

Texture Cache Approximation on GPUs

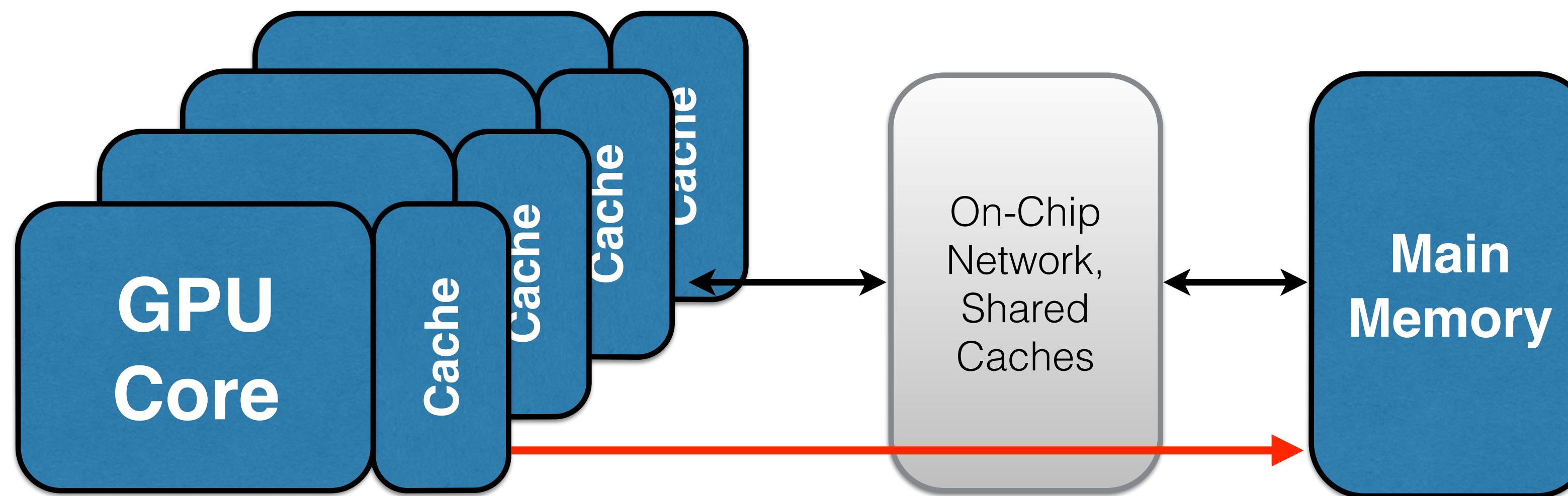
Mark Sutherland

Joshua San Miguel

Natalie Enright Jerger

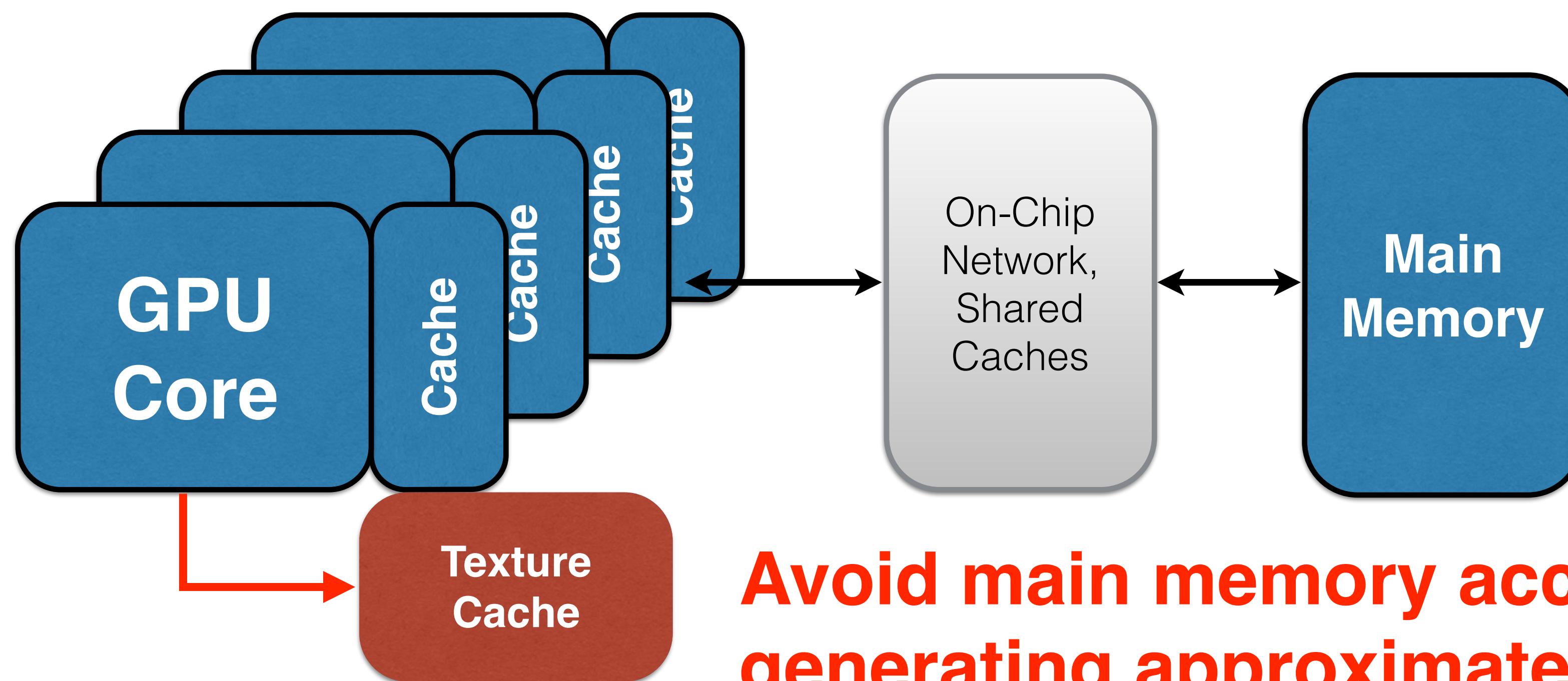
{suther68,enright}@ece.utoronto.ca, joshua.sanmiguel@mail.utoronto.ca

Our Contribution



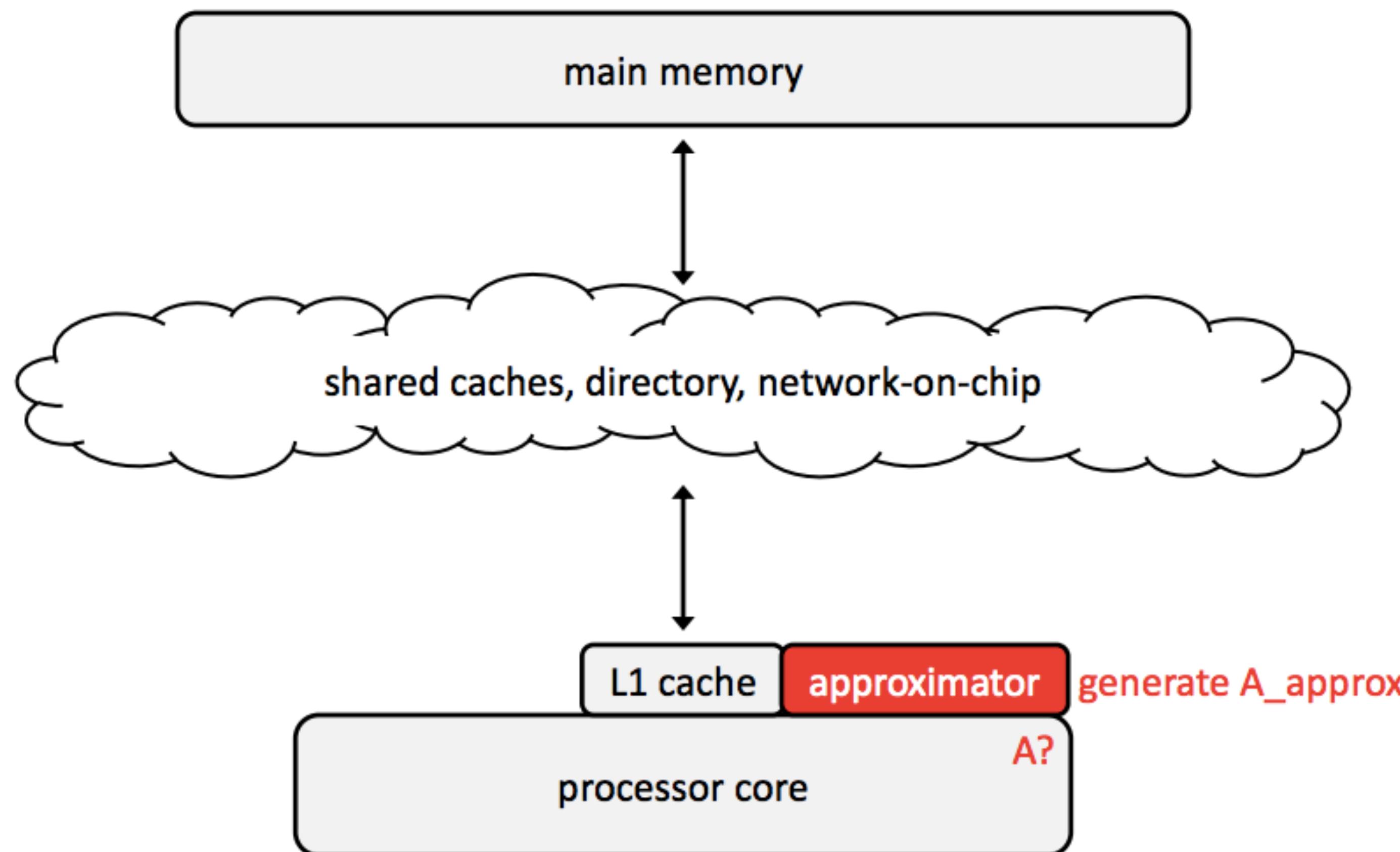
**Problem: High memory latency
requires thread swapping.**

