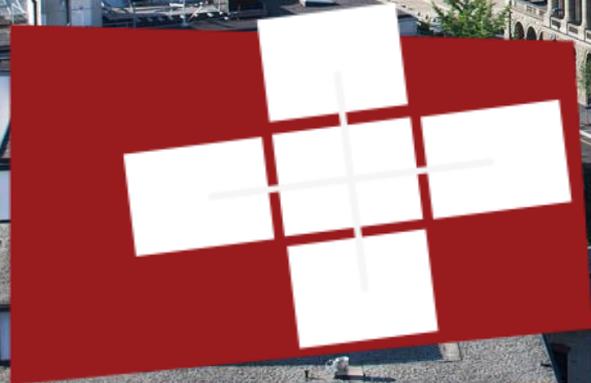


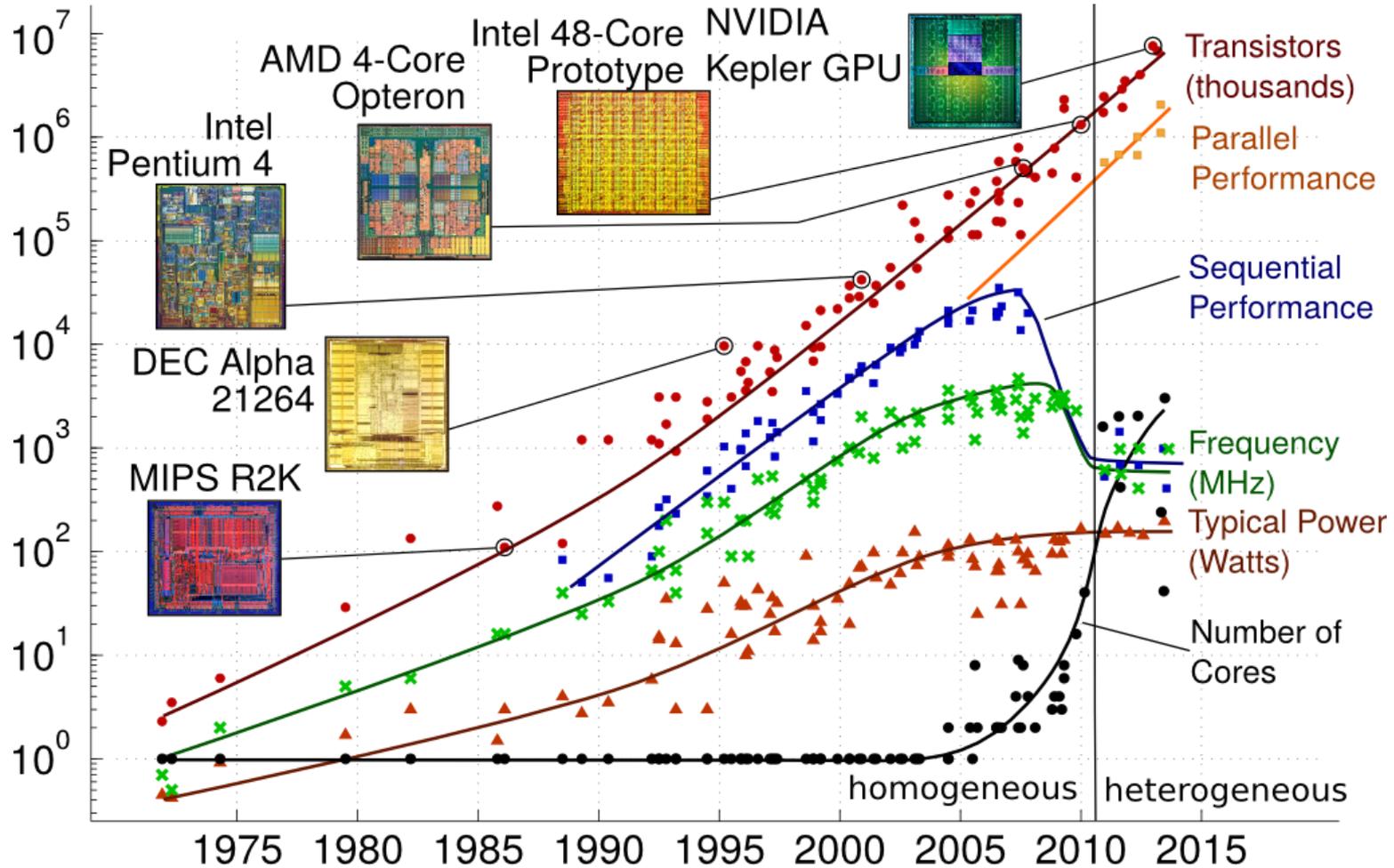
TORSTEN HOEFLER

# Progress in automatic GPU compilation and why you want to run MPI on your GPU

with Tobias Grosser and Tobias Gysi @ SPCL  
presented at High Performance Computing, Cetraro, Italy 2016



# Evading various “ends” – the hardware view

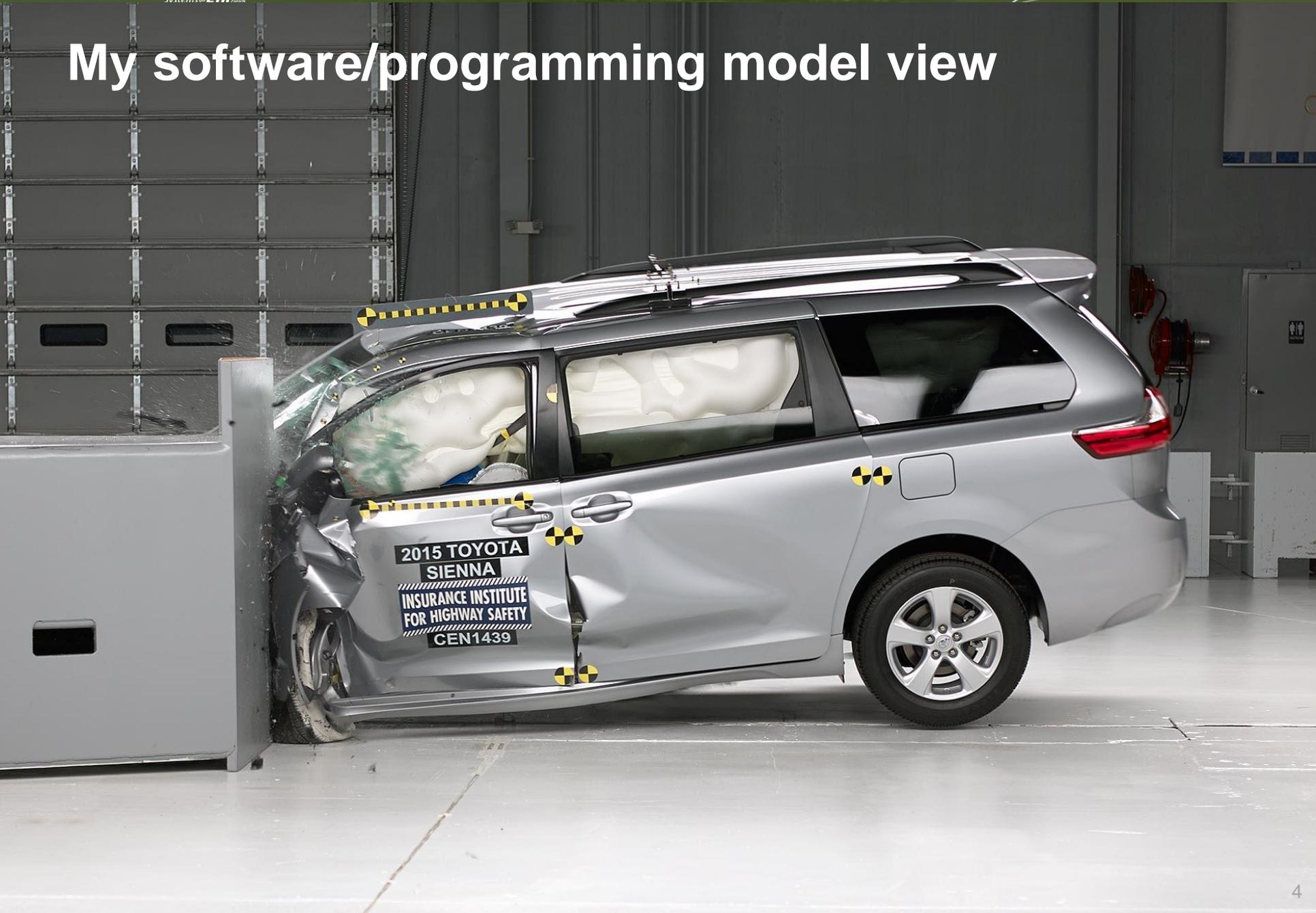


Data partially collected by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond

