

JIGSAW: Scalable Software-Defined Caches

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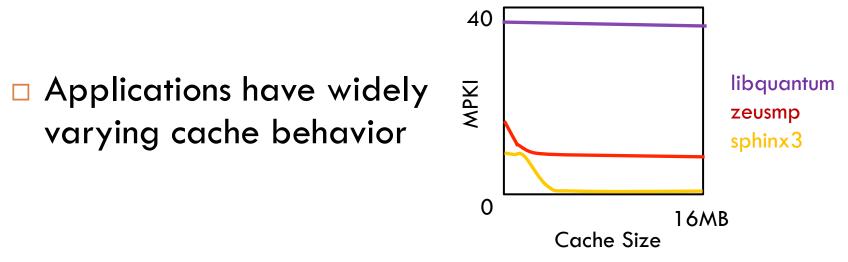
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Massachusetts Institute of Technology



Summary

□ NUCA is giving us more capacity, but further away

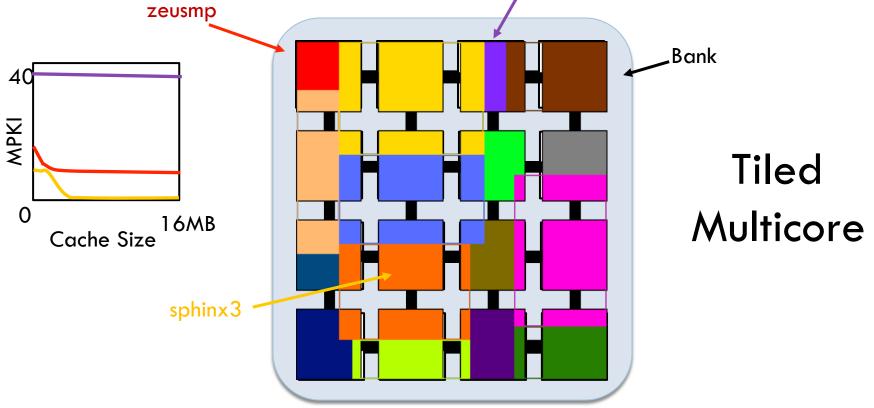


Cache organization should adapt to application

Jigsaw uses physical cache resources as building blocks of virtual caches, or shares



Jigsaw uses physical cache resources as building blocks of virtual caches, or shares zeusmp



Agenda

□ Introduction

- Background
 - Goals
 - Existing Approaches
- Jigsaw Design

Evaluation



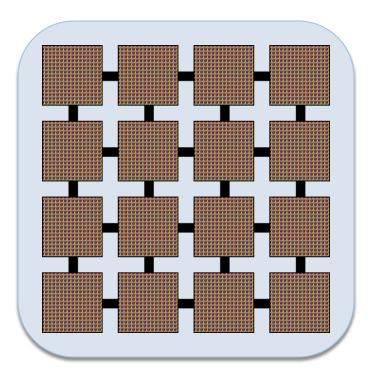
□ Make effective use of cache capacity

- Place data for low latency
- Provide capacity isolation for performance
- □ Have a simple implementation

Existing Approaches: S-NUCA

Spread lines evenly across banks

- High Capacity
- High Latency
- No Isolation
- □ Simple

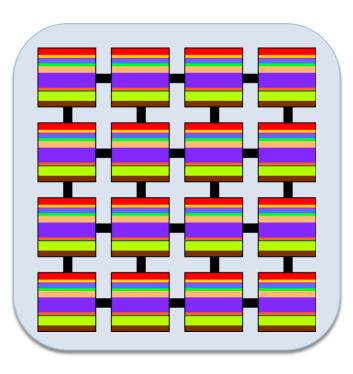


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Existing Approaches: Partitioning

Isolate regions of cache between applications.

- High Capacity
- High Latency
- Isolation
- □ Simple

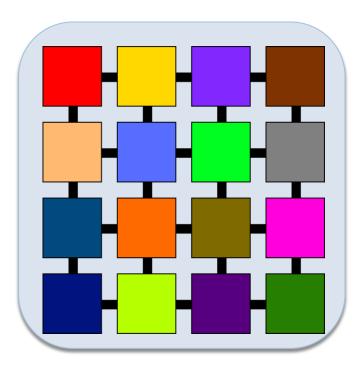


Jigsaw needs partitioning; uses Vantage to get strong guarantees with no loss in associativity

Existing Approaches: Private

Place lines in local bank

- Low Capacity
- □ Low Latency
- □ Isolation
- Complex LLC directory



Existing Approaches: D-NUCA

Placement, migration, and replication heuristics

High Capacity

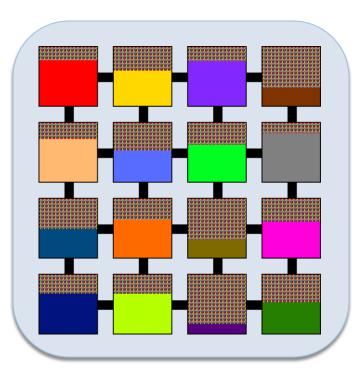
But beware of over-replication and restrictive mappings

Low Latency

- Don't fully exploit capacity vs. latency tradeoff
- No Isolation

Complexity Varies

Private-baseline schemes require LLC directory



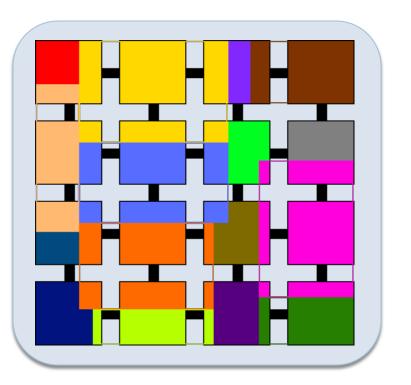
Existing Approaches: Summary

	S-NUCA	Partitioning	Private	D-NUCA
High Capacity	Yes	Yes	No	Yes
Low Latency	No	No	Yes	Yes
Isolation	No	Yes	Yes	No
Simple	Yes	Yes	No	Depends

