



# Introduction to OpenACC

## Alexander B. Pacheco

User Services Consultant LSU HPC & LONI sys-help@loni.org

LONI Parallel Programming Workshop Louisiana State University Baton Rouge June 10-12, 2013







 OpenACC Application Program Interface describes a collection of compiler directive to specify loops and regions of code in standard C, C++ and Fortran to be offloaded from a host CPU to an attached acclerator.

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provides portability across operating systems, host CPUs and accelerators









#### The Standard for GPU Directives

Simple: Directive are the easy path to accelerate compute intensive

applications

Open: OpenACC is an open GPU directives standard, making GPU programming straightforwards and portable across parallel

and multi-core processors

Powerful: GPU directives allow complete access to the massive parallel

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power of a GPU





#### OpenACC II



## High Level

- Compiler directives to specify parallel regions in C & Fortran
  - Offload parallel regions
  - Portable across OSes, host CPUs, accelerators, and compilers
- Create high-level heterogenous programs
  - Without explicit accelerator intialization
  - Without explicit data or program transfers between host and accelerator

#### High Level ... with low-level access

- Programming model allows programmers to start simple
- Compiler gives additional guidance
  - Loop mappings, data location and other performance details
- Compatible with other GPU languages and libraries
  - Interoperate between CUDA C/Fortran and GPU libraries
  - e.g. CUFFT, CUBLAS, CUSPARSE, etc









- Directives are easy and powerful.
- Avoid restructuring of existing code for production applications.
- Focus on expressing parallelism.

# OpenACC is not GPU Programming

OpenACC is Expressing Parallelism in your code









## Serial Code

```
program saxpy
  implicit none
  integer, parameter :: dp = selected_real_kind(15)
  integer, parameter :: ip = selected_int_kind(15)
  integer(ip) :: i.n
  real(dp), dimension(:), allocatable :: x, y
  real(dp) :: a, start_time, end_time
  n=500000000
  allocate(x(n),y(n))
  x = 1.0d0
 y = 2.0d0
  a = 2.0
  call cpu time(start time)
  do i = 1, n
    v(i) = v(i) + a * x(i)
  end do
  call cpu time (end time)
  deallocate(x, y)
  print '(a,f8.6)', 'SAXPY Time: ', end time - start time
end program saxpy
```

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## Simple Example II



## OpenMP Code

```
program saxpy
  implicit none
  integer, parameter :: dp = selected real kind(15)
  integer, parameter :: ip = selected int kind(15)
  integer(ip) :: i,n
  real(dp), dimension(:), allocatable :: x, y
  real(dp) :: a, start_time, end_time
  n=500000000
  allocate(x(n),y(n))
  !$omp parallel sections
  !Somp section
  x = 1.0
  !Somp section
  y = 1.0
  !Somp end parallel sections
  a = 2.0
  call cpu_time(start_time)
  !Somp parallel do default(shared) private(i)
  do i = 1, n
    y(i) = y(i) + a * x(i)
  end do
  !$omp end parallel do
  call cpu time (end time)
  deallocate(x, y)
  print '(a,f8.6)', 'SAXPY Time: ', end time - start time
end program saxpy
```

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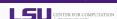




## OpenACC Code

```
program saxpy
  use omp lib
  implicit none
  integer :: i,n
  real, dimension(:), allocatable :: x, y
  real :: a, start time, end time
  allocate(x(n),y(n))
  a = 2.0
  !$acc data create(x,y) copyin(a)
  !$acc parallel
 x(:) = 1.0
  !$acc end parallel
  !Sacc parallel
  y(:) = 1.0
  !$acc end parallel
  start_time = omp_get_wtime()
  !Sacc parallel loop
  do i = 1, n
    y(i) = y(i) + a * x(i)
  end do
  !$acc end parallel loop
  end_time = omp_get_wtime()
  !$acc end data
  deallocate(x,y)
  print '(a,f15.6,a)', 'SAXPY Time: ', end time - start time, 'in secs'
end program saxpy
```

end program saxpy









## CUDA Fortran Code

```
module mymodule
contains
  attributes(global) subroutine saxpy(n, a, x, y)
    real :: x(:), v(:), a
    integer :: n, i
    attributes(value) :: a, n
   i = threadIdx%x+(blockIdx%x-1)*blockDim%x
    if (i \le n) \forall (i) = a \times x(i) + y(i)
  end subroutine saxpy
end module mymodule
program main
  use cudafor; use mymodule
  integer, parameter :: n = 100000000
  real, device :: x d(n), y d(n)
  real, device :: a d
  real :: start_time, end time
  x d = 1.0
  v d = 2.0
  a d = 2.0
  call cpu_time(start_time)
  call saxpy<<<4096, 256>>>(n, a, x_d, y_d)
  call cpu_time(end_time)
  print '(a,f15.6,a)', 'SAXPY Time: ', end_time - start_time, 'in secs'
end program main
```

