

VCSR: Mutable CSR Graph Format Using Vertex-Centric Packed Memory Array

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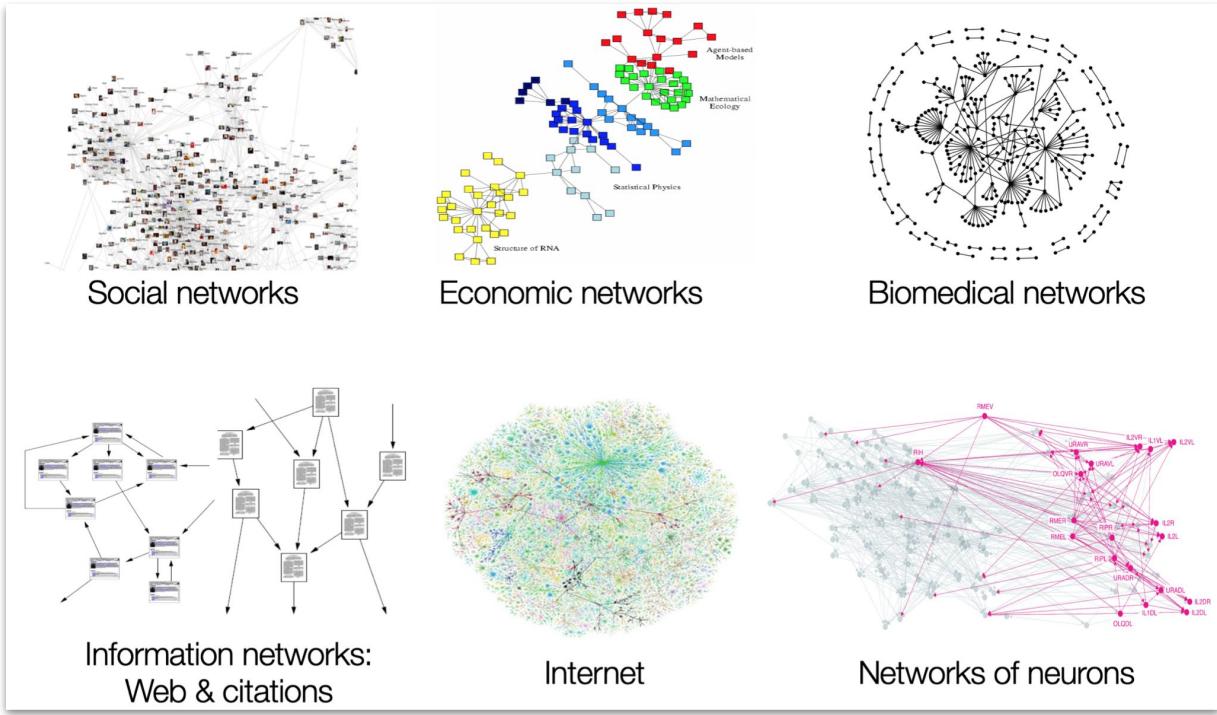
Dazhao Cheng
dcheng@whu.edu.cn

Wuhan University
Wuhan, China

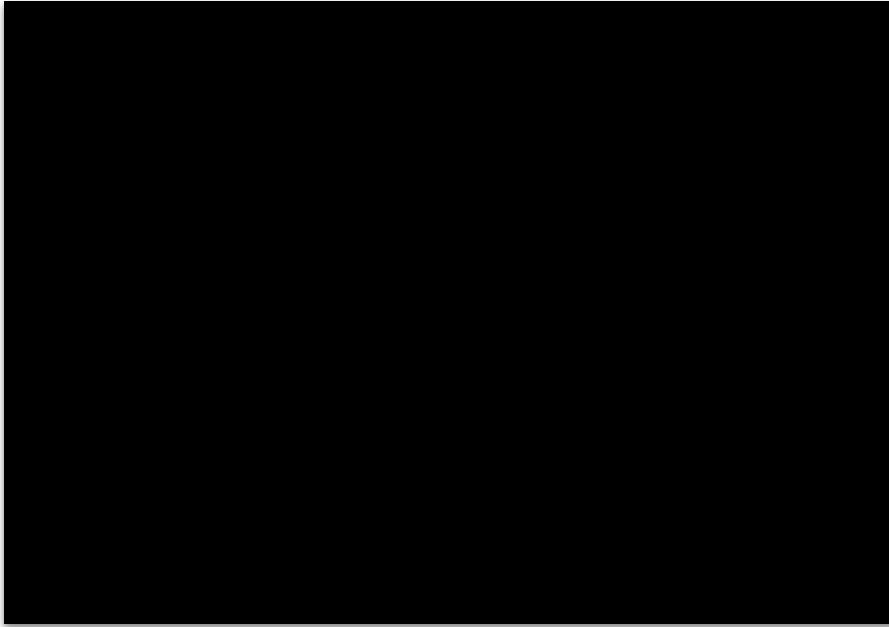


Background: Graph

Graphs are everywhere!



Background: Dynamic Graph



Graphs are dynamic!

Background: Graph Processing Systems

- Dynamic graph storage data structures

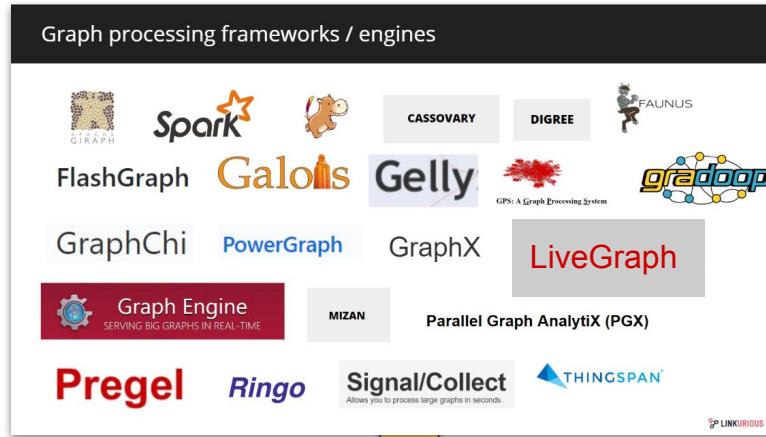
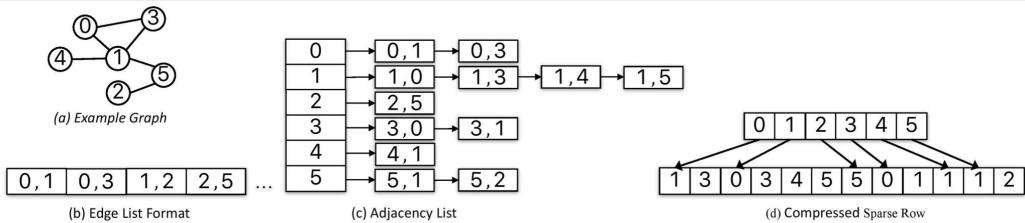


Image Source:
<https://www.cs.cmu.edu/~hranjan/BigData2013-10-GraphProcessing.pdf>

Background: Graph Processing Systems

- Dynamic graph storage data structures
- Expectations
 - Expectation-1: Efficient graph construction
 - Expectation-2: Efficient graph analysis

Dynamic In-Memory Graph Storage Data Structures

Common Graph Data-structures

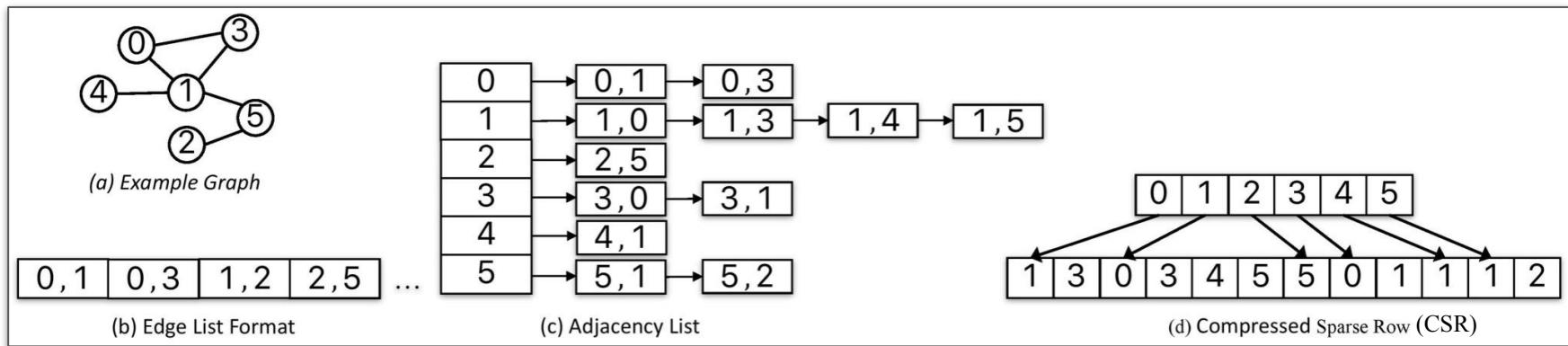
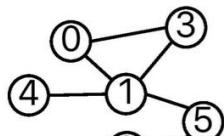


Figure: Three basic graph storage data structures.

Common Graph Data-structures

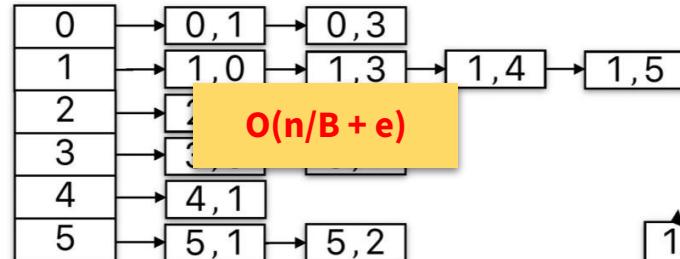
Storage cost / scanning whole graph



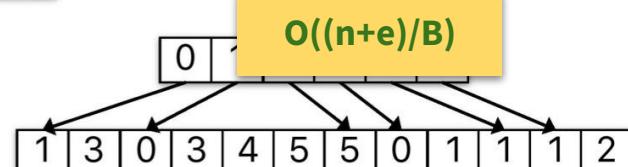
$$O(2^e/B)$$

0, 1	0, 3	1, 2	2, 5	...
------	------	------	------	-----

(b) Edge List Format



(c) Adjacency List

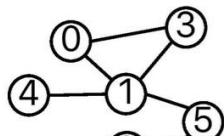


(d) Compressed Sparse Row (CSR)

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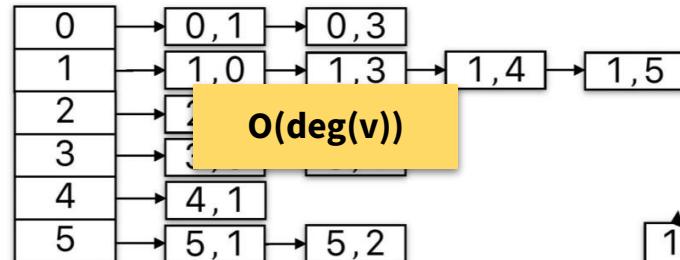
Finding all neighbors of a vertex v



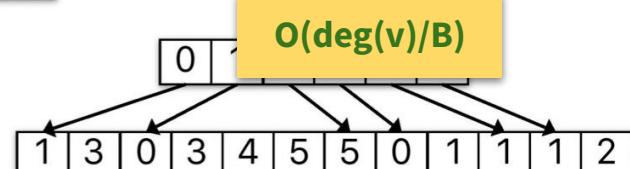
$O(2^*e/B)$

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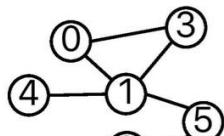


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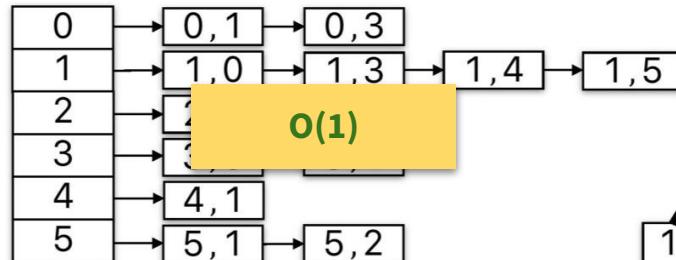
Add edge



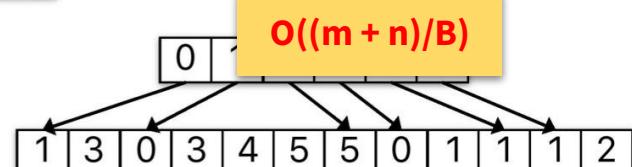
$O(1)$

0, 1	0, 3	1, 2	2, 5	...
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(b) Edge List Format



(c) Adjacency List

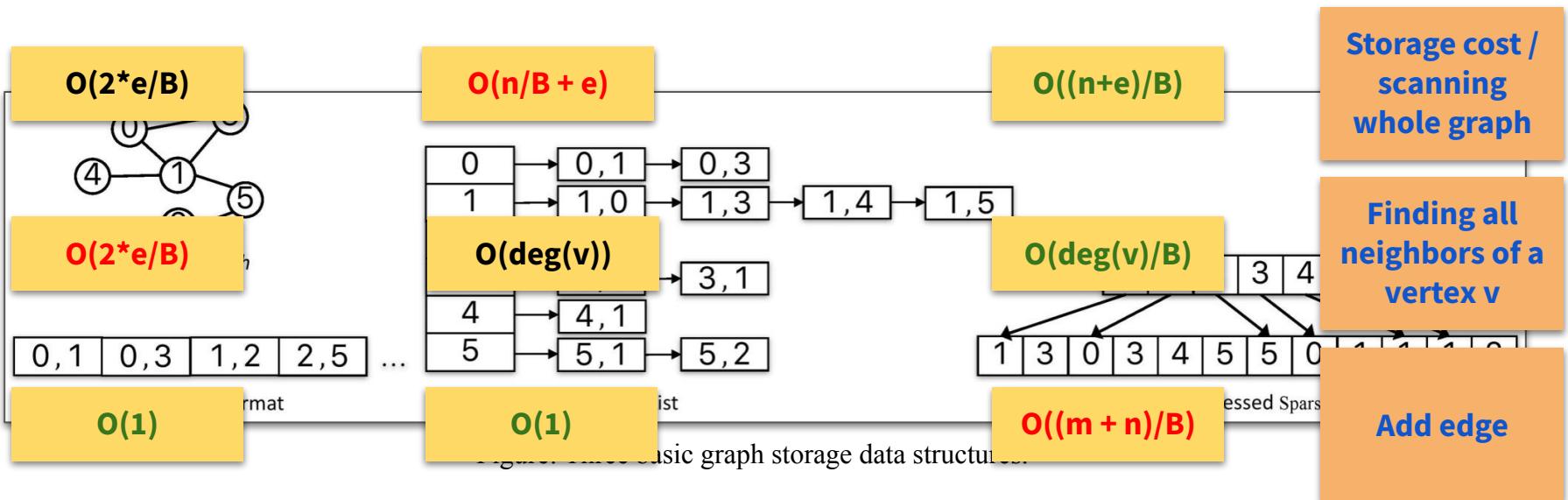


$O((m + n)/B)$

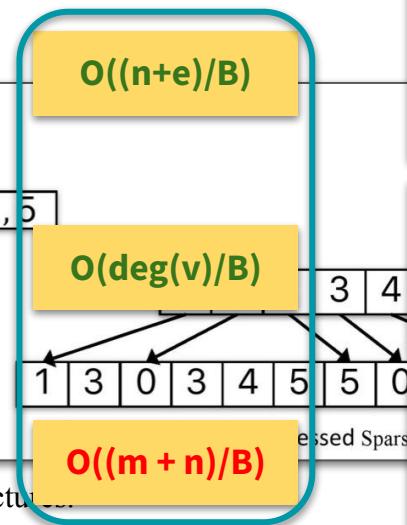
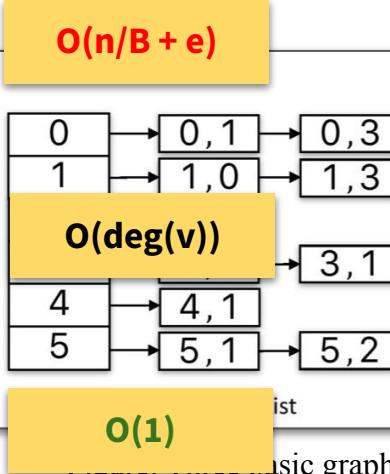
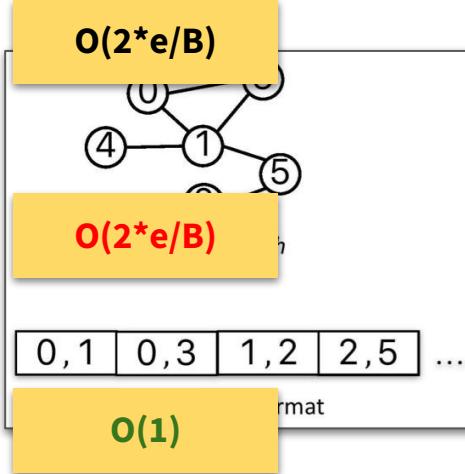
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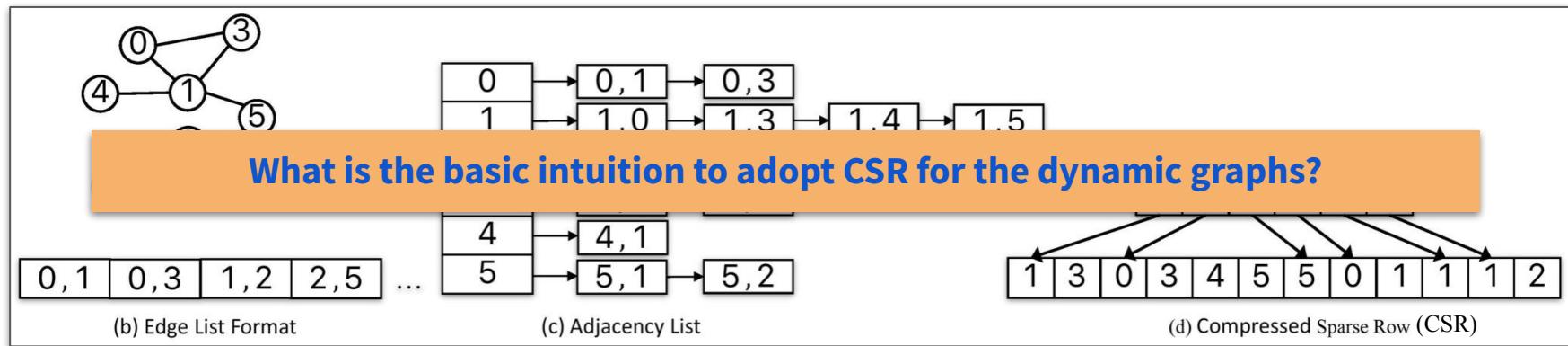


