

# ChunkGraph: Large Graph Processing with Chunk-Based Graph Representation Model

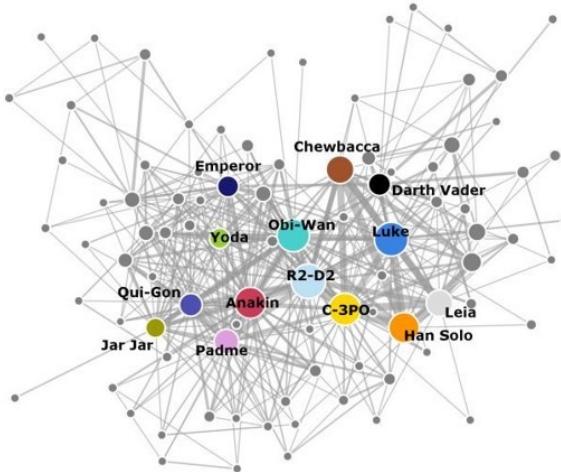
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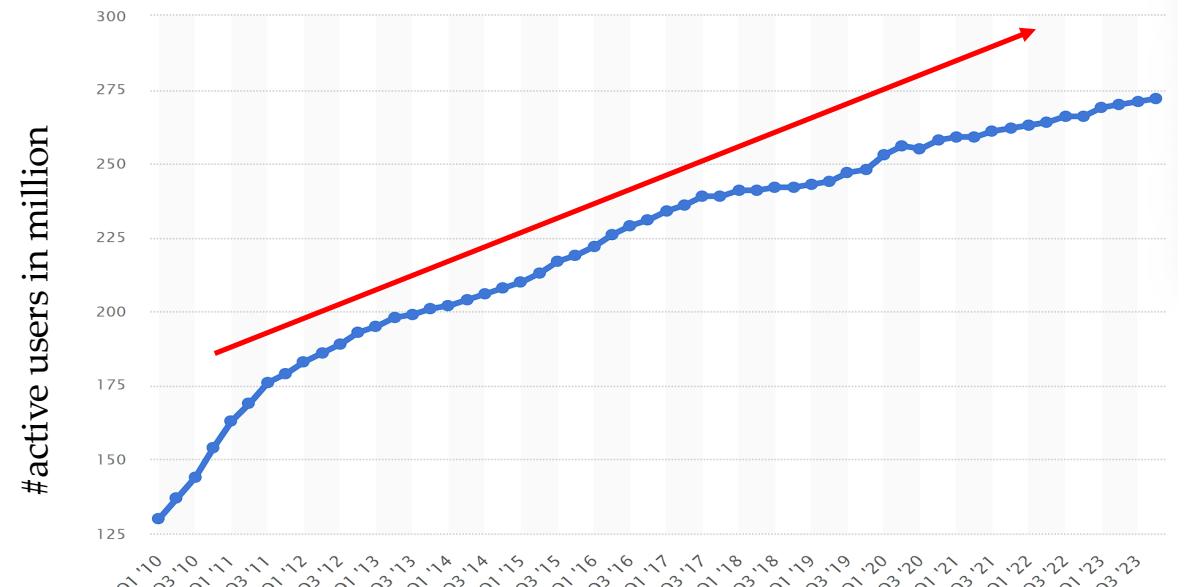
USENIX ATC 2024