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## Simple Example V



## Compile

LSU

```
[apacheco@philip038 2013-LONI]$ pgf90 -o saxpy saxpy.f90
[apacheco@philip038 2013-LONI]$ pgf90 -mp -o saxpy_omp saxpy_omp.f90
[apacheco@philip038 2013-LONI]$ pgf90 -acc -ta=nvidia -o saxpy_acc saxpy_acc.f90
[apacheco@philip038 2013-LONI]$ pgf90 -o saxpy_cuda saxpy.cuf
```

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## Speed Up

Algorithm	Device	Time (s)	Speedup
Serial	Xeon X5650	0.231282	1
OpenMP (12 threads)	Xeon X5650	0.063231	3.6x
OpenACC	M2070	0.014329	16.1x
CUDA	M2070	0.006901	33.5x

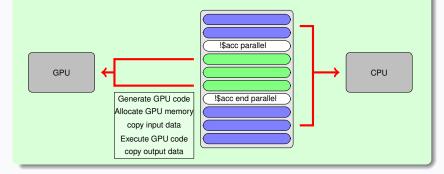




#### OpenACC Execution Model



- Application code runs on the CPU (sequential, shared or distributed memory)
- OpenACC directives indicate that the following block of compute intensive code needs to be offloaded to the GPU or accelerator.







## **Building Block of OpenACC**





- Program directives
  - Syntax
    - C/C++: #pragma acc <directive> [clause]
    - Fortran: !\$acc <directive> [clause]
  - Regions
  - Loops
  - Synchronization
  - Data Structure
  - ...
- Runtime library routines









- if (condition)
- async (expression)
- data management clauses
  - $copy(\cdots), copyin(\cdots), copyout(\cdots)$
  - create( $\cdots$ ), present( $\cdots$ )
  - $\bullet \ \, \mathsf{present\_or\_copy}\{,\mathsf{in},\mathsf{out}\}(\cdots) \,\,\mathsf{or}\,\,\mathsf{pcopy}\{,\mathsf{in},\mathsf{out}\}(\cdots) \\$
  - $\bullet \ \, \mathsf{present\_or\_create}(\cdots) \ \, \mathsf{or} \ \, \mathsf{pcreate}(\cdots) \\$
- reduction(operator:list)









- System setup routines
  - acc\_init(acc\_device\_nvidia)
  - acc set device type(acc device nvidia)
- acc set\_device\_num(acc\_device\_nvidia)
- Synchronization routines
  - acc\_async\_wait(int)
  - acc async wait all()









## C: #pragma acc kernels [clause]

#### Fortran !\$acc kernels [clause]

- The kernels directive expresses that a region may contain parallelism and the compiler determines what can be safely parallelized.
- The compiler breaks code in the kernel region into a sequence of kernels for execution on the accelerator device.
- For the codes on the right, the compiler identifies 2 parallel loops and generates 2 kernels.
- What is a kernel? A function that runs in parallel on the GPU.
- When a program encounters a kernels contruct, it will launch a sequence of kernels in order on the device.

```
!Sacc kernels
do i = 1, n
   x(i) = 1.0
   v(i) = 2.0
end do
do i = 1, n
   v(i) = v(i) + a * x(i)
end do
!$acc end kernels
#pragma acc kernels
  for (i = 0; i < n; i++) {
    x[i] = 1.0;
    v[i] = 2.0;
  for (i = 0; i < n; i++) {
    v[i] = a * x[i] + v[i];
```









- The parallel directive identifies a block of code as having parallelism.
- Compiler generates a parallel kernel for that loop.

C: #pragma acc parallel [clauses]

Fortran: !\$acc parallel [clauses]

```
!$acc parallel
do i = 1, n
   x(i) = 1.0
  v(i) = 2.0
end do
do i = 1, n
   v(i) = v(i) + a * x(i)
end do
!$acc end parallel
#pragma acc parallel
  for (i = 0; i < n; i++) {
    x[i] = 1.0;
   v[i] = 2.0;
  for (i = 0; i < n; i++) {
    y[i] = a*x[i] + y[i];
```



## OpenACC Loop Directive



- Loops are the most likely targets for Parallelizing.
- The Loop directive is used within a parallel or kernels directive indentifying a loop that can be executed on the accelerator device.

