

# ZERO TO GPU HERO WITH OPENACC

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# OUTLINE

## Topics to be covered

- What is OpenACC
- Profile-driven Development
- OpenACC with CUDA Unified Memory
- OpenACC Data Directives
- OpenACC Loop Optimizations
- Where to Get Help

# INTRODUCTION TO OPENACC

**OpenACC** is a directives-based programming approach to **parallel computing** designed for **performance** and **portability** on CPUs and GPUs for HPC.

Add Simple Compiler Directive

```
main()
{
    <serial code>
    #pragma acc kernels
    {
        <parallel code>
    }
}
```



# 3 WAYS TO ACCELERATE APPLICATIONS

Applications

Libraries

**Easy to use**  
**Most Performance**

Compiler  
Directives

**Easy to use**  
**Portable code**

**OpenACC**

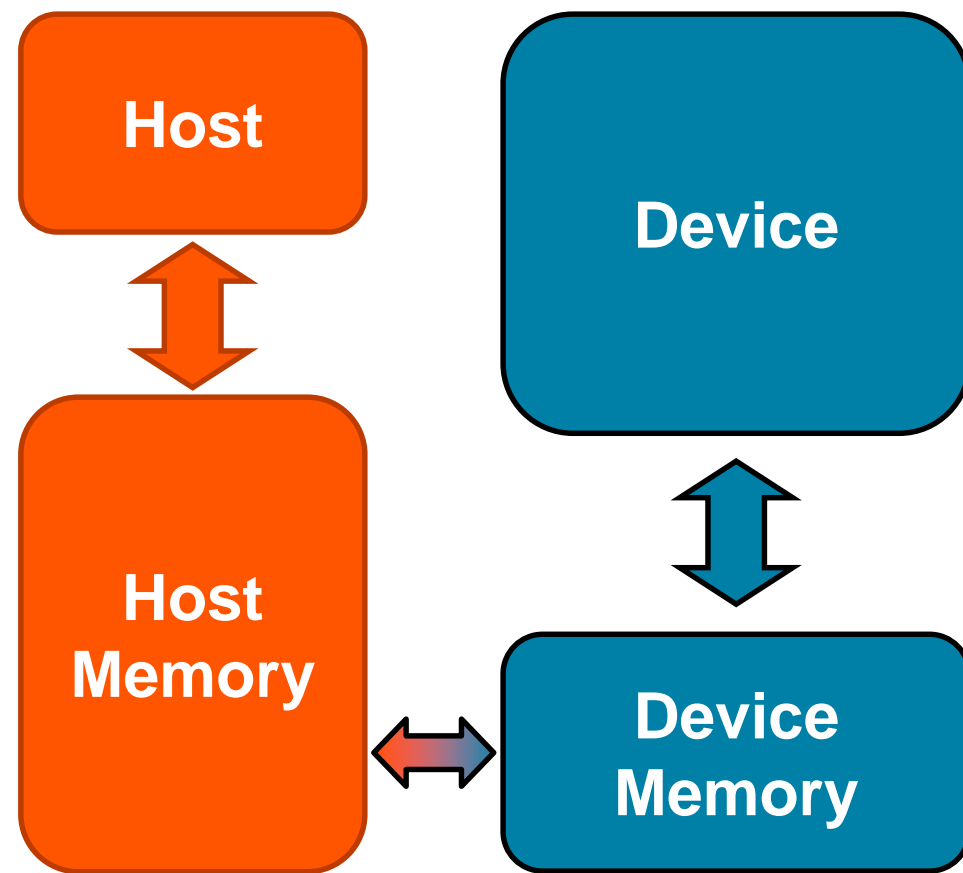
Programming  
Languages

**Most Performance**  
**Most Flexibility**

# OPENACC PORTABILITY

## Describing a generic parallel machine

- OpenACC is designed to be portable to many existing and future parallel platforms
- The programmer need not think about specific hardware details, but rather express the parallelism in generic terms
- An OpenACC program runs on a *host* (typically a CPU) that manages one or more parallel *devices* (GPUs, etc.). The host and device(s) are logically thought of as having separate memories.



# OPENACC

Three major strengths

Incremental

Single Source

Low Learning Curve

# OPENACC

## Incremental

- Maintain existing sequential code
- Add annotations to expose parallelism
- After verifying correctness, annotate more of the code

### Enhance Sequential Code

```
#pragma acc parallel loop  
for( i = 0; i < N; i++ )  
{  
    < loop code >  
}  
  
#pragma acc parallel loop  
for( i = 0; i < N; i++ )  
{  
    < loop code >  
}
```

Begin with a working sequential code.

Parallelize it with OpenACC.

Rerun the code to verify correct behavior, remove/alter OpenACC code as needed.



# OPENACC

## Incremental

- Maintain existing sequential code
- Add annotations to expose parallelism
- After verifying correctness, annotate more of the code

## Single Source

## Low Learning Curve

# OPENACC

## Supported Platforms

POWER  
Sunway  
x86 CPU  
x86 Xeon Phi  
NVIDIA GPU  
PEZY-SC

## Single Source

- Rebuild the same code on multiple architectures
- Compiler determines how to parallelize for the desired machine
- Sequential code is maintained

The compiler can **ignore** your OpenACC code additions, so the same code can be used for **parallel** or **sequential** execution.

```
int main(){  
  
...  
  
#pragma acc parallel loop  
for(int i = 0; i < N; i++)  
    < loop code >  
}
```

# OPENACC

## Incremental

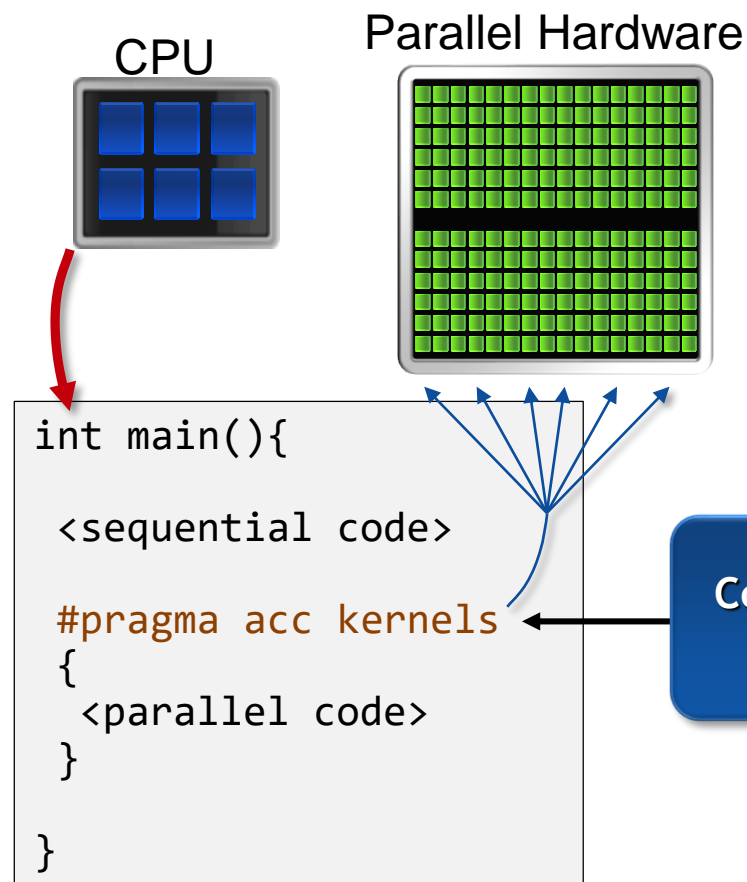
- Maintain existing sequential code
- Add annotations to expose parallelism
- After verifying correctness, annotate more of the code

## Single Source

- Rebuild the same code on multiple architectures
- Compiler determines how to parallelize for the desired machine
- Sequential code is maintained

## Low Learning Curve

# OPENACC



The programmer will give hints to the compiler about which parts of the code to parallelize.

The compiler will then generate parallelism for the target parallel hardware.

## Low Learning Curve

- OpenACC is meant to be easy to use, and easy to learn
- Programmer remains in familiar C, C++, or Fortran
- No reason to learn low-level details of the hardware.



# OPENACC

## Incremental

- Maintain existing sequential code
- Add annotations to expose parallelism
- After verifying correctness, annotate more of the code

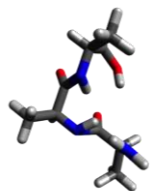
## Single Source

- Rebuild the same code on multiple architectures
- Compiler determines how to parallelize for the desired machine
- Sequential code is maintained

## Low Learning Curve

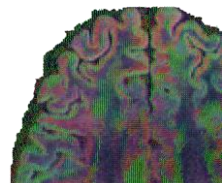
- OpenACC is meant to be easy to use, and easy to learn
- Programmer remains in familiar C, C++, or Fortran
- No reason to learn low-level details of the hardware.