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CSR Variants for Dynamic Graphs

- Delta Map
 - LLAMA [Macko et. al., ICDE'15]
 - Graphchi [Kyrola et. al., OSDI'12]
- Log-Based
 - Grace [Vijayan et. al., ATC'12]
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- Dynamic Block Linked Lists
 - PowerGraph [Gonzalez et. al., OSDI'12]
 - Graphlab [Low et. al., UAI'10]
- Packed Memory Array (PMA) Based Extensions
 - RMA [De Leo et. al., ICDE'19]
 - PCSR [Wheatman et. al., HPEC'18]
 - GPMA [Sha et. al., VLDB'17]
- Other Common Extensions
 - CSR++ [Firmlí et. al., OPODIS'20]
 - Dynamic-CSR [King et. al., Springer'16]
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Packed Memory Array (PMA)

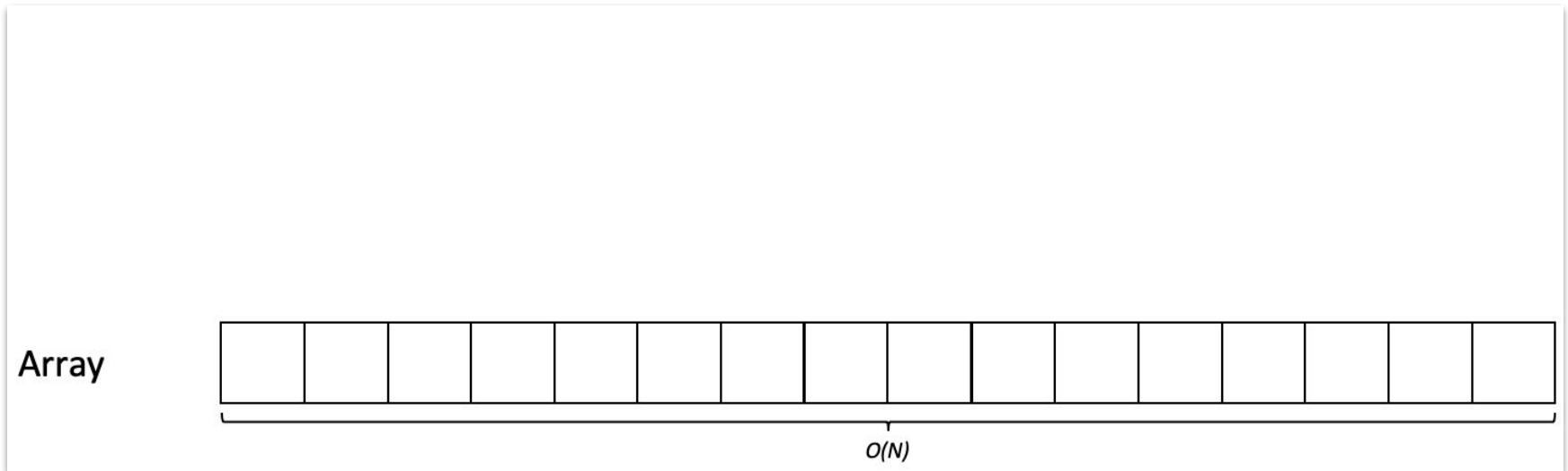


Figure: PMA data structure and one insertion example.

Packed Memory Array (PMA)

- Proposed by **Itai et. al.** in “A sparse table implementation of priority queues” [ICALP'1981]

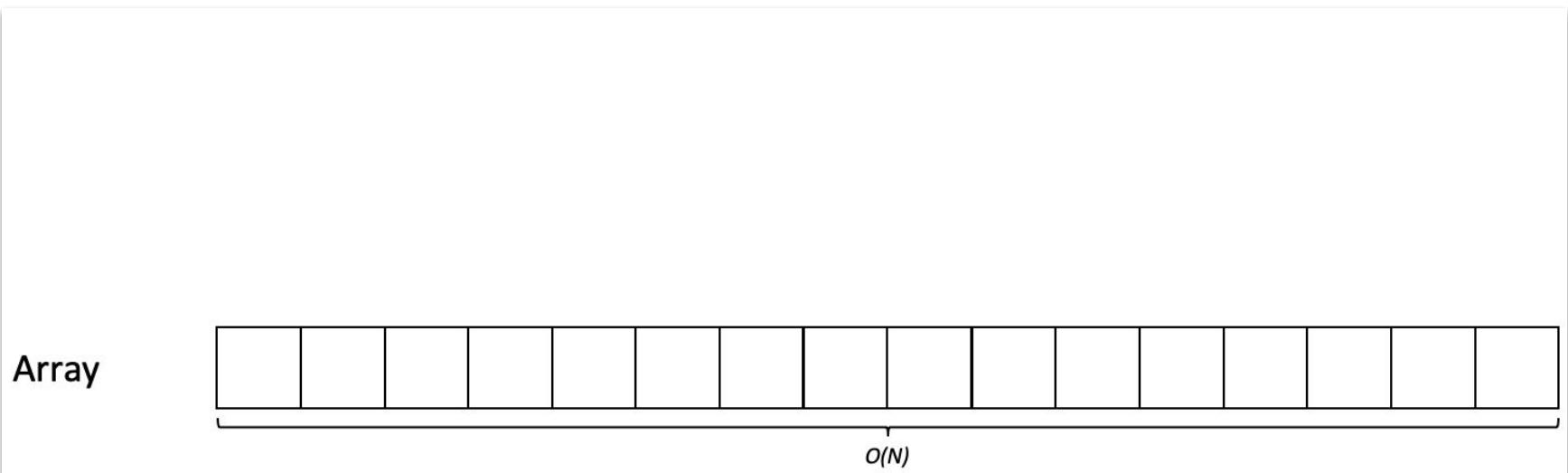


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Array

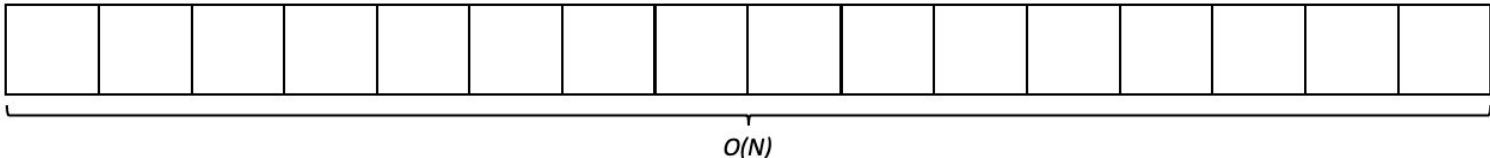


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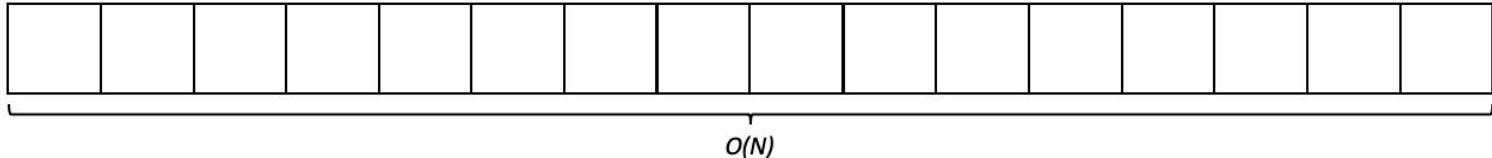


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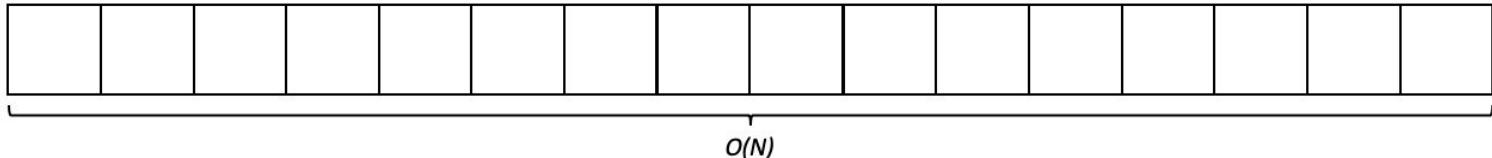


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- Insert N items
- The array will be left sorted after each insertion
- Asymptotic computational complexity of each insertion: $O(\log^2 n)$

Array

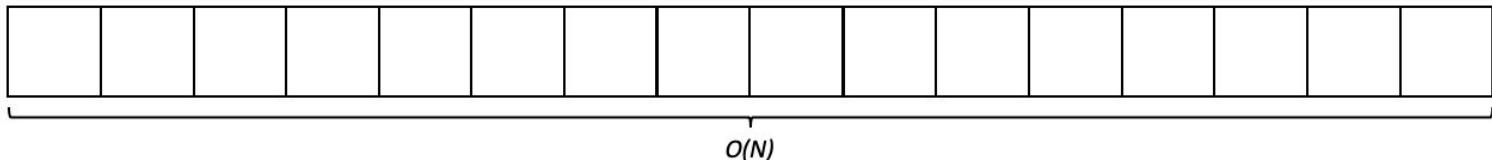


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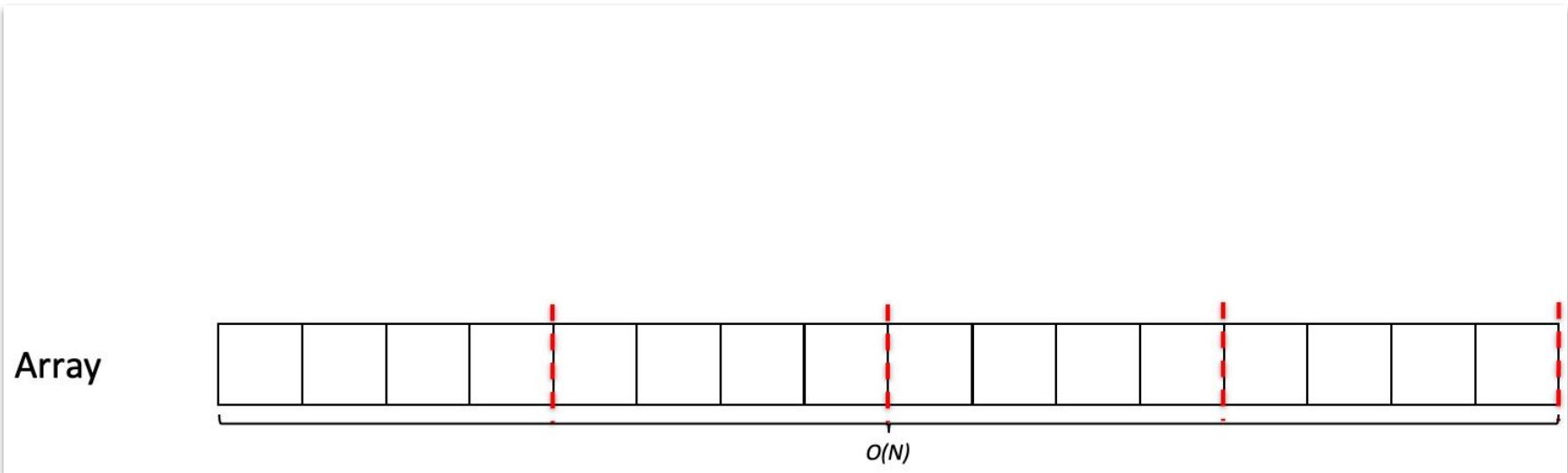


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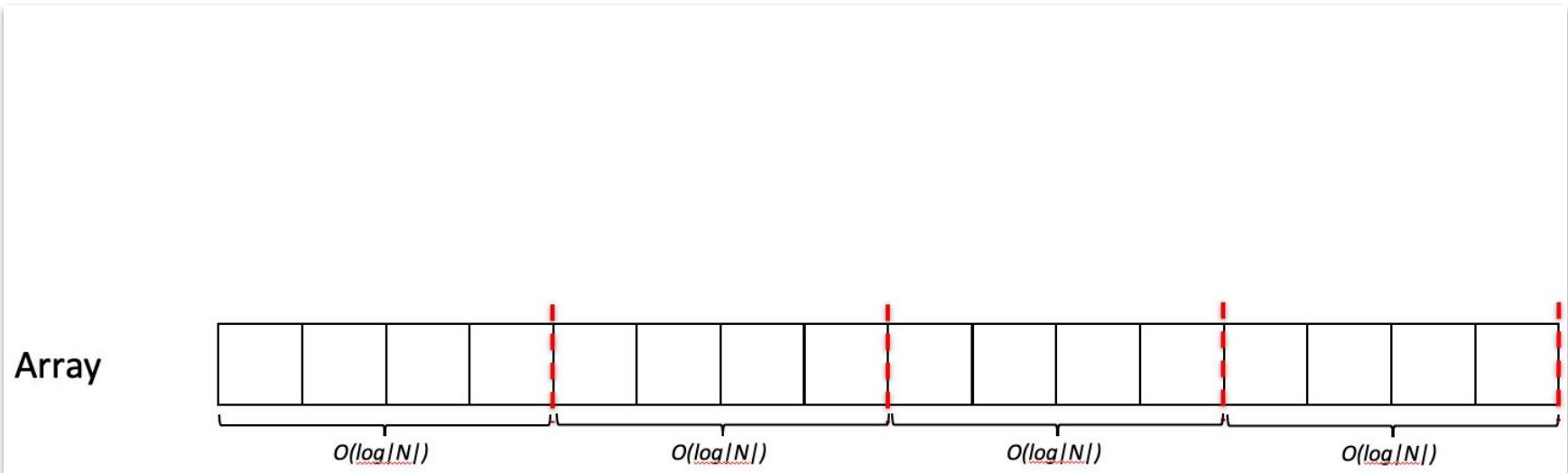


