

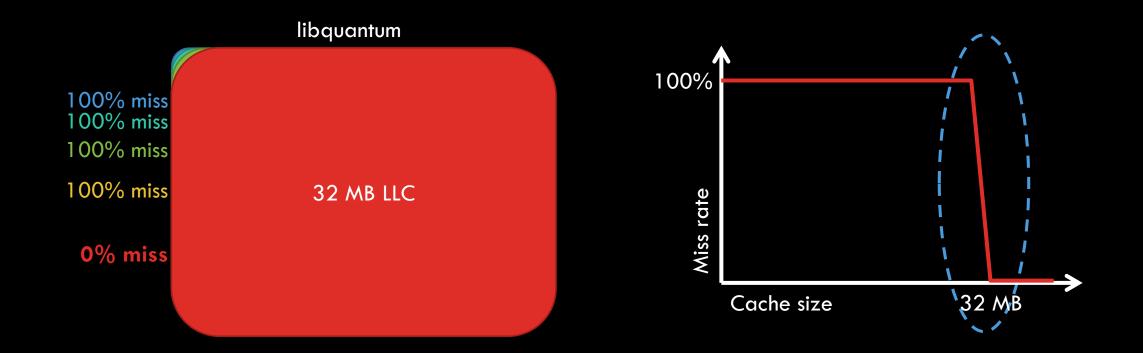
TALUS: A SIMPLE WAY TO REMOVE PERFORMANCE CLIFFS IN CACHES

Nathan Beckmann Daniel Sanchez





CACHES HAVE PERFORMANCE CLIFFS



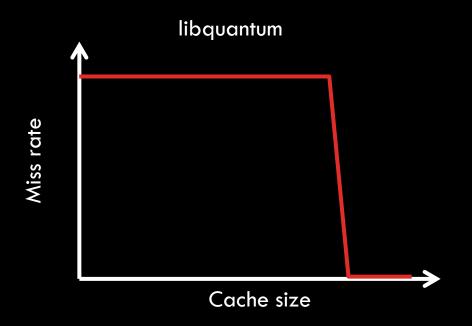
CLIFFS ARE A PROBLEM

Cliffs are wasteful

Cliffs cause annoying performance bugs

Cliffs complicate cache partitioning

NP-hard problem



PRIOR WORK: HIGH-PERFORMANCE REPLACEMENT VS. CACHE PARTITIONING

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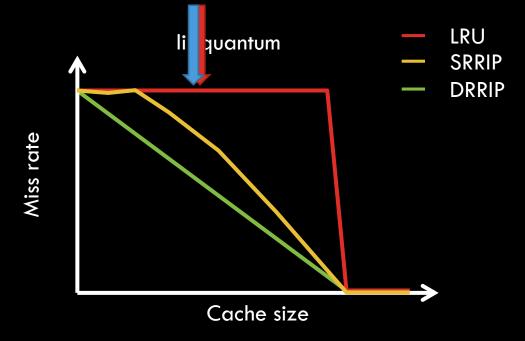
ance

Individual apps: High-performance replacement • E.g., RRIP [ISCA'10]

Shared caches: Cache partitioning • E.g., UCP [MICRO'06]

For shared • Both in perf Partitioning replacemen Can partitic

32 MB libquantum #1 0% miss



IN THIS TALK WE WILL...

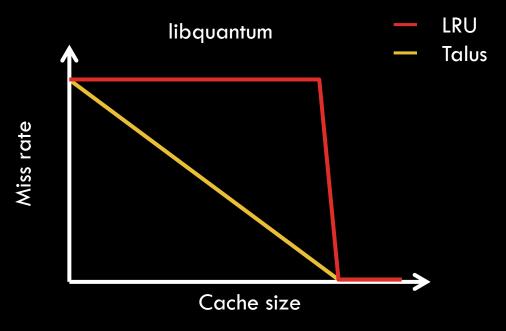
Give a simple technique to eliminate cliffs (Talus) • Talus partitions within a single access stream