# OPENACC SUCCESSES



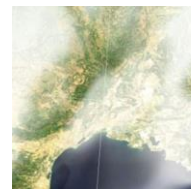
## LSDalton

Quantum Chemistry  
Aarhus University  
12X speedup  
1 week



## PowerGrid

Medical Imaging  
University of Illinois  
40 days to  
2 hours



## COSMO

Weather and Climate  
MeteoSwiss, CSCS  
40X speedup  
3X energy efficiency



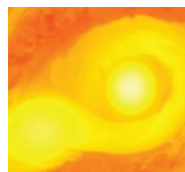
## INCOMP3D

CFD  
NC State University  
4X speedup



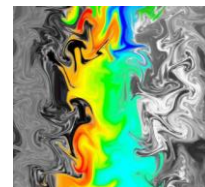
## NekCEM

Comp Electromagnetics  
Argonne National Lab  
2.5X speedup  
60% less energy



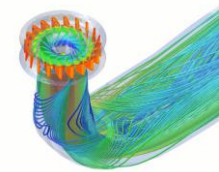
## MAESTRO CASTRO

Astrophysics  
Stony Brook University  
4.4X speedup  
4 weeks effort



## CloverLeaf

Comp Hydrodynamics  
AWE  
4X speedup  
Single CPU/GPU code



## FINE/Turbo

CFD  
NUMECA  
International  
10X faster routines  
2X faster app

# OPENACC SYNTAX

# OPENACC SYNTAX

## Syntax for using OpenACC directives in code

C/C++

```
#pragma acc directive clauses  
<code>
```

Fortran

```
!$acc directive clauses  
<code>
```

- A ***pragma*** in C/C++ gives instructions to the compiler on how to compile the code. Compilers that do not understand a particular pragma can freely ignore it.
- A ***directive*** in Fortran is a specially formatted comment that likewise instructions the compiler in its compilation of the code and can be freely ignored.
- “***acc***” informs the compiler that what will come is an OpenACC directive
- ***Directives*** are commands in OpenACC for altering our code.
- ***Clauses*** are specifiers or additions to directives.



# EXAMPLE CODE

# LAPLACE HEAT TRANSFER

Very Hot

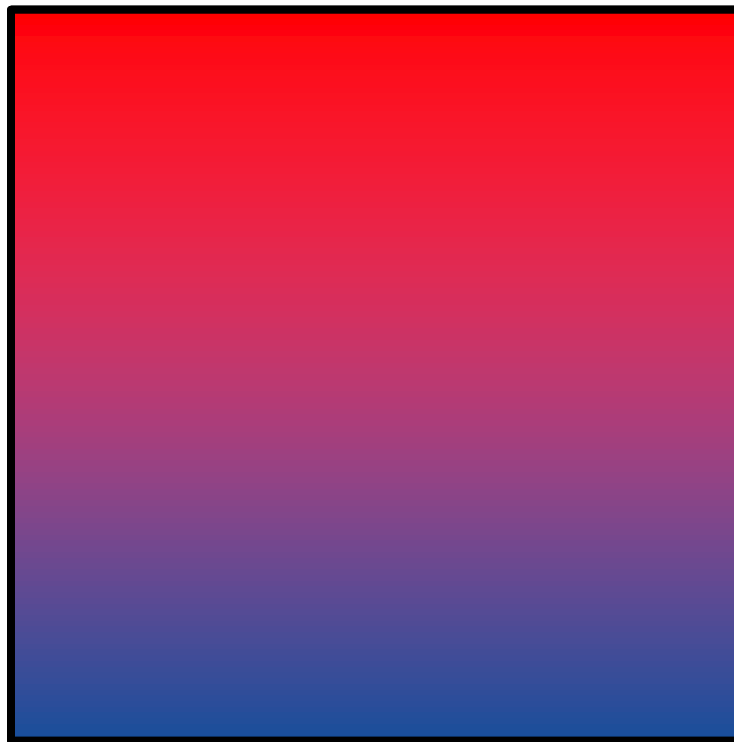
Room Temp



We will observe a simple simulation of heat distributing across a metal plate.

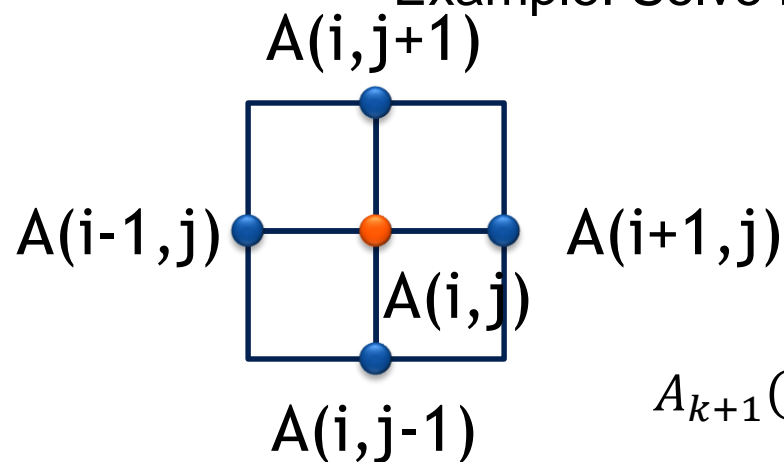
We will apply a consistent heat to the top of the plate.

Then, we will simulate the heat distributing across the plate.



# EXAMPLE: JACOBI ITERATION

- Iteratively converges to correct value (e.g. Temperature), by computing new values at each point from the average of neighboring points.
- Common, useful algorithm
- Example: Solve Laplace equation in 2D:  $\nabla^2 f(x, y) = 0$



$$A_{k+1}(i, j) = \frac{A_k(i-1, j) + A_k(i+1, j) + A_k(i, j-1) + A_k(i, j+1)}{4}$$

# JACOBI ITERATION: C CODE

```
while ( err > tol && iter < iter_max ) {  
    err=0.0;  
  
    for( int j = 1; j < n-1; j++) {  
        for(int i = 1; i < m-1; i++) {  
  
            Anew[j][i] = 0.25 * (A[j][i+1] + A[j][i-1] +  
                                A[j-1][i] + A[j+1][i]);  
  
            err = max(err, abs(Anew[j][i] - A[j][i]));  
        }  
    }  
  
    for( int j = 1; j < n-1; j++) {  
        for( int i = 1; i < m-1; i++ ) {  
            A[j][i] = Anew[j][i];  
        }  
    }  
  
    iter++;  
}
```



Iterate until converged



Iterate across matrix  
elements



Calculate new value from  
neighbors



Compute max error for  
convergence



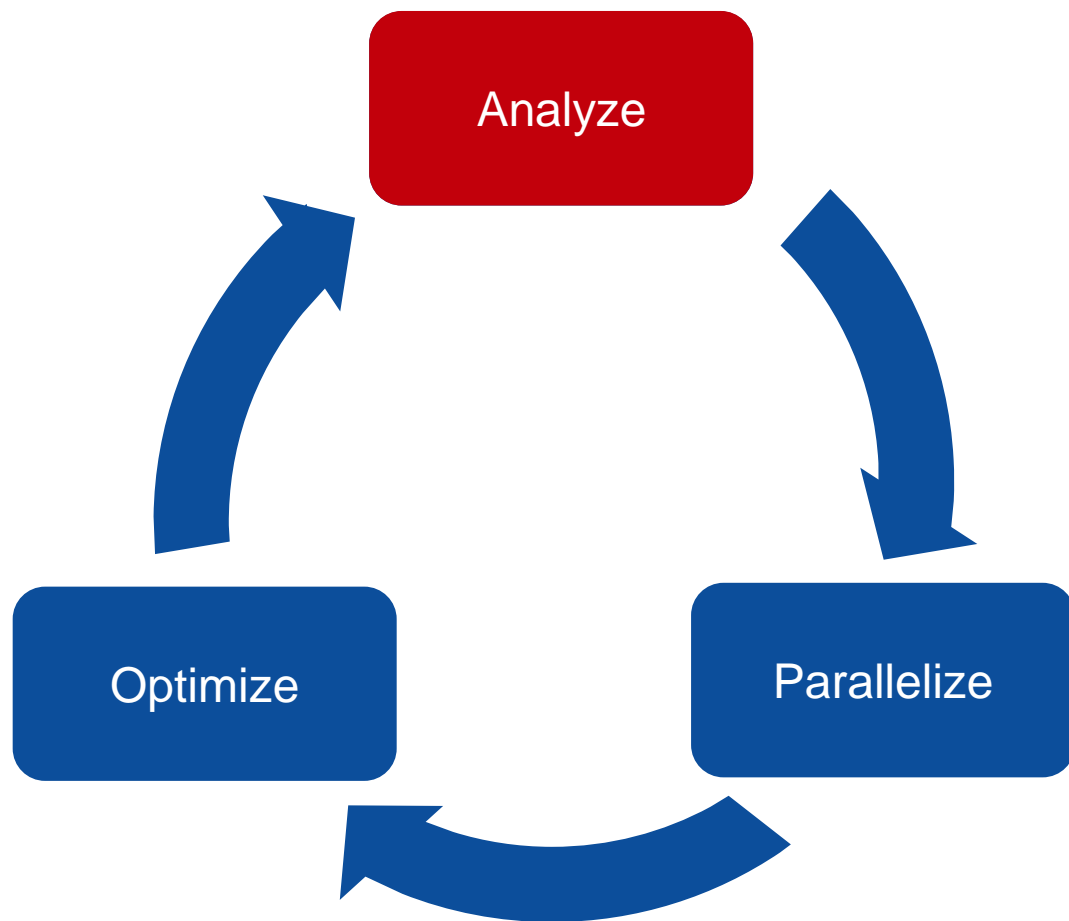
Swap input/output arrays



# PROFILE-DRIVEN DEVELOPMENT

# OPENACC DEVELOPMENT CYCLE

- **Analyze** your code to determine most likely places needing parallelization or optimization.
- **Parallelize** your code by starting with the most time consuming parts and check for correctness.
- **Optimize** your code to improve observed speed-up from parallelization.