Our Contribution

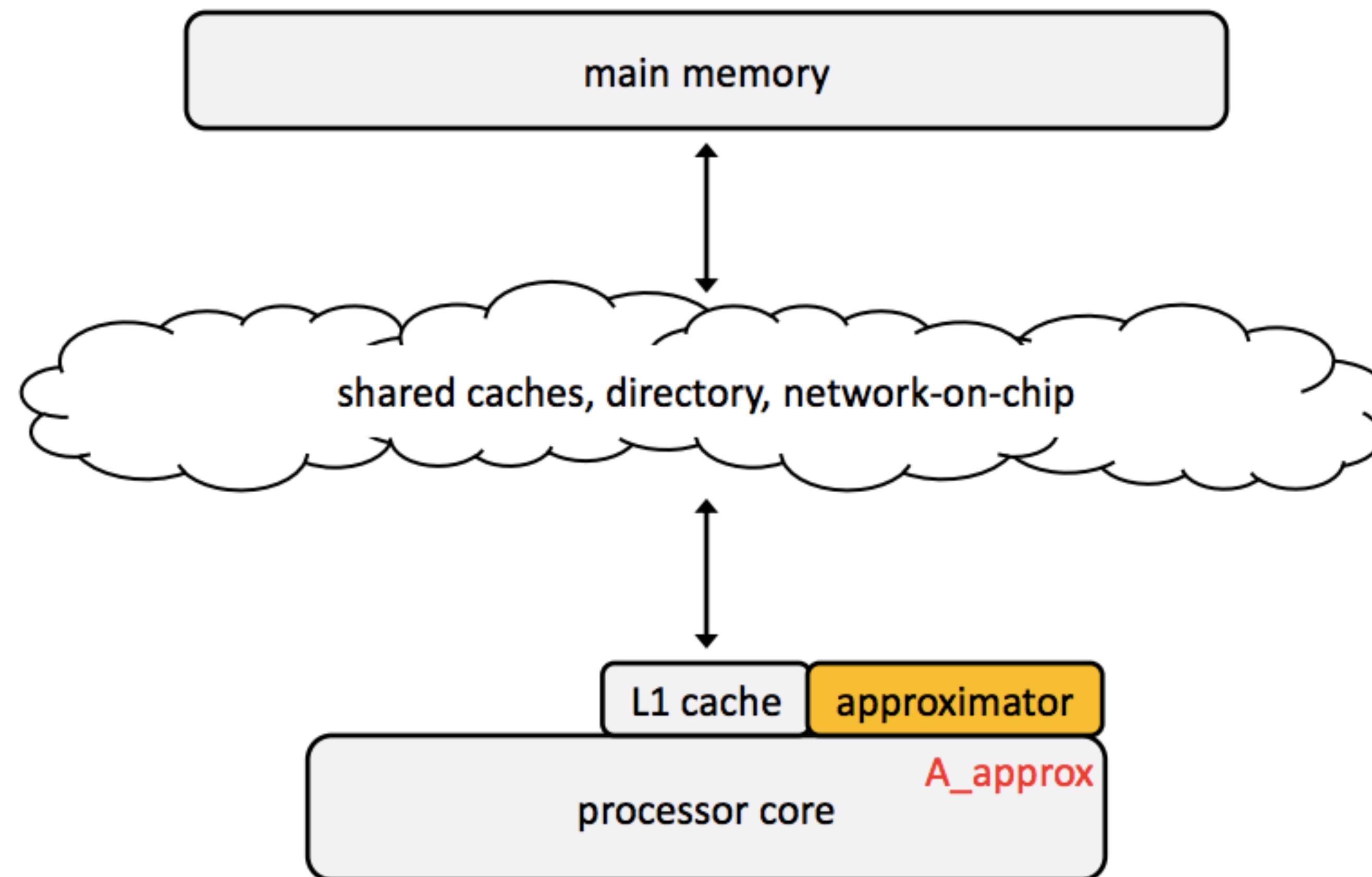


Avoid main memory accesses by generating approximate values on-chip.

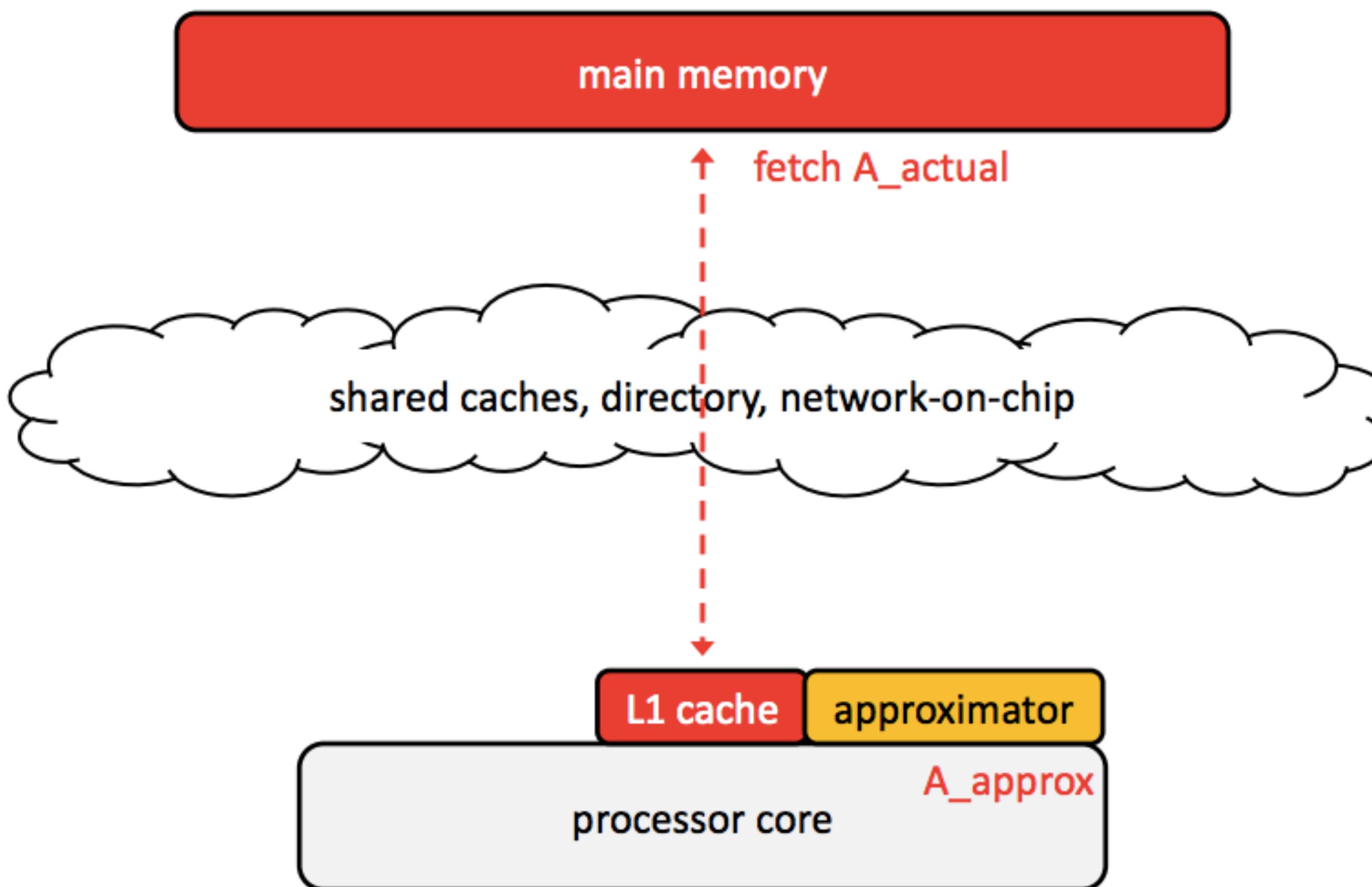
Load Value Approximation



Load Value Approximation



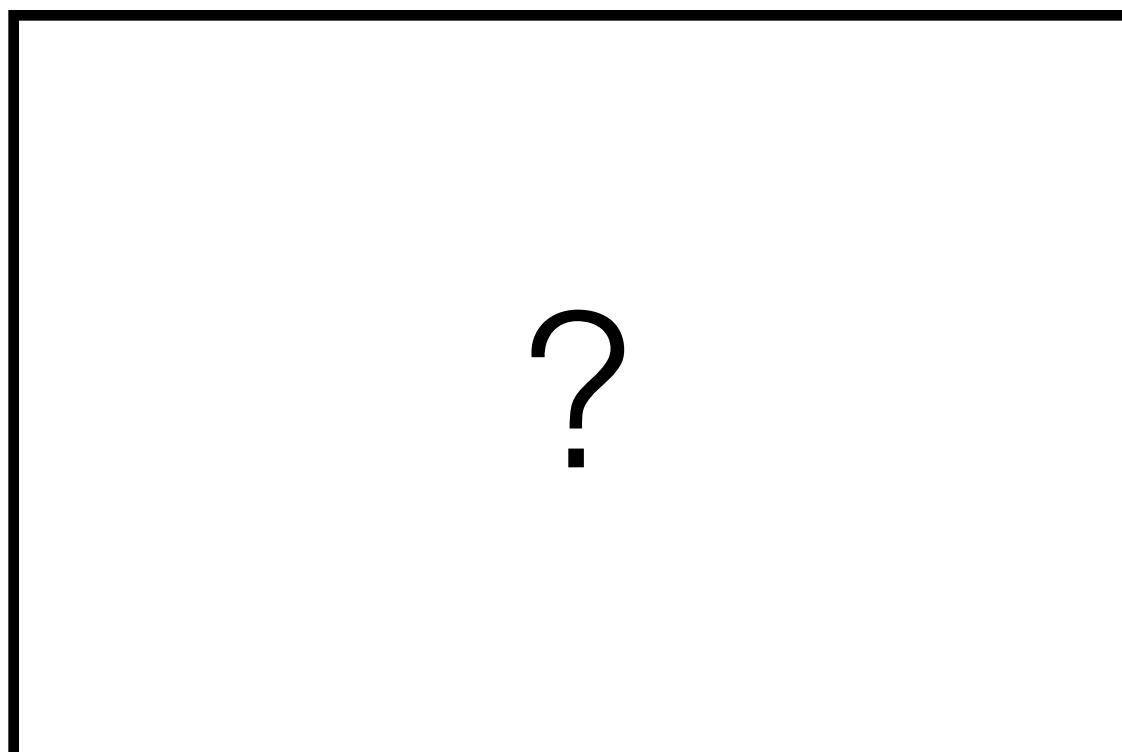
Load Value Approximation



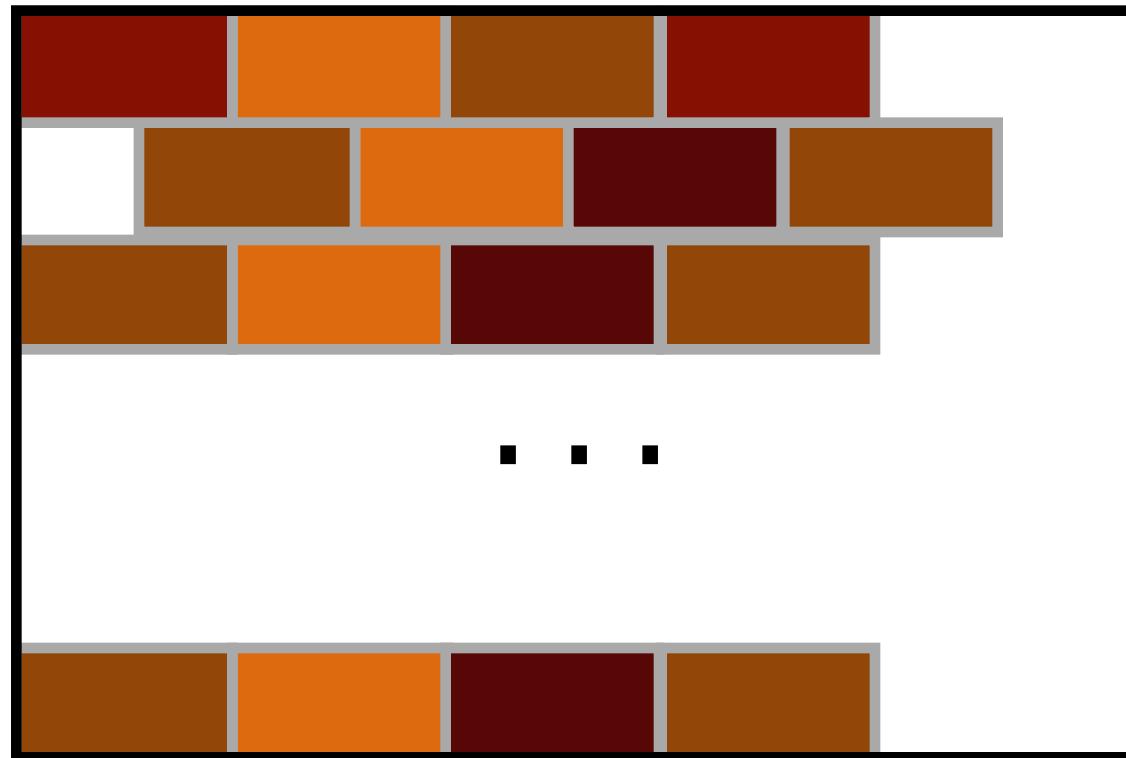
Goal: Use off-the-shelf GPU hardware to implement massively parallel value approximation.