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Jigsaw

- High Capacity Any share can take full capacity, no replication
- Low Latency Shares allocated near cores that use them
- Isolation Partitions within each bank



Simple – Low overhead hardware, no LLC directory, software-managed

Agenda

□ Introduction

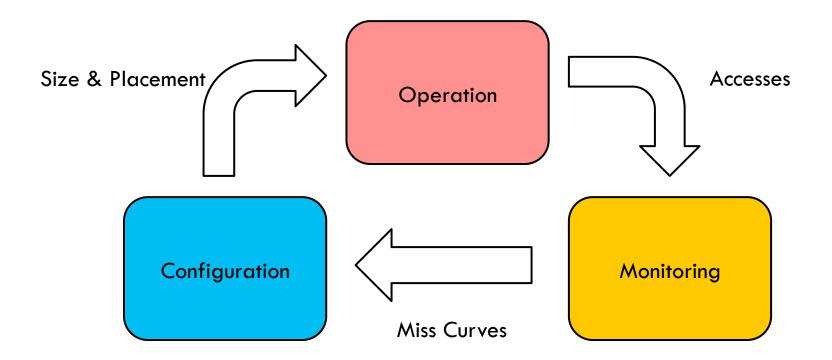
Background

Jigsaw Design

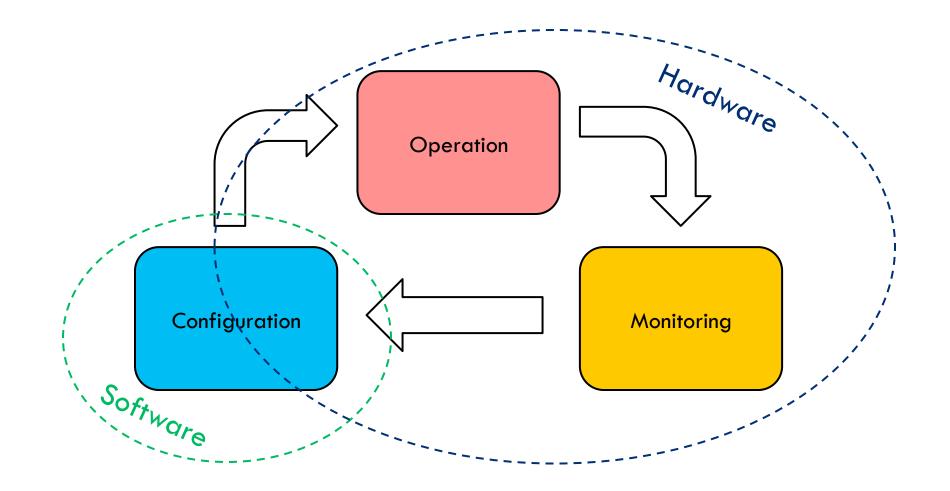
- Operation
- Monitoring
- Configuration

Evaluation

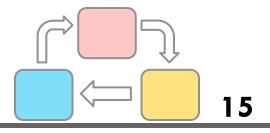
Jigsaw Components



Jigsaw Components



Agenda



□ Introduction

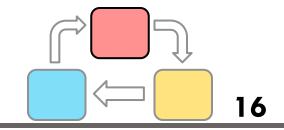
Background

Jigsaw Design

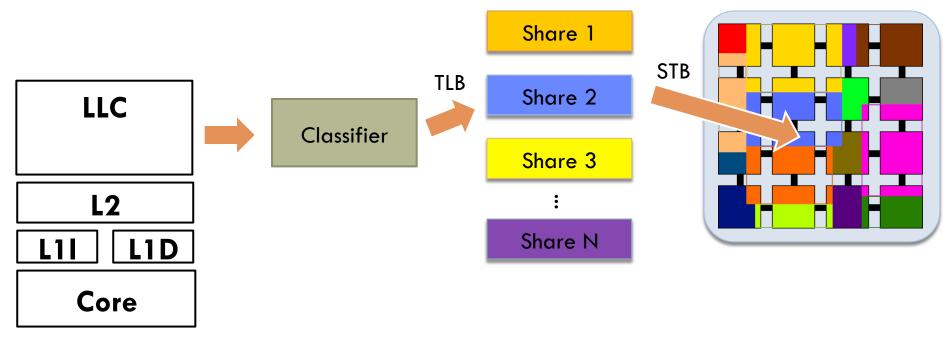
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Data \rightarrow shares, so no LLC coherence required

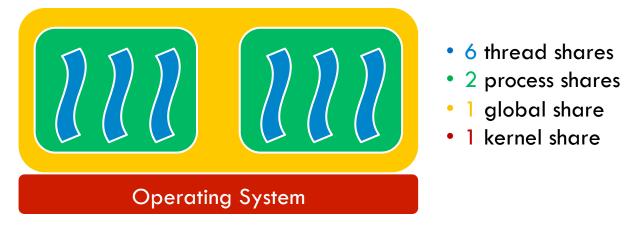


LD 0x5CA1AB1E

Data Classification

Jigsaw classifies data based on access pattern

Thread, Process, Global, and Kernel



Data lazily re-classified on TLB miss

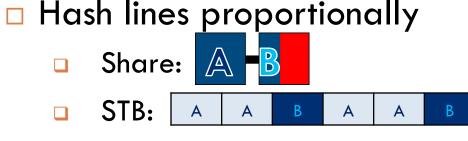
Similar to R-NUCA but...