# Explosive Growth in Graph Data Analytics

➤ Graph data



➤ Real-world graph datasets are **continuously growing**



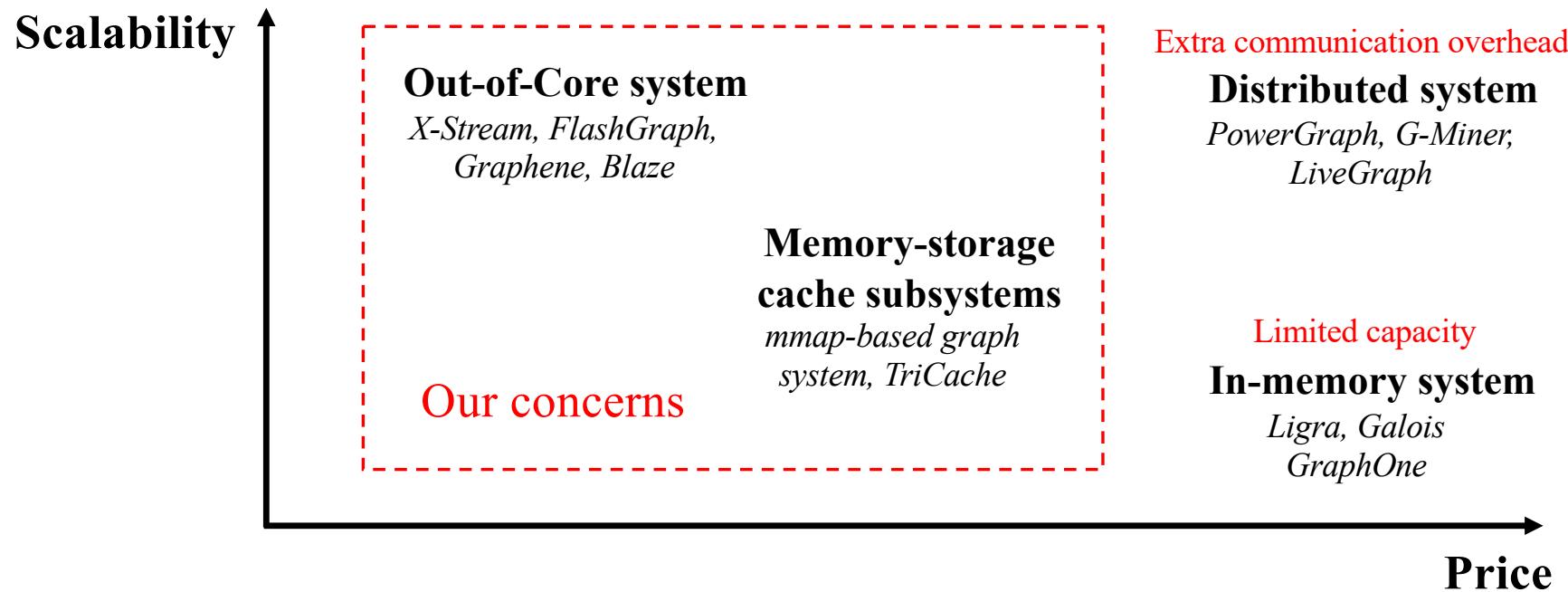
➤ Graph Application

- Social networks, webpage links, recommendation systems

#Facebook monthly active users between 2010 and 2023, Statista

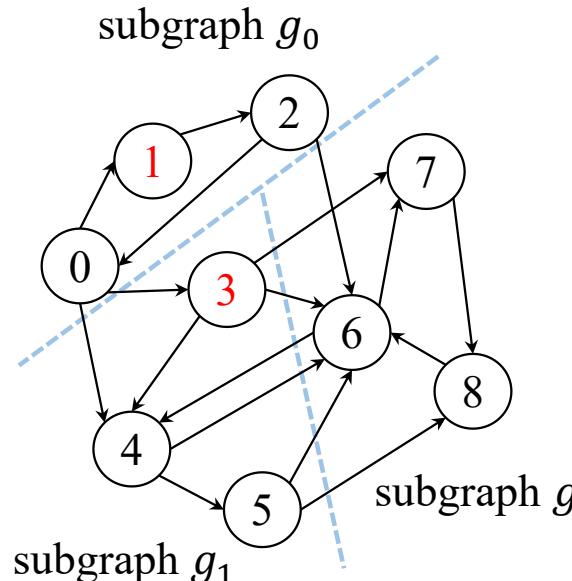
# Different Graph Systems Supporting Large Graph Processing

- Objectives for large graph processing systems:
  - Cost-effective, Scalable, Programming-friendly

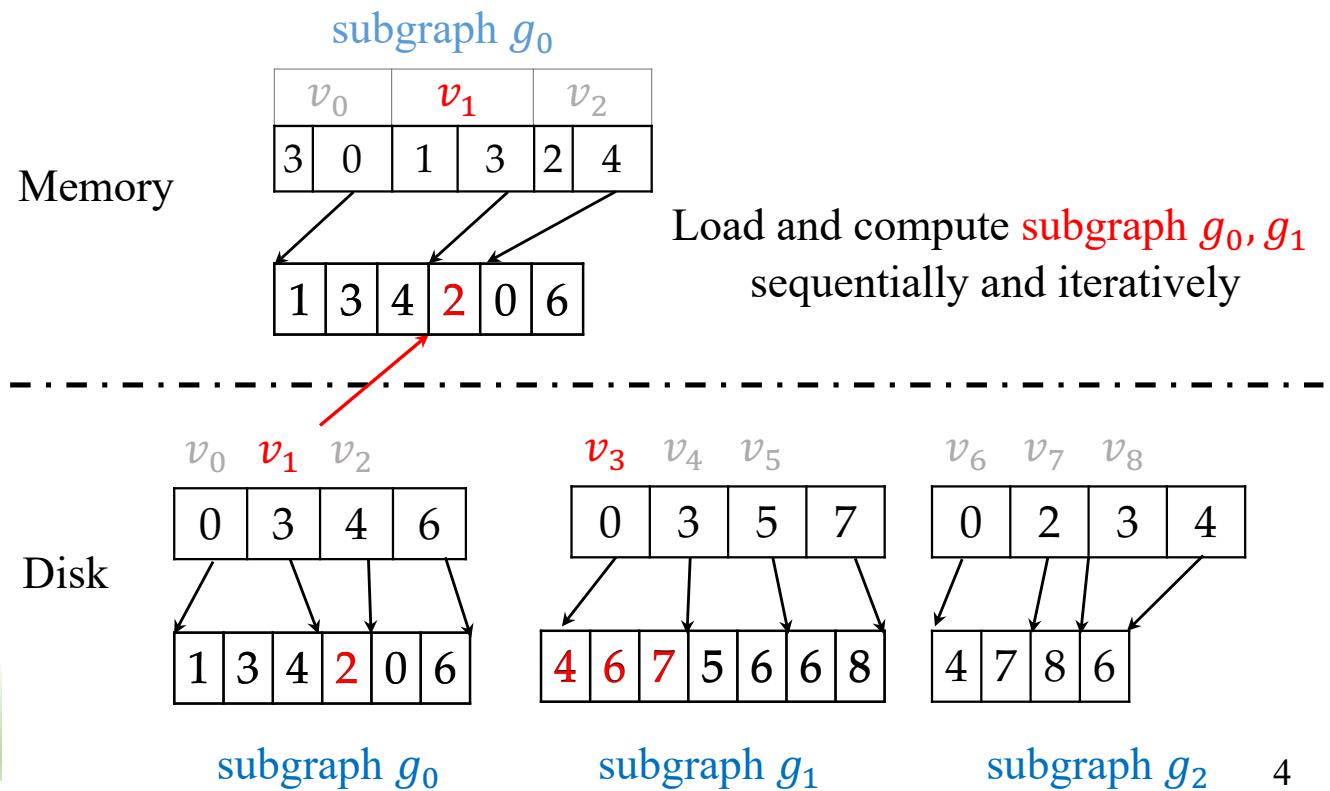


# SOTA Subgraph-based Out-of-core Graph Systems

- **Subgraph-based iterative model** divides the whole graph into **disjoint** intervals
  - Sequentially load each subgraph from disk during each iteration (e.g. access  $v_1, v_3$ )



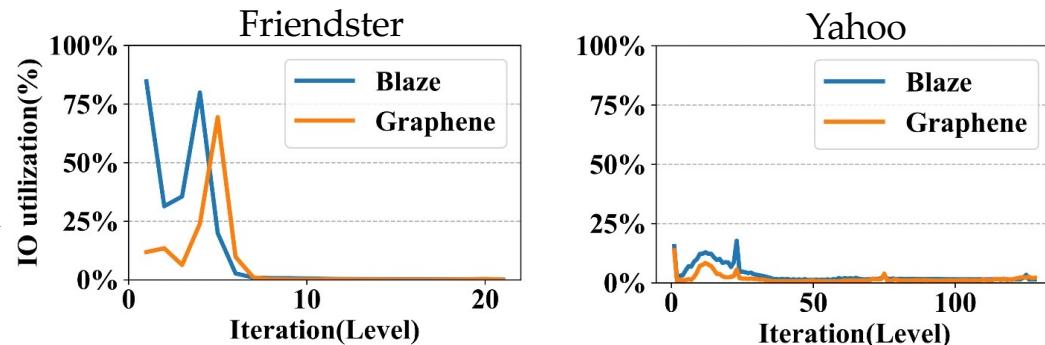
Merge multiple random I/O ops  
into fewer sequential I/O ops



# Limitations for Subgraph-based Iterative Model Graph Systems

## P1 Low I/O efficiency

Access single vertex → load whole subgraph



Average I/O utilization is lower than **13%** for BFS.

## P2 Extra computing overhead

Subgraph synchronization overhead

Blaze requires up to **154x** more CPU instructions compared to in-memory system Ligra's mmap variant.

