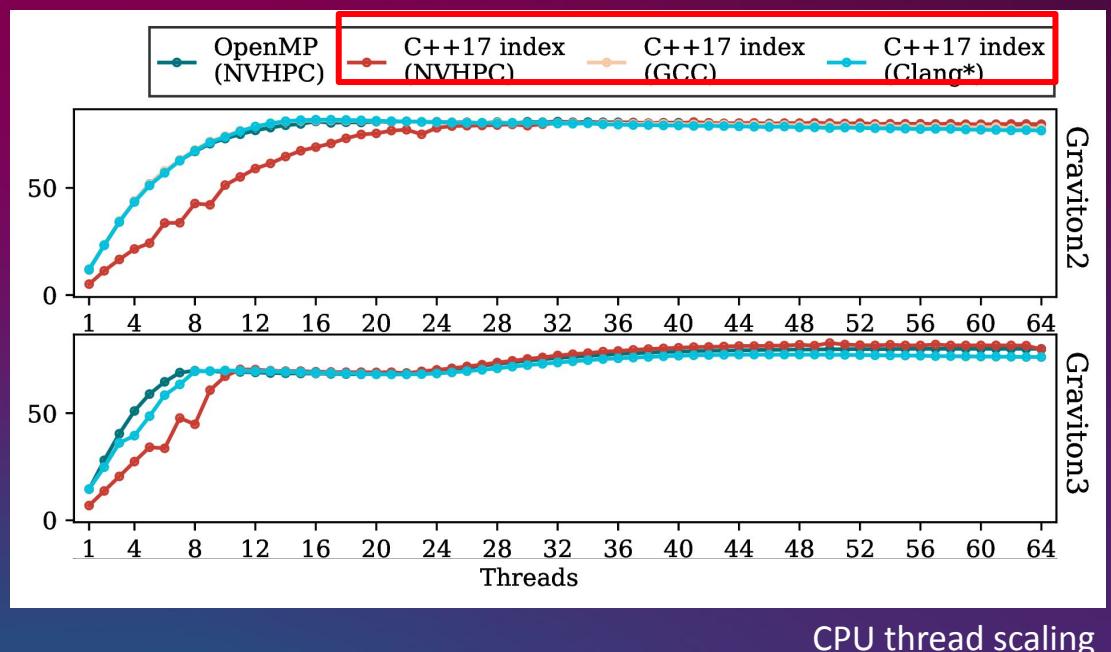
- Explicit NUMA-awareness
 - OpenMP, C++17 on OpenMP
- Bad chunking + OpenMP taskloop
 - oneDPL



Results: BabelStream AArch64 CPU

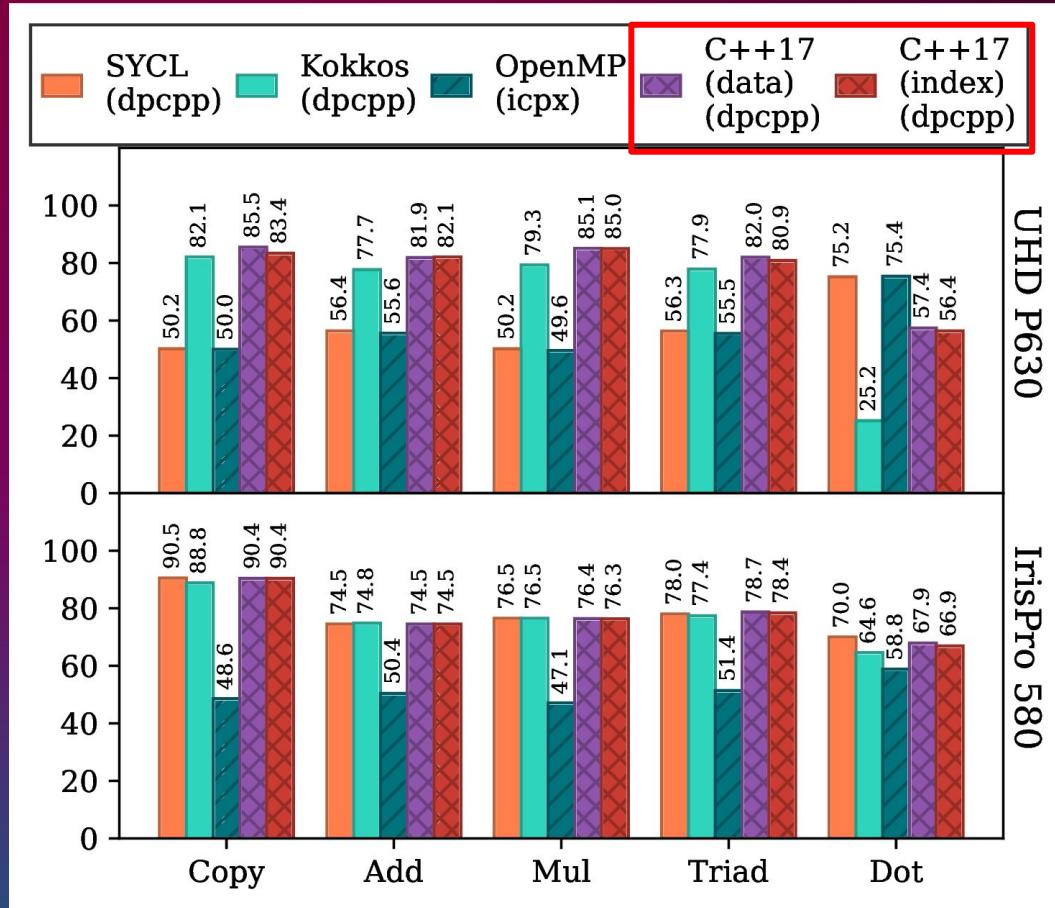
% of peak memory bandwidth per platform; higher is better