What is a Texture?

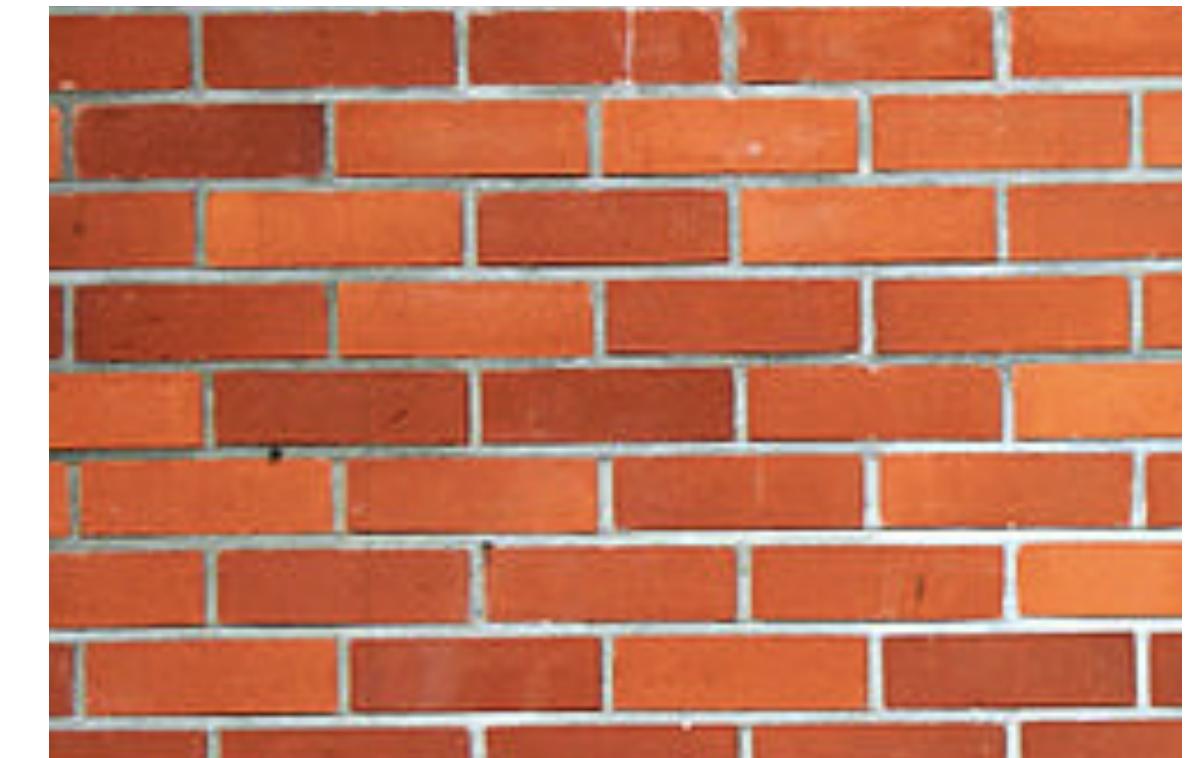
- Images are comprised of polygons, which the GPU fills using *shaders*.
- **Example:** Rendering a brick wall.
- Without a texture, the GPU can fill a polygon with a solid colour, or a gradient spanning from one vertex to another.
- Textures allow you to draw the whole wall, without filling N polygons for each brick.



Single-polygon (rectangle)



Multi-polygon solid



Single-polygon w. texture

Texture Hardware

Streaming
Multiprocessor
(Core)

- 12kB dedicated cache **per SM**, 4 dedicated fetch/interpolation units.
- Texture Unit Features:
 - Interpolating between data values, wrapping out-of-bounds indexes.
- Caveat: **Texture memory is read-only.**

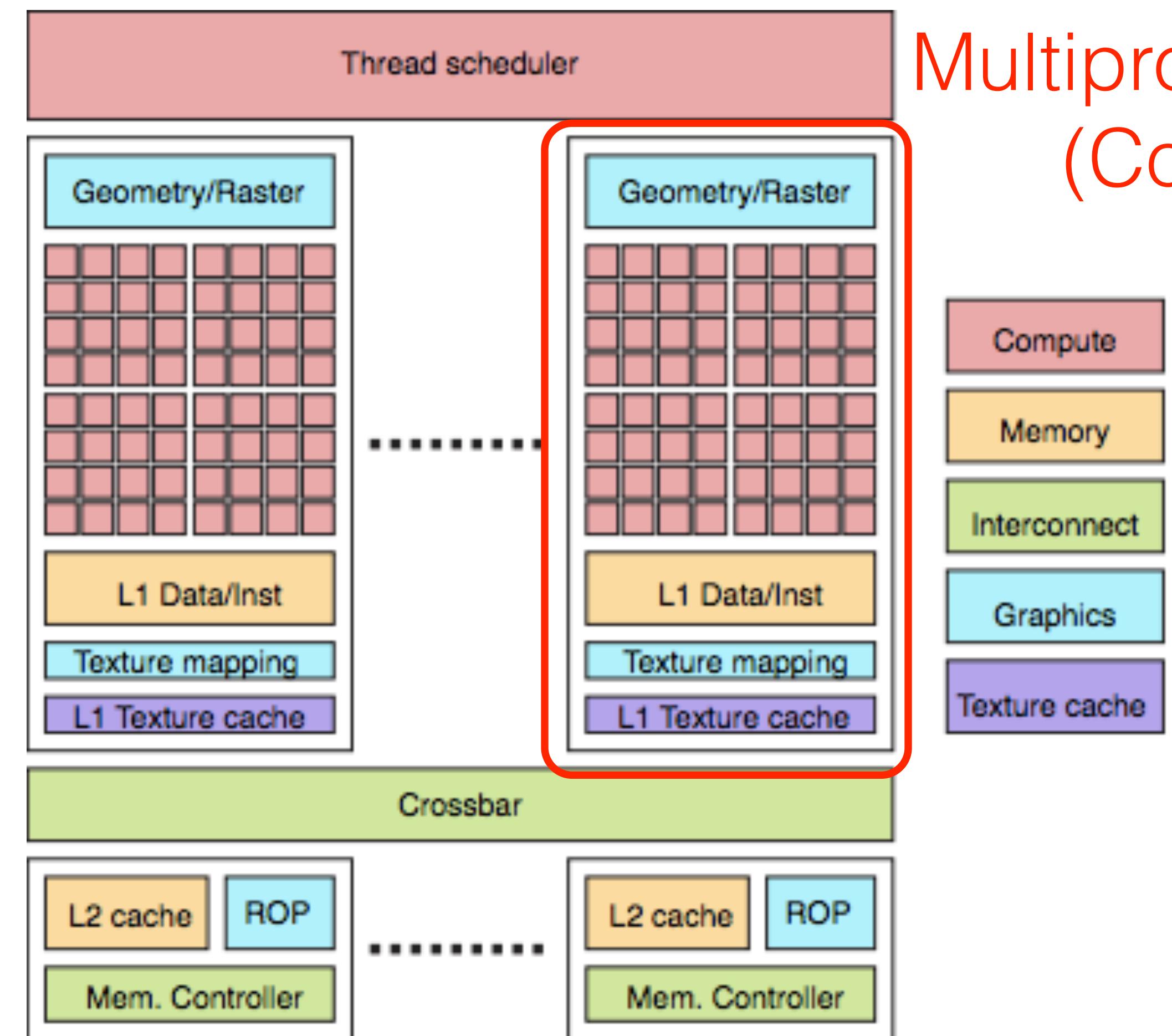


Figure source: M. Doggett. *Texture caches*. IEEE Micro, 32(3):136–141, May 2012.

Texture Cache Approximation

1. Analyze the data values read by each GPU thread, and build a training set from repetitive patterns in this data.
2. Before GPU execution, load the texture cache with this training set.
3. Replace memory accesses with approximations derived from the texture cache.