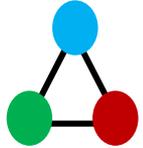
# Pete's system software view



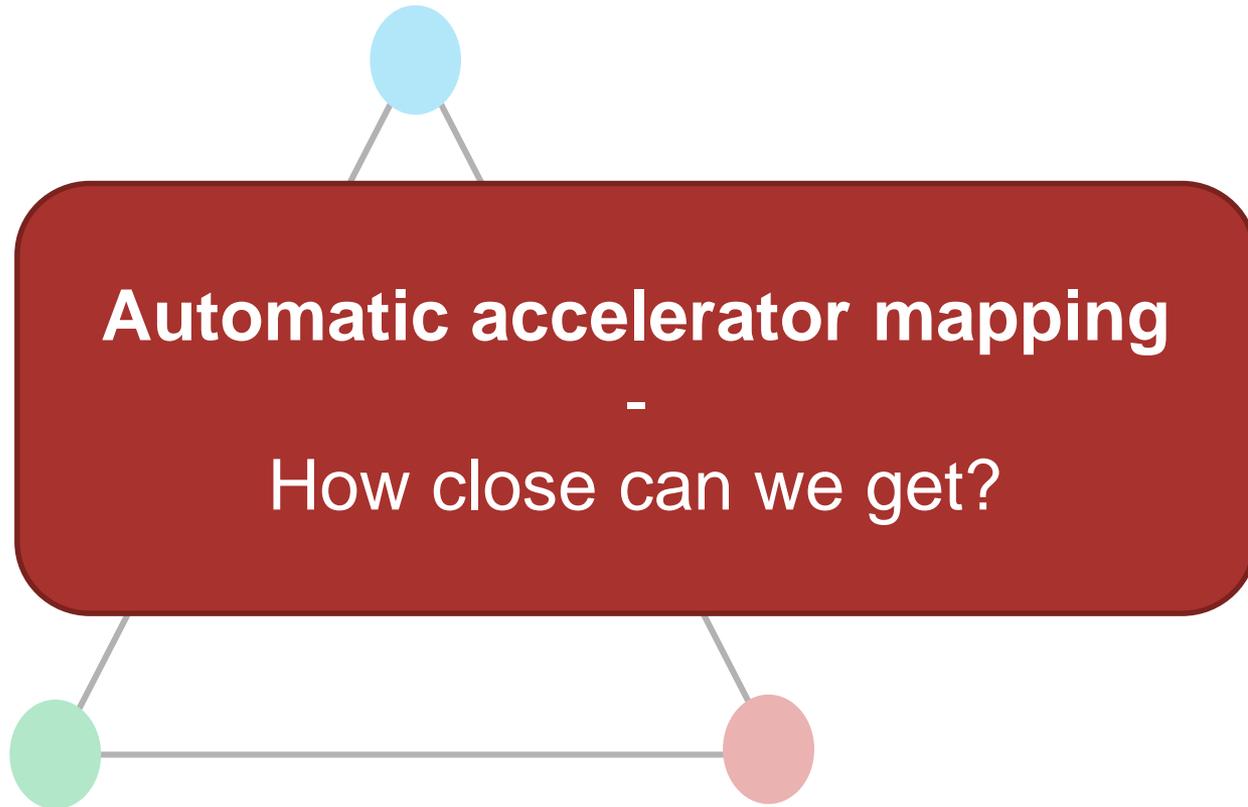
# My software/programming model view



# Holy grail – auto-parallelization heterogenization



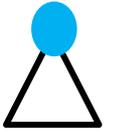
Automatic



Regression Free

High Performance

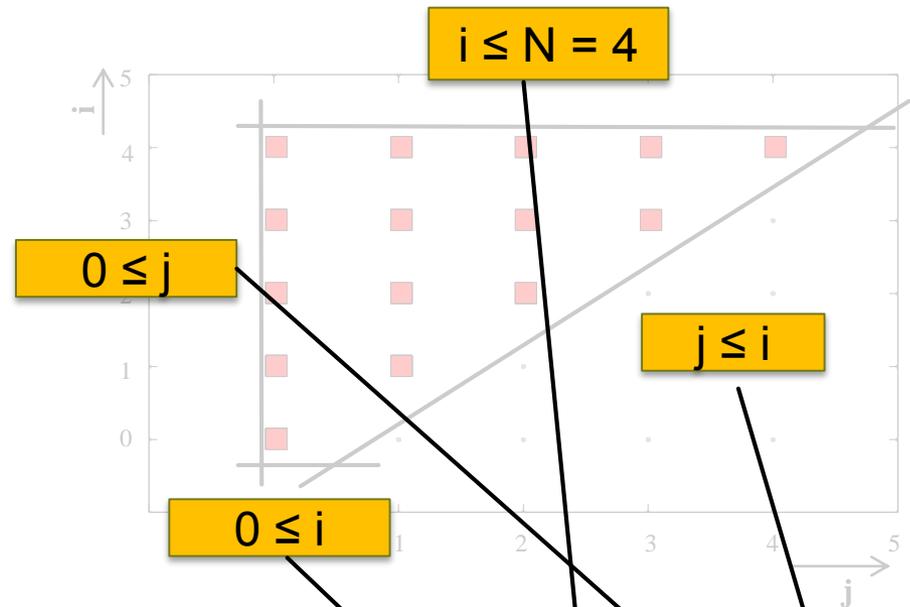
# Tool: Polyhedral Modeling



*Program Code*

```
for (i = 0; i <= N; i++)
  for (j = 0; j <= i; j++)
    S(i,j);
```

*Iteration Space*

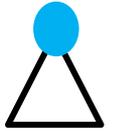


$N = 4$

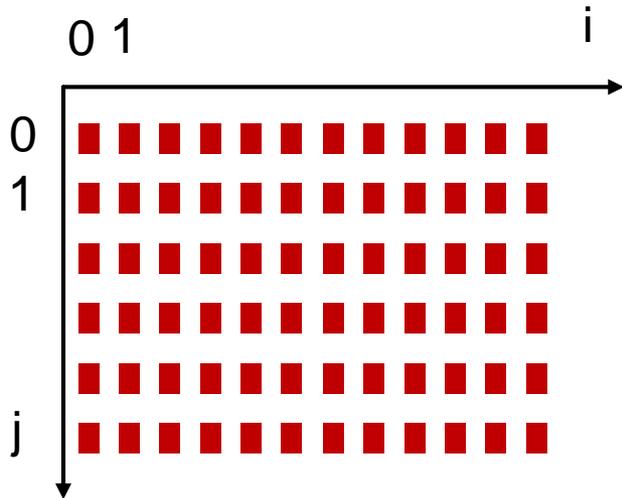
$(i, j) = (4, 4)$

$$D = \{ (i,j) \mid 0 \leq i \leq N \wedge 0 \leq j \leq i \}$$

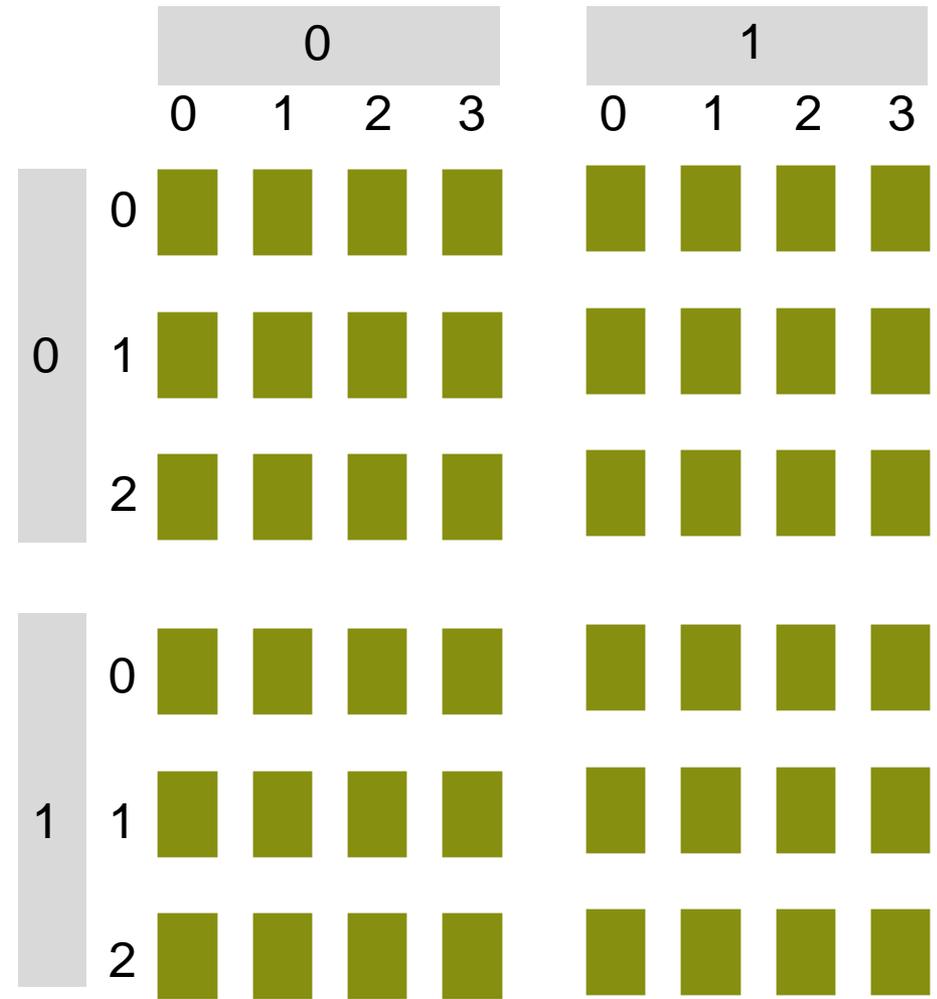
# Mapping Computation to Device



*Iteration Space*



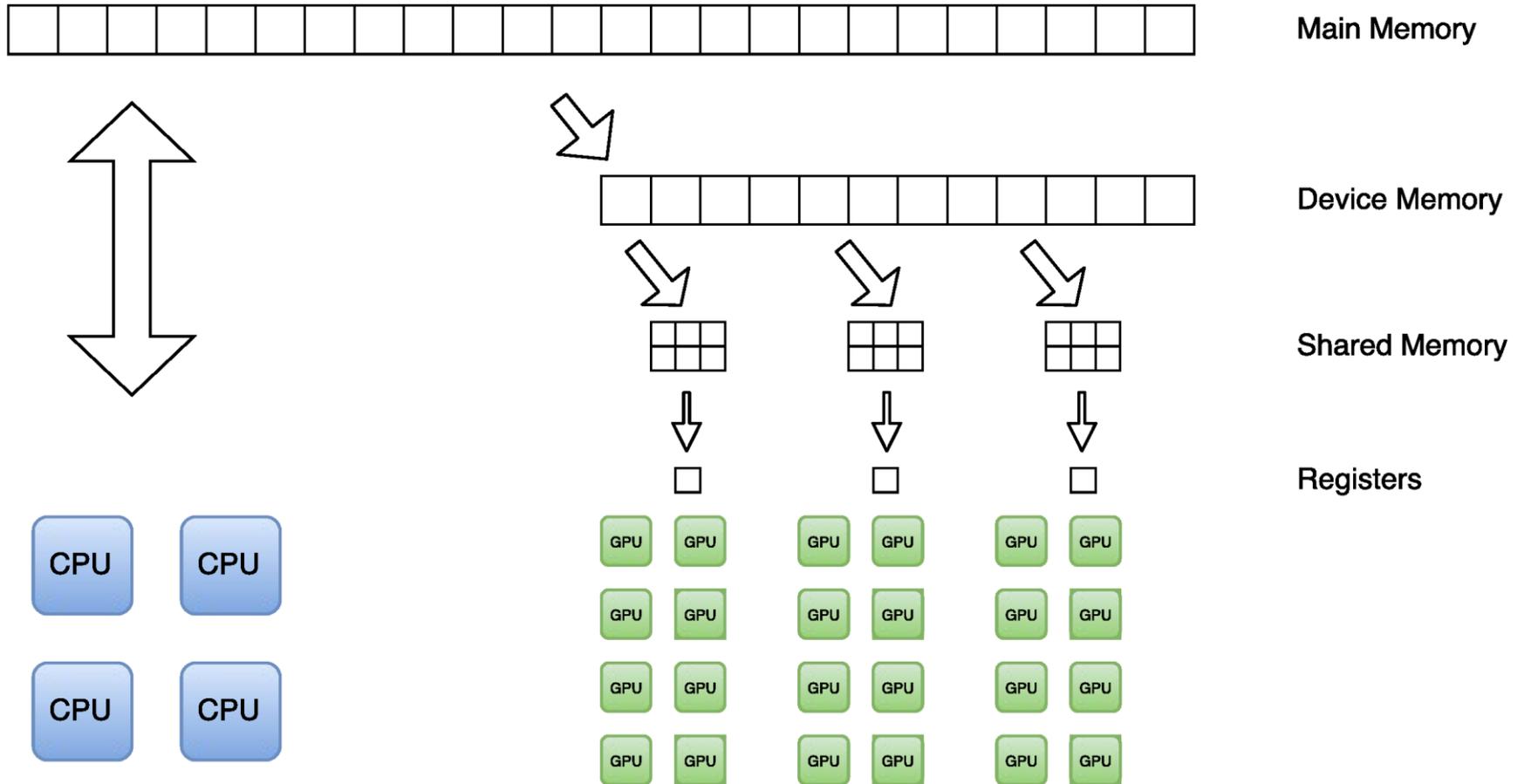
*Device Blocks & Threads*



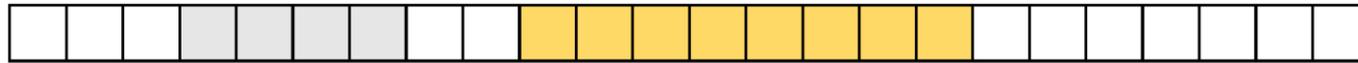
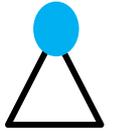
$$BID = \{(i, j) \rightarrow \left( \left\lfloor \frac{i}{4} \right\rfloor \% 2, \left\lfloor \frac{j}{3} \right\rfloor \% 2 \right)\}$$

