

# OpenACC is not an Island



- OpenACC allows very high level expression of parallelism and data movement.
- It's still possible to leverage low-level approaches such as CUDA C, CUDA Fortran, and GPU Libraries.



# Why Interoperate?



- Don't reinvent the wheel
  - Lots of CUDA code and libraries already exist and can be leveraged.
- Maximum Flexibility
  - Some things can just be represented more easily in one approach or another.
- Maximum Performance
  - Sometimes hand-tuning achieves the most performance.

# **CUDA C Primer**



### Standard C

### Parallel C

- Serial loop over 1M elements, executes 1M times sequentially.
- Data is resident on CPU.

- Parallel kernel, executes 1M times in parallel in groups of 256 elements.
- Data must be copied to/from GPU.

# **CUDA C Interoperability**



## OpenACC Main

## CUDA C Kernel & Wrapper

- It's possible to interoperate from C/C++ or Fortran.
- OpenACC manages the data and passes device pointers to CUDA.

- CUDA kernel launch wrapped in function expecting device arrays.
- Kernel is launch with arrays passed from OpenACC in main.

# CUDA C Interoperability (Reversed)



## OpenACC Kernels

## **CUDA C Main**

```
void saxpy(int n, float a, float *
restrict x, float * restrict y)
{
    #pragma acc kernels
deviceptr(x[0:n],y[0:n])
    {
       for(int i=0; i<n; i++)
         {
            y[i] += 2.0*x[i];
        }
    }
}</pre>
```

By passing a device pointer to an OpenACC region, it's possible to add OpenACC to an existing CUDA code.

Memory is managed via standard CUDA calls.

# **CUDA Fortran**



### Standard Fortran

```
module mymodule contains
  subroutine saxpy(n, a, x, y)
    real :: x(:), y(:), a
    integer :: n, i
    do i=1,n
      y(i) = a*x(i)+y(i)
    enddo
  end subroutine saxpy
end module mymodule
program main
  use mymodule
  real :: x(2**20), y(2**20)
  x = 1.0, y = 2.0
  ! Perform SAXPY on 1M elements
  call saxpy(2**20, 2.0, x, y)
end program main
```

- Serial loop over 1M elements, executes 1M times sequentially.
- Data is resident on CPU.

### Parallel Fortran

```
module mymodule contains
  attributes(global) subroutine saxpy(n, a, x, y)
    real :: x(:), y(:), a
    integer :: n, i
    attributes(value) :: a, n
    i = threadIdx%x+(blockIdx%x-1)*blockDim%x
    if (i \le n) y(i) = a*x(i)+y(i)
  end subroutine saxpy
end module mymodule
program main
  use cudafor; use mymodule
  real, device :: x_d(2**20), y_d(2**20)
  x_d = 1.0, y_d = 2.0
  ! Perform SAXPY on 1M elements
  call saxpy <<<4096,256>>>(2**20, 2.0, x_d, y_d)
end program main
```

- Parallel kernel, executes 1M times in parallel in groups of 256 elements.
- Data must be copied to/from GPU (implicit).

http://developer.nvidia.com/cuda-fortran

# **CUDA Fortran Interoperability**



# OpenACC Main

```
program main
   use mymodule
   integer, parameter :: N =
2**20
   real, dimension(N) :: X, Y

X(:) = 1.0
   Y(:) = 0.0

!$acc data copy(y) copyin(x)
   call saxpy(N, 2.0, x, y)
   !$acc end data
end program
```

## CUDA Fortran Kernel & Launcher

```
module mymodule
  contains
  attributes(global) &
  subroutine saxpy_kernel(n, a, x, y)
    real :: x(:), y(:), a
    integer :: n, i
    attributes(value) :: a, n
    i = threadIdx%x+(blockIdx%x-1)*blockDim%x
    if (i \le n) y(i) = a*x(i)+y(i)
  end subroutine saxpy_kernel
  subroutine saxpy (n, a, x, y)
    use cudafor
    real, device :: x(:), y(:)
    real :: a
    integer :: n
    call saxpy_kernel <<< 4096, 256 >>> (n, a, x, y)
  end subroutine saxpy
end module mymodule
```

- Thanks to the "device" attribute in saxpy, no host\_data is needed.
- OpenACC manages the data and passes device pointers to CUDA.

- CUDA kernel launch wrapped in function expecting device arrays.
- Kernel is launch with arrays passed from OpenACC in main.

# OpenACC with CUDA Fortran Main



## CUDA Fortran Main w/ OpenAcc Region

Using the "deviceptr" data clause makes it possible to integrate OpenACC into an existing CUDA application.

CUDA C takes a few more tricks to compile, but can be done.

In theory, it should be possible to do the same with C/C++ (including Thrust), but in practice compiler incompatibilities make this difficult.

```
program main
    use cudafor
    integer, parameter :: N = 2**20
    real, device, dimension(N) :: X, Y
    integer :: i
    real :: tmp

X(:) = 1.0
    Y(:) = 0.0

!$acc kernels deviceptr(x,y)
    y(:) = y(:) + 2.0*x(:)
    !$acc end kernels

tmp = y(1)
    print *, tmp
end program
```

# **CUBLAS Library**



### Serial BLAS Code

## Parallel cuBLAS Code

```
int N = 1<<20;
...
// Use your choice of blas library
// Perform SAXPY on 1M elements
blas_saxpy(N, 2.0, x, 1, y, 1);</pre>
```

```
int N = 1<<20;

cublasInit();
cublasSetVector(N, sizeof(x[0]), x, 1, d_x, 1);
cublasSetVector(N, sizeof(y[0]), y, 1, d_y, 1);

// Perform SAXPY on 1M elements
cublasSaxpy(N, 2.0, d_x, 1, d_y, 1);

cublasGetVector(N, sizeof(y[0]), d_y, 1, y, 1);

cublasShutdown();</pre>
```

You can also call cuBLAS from Fortran, C++, Python, and other languages

http://developer.nvidia.com/cublas

# CUBLAS Library & OpenACC



# OpenACC Main Calling CUBLAS

OpenACC can interface with existing GPU-optimized libraries (from C/C++ or Fortran).

