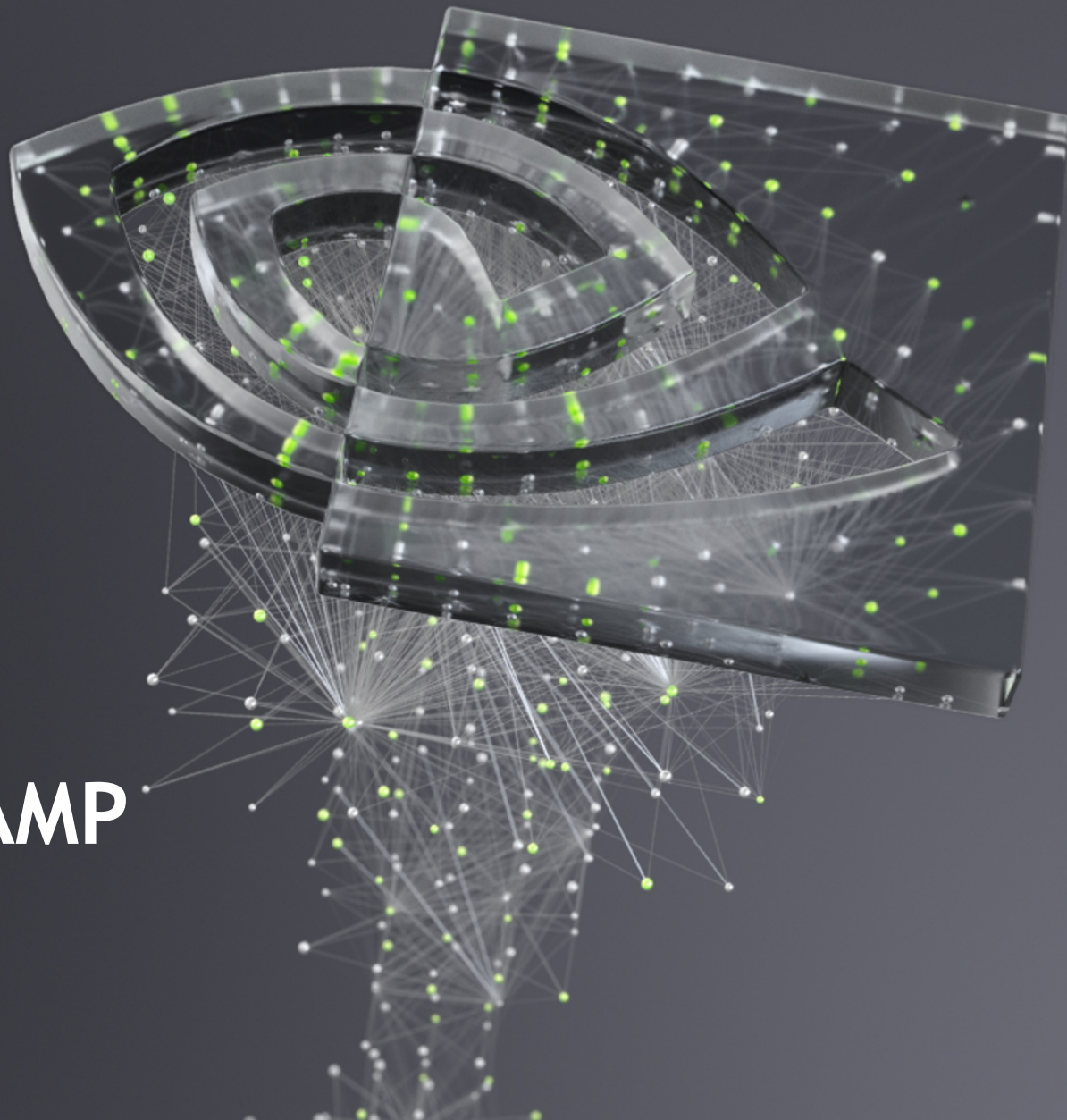




# N-WAYS GPU BOOTCAMP

## STANDARD LANGUAGES



# STANDARD LANGUAGES

## What to expect?

- C++ , Fortran ISO standard brief
- C++ `std::par` , Fortran DO-Concurrent API
- Known limitations

# BRIEF HISTORY

- Historically, accelerating your code with GPUs has not been possible in Standard C++/Fortran without using language extensions or additional libraries:
  - CUDA C++ requires the use of `__host__` and `__device__` attributes on functions and the `<<<>>` syntax for GPU kernel launches.
  - OpenACC uses `#pragmas` to control GPU acceleration
- What if you could take your Standard C++ or Fortran code and accelerate on a GPU?