Prove it works under simple assumptionsAgnostic to app or replacement policy

No cliffs

Simpler cache partitioning

Talus combines the benefits of high-performance replacement and partitioning



5

ROAD MAP

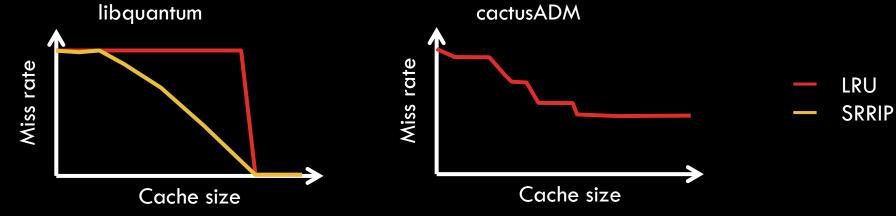
Talus example

Theory

Implementation

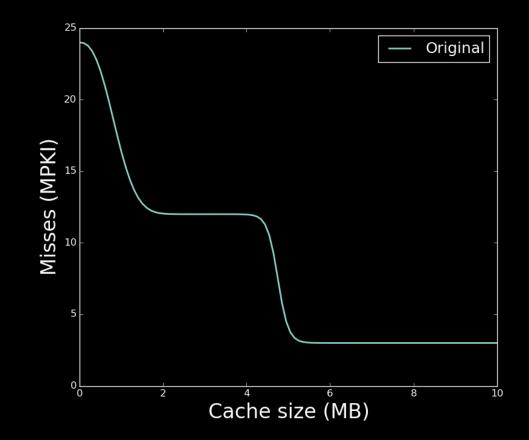
Evaluation

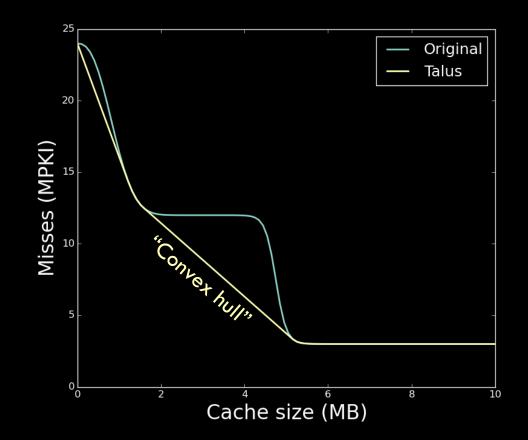
TALUS USES MISS CURVES

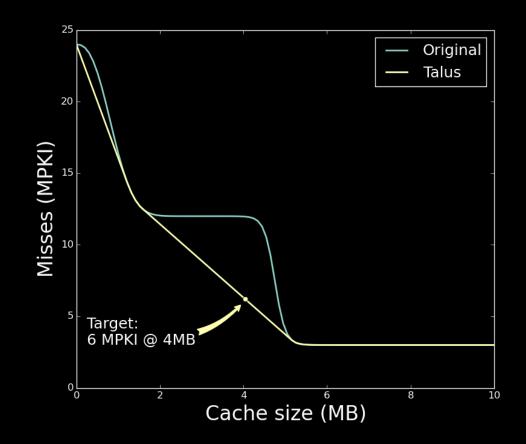


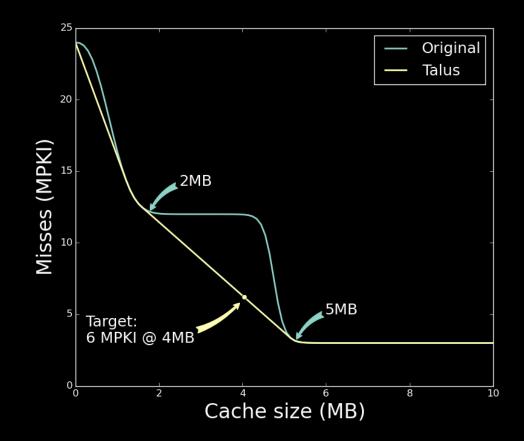
Cliffs occur under a variety of access pattern and replacement policies

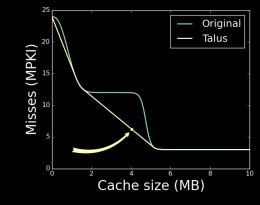
Talus works on miss curves <u>only;</u> Talus is agnostic to app and replacement policy