### This includes...

- CUBLAS
- Libsci\_acc
- CUFFT
- MAGMA
- CULA
- ...

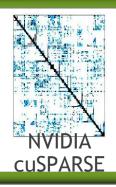
```
int N = 1 << 20;
float *x. *v
// Allocate & Initialize X & Y
cublasInit();
#pragma acc data copyin(x[0:N]) copy(y[0:N])
  #pragma acc host_data use_device(x,y)
    // Perform SAXPY on 1M elements
    cublasSaxpy(N, 2.0, d_x, 1, d_y, 1);
cublasShutdown();
```

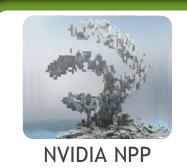
# Some GPU-accelerated Libraries













Vector Signal Image Processing



GPU Accelerated <u>Linear Algebra</u>



Matrix Algebra on GPU and Multicore





**NVIDIA** cuFFT





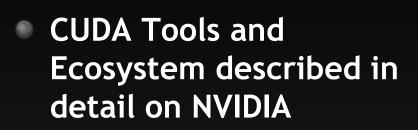


Sparse Linear Algebra open source initiative



# Explore the CUDA (Libraries) Ecosystem





**Developer Zone:** 

developer.nvidia.com/cu da-tools-ecosystem



DEVELOPER CENTERS TECHNOLOGIES TOOLS RESOURCES COMMUNITY

### GPLI-Accelerated Libraries

Adding GPU-acceleration to your application can be as easy as simply calling a library function. Check out the extensive list of high performance GPU-accelerated libraries below. If you would like other libraries added to this list please contact us.



NVIDIA CUDA Fast Fourier Transform Library (cuFFT) provides a simple interface for computing FFTs up to 10x faster, without having to develop your own custom GPU FFT implementation.



NVIDIA CUDA BLAS Library (cuBLAS) is a GPU-accelerated version of the complete standard BLAS library that delivers 6x to 17x faster performance than the latest MKI RI AS



### CULA Tools

GPU-accelerated linear algebra library by EM Photonics, that utilizes CUDA to dramatically improve the computation speed of sophisticated mathematics

### OHICKLINKS

The NVIDIA Registered Developer

Registered Developers Website

NVDeveloper (old site)

CUDA Newsletter

CUDA Downloads

CUDA GPUs

Get Started - Parallel Computing

CUDA Spotlights

CUDA Tools & Ecosystem

FEATURED ARTICLES

### MAGMA **KNOO**ET

A collection of next gen linear algebra routines. Designed for heterogeneous GPU-based architectures Supports current I APACK and RI AS standards



### IMSI, Fortran Numerical Library

Developed by RogueWave, a comprehensive set of mathematical and statistical functions that offloads



NVIDIA CUDA Sparse (cuSPARSE) 8x performance boost.



Matric library provides a collection of basic linear algebra subroutines used for sparse matrices that delivers over





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### NVIDIA CLISE

NVIDIA NPP

A GPU accelerated Open Source C++ library of generic parallel algorithms for sparse linear algebra and graph computations. Provides an easy to use high-level interface.

a GPU accelerated library with a ven

large collection of 1000's of image



### AccelerEyes ArrayFire

Comprehensive GPU function library, including functions for math, signal and image processing, statistics, and more. Interfaces for C, C++, Fortran, and Python



Generation library performs high quality GPII-accelerated random number generation (RNG) over 8x faster than typical CPU only code.



OpenACC Compiler For \$199



Local Single GPU CUDA Debugging!



CUDA Spotlight: Lorena Barba, Boston University



Stanford To Host CUDA On Campus Day, April 13, 2012







### NVIDIA CUDA Math Library

functions, providing high

An industry proven, highly accurate collection of standard mathematical

A powerful, open source library of parallel algorithms and data structures. Perform GPU-accelerated

# Thrust C++ Template Library

# Serial C++ Code with STL and Boost

## Parallel C++ Code

# Thrust C++ and OpenACC??

## OpenACC Saxpy

## Thrust Main

```
void saxpy(int n, float a, float *
restrict x, float * restrict y)
{
    #pragma acc kernels
    deviceptr(x[0:n],y[0:n])
        {
            for(int i=0; i<n; i++)
            {
                y[i] += 2.0*x[i];
            }
        }
}</pre>
```

```
int main(int argc, char **argv)
 int N = 1 << 20;
  thrust::host_vector<float> x(N), y(N);
 for(int i=0; i<N; i++)
   x[i] = 1.0f;
   y[i] = 1.0f;
  thrust::device_vector<float> d_x = x;
 thrust::device_vector<float> d_y = y;
 thrust::device_ptr<float> p_x = &d_x[0];
  thrust::device_ptr<float> p_y = &d_y[0];
 saxpy(N,2.0,p_x.get(),p_y.get());
 y = d_y;
 return 0;
```

# How to play well with others



## My advice is to do the following:

- 1. Start with OpenACC
  - Expose high-level parallelism
  - Ensure correctness
  - Optimize away data movement last
- 2. Leverage other work that's available (even if it's not OpenACC)
  - Common libraries (good software engineering practice)
  - Lots of CUDA already exists
- 3. Share your experiences
  - OpenACC is still very new, best practices are still forming.
  - Allow others to leverage your work.