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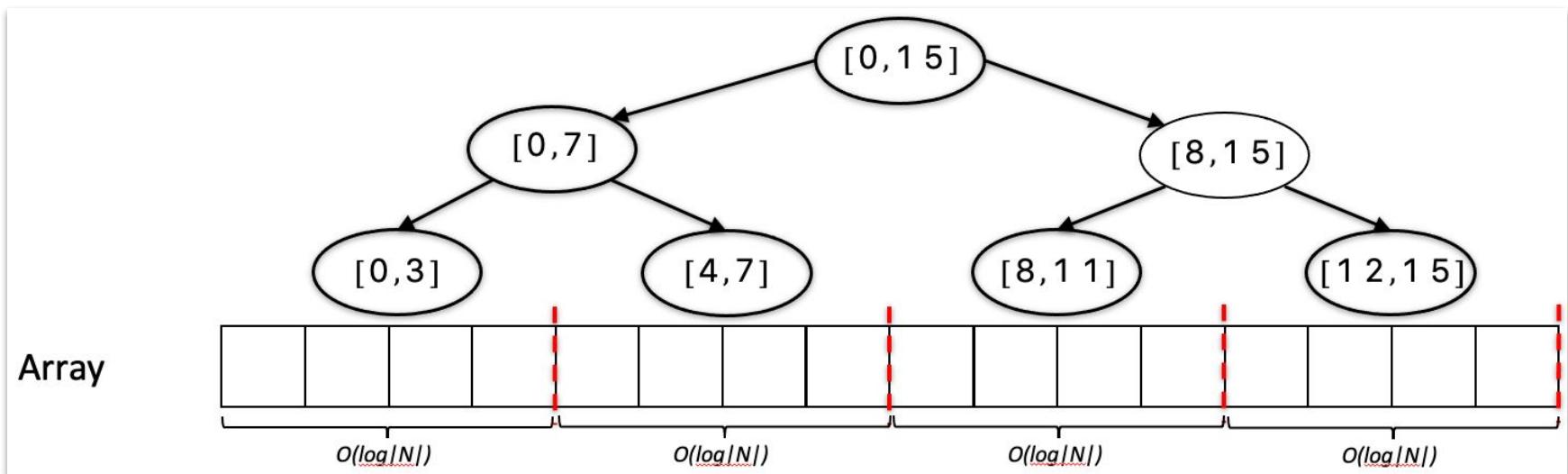


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Packed Memory Array (PMA)

Level 2: [0.5, 0.8]

Level 1: [0.375, 0.85]

Leaf: [0.25, 0.9]

PMA

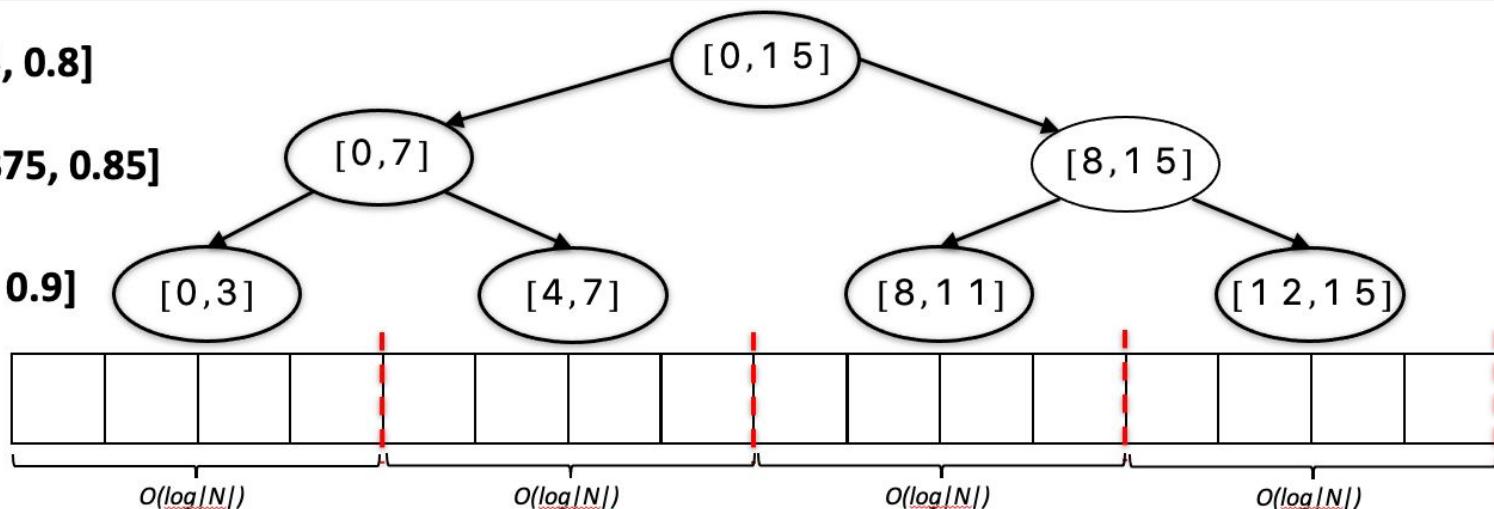


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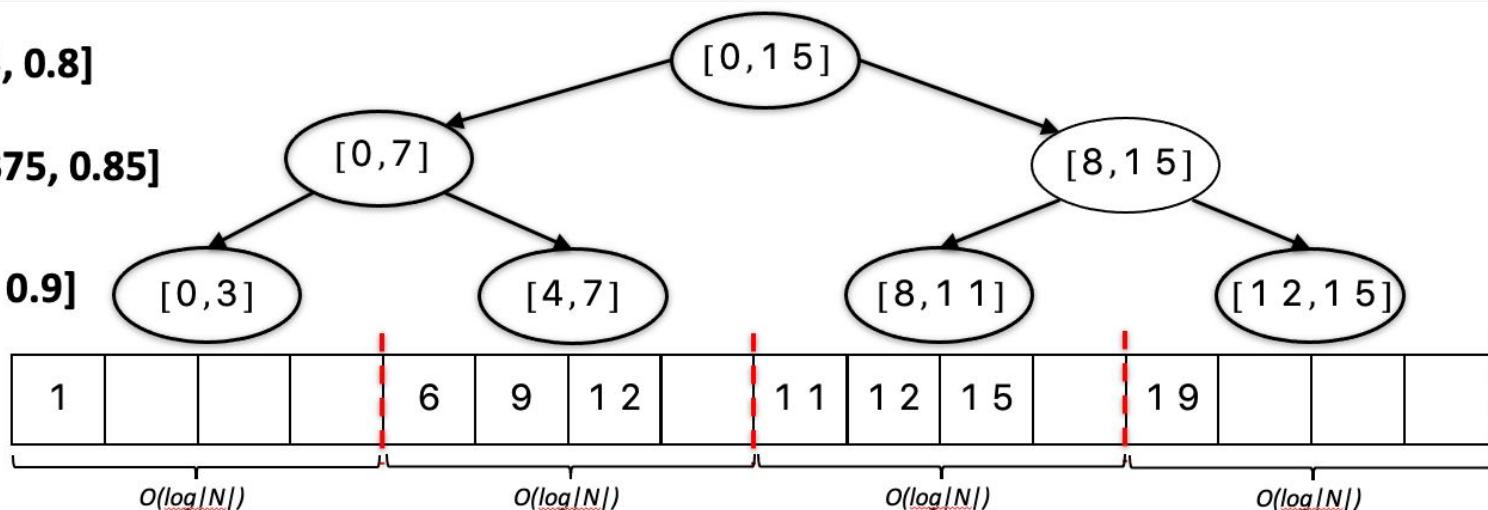


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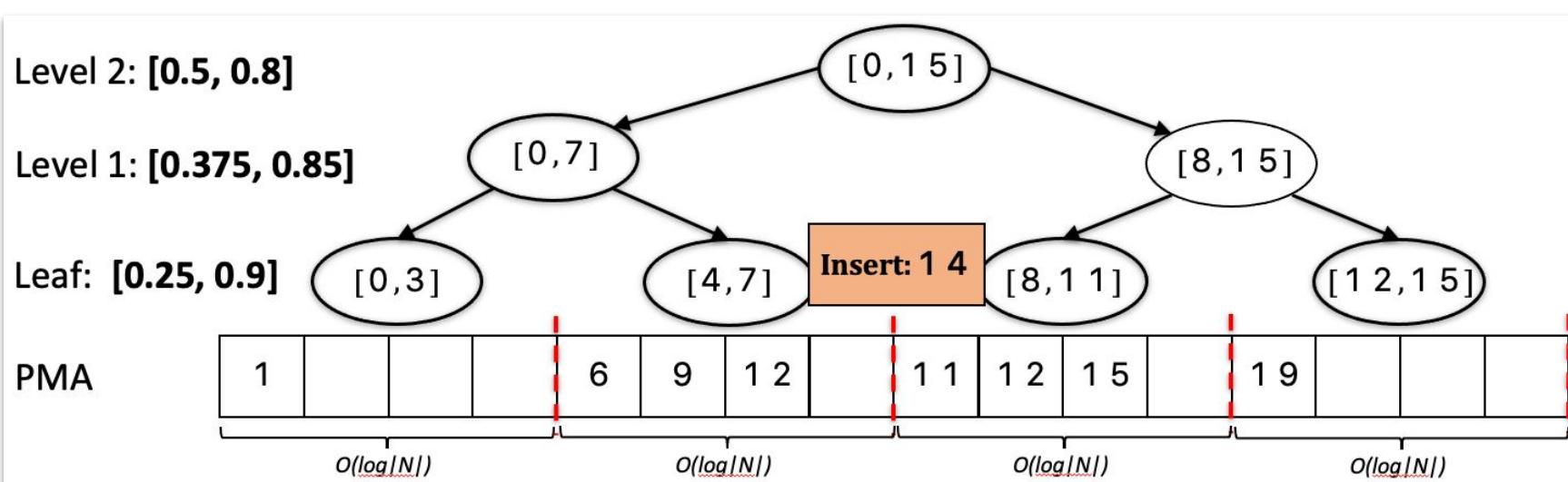


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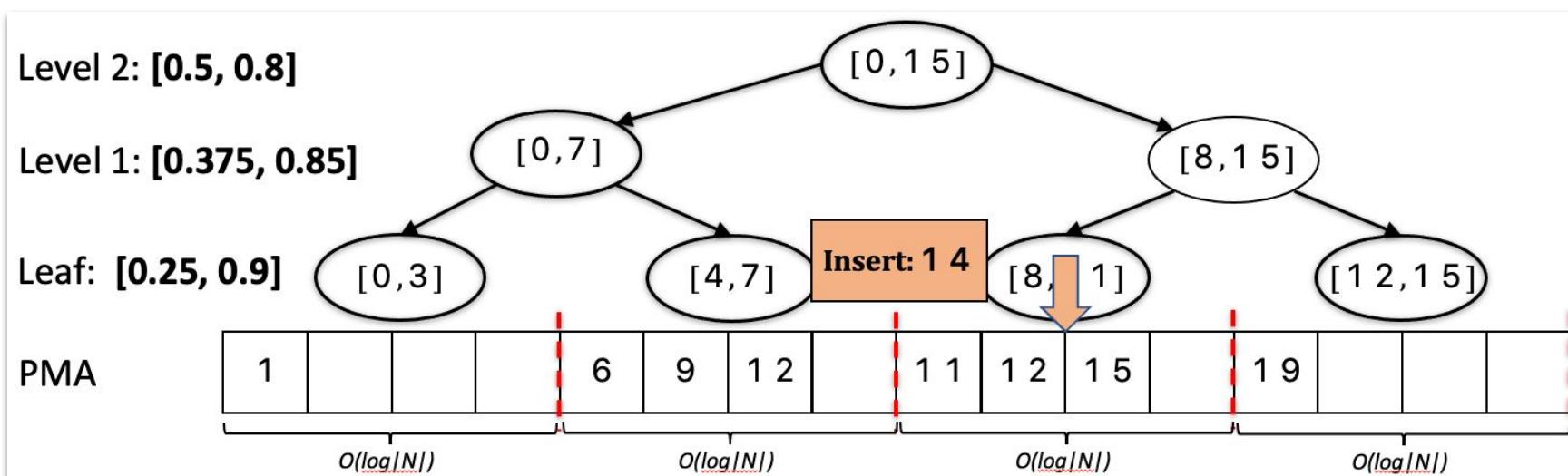


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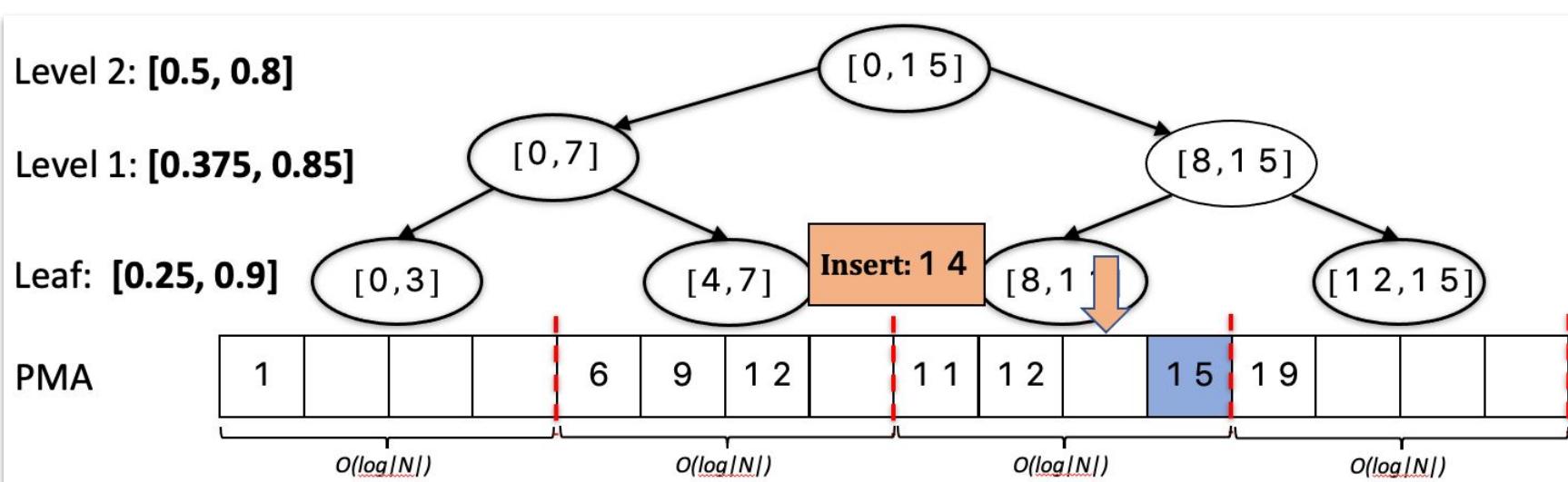


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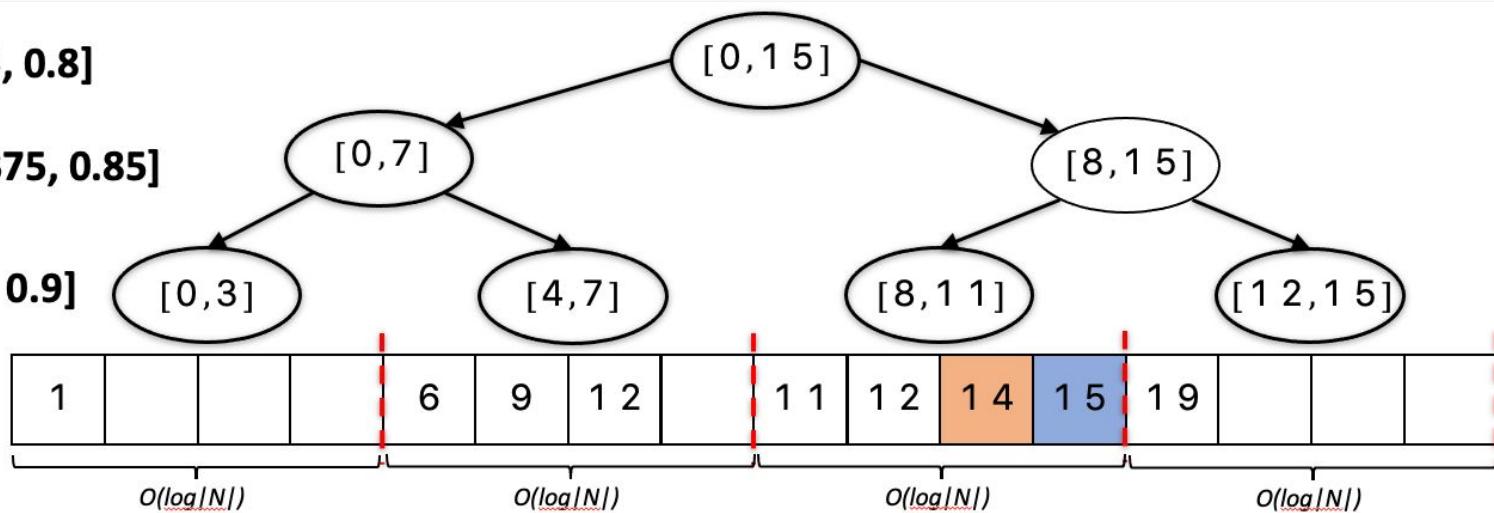