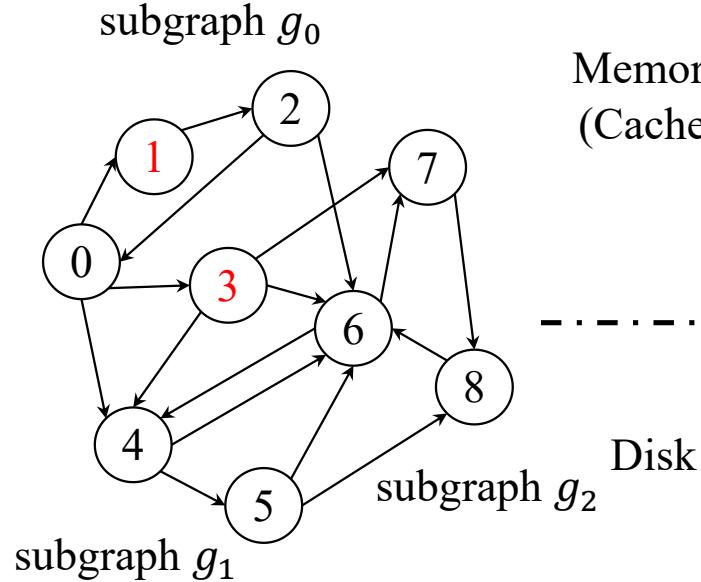
## P3 Expensive algorithm development costs

Extra implementation for I/O management

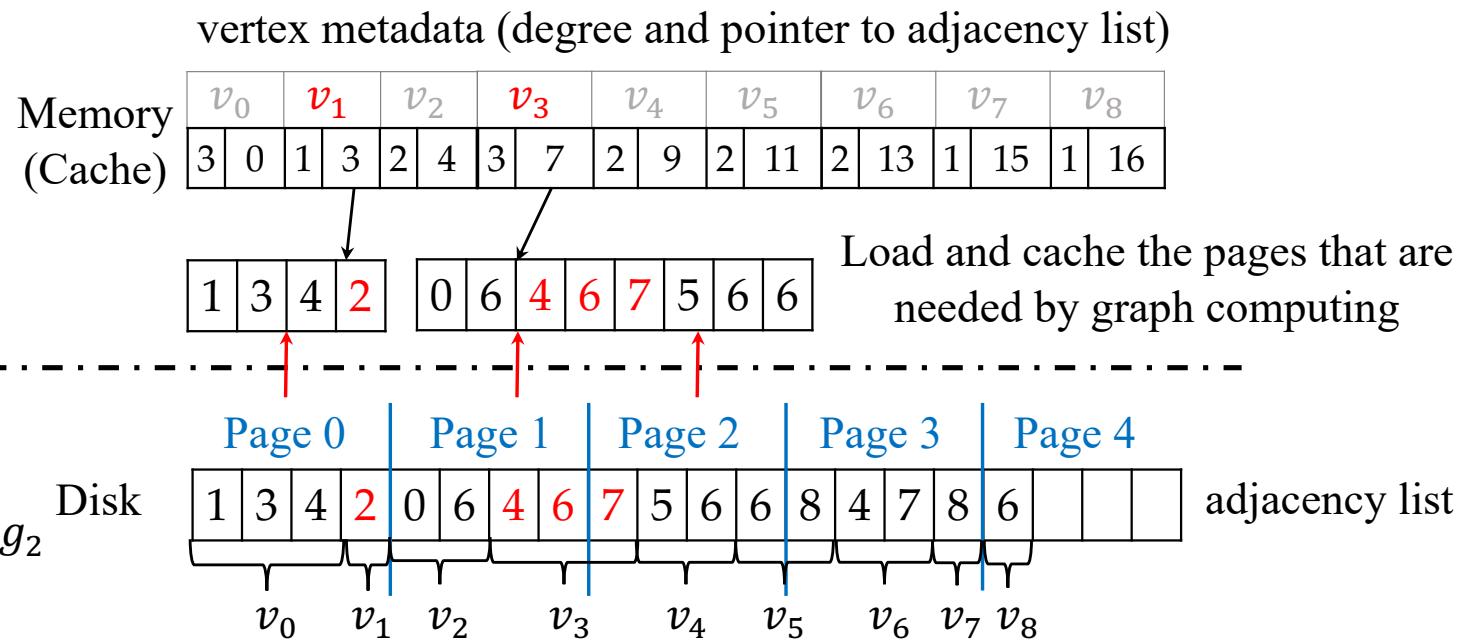
#line of codes	BFS	BC	PageRank	KCore
Ligra	34	123	77	86
Graphene	420	-	763	476
Blaze	75	197	159	133

# Alternative: Memory-Storage Cache Subsystems

- Using **page cache based mechanism** to cache data from external storage



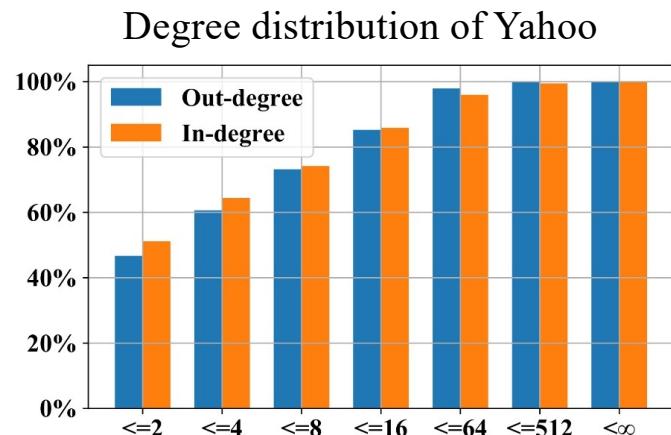
Fine-grained for page access  
compared to subgraph access



( $v_i$  indicate the adjacency list of vertex  $i$ )

# Limitations for Memory-Storage Cache Subsystems

- **Problem 1:** Mismatch between **page granularity** and vertex access
  - Real graph datasets behave power law degree distribution



Low-degree vertices → **Poor I/O efficiency**

- **51.17%** of non-sink vertices have only **one or two** in-neighbors

High-degree vertices → **Massive page table entries**

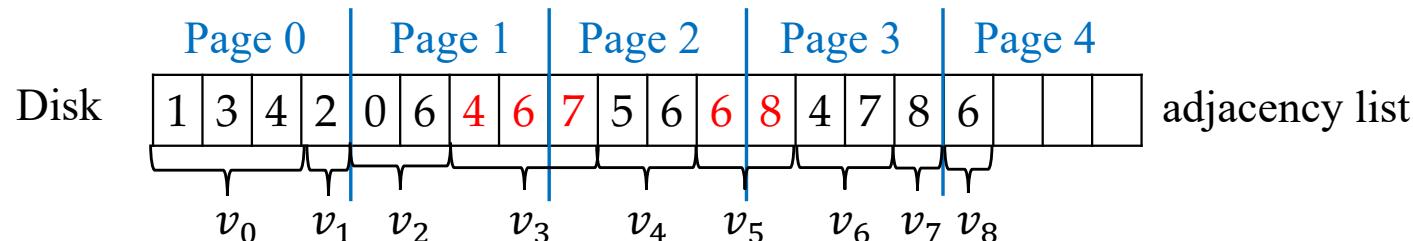
- **0.09%** of non-sink vertices account for **58.44%** of total edges
- **7459** 4KB-pages are needed for largest vertex's neighbors