# PROFILING SEQUENTIAL CODE

## Profile Your Code

Obtain detailed information about how the code ran.

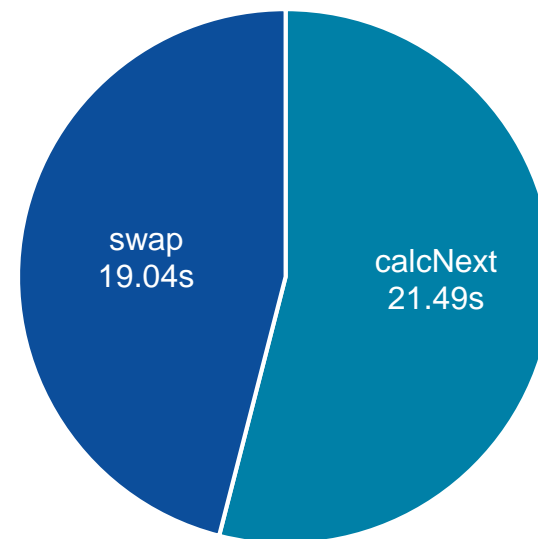
This can include information such as:

- Total runtime
- Runtime of individual routines
- Hardware counters

Identify the portions of code that took the longest to run. We want to focus on these “hotspots” when parallelizing.

## Sample Code: Laplace Heat Transfer

Total Runtime: 39.43 seconds



# OPENACC PARALLEL DIRECTIVE

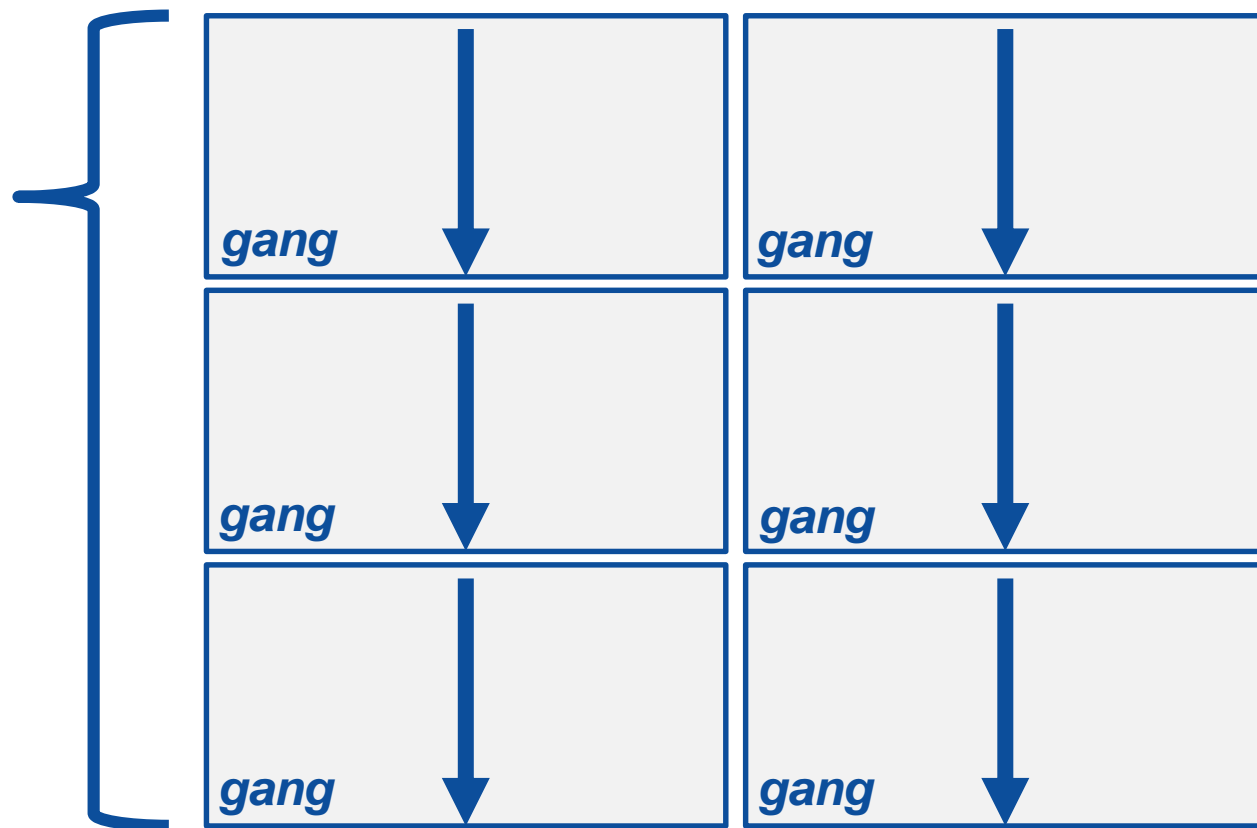
# OPENACC PARALLEL DIRECTIVE

## Expressing parallelism

```
#pragma acc parallel  
{
```

When encountering the ***parallel*** directive, the compiler will generate *1 or more parallel gangs*, which execute redundantly.

```
}
```





# OPENACC PARALLEL DIRECTIVE

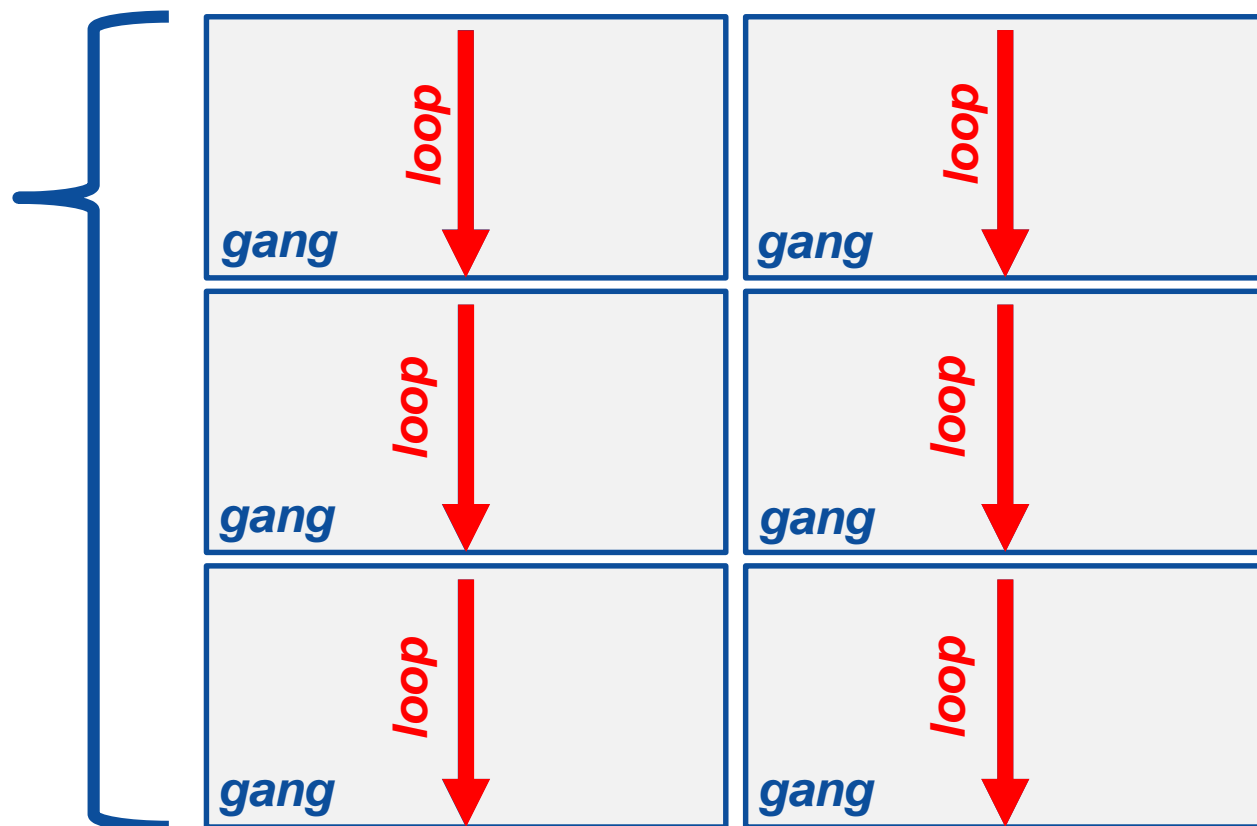
## Expressing parallelism

```
#pragma acc parallel  
{
```

```
    loop  
    for(int i = 0; i < N; i++)  
    {  
        // Do Something  
    }
```

```
}
```