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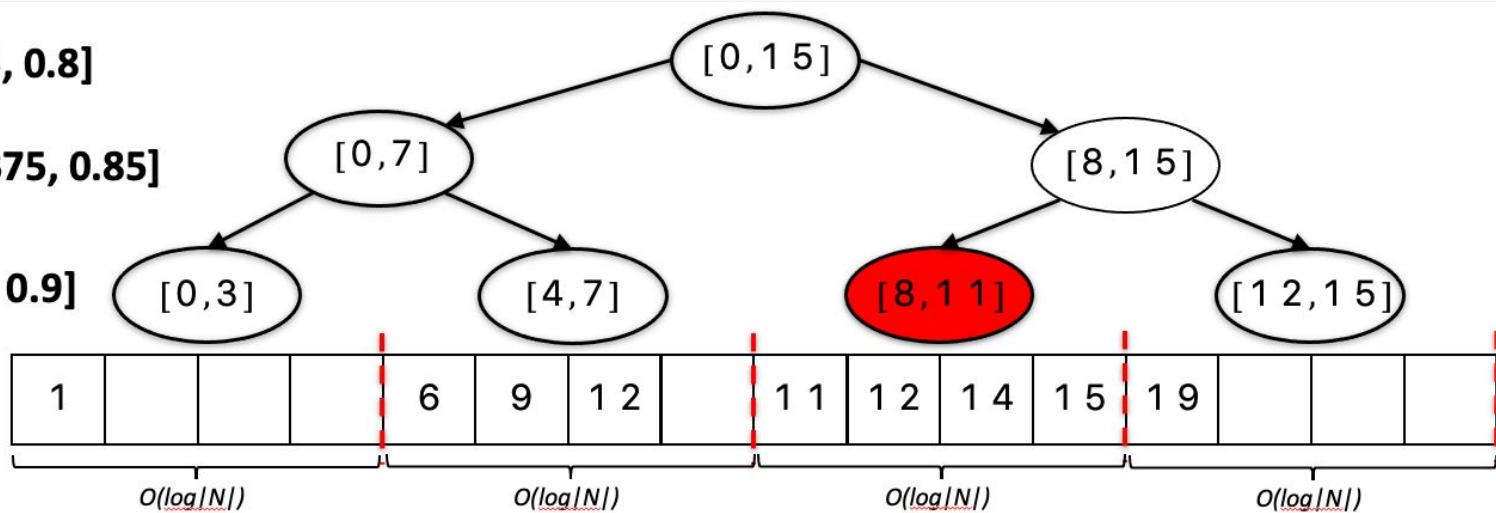


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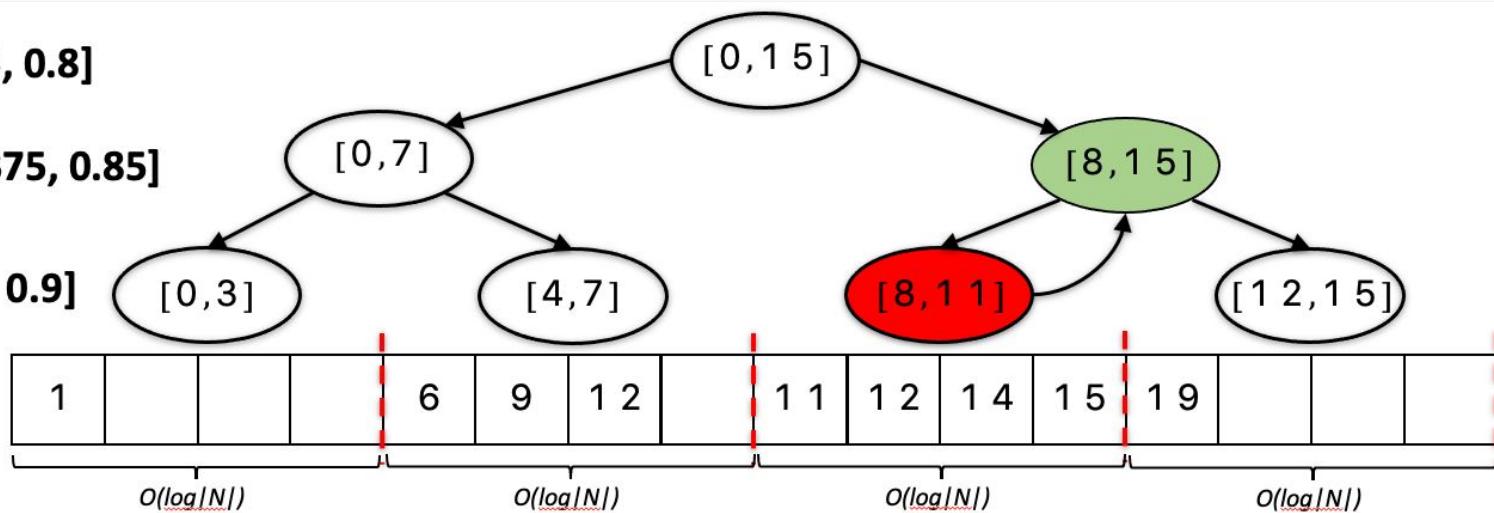


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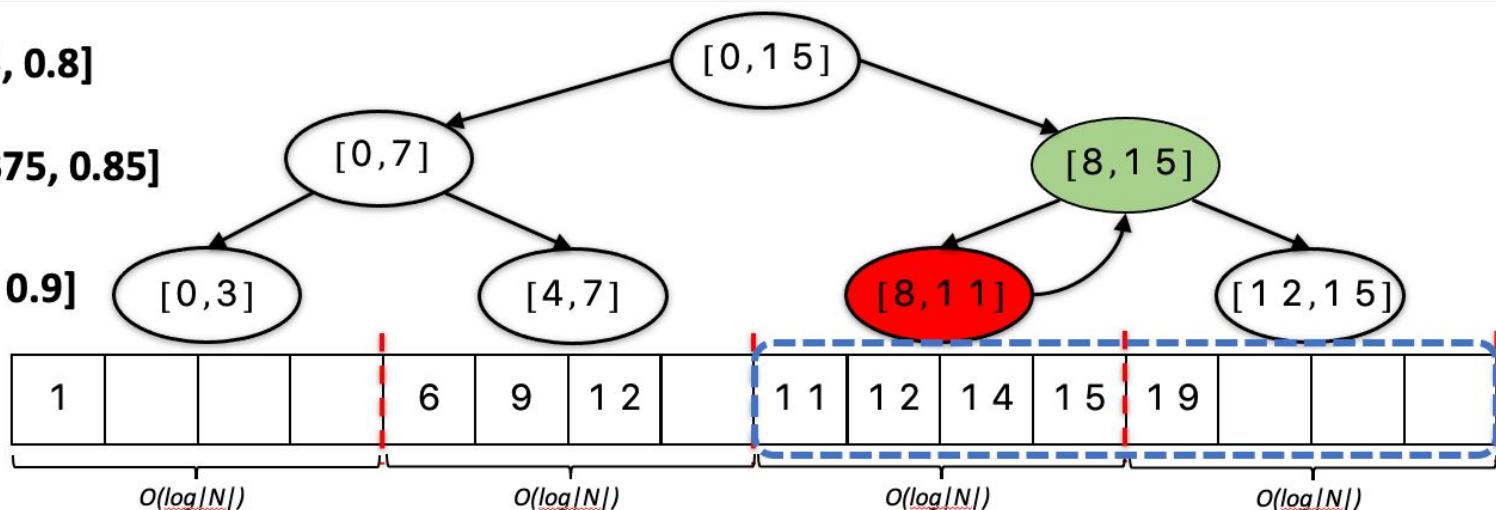


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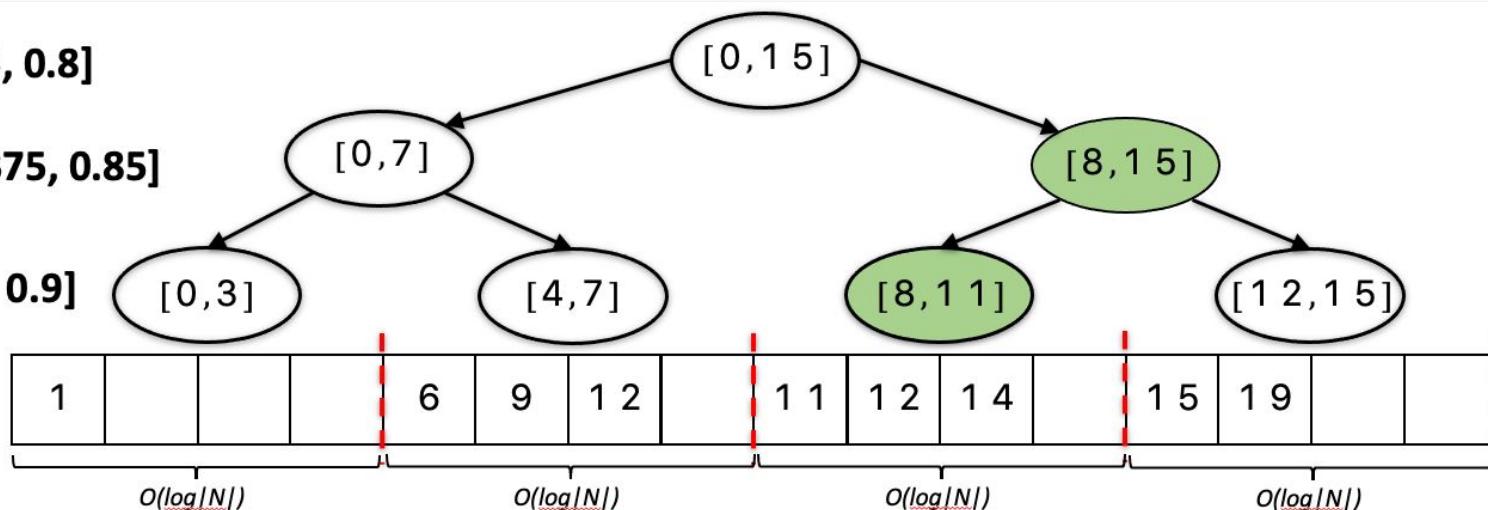


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PMA-Based CSR Extension: PCSR

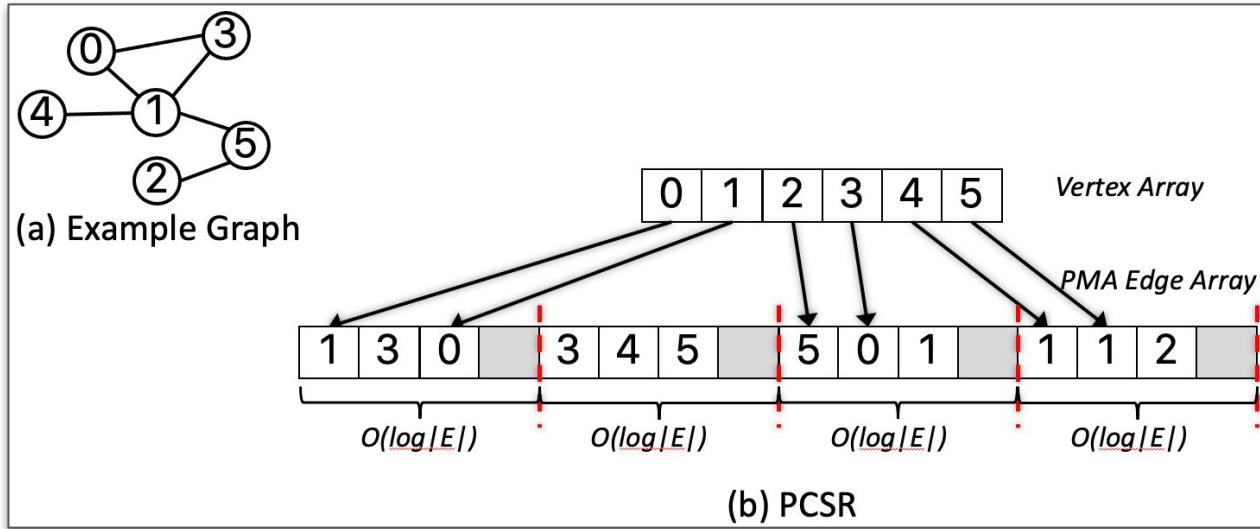


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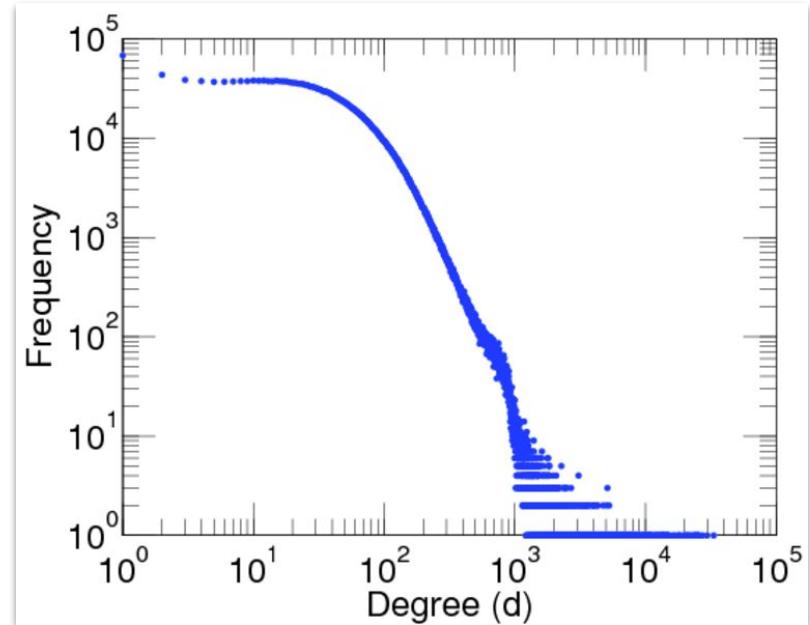
Graph Skewness

Name: Orkut Graph

Category: Online social network

Number of nodes: 3,072,441

Number of edges: 117,184,899



Graph's Mutation Pattern

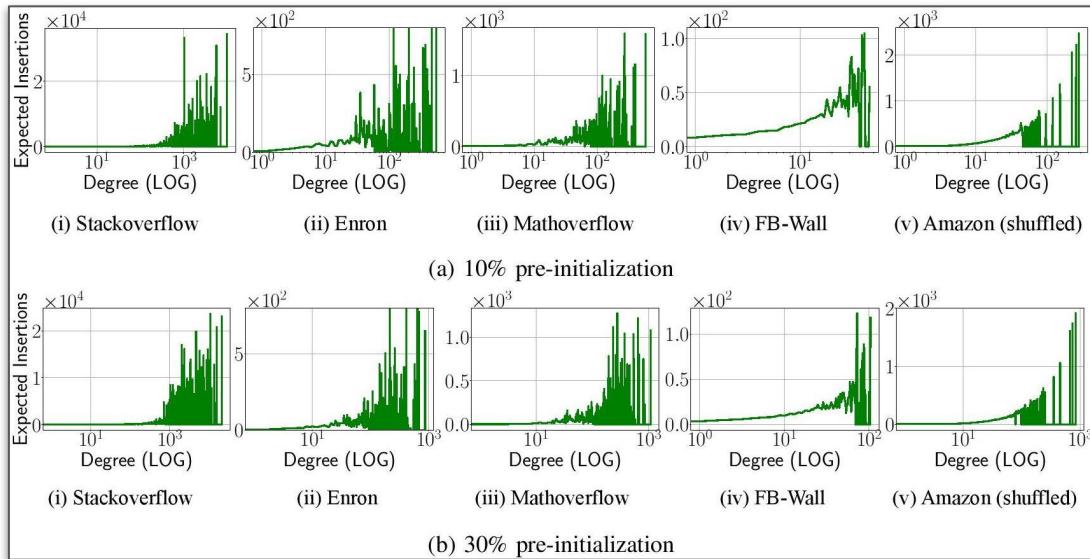


Figure: The relationship between vertex' base degree in both the 10% and 30% pre-initialization cases and their expected new edge insertions. x-axis is the degree; y-axis is the expected insertions; both are in log-scale.

