Domain-Specialized Cache Management for Graph Analytics



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Cache management in the age of big data

Variety of application domains

Data Analytics



Graph Analytics



Machine Learning



Working set size much larger than typical SPEC benchmarks - Vastly different cache access patterns across domains



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Working set size much larger than typical SPEC benchmarks - Vastly different cache access patterns across domains

Yet, cache management mechanisms are "domain-agnostic" - Assumption: one size fits all





A case for domain-specialized cache management



Domain-agnostic techniques for graph analytics

SHiP-MEM Hawkeye Leeway





Domain-agnostic techniques for graph analytics





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SHiP-MEM Hawkeye Leeway





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I-15% geomean slowdown



Outline

Performance of domain-agnostic cache management

- Graph analytics
- GRASP: domain-specialized cache management
 - Software-guided reuse-prediction
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- Performance evaluation





Applications of graph analytics

Extract meaningful information out of complex many-to-many relationships among objects

- **Community Analysis**
 - Identify customers with similar interests







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Extract meaningful information out of complex many-to-many relationships among objects

- **Community Analysis**
 - Identify customers with similar interests
- **Connectivity Analysis**
 - Find weakness in a network
- Path Analysis

- Route optimization for distribution and supply chain
- **Centrality Analysis**
 - Most influential people and information in social media
- And many others ...



Real-world graphs & power-law degree distribution

Small fraction of vertices have high connectivity – hot vertices

Large fraction of vertices have low connectivity – cold vertices

Prevalent in many domains – e.g., Twitter user-follower graph



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Donald Trump



~72M



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Donald Trump



How does connectivity influence cache locality?



A canonical example of graph analytics

Computes property for a vertex based on its neighbors' properties



Vertex Properties





A canonical example of graph analytics





A canonical example of graph analytics

Computes property for a vertex based on its neighbors' properties





Cache Accesses in Time



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Key observation: vertex reuse is proportional to its degree





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Hot vertices \rightarrow Small footprint + High reuse



Domain-agnostic techniques rely on purely hardware mechanisms



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Domain-agnostic techniques rely on purely hardware mechanisms

Vertex Properties

Example Graph P_0 V_0 V P Vo **V**3 V_2 V_2 P_2 Hot V_3 P_3 V_5 V_4 V_4 P_4 Hot V_5 P_5 Reason 1 Irregular Accesses





Domain-agnostic techniques rely on purely hardware mechanisms

Vertex Properties Example Graph P_0 V_0 V P V₀ **V**₂ V_3 V_2 P_2 Hot V_3 P_3 V_5 V_4 V_4 P_4 Hot V_5 P_5 Reason 1 Irregular Accesses Reason **2** Long Reuse Distances V_0 P₅ V_4 ۷_۶ P. V P₄ ٧, ٧, P₄ Cache Accesses in Time

P₄



Domain-agnostic techniques rely on purely hardware mechanisms

Vertex Properties Example Graph P_0 V_0 V P Vo **V**2 P_2 V_2 Hot P_3 V_3 V_5 V_4 V_4 P_4 Hot V_5 P_5 Reason 1 Irregular Accesses Reason **2** Long Reuse Distances

Idea: Leverage domain-knowledge for reuse prediction



Proposal: GRASP – a software-hardware co-design

Software aids hardware in identifying hot vertices



Hardware preferentially caches hot vertices



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GRASP: Software-guided reuse-prediction

Task: Let software aid hardware in identifying hot vertices



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Challenge: Non-trivial due to sparse distribution of hot vertices in memory Vertex Properties




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Idea: Leverage prior graph reordering optimization



Optimization: skew-aware graph reordering

Vertices are ordered in memory based on their assigned IDs

Changing vertex order to improve cache locality [IISWC'19]



HPCA'20

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Original Vertex Order



HPCA'20

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Original Vertex Order



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Vertices are ordered in memory based on their assigned IDs

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Original Vertex Order

New Vertex Order



HPCA'20

Vertices are ordered in memory based on their assigned IDs

Changing vertex order to improve cache locality [IISWC'19]



Easy to communicate the region boudary to hardware



GRASP: Region-based lightweight interface

Hot Vertices Cold Vertices 1 Preprocessing: Software applies skew-aware reordering



Hot

Vertices

Cold

Vertices

ics

GRASP: Region-based lightweight interface

HPCA'20



Region

Start

Architecturally exposed configuration registers

Preprocessing:
Software applies
skew-aware
reordering



GRASP: Region-based lightweight interface



 Preprocessing: Software applies skew-aware reordering

2 Initialization:Software populatesconfiguration registers





GRASP: Region-based lightweight interface





GRASP: Region-based lightweight interface





GRASP: Region-based lightweight interface



Software involvement is limited to initialization



GRASP: Reuse prediction at runtime





GRASP: Reuse prediction at runtime



Prediction is entirely done in hardware



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GRASP: Hardware-enforced cache management

Task: Preferentially cache hot vertices

Challenge: LLC capacity is limited - Not all hot vertices can fit



GRASP: Hardware-enforced cache management

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GRASP: Hardware-enforced cache management

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Requirement: Keep cache management flexible



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GRASP: Preferential but flexible cache management





GRASP: Preferential but flexible cache management





GRASP is simple!

Software

- Off the shelf skew-aware reordering optimization
- Compatible with multiple skew-aware reordering techniques



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Lightweight Interface

- Software configures a pair of registers at initialization
- No software dependency after initialization



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Hardware

- Lightweight address comparison logic to infer the reuse hint
- Trivial policy changes
- Minimal modifications to cache structure no additional metadata



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Accelerating graph analytics at minimal cost



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Performance evaluation



Evaluation methodology

Evaluated 25 benchmarks (5 applications x 5 graph datasets)

- Graph applications from the Ligra framework [PPoPP'I3]
- Graph datasets are 0.3GB 8GB in Compressed Sparse Row (CSR) format

Datasets are reordered using DBG [IISWC'19]

- Degree-Based Grouping is state-of-the-art skew-aware reordering

Evaluated on the Sniper simulator [TACO'14]

- 8 Out of Order cores
- I6MB shared LLC (2MB per core)





Domain-agnostic techniques vs GRASP

SHiP-MEM Hawkeye Leeway





Domain-agnostic techniques vs GRASP

SHiP-MEM Hawkeye Leeway GRASP





Domain-agnostic techniques vs GRASP





More results in paper

Evaluation of pinning-based techniques

Evaluation of GRASP on low-/no-skew graph datasets

Evaluation of GRASP on top of other reordering schemes

... and more


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Key take away: one size does NOT fit all



Look beyond domain-agnostic cache management



HPCA'20

Thank You



Priyank Faldu

Source code https://github.com/faldupriyank

Personal website www.faldupriyank.com

I am on the job market