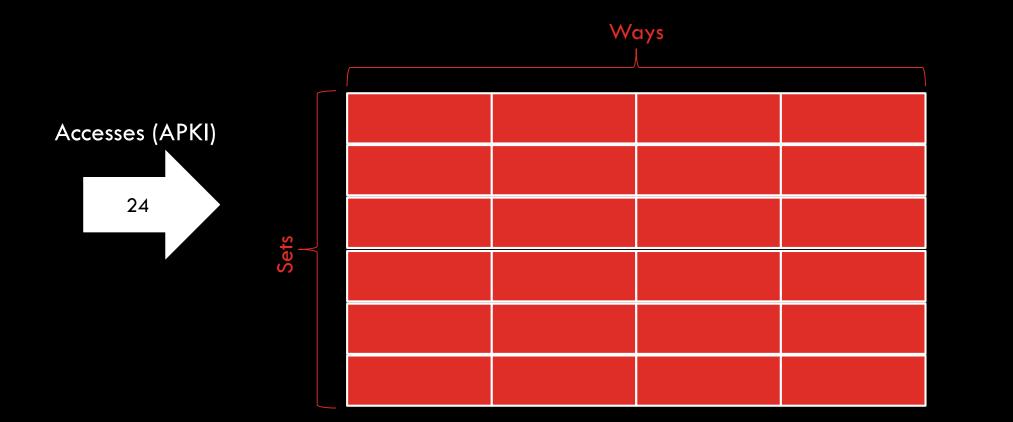


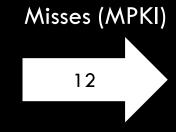


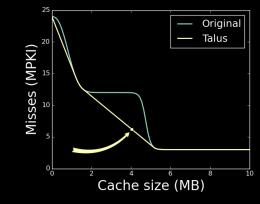




(HYPOTHETICAL) BASELINE CACHE AT 2 MB

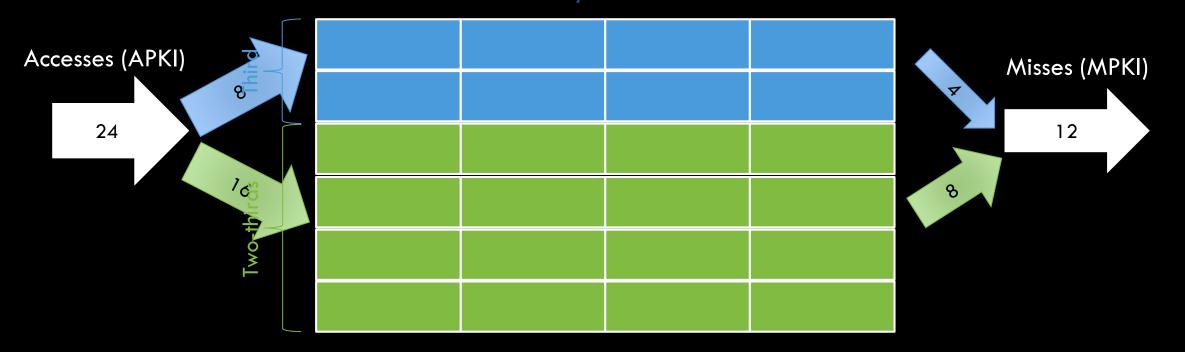




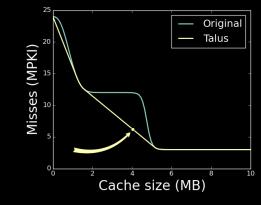


(HYPOTHETICAL) BASELINE CACHE AT 2 MB

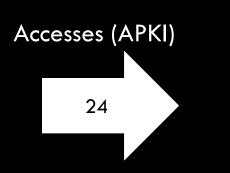
2/3 MB

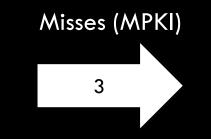


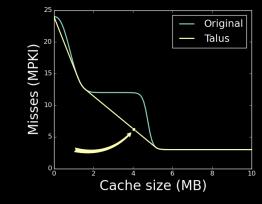
4/3 MB



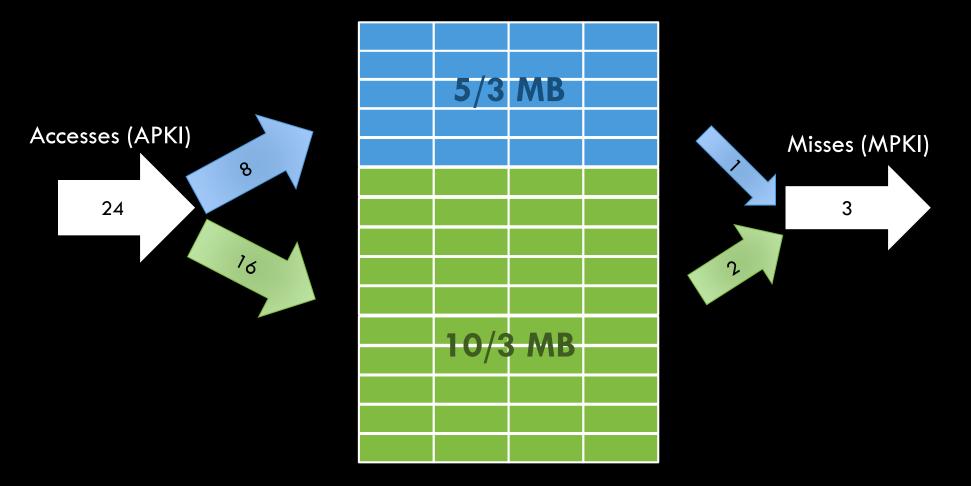
(HYPOTHETICAL) BASELINE CACHE AT 5 MB







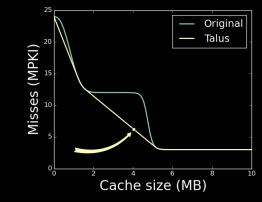
(HYPOTHETICAL) BASELINE CACHE AT 5 MB

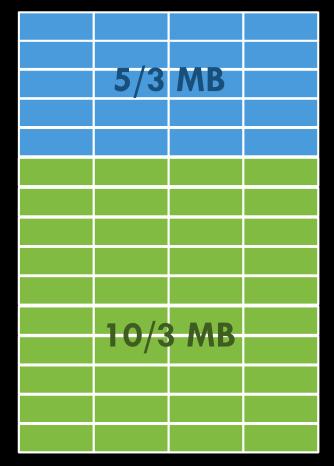


TALUS AT 4 MB

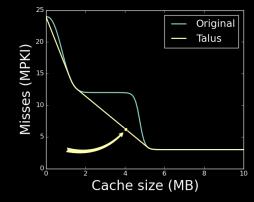
Combine hypothetical baseline 2 MB & 5 MB

2/3	MB	
4/3	MB	
.,.		

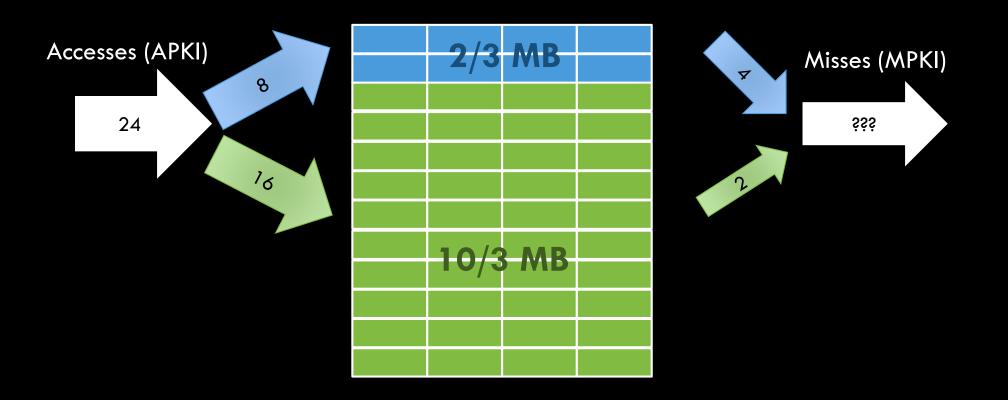




TALUS AT 4 MB



Spread accesses disproportionally across partitions to match baselines



EXAMPLE SUMMARY

Talus avoids cliffs by combining efficient cache sizes of baseline

Does not know or care about app or replacement details

Just needs miss curve!

Nothing special about set partitioning; Talus works on other partitioning techniques

But how to choose partition configuration?

ROAD MAP

Talus example

Theory

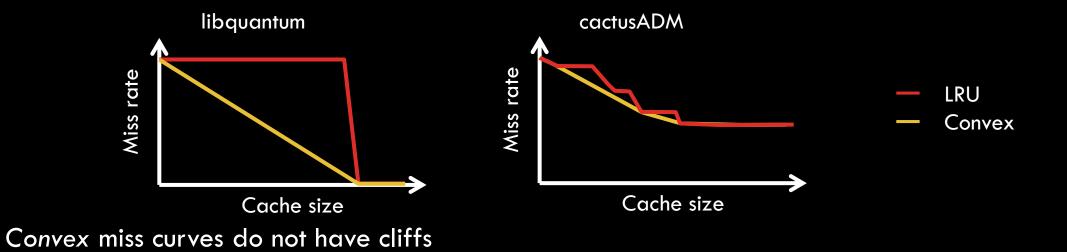
Proof sketch

Talus vs prior policies

Implementation

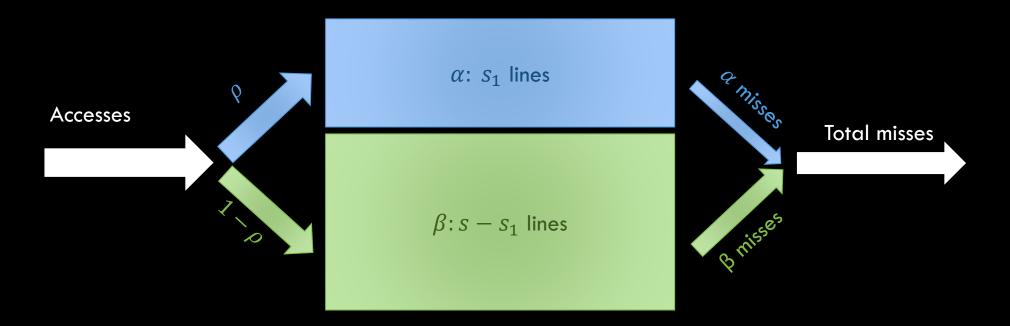
Evaluation

GOAL: CONVEXITY AVOIDS CLIFFS



SHADOW PARTITIONING

Talus divides the cache (of size S) into shadow partitions, invisible to software



Talus ensures convexity under simple assumptions

ASSUMPTIONS

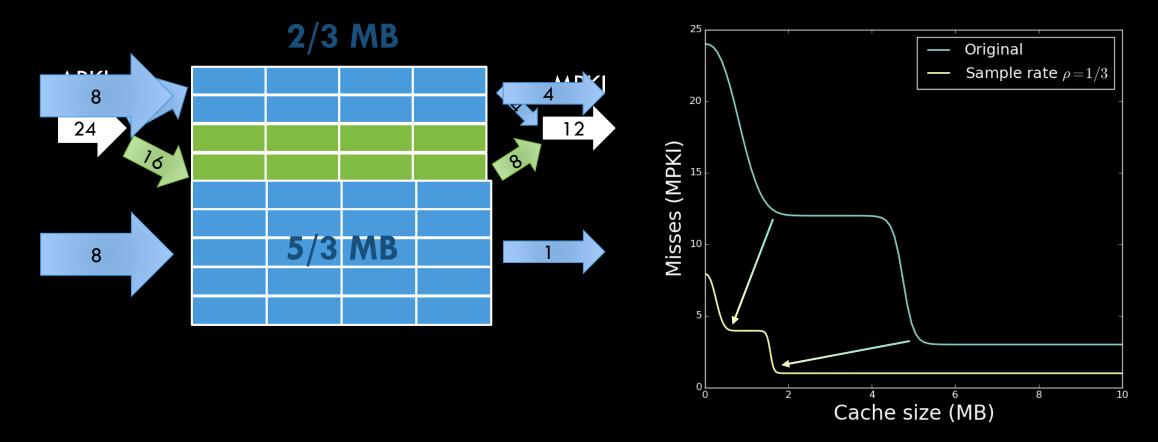
Miss curves are stable (eg, across tens of milliseconds)

