

UC Berkeley 2012 Short Course on Parallel Programming

# Parallel Programming on Windows and Porting CS267

Matej Ciesko

Technology Policy Group (TPG)

Microsoft

# Agenda

- Overview of parallel programming landscape
- C++ AMP
- CS267 port to Windows

Overview to

# **PARALLEL COMPUTING LANDSCAPE**

## Welcome to the jungle

The free lunch  
is so over

1975-2005

Put a **computer**  
on every desk, in  
every home, in  
every pocket.

2005-2011

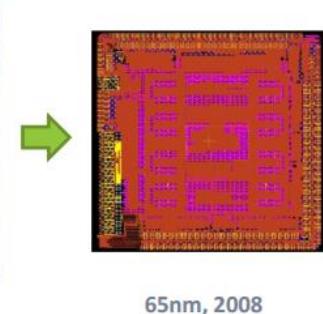
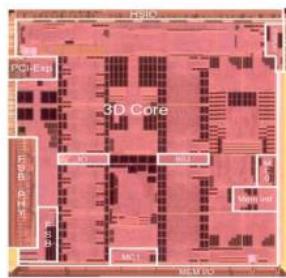
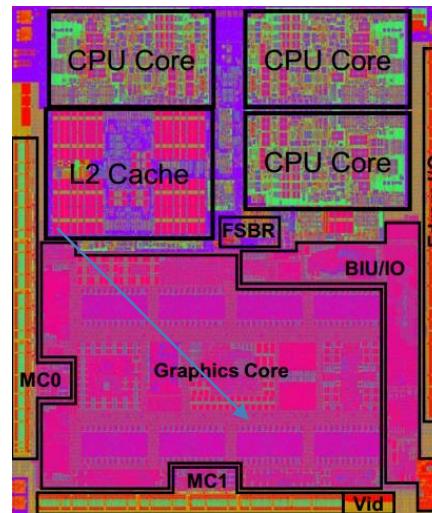
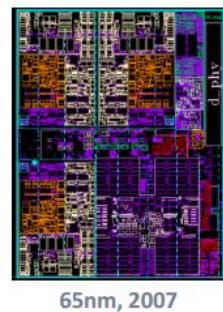
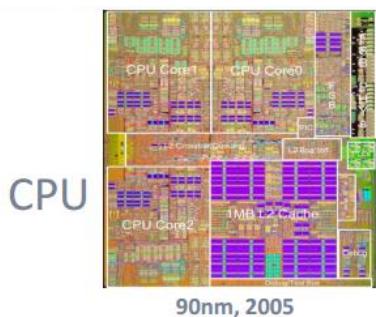
Put a **parallel  
supercomputer**  
on every desk, in  
every home, in  
every pocket.

2011-201x

Put a **massively parallel  
heterogeneous super-  
computer** on every  
desk, in every home,  
in every pocket.

Herb Sutter – “Welcome to the jungle”  
David Callahan - AMP

# Xbox360



**commercially available  
to millions**

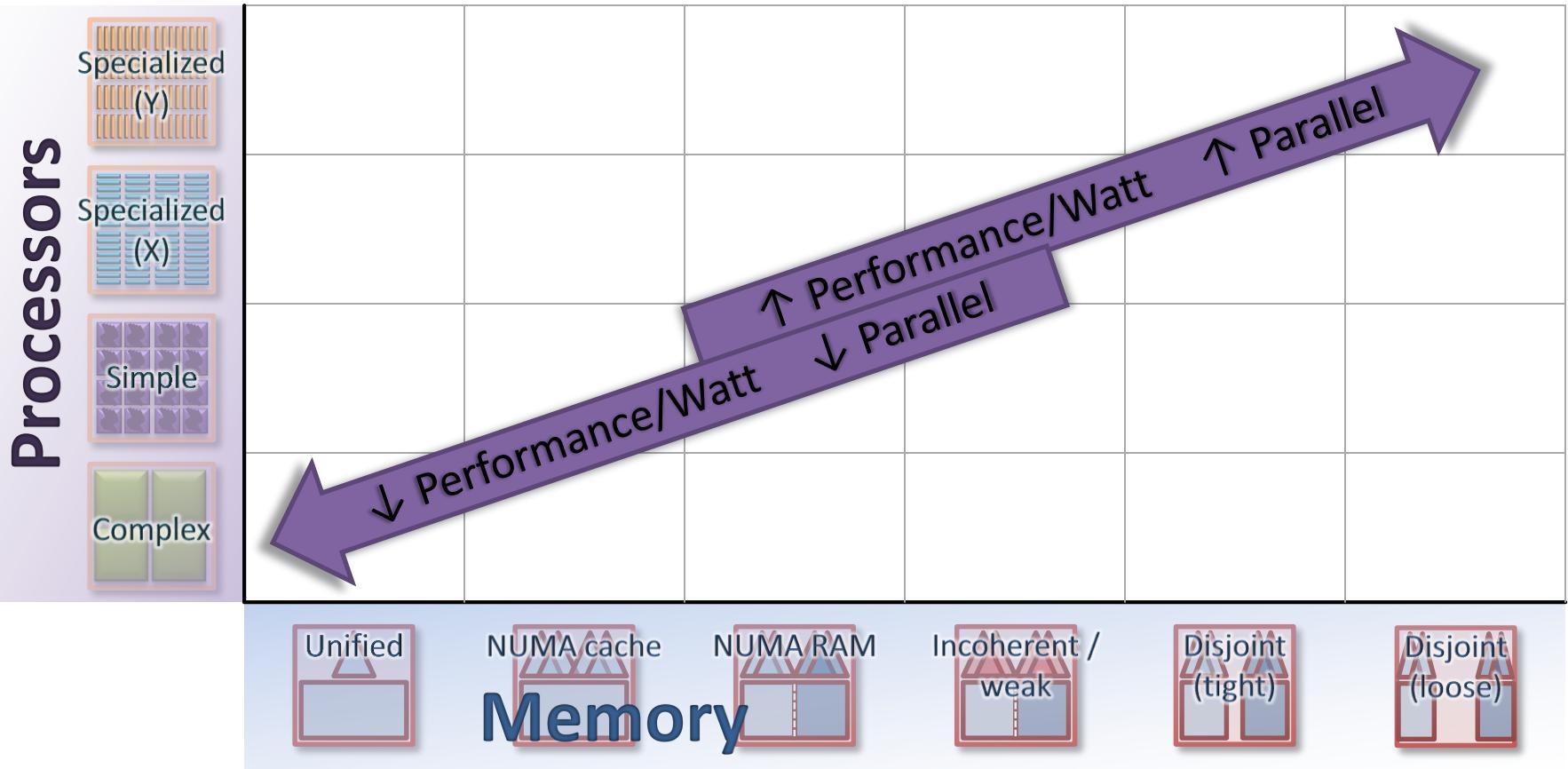
**commercially affordable  
for millions**

**commercially programmable  
by millions**

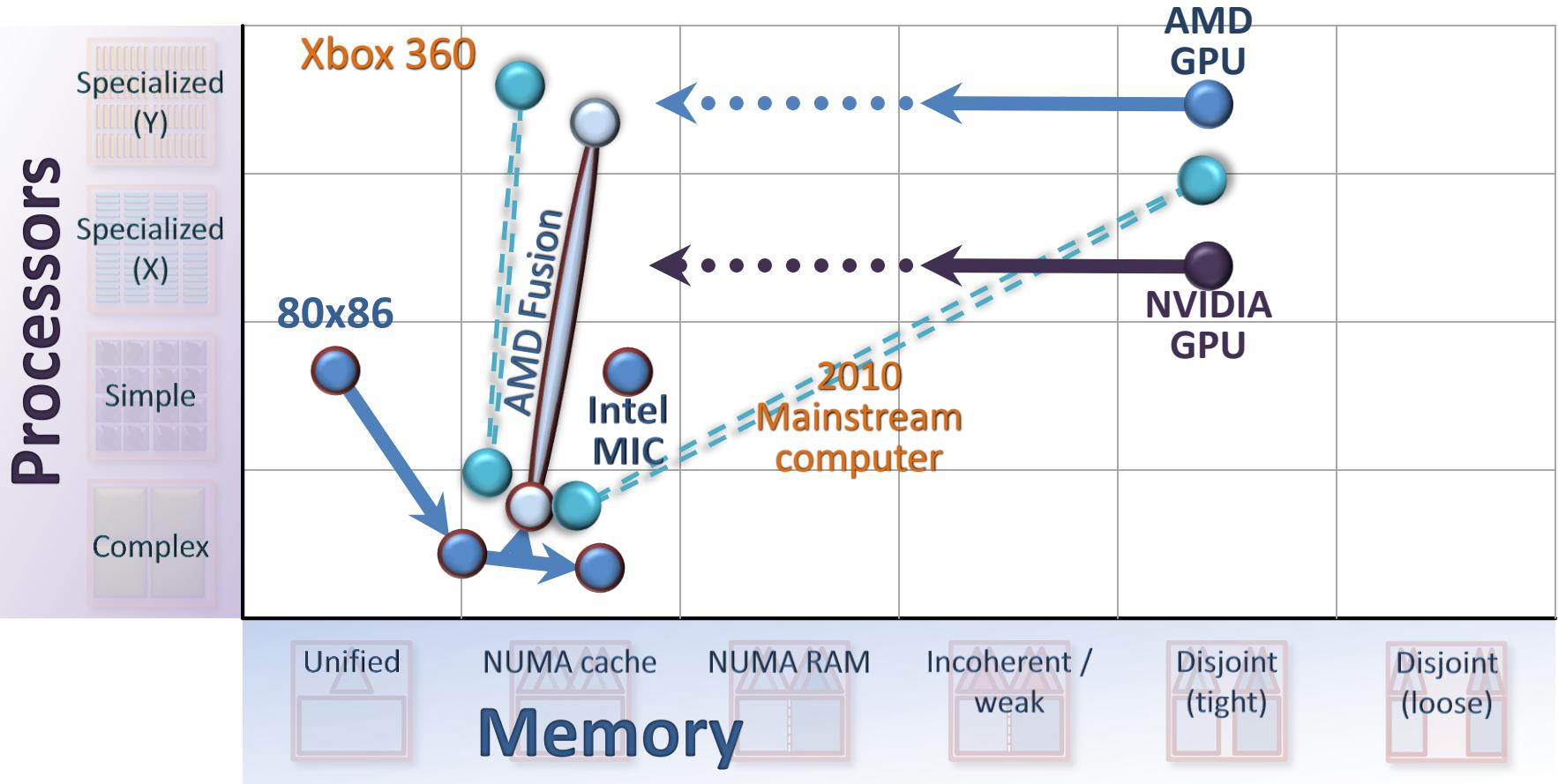
*note: everything in “the mainstream” starts out in “the exotic”*