$$TID = \{(i, j) \rightarrow (i \% 4, j \% 3)\}$$

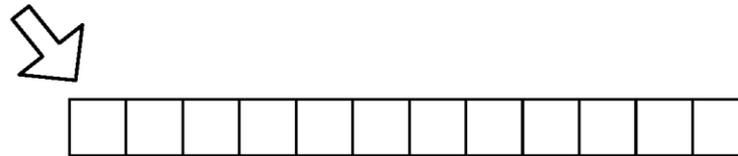
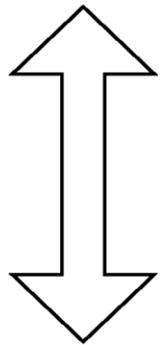
# Memory Hierarchy of a Heterogeneous System



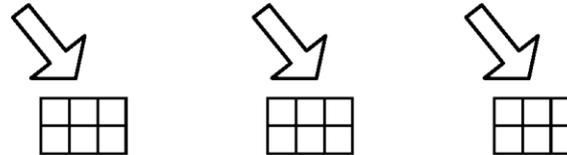
# Host-device data transfers



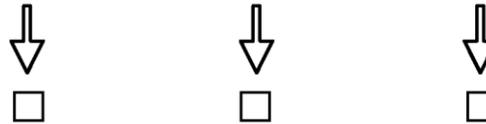
Main Memory



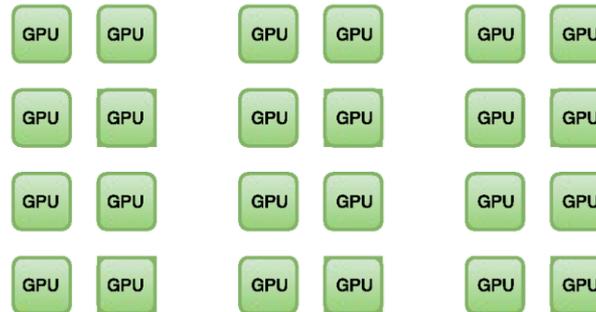
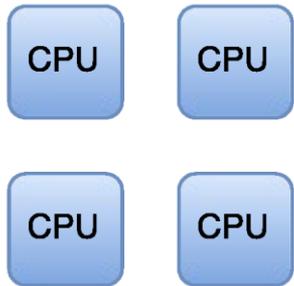
Device Memory



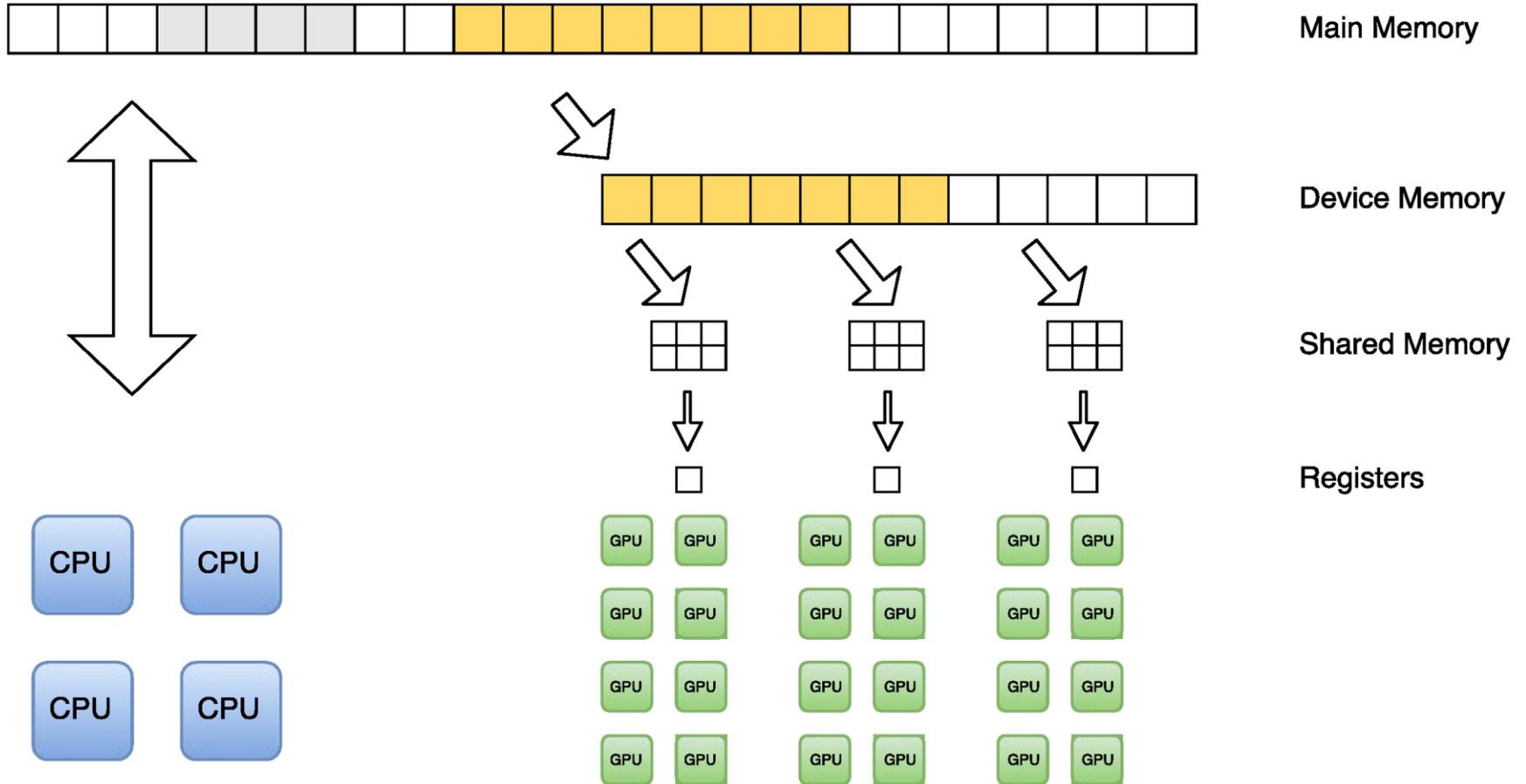
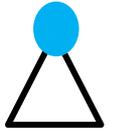
Shared Memory



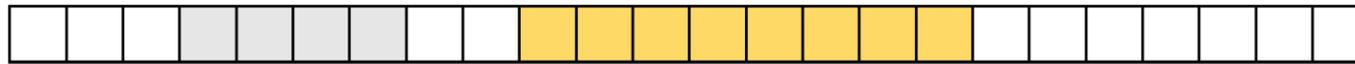
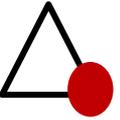
Registers



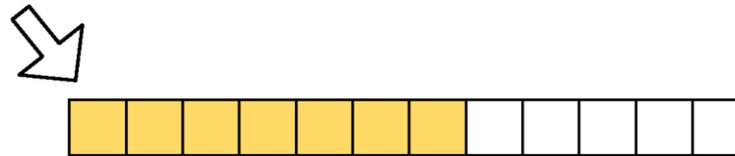
# Host-device data transfers



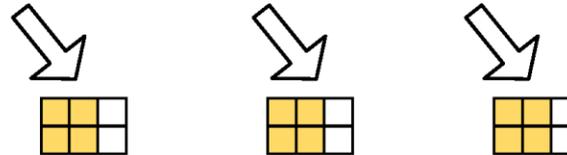
# Mapping onto fast memory



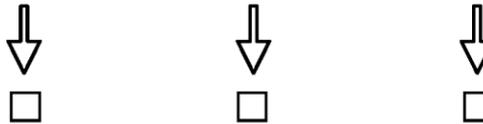
Main Memory



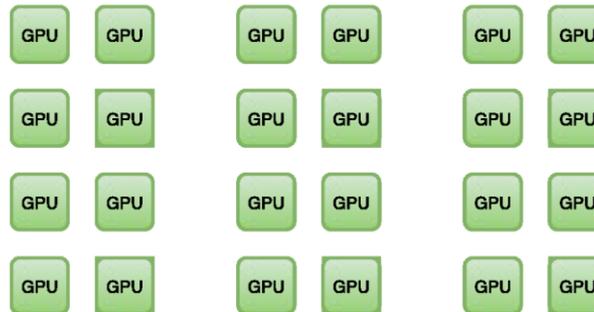
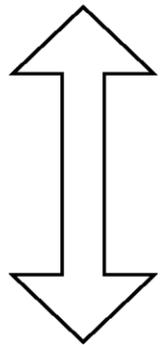
Device Memory



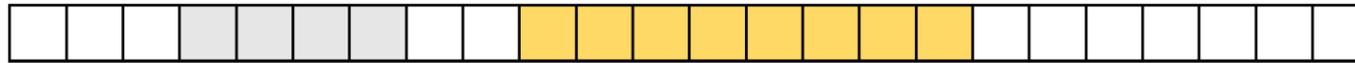
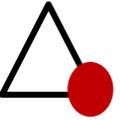
Shared Memory



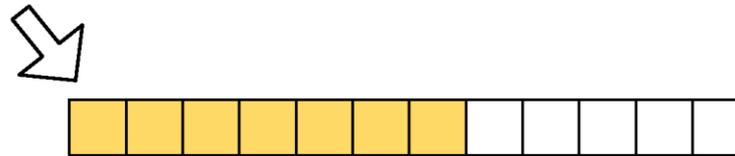
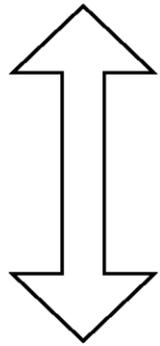
Registers