Cache size is the dominant factor in miss rate (ie, not associativity)

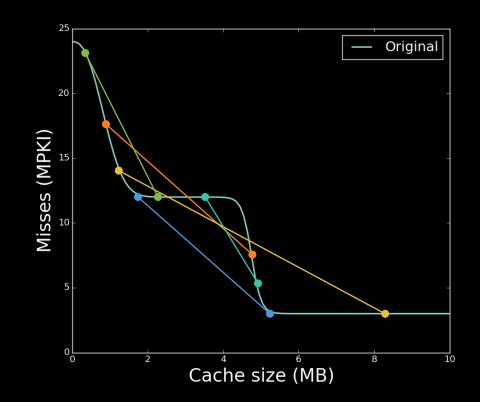
Pseudo-random sampling of an access stream yields a statistically self-similar stream

These assumptions are implicit in prior work (see paper)

SAMPLING SCALES THE MISS CURVE

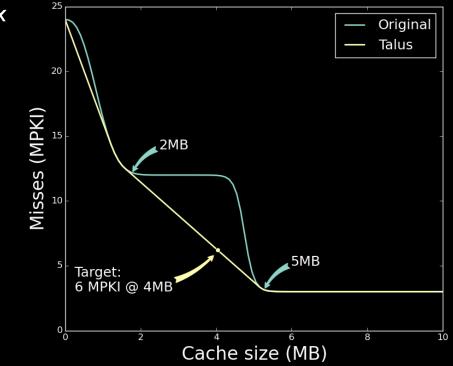


SHADOW PARTITIONING INTERPOLATES MPKI OF THE ORIGINAL MISS CURVE



TALUS GUARANTEES CONVEXITY

Just interpolate the convex hull of the original miss curve!



THERE'S MATH!

Miss curve scaling:

$$m'(s') = \rho \ m\left(\frac{s'}{\rho}\right)$$

Shadow partitioned miss rate:

$$m_{\text{shadow}(s)} = \rho m\left(\frac{s_1}{\rho}\right) + (1-\rho) m\left(\frac{s-s_1}{1-\rho}\right)$$

How to interpolate between α and β :

$$\rho = \frac{\beta - s}{\beta - \alpha}, \ s_1 = \rho \alpha$$

ROAD MAP

Motivation

Talus example

Theory

- Proof sketch
- Talus vs prior policies

Implementation

Evaluation

PRIOR TECHNIQUE: BYPASSING

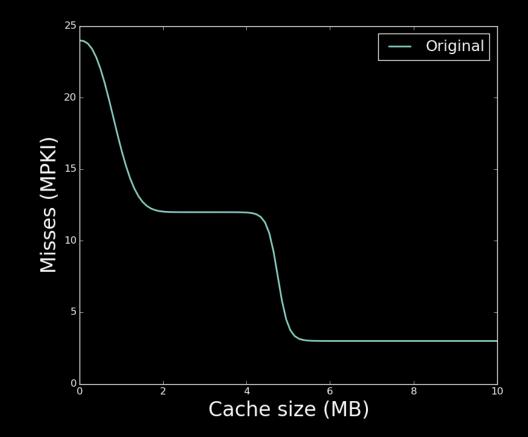
Bypassing is a common replacement technique to avoid thrashing • E.g., BIP [ISCA'07] bypasses 31/32 accesses

We compute optimal bypassing rate from miss curve

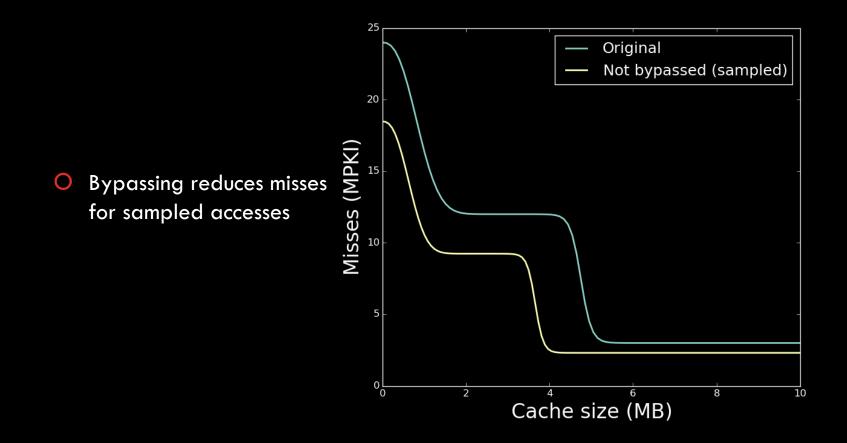
Bypassing handles some kinds of cliffs, but not all

→ Talus outperforms bypassing on some access patterns

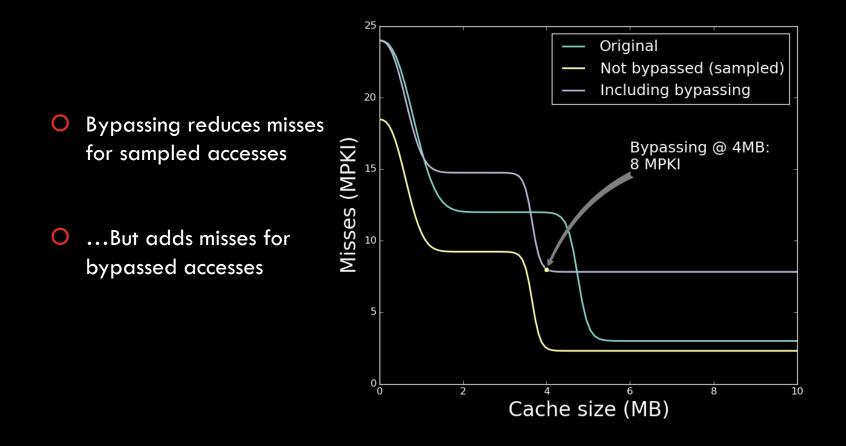
BYPASSING PRODUCES COMPETING EFFECTS



BYPASSING PRODUCES COMPETING EFFECTS

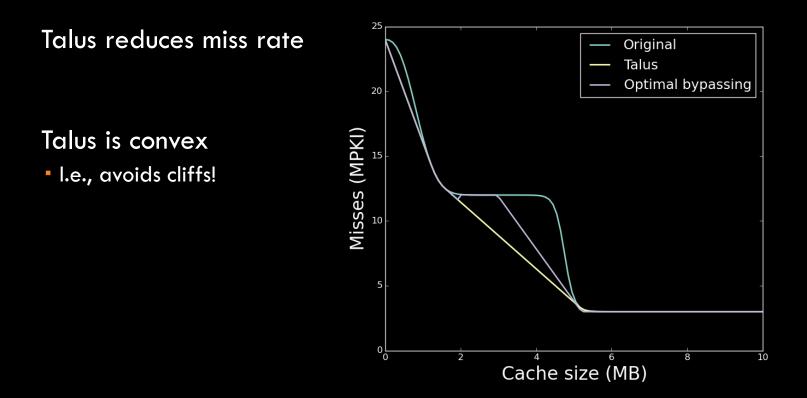


BYPASSING PRODUCES COMPETING EFFECTS



See paper for details!