# QUICK BACKGROUND

## C++ STL Containers

- One driving feature of C++ are its templates and the STL library. C++11 is further pushing these ideas and shows no sign of slowing.
- C++ templates are probably most widely used through the STL containers.
  - `std::vector`, `std::string`, `std::map`, `std::list`, etc...
- Besides the OO features and convenience, these containers are designed to rise-above basic C pointers, providing more safety from memory violations, while maintaining the bare-metal performance.
- For example `std::vector` □ The vector template is designed to replace C's arrays.

```
std::vector<int> my_ints(4, 100); // four ints with value 100
```

# STD::PAR

## What is std::par?

- Use standard C++ constructs to make code run parallel on heterogeneous hardware
- C++11 introduced a memory model, concurrent execution model, and concurrency library, providing a standard way to take advantage of multicore processors
- The C++17 Standard introduced higher-level parallelism features that allow users to request parallelization of Standard Library algorithms.

## Advantage:

- No language extensions, pragmas, directives, or non-standard libraries
- Write Standard C++, which is portable to other compilers and systems
- Compiler automatically accelerates code with high-performance NVIDIA GPUs and hence less time porting and more time on what really matters

# STD::PAR

## Parallelism in Standard C++

- Parallelism is expressed by adding an execution policy as the first parameter to any algorithm that supports execution policies
- Most of the existing Standard C++ algorithms were enhanced to support execution policies

Execution policies can be applied to most standard algorithms

- `std::execution::seq` = sequential: Sequential execution. No parallelism is allowed.
- `std::execution::par` = **parallel**: Parallel execution on one or more threads.
- `std::execution::par_unseq` = parallel + vectorized: Parallel execution on one or more threads, with each thread possibly vectorized.



# C++17 PARALLEL ALGORITHMS

## Example

C++98: `std::sort(c.begin(), c.end());`

C++17: `std::sort(std::execution::par, c.begin(), c.end());`



**BUILD AND RUN THE CODE**



# NVIDIA HPC SDK

- Comprehensive suite of compilers, libraries, and tools used to GPU accelerate HPC modeling and simulation application
- The NVIDIA HPC SDK includes the new NVIDIA HPC C++ compiler, NVC++. NVC++ supports C++17, C++ Standard Parallelism (stdpar) for CPU and GPU
- NVC++ can compile Standard C++ algorithms with the parallel execution policies `std::execution::par` execution on NVIDIA GPUs.
- An NVC++ command-line option, `-stdpar`, is used to enable GPU-accelerated C++ Parallel Algorithms

```
nvc++ -stdpar program.cpp -o program
```

# RDF

## Pseudo Code

```
for (int frame=0;frame<nconf;frame++){  
    for(int id1=0;id1<numatm;id1++){  
        for(int id2=0;id2<numatm;id2++){  
            {  
                dx=d_x[id1]-d_x[id2];  
                dy=d_y[id1]-d_y[id2];  
                dz=d_z[id1]-d_z[id2];  
  
                r=sqrtf(dx*dx+dy*dy+dz*dz);  
  
                if (r<cut) {  
                    ig2=(int)(r/del);  
                    d_g2[ig2] = d_g2[ig2] +1 ;  
                }  
            }  
        }  
    }  
}
```

- Across Frames
- Find Distance
- Reduction

# STEPS

## Step 1: Replace for with std::for\_each

```
std::for_each (InputIterator first, InputIterator last, Function fn)
```

`start_iter` : The beginning position from where function operations has to be executed.

`last_iter` : This ending position till where function has to be executed.

`fnc/obj_fnc` : The 3rd argument is a function or an object function which operation would be applied to each element.