PCSR: Edge Insertion

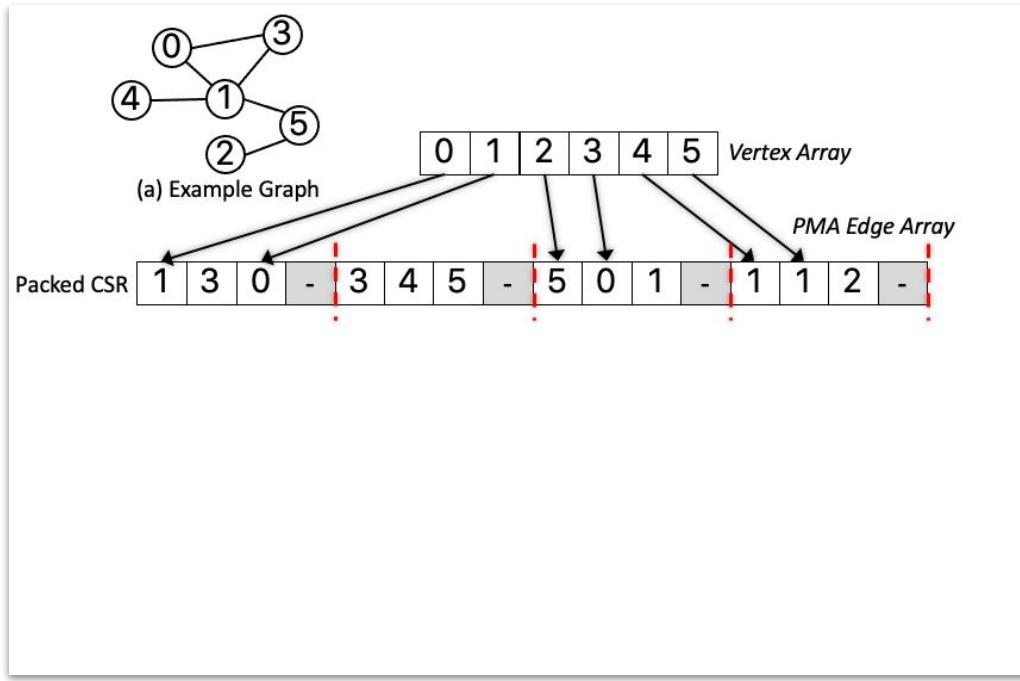


Figure: PMA-based CSR extension, PCSR.

Wheatman, B., & Xu, H. (2018). Packed Compressed Sparse Row: A Dynamic Graph Representation. 2018 IEEE High Performance Extreme Computing Conference, HPEC 2018.

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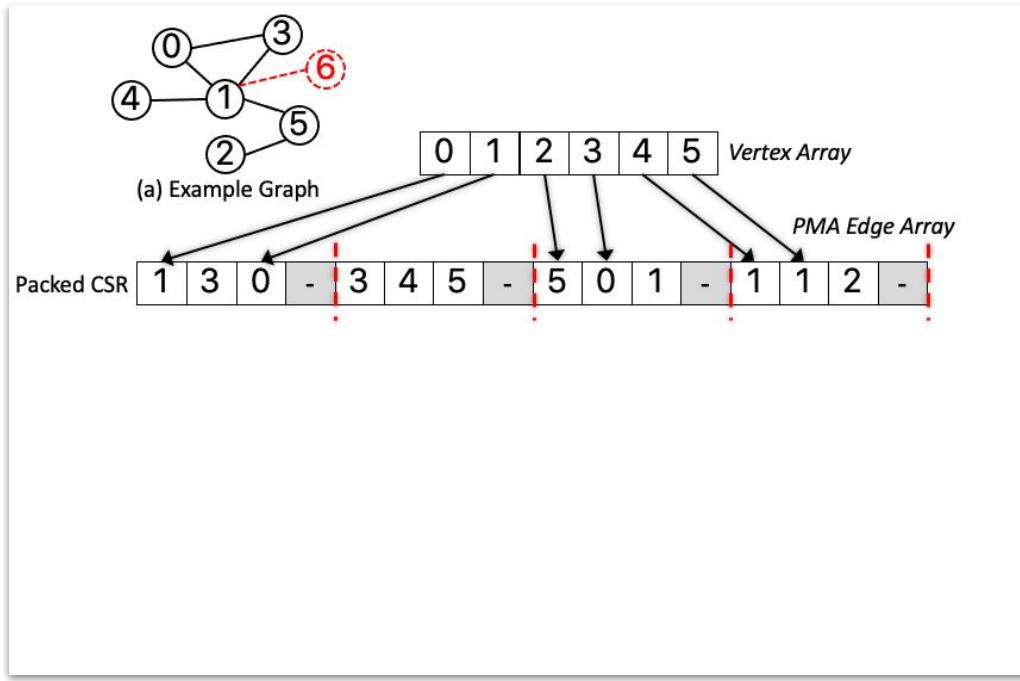


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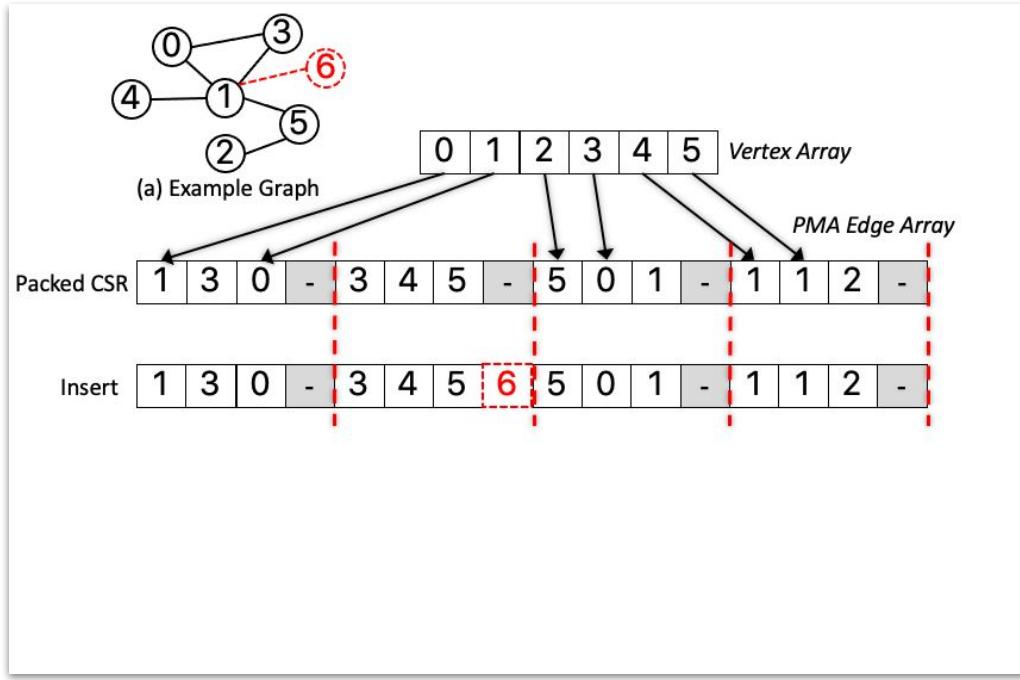


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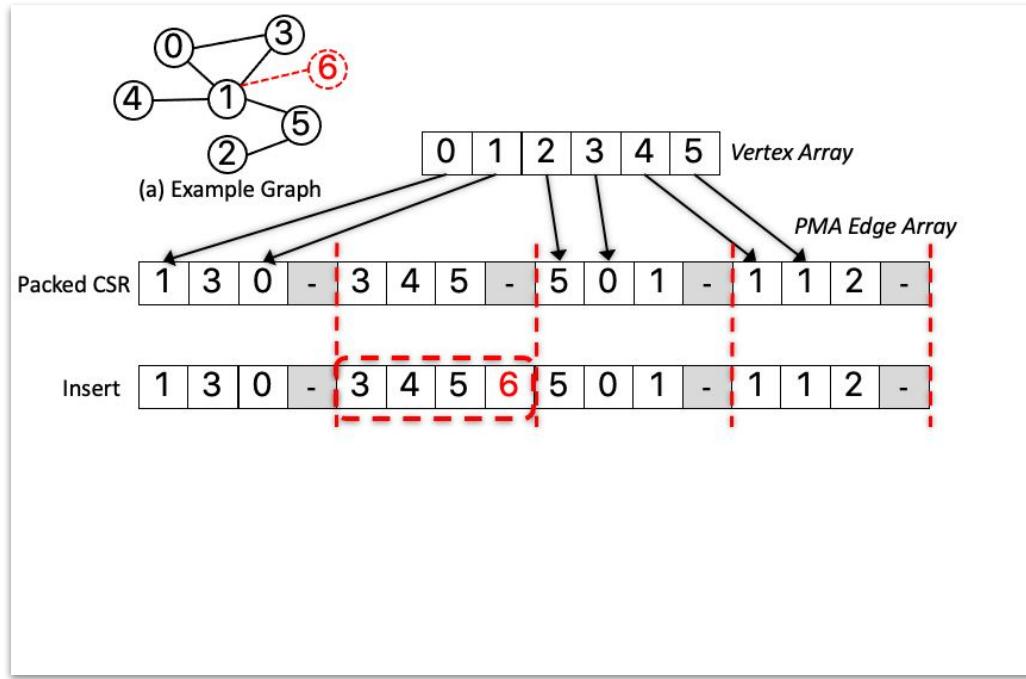


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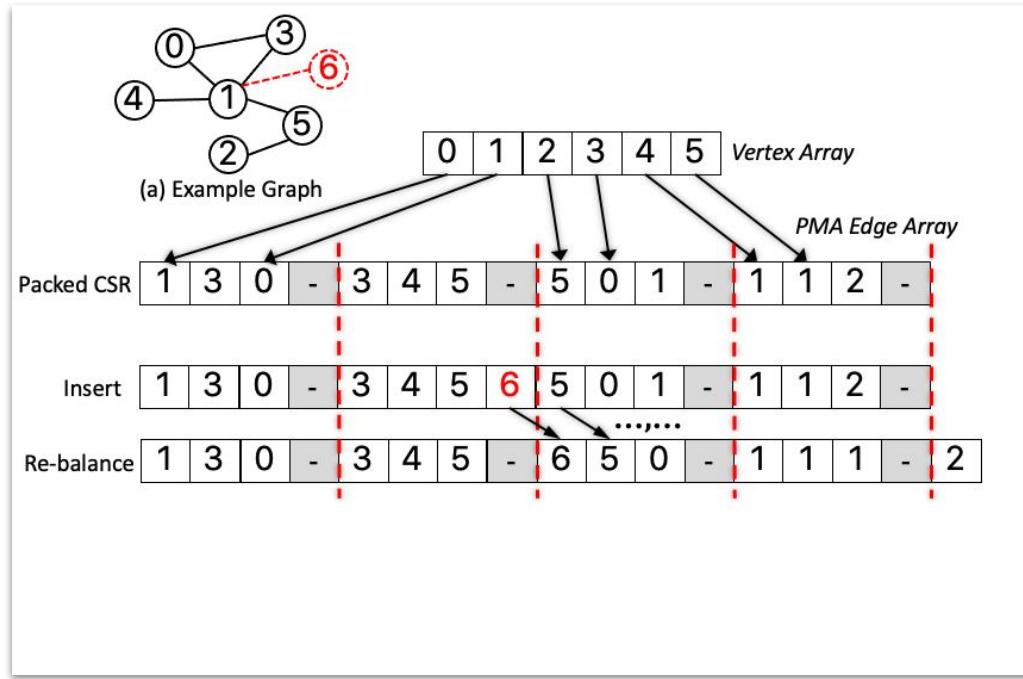


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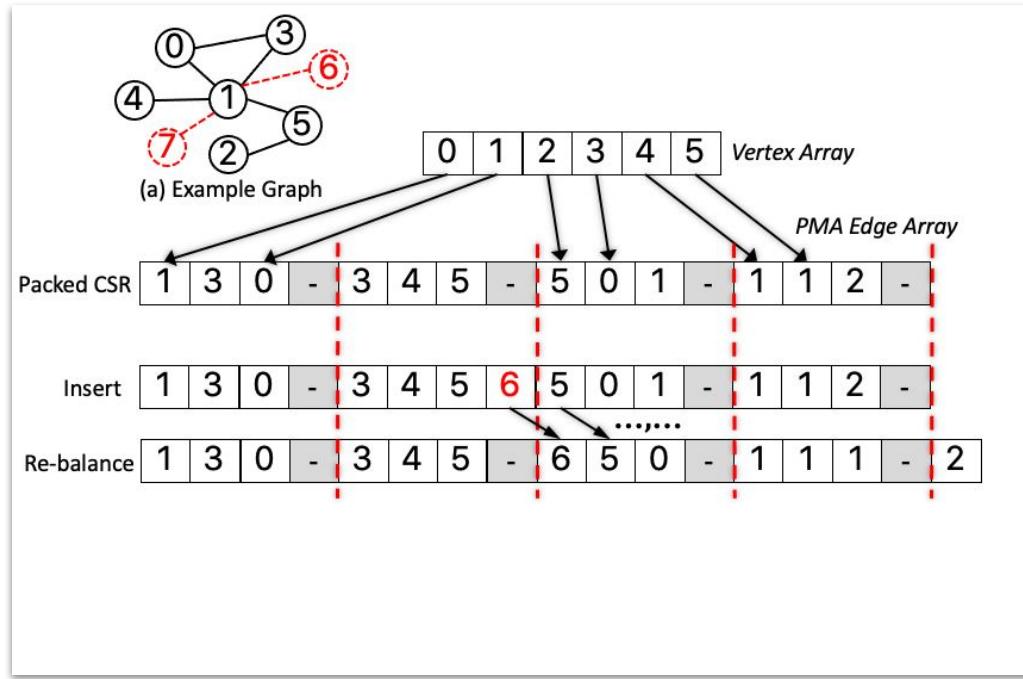


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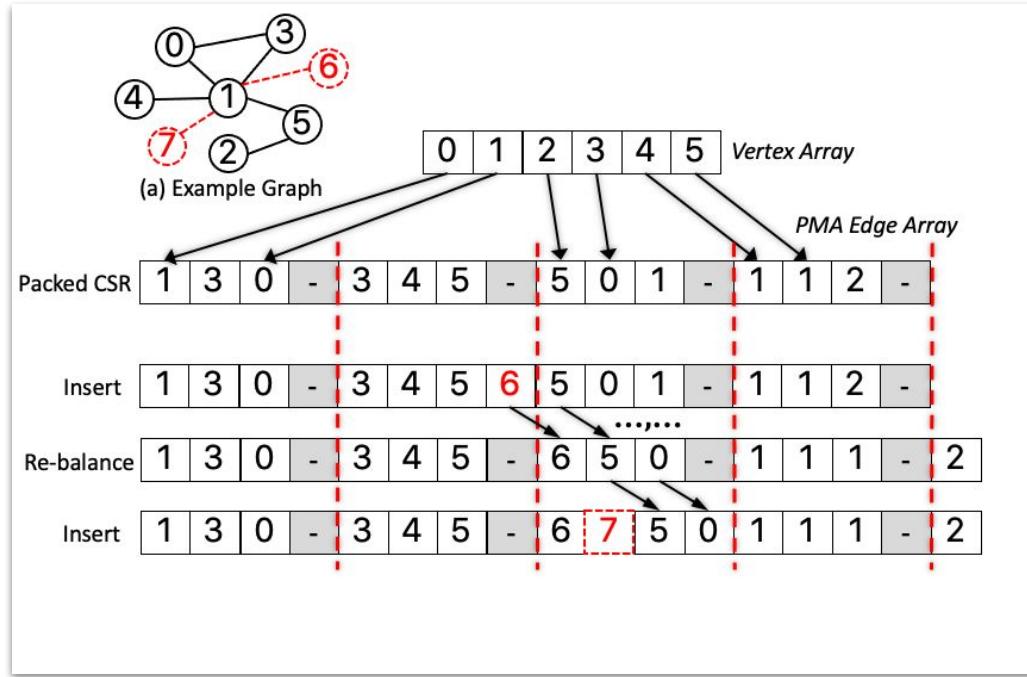