- Graviton nodes lack NUMA regions
 - Good performance overall, mature compilers

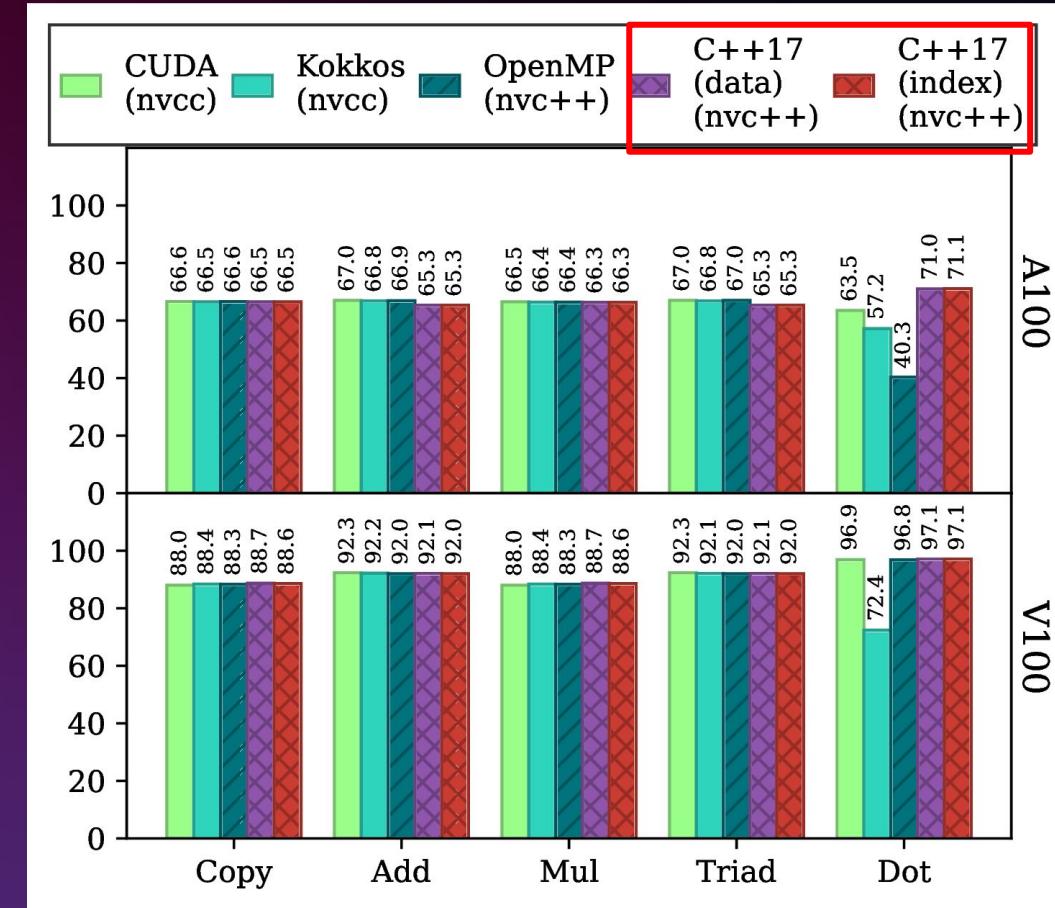


Results: BabelStream GPUs

% of peak memory bandwidth per platform; higher is better



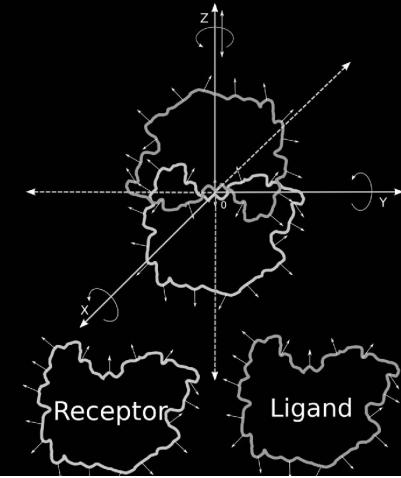
Intel GPUs



NVIDIA GPUs

Mini-app: miniBUDE

- Proxy application of the Bristol University Docking Engine (BUDE)
- Source code available on GitHub
 - <https://github.com/UoB-HPC/miniBUDE>
- Compute bound, measurements in GFLOP/s or total runtime
 - Abstraction libraries:
 - Kokkos, RAJA
 - C/C++ dialects
 - SYCL, OpenCL, CUDA, HIP
 - C/C++ directives
 - OpenMP, OpenMP target, OpenACC
 - Libraries
 - Intel TBB, NVIDIA Thrust,
 - Languages
 - ISO C++17 (Index), Julia