This loop will be  
executed redundantly  
on each gang



# OPENACC PARALLEL DIRECTIVE

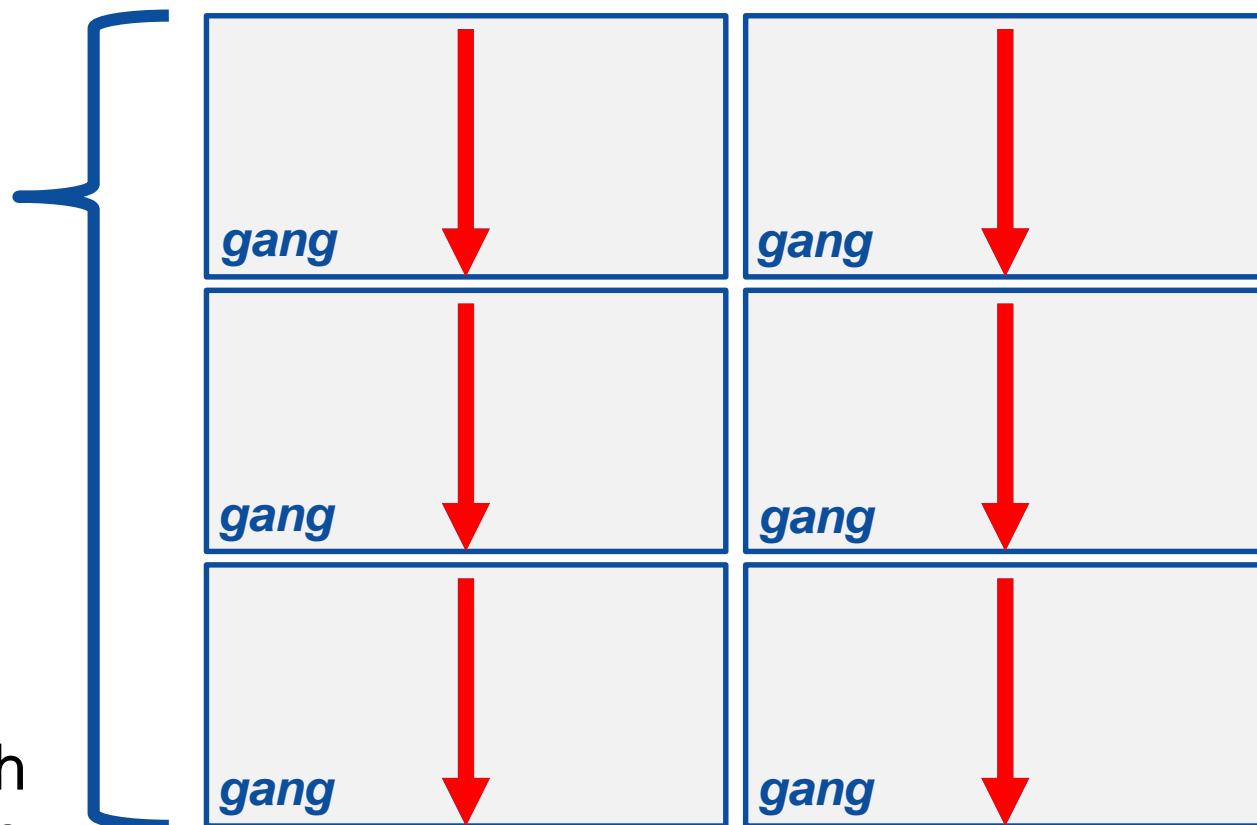
## Expressing parallelism

```
#pragma acc parallel  
{
```

```
  for(int i = 0; i < N; i++)  
  {  
    // Do Something  
  }
```

```
}
```

This means that each **gang** will execute the entire loop



# OPENACC PARALLEL DIRECTIVE

## Parallelizing a single loop

### C/C++

```
#pragma acc parallel
{
    #pragma acc loop
    for(int i = 0; i < N; i++)
        a[i] = 0;
}
```

### Fortran

```
!$acc parallel
!$acc loop
do i = 1, N
    a(i) = 0
end do
!$acc end parallel
```

- Use a **parallel** directive to mark a region of code where you want parallel execution to occur
- This parallel region is marked by curly braces in C/C++ or a start and end directive in Fortran
- The **loop** directive is used to instruct the compiler to parallelize the iterations of the next loop to run across the parallel gangs

# OPENACC PARALLEL DIRECTIVE

## Parallelizing a single loop

### C/C++

```
#pragma acc parallel loop  
for(int i = 0; i < N; i++)  
    a[i] = 0;
```

### Fortran

```
!$acc parallel loop  
do i = 1, N  
    a(i) = 0  
end do
```

- This pattern is so common that you can do all of this in a single line of code
- In this example, the parallel loop directive applies to the next loop
- This directive both marks the region for parallel execution and distributes the iterations of the loop.
- When applied to a loop with a data dependency, parallel loop may produce incorrect results

# OPENACC PARALLEL DIRECTIVE

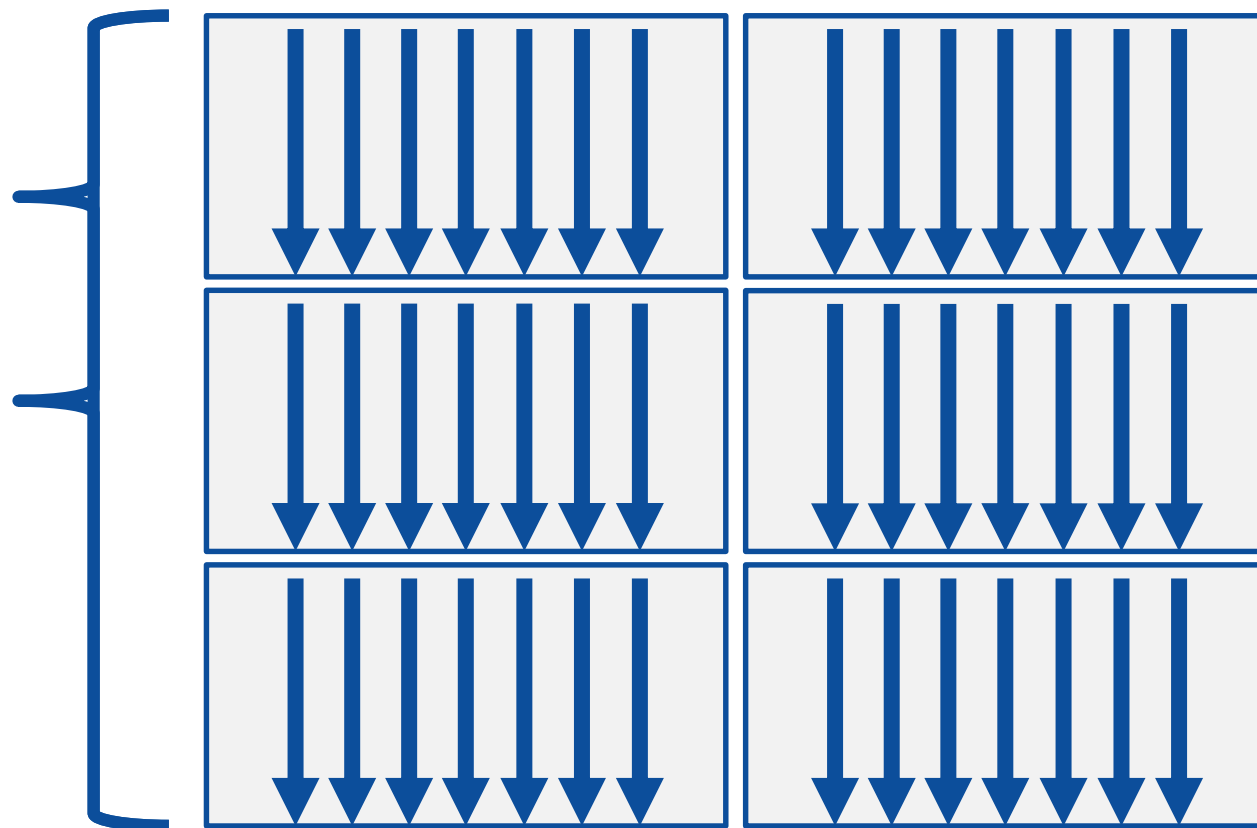
## Expressing parallelism

```
#pragma acc parallel  
{
```

```
    #pragma acc loop  
    for(int i = 0; i < N; i++)  
    {  
        // Do Something  
    }
```

```
}
```

The **loop** directive informs the compiler which loops to parallelize.



# OPENACC PARALLEL LOOP DIRECTIVE

## Parallelizing many loops

```
#pragma acc parallel loop
for(int i = 0; i < N; i++)
    a[i] = 0;

#pragma acc parallel loop
for(int j = 0; j < M; j++)
    b[j] = 0;
```

- To parallelize multiple loops, each loop should be accompanied by a parallel directive
- Each parallel loop can have different loop boundaries and loop optimizations
- Each parallel loop can be parallelized in a different way
- This is the recommended way to parallelize multiple loops. Attempting to parallelize multiple loops within the same parallel region may give performance issues or unexpected results

# PARALLELIZE WITH OPENACC PARALLEL LOOP

```
while ( err > tol && iter < iter_max ) {  
    err=0.0;  
  
    #pragma acc parallel loop reduction(max:err)  
    for( int j = 1; j < n-1; j++) {  
        for(int i = 1; i < m-1; i++) {  
  
            Anew[j][i] = 0.25 * (A[j][i+1] + A[j][i-1] +  
                                A[j-1][i] + A[j+1][i]);  
  
            err = max(err, abs(Anew[j][i] - A[j][i]));  
        }  
    }  
  
    #pragma acc parallel loop  
    for( int j = 1; j < n-1; j++) {  
        for( int i = 1; i < m-1; i++ ) {  
            A[j][i] = Anew[j][i];  
        }  
    }  
    iter++;  
}
```

Parallelize first loop nest,  
max *reduction* required.