*GUIs ✓ objects ✓ parallelism ✓*

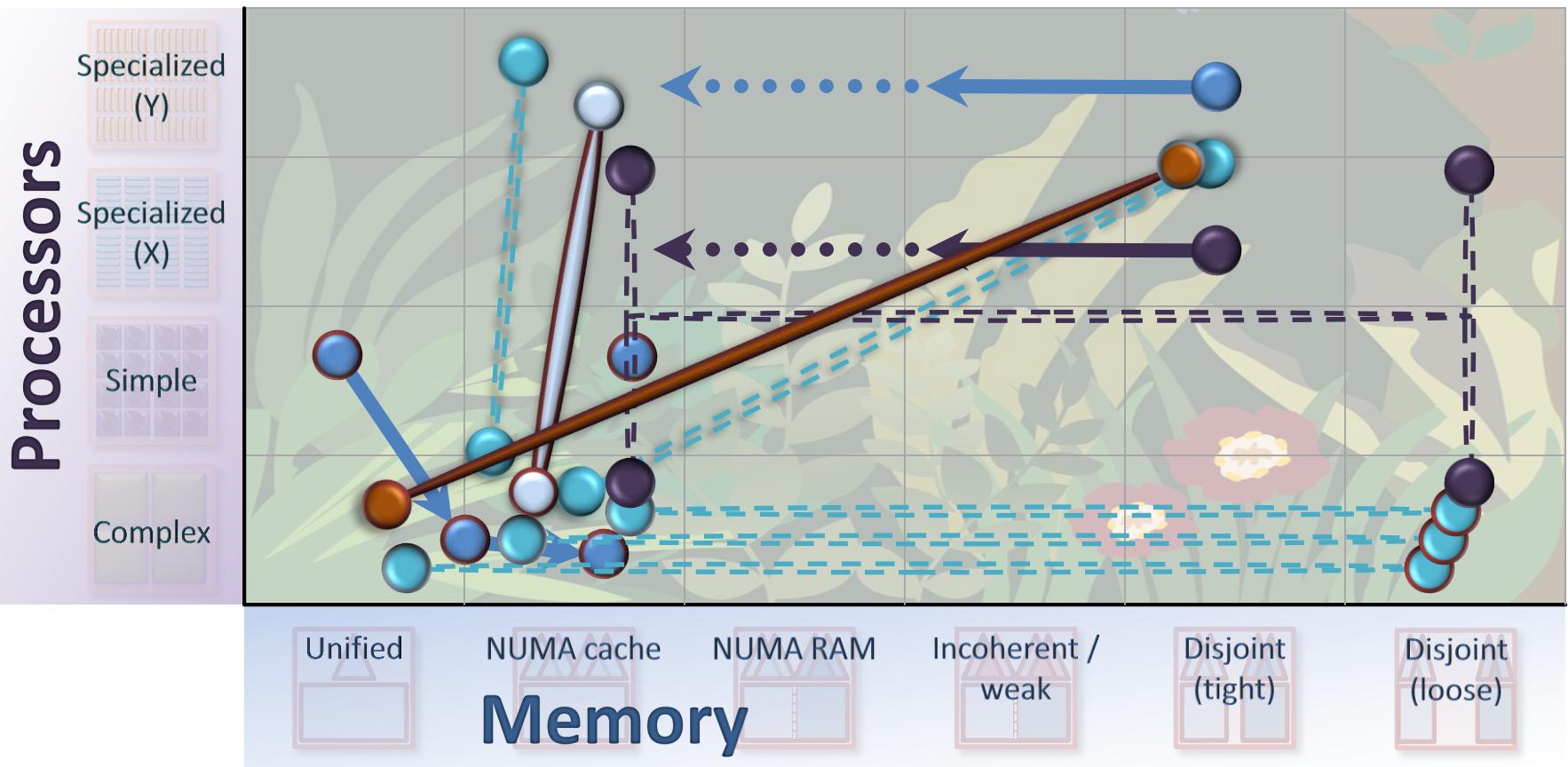
# Charting the Landscape



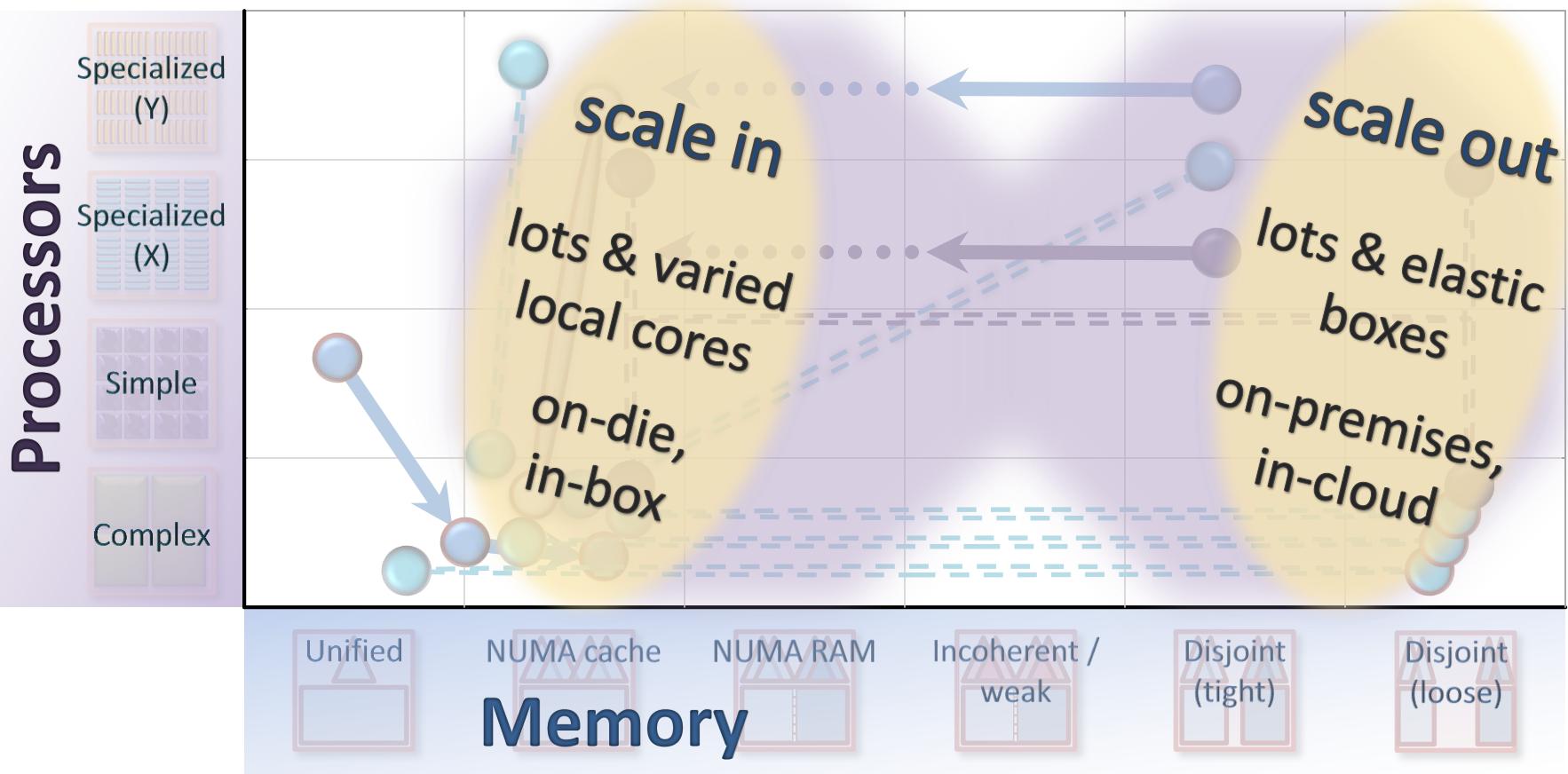
# Hardware



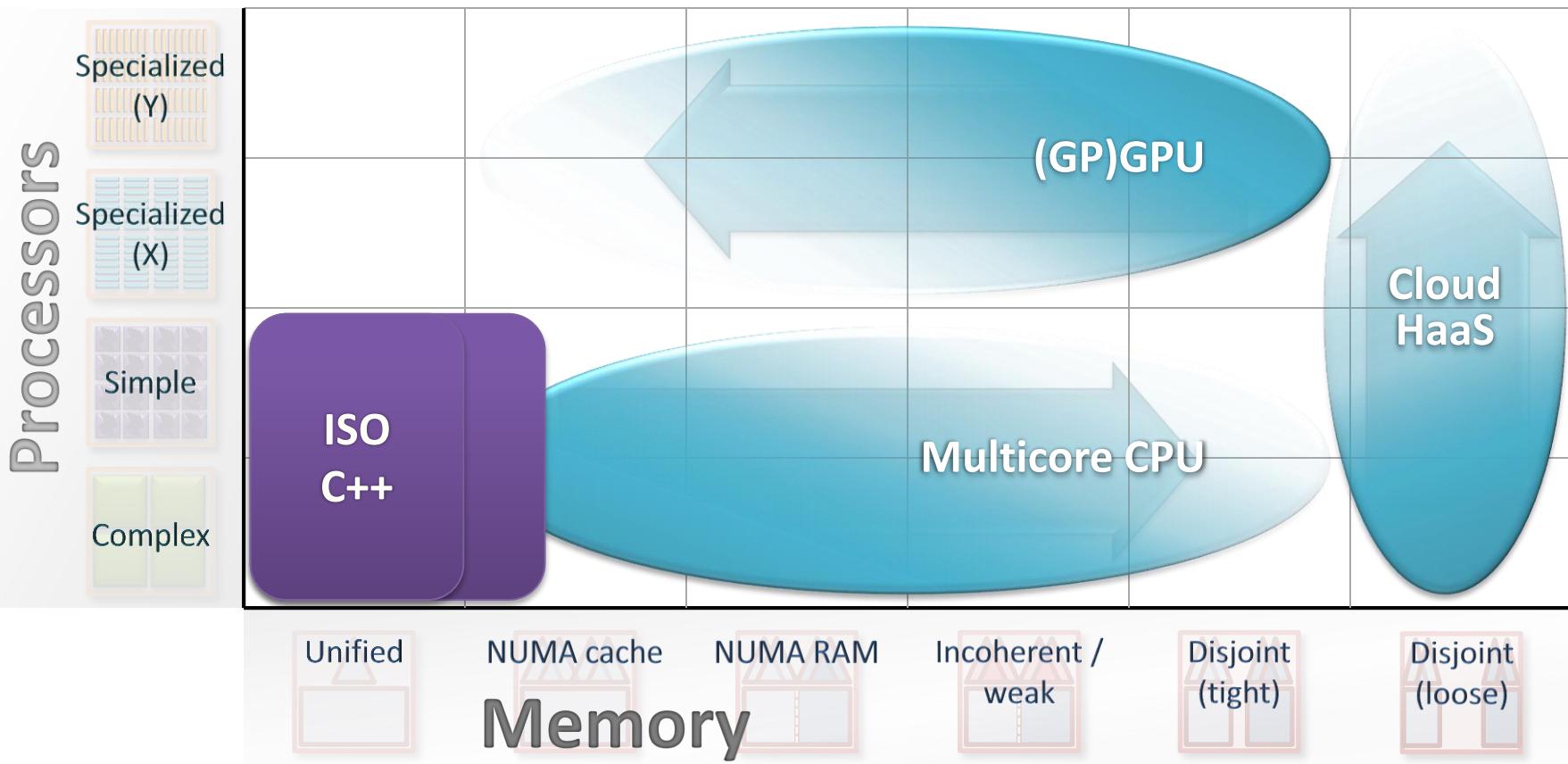
# The Jungle



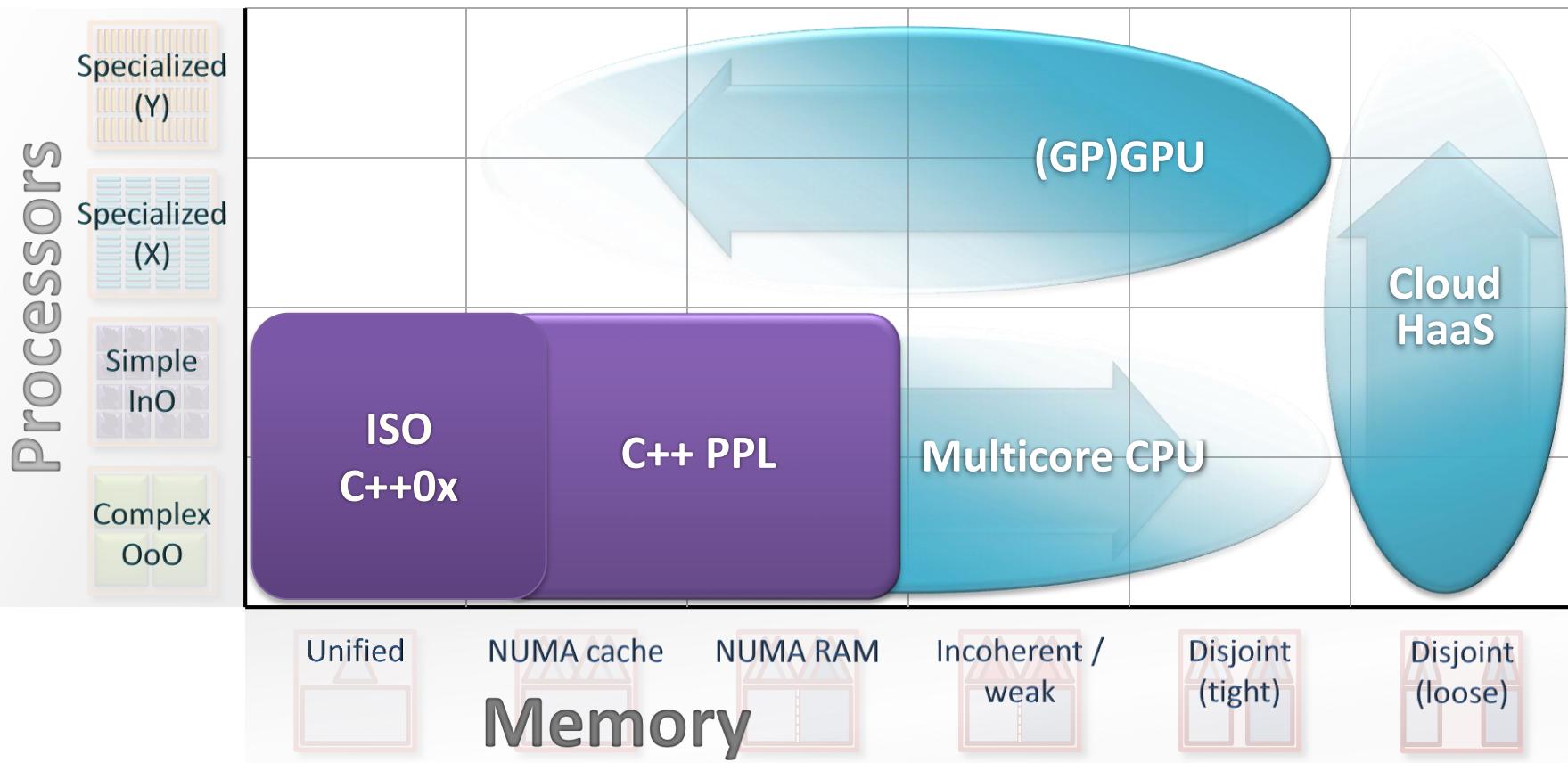
# Hardware Evolution



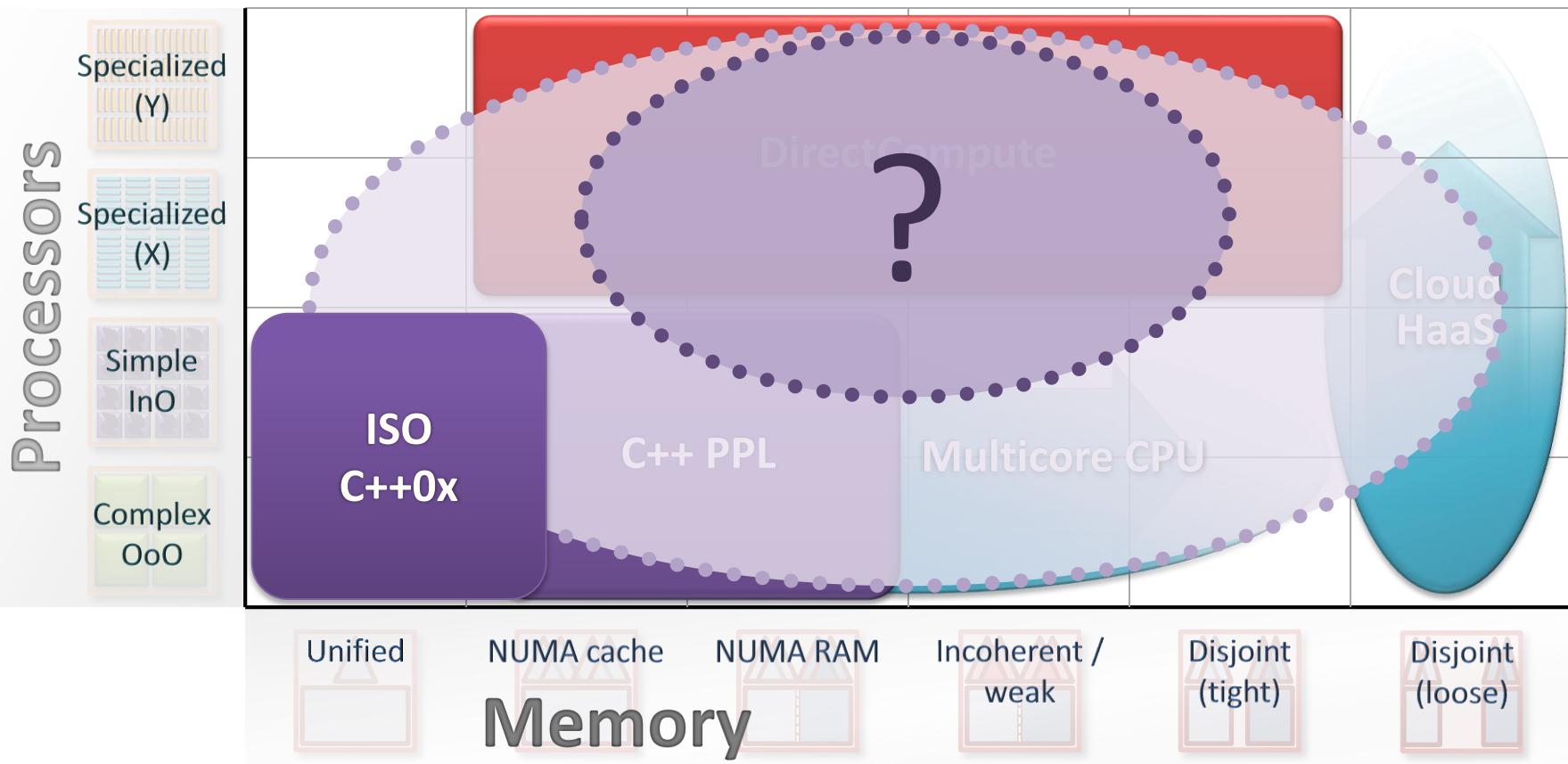
# Programming Models & Languages



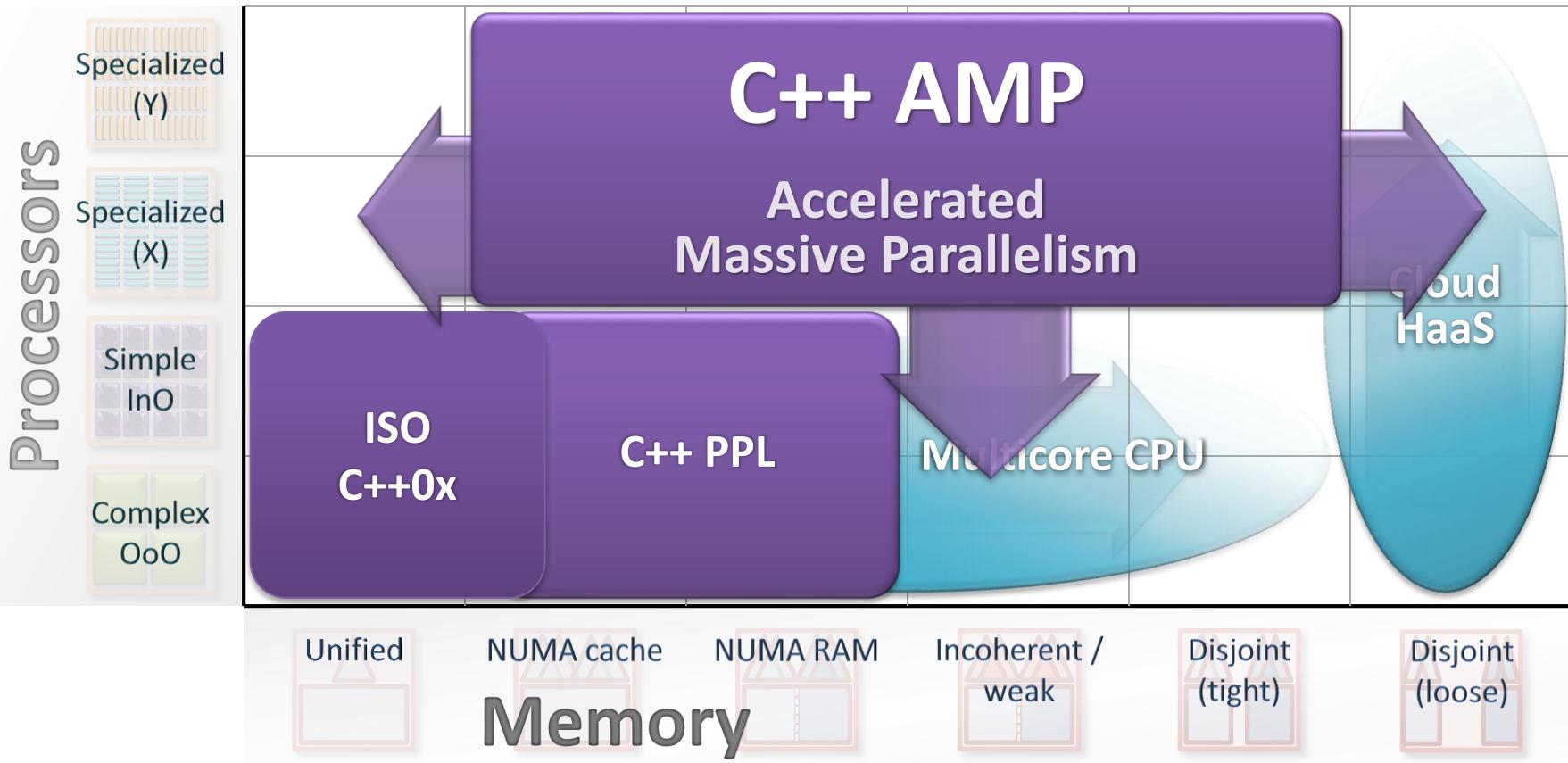
# Programming Models & Languages



# Programming Models & Languages



# Programming Models & Languages



# Matrix Multiply

Convert this (serial loop nest)

```
void MatrixMult( float* C, const vector<float>& A, const vector<float>& B,
                 int M, int N, int W )
{
    for (int y = 0; y < M; y++)
        for (int x = 0; x < N; x++) {
            float sum = 0;
            for(int i = 0; i < W; i++)
                sum += A[y*W + i] * B[i*N + x];
            C[y*N + x] = sum;
        }
}
```