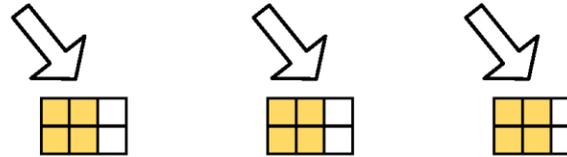
# Mapping onto fast memory



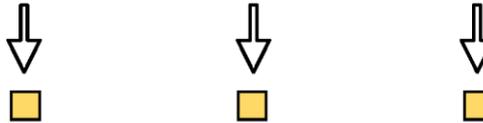
Main Memory



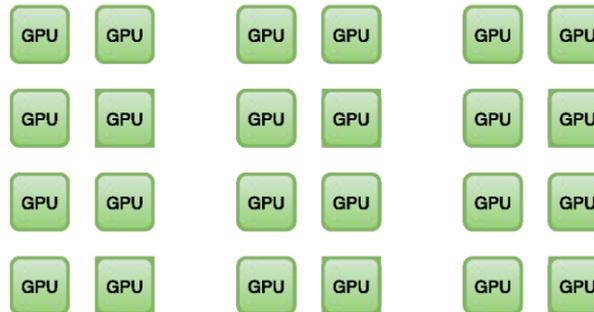
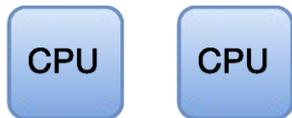
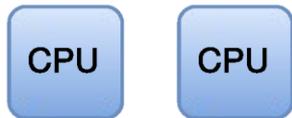
Device Memory



Shared Memory



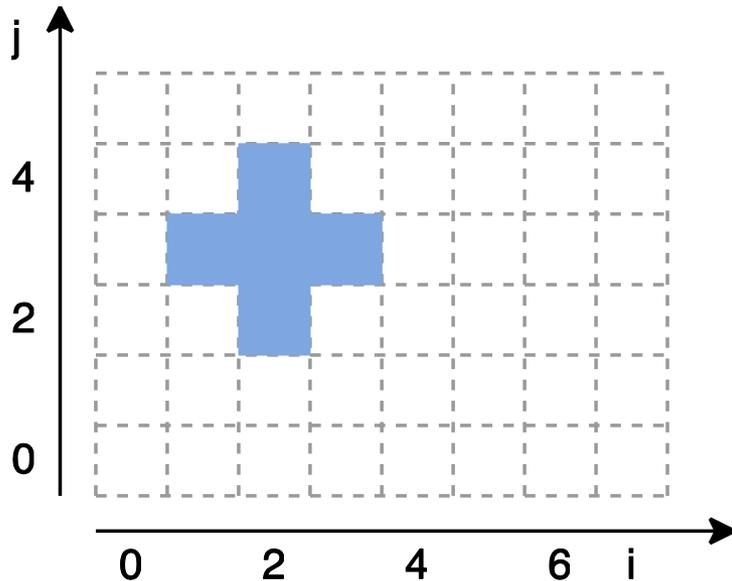
Registers



# Accessed Data (for a 2x2 thread block)



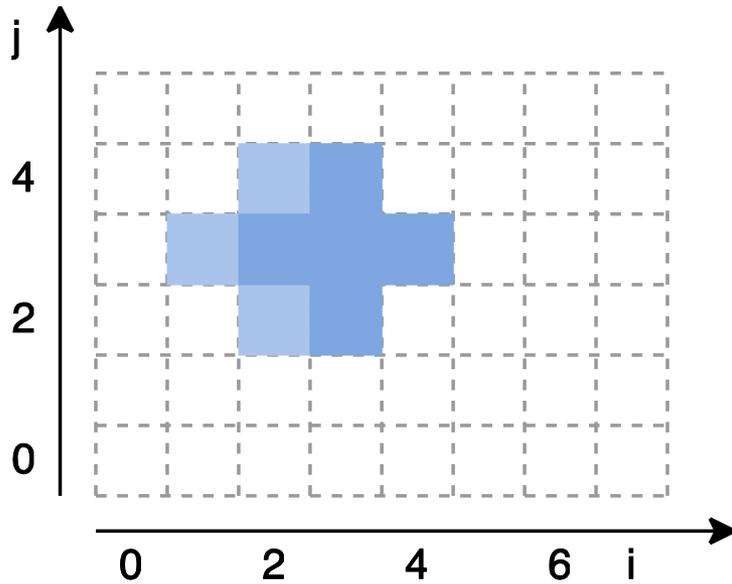
```
for (i = 1; i <= 6; i++)  
  for (j = 1; j <= 4; j++)  
    ... = A[i+1][j] + A[i-1][j] + A[i][j] + A[i][j-1] + A[i][j+1];
```



# Accessed Data (for a 2x2 thread block)



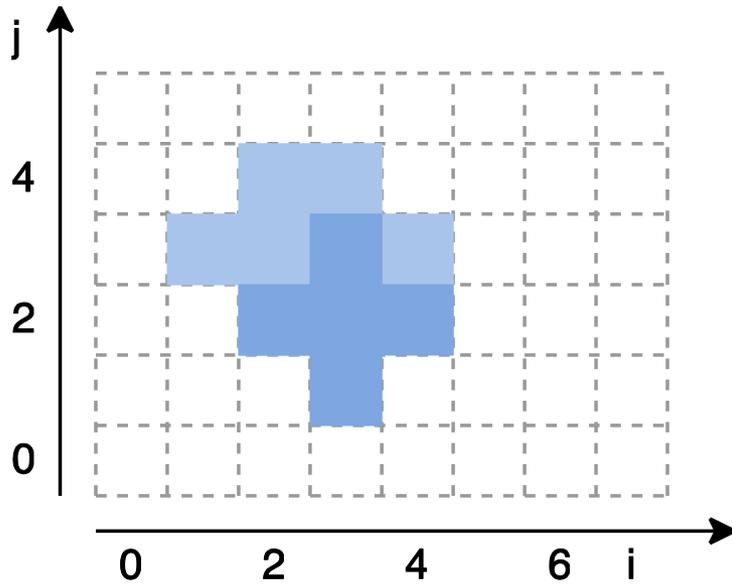
```
for (i = 1; i <= 6; i++)  
  for (j = 1; j <= 4; j++)  
    ... = A[i+1][j] + A[i-1][j] + A[i][j] + A[i][j-1] + A[i][j+1];
```



# Accessed Data (for a 2x2 thread block)



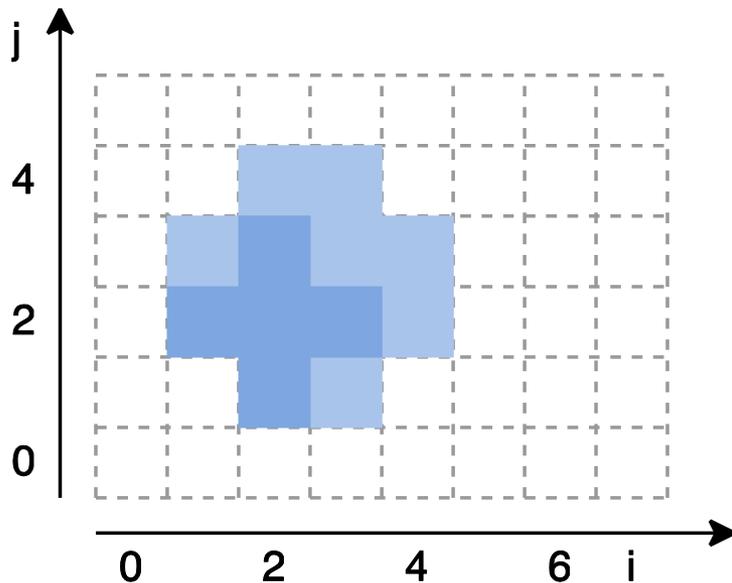
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```



# Accessed Data (for a 2x2 thread block)

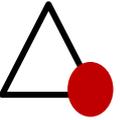


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  for (j = 1; j <= 4; j++)  
    ... = A[i+1][j] + A[i-1][j] + A[i][j] + A[i][j-1] + A[i][j+1];
```

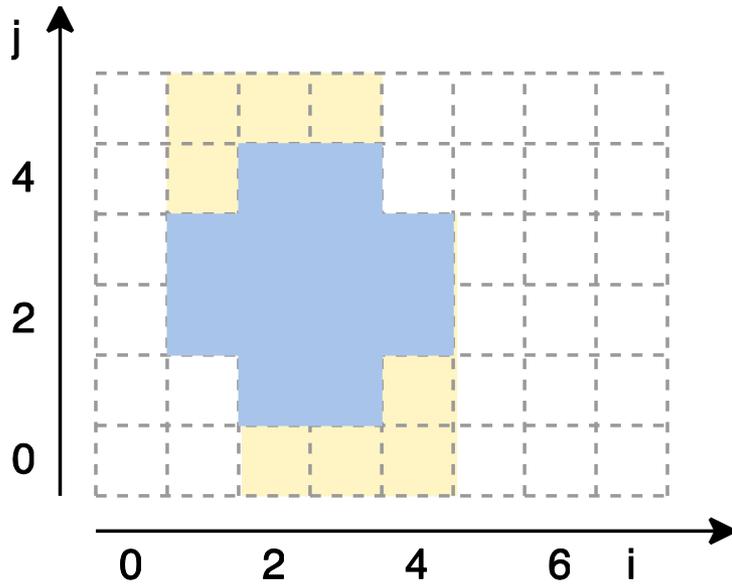


- Data needed on device
- 12 elements
- Minimal data, but complex transfer

# Accessed Data (for a 2x2 thread block)

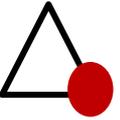


