Graph Prefetching Using Data Structure Knowledge

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Background and Motivation

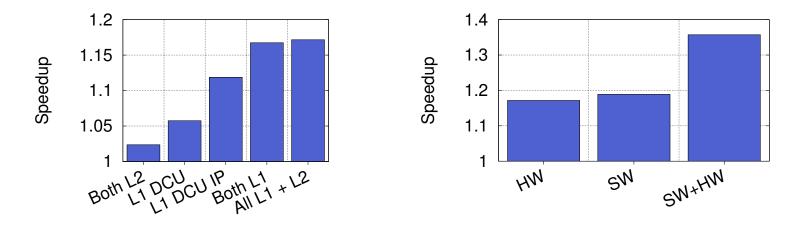
Graph applications are memory latency bound

- Caches & Prefetching are existing solution for memory latency
- However, irregular access patterns hinder their usefulness

Key insight: accesses seem irregular at individual load/store level, but have **predictable structure when we consider the high-level algorithm** •e.g. Breadth-first search (BFS)

Background and Motivation

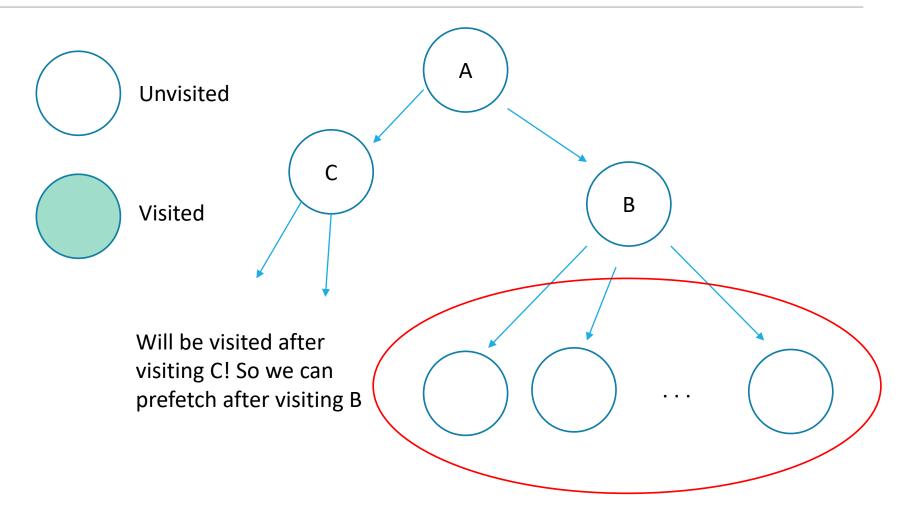
Existing SW/HW prefetching is insufficient



(a) Hardware prefetchers (b) Hardware vs software

Figure 3: Hardware and software prefetching on Graph 500 search with scale 21, edge factor 10.

Background and Motivation - BFS



Prefetching with Algorithmic Knowledge

Design a hardware prefetcher that relies on access patterns specific to algorithms

- Target BFS, but can support a wider range of algorithms/access patterns
- Specific to Compressed Sparse Row (CSR) format
- Prefetcher snoop reads/writes from L1 cache

Achieve an average of 2.3x speedup

Review: Compressed Sparse Row (CSR)