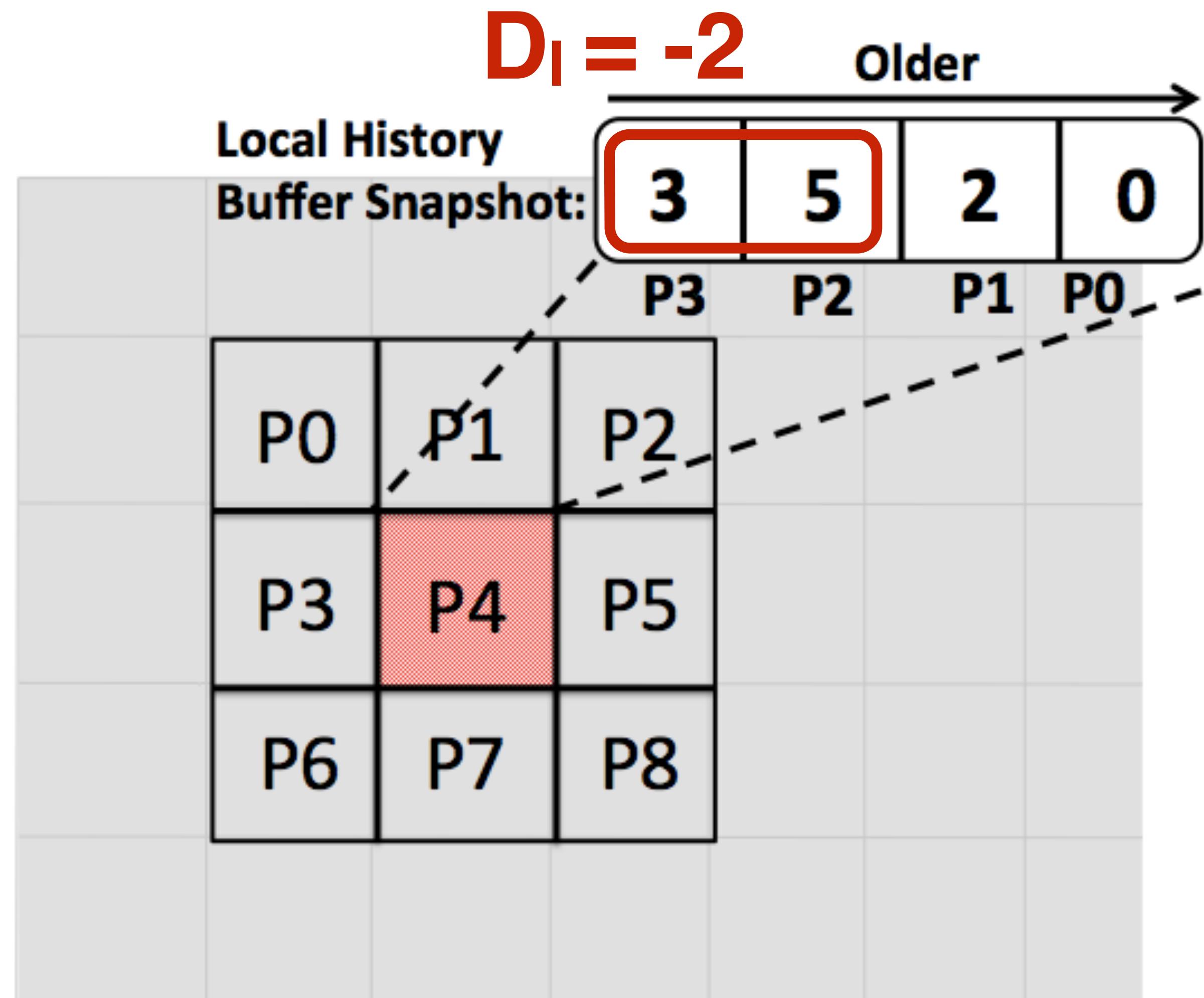
Cache Contents

- Want our approximations to be compact, so they fit in the texture cache, as well as portable to many value patterns.
- Each approximation is the sum of the last data value, and a **delta approximation** from the texture cache:
 - $X_{apx} = X_{last} + D_{apx}$

Training Set Generation

Example: Image access pattern

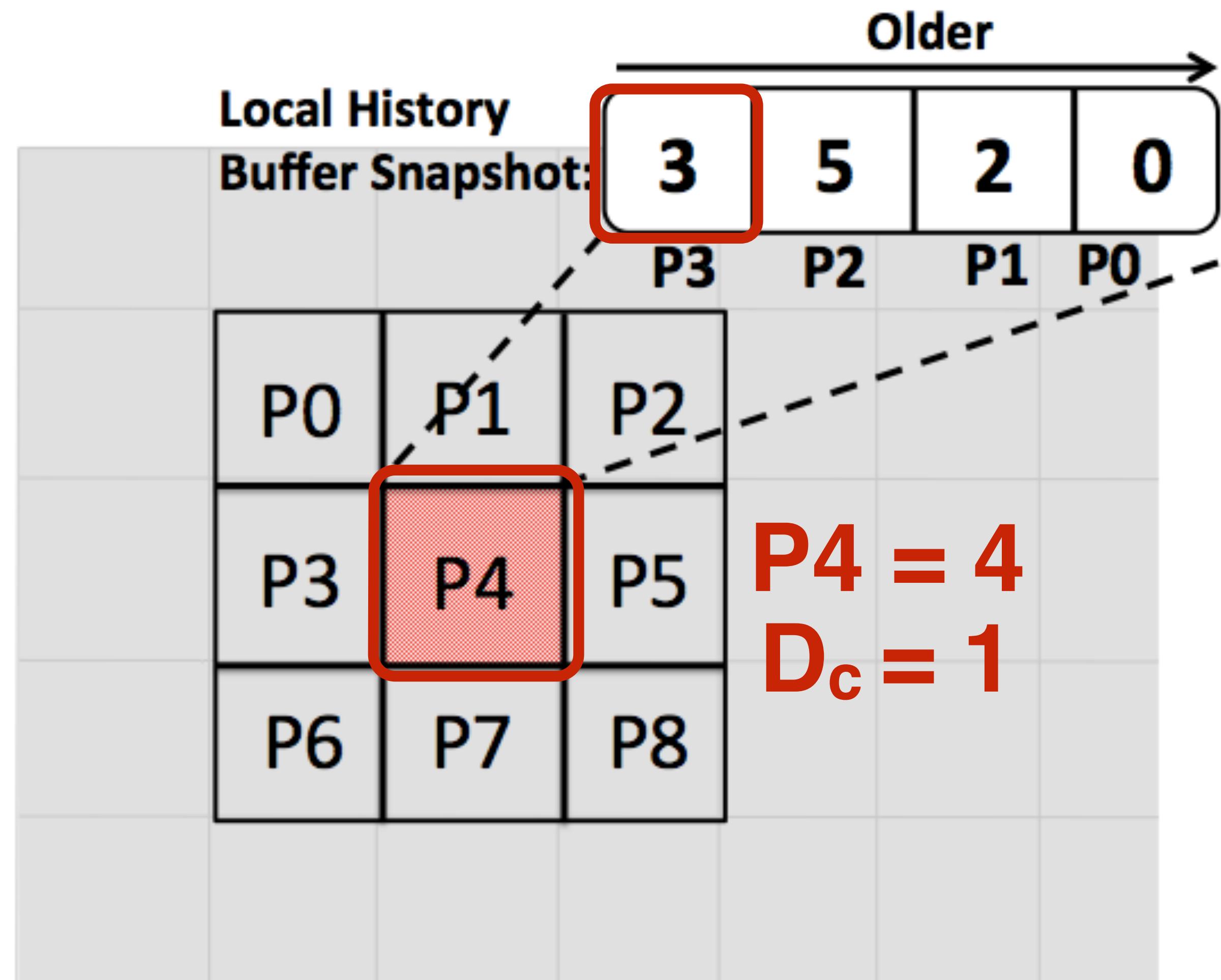
1. Upon accessing P4, thread inspects its LHB and calculates the **last** delta.
2. Baseline code then reads P4 from memory, and calculates the **current** delta.
3. Output a pair to the trace: $[D_l, D_c]$



Training Set Generation

Example: Image access pattern

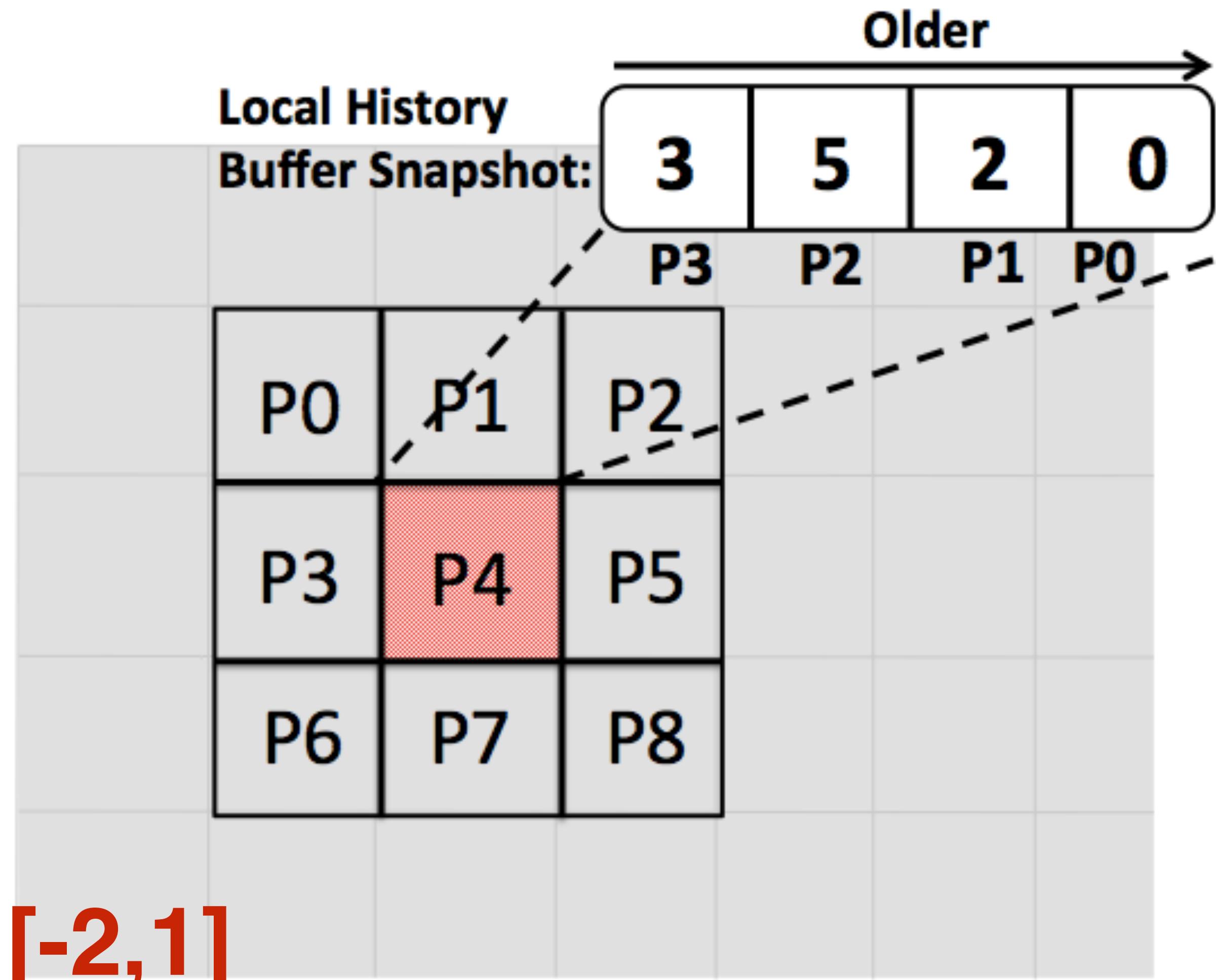
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Training Set Generation