# STEPS

Step 2: Pass execution policy as `std::execution::par`

```
for_each (std::execution::par , InputIterator first, InputIterator last, Function fn)
```

Execution policy as the first parameter will dictate to run the loop body in parallel across threads

# STEPS

## Step 3: Change indexing to use `counting::iterator`

```
std::for_each(std::execution::par,  
             thrust::counting_iterator<unsigned int>(0u), thrust::counting_iterator<unsigned int>(numatm*numatm)
```

```
std::vector<unsigned int> indices(numatm * numatm);  
std::generate(indices.begin(), indices.end(), [n = 0]() mutable { return n++; });
```

```
std::for_each(std::execution::par,  
             indices.begin(), indices.end(),
```

- Counting Iterator helps in filling up a vector with the numbers zero through N
- In our case from 0 to number of atoms
- GPU We will be using Thrust library for counting iterator for GPU
  - High-Level Parallel Algorithms Library
  - Parallel Analog of the C++ Standard Template Library (STL)

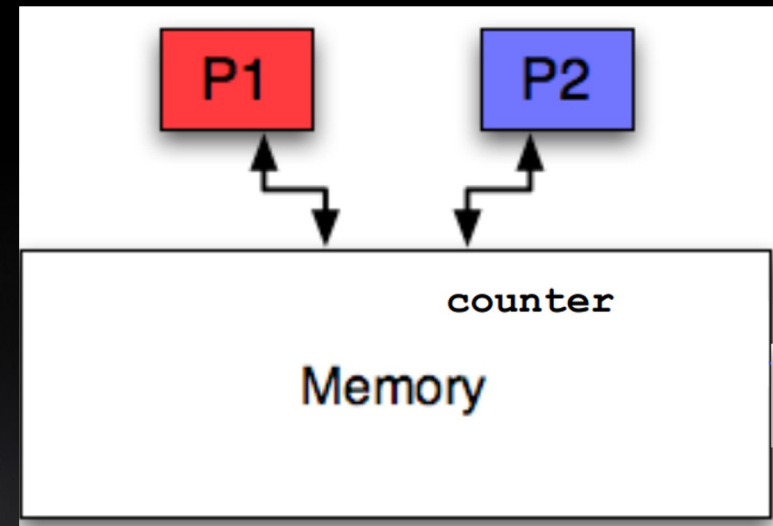


# ATOMIC

## Step 4: Remove Datarace

```
std::atomic<int>* h_g2 = new std::atomic<int>[nbin];
```

```
void *do_stuff(void * arg)
{
    for (int i = 0 ; i < 200000000 ; ++ i)
    { counter ++; }
    return arg;
}
```



Since the variable counter is shared, we can get a data race

# STEPS

## Step 5: Compile for Multicore and GPU

```
std::atomic<int>* h_g2 = new std::atomic<int>[nbin];

std::for_each(std::execution::par, thrust::counting_iterator<unsigned int>(0u),
              thrust::counting_iterator<unsigned int>(numatm*numatm),
              [...](unsigned int index)
              {
    for(int id2=0;id2<numatm;id2++)
    {
        dx=d_x[id2]-d_x[id2];
        dy=d_y[id2]-d_y[id2];
        dz=d_z[id2]-d_z[id2];
        r=sqrtf(dx*dx+dy*dy+dz*dz);

        if (r<cut) {
            ig2=(int)(r/del);
            ++d_g2[ig2];
        }
    }
}
}
```

- Atomic Declaration
- Counting Iterator
- Find Distance
- Atomic Increment

```
nvc++ -stdpar=gpu,multicore program.cpp -o program
```

# STD::PAR SPEEDUP



HPC SDK 20.11, NVIDIA Tesla V100, DGX1

# FORTRAN

## DO CONCURRENT :: ISO Standard Fortran

- ISO Standard Fortran 2008 introduced the **DO CONCURRENT** construct to allow you to express loop-level parallelism, one of the various mechanisms for expressing parallelism directly in the Fortran language
- HPC SDK 20.11 release of the NVIDIA HPC SDK, the included NVFORTRAN compiler automatically accelerates **DO CONCURRENT**

```
1 subroutine saxpy(x,y,n,a)
2   real :: a, x(:), y(:)
3   integer :: n, i
4   do i = 1, n
5     y(i) = a*x(i)+y(i)
6   enddo
7 end subroutine saxpy
```

```
1 subroutine saxpy(x,y,n,a)
2   real :: a, x(:), y(:)
3   integer :: n, i
4   do concurrent (i = 1: n)
5     y(i) = a*x(i)+y(i)
6   enddo
7 end subroutine saxp
```

```
nvfortran -stdpar=gpu,multicore program.f90 -o program
```