**Solution:** Use **different storing strategy** for vertices with different degrees

# Limitations for Memory-Storage Cache Subsystems

## ➤ Problem 2: Vertex cut

- A vertex smaller than one page is placed **across two adjacent pages** (e.g.  $v_3, v_5$ )



( $v_i$  indicate the adjacency list of vertex  $i$ )

↓  
access  $v_1, v_3$

CSR format: access 3 Pages

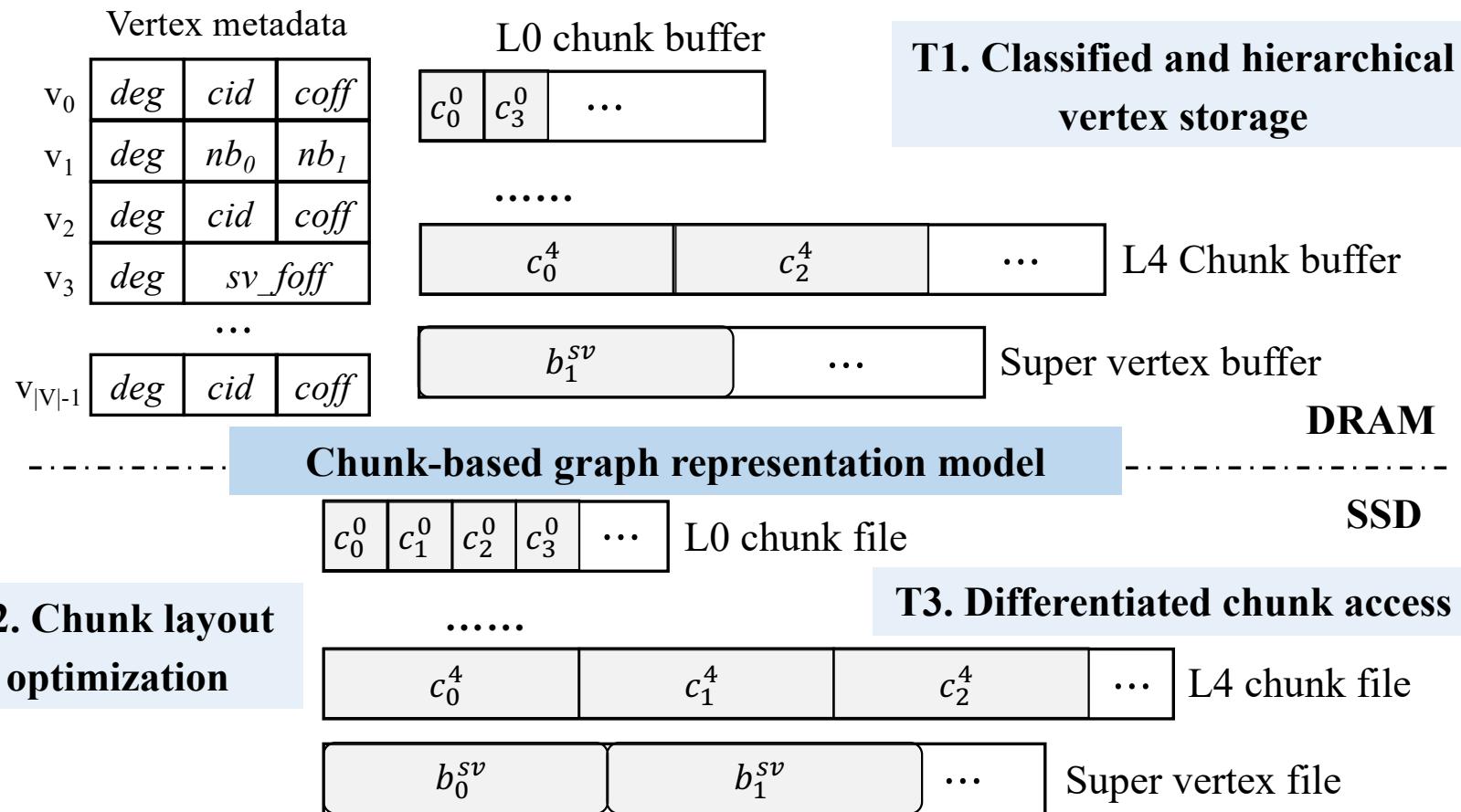
1	3	4	2	0	6	4	6	7	5	6	6
---	---	---	---	---	---	---	---	---	---	---	---

Best case: access 1 page

2	4	6	7
---	---	---	---

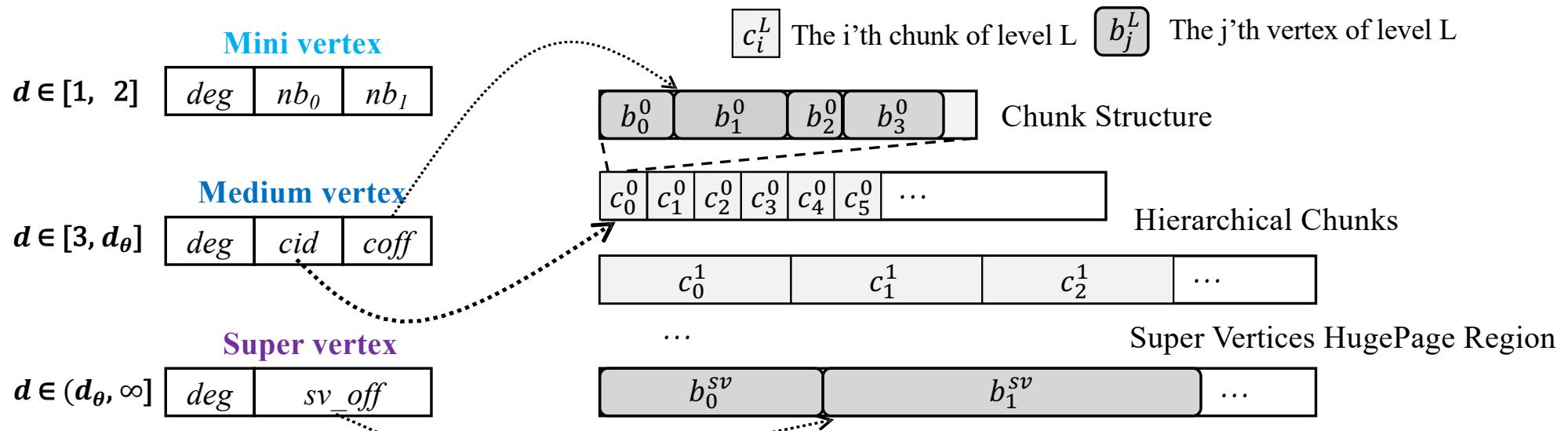
Target: **Minimize the number of page accesses for each query**