C: #pragma acc loop [clauses]

Fortran: !\$acc loop [clauses]

 The loop directive can be combined with the enclosing parallel or kernels

C: #pragma acc kernels loop [clauses]

Fortran: !\$acc parallel loop [clauses]

 The loop directive clauses can be used to optimize the code. This however requires knowledge of the accelerator device.

Clauses: gang, worker, vector, num\_gangs, num workers

```
!$acc loop
do i = 1, n
    y(i) = y(i) + a * x(i)
end do
!$acc end loop

#pragma acc loop
for (i = 0; i < n; i++) {
    y[i] = a*x[i] + y[i];
}</pre>
```





## OpenACC parallel vs. kernels



#### **PARALLEL**

- Requires analysis by programmer to ensure safe parallelism.
- Straightforward path from OpenMP

#### **KERNELS**

- Compiler performs parallel analysis and parallelizes what it believes is safe.
- Can cover larger area of code with single directive.

Both approaches are equally valid and can perform equally well.





#### **Compilation**



- C:
  - pgcc -acc [-Minfo=accel] [-ta=nvidia] -o saxpyc\_acc saxpy\_acc.c
- Fortran 90:
  - pgf90 -acc [-Minfo=accel] [-ta=nvidia] -o saxpyf\_acc saxpy\_acc.f90

```
Compiler Output
[apacheco@mikel nodataregion] $ pgcc -acc -ta=nvidia, time -Minfo=accel -o saxpvc acc saxpv acc.c
main:
     19, Generating copyin(x[0:500000000])
         Generating copy(y[0:500000000])
         Generating compute capability 1.0 binary
         Generating compute capability 2.0 binary
     21. Loop is parallelizable
         Accelerator kernel generated
         21, #pragma acc loop gang, vector(128) /* blockIdx.x threadIdx.x */
            CC 1.0 : 10 registers; 44 shared, 0 constant, 0 local memory bytes
            CC 2.0 : 15 registers; 0 shared, 60 constant, 0 local memory bytes
[apacheco@mikel nodataregion] $ pqf90 -acc -ta=nvidia,time -Minfo=accel -o saxpyf acc saxpy acc.f90
saxpy:
     17, Accelerator kernel generated
         17, CC 1.0 : 7 registers; 40 shared, 4 constant, 0 local memory bytes
            CC 2.0 : 15 registers; 0 shared, 56 constant, 0 local memory bytes
         18, !$acc loop gang, vector(256) ! blockidx%x threadidx%x
     17, Generating copy(v(1:500000000))
         Generating copyin(x(1:500000000))
         Generating compute capability 1.0 binary
```





[apacheco@mikel nodataregion]\$

Generating compute capability 2.0 binary

## Running





- The PGI compiler provides automatic instrumentation when PGI\_ACC\_TIME=1 at runtime
- You can also obtain automatic instrumentation by using the -ta=nvidia,time compiler flag.

```
[apacheco@mike381 nodataregion]$ ./saxpyc_acc
Accelerator Kernel Timing data
/work/apacheco/2013-LONI/openmp/saxpv/nodataregion/saxpv acc.c
  main
    19: region entered 1 time
        time(us): total=9.195.380 init=7.246.895 region=1.948.485
                  kernels=58,028 data=1,835,336
        w/o init: total=1,948,485 max=1,948,485 min=1,948,485 avg=1,948,485
        21: kernel launched 1 times
            grid: [65535] block: [128]
            time(us): total=58,028 max=58,028 min=58,028 avg=58,028
SAXPY Time: 9.195481
[apacheco@mike381 nodataregion]$ ./saxpvf acc
Accelerator Kernel Timing data
/work/apacheco/2013-LONI/openmp/saxpy/nodataregion/saxpy acc.f90
  saxpy
    17: region entered 1 time
        time(us): total=9,180,978 init=7,254,065 region=1,926,913
                  kernels=59,013 data=1,923,915
        w/o init: total=1,926,913 max=1,926,913 min=1,926,913 avg=1,926,913
        17: kernel launched 1 times
            grid: [65535] block: [256]
            time(us): total=59,013 max=59,013 min=59,013 avg=59,013
SAXPY Time:
                   9.181015
```









Execution	С		Fortran	
	Time	SpeedUp	Time	Speedup
Serial	0.511		0.993	
OpenMP (16 Threads)	0.186	2.75	0.244	4.07
OpenACC (M2090)	9.195	0.056	9.181	0.108

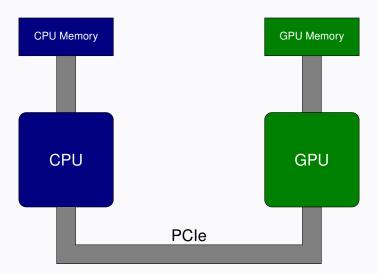
- What's going with OpenACC code?
- Why even bother with OpenACC if performance is so bad?