TALUS VS BYPASSING



ROAD MAP

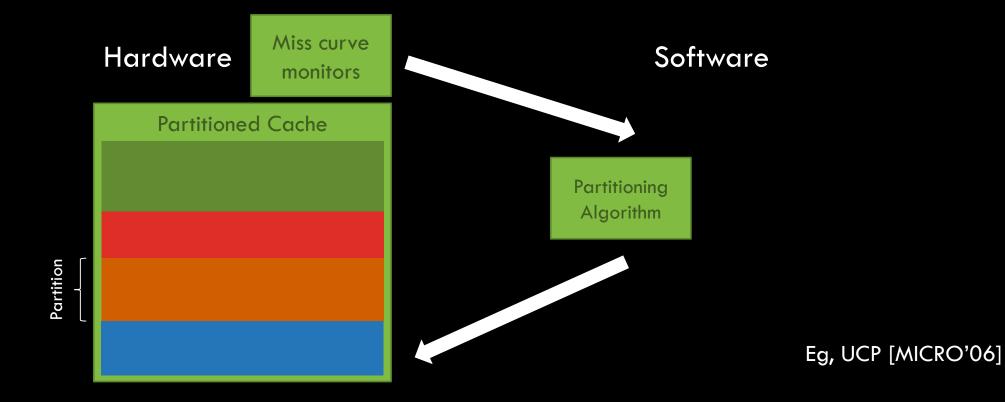
Talus example

Theory

Implementation

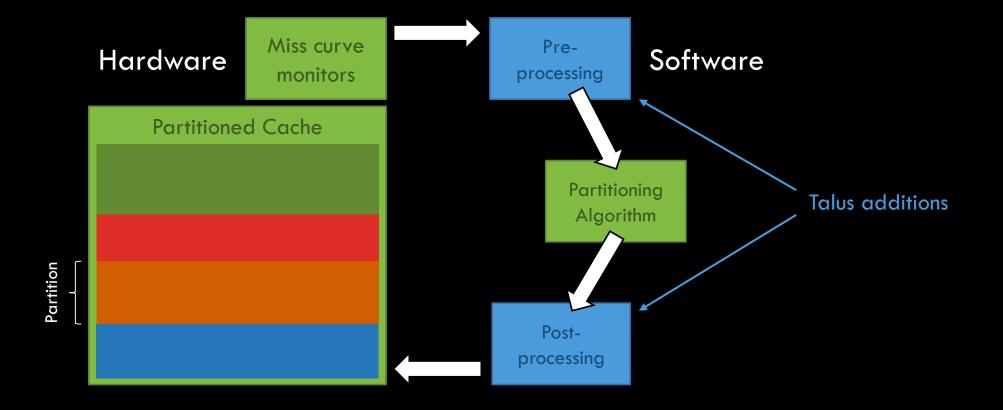
Evaluation

CONVENTIONAL PARTITIONED CACHE

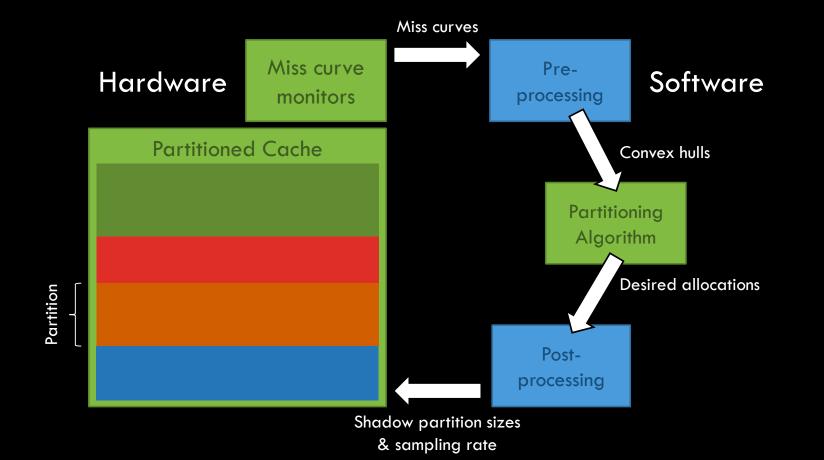


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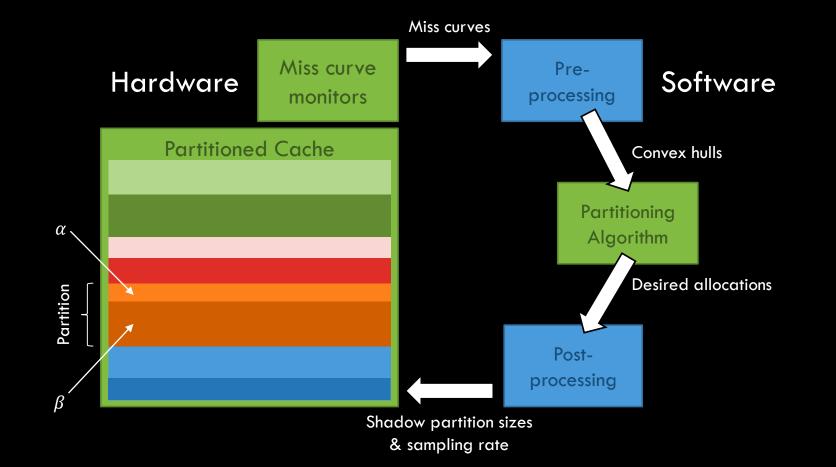
EFFICIENT TALUS IMPLEMENTATION



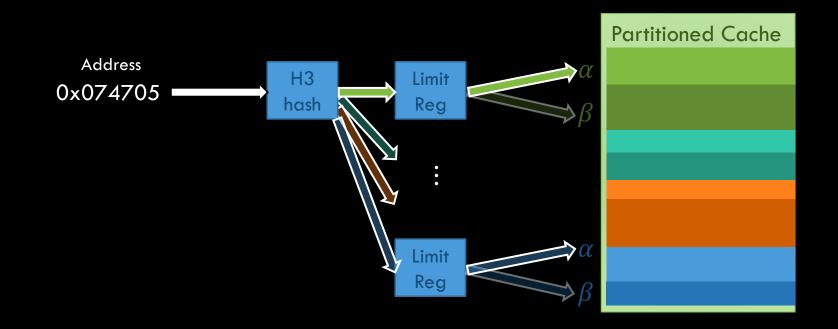
EFFICIENT TALUS IMPLEMENTATION



EFFICIENT TALUS IMPLEMENTATION



EFFICIENT TALUS IMPLEMENTATION



TALUS IMPOSES LOW OVERHEADS

Computing convex hulls is cheap: O(N)

