## Footprint Modeling of Cache Associativity and Granularity

Hao Luo<sup>1</sup> (UR), Guoyang Chen<sup>2</sup> (NCSU), *Fangzhou Liu* (UR), Pengcheng Li<sup>1</sup> (UR), Chen Ding (UR), Xipeng Shen (NCSU)

University of Rochester, Rochester, NY North Carolina State University, Raleigh, NC



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1. Now at Google Inc. This work was done when the student was a graduate student

2. Now at Alibaba Group US Inc. This work was done when the student was a graduate student

#### Motivation Data placement problem



Array in a program

3X performance difference

### Motivation

PORPLE<sup>1</sup> - Online Optimization of Data Placement



1. Guoyang Chen, Bo Wu, Dong Li and Xipeng Shen. 2014. PORPLE: An Extensible Optimizer for Portable Data Placement on GPU. *In Proceedings of MICRO.* 

## Motivation

- Data Non-uniformity
  - A program uses multiple arrays, which have different access frequency and locality.
- Uneven Address Mapping
  - Data addresses may be mapped unevenly to cache sets.
  - Cache can be reconfigured to different capacity
    - Which changes address mapping
- Dual-grained Cache
  - Data was fetched and managed in different granularity.



- Footprint Metric
- Three footprint models
- Evaluation

## Background

Footprint Metric

• Trace is a series of memory accesses (interleaved)

- Working Set Size is the number of distinct elements within a window (fixed-length time interval)
- Footprint is the average number of distinct items accessed in windows of length "x"

$$fp(x) = \frac{1}{N-x+1} \sum_{t=x}^{N} \omega(t,x)$$

### Background Footprint Metric



• **Footprint** is the average number of distinct items accessed in windows of length "x"

Window length	1	2	3	4	5	≥ 6
Footprint	1	2	2.75	3	3.5	4

#### Background Denning-HOTL Conversion

• The **Cache Miss ratio** can be computed by the derivative of footprint function.



1.Xiaoya Xiang, Chen Ding, Hao luo, and Bin Bao. 2013. HOTL: a higher order theory of locality. *In Proceedings of ASPLOS*. pp 343-356



- Footprint Metric
- Three new footprint models
  - Nonuniform data locality -> Partial Footprint (This talk)
  - Uneven address mapping -> Mapped Footprint (See paper)
  - Dual-grained Cache -> Dual-grained Footprint (See paper)
- Evaluation

#### **Partial Footprint** *Time-preserving trace (TP)*

• **Time-preserving trace** is a copy of the original trace with 0 or more accesses replaced with placeholders.



Partial Footprint Model

• Let  $s_i$  be a TP trace, its **Partial Footprint** is defined as:

$$pfp(x) = \frac{1}{N-x+1} \sum_{t=x}^{N} \omega(s_i, t, x)$$
TP:  $a[1] - a[6] - a[6] - a[6]$ 

Let Ω = {d<sub>i</sub>} be the set of arrays used by the program. We can define the partial footprint of the TP trace that preserves array d<sub>i</sub>, denoted as pfp(d<sub>i</sub>, x).

#### Partial Footprint Model

	TP:	a[1]		- a[6]	—		
Window length	1	2	3	4	5	≥ 6	
Partial Footprint	1/3	3/5	1	4/3	3/2	2	
Γ.							
Cache Size	1	2	3	4	5	≥ 6	
Miss Ratio	4/15	2/5	1/3	1/6	1/2	0	

#### All-group Cache Performance

- The all-group cache miss ratio can be calculated in two ways
  - 1. Compute the all-group cache miss ratio using the Denning-HOTL conversion.



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$$mr(\Omega, c) = \frac{d}{dx} pfp(\Omega, x) \Big|_{pfp(\Omega, x) = c}$$

2. Compute the cache miss ratio for each data item and then sum them up.

$$mr(d_i, c) = \frac{d}{dx} pfp(d_i, x) \Big|_{pfp(\Omega, x) = c}$$
$$mr(\Omega, c) = \sum_{d_i \in \Omega} mr(d_i, c)$$

The equality of this two solutions shows the Composition Invariance.



- Footprint Metric
- Three footprint models
- Evaluation

#### Evaluation Setup

- 13 benchmarks from SHOC and CUDA Code Samples
- Traces were collected on Tesla M2075 GPU
- 48KB 6-way L1 Cache and 12 KB, 8-way Texture Cache
- Compared our models with Simulation (Baseline), and two modeling techniques, Set-RD<sup>1</sup> and PORPLE<sup>2</sup>

<sup>1.</sup> Rathijit Sen and David A. Wood, 2013, Reuse-based online models for caches. *In Proceedings of SIGMETRICS*. pp279-292

<sup>2.</sup> Guoyang Chen, Bo Wu, Dong Li and Xipeng Shen. 2014. PORPLE: An Extensible Optimizer for Portable Data Placement on GPU. *In Proceedings of MICRO.* 

## **Evaluation**

#### **Experiment Result**

	Simulator	Footprint	PORPLE	Set-RD
Accuracy	100%	99.2%	96.6%	100%
Cost	14.0	1.6	4.7	524

PORPLE is **111** times faster than Set-RD, and Footprint analysis is **3** times faster than PORPLE.

The arithmetic average across the 13 benchmarks is **0.0%** for Set-RD, **3.4%** error for PORPLE and **0.8%** for Footprint, which is **4.3 times** smaller error than PORPLE.



Three models solve Data Non-uniformity, Uneven Address
Mapping and Dual Granularity.



Predict the per-array cache misses on heterogeneous memory system with all possible cache sizes.

Experiment results show lower overhead and higher accuracy.
 327 times faster than Set-RD, 3 times faster than PORPLE
 99.2% accurate, 4.3 times lower error than PORPLE

### Thanks for Listening Any Questions?