# Matrix Multiply

Convert this (serial loop nest)

void Mat

... to this (parallel loop, CPU or GPU)

```
void MatrixMult( float* C, const vector<float>& A, const vector<float>& B,
                  int M, int N, int W )
{
    for (int i = 0; i < M; i++) {
        for (int j = 0; j < N; j++) {
            float sum = 0;
            for (int k = 0; k < W; k++) {
                sum += A[i * W + k] * B[k * N + j];
            }
            C[i * N + j] = sum;
        }
    }
}
```

# Why C++ AMP?

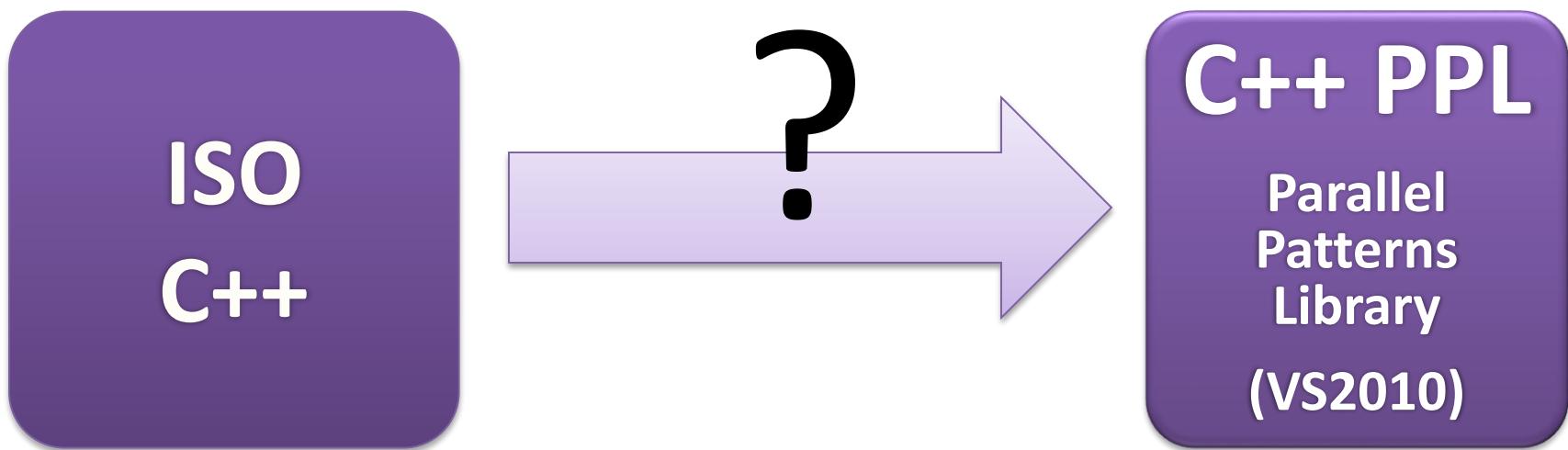
## Processors



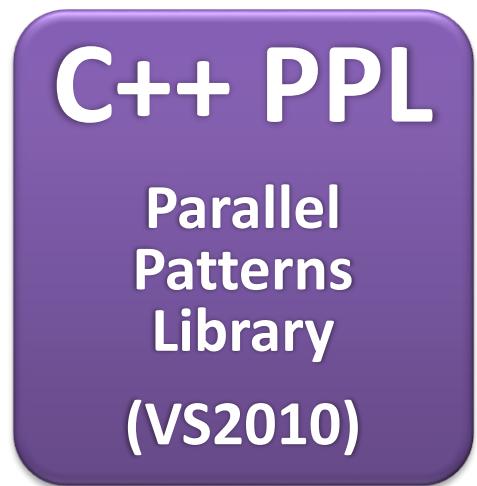
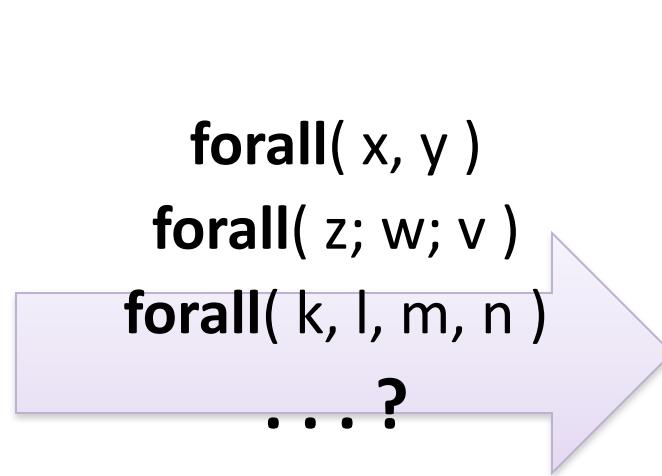
C++, not C					
<b>mainstream</b> , programmable by millions					
<b>minimal</b> , just one general language extension					
<b>portable</b> , mix & match hardware from any vendor, one EXE					
<b>general and future-proof</b> , designed to cover the full range of hardware heterogeneity – hardware is still in motion					



# Language Design: Parallelism Phase 1

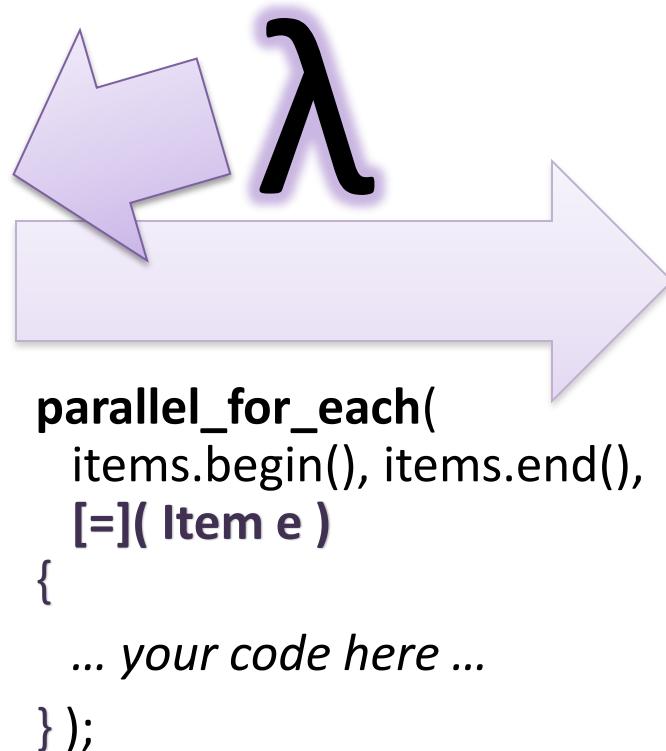


# Language Design: Parallelism Phase 1



# Language Design: Parallelism Phase 1

ISO  
C++0x



*Single-core to multi-core*

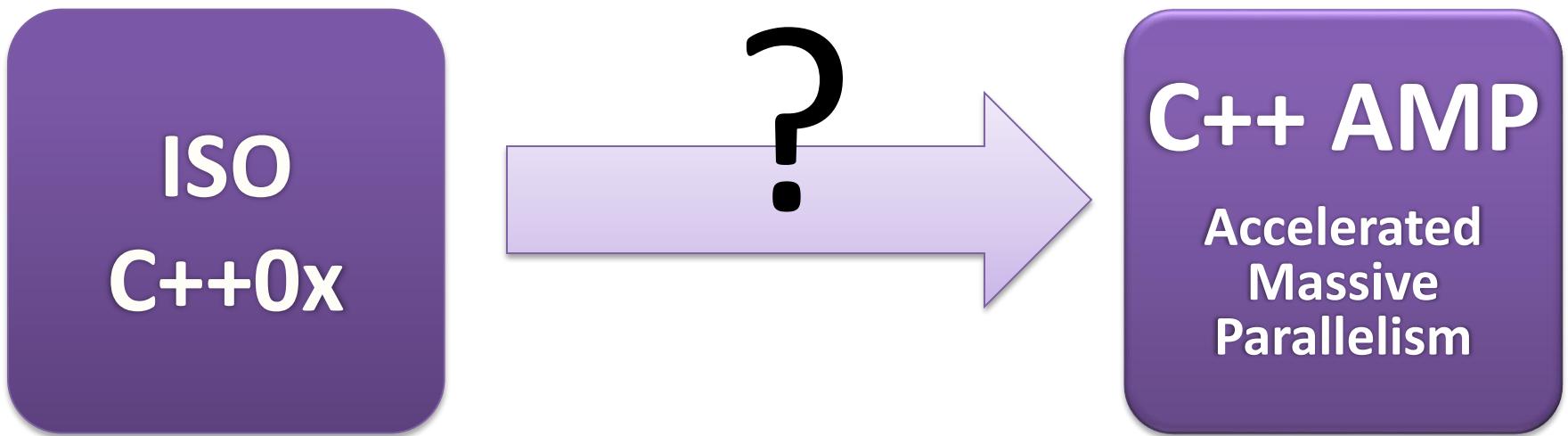
C++ PPL  
Parallel  
Patterns  
Library  
(VS2010)

# 1

language feature for multicore

and STL, functors, callbacks, events, ...