Computing shadow partition sizes is cheap: O(1)

Talus reduces software overheads by making simple algorithms perform well

Shadow partitioning is cheap: similar monitors to prior work (see paper), 1 bit per tag, 8 bits per partition, simple hash function

Talus improves cache performance and adds <1% state



EVALUATION CLAIMS

We compare Talus to high-performance replacement policies and partitioning schemes

Talus is convex in practice

Single-program: Talus gets similar performance to prior replacement policies

Multi-program: Talus greatly simplifies cache partitioning and slightly outperforms prior, complex partitioning algorithms

Talus combines the benefits of high-performance replacement and partitioning

METHODOLOGY

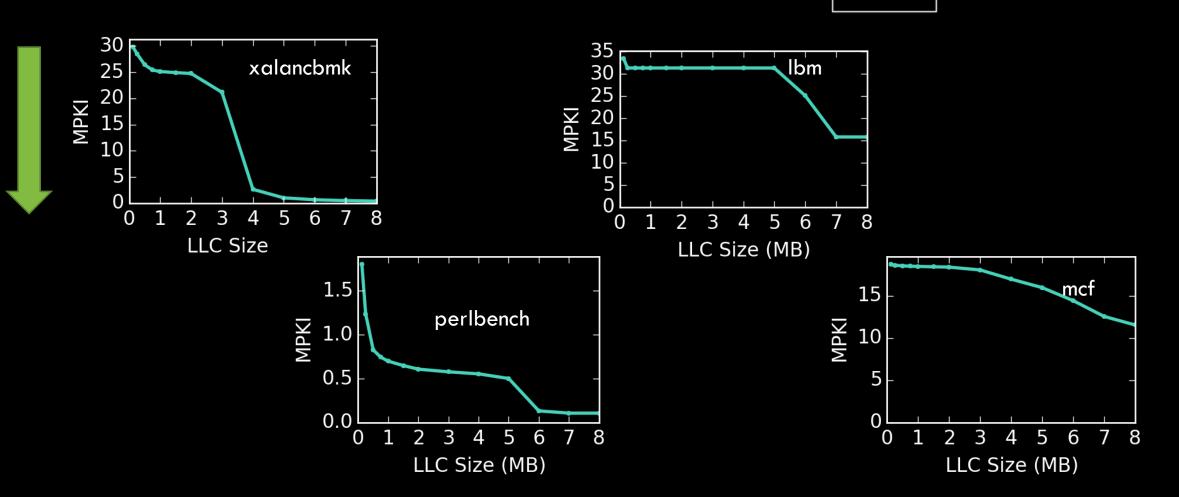
Evaluate 1- and 8-core system similar to Silvermont on zsim

See paper for details

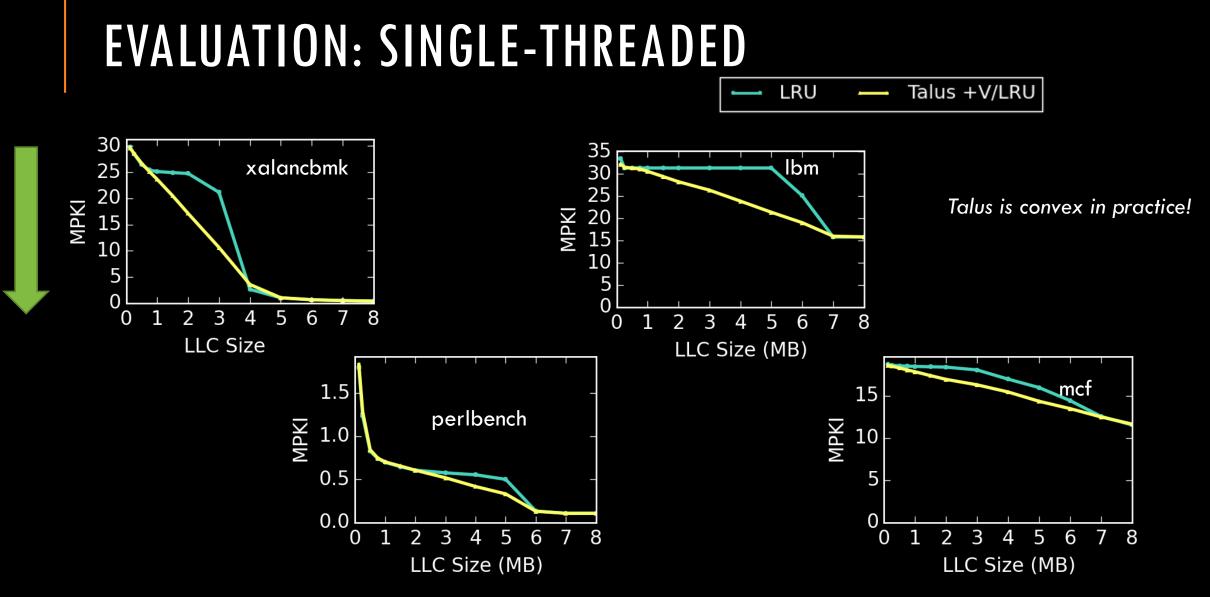
Individual SPEC CPU2006 benchmarks + random mixes

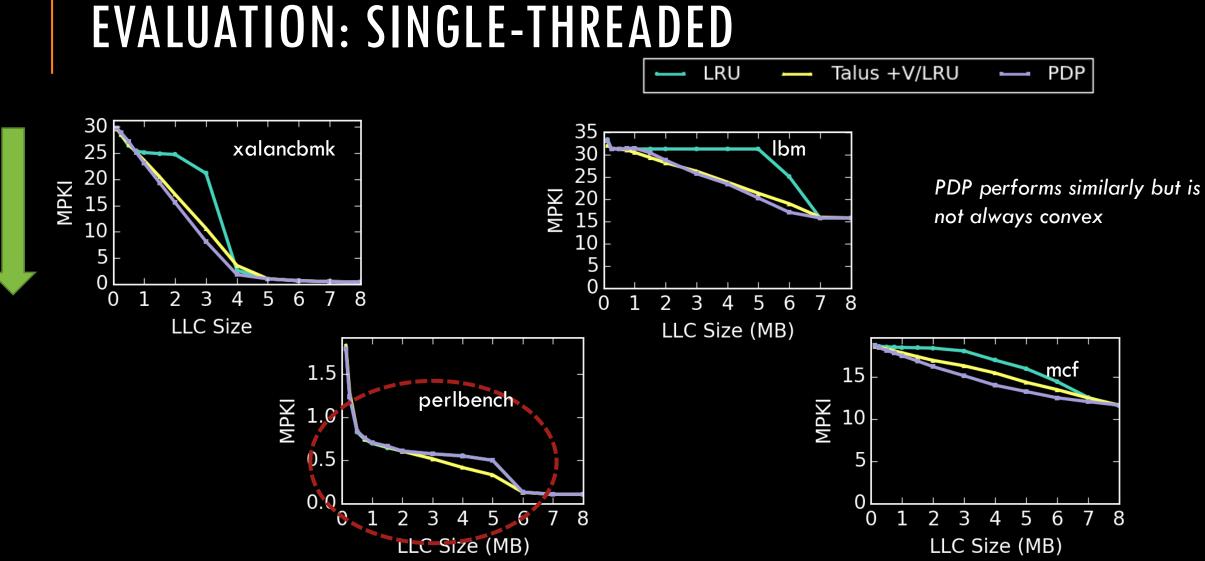
Talk only shows Talus on LRU with Vantage partitioning (Talus +V/LRU)