R-NUCA: Classification → Location

Negligible overhead

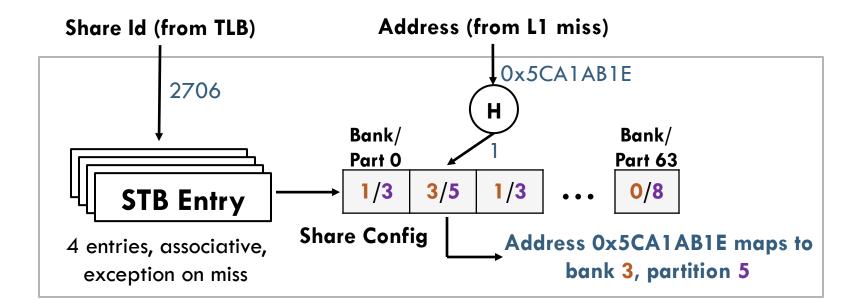


Address, Share →
 Bank, Partition

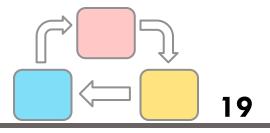


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400 bytes; low overhead



Agenda



□ Introduction

Background

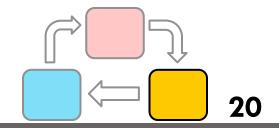
Jigsaw Design

Operation

- Monitoring
- Configuration

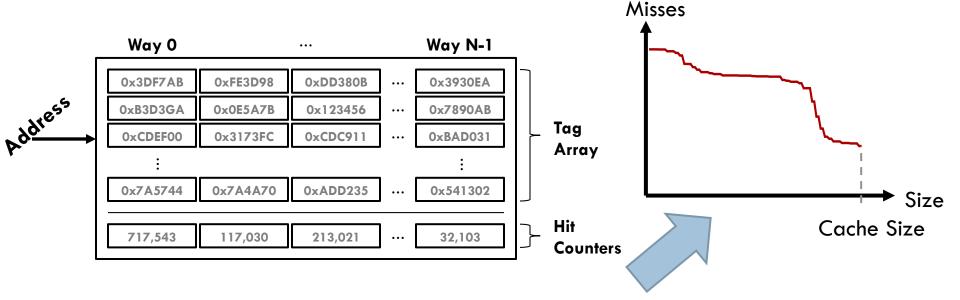
Evaluation



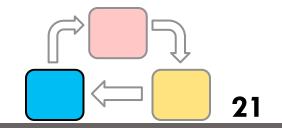


□ Software requires miss curves for each share

- □ Add utility monitors (UMONs) per tile to produce miss curves
- Dynamic sampling to model full LLC at each bank; see paper

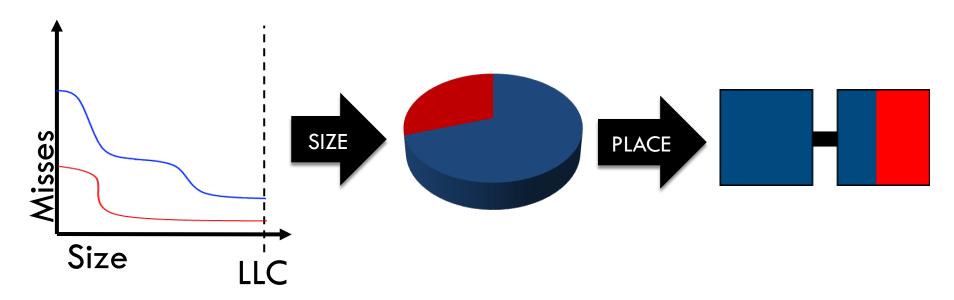


Configuration

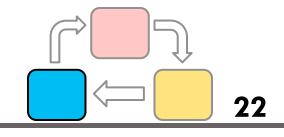


Software decides share configuration

- □ Approach: Size → Place
 - Solving independently is simple
 - Sizing is hard, placing is easy



Configuration: Sizing

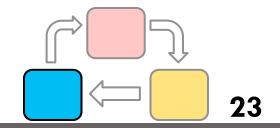


- Partitioning problem: Divide cache capacity of S among P partitions/shares to maximize hits
- Use miss curves to describe partition behavior
- □ NP-complete in general
- Existing approaches:

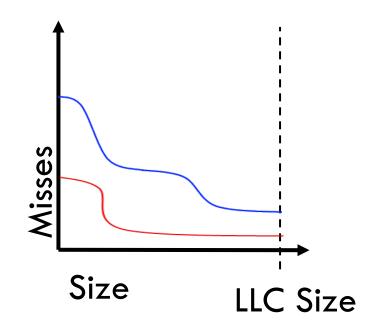
Hill climbing is fast but gets stuck in local optima-

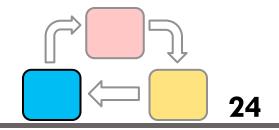
UCP Lookahead is good but scales quadratically: O(P x S²) Utility-based Cache Partitioning, Qureshi and Patt, MICRO'06

Can we scale Lookahead?

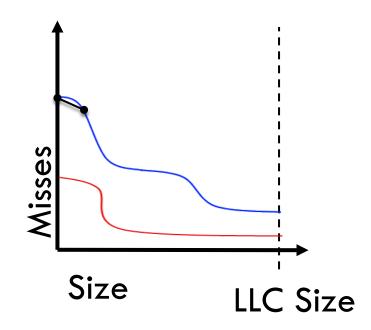


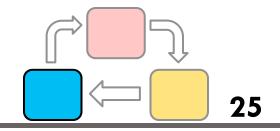
UCP Lookahead:



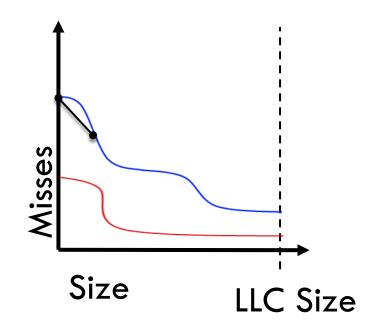


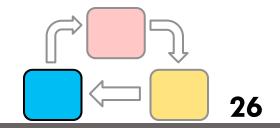
UCP Lookahead:



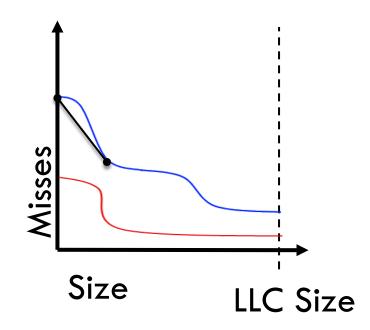


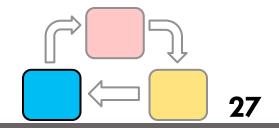
UCP Lookahead:



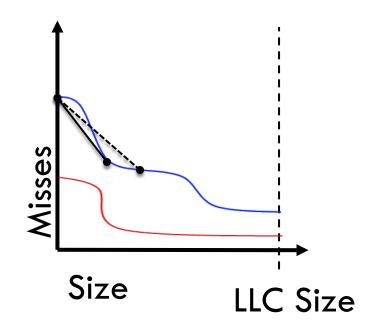


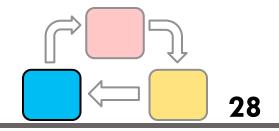
UCP Lookahead:



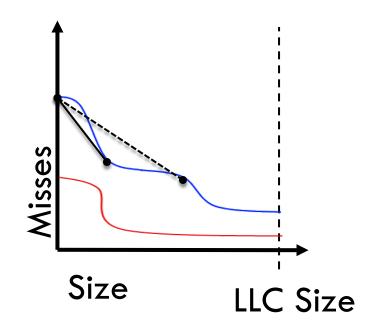


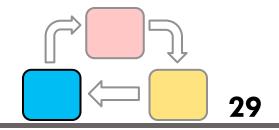
UCP Lookahead:



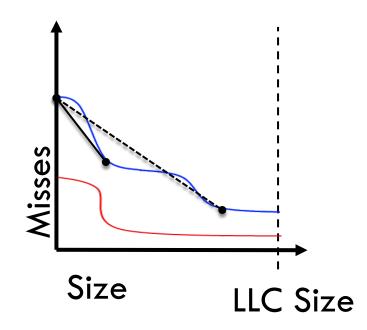


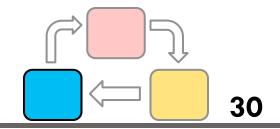
UCP Lookahead:



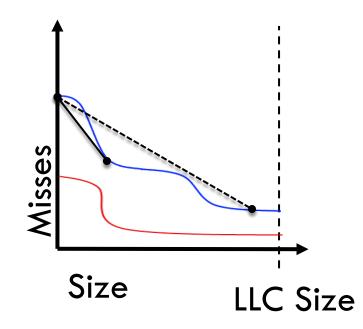


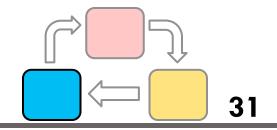
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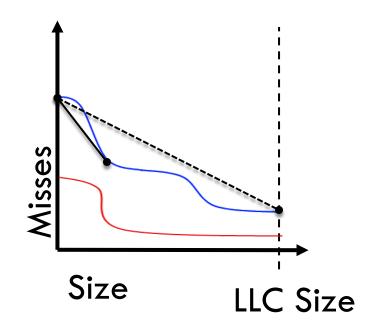


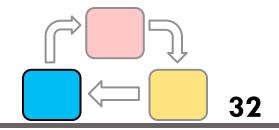
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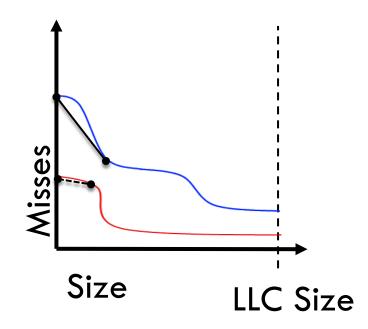


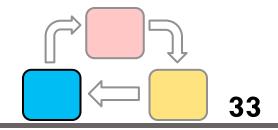
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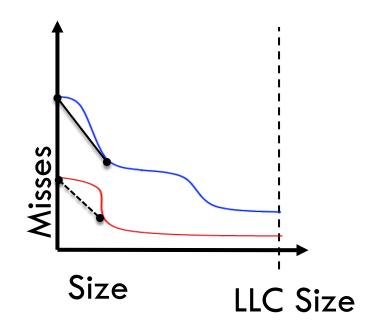


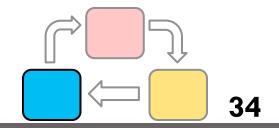
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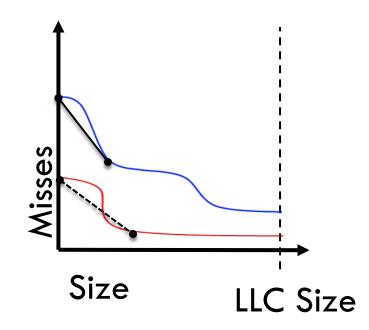


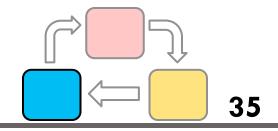
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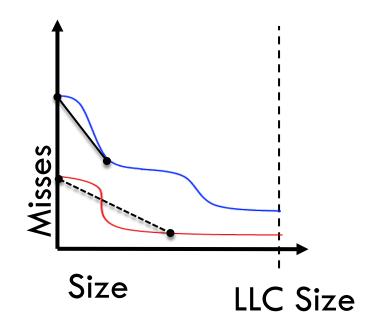


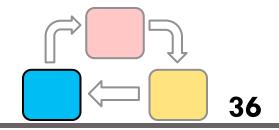
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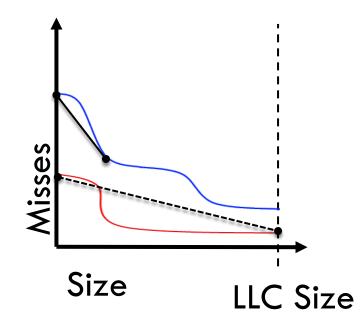


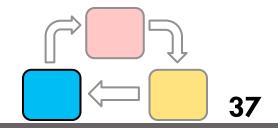
UCP Lookahead:





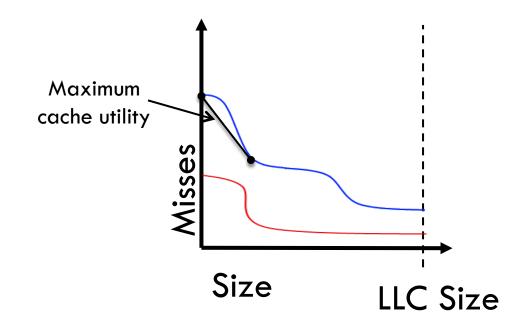
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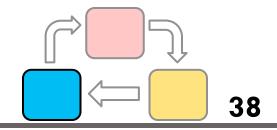




UCP Lookahead:

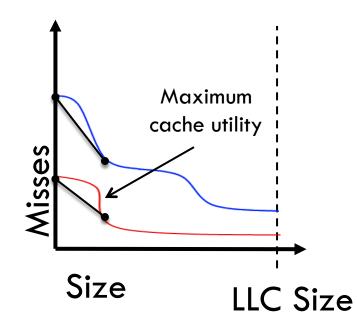
Scan miss curves to find allocation that maximizes average cache utility (hits per byte)

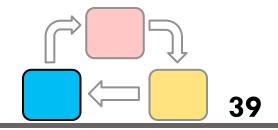




UCP Lookahead:

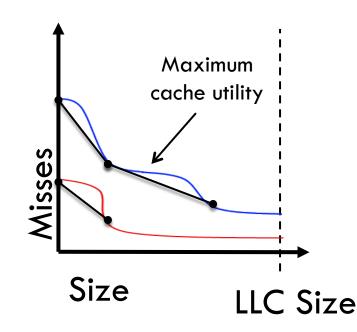
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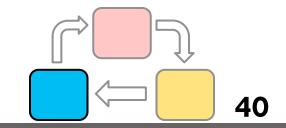




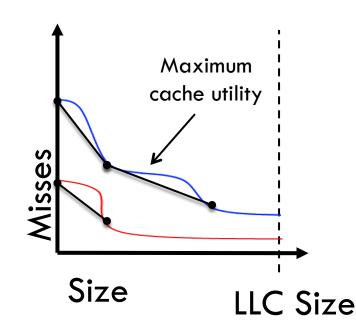
UCP Lookahead:

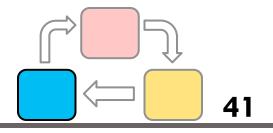
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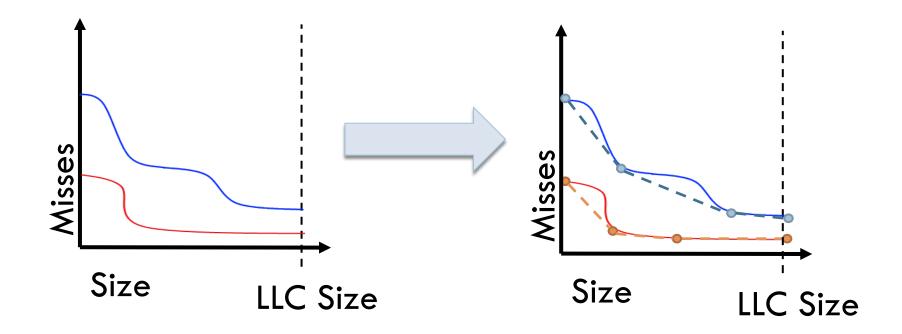


Observation: Lookahead traces the convex hull of the miss curve

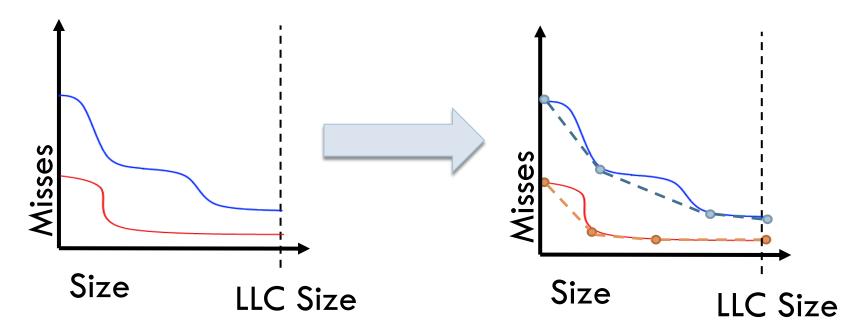




The convex hull of a curve is the set containing all lines between any two points on the curve, or "the curve connecting the points along the bottom"

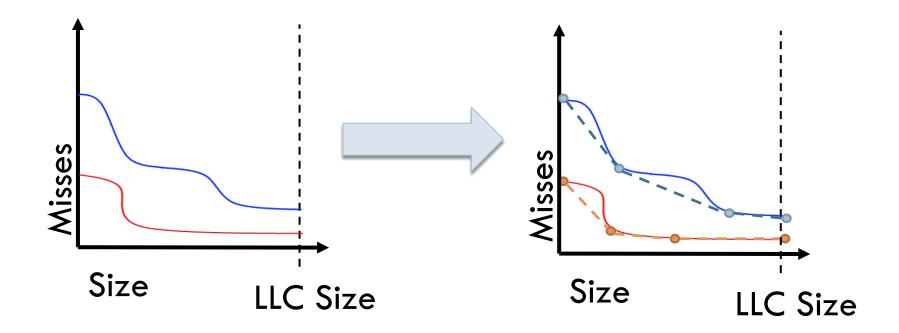


- **42**
- There are well-known linear algorithms to compute convex hulls
- Peekahead algorithm is an exact, linear-time implementation of UCP Lookahead