# FORTRAN

## Nested Loop Parallelism

- Nested loops are a common code pattern encountered in HPC applications
- It is straightforward to write such patterns with a single DO CONCURRENT statement, as in the following example

```
do i=2, n-1
  do j=2, m-1
    a(i,j) = w0 * b(i,j)
  enddo
enddo
```

```
do concurrent(i=2 : n-1, j=2 : m-1)
  a(i,j) = w0 * b(i,j)
enddo
```



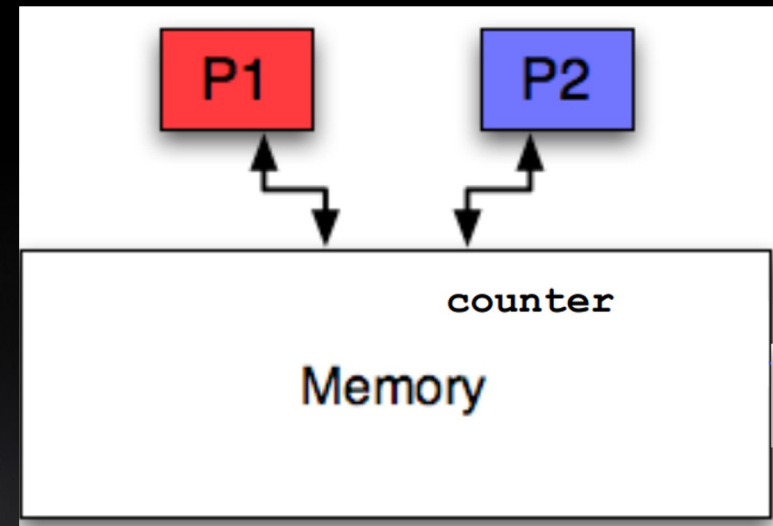
# ATOMIC

## Limitation

```
!$acc atomic  
g(ind)=g(ind)+1.0d0
```

```
void *do_stuff(void * arg)  
{  
    for (int i = 0 ; i < 200000000 ; ++ i)  
    { counter ++; }  
    return arg;  
}
```

- Do-Concurrent implementation of GPC SDK currently does not support Atomic constructs
- Hence we use the OpenACC Construct to solve data race



# STEPS

## Compile for Multicore and GPU

```
do iconf=1,nframes

  do concurrent(i=1 : natoms, j=1:natoms)
    dx=x(iconf,i)-x(iconf,j)
    dy=y(iconf,i)-y(iconf,j)
    dz=z(iconf,i)-z(iconf,j)

    ...
    r=dsqrt(dx**2+dy**2+dz**2)
    if(r<cut)then
      !$acc atomic
      g(ind)=g(ind)+1.0d0
    endif
  enddo
enddo
```

- Do Concurrent
- Find Distance
- Atomic Increment

```
nvfortran -stdpar=gpu,multicore program.f90 -o program
```

A network diagram consisting of numerous small circular nodes connected by thin, light-colored lines. The nodes are primarily white, with several highlighted in a bright green color. The connections form a complex web of lines, with some nodes having a higher degree of connectivity than others. The background is a dark, solid color, making the network structure stand out.

**WE WILL BE BACK AT 13: 00**

# REFERENCES

<https://developer.nvidia.com/blog/accelerating-fortran-do-concurrent-with-gpus-and-the-nvidia-hpc-sdk/>

<https://developer.nvidia.com/blog/accelerating-standard-c-with-gpus-using-stdpar/>

<https://developer.download.nvidia.com/video/gputechconf/gtc/2019/presentation/s9770-c++17-parallel-algorithms-for-nvidia-gpus-with-pgi-c++.pdf>



**THANK YOU**