```
for (i = 1; i <= 6; i++)
  for (j = 1; j <= 4; j++)
    ... = A[i+1][j] + A[i-1][j] + A[i][j] + A[i][j-1] + A[i][j+1];
```



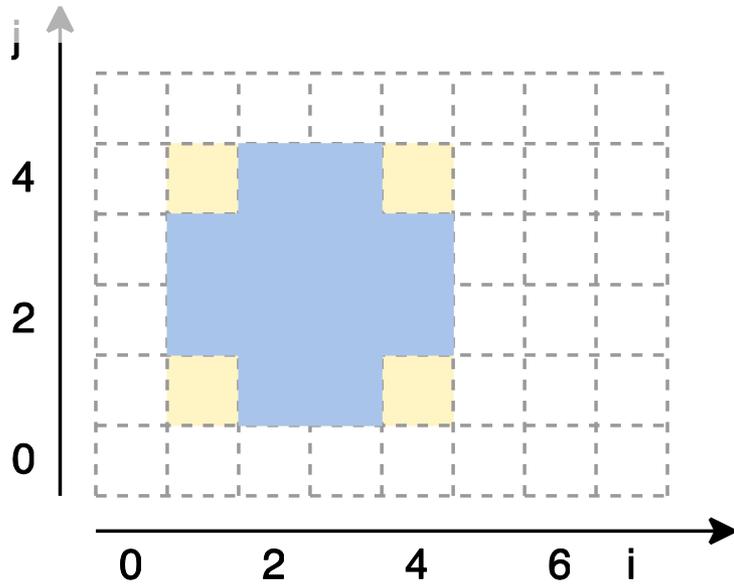
- One-dimensional hull
- 20 elements
- Simple transfer, but redundant data

# Accessed Data (for a 2x2 thread block)



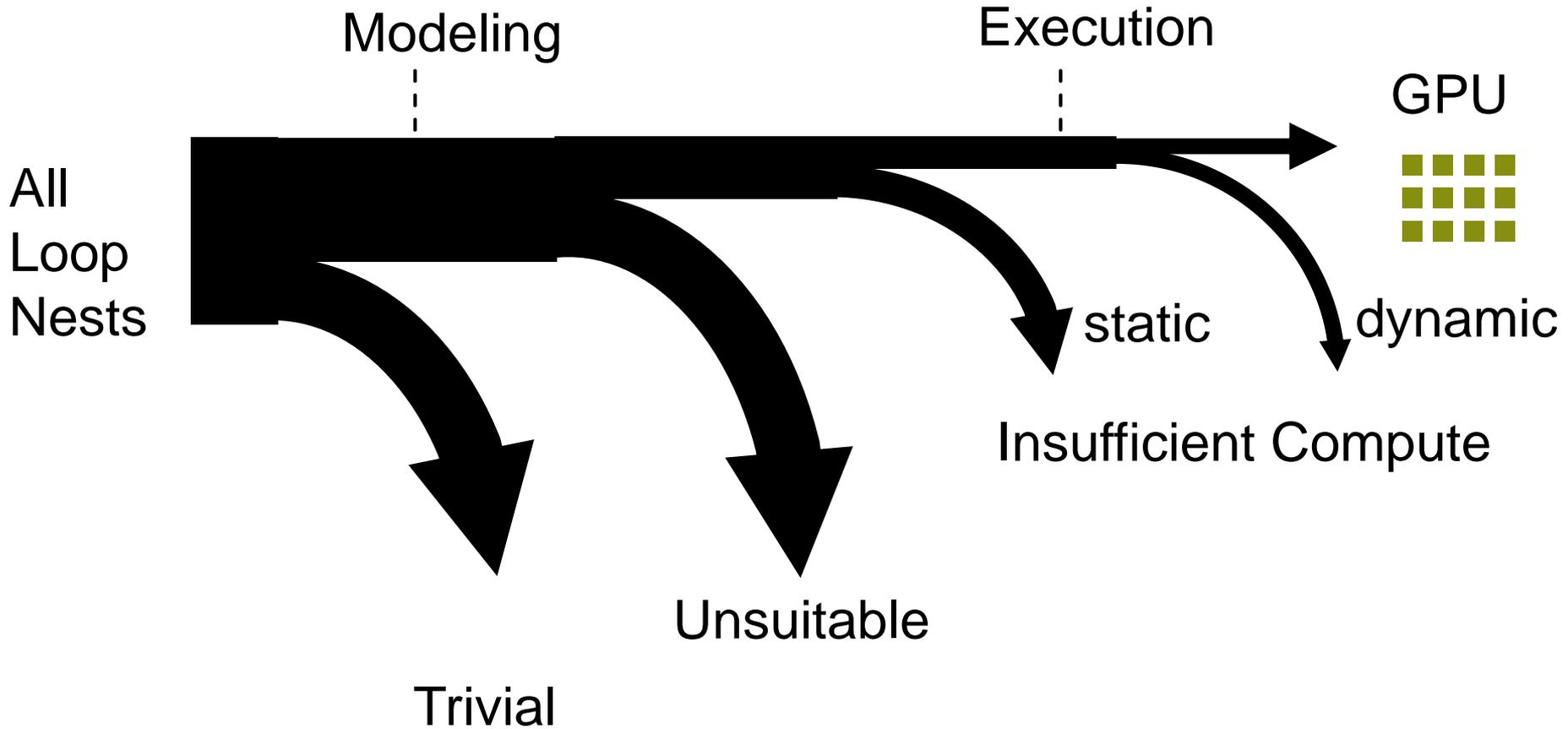
```
for (i = 1; i <= 5; i++)
  for (j = 1; j <= 5; j++)
    ...
```

Modeling multi-dimensional access behavior is important

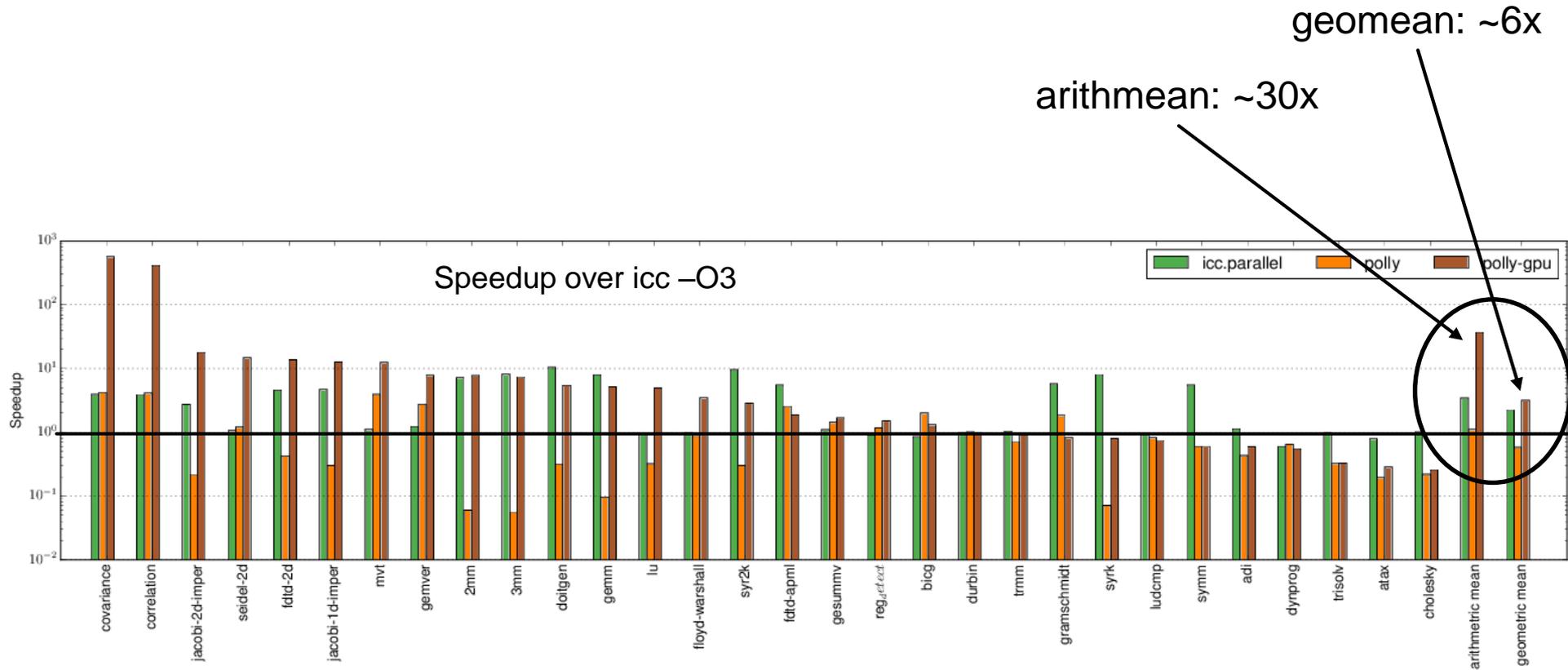


- Two-dimensional hull
- 16 elements
- Simple transfer, less redundant data

# Profitability Heuristic

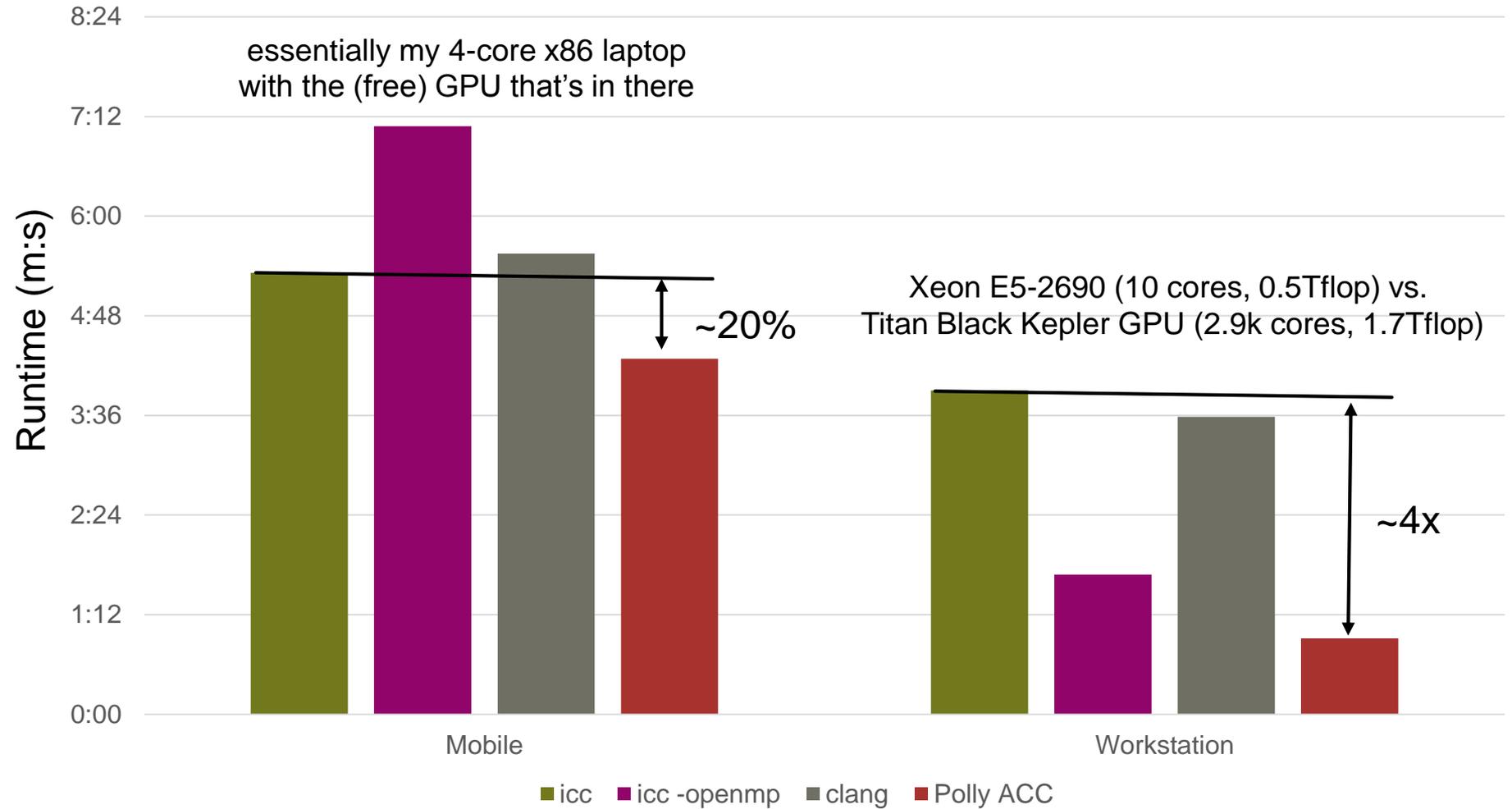


# Some results: Polybench 3.2



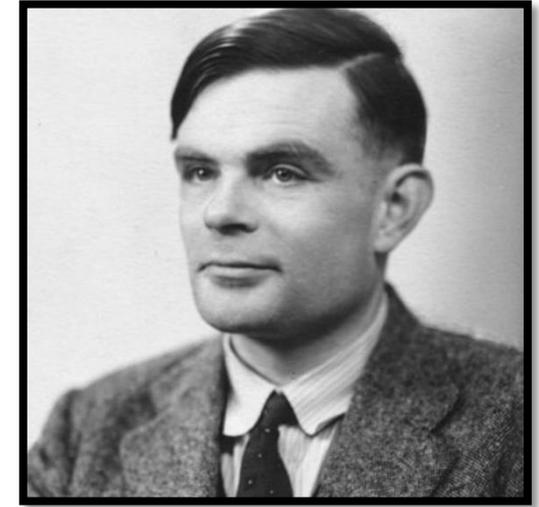
Xeon E5-2690 (10 cores, 0.5Tflop) vs. Titan Black Kepler GPU (2.9k cores, 1.7Tflop)

# Compiles all of SPEC CPU 2006 – Example: LBM



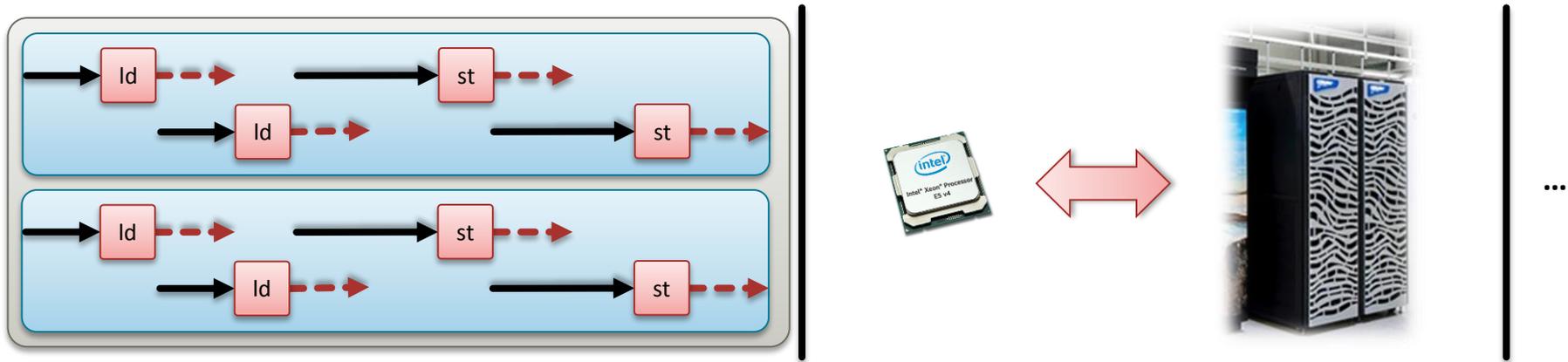