Peekahead computes all convex hulls encountered during allocation in linear time

- Starting from every possible allocation
- Up to any remaining cache capacity



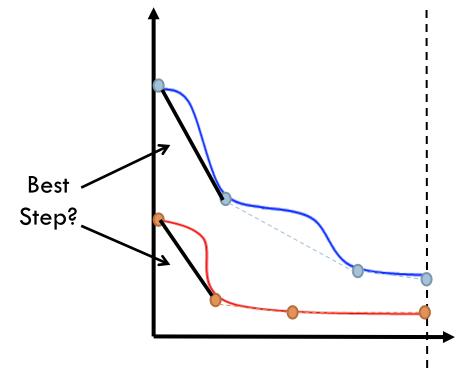
□ Knowing the convex hull, each allocation step is O(log P)

Convex hulls have decreasing slope

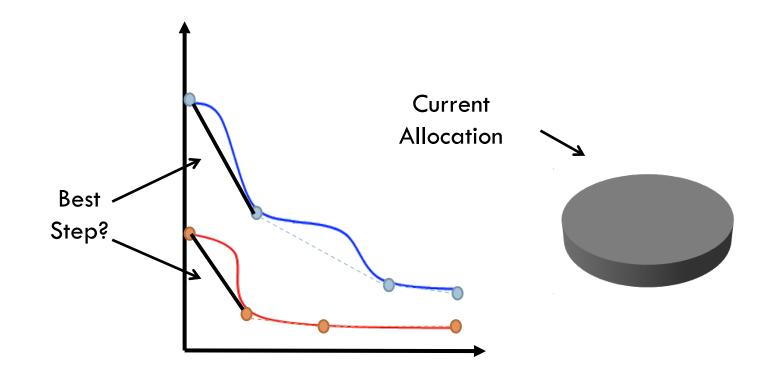
 decreasing average cache utility
 only consider next point on hull

ΔΔ

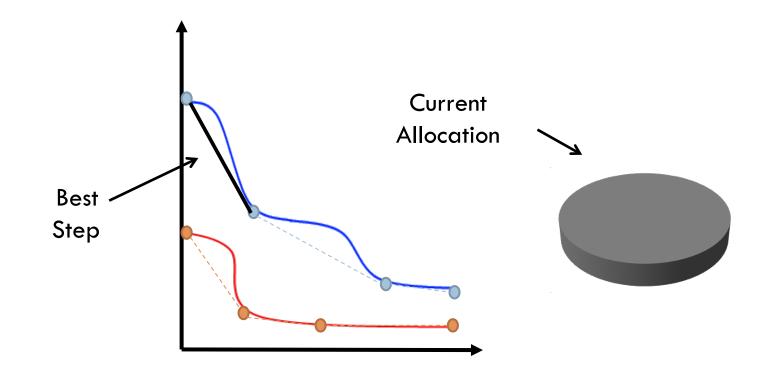
Use max-heap to compare between partitions



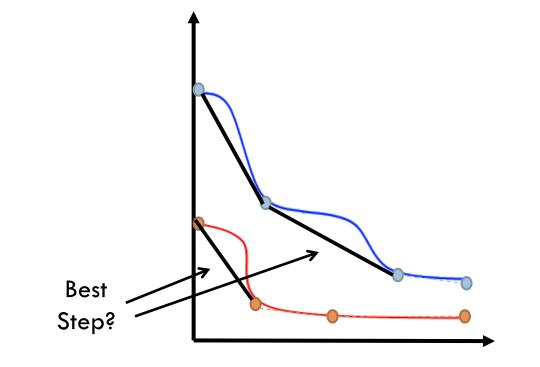
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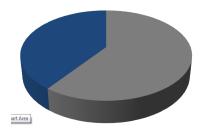


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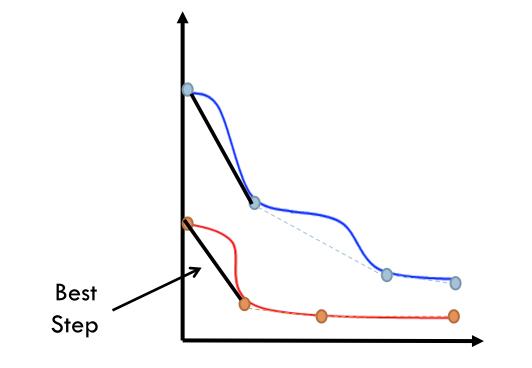


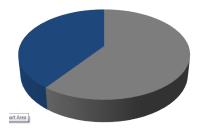
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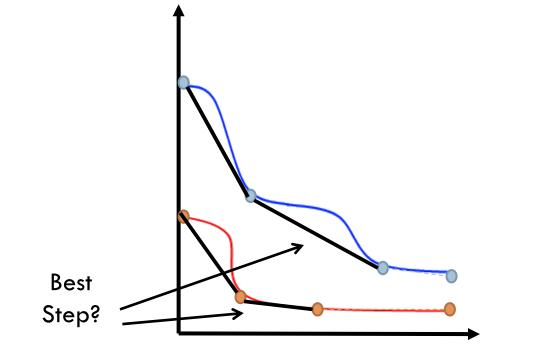


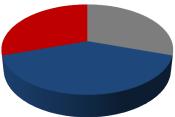




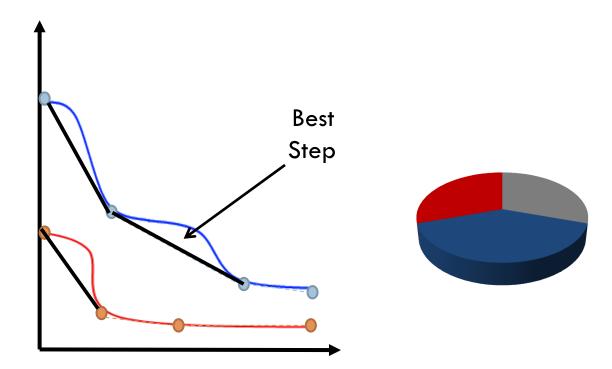


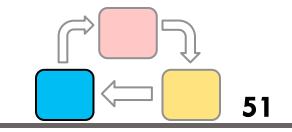




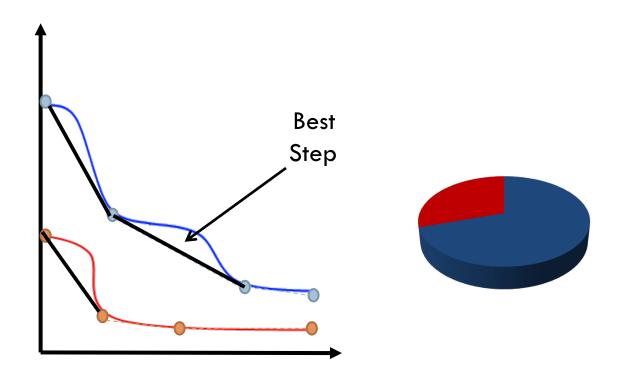


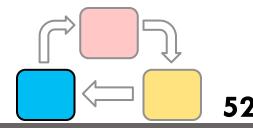
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□ Knowing the convex hull, each allocation step is O(log P)