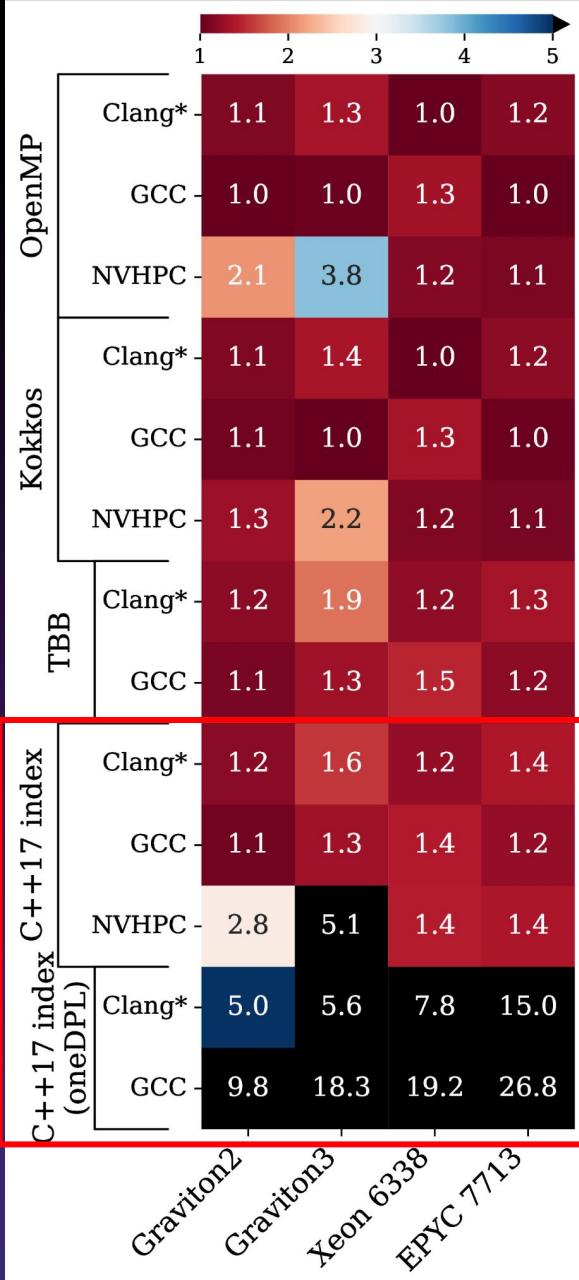
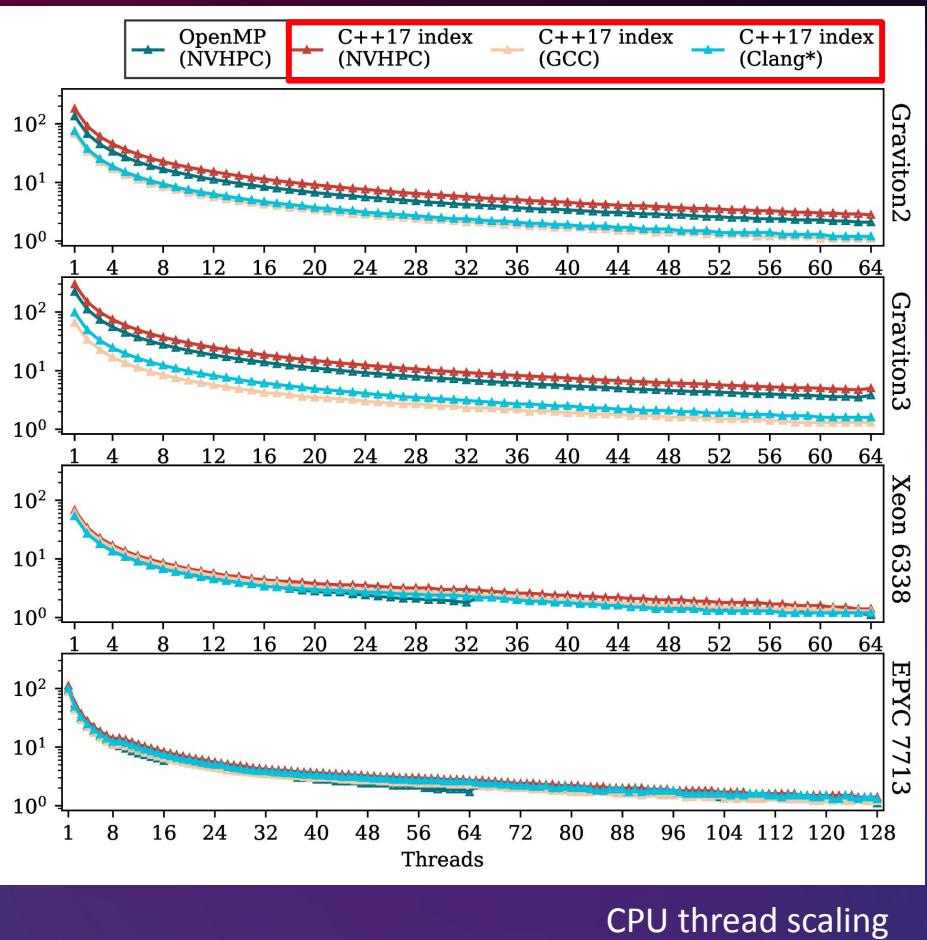
Algorithm 2 miniBUDE Fasten Kernel

```
1: procedure FASTEN( const  $i$ , const  $xform_{3 \times 3}$ [],  
    const  $proteins[ps]$ , const  $ligands[ls]$ , out  $energy[]$ )  
     $\triangleright$  Values  $R$ ,  $DSLV$ ,  $DSLV_R$ ,  $NZ$ ,  $DST_1$ ,  $DST$ ,  $HRD$ ,  $T$   
    are part of the simulation constants  
2:   for  $il \leftarrow 0, ls$  do  
3:      $lpos_{1 \times 3} \leftarrow xform \cdot ligands[il].pos_{1 \times 3}$   
4:     for  $ip \leftarrow 0, ps$  do  
         $\triangleright$  Atom distance and sphere radii sum  
5:            $dist \leftarrow distance(lpos, proteins[ip].pos_{1 \times 3})$   
6:            $d \leftarrow dist - R$   
         $\triangleright$  Steric energy, formal/dipole charge interactions  
7:            $energy[i] \leftarrow energy[i] +$   
            $(1 - dist * (1/R)) * (d < 0?2 * HRD : 0)$   
8:            $e \leftarrow init *$   
            $(d < 0.f?1 : (1 - d * DST_1)) *$   
            $(d < DST?1 : 0)$   
9:            $energy[i] \leftarrow energy[i] + (typeE? - |e| : e) * T$   
         $\triangleright$  Nonpolar-Polar repulsive interactions  
10:           $dslvE = dslvInit *$   
             $((d < DSLV \wedge NZ)?1 : 0.f) *$   
             $(d < 0?1 : (1 - d * DSLV_R))$   
11:           $energy[i] \leftarrow energy[i] + dslvE * 0.5$ 
```