# Brave new compiler world!?

- **Unfortunately not ...**
  - Limited to affine code regions
  - Maybe generalizes to control-restricted programs
  - No distributed anything!!
- **Good news:**
  - Much of traditional HPC fits that model
  - Infrastructure is coming along



- **Bad news:**
  - Modern data-driven HPC and Big Data fits less well
  - Need a programming model for **distributed** heterogeneous machines!

# How do we program GPUs today?



## CUDA

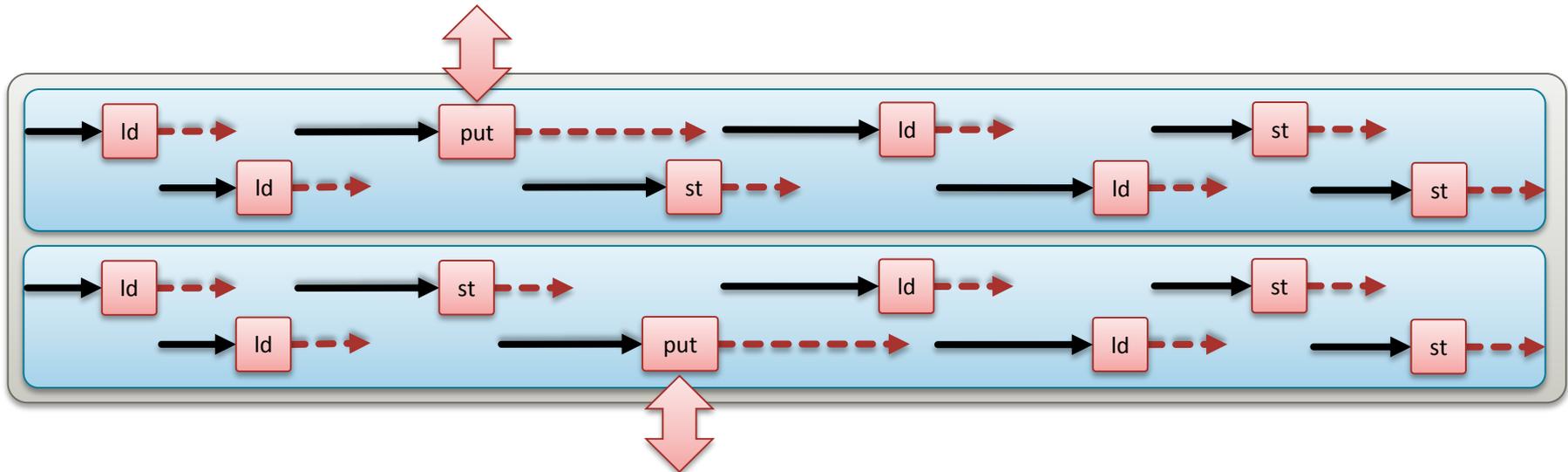
- over-subscribe hardware
- use spare parallel slack for latency hiding

## MPI

- host controlled
- full device synchronization



# Latency hiding at the cluster level?



## dCUDA (distributed CUDA)

- unified programming model for GPU clusters
- avoid unnecessary device synchronization to enable system wide latency hiding



# dCUDA extends CUDA with MPI-3 RMA and notifications