# Offloading a Parallel Kernel



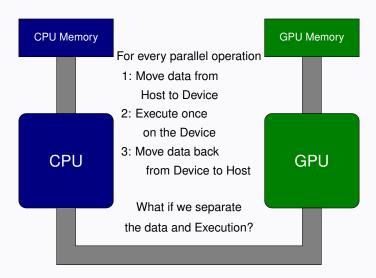


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## Offloading a Parallel Kernel



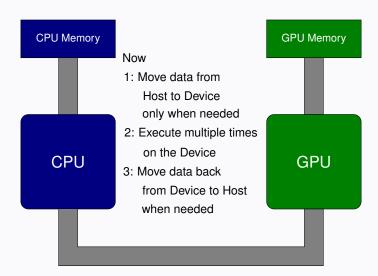


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## Offloading a Parallel Kernel





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 The data construct defines a region of code in which GPU arrays remain on the GPU and are shared among all kernels in that region

```
!$acc data [clause]
    !$acc parallel loop
    ...
    !$acc end parallel loop
!$acc end data
```

Arrays used within the data region will remain on the GPU until the end of the data region.



#### Data Clauses



- copy(list) Allocates memory on GPU and copies data from host to GPU when entering region and copies data to the host when exiting region.
- copyin(list) Allocates memory on GPU and copies data from host to GPU when entering region.
- copyout(list) Allocates memory on GPU and copies data to the host when exiting region.
  - create(list) Allocates memory on GPU but does not copy.
- present(list) Data is already present on GPU from another containing data region.
- Other clauses: present\_or\_copy[in|out], present\_or\_create, deviceptr.





## Array Shaping



- Compiler sometime cannot determine size of arrays
  - Must specify explicitly using the data clauses and array "shape"
- C #pragma acc data copyin(a[0:size]), copyout(b[s/4:3\*s/4])
- Fortran !\$acc data copyin(a(1:size)), copyout(b(s/4:3\*s/4))
  - Note: data clauses can be used on data, parallel or kernels





### **Update Construct**





- Used to update existing data after it has changed in its corresponding copy (e.g. upate device copy after host copy changes).
- Move data from GPU to host, or host to GPU.
- Data movement can be conditional and asynchronous.
- Fortran

```
\verb|!acc update [clause \cdots||\\
```

C

```
\#pragma acc update [clause \cdots]
```

- Clause
  - host(list)
  - device(list)
  - if(expression)
  - async(expression)





#### SAXPY using data clause



```
program saxpy
  use omp_lib
  implicit none
  integer :: 1,n
  real, dimension(:), allocatable :: x, y
  real :: a,start_time, end_time
  allocate(x(n),v(n))
  !$acc data create(x,y) copyin(a)
  !$acc parallel
  !Sacc end parallel
  !Sacc parallel
  !$acc end parallel
  !Sacc parallel loop
  do i = 1, n
  !$acc end parallel loop
  end_time = omp_get_wtime()
  !Sacc end data
  deallocate(x,y)
  print '(a,f15.6,a)', 'SAXPY Time: ', end_time - start_time, 'in secs'
end program saxpy
```

```
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```

Execution	С		Fortran	
	Time	SpeedUp	Time	Speedup
Serial	0.511		0.993	
OpenMP (16 Threads)	0.186	2.75	0.244	4.07
OpenACC (M2090)	0.058	8.81	0.059	16.83





## Exercise: Redo Matrix Multiplication Problem I



#### С

Execution	Time	SpeedUp	GFlops/s
Serial	6.226		0.964
OpenMP 16 CPUs	0.444	14.022	13.03
OpenACC	0.175	35.577	34.265

#### Fortran

Execution Time SpeedUp		SpeedUp	GFlops/s
Serial	7.113		0.844
OpenMP 16 CPUs	0.494	14.399	12.146
OpenACC	0.257	27.677	23.346

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### Exercise: Redo Matrix Multiplication Problem II