Sparse representation, with a vertex list indirecting to an edge list

• Authors add a visited list and work list specifically for BFS

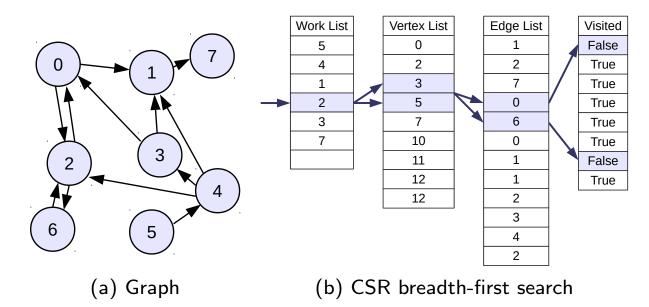
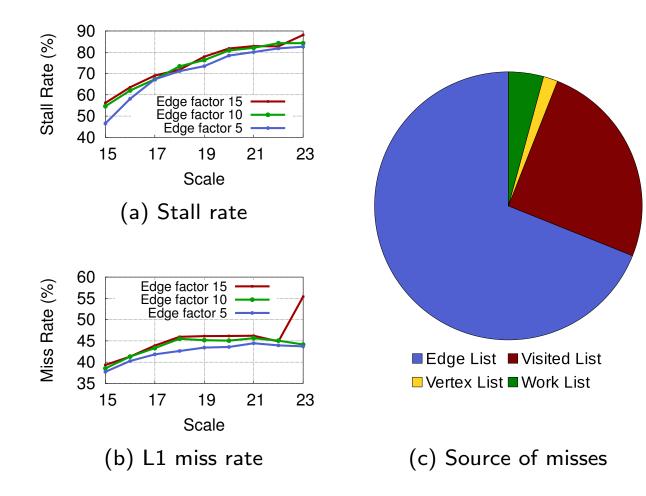


Figure 1: A compressed sparse row format graph and breadth-first search on it.

Poor Locality of Accesses in Graphs



Overview of Approach

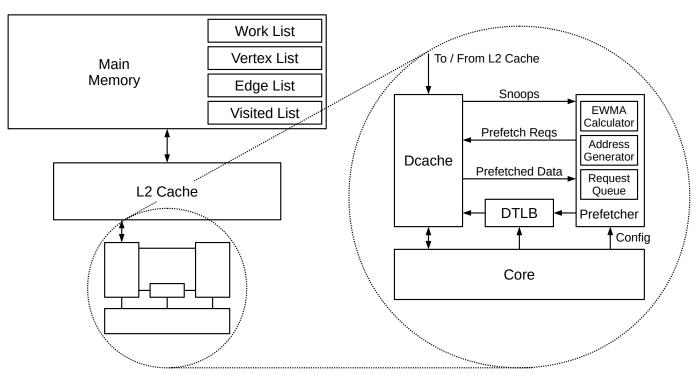
Prefetch all relevant data of *o*-distance away from the current worklist entry:

visited[edgeList[vertexList[workList[n+o]]]]

Prefetcher snoops the core-to-L1 mem. accesses to determine which data to prefetch

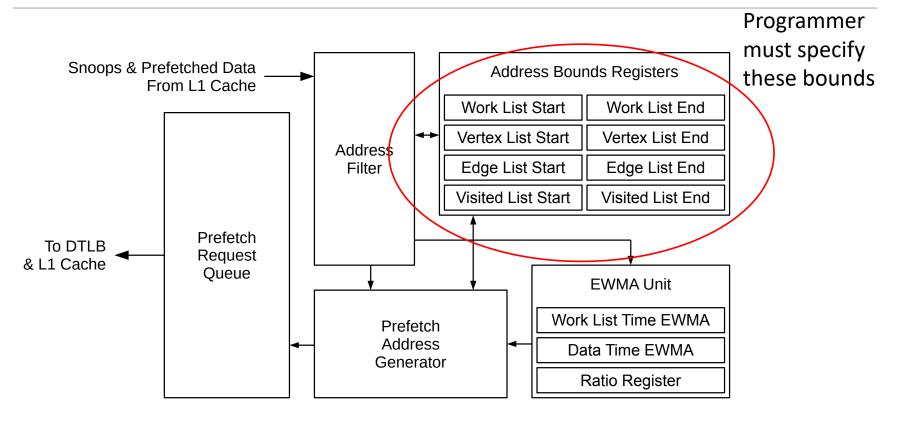
Vertex-Offset Mode	
Observation	Action
Load from workList[n] Prefetch vid = workList[n] Prefetch from vertexList[vid]	Prefetch workList[n+o] Prefetch vertexList[vid] Prefetch edgeList[vertexList[vid]] to
Prefetch vid = edgeList[eid]	<pre>edgeList[vertexList[vid+1]] (12 lines max) Prefetch visited[vid]</pre>

System Architecture



(a) System overview

Prefetcher Microarchitecture



(b) Prefetcher microarchitecture detail

Determining Prefetch Distance

Easy Case: Time to process a vertex (*work_list_time*) is less than time to pre-fetch the next vertex (*data_time*)

$$o * work_list_time = data_time$$

 work_list_time and data_time vary wildly => use exponentially weighted moving averages (EWMA)