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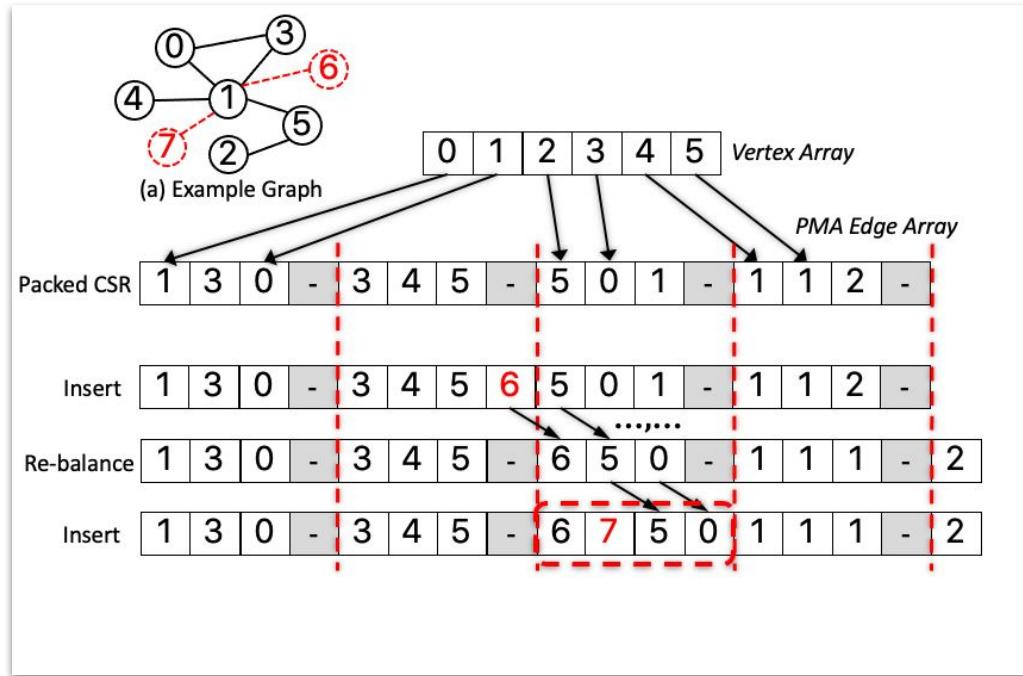


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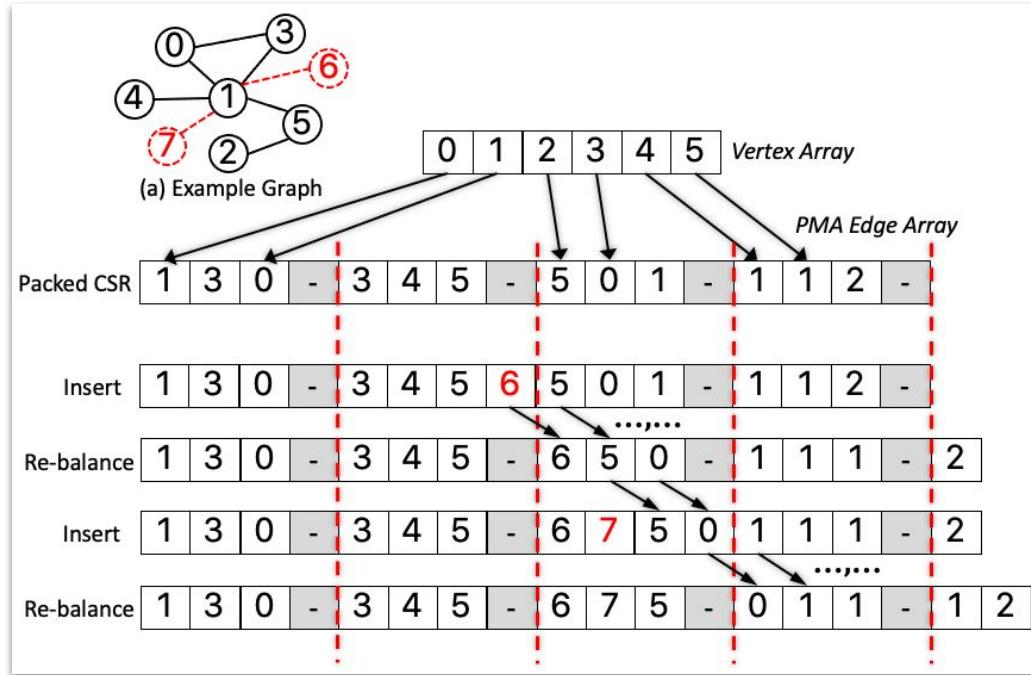


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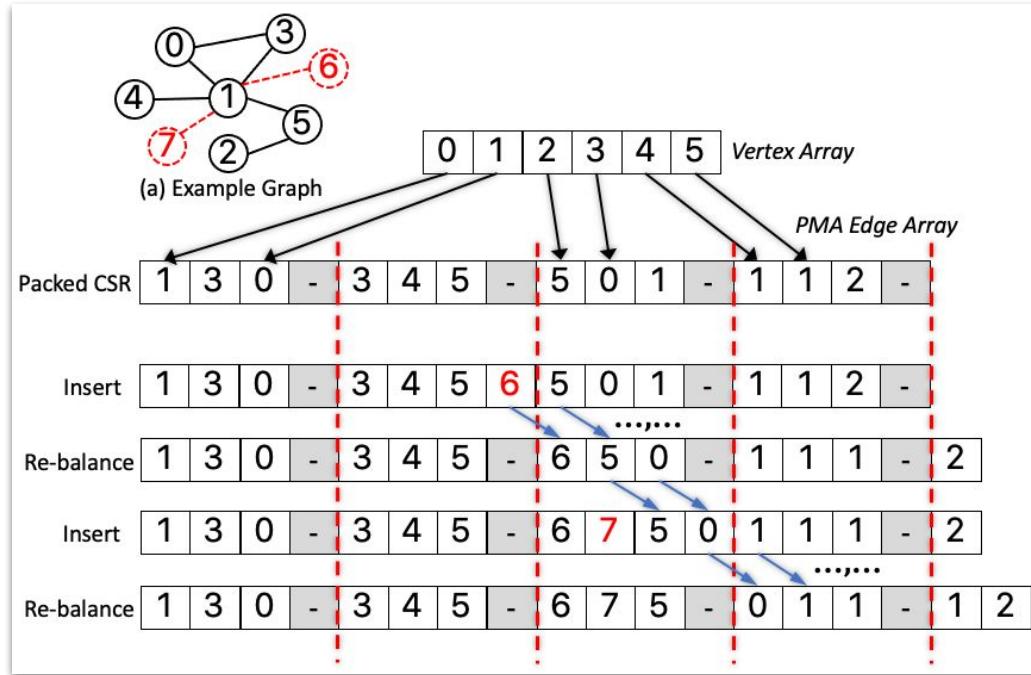


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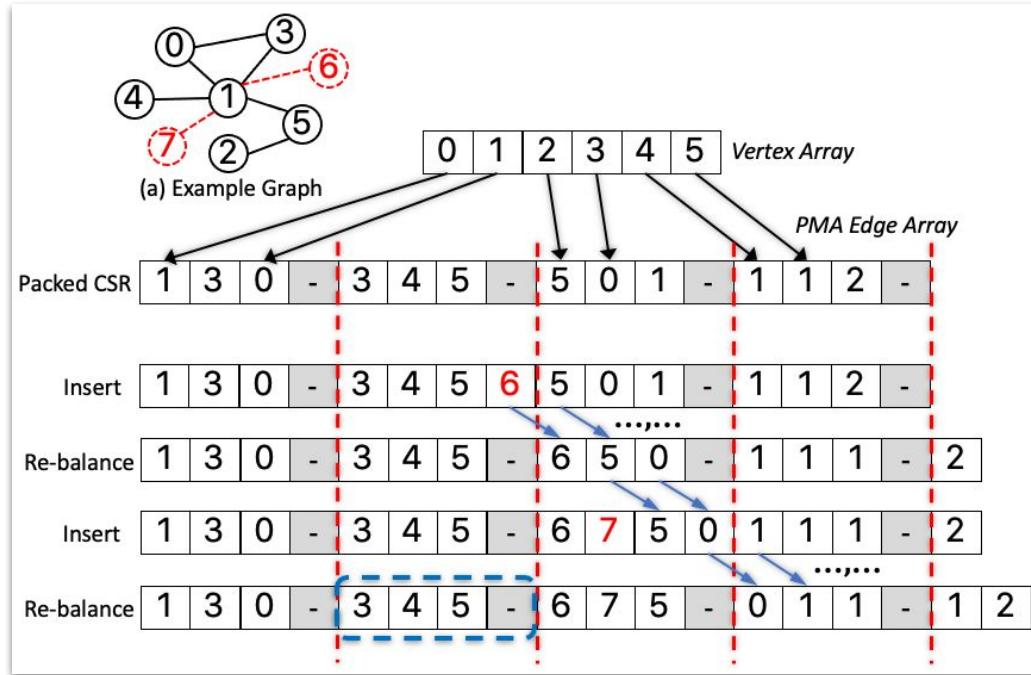
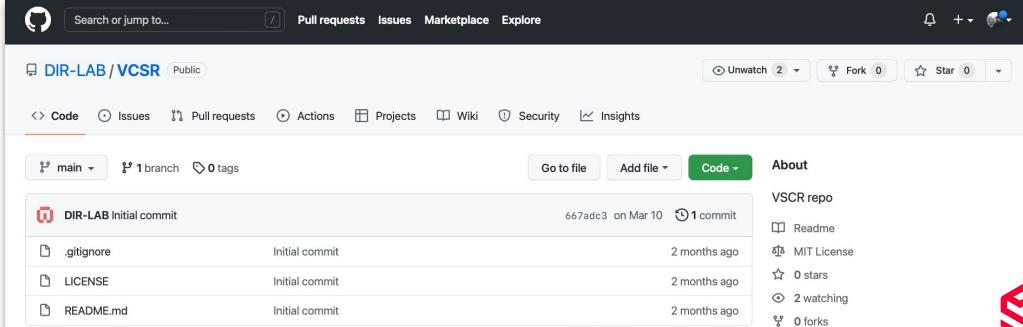


Figure: PMA-based CSR extension, PCSR.

Our Contributions

- Design a novel vertex-centric CSR extension, VCSR, to solve the fundamental limitations in handling graph imbalances
- 1.41x-3.81x better performance in graph insertions
- 1.22x-2.05x better performance in running typical graph analytic algorithms

- We release our code at github



The screenshot shows the GitHub repository page for 'DIR-LAB / VCSR'. The repository is public and has 1 branch and 0 tags. The 'Code' tab is selected. The commit history shows three initial commits:

Commit	Message	Date	Author
667adc3	DIR-LAB Initial commit	on Mar 10	1 commit
.gitignore	Initial commit	2 months ago	
LICENSE	Initial commit	2 months ago	
README.md	Initial commit	2 months ago	

On the right side of the repository page, there is an 'About' section with the following details:

- VCSR repo
- Readme
- MIT License
- 0 stars
- 2 watching
- 0 forks

<https://github.com/DIR-LAB/VCSR>

Outlines

Novel Mutable CSR Design

Evaluation and Results



VCSR: Design

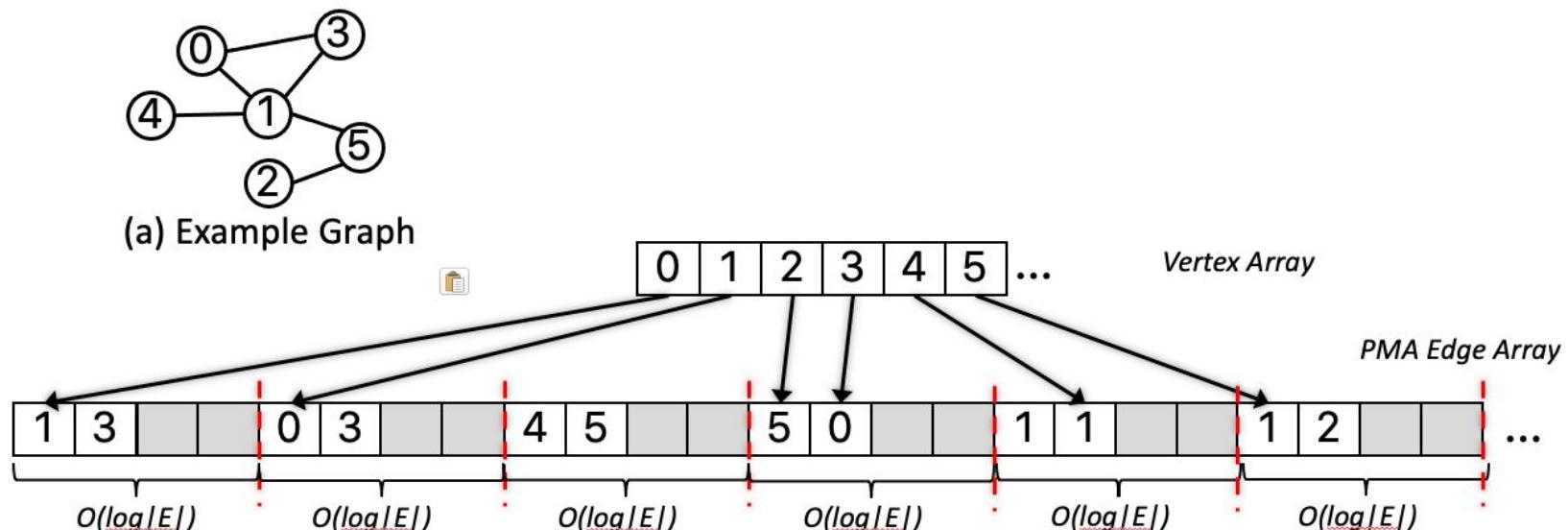


Figure: PMA-based Vertex-Centric Mutable CSR Extension, VCSR.

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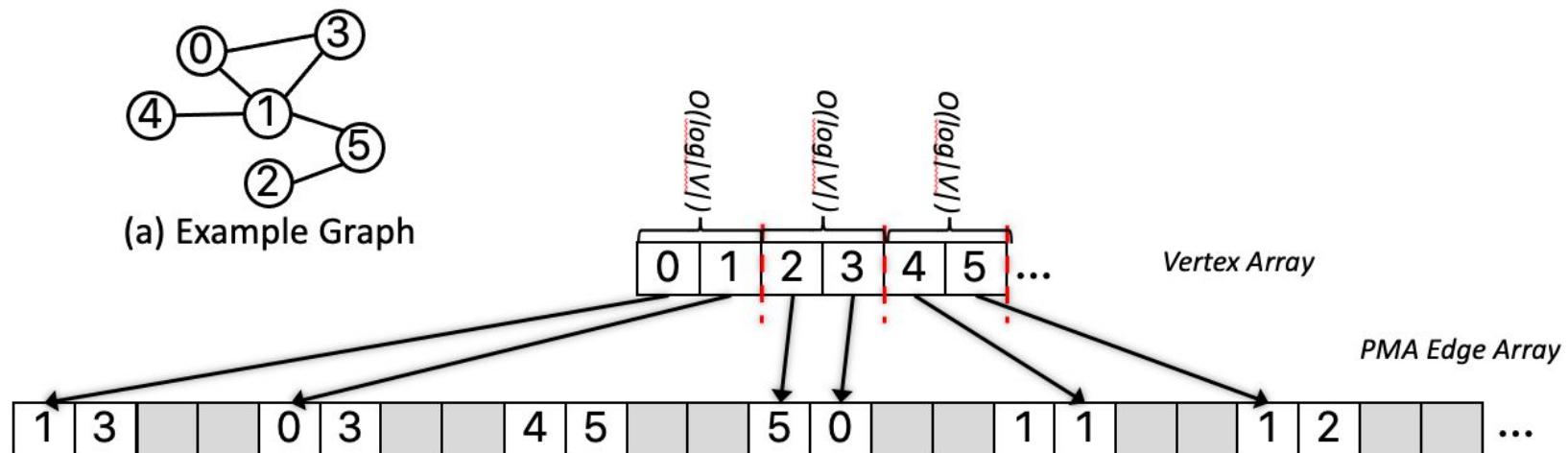


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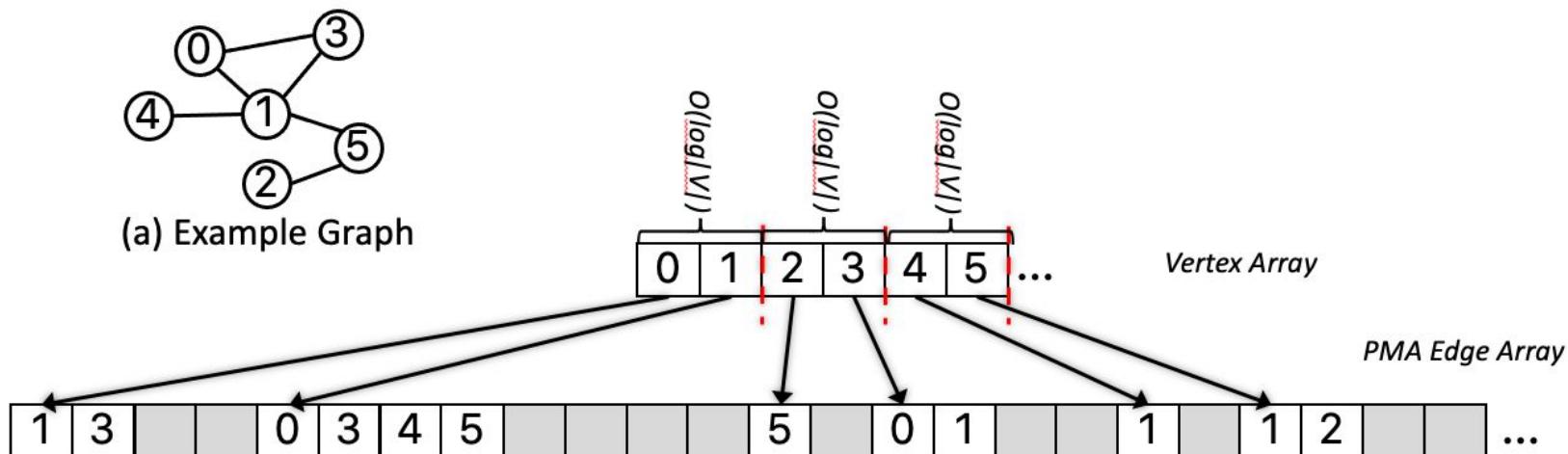