Parallelize second loop.

We didn't detail *how* to  
parallelize the loops, just *which*  
loops to parallelize.



# BUILDING THE CODE (GPU)

```
$ pgcc -fast -ta=tesla:managed -Minfo=accel laplace2d_uvm.c  
main:
```

```
63, Generating Tesla code  
64, #pragma acc loop gang /* blockIdx.x */  
    Generating reduction(max:error)  
66, #pragma acc loop vector(128) /* threadIdx.x */  
63, Generating implicit copyin(A[:])  
    Generating implicit copyout(Anew[:])  
    Generating implicit copy(error)  
66, Loop is parallelizable  
74, Generating Tesla code  
75, #pragma acc loop gang /* blockIdx.x */  
77, #pragma acc loop vector(128) /* threadIdx.x */  
74, Generating implicit copyin(Anew[:])  
    Generating implicit copyout(A[:])  
77, Loop is parallelizable
```

# BUILDING THE CODE (MULTICORE)

```
$ pgcc -fast -ta=multicore -Minfo=accel laplace2d_uvm.c
```

```
main:
```

```
55, Generating Multicore code
```

```
56, #pragma acc loop gang
```

```
64, Accelerator restriction: size of the GPU copy of Anew,A is unknown  
Generating reduction(max:error)
```

```
66, Loop is parallelizable
```

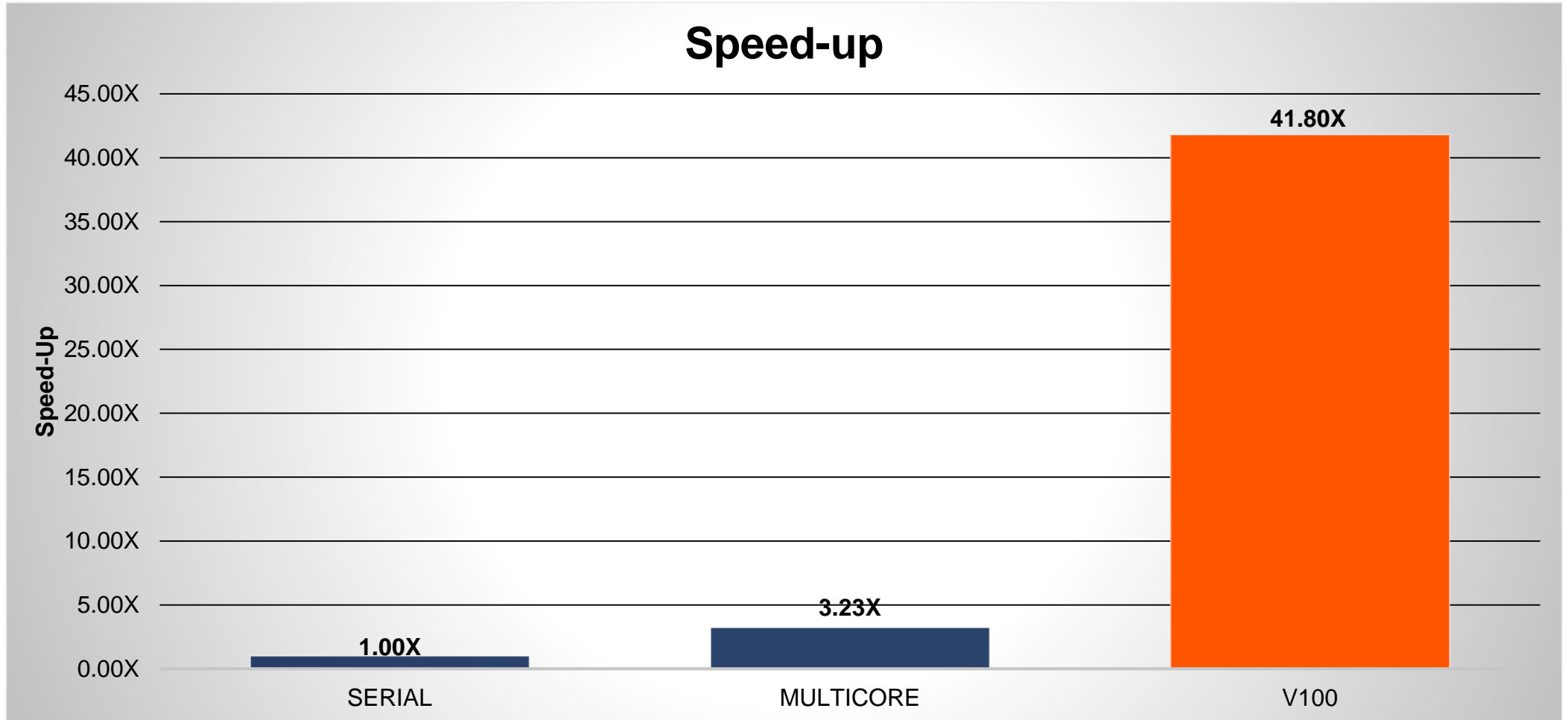
```
74, Generating Multicore code
```

```
75, #pragma acc loop gang
```

```
75, Accelerator restriction: size of the GPU copy of Anew,A is unknown
```

```
77, Loop is parallelizable
```

# OPENACC SPEED-UP



# BUILDING THE CODE (GPU)

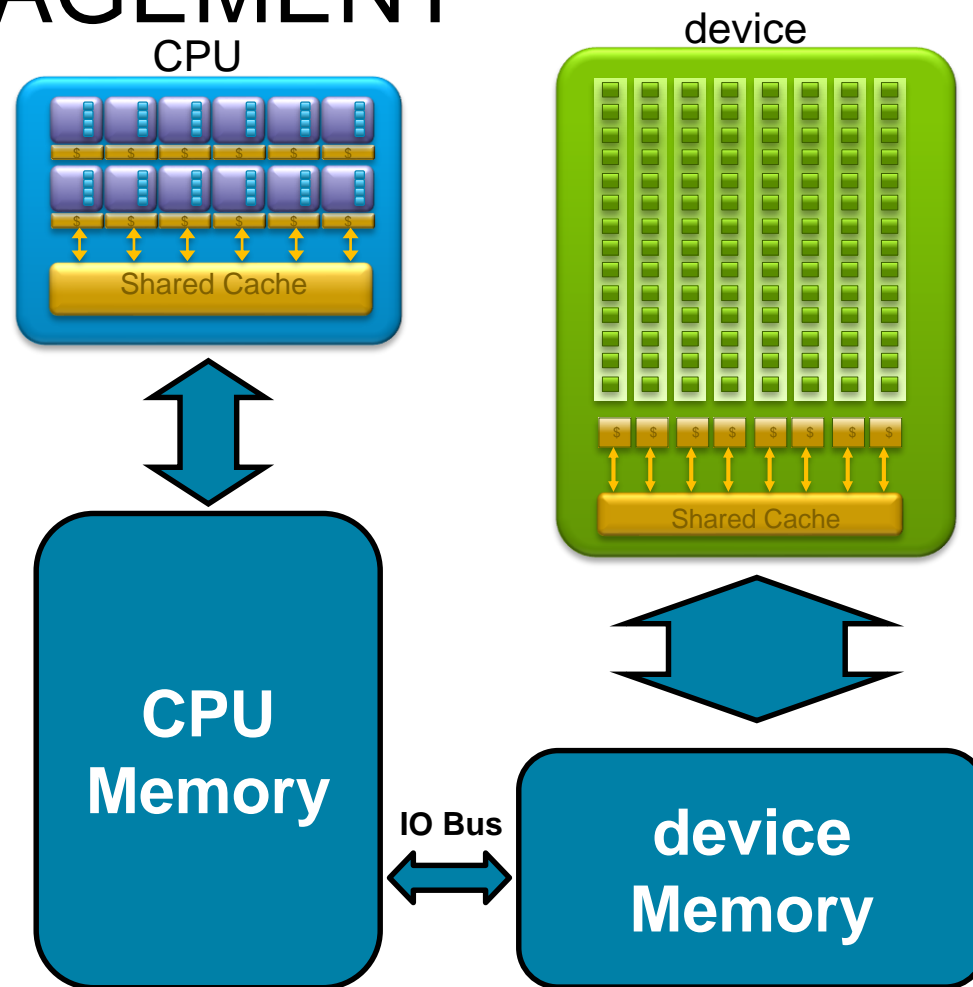
```
$ pgcc -fast -ta=tesla -Minfo=accel laplace2d_uvm.c
PGC-S-0155-Compiler failed to translate accelerator region (see -Minfo messages):
Could not find allocated-variable index for symbol (laplace2d_uvm.c: 63)
PGC-S-0155-Compiler failed to translate accelerator region (see -Minfo messages):
Could not find allocated-variable index for symbol (laplace2d_uvm.c: 74)
main:
    63, Accelerator kernel generated
        Generating Tesla code
        63, Generating reduction(max:error)
        64, #pragma acc loop gang /* blockIdx.x */
        66, #pragma acc loop vector(128) /* threadIdx.x */
    64, Accelerator restriction: size of the GPU copy of Anew,A is unknown
    66, Loop is parallelizable
    74, Accelerator kernel generated
        Generating Tesla code
        75, #pragma acc loop gang /* blockIdx.x */
        77, #pragma acc loop vector(128) /* threadIdx.x */
    75, Accelerator restriction: size of the GPU copy of Anew,A is unknown
    77, Loop is parallelizable
```