Results: miniBUDE CPUs

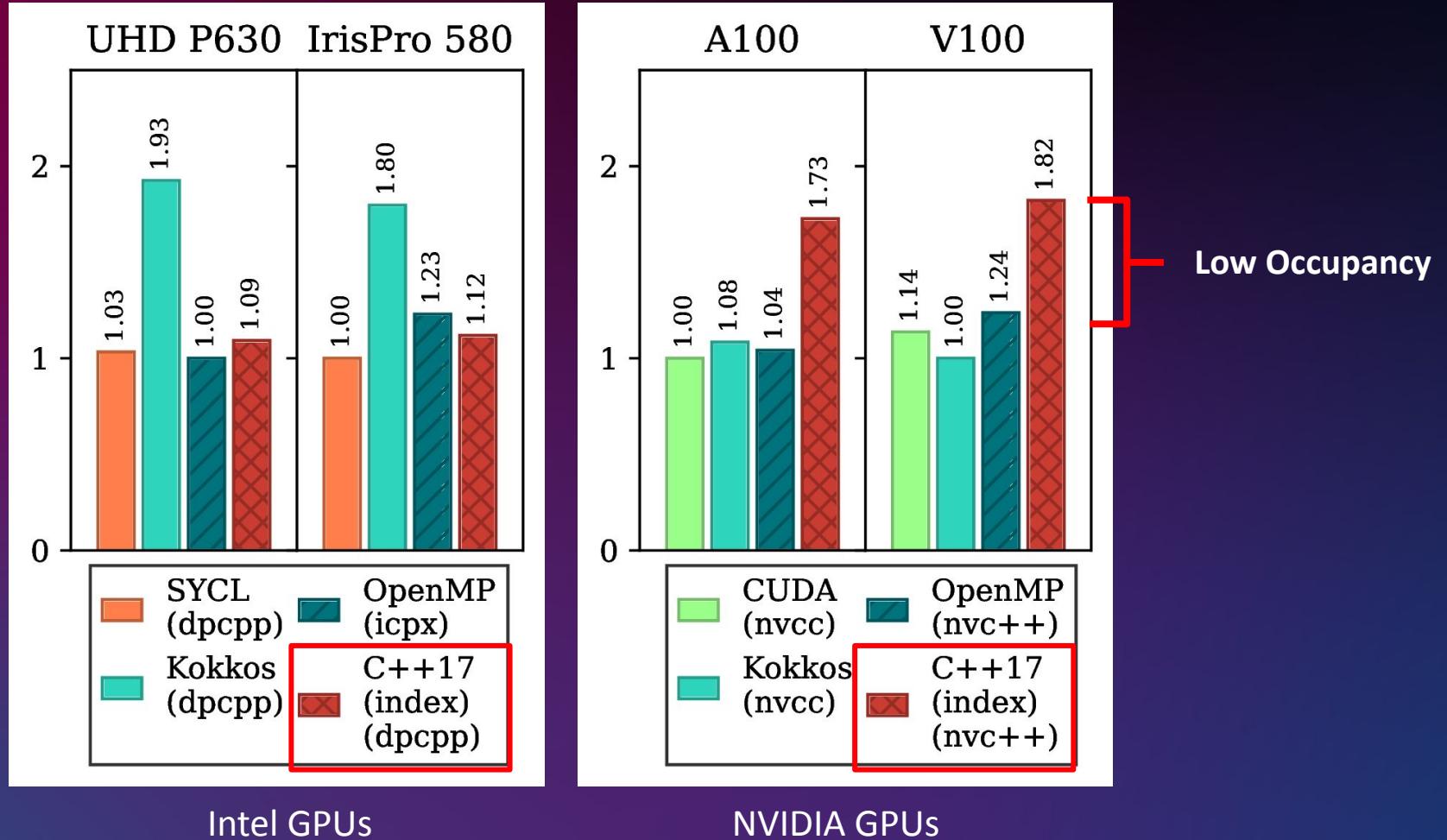
- Unaffected by NUMA
- NVHPC unoptimised for ARM
 - ARM support in NVHPC is new
- Same chunking+OpenMP taskloop issue on oneDPL

Results as normalised runtime per platform; lower is better



Results: miniBUDE GPUs

Results as normalised runtime per platform; lower is better





Mini-app: CloverLeaf

- Proxy application for 2D hydrodynamics
- Source code available on GitHub
 - https://github.com/UoB-HPC/cloverleaf_stdpar
- Mixed memory-bandwidth bound; measurements in total runtime
 - Structured grid; stencil access pattern
 - Reductions
- Complex application, >100 unique kernels + MPI halo exchange
 - Abstraction libraries:
 - **Kokkos**
 - C/C++ dialects
 - **SYCL**, OpenCL, **CUDA**
 - C/C++ directives
 - **OpenMP**, OpenMP target
 - Libraries
 - **Intel TBB**:
 - Languages
 - **ISO C++17 (Index)**:

Algorithm 3 High-level CloverLeaf kernel overview

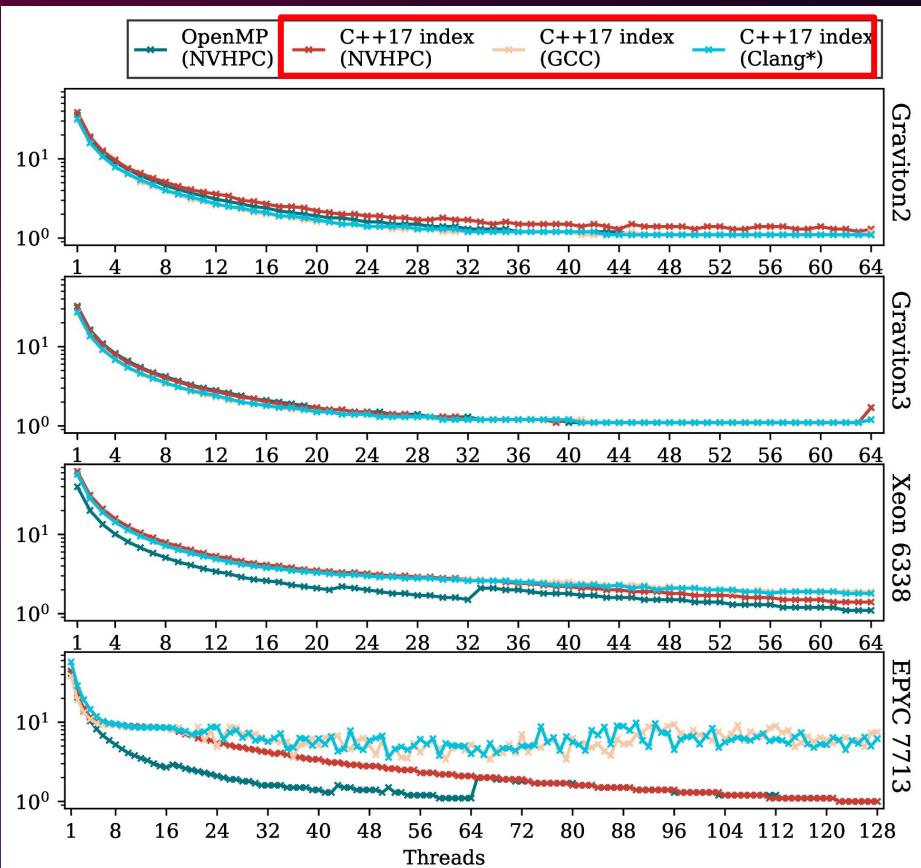
```
▷ Each procedure traverses the full  $W \times H$  grid, some procedures may
▷ invoke multiple kernels
1: while  $step < maxStep$  do
2:   procedure IDEAL_GAS
    ▷ Pressure/sound speed via ideal gas equation of state with a fixed gamma
3:   procedure VISCOSITY
    ▷ Artificial viscosity via the Wilkin's method to smooth out shock front
    and prevent oscillations
4:   procedure PDV
    ▷ Cell energy/density  $\delta$  via velocity gradients
5:   procedure CALC_DT
    ▷ Compute the minimum timestep based on CFL conditions, velocity
    gradient, and velocity divergence.
6:   procedure ACCELERATE
    ▷ Update velocity field via cell pressure/viscosity gradients
7:   procedure FLUX_CALC
    ▷ Edge volume fluxes using the velocity fields
8:   procedure ADVECTION
    ▷ Setup fields for the next iteration
9:   procedure RESET_FIELD
    ▷ Edge volume fluxes based on the velocity fields
10:  procedure FIELD_SUMMARY
    ▷ Total mass, internal energy, kinetic energy, and volume weighted
    pressure
11:   $step \leftarrow step + 1$ 
```



Results: CloverLeaf CPUs

- Strong NUMA effects
 - Similar to BabelStream
- Degradation proportional to NUMA domain count