# ChunkGraph: I/O efficient chunk-based graph representation model



# Technique 1.1: Classified Hierarchical Vertex Storage

- All vertices are classified into **three categories** according to their degrees
  - **Mini vertex:** in-index storing without additional storage cost and indirect addressing
  - **Medium vertex:** chunk-based storing **without vertex cutting issues**
  - **Super vertex:** HugePage-based storing with lower page table and TLB overhead



## Technique 1.2: Hierarchical Chunk Management

- Hierarchical chunk size according to medium vertices' degree

- E.g. a four-layer hierarchical chunk implementation

$c_i^L$  The i'th chunk of level L

$d \in [3, 1021]$   
L0 4KB chunks

$b_j^L$  The j'th vertex of level L

$d \in [1022, 7168]$   
L1 32KB chunks

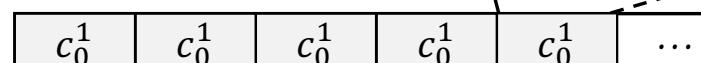
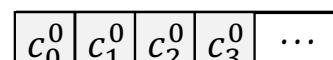
chunk level

$deg$   $cid$   $coff$

Medium vertex

$d \in [7169, 58365]$   
L2 256KB chunks

$d \in [58366, 465920]$   
L3 2MB chunks



Chunk Structure

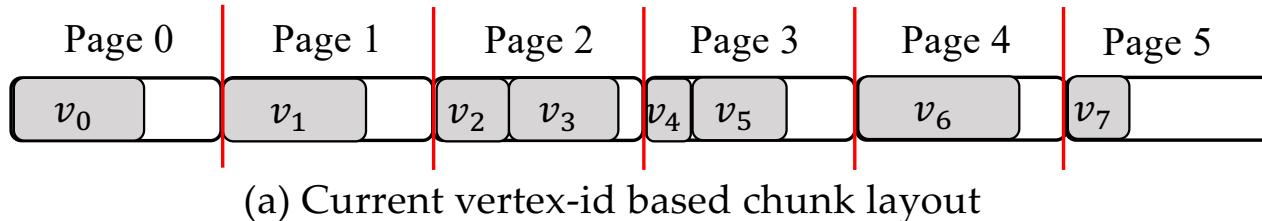
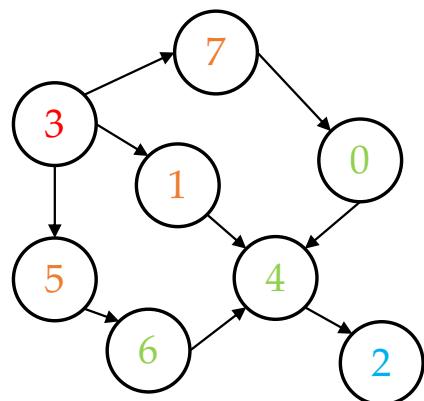


Differentiated chunk buffer sizes according to each layer proportion

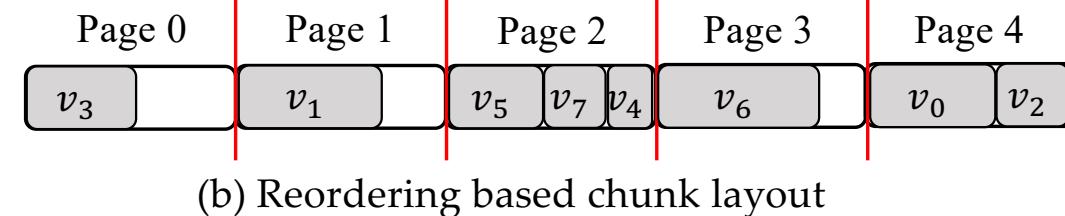
$$S_i = \frac{S_i \times M}{\sum_1^L S_j} \quad \begin{aligned} M: & \text{total available memory size} \\ S_i: & \text{chunk file size of layer } i \end{aligned}$$

## Technique 2.1: Reordering based Chunk Layout Optimization

- Observation: A vertex is likely to be accessed after its neighbors or sibling vertices accessed
- E,g. Run BFS on root 3. (Vertex access order: 3, 1, 5, 7, 4, 6, 0, 2)



Iter1: P2  
Iter2: P1, P3, P5  
Iter3: P0, P3, P4  
Iter4: P2



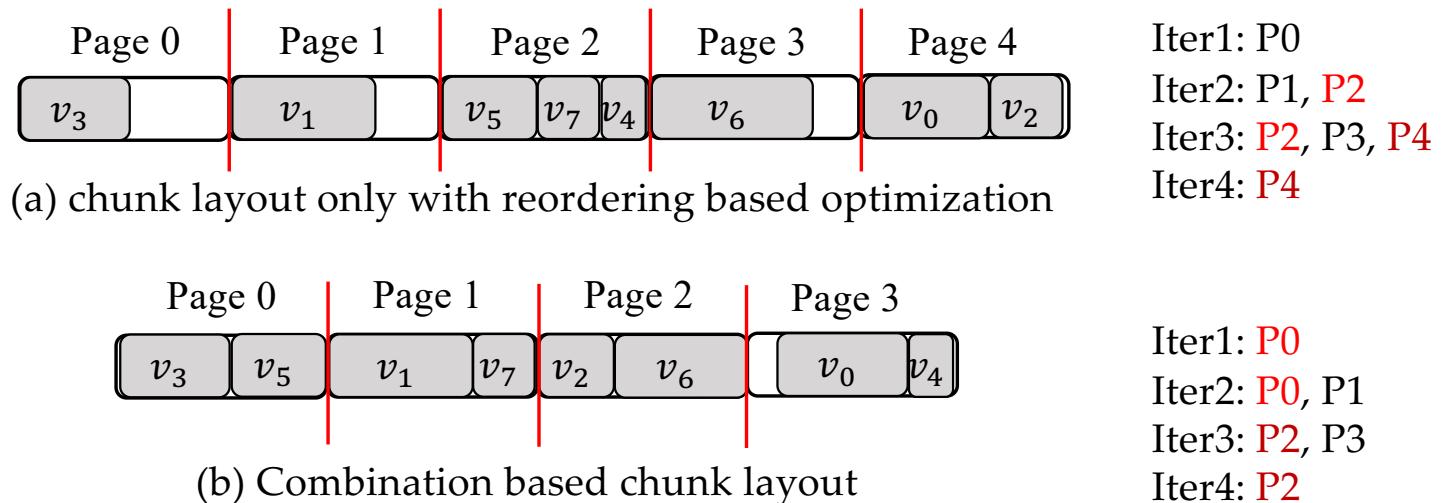
Iter1: P0  
Iter2: P1, P2  
Iter3: P2, P3, P4  
Iter4: P4