# OPTIMIZE DATA MOVEMENT

# EXPLICIT MEMORY MANAGEMENT

## Key problems

- Many parallel accelerators (such as devices) have a separate memory pool from the host
- These separate memories can become out-of-sync and contain completely different data
- Transferring between these two memories can be a very time consuming process



# OPENACC DATA DIRECTIVE

## Definition

- The data directive defines a lifetime for data on the device
- During the region data should be thought of as residing on the accelerator
- Data clauses allow the programmer to control the allocation and movement of data

```
#pragma acc data clauses  
{  
    < Sequential and/or Parallel code >  
}
```

```
!$acc data clauses  
    < Sequential and/or Parallel code >  
!$acc end data
```



# 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.**

**Principal use:** For many important data structures in your code, this is a logical default to input, modify and return the data.

`copyin( list )`

**Allocates memory on GPU and copies data from host to GPU when entering region.**

**Principal use:** Think of this like an array that you would use as just an input to a subroutine.

`copyout( list )`

**Allocates memory on GPU and copies data to the host when exiting region.**

**Principal use:** A result that isn't overwriting the input data structure.

`create( list )`

**Allocates memory on GPU but does not copy.**

**Principal use:** Temporary arrays.

# ARRAY SHAPING

- Sometimes the compiler needs help understanding the *shape* of an array
- The first number is the start index of the array
- In C/C++, the second number is how much data is to be transferred
- In Fortran, the second number is the ending index

```
copy(array[starting_index:length])
```

C/C++

```
copy(array(starting_index:ending_index))
```

Fortran

# ARRAY SHAPING (CONT.)

## Multi-dimensional Array shaping

```
copy(array[0:N][0:M])
```

C/C++

Both of these examples copy a 2D array to the device

```
copy(array(1:N, 1:M))
```

Fortran

# ARRAY SHAPING (CONT.)

## Partial Arrays

```
copy(array[i*N/4:N/4])
```

C/C++

Both of these examples copy only  $\frac{1}{4}$  of the full array

```
copy(array(i*N/4:i*N/4+N/4))
```

Fortran

# STRUCTURED DATA DIRECTIVE

## Example

```
#pragma acc data copyin(a[0:N], b[0:N]) copyout(c[0:N])
{
    #pragma acc parallel loop
    for(int i = 0; i < N; i++){
        c[i] = a[i] + b[i];
    }
}
```

Action

Device to Device  
Device to Device  
Device to Device  
Device to Device

Host Memory



Device memory



# OPTIMIZED DATA MOVEMENT

```
#pragma acc data copy(A[:n*m]) copyin(Anew[:n*m])
while ( err > tol && iter < iter_max ) {
    err=0.0;
```

```
#pragma acc parallel loop reduction(max:err)
for( int j = 1; j < n-1; j++) {
    for(int i = 1; i < m-1; i++) {

        Anew[j][i] = 0.25 * (A[j][i+1] + A[j][i-1] +
                             A[j-1][i] + A[j+1][i]);

        err = max(err, abs(Anew[j][i] - A[j][i]));
    }
}
```

```
#pragma acc parallel loop
for( int j = 1; j < n-1; j++) {
    for( int i = 1; i < m-1; i++ ) {
        A[j][i] = Anew[j][i];
    }
}
iter++;
```

Copy A to/from the accelerator only when needed.

Copy initial condition of Anew, but not final value

# REBUILD THE CODE

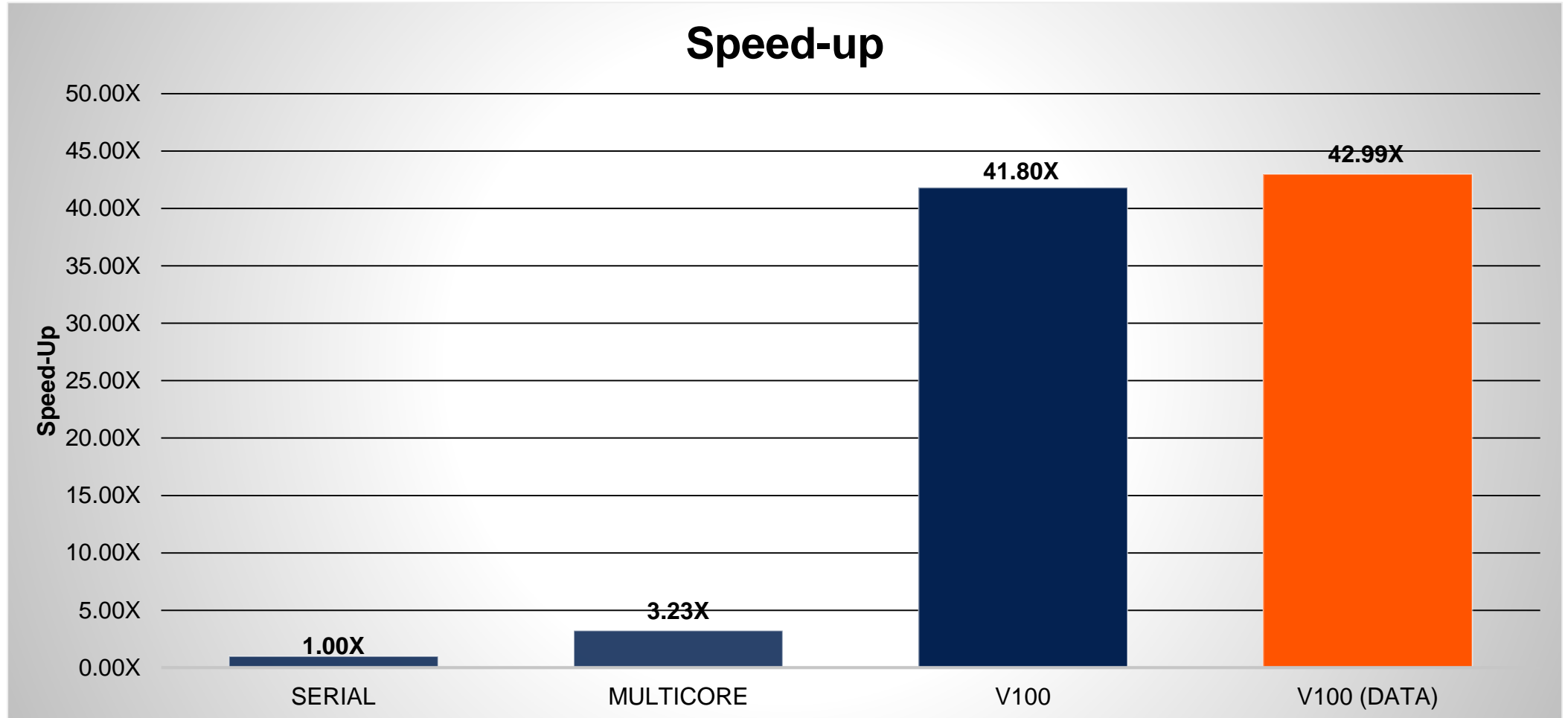
```
pgcc -fast -ta=tesla -Minfo=accel laplace2d_uvm.c  
main:
```

```
60, Generating copy(A[:m*n])  
    Generating copyin(Anew[:m*n])  
64, Accelerator kernel generated  
    Generating Tesla code  
    64, Generating reduction(max:error)  
    65, #pragma acc loop gang /* blockIdx.x */  
    67, #pragma acc loop vector(128) /* threadIdx.x */  
67, Loop is parallelizable  
75, Accelerator kernel generated  
    Generating Tesla code  
    76, #pragma acc loop gang /* blockIdx.x */  
    78, #pragma acc loop vector(128) /* threadIdx.x */  
78, Loop is parallelizable
```



Now data movement only  
happens at our data  
region.