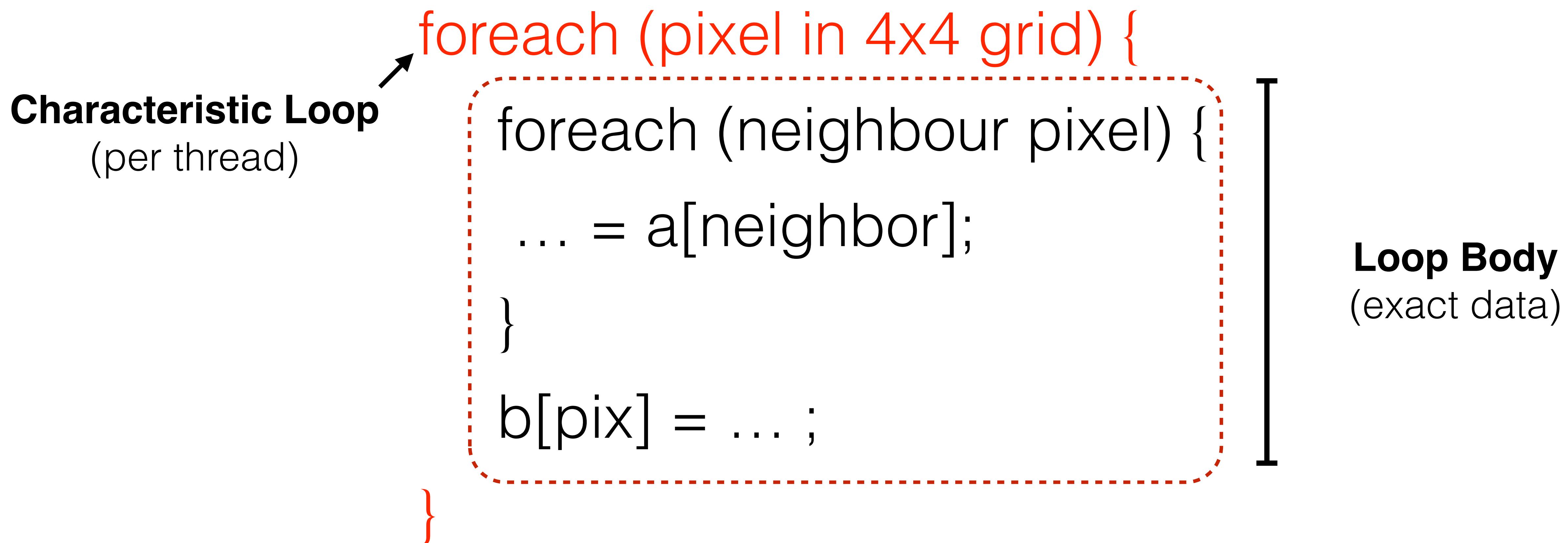
Example: Image access pattern

1. Upon accessing P4, thread inspects its LHB and calculates the **last** delta.
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3. Output a pair to the trace: $[D_l, D_c]$



$$[D_l, D_c] = [-2, 1]$$

Kernel Code Transformation



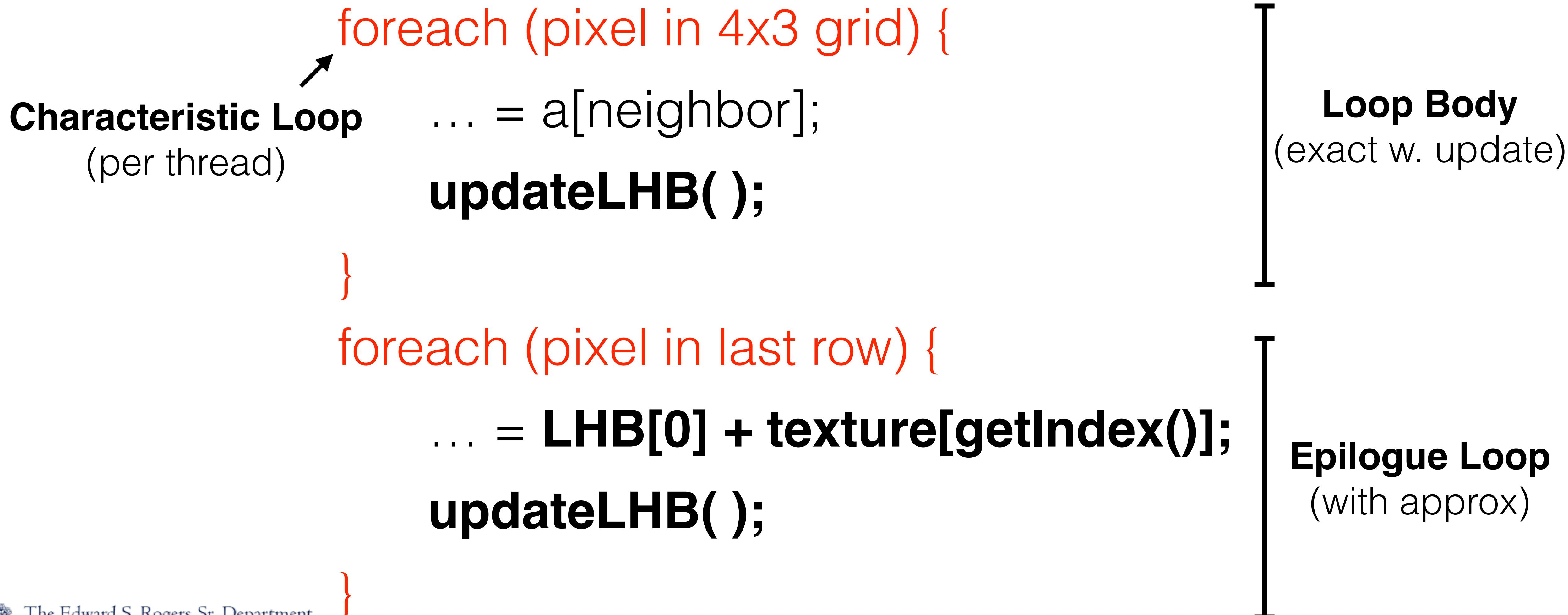
Kernel Code Transformation

```
foreach (pixel in 4x3 grid) {  
    ... = a[neighbor];  
    updateLHB();  
}  
  
foreach (pixel in last row) {  
    ... = LHB[0] + texture[getIndex()];  
    updateLHB();  
}
```

Characteristic Loop
(per thread)

Loop Body
(exact w. update)

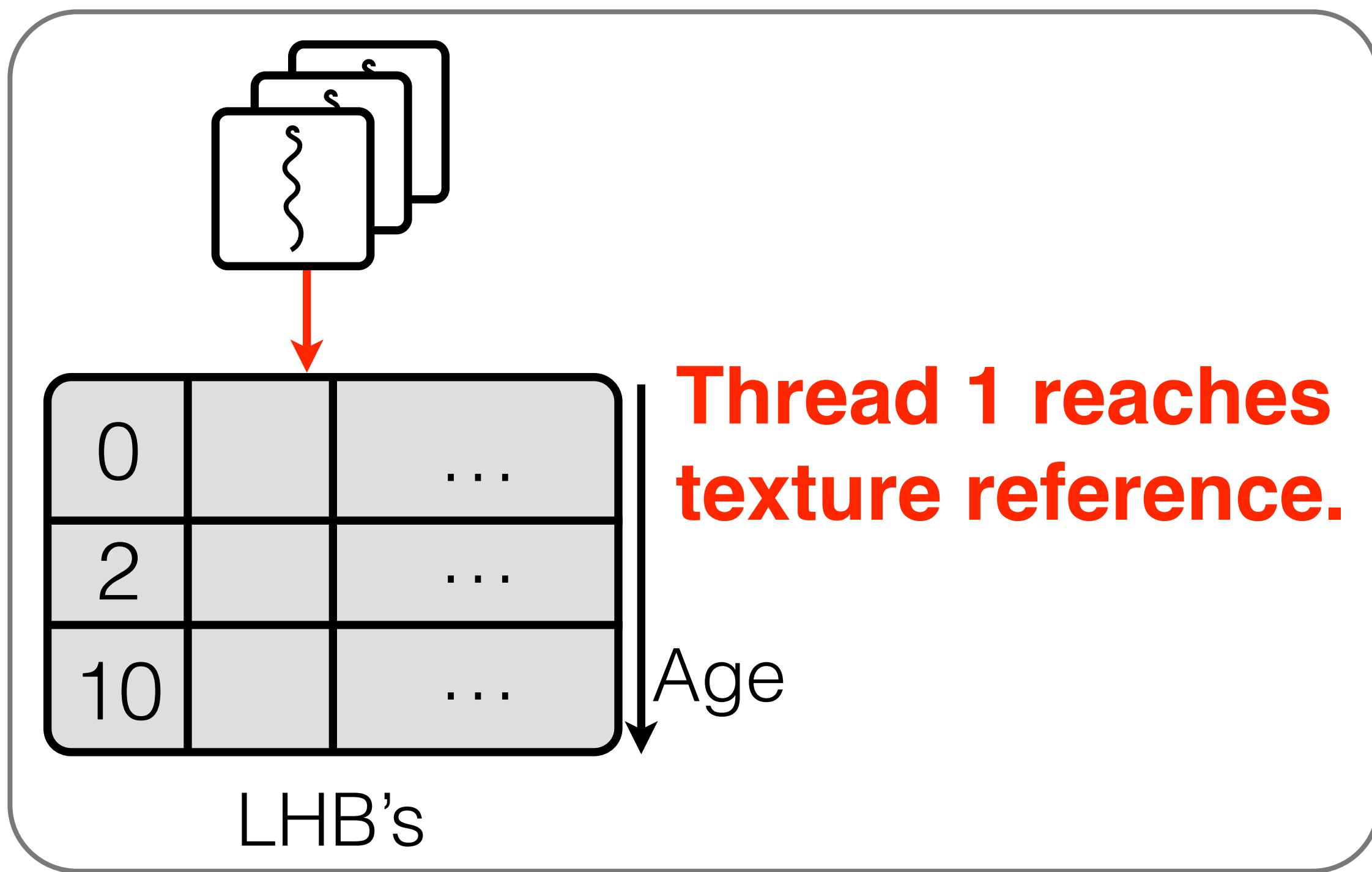
Epilogue Loop
(with approx)