Better temporal locality, buffer cache hit ratio, providing opportunities for sequential I/O

Problem: reordered optimization still suffers from **inter-fragementation** within chunk

## Technique 2.2: Vertex-combination Chunk Layout Optimization

- Solution: combine the vertices with complementary degree into one chunk

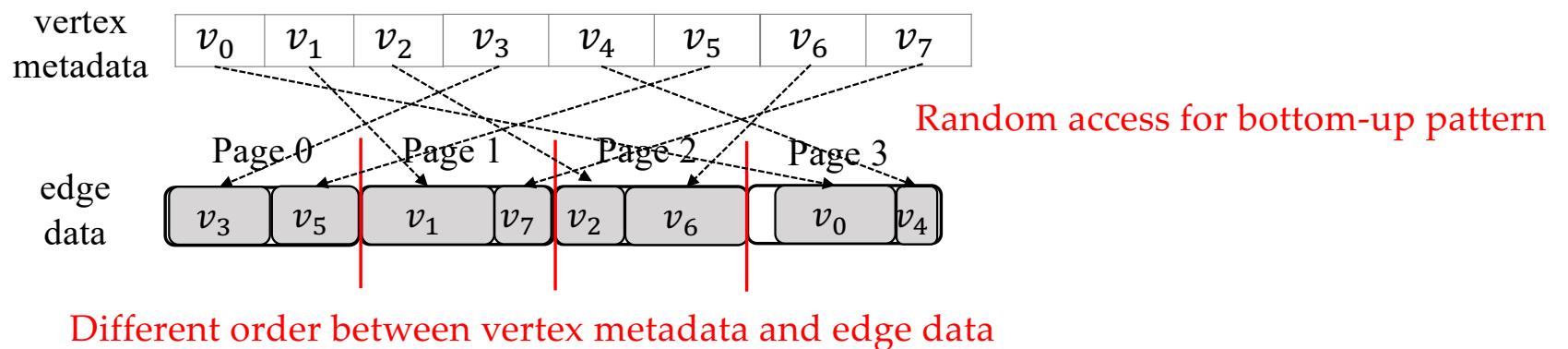


Less intra-chunk fragmentation and better spatial locality, minimizing chunk access

#Fragmentation chunk is reduced from **95.24% to 52.77%** on YahooWeb dataset.

## Technique 3: Differentiated Chunk Access Optimization

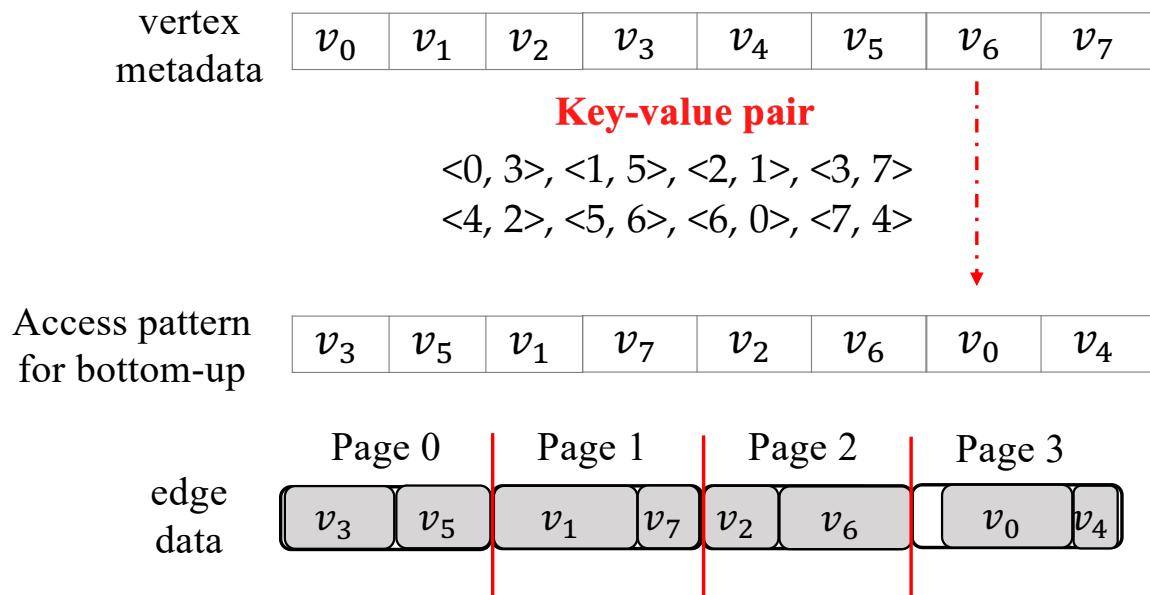
- Graph algorithms usually involves different graph access pattern
  - **Top-down:** sparse access, only activated vertices
  - **Bottom-up:** dense access, scan the whole graph in vertex ID order



For bottom-up algorithm, we should traverse all vertices according to **edge data order**