```
program matrix_mul
  implicit none
  integer, parameter :: do - selected real kind(14)
 integer :: 1, 1, k
 integer, parameter :: nra-1500, nca-2000, ncb-1000
 real(dp) :: a(nra,nca) , b(nca,ncb) , c(nra,ncb)
 real(dp) :: flops, sum
 real(dp) :: init_time, start_time, end_time
 integer :: cl, c2, c3, cr
 integer, dimension(8) :: value
 flops = 2d0 * float(nra) * float(nca) * float(ncb)
 !$acc data create(a,b,c)
 call date_and_time (VALUES=value)
 init time = float(value(6)+60) + float(value(7)) + float(value(8))/1000d0
 c = 0d0
 do 1 - 1.nra
    do 1 - 1, nca
     end do
  end do
  do 1 - 1.nca
    do 1 = 1, nch
    end do
 end do
 call date and time (VALUES-value)
 start_time = float(value(6) + 60) + float(value(7)) + float(value(8))/1000d0
  !$acc parallel loop private(sum)
 do j = 1, nca
    do k = 1, ncb
        sum - 0d0
        !Sacc loop reduction (+: sum)
         sum = sum + a(1, j) + b(j, k)
        end do
        c(i,k) = sum
    end do
  !$acc end parallel loop
 call date and time (VALUES=value)
 end_time = float(value(6) +60) + float(value(7)) + float(value(8))/1000d0
 print '(a, f6.3, a, f6.3, a, f7.3)', 'Init Time: ', start_time - init_time, &
      ' Calc Time: ', end_time - start_time, &
      ' GFlops: ', 1d-9 * flops/(end_time - start_time)
end program matrix_mul
```

```
#include <stdio.h>
#include <stdlib.h>
#include <time.h>
#define dt(start, end) ((end.tv_sec - start.tv_sec) + \
                        1/1000000.0+(end.tv_usec - start.tv_usec))
 int 1, 1, k;
 double a[nra][nca],b[nca][ncb],c[nra][ncb];
 struct timeval icalc, scalc, ecalc;
 double flops, sum, timing ;
 flops = 2.0 * nra * nca * ncb;
#pragma acc data create(a,b,c)
   gettimeofday(Sicalc, NULL):
   for (i = 0; i < nra; i++) (
     for (j = 0; j < nca; j++) {
       a[i][j] = (double)(i+j);
   for (1 = 0; 1 < nca; 1++) (
     for (k = 0; k < ncb; k++) {
       b[j][k] = (double)(i+1);
     for (k = 0; k < ncb; k++) {
   gettimeofday(sscalc, NULL);
#pragma acc parallel loop private(sum)
     for (k = 0; k < ncb; k++) (
       sum = 0.0;
#pragma acc loop seq
         sum = sum + a[1][j] + b[j][k];
       c[i][k] = sum:
   gettimeofday (secalc, NULL);
 timing = dt(scalc.ecalc):
 printf("Init Time: %6.3f Calc Time: %6.3f GFlops: %7.3f\n",dt(icalc,scalc),timing,1
```



#### Reduction



Reduction clause is allowed on parallel and loop constructs

```
Fortran

!$acc parallel reduction(operation: var)
    structured block with reduction on var
!$acc end parallel

C

#pragma acc kernels reduction(operation: var) {
    structured block with reduction on var
}
```

 Redo Calculation of Pi in OpenACC and compare timing results with serial and OpenMP.





## Further Speedups



- OpenACC gives us more detailed control over parallelization
  - Via gang, worker and vector clauses
- By understanding more about specific GPU on which you're running, using these clauses may allow better performance.
- By understanding bottlenecks in the code via profiling, we can reorganize the code for even better performance.





## General Principles: Finding Parallelism in Code



- (Nested) for/do loops are best for parallelization
- Large loop counts are best
- Iterations of loops must be independent of each other
  - To help compiler: restrict keyword (C), independent clause
  - Use subscripted arrays, rather than pointer-indexed arrays
- Data regions should avoid wasted bandwidth
  - Can use directive to explicitly control sizes
- Various annoying things can interfere with accelerated regions.
  - Function calls within accelerated region must be inlineable.

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No IO





## OpenACC: Is it worth it?



- High-level. No involvement of OpenCL, CUDA, etc
- Single source. No forking off a separate GPU code. Compile the same program for accelerators or serial, non-GPU programmers can play along.
- Efficient. Experience shows very favorable comparison to low-level implementations of same algorithms.
- Performance portable. Supports GPU accelreators and co-processors from multiple vendors, current and future versions.
- Incremental. Developers can port and tune parts of their application as resources and profiling dictates. No wholesale rewrite required. Which can be quick.









Lecture derived from slides and presentations by

- Michael Wolfe, PGI
- Jeff Larkin, NVIDIA
- John Urbanic, PSC

Search for OpenACC presentations at the GPU Technology Conference Website for further study

http://www.gputechconf.com/gtcnew/on-demand-gtc.php