Online Approximations

$\dots = \text{LHB}[0] + \text{texture[getIndex()]};$

In-Core Activity



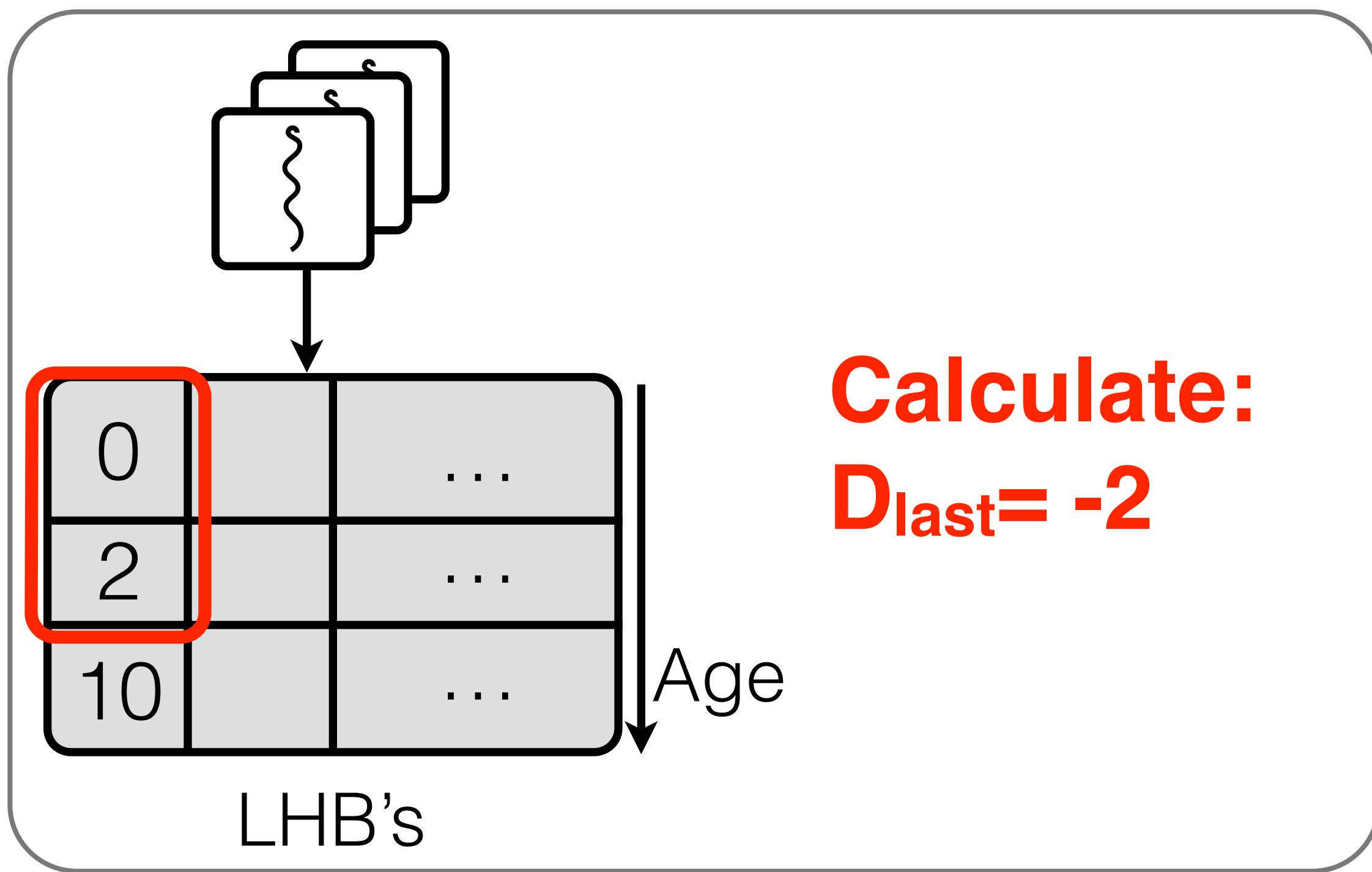
Texture Cache

Index	Value
0	1
1	4
2	8
3	-2

Online Approximations

$\dots = \text{LHB}[0] + \text{texture}[\text{getIndex}(\text{)}];$

In-Core Activity



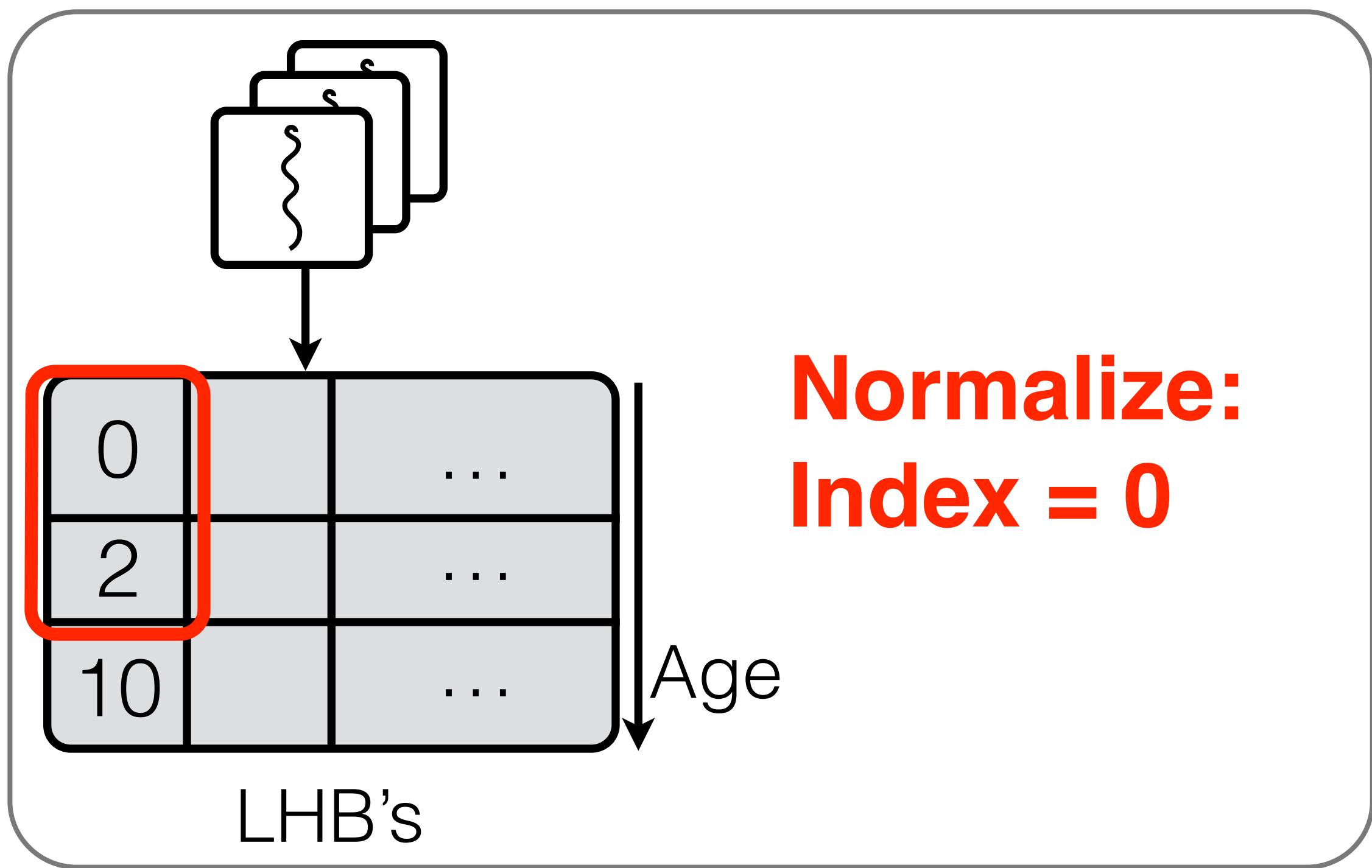
Texture Cache

Index	Value
0	1
1	4
2	8
3	-2

Online Approximations

$$\dots = \text{LHB}[0] + \text{texture}[\mathbf{0}];$$

In-Core Activity



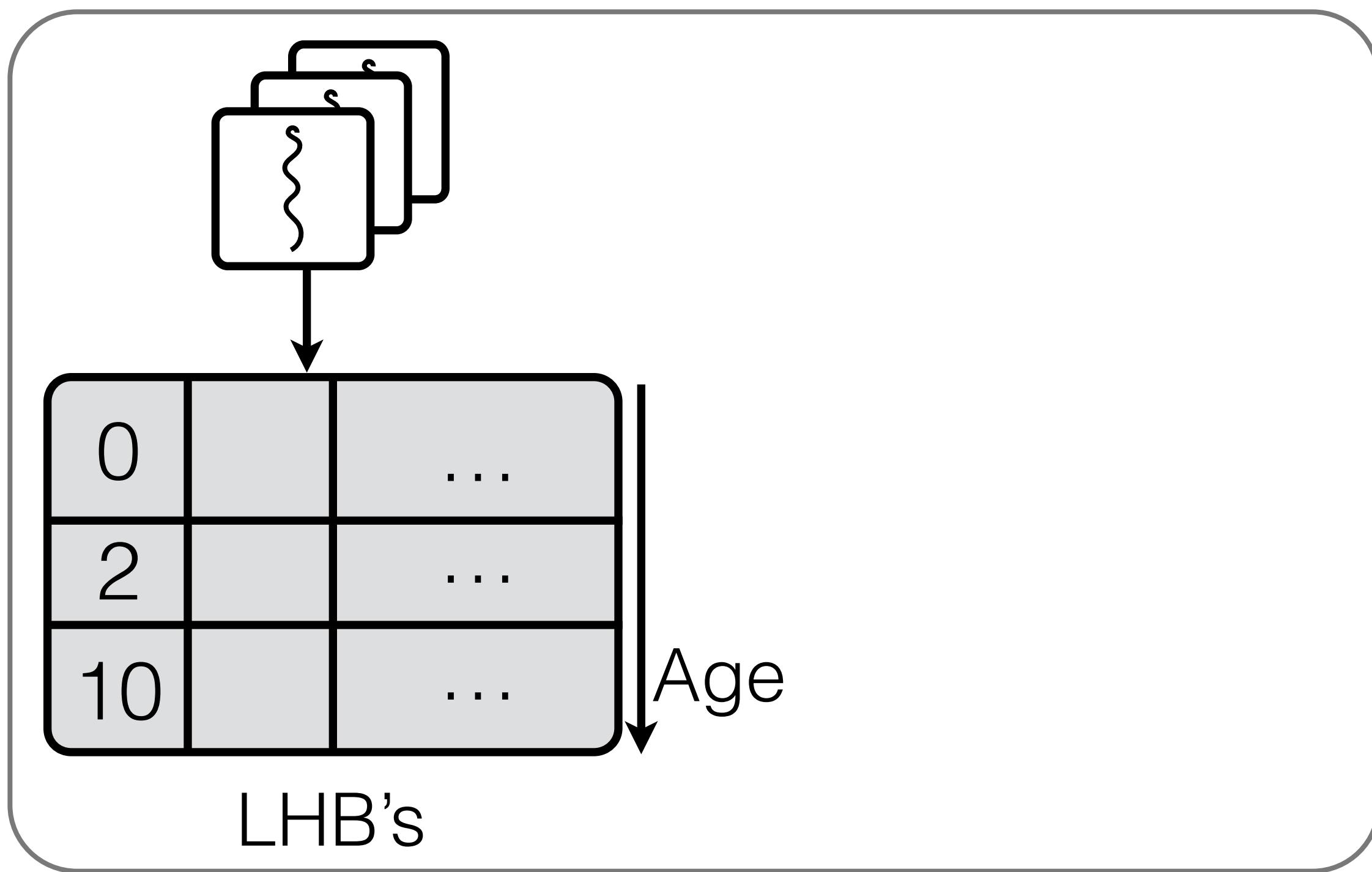
Texture Cache

Index	Value
0	1
1	4
2	8
3	-2

Online Approximations

$$\dots = \text{LHB}[0] + \text{texture}[0];$$

In-Core Activity



Texture Cache

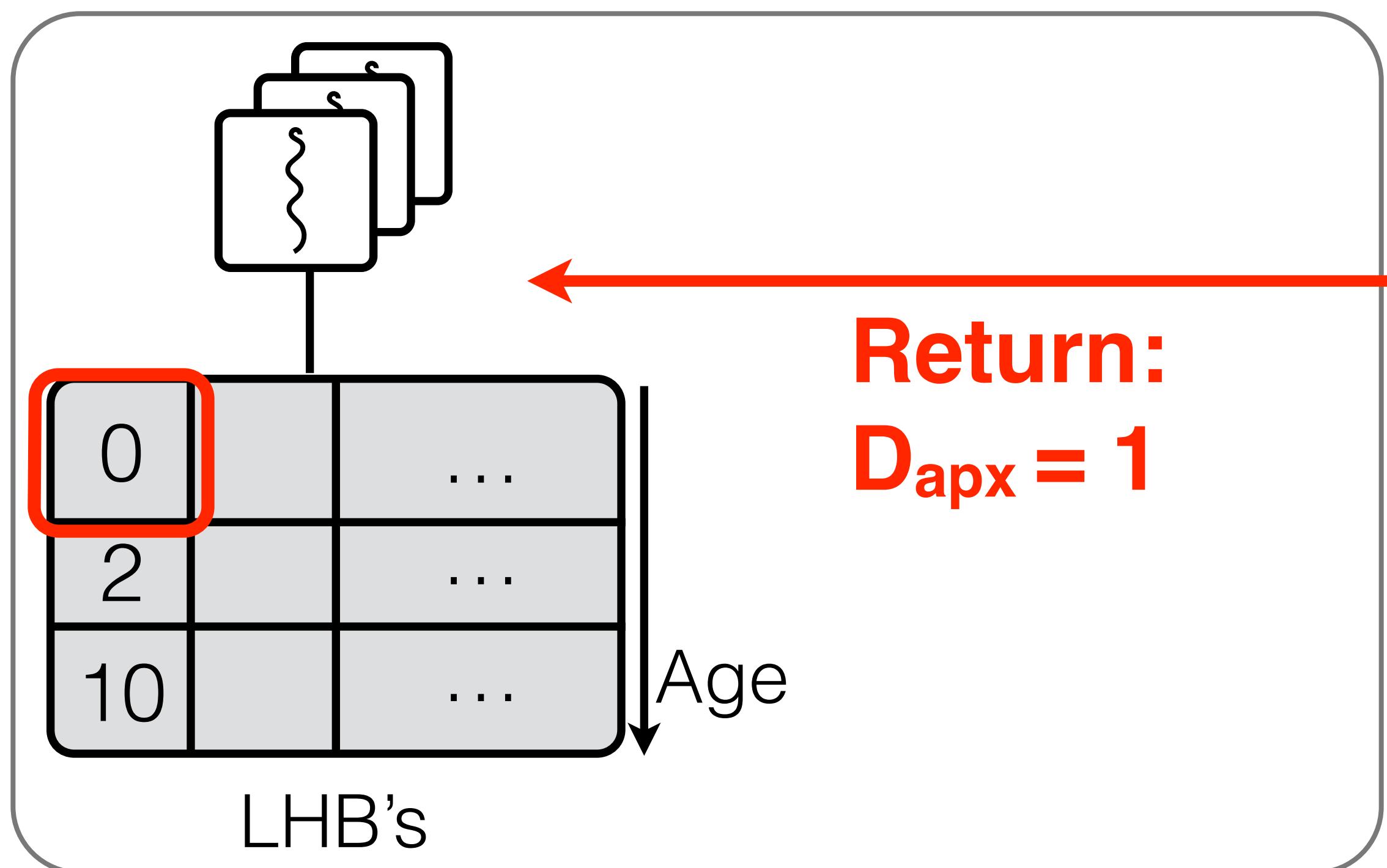
Index	Value
0	1
1	4
2	8
3	-2

Access:
texture[0]

Online Approximations

$$\dots = 0 + 1;$$

In-Core Activity



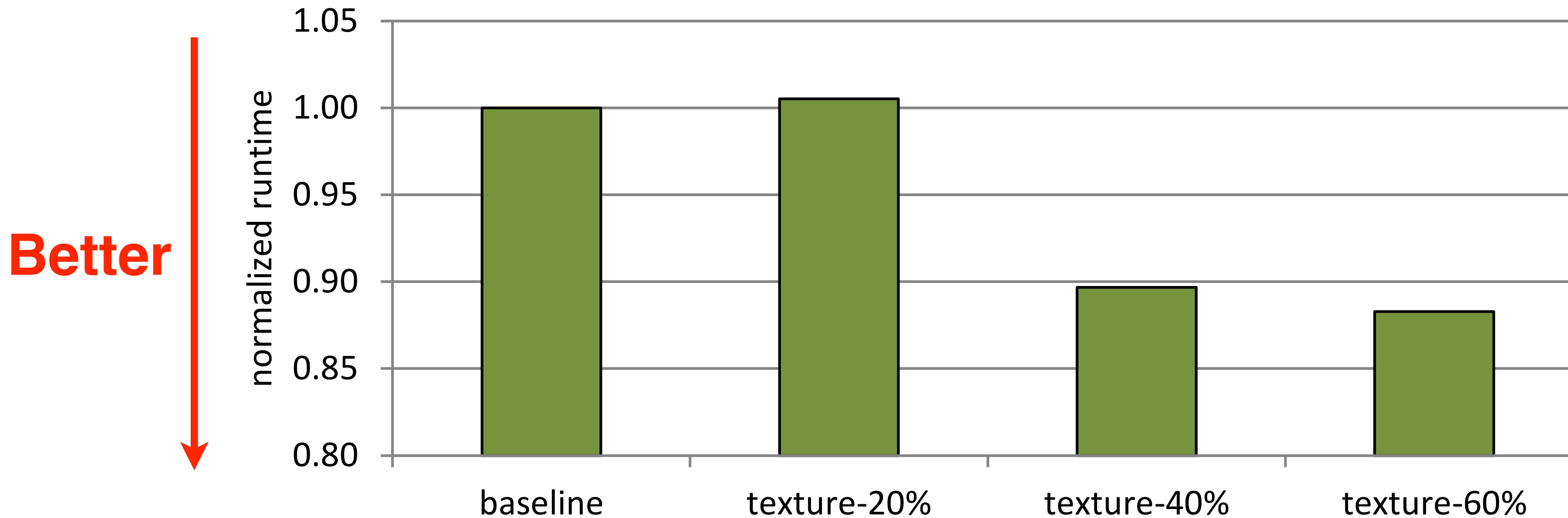
Texture Cache

Index	Value
0	1
1	4
2	8
3	-2

Evaluation

- Commodity hardware: NVIDIA 780GTX GPU (Kepler u-architecture)