 • Use a safe bound because EWMA often underestimates data_time:

 $o = 1 + \frac{k * data_time}{work_list_time}$

Determining Prefetch Distance

Problem: work_list_time > data_time

- Pre-fetched data is not used timely, might get kicked out of cache before it is used!
- Happens with high-degree vertices

Solution: Large vertex mode

- Base prefetch on how far along we have processed the high-degree vertex
 - »Possible because we know the range of the edge indices
- Prefetch within edgeList for larger vertex
- Fetch need vertex in worklist when almost done with current vertex's edges

Extensions

Technique can be extended to other algorithms: • Parallel BFS

Sequentially scanning vertex and edge data (e.g. PageRank)

Methodology

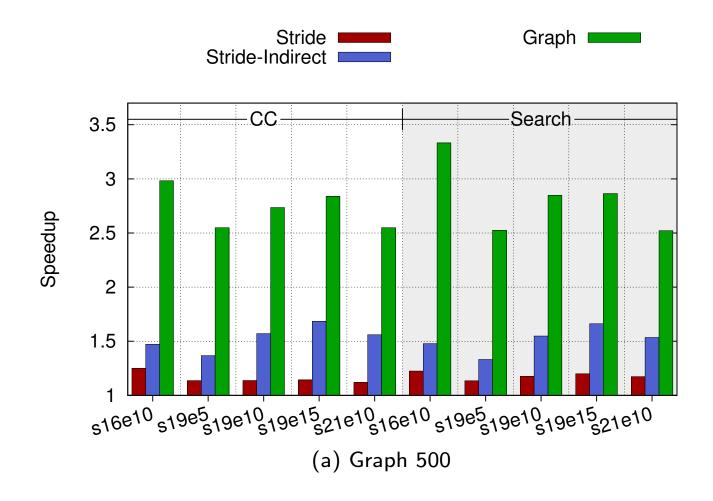
gem5 simulator

Set of algorithms from Graph500 and the Boost Graph Library:

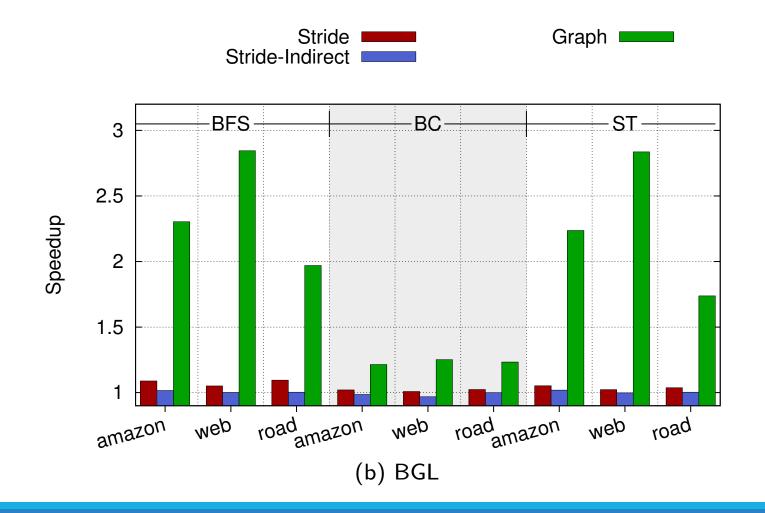
 BFS-like traversal: Connected components, BFS, betweenness-centrality, ST connectivity

•Sequential access: PageRank, sequential coloring

Evaluation - BFS-like traversal



Evaluation - BFS-like traversal



Significantly improved L1 hit-rate

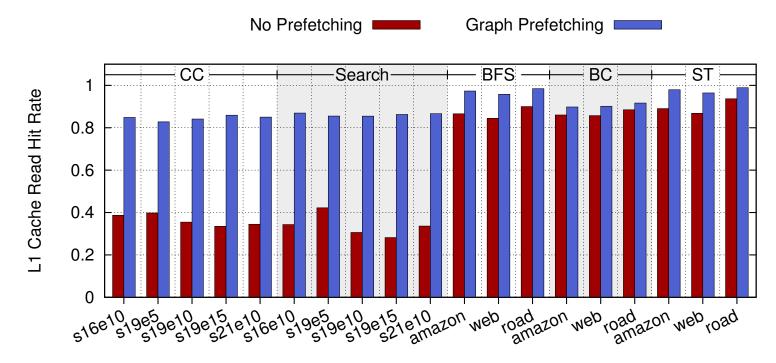


Figure 7: Hit rates in the L1 cache with and without prefetching.