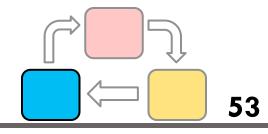




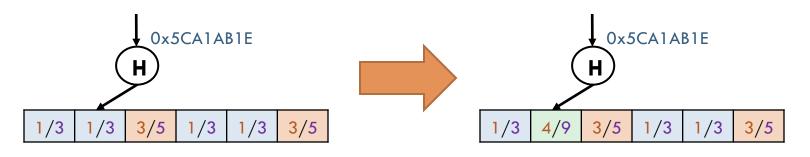
- \Box Full runtime is O(P x S)
 - P number of partitions
 - S cache size
- See paper for additional examples, algorithm, and corner cases

- See technical report for additional detail, proofs, and run-time analysis
 - Jigsaw: Scalable Software-Defined Caches (Extended Version), Nathan Beckmann and Daniel Sanchez, Technical Report MIT-CSAIL-TR-2013-017, Massachusetts Institute of Technology, July 2013





When STB changes, some addresses hash to different banks

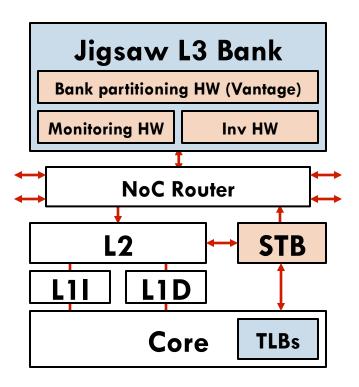


- Selective invalidation hardware walks the LLC and invalidates lines that have moved
- Heavy-handed but infrequent and avoids directory
 Maximum of 300K cycles / 50M cycles = 0.6% overhead

Design: Hardware Summary

- Operation:
 - Share-bank translation buffer (STB) handles accesses
 - TLB augmented with share id
- Monitoring HW: produces miss curves
- Configuration: invalidation HW
- Partitioning HW (Vantage)

Tile Organization



Modified structures
New/added structures

Agenda

□ Introduction

Background

Jigsaw Design

Evaluation

- Methodology
- Performance
- Energy

Methodology

Execution-driven simulation using zsim

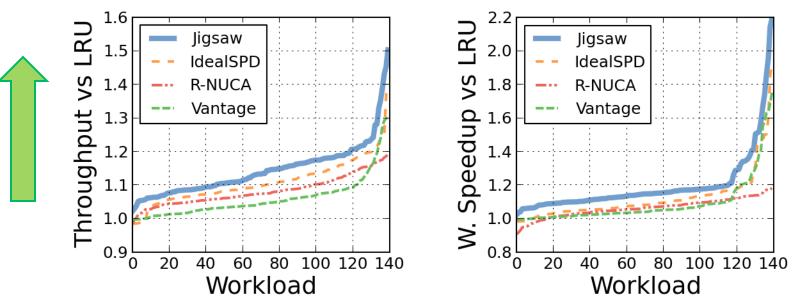
- Workloads:
 - 16-core singlethreaded mixes of SPECCPU2006 workloads
 - 64-core multithreaded (4x16-thread) mixes of PARSEC

Cache organizations

- LRU shared S-NUCA cache with LRU replacement; baseline
- Vantage S-NUCA with Vantage and UCP Lookahead
- R-NUCA state-of-the-art shared-baseline D-NUCA organization
- IdealSPD ("shared-private D-NUCA") private L3 + shared L4
 - 2x capacity of other schemes
 - Upper bound for private-baseline D-NUCA organizations
- Jigsaw

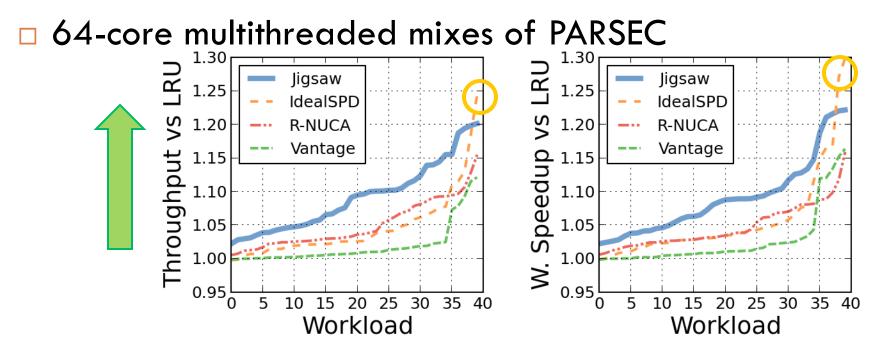
Evaluation: Performance





- Jigsaw achieves best performance
 - Up to 50% improved throughput, 2.2x improved w. speedup
 - Gmean +14% throughput, +18% w. speedup
- Jigsaw does even better on the most memory intensive mixes
 - Top 20% of LRU MPKI
 - Gmean +21% throughput, +29% w. speedup