## Technique 3: Differentiated Chunk Access Optimization

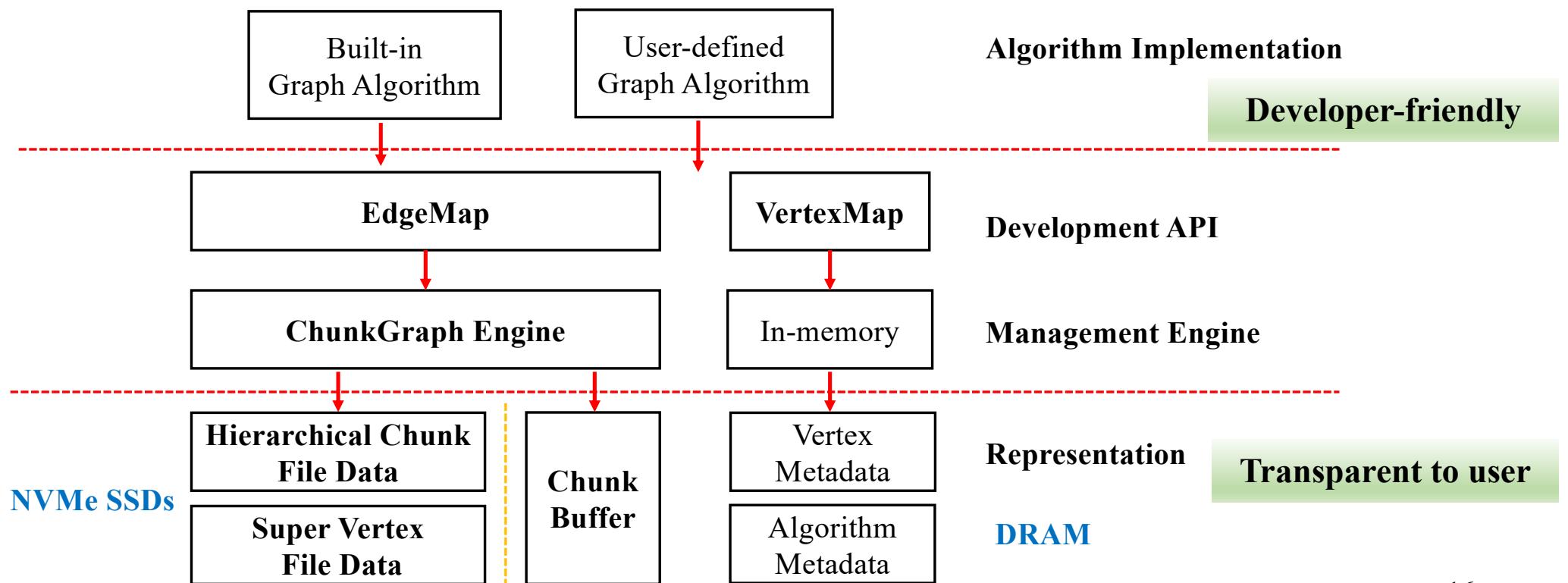
- **Differentiated chunk access** pattern for bottom-up access
  - Store **key-value pair** <reordered\_id, vid> to support chunk order access



**Avoid random access** for vertices due to reordering optimization

# Prototype System and Implementations

- **ChunkGraph** is implemented based on Ligra's graph interface



# Evaluation settings

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## ➤ Testbed

- A server with 2 sockets, each with 24 physical cores
- **8 \* 16GB = 128GB DRAM + 2 \* 3.84TB SSD**

## ➤ Graph datasets

Dataset	V	E	CSR Size	Chunk Size
Twitter (TT)	61.6M	1.5B	11.9GB	13.5GB
Friendster (FS)	68.3M	2.6B	20.3GB	21.2GB
UKdomain (UK)	101.7M	3.1B	26.2GB	27.5GB
YahooWeb (YW)	1.4B	6.6B	70.5GB	77.8GB
Kron29 (K29)	512M	8B	72GB	78.2GB
Kron30 (K30)	1B	16B	144GB	156.3GB

Real World Graph

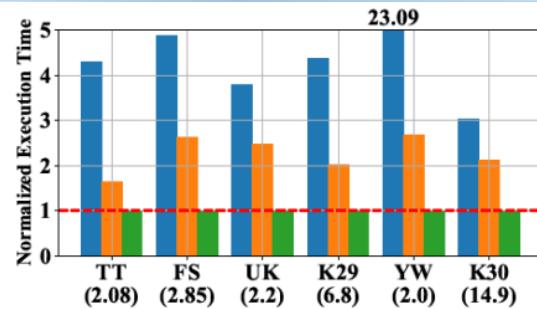
Synthetic Graph

# Evaluation settings

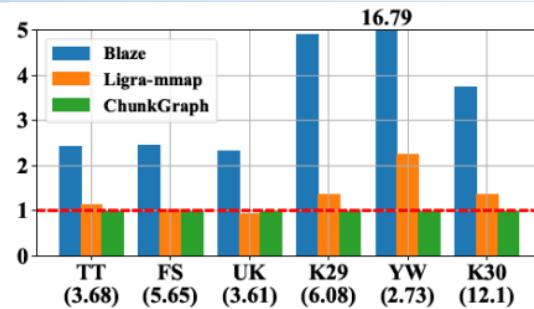
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- Comparison systems
  - **Blaze**
    - ✓ The SOTA out-of-core graph system optimized for modern fast SSDs
  - **Ligra-mmap**
    - ✓ Ligra's variant using mmap to map the graph data files into the virtual memory space
- Evaluation metrics
  - Graph query performance
    - ✓ BFS, SSSP, BC, Kcores, Radii, PageRank
  - I/O overhead
  - Computation overhead