Results as normalised runtime per platform; lower is better



CPU thread scaling

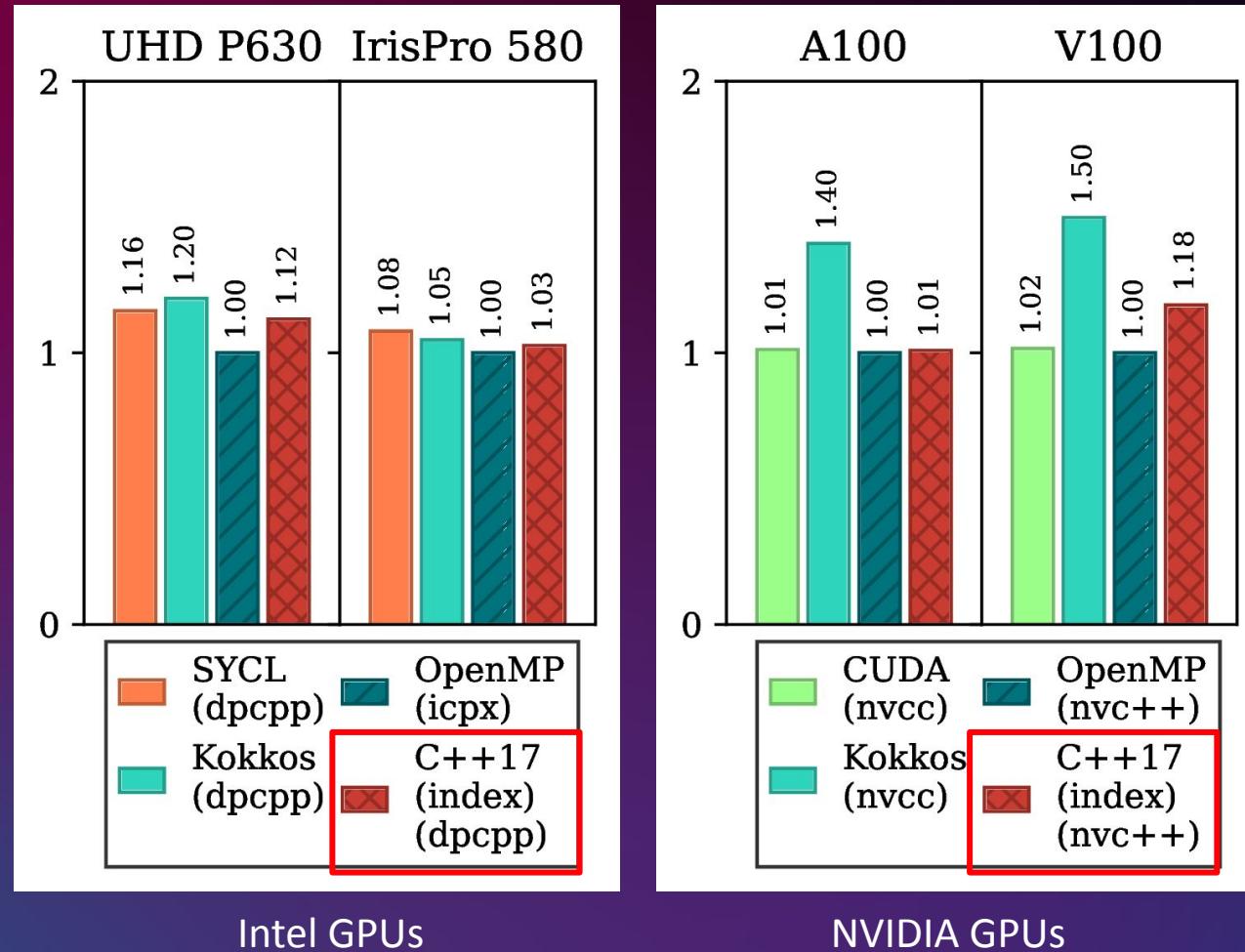
		1	2	3	4	5
		Clang*	1.2	1.0	1.3	1.0
OpenMP		GCC	1.0	1.0	1.0	1.0
NVHPC	Clang*	1.1	1.1	1.2	1.0	
	GCC	1.1	1.1	1.1	1.1	
Kokkos	Clang*	1.0	1.1	1.1	1.1	
	GCC	1.1	1.0	1.1	1.2	
NVHPC	Clang*	1.1	1.1	1.1	1.1	
	GCC	1.1	1.1	1.1	1.1	
TBB	Clang*	1.8	1.5	1.9	9.0	
	GCC	1.8	1.5	4.4	9.6	
C++17 index (oneDPL)	Clang*	1.1	1.1	3.9	5.9	
	GCC	1.1	1.1	1.9	6.6	
NVHPC	Clang*	1.3	1.1	1.5	1.0	
	GCC	1.3	1.1	1.9	4.7	
Graviton2	Clang*	1.4	1.2	1.8	3.7	
	GCC	1.4	1.2	1.8	3.7	