# OPENACC SPEED-UP





# DATA SYNCHRONIZATION

# OPENACC UPDATE DIRECTIVE

**update:** Explicitly transfers data between the host and the device

Useful when you want to synchronize data in the middle of a data region

Clauses:

**self:** makes host data agree with device data

**device:** makes device data agree with host data

```
#pragma acc update self(x[0:count])  
#pragma acc update device(x[0:count])
```

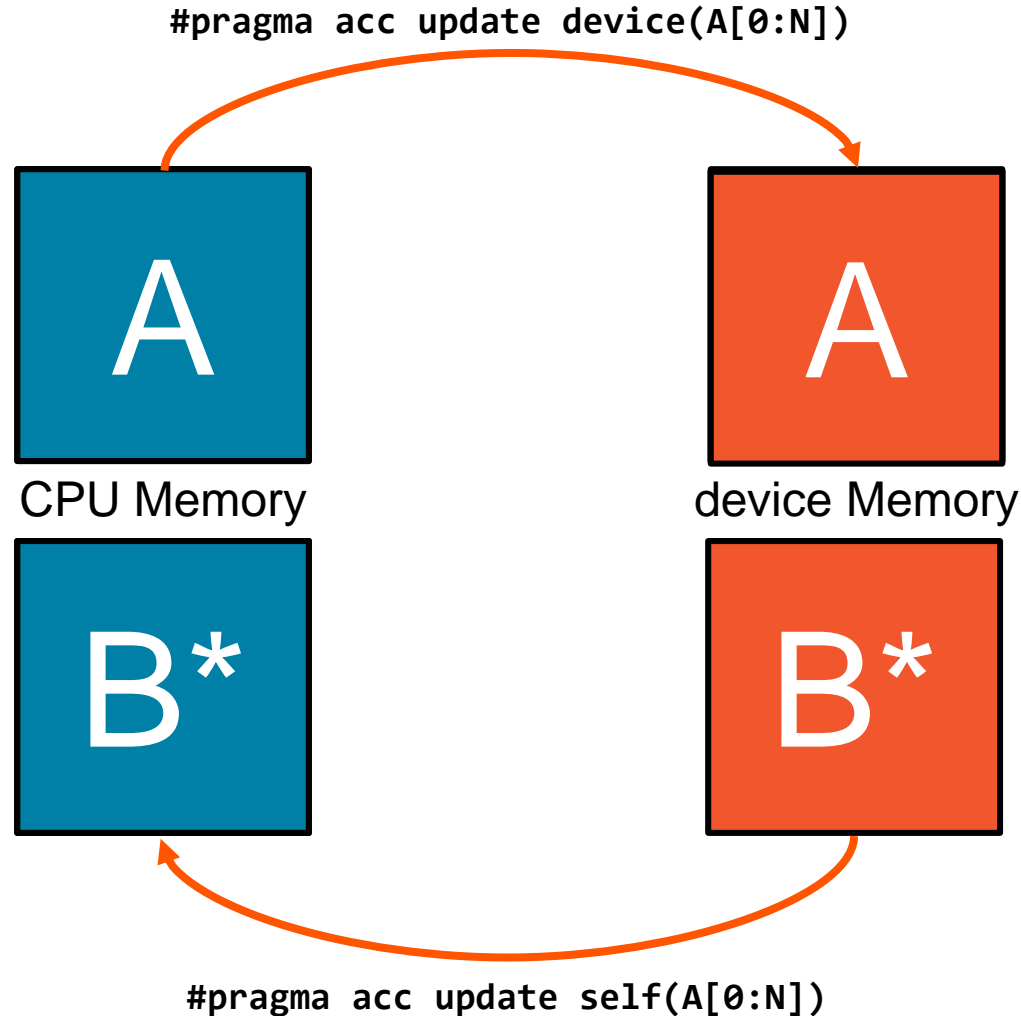
C/C++

```
!$acc update self(x(1:end_index))  
!$acc update device(x(1:end_index))
```

Fortran

# OPENACC UPDATE DIRECTIVE

The data must exist on both the CPU and device for the update directive to work.



# SYNCHRONIZE DATA WITH UPDATE

```
int* allocate_array(int N){
    int* A=(int*) malloc(N*sizeof(int));
    #pragma acc enter data create(A[0:N])
    return A;
}

void deallocate_array(int* A){
    #pragma acc exit data delete(A)
    free(A);
}

void initialize_array(int* A, int N){
    for(int i = 0; i < N; i++){
        A[i] = i;
    }
    #pragma acc update device(A[0:N])
}
```

- Inside the **initialize** function we alter the host copy of '**A**'
- This means that after calling **initialize** the host and device copy of '**A**' are out-of-sync
- We use the **update** directive with the **device** clause to update the device copy of '**A**'
- Without the **update** directive later compute regions will use incorrect data.

# LOOP OPTIMIZATIONS

# COLLAPSE CLAUSE

- **collapse( N )**
- Combine the next N tightly nested loops
- Can turn a multidimensional loop nest into a single-dimension loop
- This can be extremely useful for increasing memory locality, as well as creating larger loops to expose more parallelism

```
#pragma acc parallel loop collapse(2)
for( i = 0; i < size; i++ )
    for( j = 0; j < size; j++ )
        double tmp = 0.0f;
        #pragma acc loop reduction(+:tmp)
        for( k = 0; k < size; k++ )
            tmp += a[i][k] * b[k][j];
        c[i][j] = tmp;
```

# COLLAPSE CLAUSE

**collapse( 2 )**

(0,0)	(0,1)	(0,2)	(0,3)
(1,0)	(1,1)	(1,2)	(1,3)
(2,0)	(2,1)	(2,2)	(2,3)
(3,0)	(3,1)	(3,2)	(3,3)

```
#pragma acc parallel loop collapse(2)
for( i = 0; i < 4; i++ )
    for( j = 0; j < 4; j++ )
        array[i][j] = 0.0f;
```

# TILE CLAUSE

- **tile ( x , y , z , ... )**
- Breaks multidimensional loops into “tiles” or “blocks”
- Can increase data locality in some codes
- Will be able to execute multiple “tiles” simultaneously

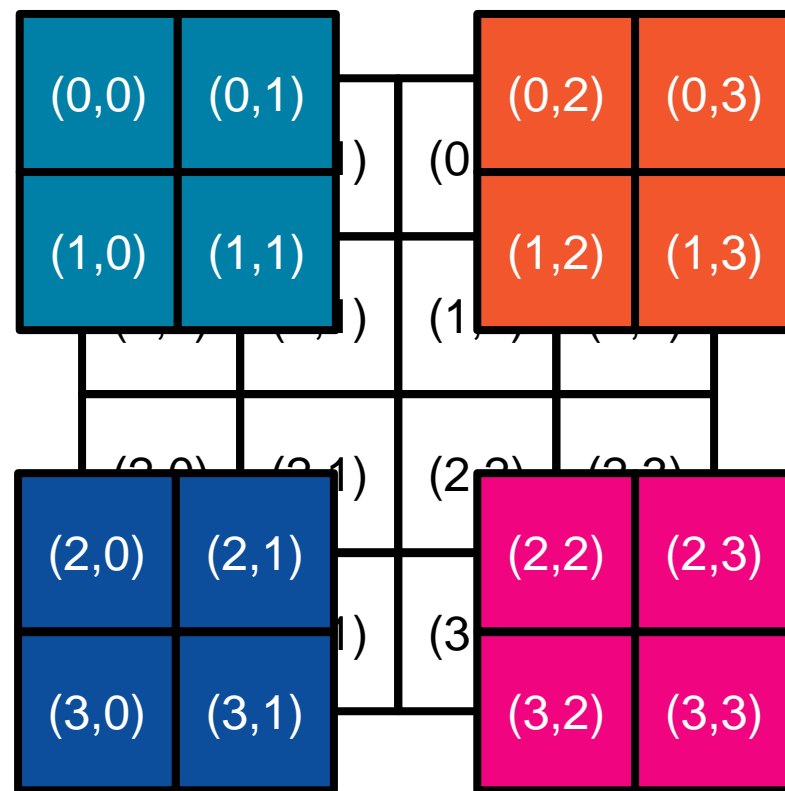
```
#pragma acc kernels loop tile(32, 32)
for( i = 0; i < size; i++ )
    for( j = 0; j < size; j++ )
        for( k = 0; k < size; k++ )
            c[i][j] += a[i][k] * b[k][j];
```



# TILE CLAUSE

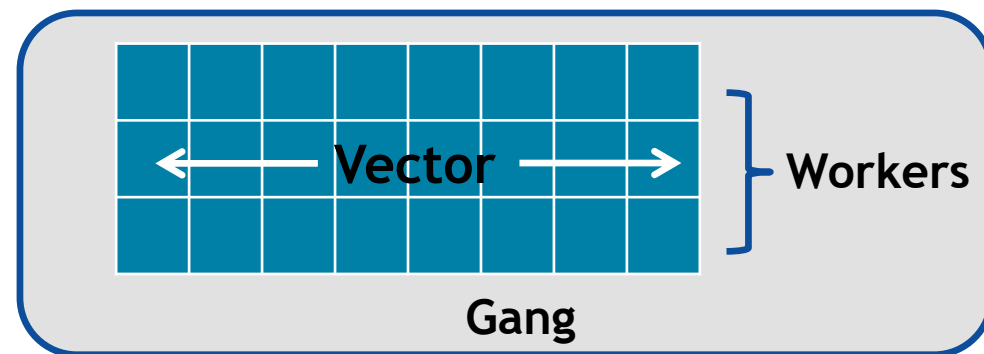
```
#pragma acc kernels loop tile(2,2)
for(int x = 0; x < 4; x++){
    for(int y = 0; y < 4; y++){
        array[x][y]++;
    }
}
```

tile ( 2 , 2 )



# GANG WORKER VECTOR

- Gang / Worker / Vector defines the various levels of parallelism we can achieve with OpenACC
- This parallelism is most useful when parallelizing multi-dimensional loop nests
- OpenACC allows us to define a generic Gang / Worker / Vector model that will be applicable to a variety of hardware, but we will focus a little bit on a GPU specific implementation



# OPTIMIZED LOOP

```
#pragma acc data copy(A[:n*m]) copyin(Anew[:n*m])
while ( err > tol && iter < iter_max ) {
    err=0.0;

    #pragma acc parallel loop reduction(max:err) tile(32,32)
    for( int j = 1; j < n-1; j++) {
        for(int i = 1; i < m-1; i++) {

            Anew[j][i] = 0.25 * (A[j][i+1] + A[j][i-1] +
                                A[j-1][i] + A[j+1][i]);

            err = max(err, abs(Anew[j][i] - A[j][i]));
        }
    }

    #pragma acc parallel loop tile(32,32)
    for( int j = 1; j < n-1; j++) {
        for( int i = 1; i < m-1; i++ ) {
            A[j][i] = Anew[j][i];
        }
    }
    iter++;
}
```

Create 32x32 tiles of the loops to better exploit data locality.