Evaluation: Performance

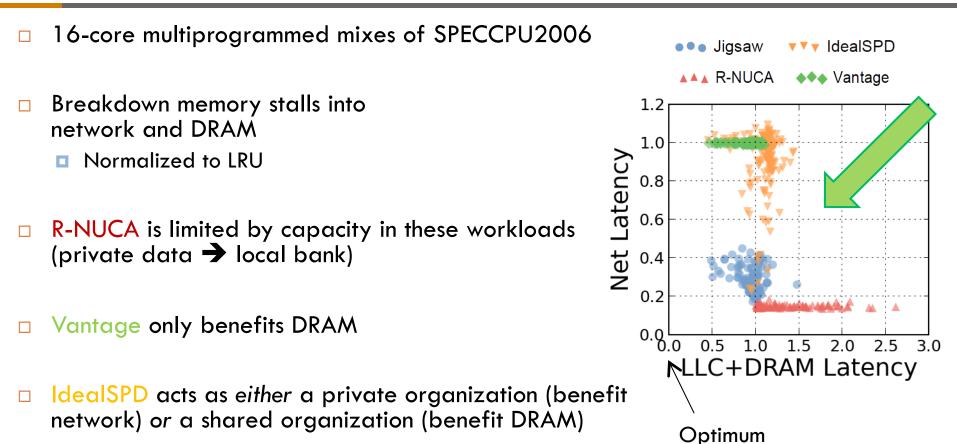


Jigsaw achieves best performance

□ Gmean +9% throughput, +9% w. speedup

Remember IdealSPD is an upper bound with 2x capacity

Evaluation: Performance Breakdown



Jigsaw is the only scheme to simultaneously benefit network and DRAM latency

Evaluation: Energy

16-core multiprogrammed mixes 1.4 1.3 LRU 1.2 1.11.0 Energy 0.9 ligsaw 0.8 dealSPD R-NUCA 0.6 Vantage 0.5 0 20 40 60 80 100 120 140 Workload

McPAT models of full-system energy (chip + DRAM)

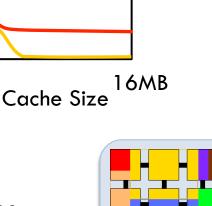
- Jigsaw achieves best energy reduction
 - □ Up to 72%, gmean of 11%
 - Reduces both network and DRAM energy

Conclusion

NUCA is giving us more capacity, but further away

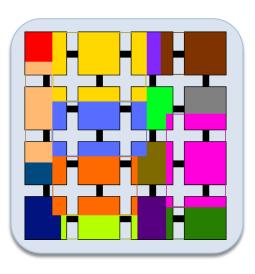
- Applications have widely varying cache behavior
- Cache organization should adapt to meet application needs
- Jigsaw uses physical cache resources as building blocks of virtual caches, or shares
 Sized to fit working set
 - Placed near application for low latency

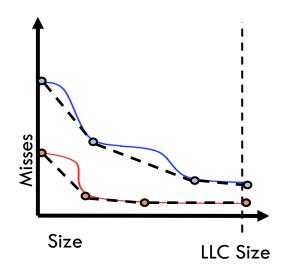
Jigsaw improves performance up to 2.2x and reduces energy up to 72%



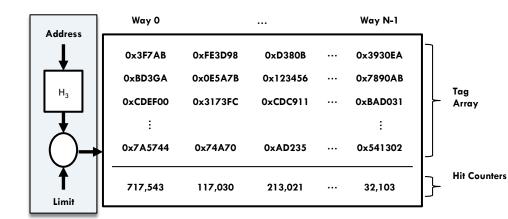
40

MPKI













Massachusetts Institute of Technology

Placement

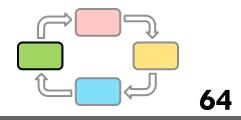
□ Greedy algorithm

Each share is allocated budget

Shares take turns grabbing space in "nearby" banks
 Banks ordered by distance from "center of mass" of cores accessing share

Repeat until budget & banks exhausted

Monitoring

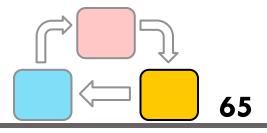


□ Software requires miss curves for each share

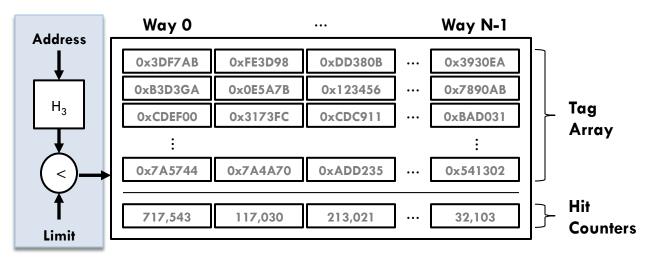
- Add UMONs per tile
 - Small tag array that models LRU on sample of accesses
 - Tracks # hits per way, # misses miss curve
- Changing sampling rate models a larger cache
 Sampling Rate = UMON Lines
 Modeled Cache Lines
- STB spreads lines proportionally to partition size, so sampling rate must compensate

Sampling Rate = $\frac{\text{Share size}}{\text{Partition size}} \times \frac{\text{UMON Lines}}{\text{Modeled Cache Lines}}$

Monitoring



- \Box STB spreads addresses unevenly \rightarrow change sampling rate to compensate
- Augment UMON with hash (shared with STB) and 32-bit limit register that gives fine control over sampling rate



- UMON now models full LLC capacity exactly
 - Shares require only one UMON
 - Max four shares / bank → four UMONs / bank → 1.4% overhead

Evaluation: Extra

- □ See paper for:
 - Out-of-order results
 - Execution time breakdown
 - Peekahead performance

32KB

128 256 512

Associativity

64KB

512 2K

Sensitivity studies

4KB

1.20

1.15

1.10

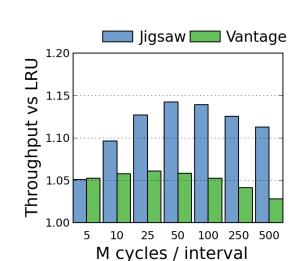
..05

1.00

32

64

Throughput vs LRU



Exec

V R

All

1.2

1.0 0.8

0.6 0.4

0.2

Cycles vs LRU