# Language Design: Parallelism Phase 2



# Language Design: Parallelism Phase 2

*Multi-core to hetero-core*

ISO  
C++0x

**restrict**

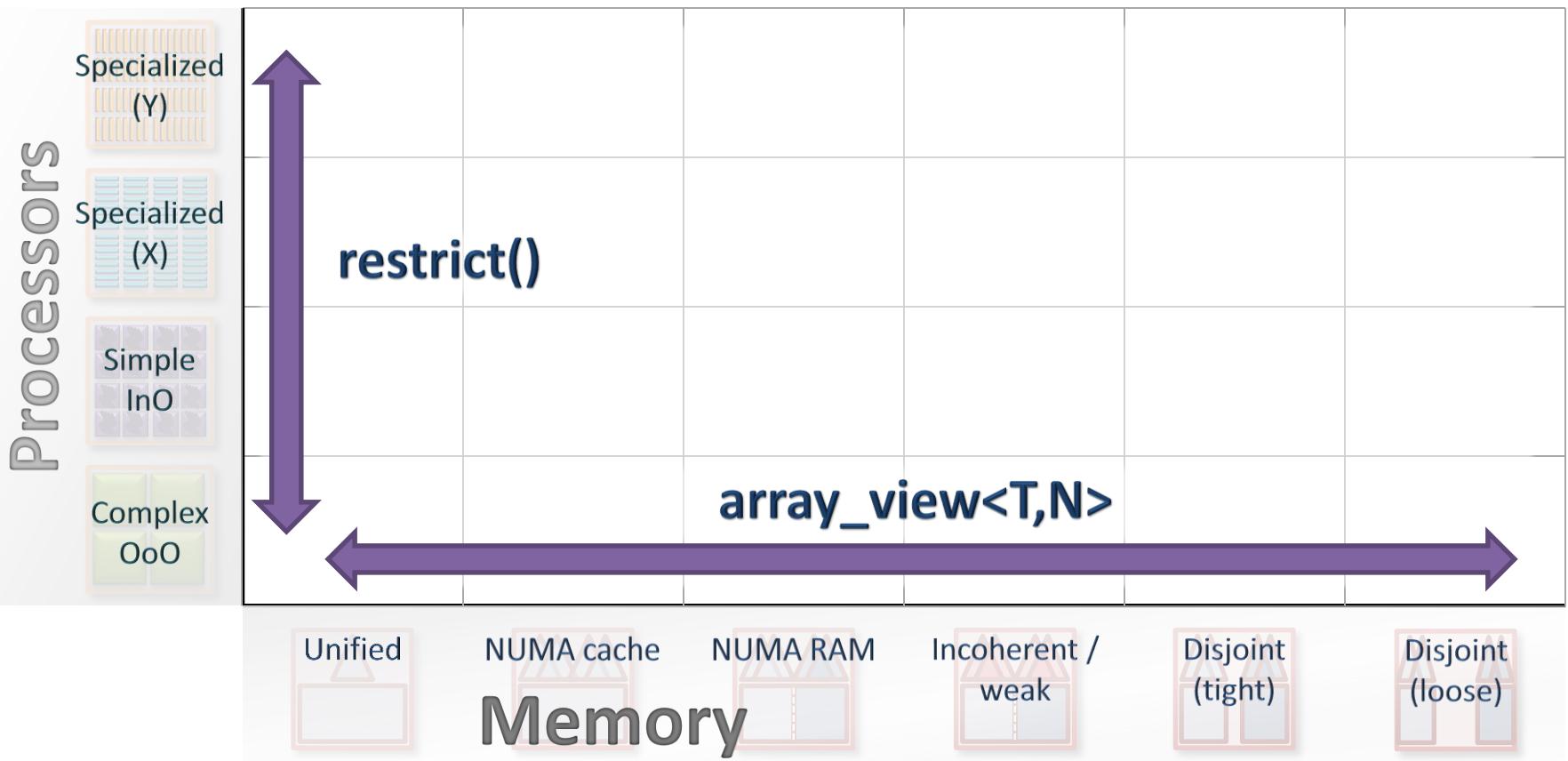
```
parallel_for_each(  
    items.grid,  
    [=](index<2> i) restrict(direct3d)  
{  
    ... your code here ...  
};
```

C++ AMP  
**Accelerated  
Massive  
Parallelism**

# 1

language feature for  
heterogeneous cores

# C++ AMP at a Glance



# restrict()

- **Problem:** Some cores don't support the entire C++ language.
- **Solution:** General restriction qualifiers enable expressing language subsets within the language. Direct3d math functions in the box.

## Example

```
double sin( double ) restrict(cpu,direct3d);      // 1: same code for either
double cos( double );                            // 2a: general code
double cos( double ) restrict(direct3d);          // 2b: specific code

parallel_for_each( c.grid, [=](index<2> idx) restrict(direct3d) {
    ...
    sin( data.angle ); // ok
    cos( data.angle ); // ok, chooses overload based on context
    ...
});
```

# **restrict()**

- Initially supported restriction qualifiers:
  - **restrict(cpu)**: The implicit default.
  - **restrict(direct3d)**: Can execute on any DX11 device via DirectCompute.
    - Restrictions follow limitations of DX11 device model (e.g., no function pointers, virtual calls, goto).
- Potential future directions:
  - **restrict(pure)**: Declare and enforce a function has no side effects. Great to be able to state declaratively for parallelism.
  - General facility for language subsets, not just about compute targets.

# Functionally Restricted Processors

```
int myFunction( int x, int y) restrict(amp)
{
    // only call similarly restricted functions
    // no partial word data types
    // no exceptions, try or catch
    // no goto
    // no indirect function calls
    // no varargs routines
    // no new/delete
    // very restricted pointer usage
    // no static variables
}
```

// overload resolution

```
int myFunction(int x, int y) restrict(amp);
int myFunction(int x, int y) restrict(cpu);
int myFunc(int x) restrict(cpu,amp);
```

// evolution over time

```
int myFunction(int x, int y) restrict(amp:2);
```

# array\_view

- **Problem:** Memory may be flat, nonuniform, incoherent, *and/or* disjoint.
- **Solution:** Portable view that works like an N-dimensional “iterator range.”
  - Future-proof: No explicit `.copy()`/`.sync()`. As needed by each actual device.

## Example

```
void MatrixMult( float* C, const vector<float>& A, const vector<float>& B,
                 int M, int N, int W )
{
    array_view<const float,2> a(M,W,A), b(W,N,B); // 2D view over C++ std::vector
    array_view<writeonly<float>,2> c(M,N,C);      // 2D view over C array
    parallel_for_each( c.grid, [=](index<2> idx) restrict(direct3d) {
        ...
    } );
}
```

# Example Matrix Multiplication

```
void MatrixMultiplySerial( vector<float>& vC,
    const vector<float>& vA,
    const vector<float>& vB, int M, int N, int W )
{
    for (int row = 0; row < M; row++) {
        for (int col = 0; col < N; col++){
            float sum = 0.0f;
            for(int i = 0; i < W; i++)
                sum += vA[row * W + i] * vB[i * N + col];
            vC[row * N + col] = sum;
        }
    }
}
```

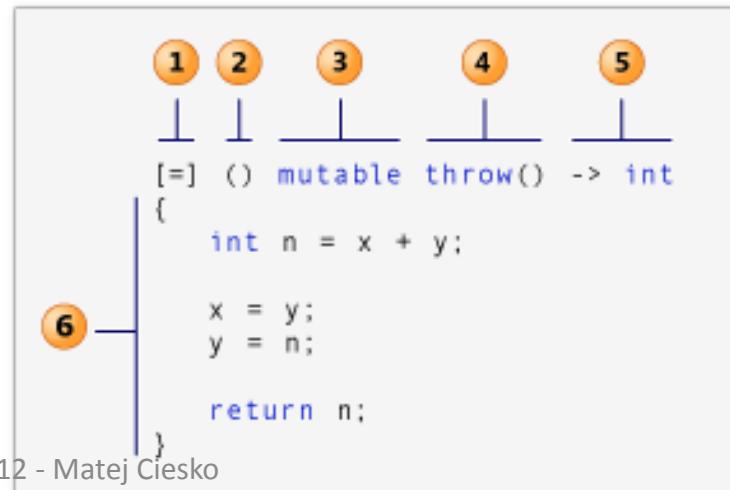
```
void MatrixMultiplyAMP( vector<float>& vC,
    const vector<float>& vA,
    const vector<float>& vB, int M, int N, int W )
{
    array_view<const float,2> a(M,W,vA),b(W,N,vB);
    array_view<float,2> c(M,N,vC);
    c.discard_data();
    parallel_for_each(c.extent,
        [=](index<2> idx) restrict(amp) {
            int row = idx[0]; int col = idx[1];
            float sum = 0.0f;
            for(int i = 0; i < W; i++)
                sum += a(row, i) * b(i, col);
            c[idx] = sum;
        });
    c.synchronize();
}
```

# Disjoint Address Space, One Name Space

- `array_view<>` enables copy-as-needed semantics
- Build on C++ lambda capture rules
- Explicit allocation via `array<>` & accelerators
- → Future-proofing for shared memory architectures

```
array_view<const float,2> a(M,W,vA), b(W,N,vB);
array_view<float,2> c(M,N,vC);

parallel_for_each(c.extent,
    [=](index<2> idx) restrict(amp) {
        int row = idx[0]; int col = idx[1];
        float sum = 0.0f;
        for(int i = 0; i < W; i++)
            sum += a(row, i) * b(i, col);
        c[idx] = sum;
    }
);
```



# Explicit Caching

- Embrace a multi-core with private caches
- Extend data-parallel model with explicit tiling
- Within tiles, barrier coordination + shared storage

Key uses:

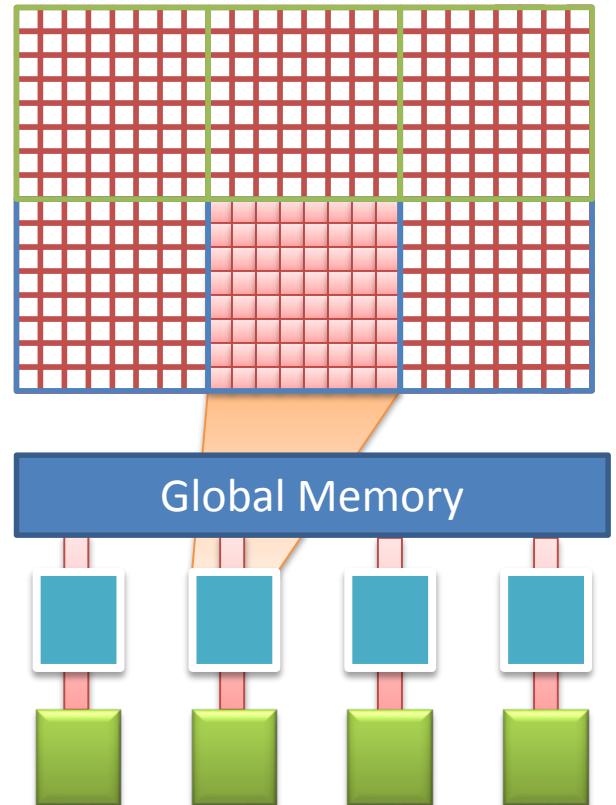
Capture cross-thread data reuse

Fast communication

    Reductions, Scans

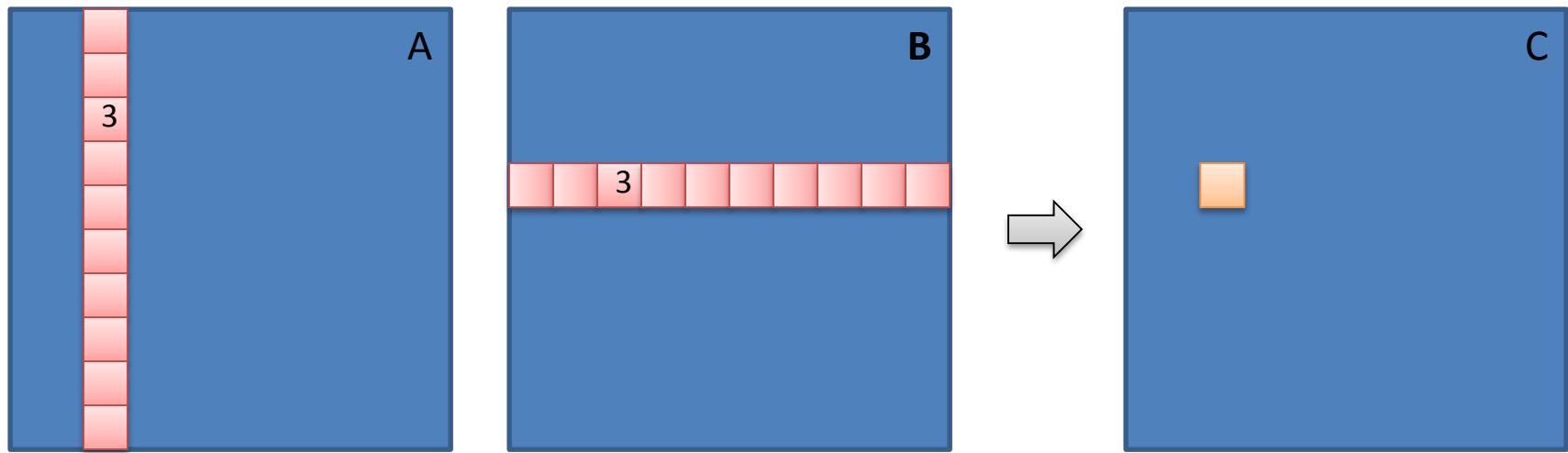
    Nearest-neighbor

Caches  
Vector  
Processors



# Improve Matrix Multiply with tiling

In block form, each block-level multiply-accumulate can be done in cache



# Explicit Caching: Tiles

```
void MatrixMultSimple(vector<float>& vC, const
vector<float>& vA, const vector<float>& vB, int M, int N,
int W )
{
    array_view<const float,2> a(M, W, vA), b(W, N, vB);
    array_view<float,2> c(M,N,vC); c.discard_data();
    parallel_for_each(c.extent,
        [=] (index<2> idx) restrict(amp)
    {
        int row = idx[0];
        int col = idx[1];

        float sum = 0.0f;
        for(int k = 0; k < W; k++)
            sum += a(row, k) * b(k, col);

        c[idx] = sum;
    });
}
```

```
void MatrixMultTiled(vector<float>& vC, const
vector<float>& vA, const vector<float>& vB, int M, int N,
int W )
{
    static const int TS = 16;
    array_view<const float,2> a(M, W, vA), b(W, N, vB);
    array_view<float,2> c(M,N,vC); c.discard_data();
    parallel_for_each(c.extent.tile< TS, TS >(),
        [=] (tiled_index< TS, TS > t_idx) restrict(amp)
    {
        int row = t_idx.global[0];
        int col = t_idx.global[1];

        float sum = 0.0f;
        for(int k = 0; k < W; k++)
            sum += a(row, k) * b(k, col);

        c[t_idx.global] = sum;
    });
}
```

# Explicit Caching: Coordination

```
void MatrixMultTiled(vector<float>& vC, const vector<float>& vA,  
const vector<float>& vB, int M, int N, int W )  
{  
    static const int TS = 16;  
    array_view<const float,2> a(M, W, vA), b(W, N, vB);  
    array_view<float,2> c(M,N,vC); c.discard_data();  
    parallel_for_each(c.extent.tile< TS, TS >(),  
        [=] (tiled_index< TS, TS > t_idx) restrict(amp) {  
  
        int row = t_idx.global[0]; int col = t_idx.global[1];  
        float sum = 0.0f;  
  
        for(int k = 0; k < W; k++)  
            sum += a(row, k) * b(k, col);  
  
        c[t_idx.global] = sum;  
    } );  
}
```

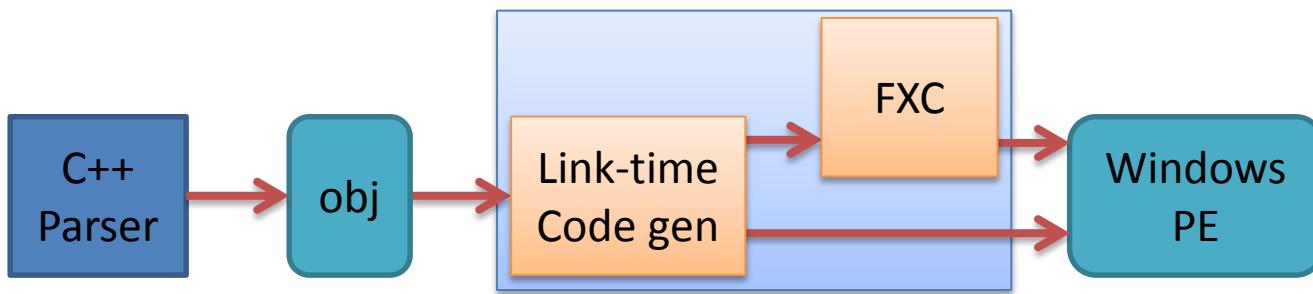
```
void MatrixMultTiled(vector<float>& vC, const vector<float>& vA,  
const vector<float>& vB, int M, int N, int W )  
{  
    static const int TS = 16;  
    array_view<const float,2> a(M, W, vA), b(W, N, vB);  
    array_view<float,2> c(M,N,vC); c.discard_data();  
    parallel_for_each(c.extent.tile< TS, TS >(),  
        [=] (tiled_index< TS, TS > t_idx) restrict(amp)  
            tile_static float locA[TS][TS], locB[TS][TS];  
            int row = t_idx.local[0]; int col = t_idx.local[1];  
            float sum = 0.0f;  
            for (int i = 0; i < W; i += TS) {  
                locA[row][col] = a(t_idx.global[0], col + i);  
                locB[row][col] = b(row + i, t_idx.global[1]);  
                t_idx.barrier.wait();  
            }  
            for (int k = 0; k < TS; k++)  
                sum += locA[row][k] * locB[k][col];  
            t_idx.barrier.wait();  
        }  
        c[t_idx.global] = sum;  
    } );  
}
```



Phase 1

Phase 2

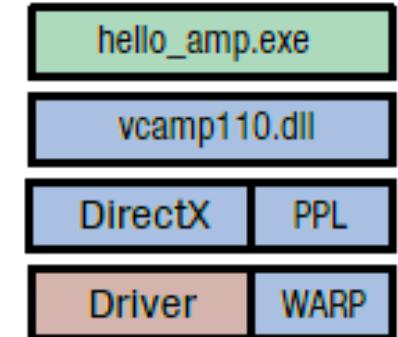
# C++ compilation & deployment



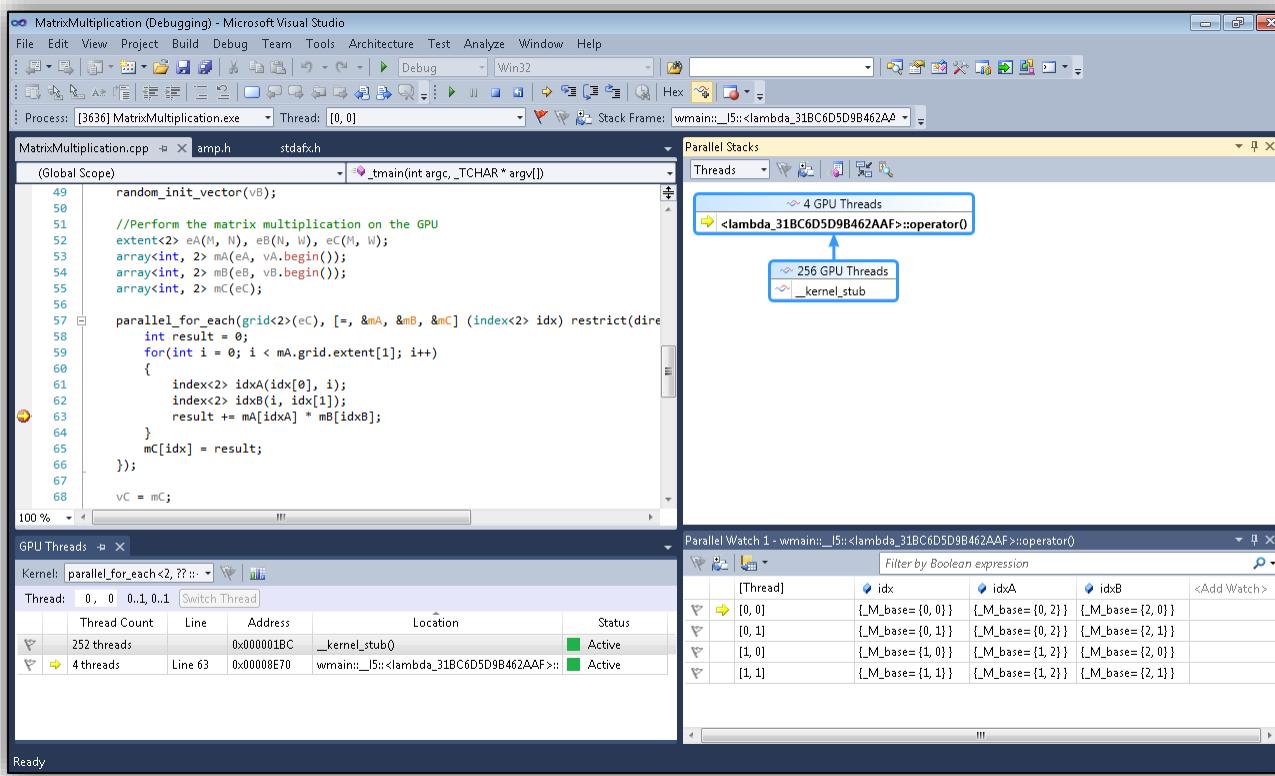
C++ compilation units, separate until linking