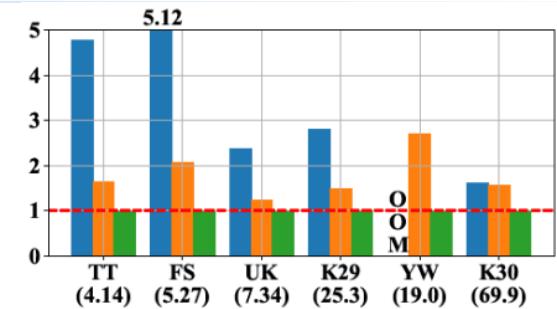
# Evaluation 1: Graph query performance



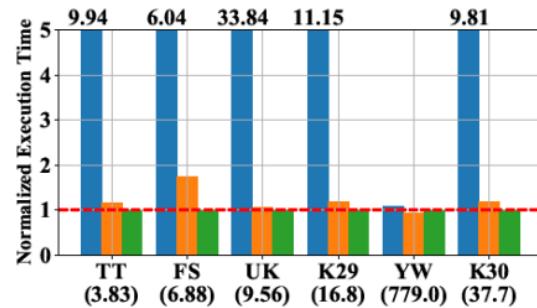
(a) BFS



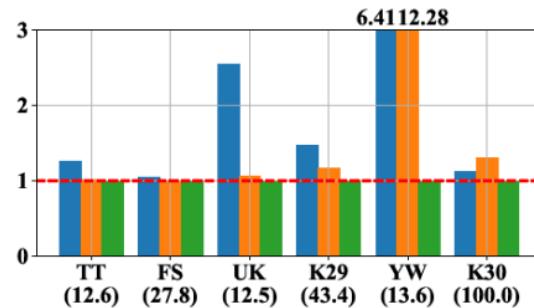
(b) SSSP



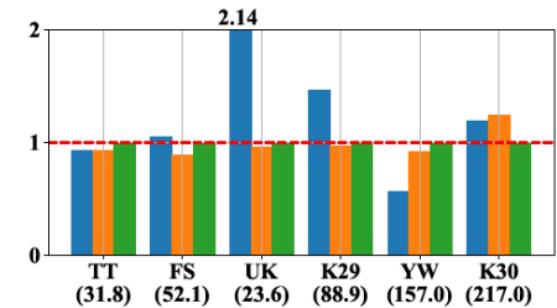
(c) BC



(d) Kcores



(e) Radii

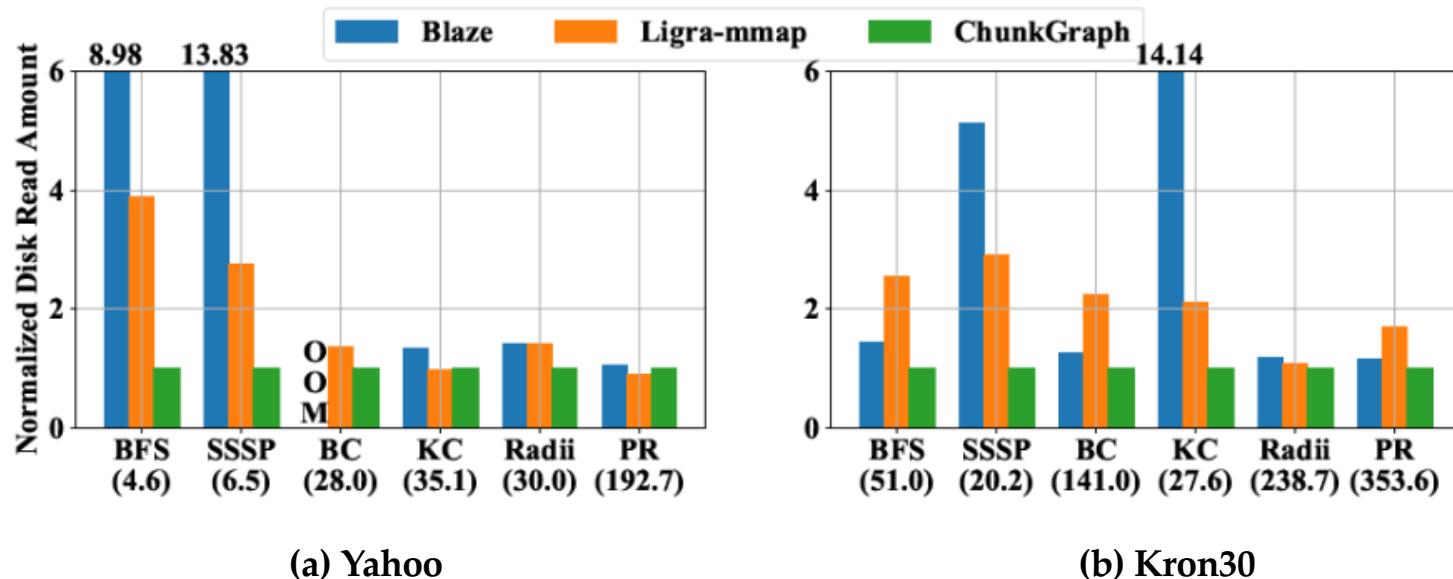


(f) PageRank

ChunkGraph achieves **1.62x-23.09x** speedup upon Blaze, and **1.08x-2.94x** compared to Ligra-mmap on sparsely accessed algorithms BFS, SSSP, BC.

## Evaluation 2: I/O overhead

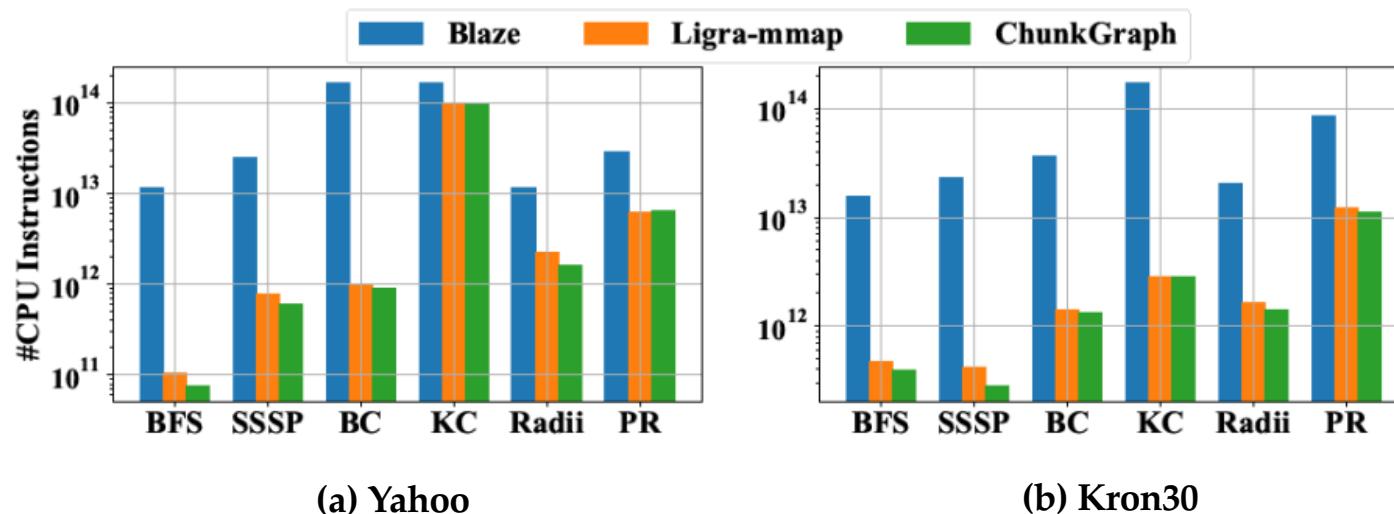
- Disk read amount of different algorithms on Yahoo and Kron30



ChunkGraph reduces disk read amount by **4.68×** and **1.98×** on average, compared to Blaze and Ligra-mmap respectively.

## Evaluation 3: Computation overhead

- CPU instructions executed during different algorithms' execution



ChunkGraph reduces the number of CPU instructions by **185.01×** compared to the external memory graph system Blaze.

## Conclusion

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- **ChunkGraph**: an I/O efficient external graph system for processing large-scale graphs
  - **Classified and hierarchical** vertex storage strategy
  - Chunk layout optimization based on **vertex reordering and combination**
  - **Differentiated chunk access** optimization
  - Encompass both out-of-core systems and memory-storage cache subsystems
- More evaluation results and analysis are in the paper
- The source code is at <https://github.com/ZoRax-A5/ChunkGraph>

Thanks for your attention!

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