```

for (int i = 0; i < steps; ++i) {
  for (int idx = from; idx < to; idx += jstride)
    out[idx] = -4.0 * in[idx] +
      in[idx + 1] + in[idx - 1] +
      in[idx + jstride] + in[idx - jstride];

  if (lsend)
    dcuda_put_notify(ctx, wout, rank - 1,
      len + jstride, jstride, &out[jstride], tag);
  if (rsend)
    dcuda_put_notify(ctx, wout, rank + 1,
      0, jstride, &out[len], tag);

  dcuda_wait_notifications(ctx, wout,
    DCUDA_ANY_SOURCE, tag, lsend + rsend);

  swap(in, out); swap(win, wout);
}

```

computation

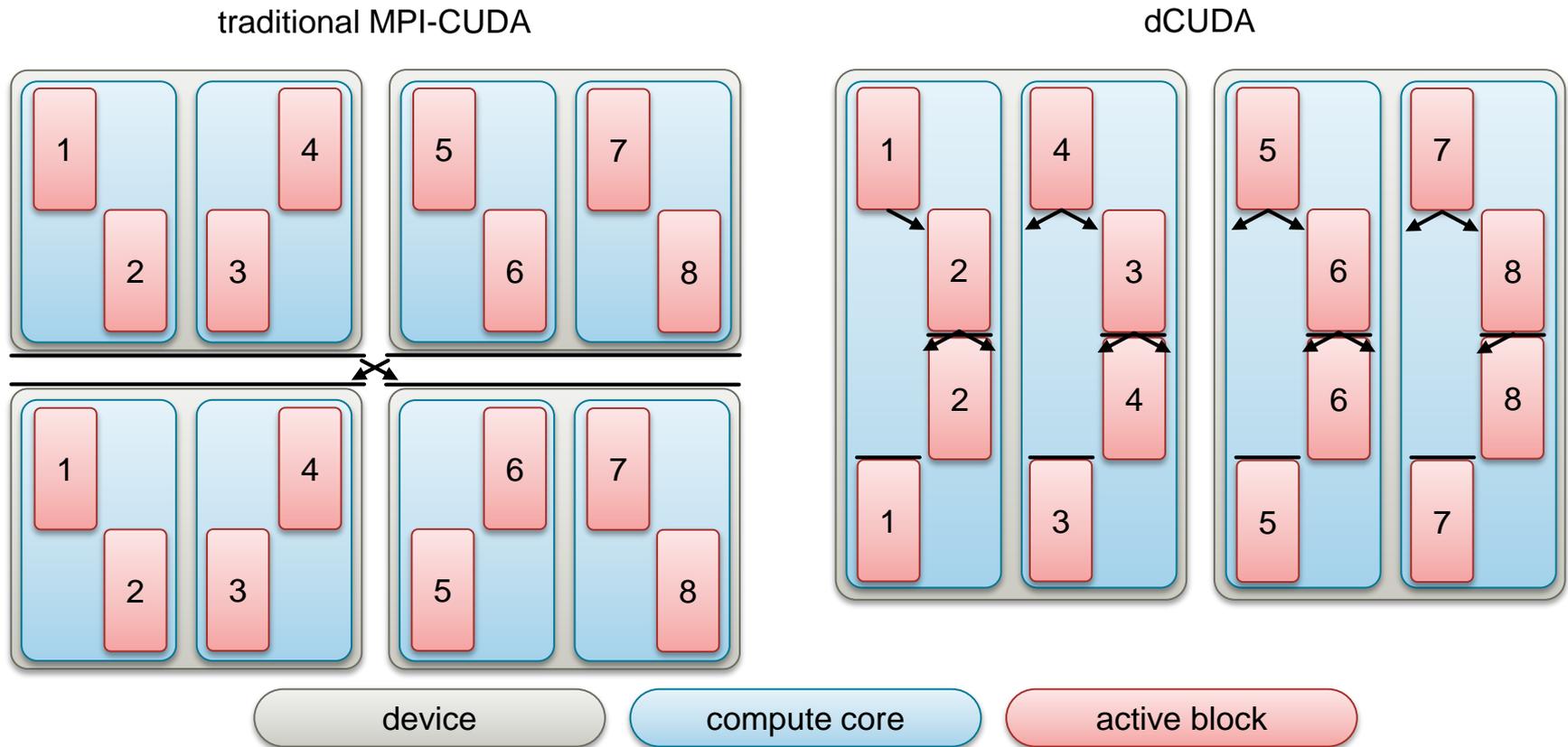
communication

- iterative stencil kernel
- thread specific idx

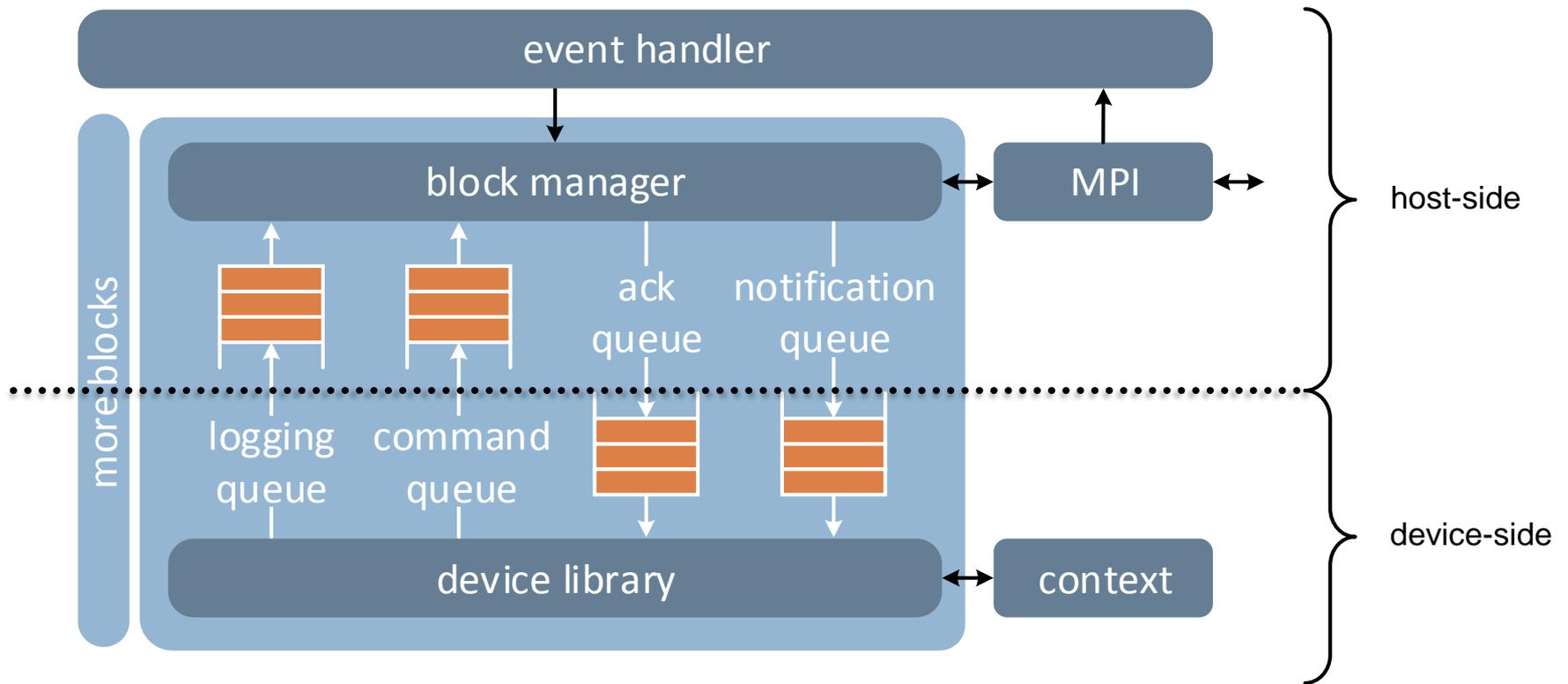


- map ranks to blocks
- device-side put/get operations
- notifications for synchronization
- shared and distributed memory

# Hardware supported communication overlap

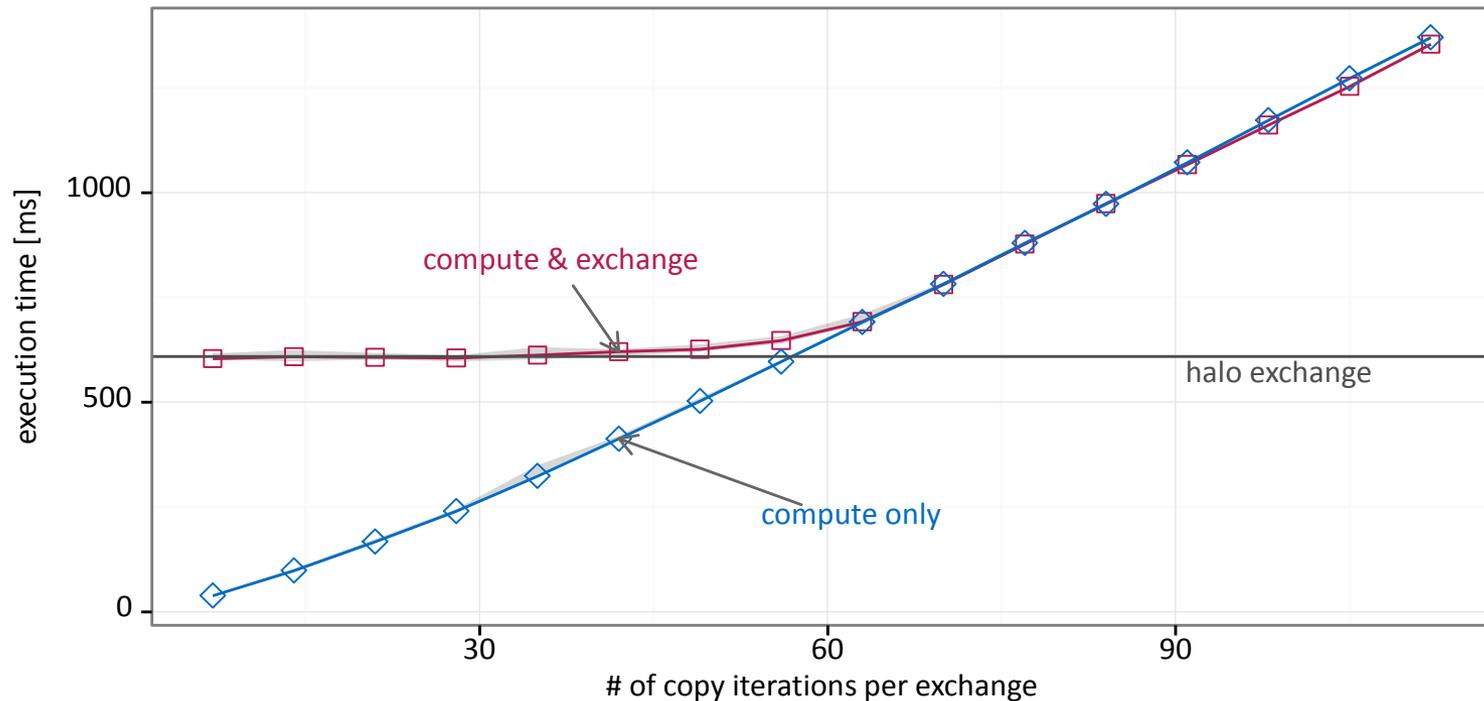


# Implementation of the dCUDA runtime system



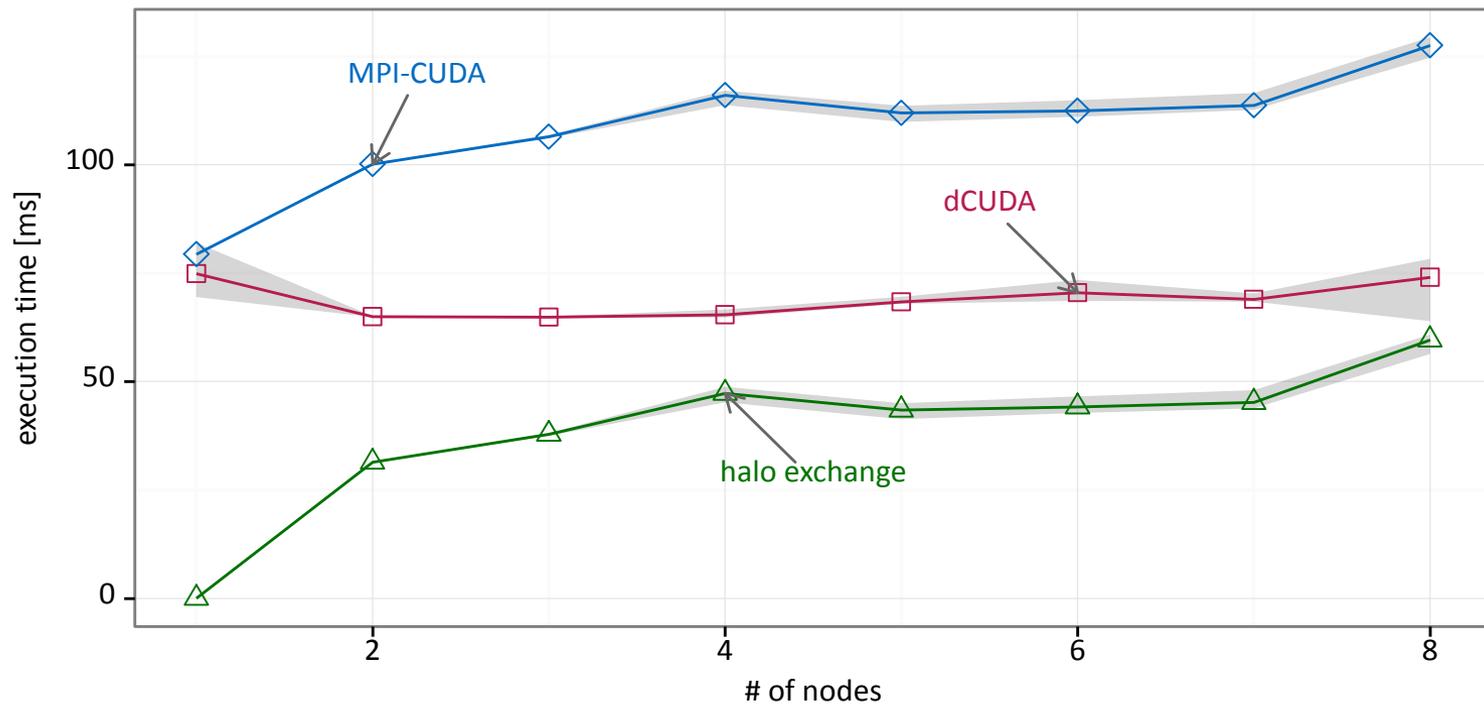
# Overlap of a copy kernel with halo exchange communication

- Benchmarked on Greina (8 Haswell nodes with 1x Tesla K80 per node)



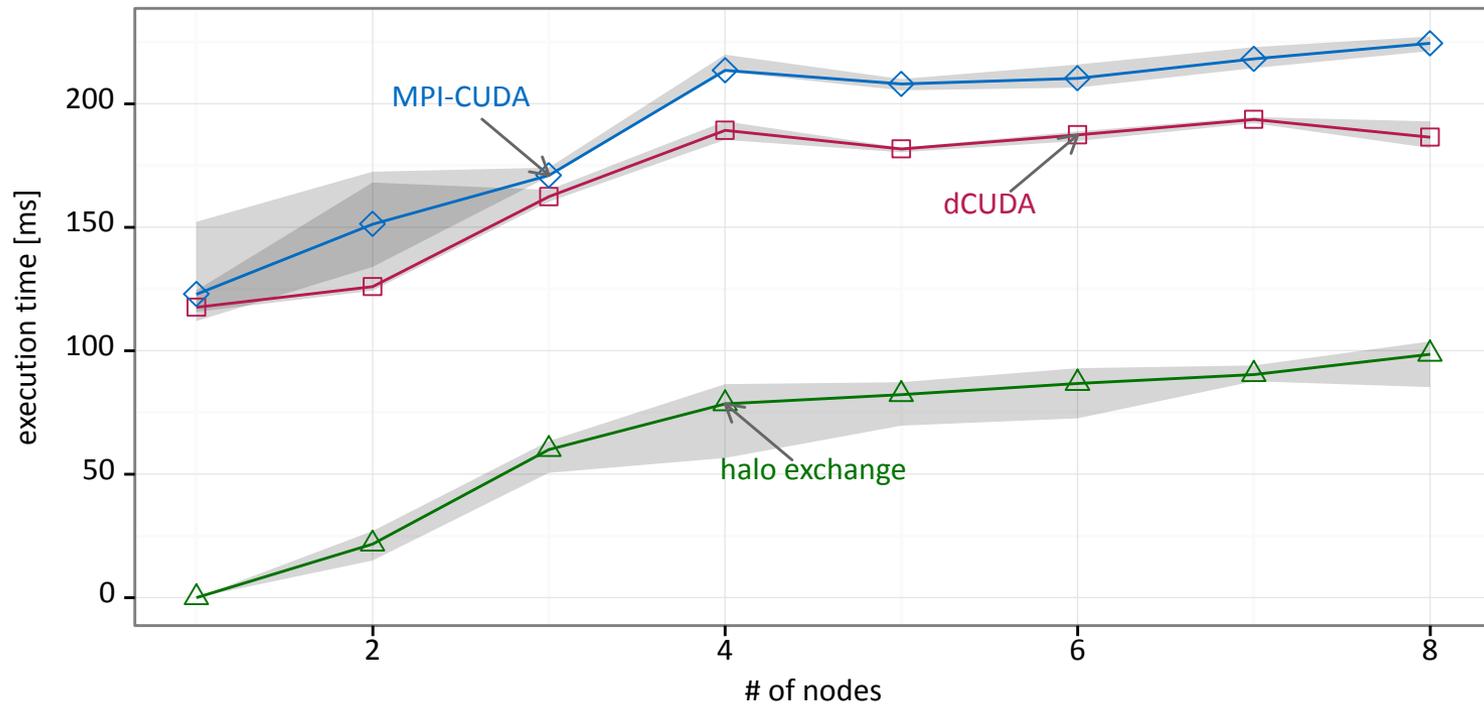
# Weak scaling of MPI-CUDA and dCUDA for a stencil program

- Benchmarked on Greina (8 Haswell nodes with 1x Tesla K80 per node)



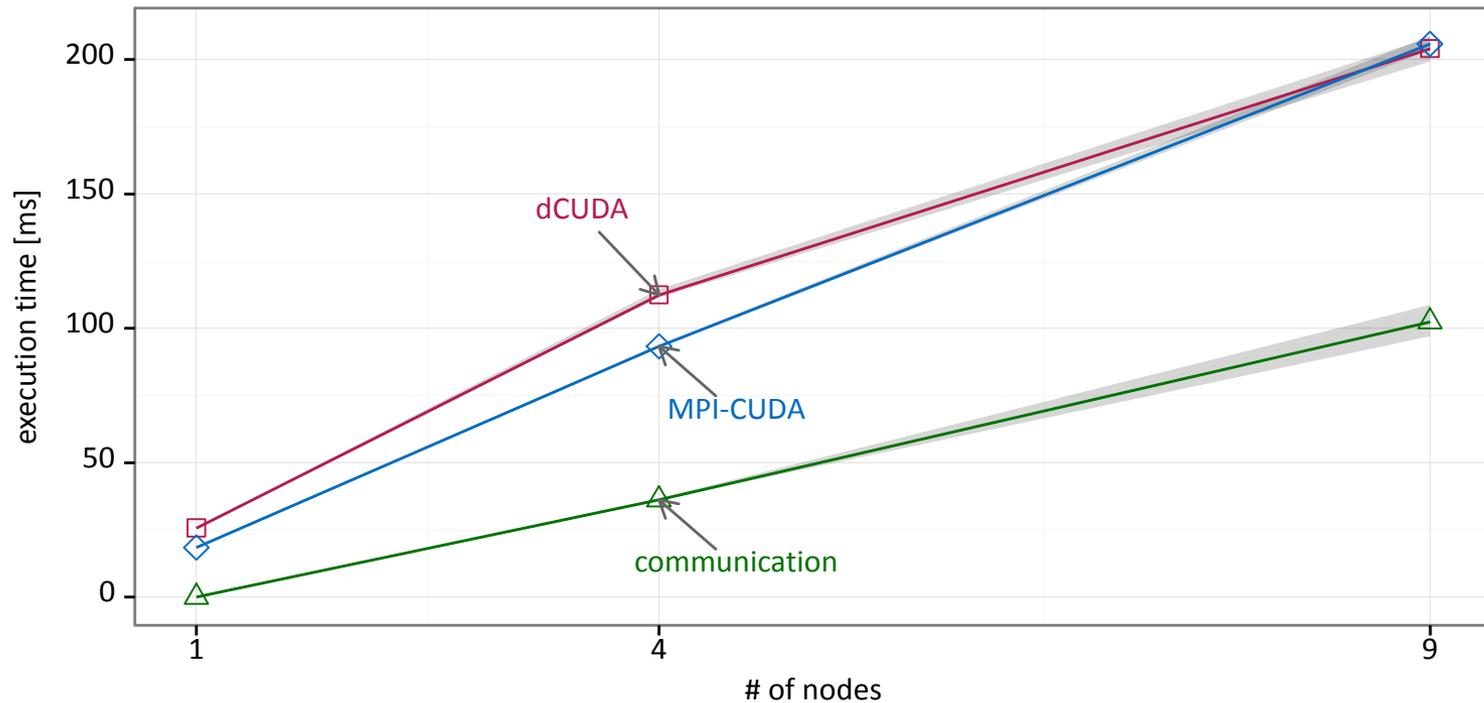