- Image blur kernel derived from San Diego Computer Vision Benchmark Suite [Venkata, IISWC '09].
- Use CUDA Toolkit Profiler to measure:
 - Kernel runtime (cycles), texture cache hit rate.
 - Error metric: Mean Pixel Difference [Samadi, MICRO '13]

Kernel Runtime



- Replaced X global memory reads (baseline has 5).
- Texture cache hit rate: **> 99%** in all cases.

Image Comparison

Exact

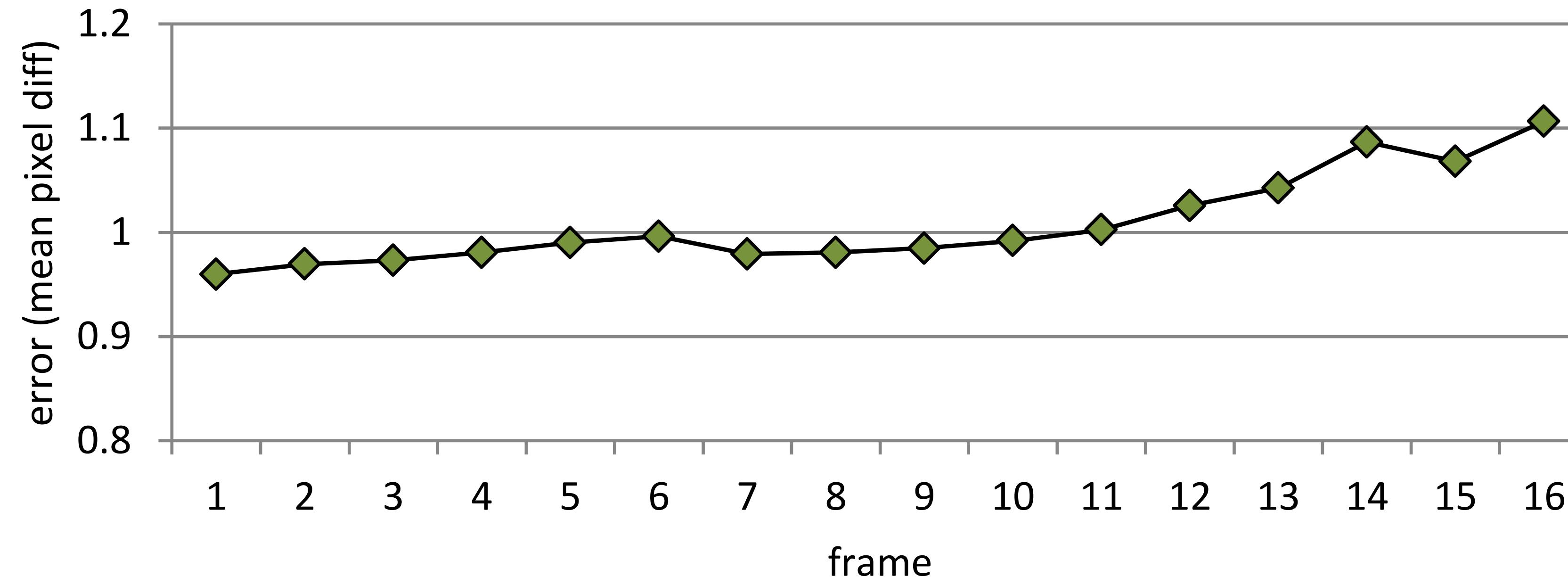


Approximate



Error: 0.4%

Error Evolution



- Used same training set (from F1) for 16 images in a video sequence, evaluated error with 40% of loads replaced.

Future Work

- Evaluate on applications from different application domains: machine learning, physics & fluid simulations, data queries.
- Can we eliminate the need for training sets?
- Improve speedup on floating point benchmarks.

Conclusion:

Under-utilized texture hardware on GPUs can be used to accelerate kernel execution using value approximation.

?