# REBUILD THE CODE

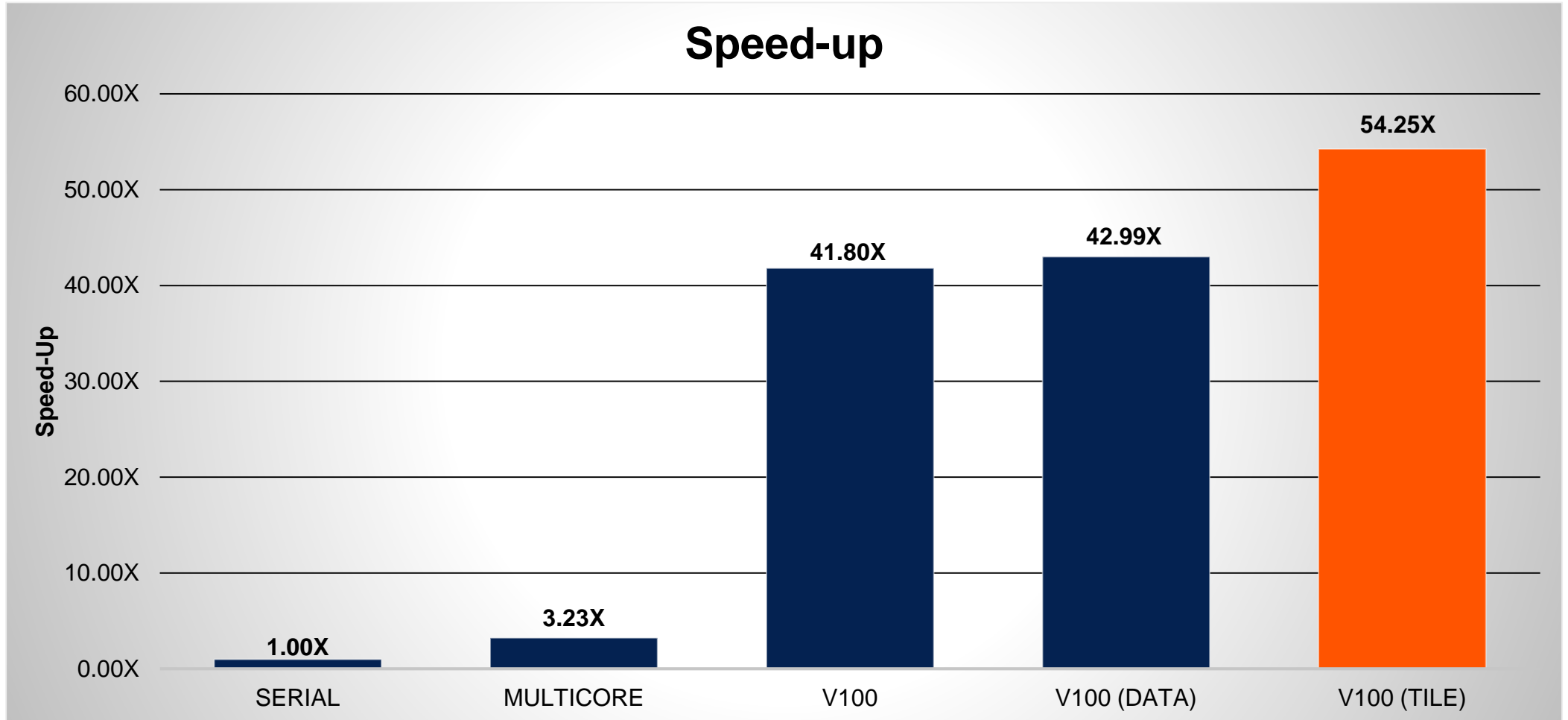
```
pgcc -fast -ta=tesla -Minfo=accel laplace2d_uvm.c  
main:
```

```
60, Generating copy(A[:m*n])  
    Generating copyin(Anew[:m*n])  
64, Accelerator kernel generated  
    Generating Tesla code  
    64, Generating reduction(max:error)  
    65, #pragma acc loop gang /* blockIdx.x */  
    67, #pragma acc loop vector(128) /* threadIdx.x */  
67, Loop is parallelizable  
75, Accelerator kernel generated  
    Generating Tesla code  
    76, #pragma acc loop gang /* blockIdx.x */  
    78, #pragma acc loop vector(128) /* threadIdx.x */  
78, Loop is parallelizable
```



Now data movement only  
happens at our data  
region.

# OPENACC SPEED-UP



# LOOP OPTIMIZATION RULES OF THUMB

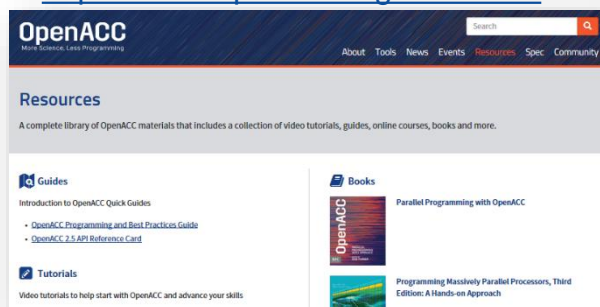
- It is rarely a good idea to set the number of gangs in your code, let the compiler decide.
- Most of the time you can effectively tune a loop nest by adjusting only the vector length.
- It is rare to use a worker loop. When the vector length is very short, a worker loop can increase the parallelism in your gang.
- When possible, the vector loop should step through your arrays
- Use the `device_type` clause to ensure that tuning for one architecture doesn't negatively affect other architectures.

# OPENACC RESOURCES

Guides • Talks • Tutorials • Videos • Books • Spec • Code Samples • Teaching Materials • Events • Success Stories • Courses • Slack • Stack Overflow

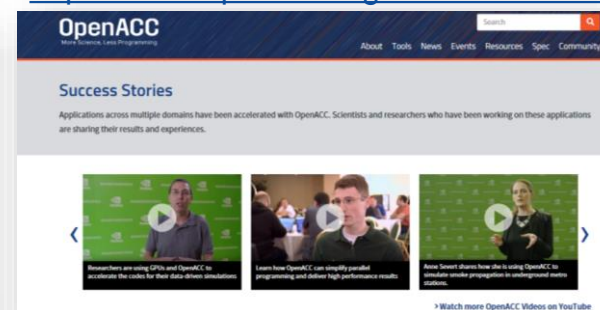
## Resources

<https://www.openacc.org/resources>



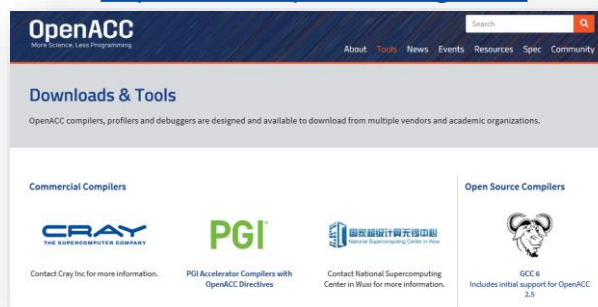
## Success Stories

<https://www.openacc.org/success-stories>



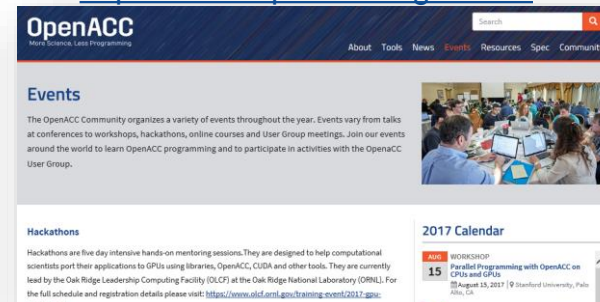
## Compilers and Tools

<https://www.openacc.org/tools>



## Events

<https://www.openacc.org/events>



# CLOSING REMARKS



# KEY CONCEPTS

In this lecture we discussed...

- How to profile a serial code to identify loops that should be accelerated
- How to use OpenACC's parallel loop directive to parallelize key loops
- How to use OpenACC's data clauses to control data movement
- How to optimize loops in the code for better performance