EVALUATION: SINGLE-THREADED

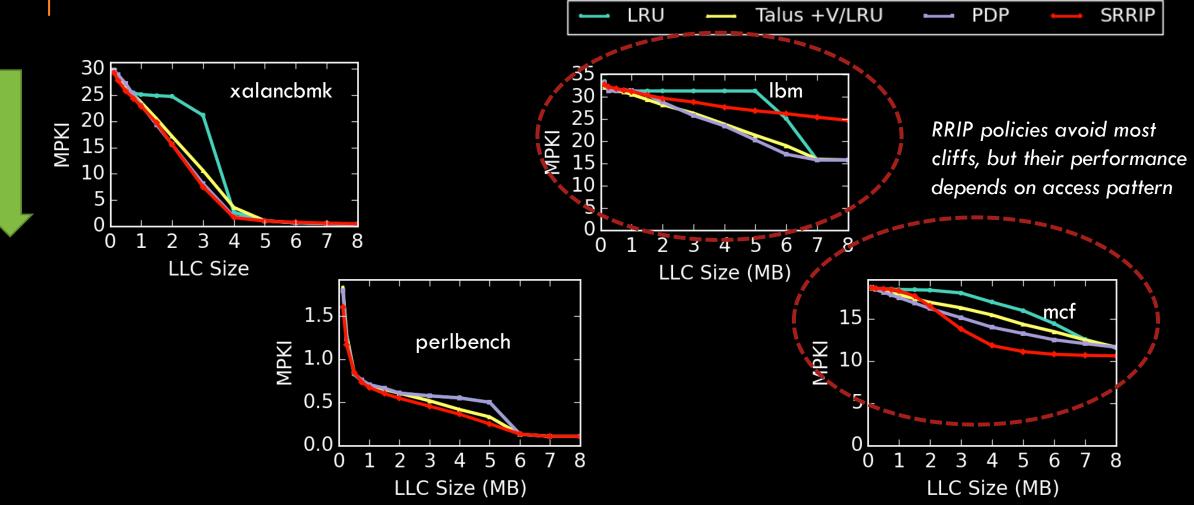


LRU

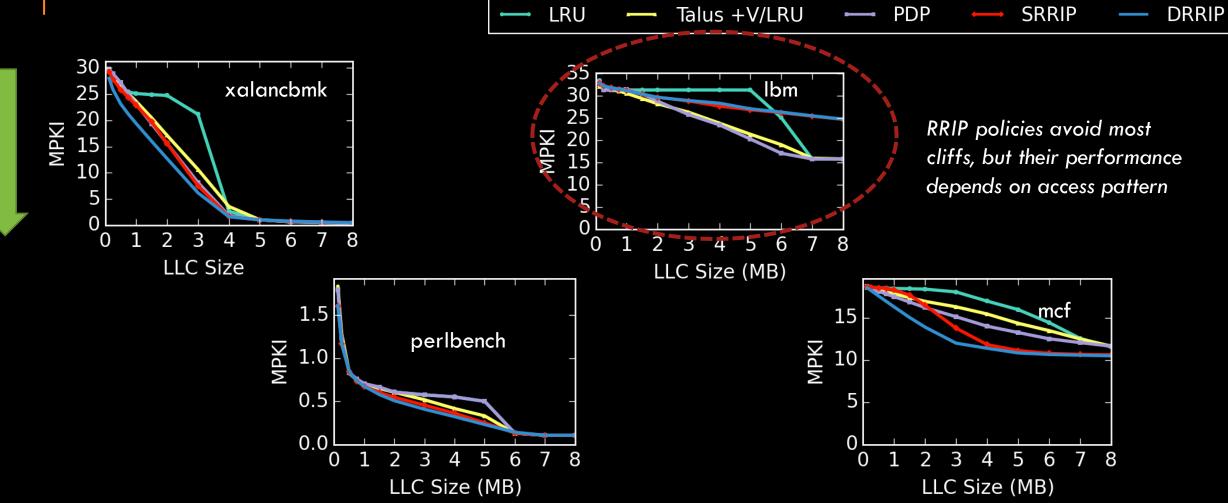




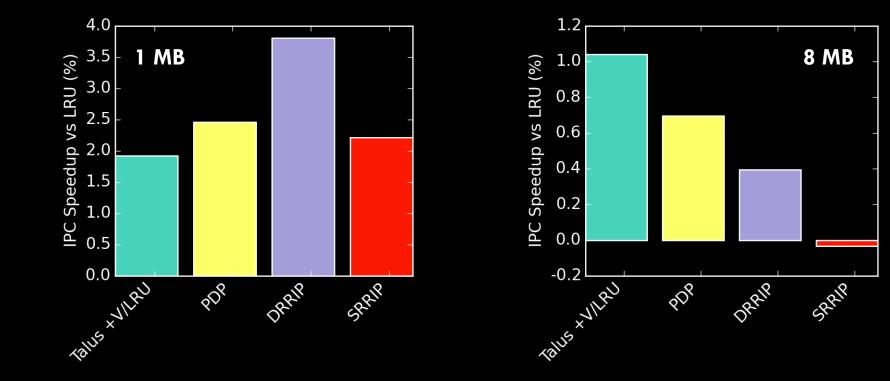
EVALUATION: SINGLE-THREADED



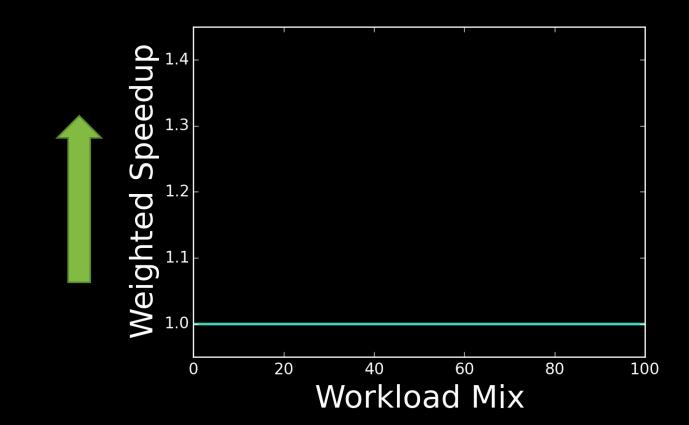
EVALUATION: SINGLE-THREADED



GMEAN IPC IMPROVEMENT VS LRU



Talus on LRU gets similar speedups to prior policies.