Comparison to Reduced Blur

Large Radius



Small Radius



Approximate Value Computing

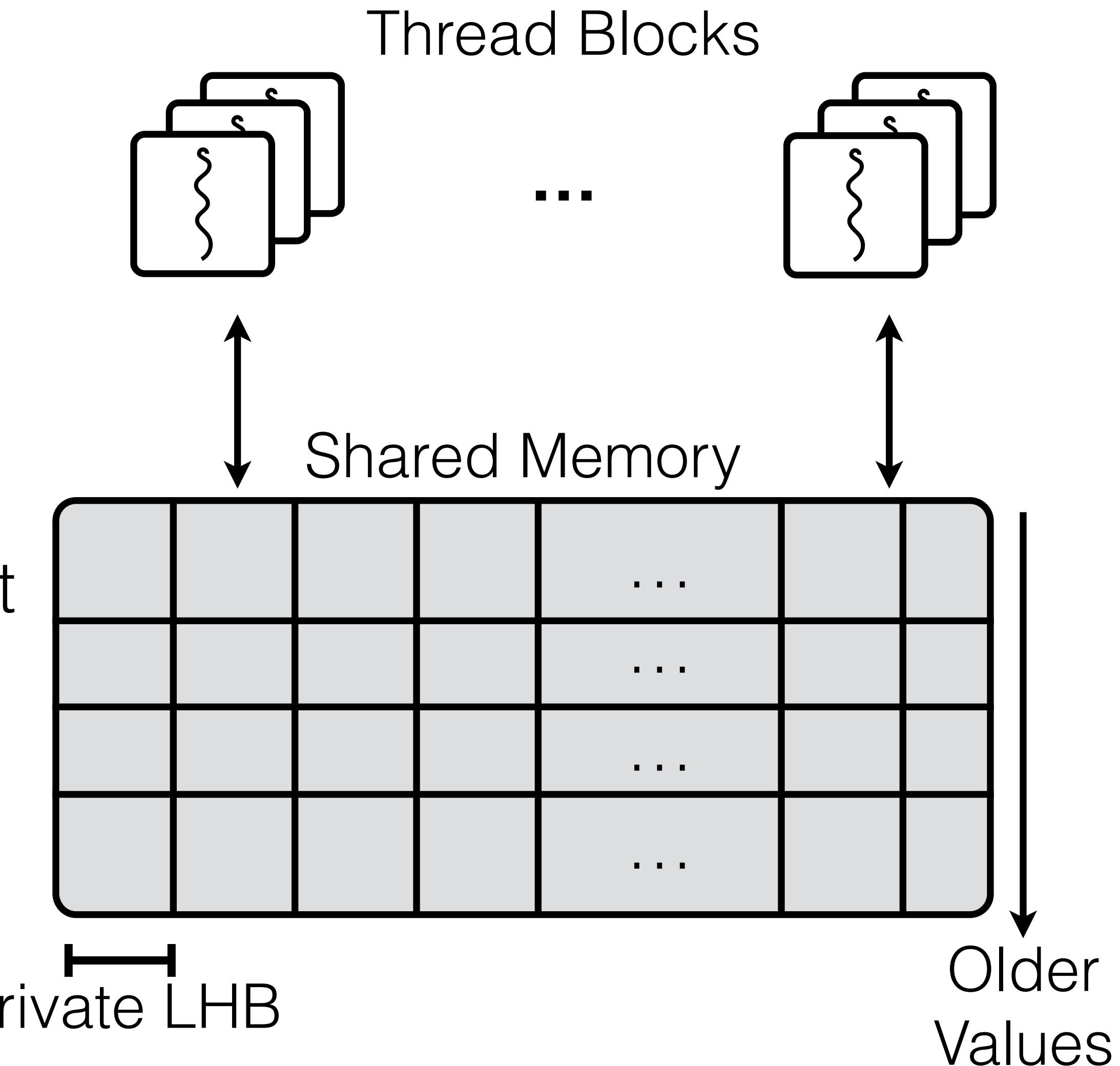
1. Certain applications are robust to inexact data values.
 - Data mining and pattern recognition [Chippa, DAC '13]
2. Where can these values come from?
 - Reduced-voltage DRAM [Liu, ASPLOS '12]
 - Kernels edited to remove synchronization [Samadi, MICRO '13]
 - Load Value Approximation [San Miguel, MICRO '14]

Why Approximate?

1. GPU memory architecture has similar drawbacks to that of a CPU.
2. Modern GPU's choose to hide latency with thread swapping and context storage.
 - **Could easily put these transistors to better use!**

Training Set Generation

1. Give every thread its own Local History Buffer (LHB) in shared memory.
2. Upon every memory access, output the **previous** and **next** deltas to a trace.
 - This allows us to analyze different **value** patterns, and generate approximations that are accurate for many thread blocks.



Kernel Code Transformation

Characteristic Loop
(per thread)

```
foreach (pixel in 4x3 grid) {  
    foreach (neighbour pixel) {  
        ... = a[np];  
        updateLHB(a[np]);  
    }  
    b[p] = ... ;  
}  
foreach (pixels left in 4x1 stripe) {  
    ...  
}
```

Loop Body
(exact w. update)

**Epilogue Loop
Body**

Kernel Code Transformation

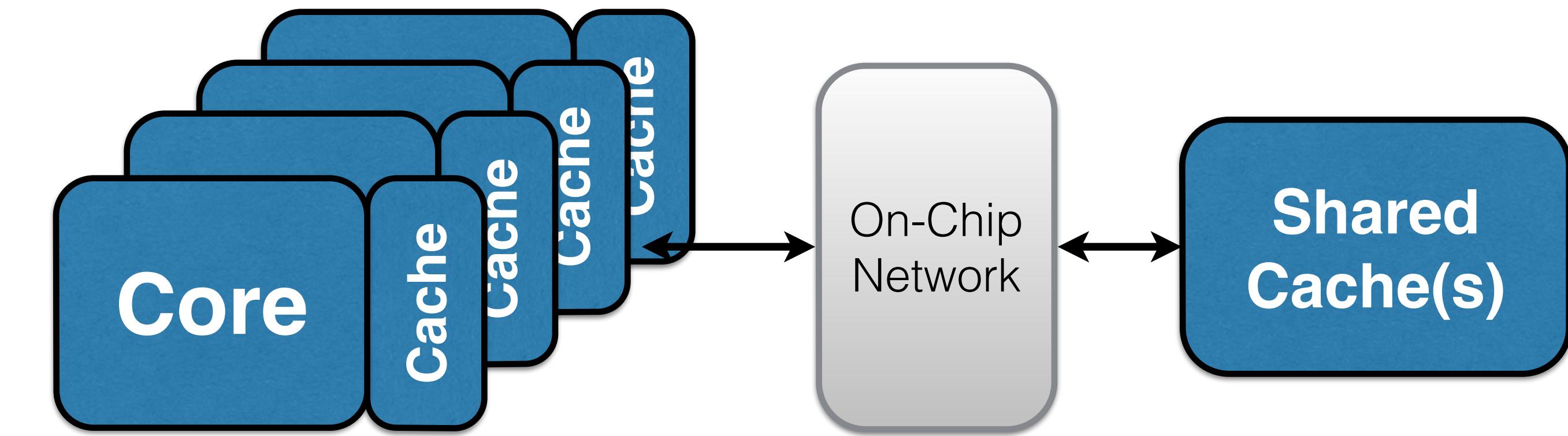
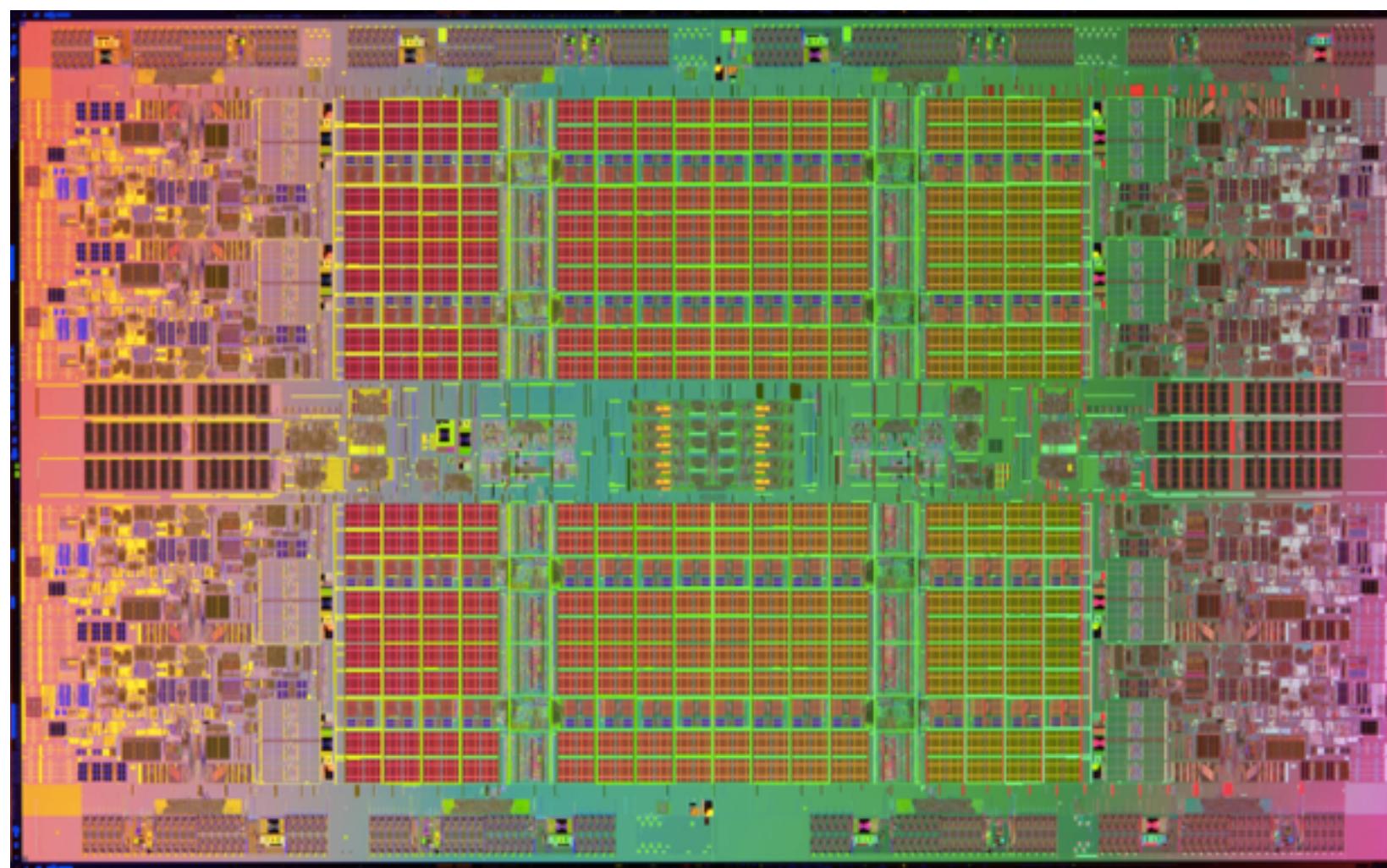
```
foreach (pixel in 4x3 grid) {  
    ...  
    updateLHB(a[np]);  
}  
foreach (pixels left in 4x1 stripe) {  
    foreach(neighbour pixel) {  
        ... = getTexture(my_tex, getDeltaFromLHB( ) );  
        updateLHB(...);  
    }  
    b[p] = ... ;  
}
```



Loop Body
(exact w. update)

Epilogue Loop Body
(replace loads with
texture references)

Processing Continuum



Accelerator¹

Multi-Core CPU

GPU Niche

GPGPU



Characteristics

- Massively multi-core.
- Trades latency for throughput.
- Very few optimizations for single-thread performance.