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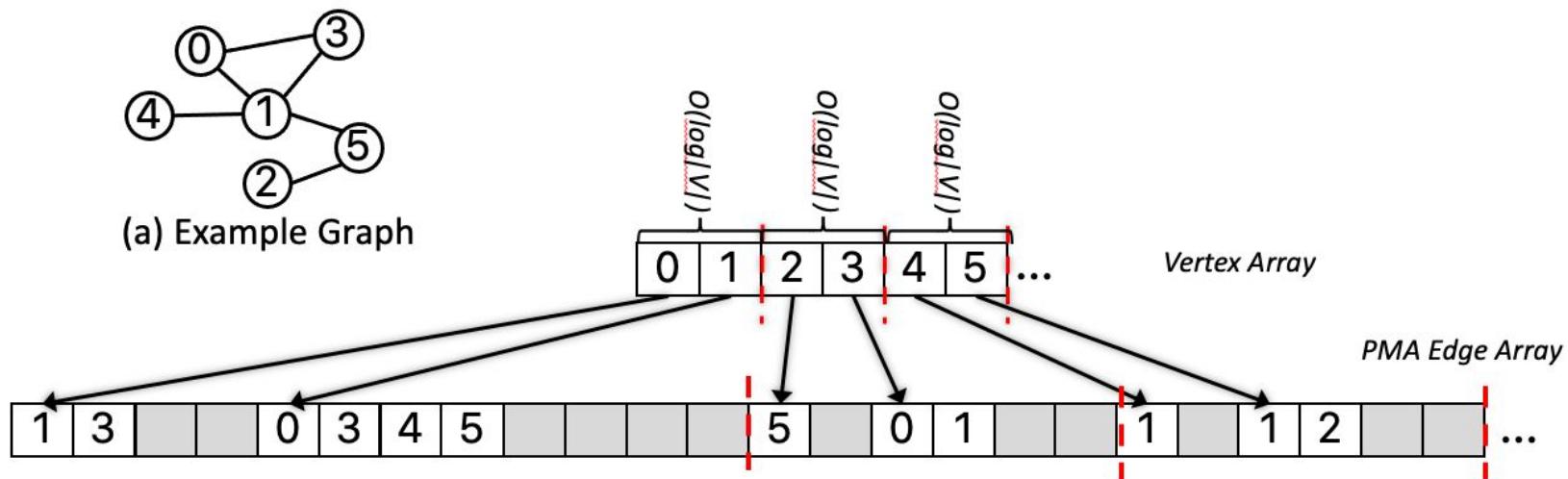


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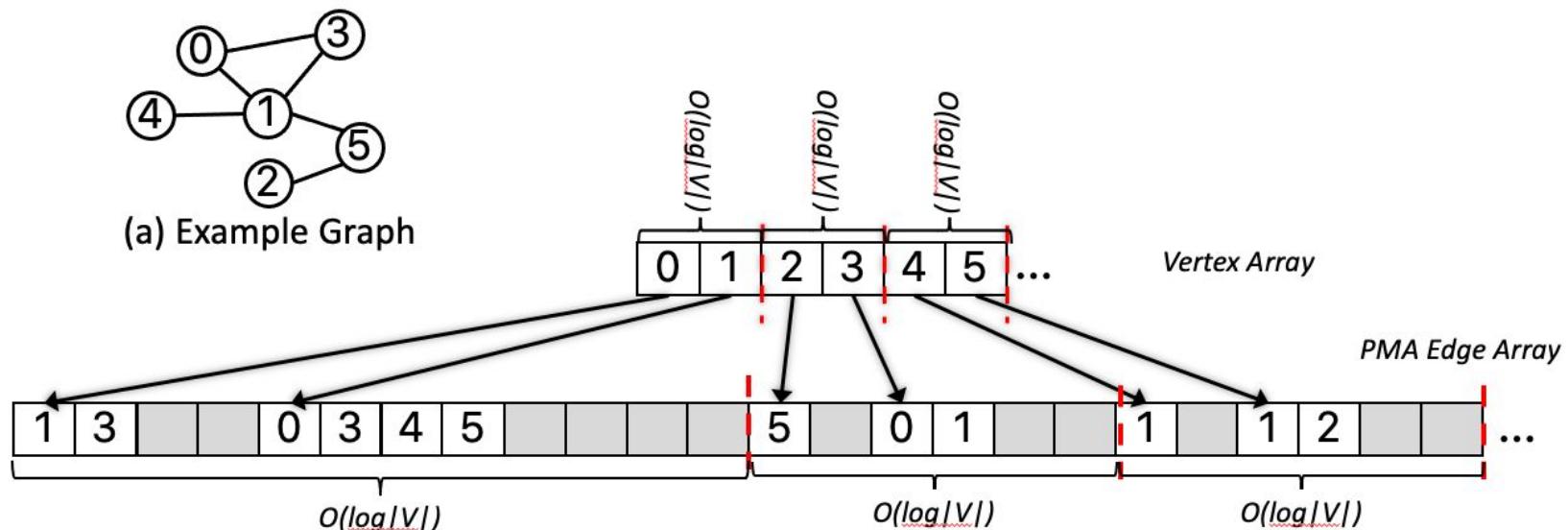


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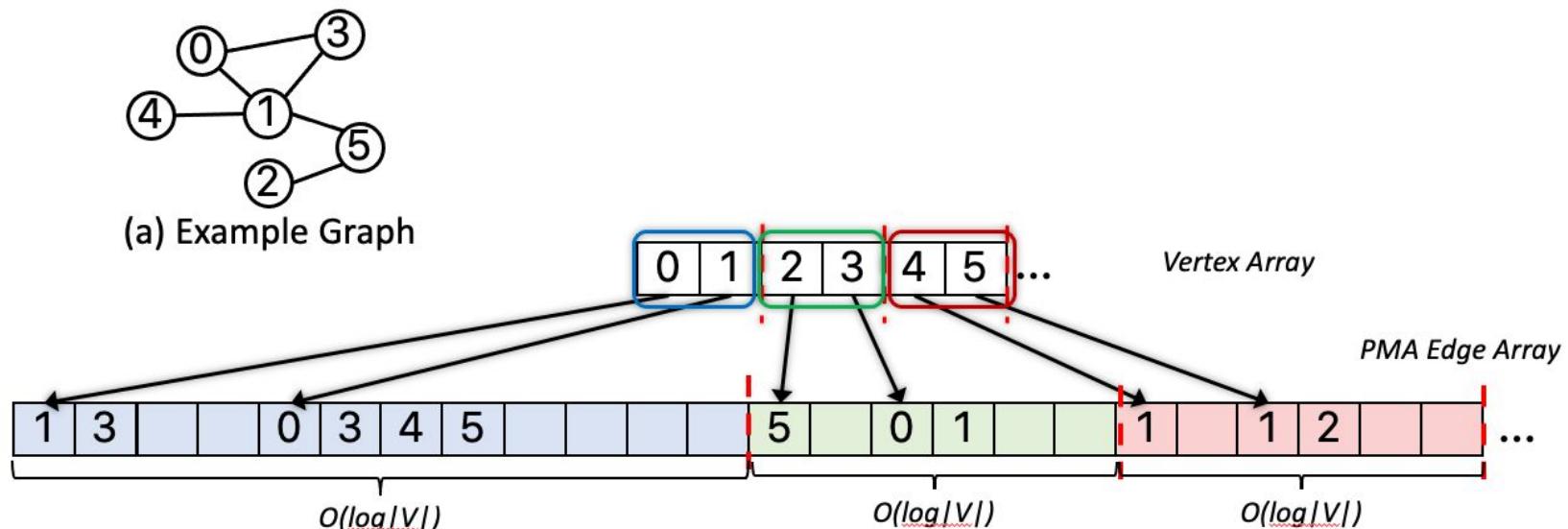


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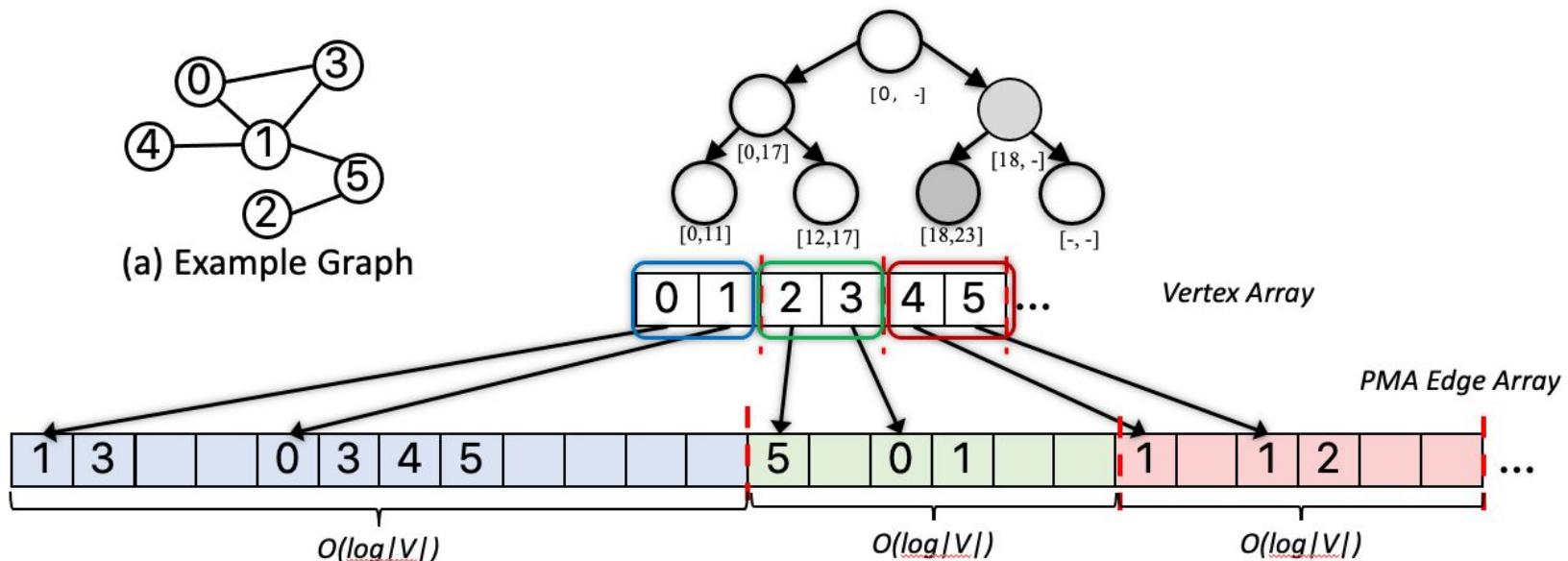


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Evaluation: Platform

- Dell R740 rack server with two sockets
- Each socket installs a 2nd generation Intel Xeon Scalable Processor (Gold 6254 @ 3.10G) with 18 physical (36 virtual) cores
- Ubuntu 18.04, Linux kernel version 4.15.0
- 6 DRAM DIMMS with 32GB each (192GB in total)

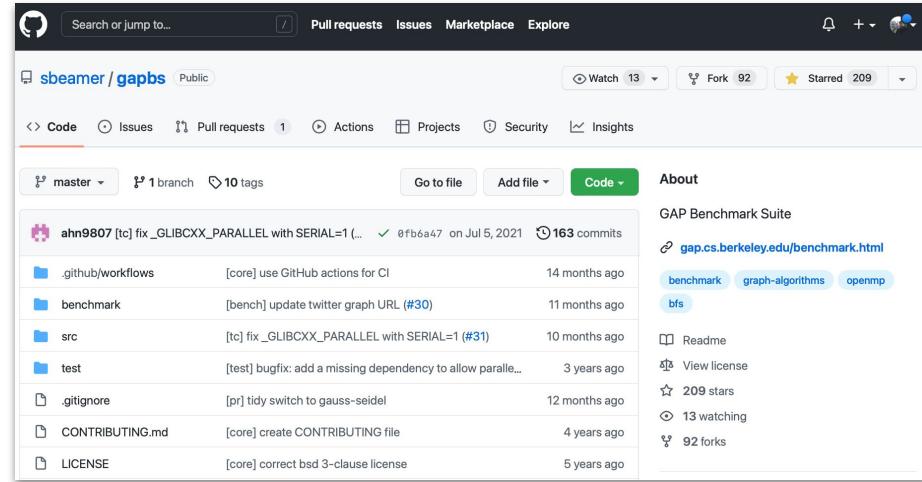
Evaluation: System Implementation

- Compressed Sparse Row (CSR)
- Blocked Adjacency-List (BAL)
- PMA-based CSR extension, PCSR
- Proposed vertex-centric PMA-based CSR extension, VCSR

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● We used GAPBS
for benchmark

The screenshot shows the GitHub repository page for `sbeamer/gapbs`. The repository is public and has 163 commits. The repository URL is <https://github.com/sbeamer/gapbs>. The repository page shows code, issues, pull requests, actions, projects, security, and insights sections. The code section shows files like `.github/workflows`, `benchmark`, `src`, `test`, `.gitignore`, `CONTRIBUTING.md`, and `LICENSE`. The repository has 13 watching, 92 forks, and 209 stars.

<https://github.com/sbeamer/gapbs>

Evaluation: Input Graphs

Datasets	Domain	$ V $	$ E $	$ E / V $
Amazon	purchase	403393	4886816	12
Orkut	social	3072626	234370166	76
Live-journal	social	4847570	85702474	18
Cit-Patents	citation	6009554	33037894	6
Road	geo	1971280	5533214	3
as-Skitter	network	1696414	22190596	13
sx-stackoverflow	temporal	6024270	57724802	10
enron	temporal	87273	594912	7
sx-mathoverflow	temporal	88580	375972	4
fb-wall	temporal	63891	366824	6

Table: Graph inputs and their key properties.

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Random Workload

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Evaluation: Graph Algorithm Kernels

Graph kernel	Kernel Type	Input	Output	Notes
PageRank (PR)	Link Analysis	-	$ V $ -sized array of ranks	Fixed number (20) of iterations
Breadth-First Search (BFS)	Graph Traversal	Source vertex	$ V $ -sized array of parent IDs	Direction-Optimizing approach [27]
Single-Source Shortest Paths (SSSP)	Shortest Path	Source vertex	$ V $ -sized array of distances	δ -stepping [28]
Connected Components (CC)	Connectivity	-	$ V $ -sized array of component labels	Afforest subgraph sampling [29, 30]

Table: A list of graph kernels and inputs and outputs used to evaluate graph data-structures.