Single “fat” binary with native CPU code and DX11 bytecode

After launch, hardware-specific JIT, same as for graphics

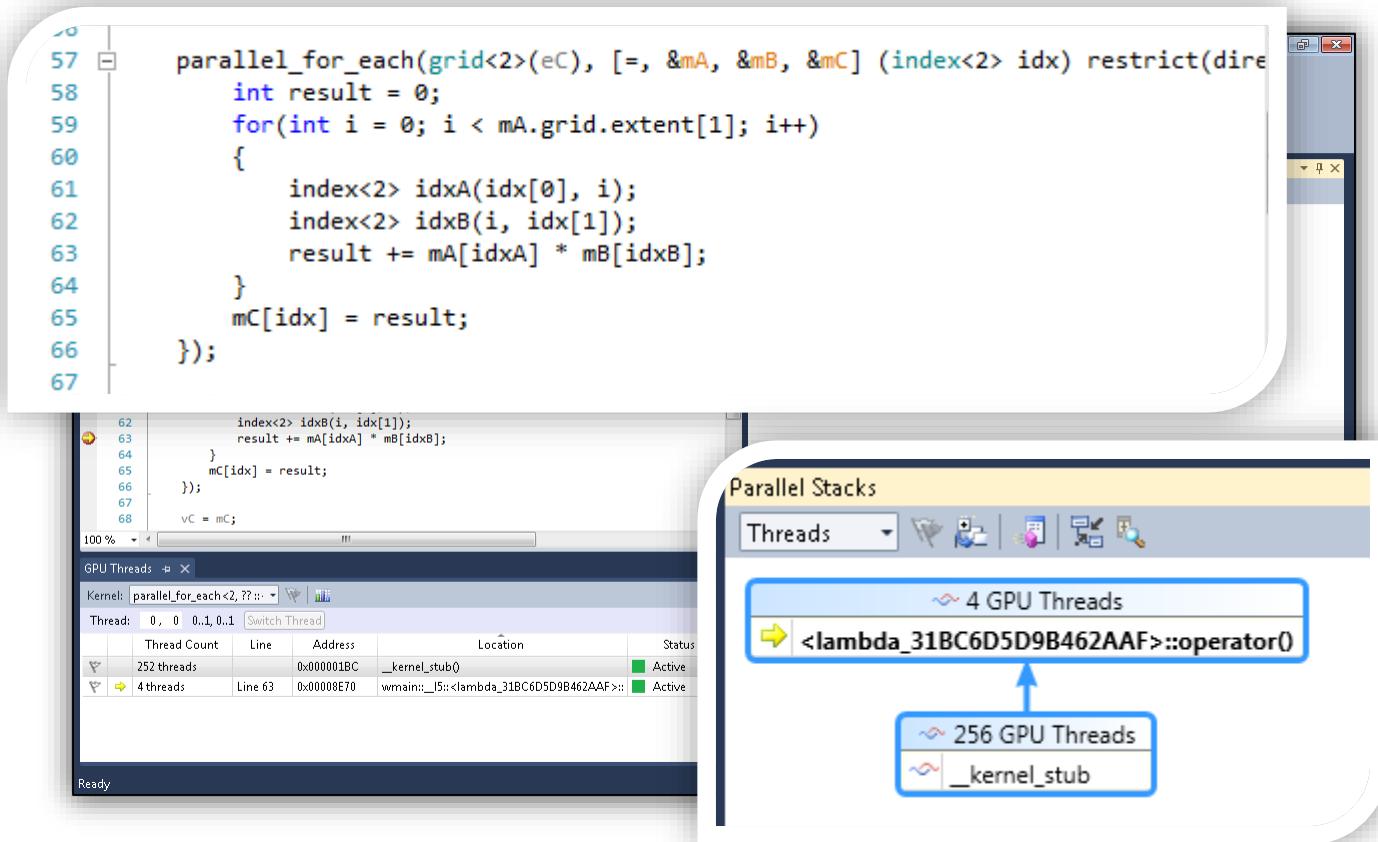


# GPU Debugging



Bring CPU  
debugging  
experience  
to the GPU

# GPU Debugging



Bring CPU  
debugging  
experience  
to the GPU

# GPU Debugging

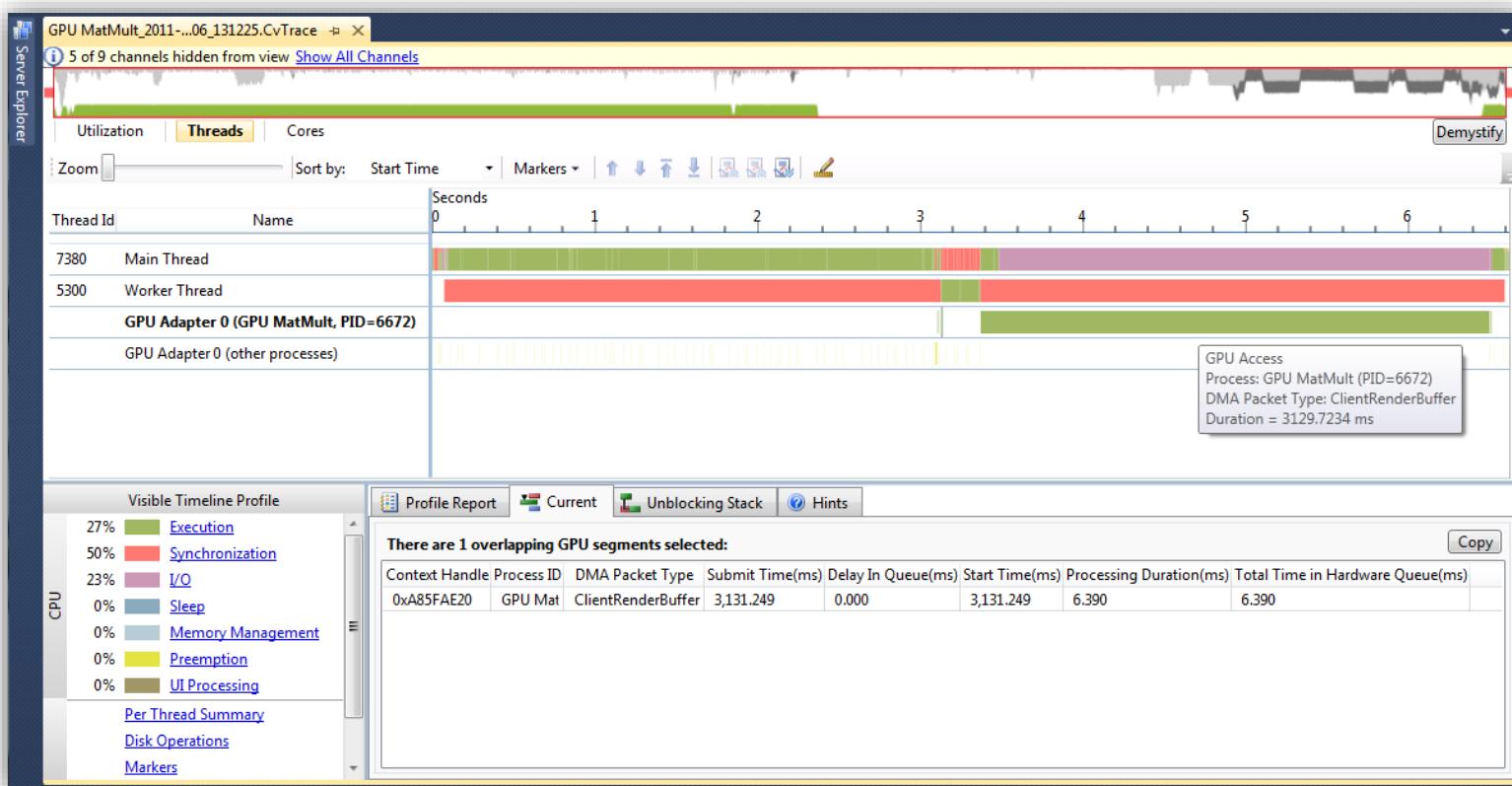
GPU Threads

Kernel: parallel\_for\_each<16,16, ... | Thread: 0, 0 0..15, 0..15 | Switch Thread

Tile	0, 0 0..15, 0..15	Thread	0, 0 0..15, 0..15	Switch Thread		
	Thread Count	Line	Address	Location	Status	Tile
^ Thread Group: [0, 0] (256 Threads)						
252 threads		0x000127D8	_52C1BEEA_3B8D_4BFC_86DE_7D17D68D02A2_	Active	[0, 0]	
2 threads	Line 22	0x00012A88	ObtainAbsoluteIndex	Diverged	[0, 0]	
2 threads	Line 13	0x00012BAC	CalculateAbsoluteIndex	Active	[0, 0]	
^ Thread Group: [0, 11] (256 Threads)						
252 threads		0x000127D8	_52C1BEEA_3B8D_4BFC_86DE_7D17D68D02A2_	Active	[0, 1]	
2 threads	Line 22	0x00012A88	ObtainAbsoluteIndex	Diverged	[0, 1]	
2 threads	Line 13	0x00012BAC	CalculateAbsoluteIndex	Active	[0, 1]	
^ Thread Group: [0, 10] (256 Threads)						
252 threads		0x000127D8	_52C1BEEA_3B8D_4BFC_86DE_7D17D68D02A2_	Active	[0, 10]	
2 threads	Line 22	0x00012A88	ObtainAbsoluteIndex	Diverged	[0, 10]	
2 threads	Line 13	0x00012BAC	CalculateAbsoluteIndex	Active	[0, 10]	
^ Thread Group: [0, 11] (256 Threads)						
252 threads		0x000127D8	_52C1BEEA_3B8D_4BFC_86DE_7D17D68D02A2_	Active	[0, 11]	
2 threads	Line 22	0x00012A88	ObtainAbsoluteIndex	Diverged	[0, 11]	

1 warp; 1 diverged warp:  
warp 0  
2 active threads at CalculateAbsoluteIndex line 13 (address 0x00012BAC)  
2 diverged threads at ObtainAbsoluteIndex line 22 (address 0x00012A88)

# GPU Profiling





# PORTING CS 267

# Assignment 1

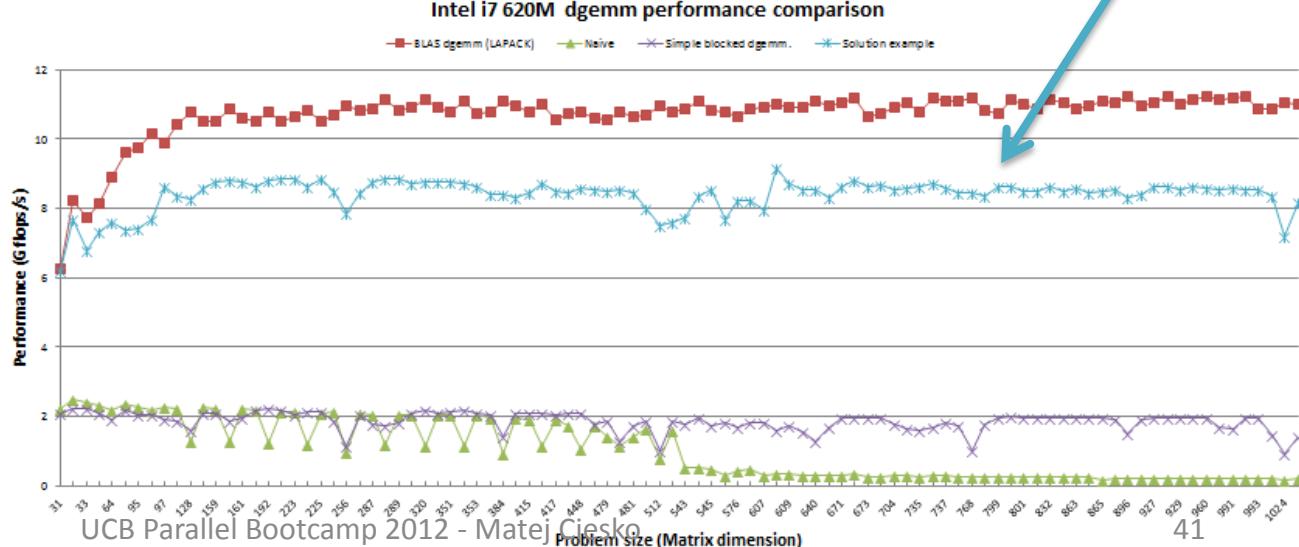
## Serial Performance

- **Assignment target:** Single thread performance
- **Learning goal:**
  - Identify performance bottlenecks
  - Use libraries
  - Optimize kernel
- **Problem:**  $C \leftarrow A \times B + C \quad A, B, C \in \mathbb{R}^{n \times n}$
- **Solution:** **Performance:**

### Optimizations

- Blocking
- Loop unrolling
- Prefetching
- Vectorization
- Repacking data
- [...]