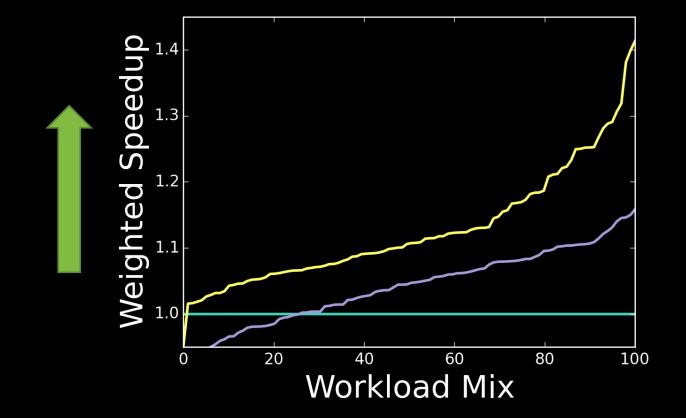
LRU

Speedup 1.3 1.3 1.2 Weighted 1.11.0 20 40 60 80 100 0 Workload Mix

LRU

Talus is convex → naïve hill climbing yields large performance gains

Talus +V/LRU (Hill)



LRU

Hill climbing alone does not improve performance much

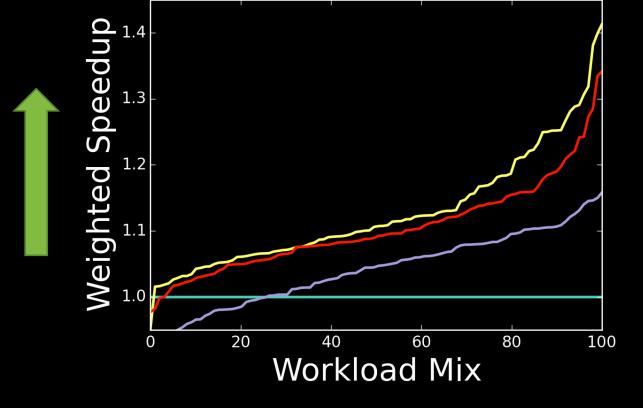
Hill

Talus +V/LRU (Hill)

Lookahead is close to Talus, but more expensive

Lookahead

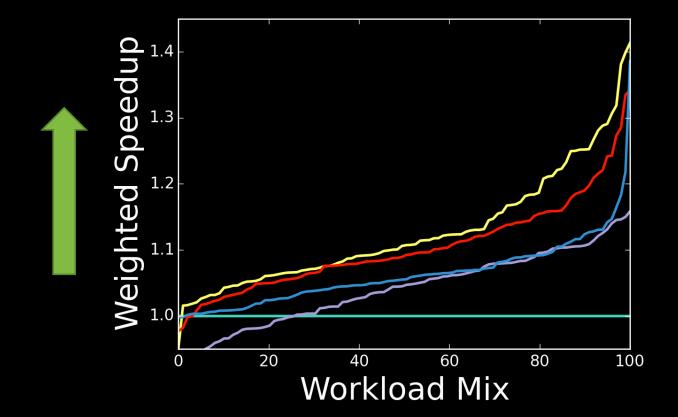
Hill



LRU

Talus +V/LRU (Hill)

LRU



Talus +V/LRU (Hill)

Hill

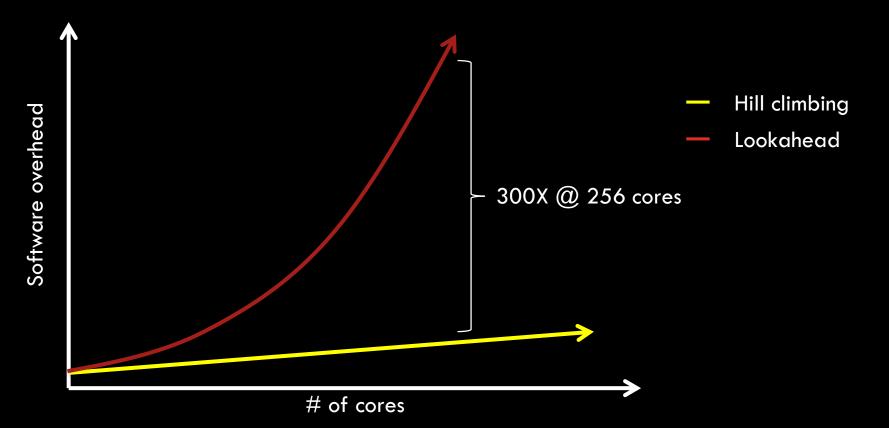
Partitioning techniques outperform high-performance policies on shared caches

Lookahead

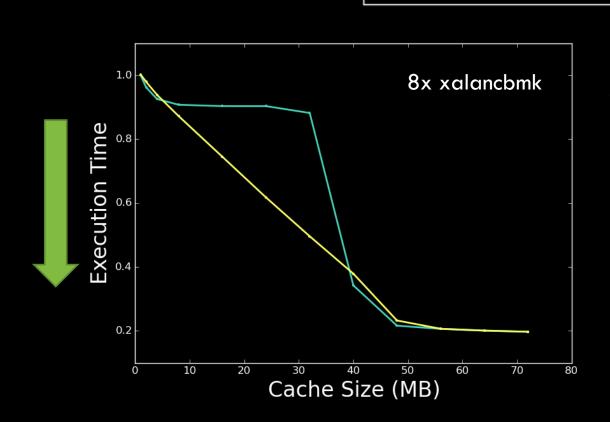
TA-DRRIP

TALUS SIMPLIFIES PARTITIONING ALGORITHMS AND REDUCES OVERHEADS

Efficient alternatives to Lookahead add significant complexity!



MULTI-PROGRAMMED FAIRNESS



LRU

Talus with fair (equal-sized) partitions decreases execution time without degrading fairness.

Talus +V/LRU (Fair)

See paper for other apps & schemes!

MORE CONTENT IN PAPER!

Detailed proofs

Prove optimal replacement is convex

Evaluation:

- Talus works on way partitioning
- Talus works with SRRIP
- More benchmarks
- Talus works with pre-fetching and multi-threading

THANK YOU!

- Talus avoids cliffs and ensures convexity
 - Proven under simple assumptions
 - Verified by experiment
- Analysis of shadow partitioning shows advantages vs bypassing
- Talus improves performance and simplifies cache partitioning
- Talus combines the benefits of high-performance replacement and partitioning