Prefetching has low overheads

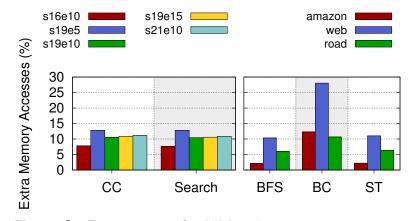


Figure 8: Percentage of additional memory accesses as a result of using our prefetcher.

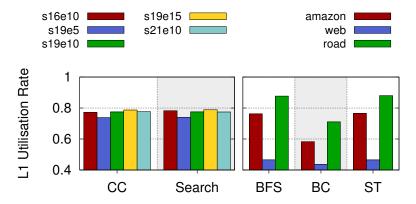


Figure 9: Rates of prefetched cache lines that are used before leaving the L1 cache.

Prefetching Analysis

Most of the benefit comes from prefetching visited & edge lists -> as expected!

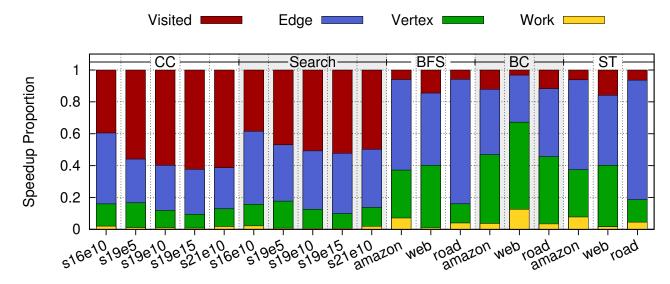


Figure 10: The proportion of speedup from prefetching each data structure within the breadth first search.

Prefetching works for other traversal types

Example: parallel BFS

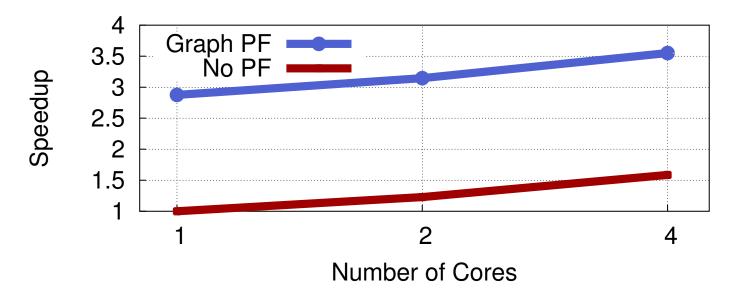


Figure 11: Speedup relative to 1 core with a parallel implementation of Graph500 search with scale 21, edge factor 10 using OpenMP.

Prefetching works for other traversal types

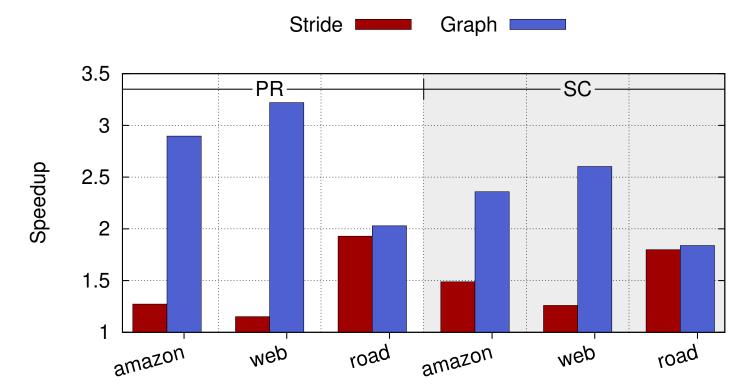


Figure 12: Speedup for different types of prefetching when running PageRank and Sequential Colouring.

Conclusion

Prefetching with knowledge of the graph traversal order significantly improves its performance

Works for different traversal types (BFS, sequential scan, ...)