Using Visual Studio students can achieve near optimal performance by applying standard optimization techniques.



# Assignment 1 (Cont.)

Current View: Processes

Function Code View

C:\Users\matej\Desktop\U

```
1 #include <math.h>
2 #include <immintrin.h>
3
4 C void dgemm_(const char *transA, const char *transB, const int m, const int n,
5   { const double alpha = 1.0;
6     const double beta = 0.0;
7     const int k = min(m, n);
8
9     if (m <= 0 || n <= 0)
10    return;
11
12    if (k == 0)
13      return;
14
15    if (transA[0] == 'N' && transB[0] == 'N') {
16      for (int i = 0; i < m; ++i) {
17        for (int j = 0; j < n; ++j) {
18          double sum = 0.0;
19          for (int t = 0; t < k; ++t) {
20            sum += a[i * k + t] * b[t * n + j];
21          }
22          c[i * n + j] = sum * alpha + c[i * n + j] * beta;
23        }
24      }
25    } else if (transA[0] == 'N' && transB[0] == 'T') {
26      for (int i = 0; i < m; ++i) {
27        for (int j = 0; j < n; ++j) {
28          double sum = 0.0;
29          for (int t = 0; t < k; ++t) {
30            sum += a[i * k + t] * b[t * m + j];
31          }
32          c[i * n + j] = sum * alpha + c[i * n + j] * beta;
33        }
34      }
35    } else if (transA[0] == 'T' && transB[0] == 'N') {
36      for (int i = 0; i < m; ++i) {
37        for (int j = 0; j < n; ++j) {
38          double sum = 0.0;
39          for (int t = 0; t < k; ++t) {
40            sum += a[t * m + i] * b[t * n + j];
41          }
42          c[i * n + j] = sum * alpha + c[i * n + j] * beta;
43        }
44      }
45    } else if (transA[0] == 'T' && transB[0] == 'T') {
46      for (int i = 0; i < m; ++i) {
47        for (int j = 0; j < n; ++j) {
48          double sum = 0.0;
49          for (int t = 0; t < k; ++t) {
50            sum += a[t * m + i] * b[t * m + j];
51          }
52          c[i * n + j] = sum * alpha + c[i * n + j] * beta;
53        }
54      }
55    }
56  }
```

Performance Wizard -- Page 1 of 3

Specify the profiling method

Profiling your application can help diagnose performance problems and identify the most common expensive methods in your application. To begin, choose a profiling method from the options below.

What method of profiling would you like to use?

CPU sampling (recommended)  
Monitor CPU-bound applications with low overhead

Instrumentation  
Measure function call counts and timing

.NET memory allocation (sampling)  
Track managed memory allocation

Resource contention data (concurrency)  
Detect threads waiting for other threads

[Read more about profiling methods](#)

< Previous Next > Finish Cancel

Using Visual Studio student can identify performance bottlenecks.

0.00 msvcr110.dll  
1.52 Matmul.exe  
0.04 Matmul.exe  
0.03 Matmul.exe  
0.00 Matmul.exe

UCB Parallel Bootcamp 2012 - Matej Ciesko

8/12/2012

42

# Assignment 2

## Parallel Programming

- **Assignment target & learning goal:**

- Parallelize given problem
- Distributed & shared memory parallelism
- Accelerators (CUDA)

- **Problem: n-body**

On Windows:

- Native threads
- OpenMP, PPL
- MPI
- C++ AMP

Using Visual Studio  
students can easily write  
parallel code in native  
threads, OpenMP, MPI,  
C++ AMP.

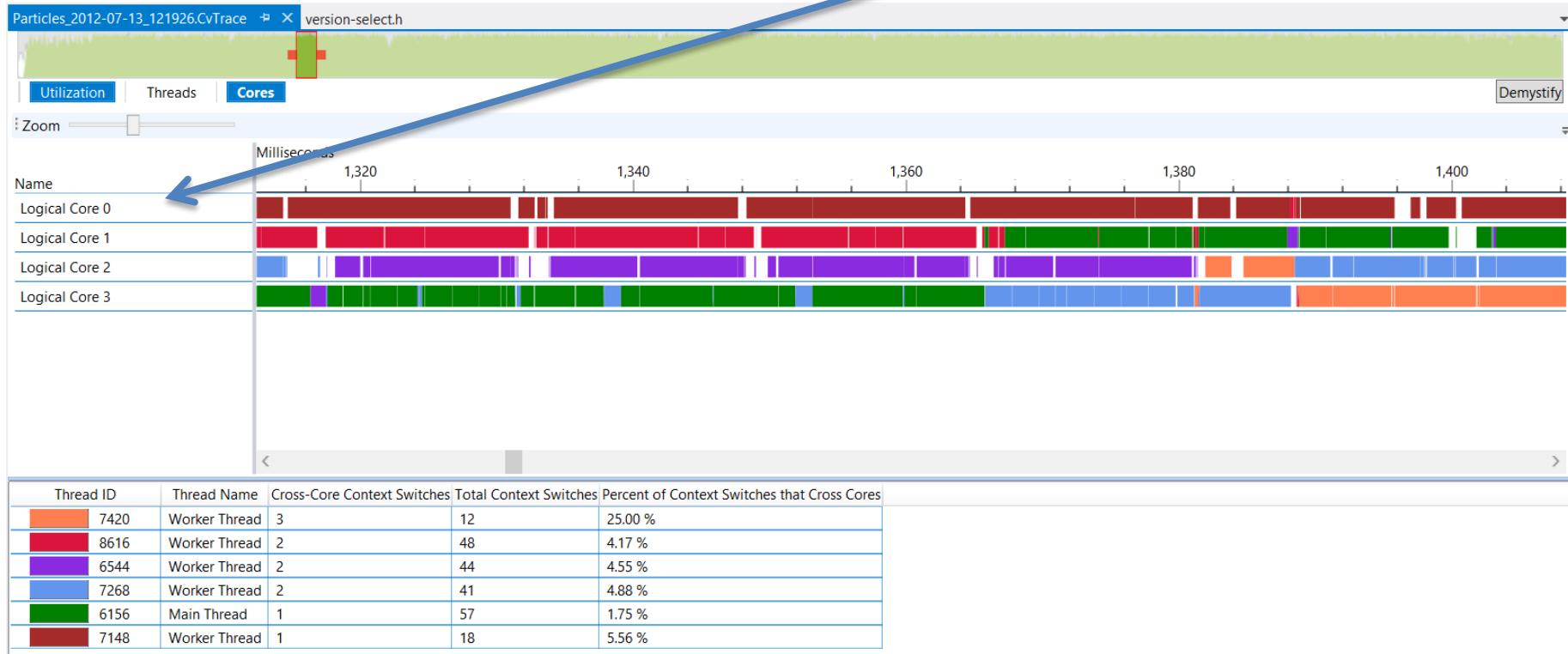
```
#include <amp.h>
array_view<particle_t, 1> particles_av(n, particles);
for( int step = 0; step < NSTEPS; step++ )
{
    parallel_for_each(particles_av.extent, [=] (index<1> i) restrict(amp){
        particles_av[i].ax = 0;
        particles_av[i].ay = 0;
        for (int j = 0; j < n; j++ ){
            apply_force( particles_av[i], particles_av[j] );
        }
    });
    parallel_for_each(particles_av.extent, [=] (index<1> i) restrict(amp){
        move( particles_av[i], size );
    });
    particles_av.synchronize();
}
```

- Performance comparison: (initial/ trivial solution, n = 1000)

Threading	OpenMP	PPL	MPI	C++ AMP*
3.49 sec	2.21 sec	2.33 sec	2.70 sec	2.41 sec

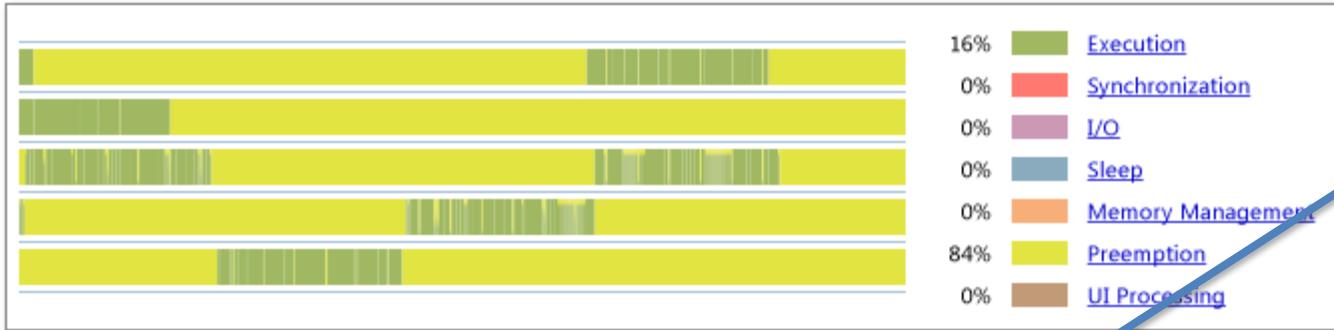
# Assignment 2 (Cont.)

Using Visual Studio students can deep dive into synchronization patterns and timing.



# Assignment 2 (Cont.)

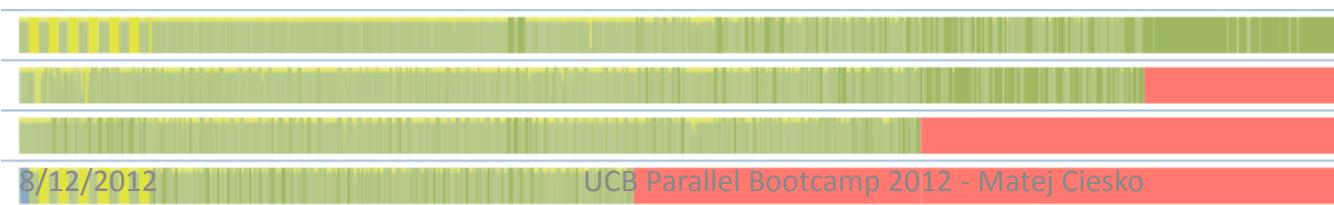
- Oversubscription



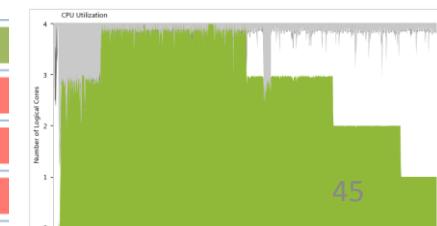
- Lock Convoys



- Uneven Workload distribution



Using Visual Studio  
students can deep dive  
into synchronization  
patterns and timing.



# Lab

- Get hands-on experience using Visual Studio profiling and parallel debugging functionality
- Experiment with C++ AMP

## Sources

- **AMP blog:** <http://blogs.msdn.com/b/nativeconcurrency/>
- **H. Sutter blog:** <http://herbsutter.com/>
- **NsfPPC:** <http://www.nsfppc.net>

Thank you for your attention!