Relative runtime



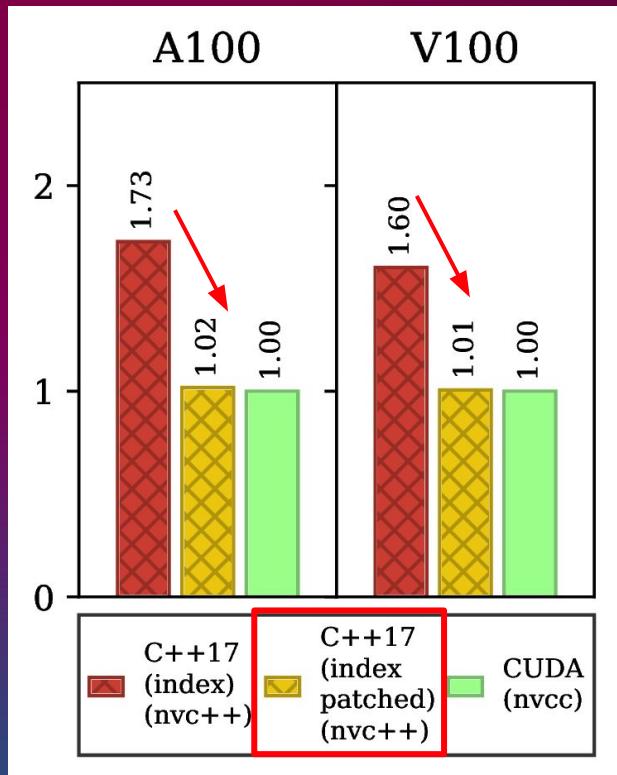
Results: CloverLeaf GPUs

Results as normalised runtime per platform; lower is better

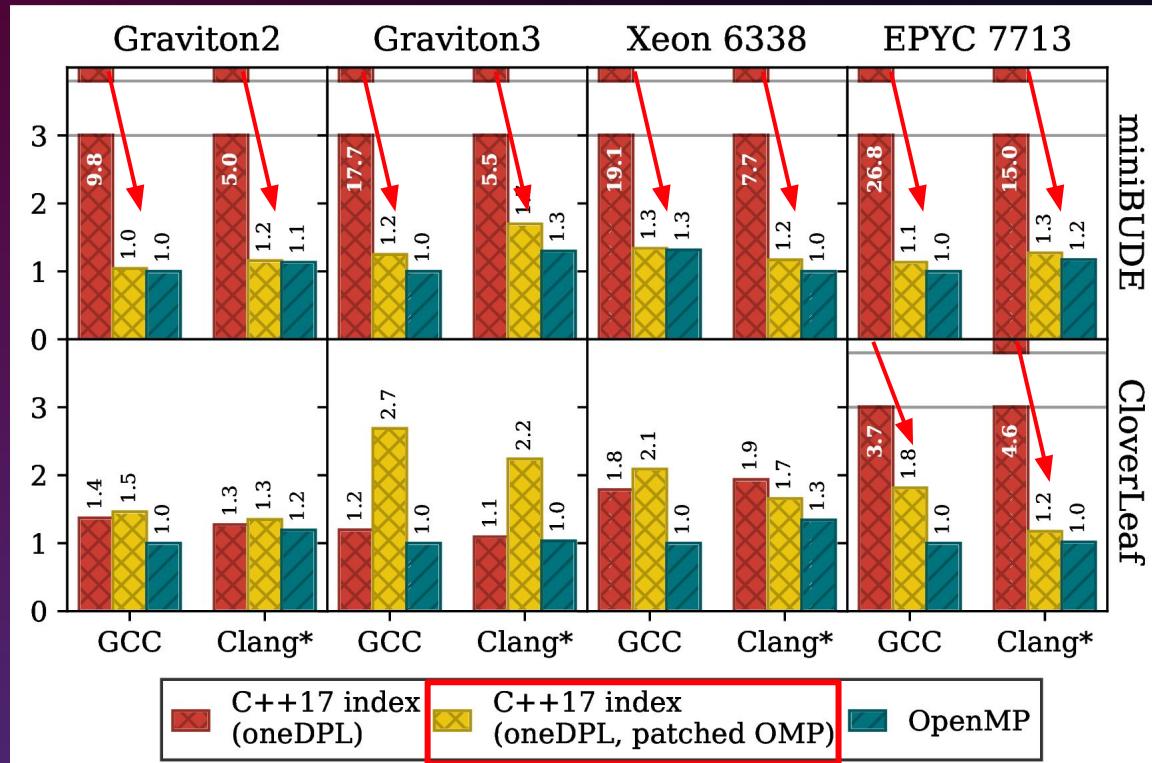


Performance issues

miniBUDE Thrust header patch uplift



oneDPL OpenMP chunk patch uplift



Conclusion

- Three ISO C++ ports of representative mini-apps
 - In high productivity, portable, idiomatic C++
 - Reductions
 - Transformations
 - Traditional index-central traversal works
 - Stencil
 - Multiple allocations in kernel
- Performance portable
 - Directly comparable to established models in most cases
- Missing pieces
 - Device control - no API
 - Async dispatch - senders/receivers coming soon
 - \geq C++20 features - Coming soon to NVHPC and GCC/Clang



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