# Weak scaling of MPI-CUDA and dCUDA for a particle simulation

- Benchmarked on Greina (8 Haswell nodes with 1x Tesla K80 per node)



# Weak scaling of MPI-CUDA and dCUDA for sparse-matrix vector multiplication

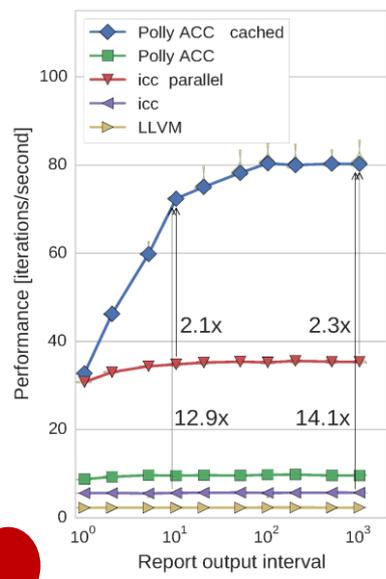
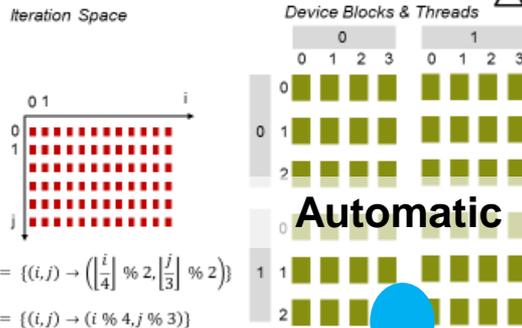
- Benchmarked on Greina (8 Haswell nodes with 1x Tesla K80 per node)



# http://spcl.inf.ethz.ch/Polly-ACC

# dCUDA – distributed memory

## Mapping Computation to Device

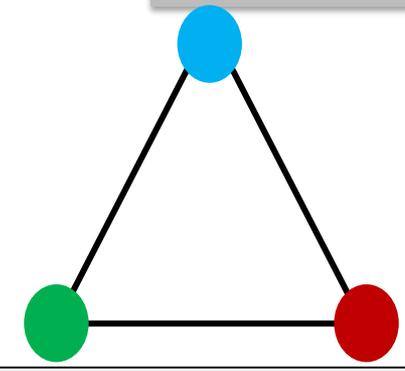
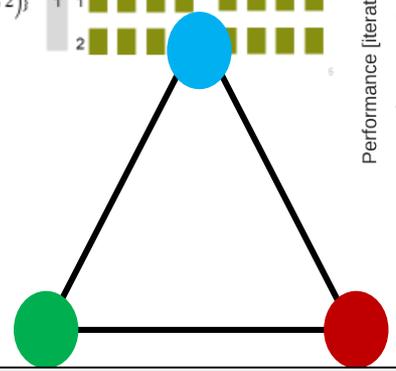


```
for (int i = 0; i < steps; ++i) {
  for (int idx = from; idx < to; idx += jstride)
    out[idx] = -4.0 * in[idx] +
              in[idx + 1] + in[idx - 1] +
              in[idx + jstride] + in[idx - jstride];

  if (lsend)
    dcuda_put_notify(ctx, wout, rank - 1,
                    len + jstride, jstride, &out[jstride], tag);
  if (rsend)
    dcuda_put_notify(ctx, wout, rank + 1,
                    0, jstride, &out[len], tag);

  dcuda_wait_notifications(ctx, wout,
                           DCUDA_WAIT_SOURCE, tag, lsend + rsend);
  swap(in, out); swap(win, wout);
}
```

**Automatic**



English Spanish French Detect language

English Spanish German Translate

a bad day for Europe

ein guter Tag für Europa

Suggest an edit