Evaluation: Graph Insertion Performance

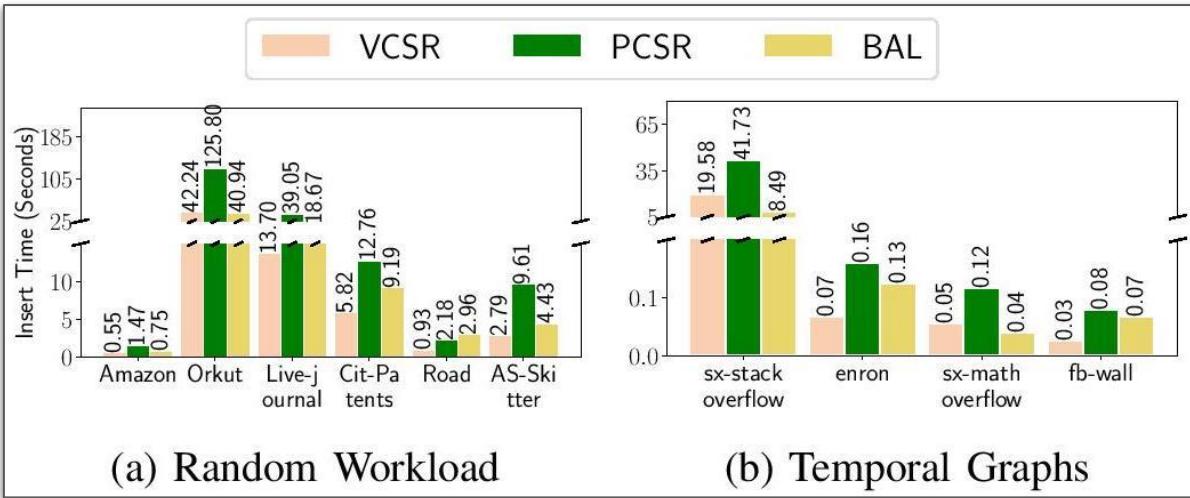


Figure: Comparing VCSR's dynamic graph insertion performance (in seconds) for 10% pre-initialization.

Evaluation: Graph Insertion Performance

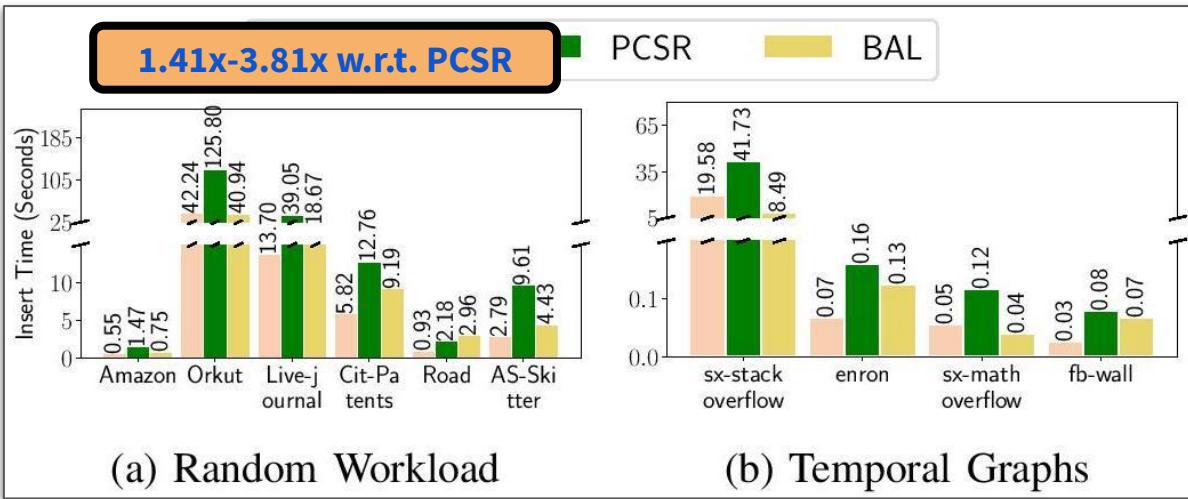


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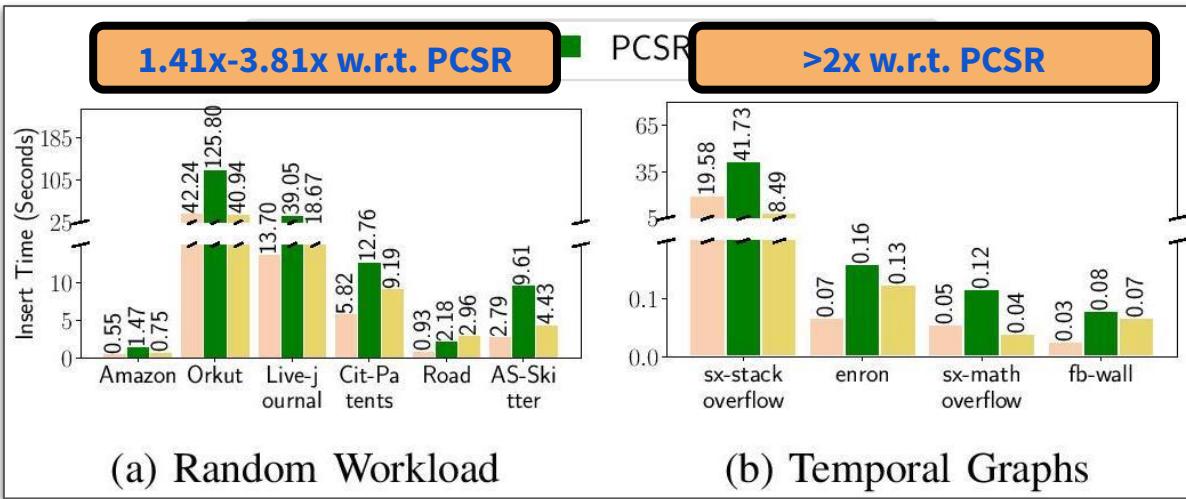


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Evaluation: Graph Insertion Performance

Dataset	Tree Levels	VCSR (Mem. Acc.)	PCSR (Mem. Acc.)
Amazon (shuffled)	[1-3)	710 (0.84 M)	251077 (35.50 M)
	[3-7)	4 (0.02 M)	12964 (8.49 M)
	≥ 7	0 (0.00 M)	45 (0.41 M)
sx-Stack overflow	[1-3)	904363 (457.23 M)	13876752 (2262.72 M)
	[3-7)	357575 (1056.19 M)	3053307 (2720.56 M)
	[7-15)	26530 (1594.56 M)	70307 (1704.70 M)
	≥ 15	21 (438.37 M)	288 (2531.68 M)

Table: Number of re-balancing operations triggered by edge insertions in VCSR and PCSR on two graphs.

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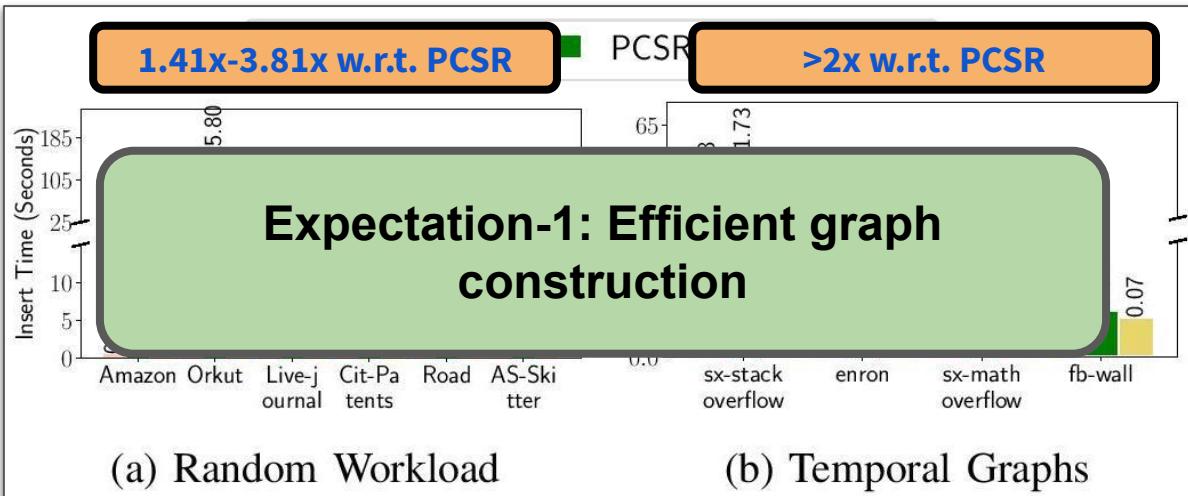


Figure: Comparing VCSR's dynamic graph insertion performance (in seconds) for 10% pre-initialization.

Evaluation: Graph Analytic Algorithms Performance

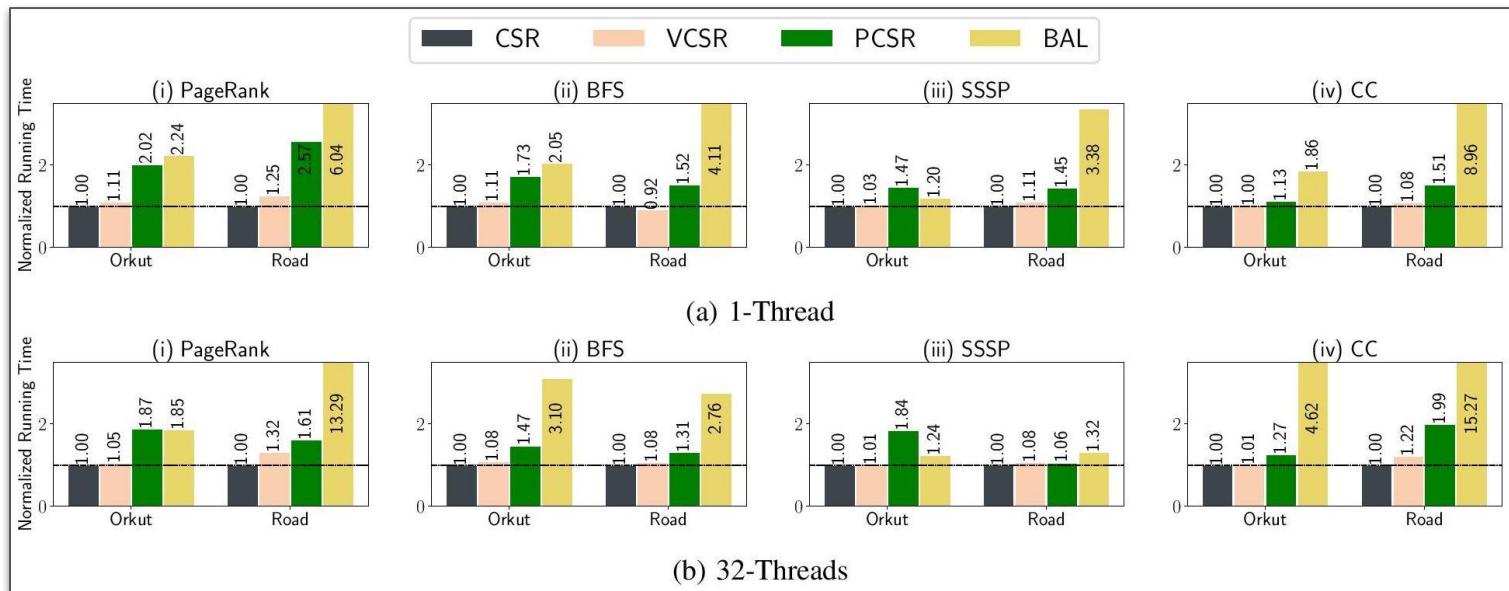


Figure: Comparing graph analysis runtime normalized to CSR.

Evaluation: Graph Analytic Algorithms Performance

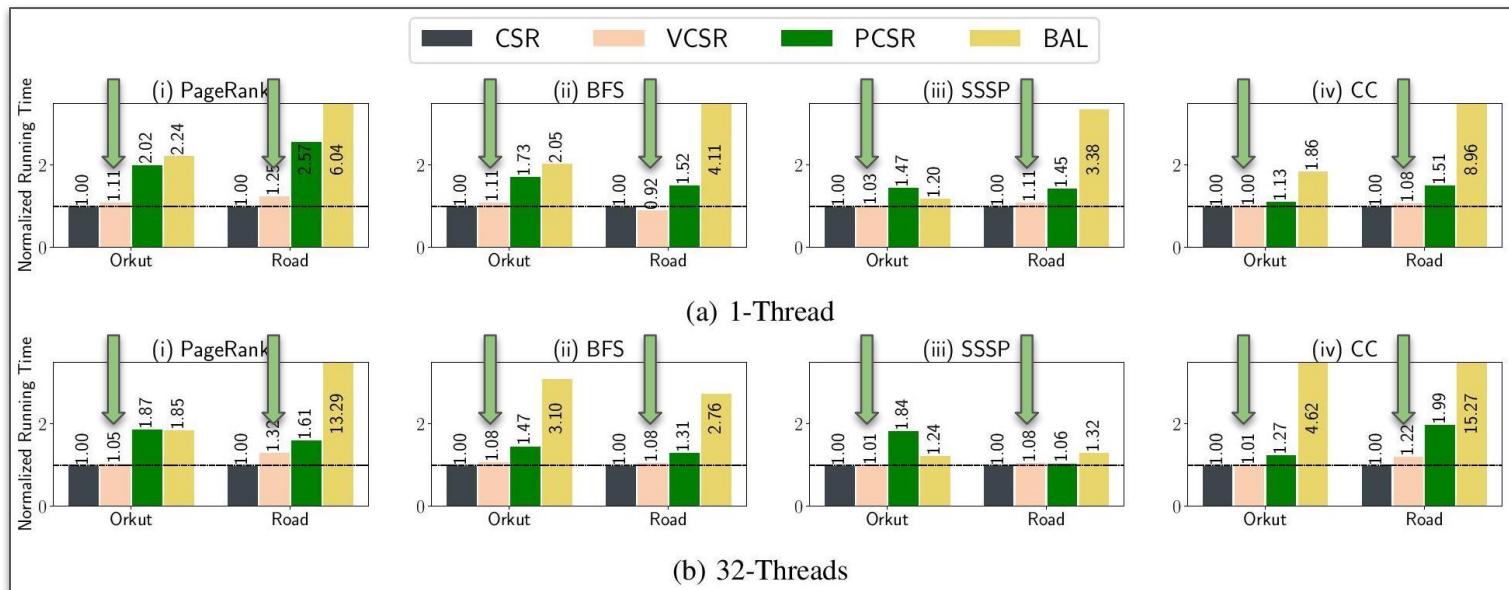


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Evaluation: Graph Analytic Algorithms Performance

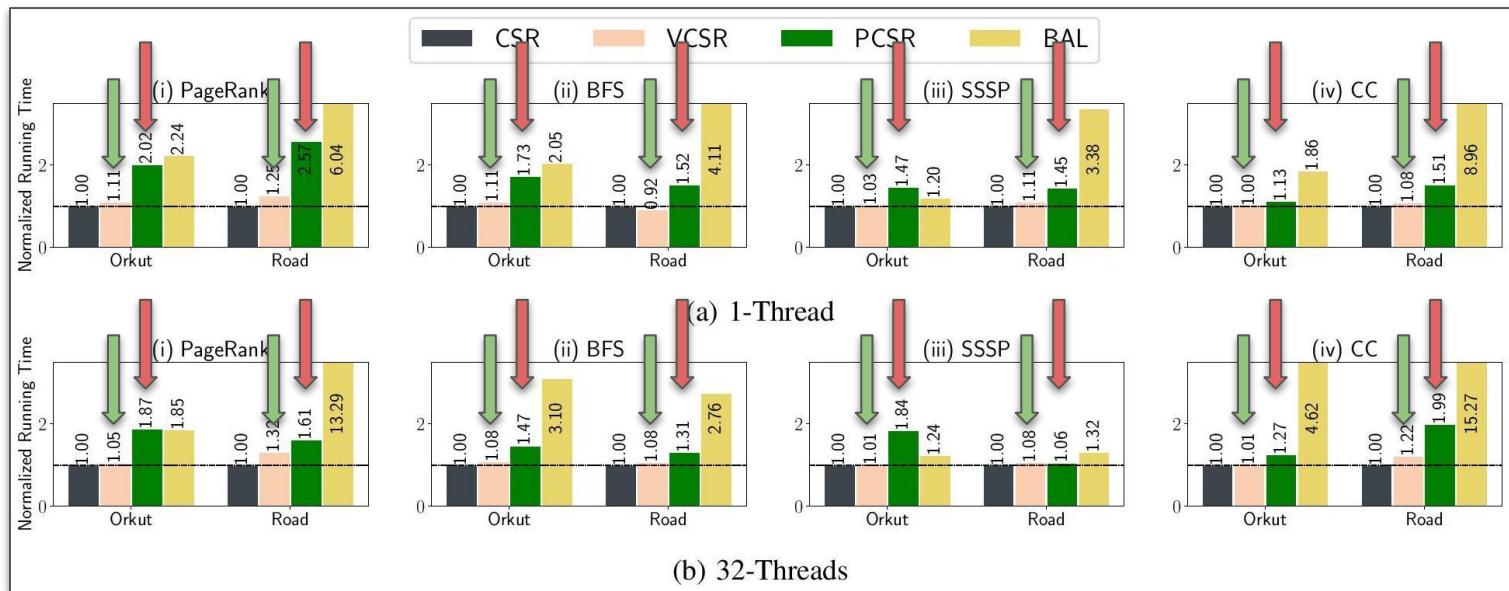


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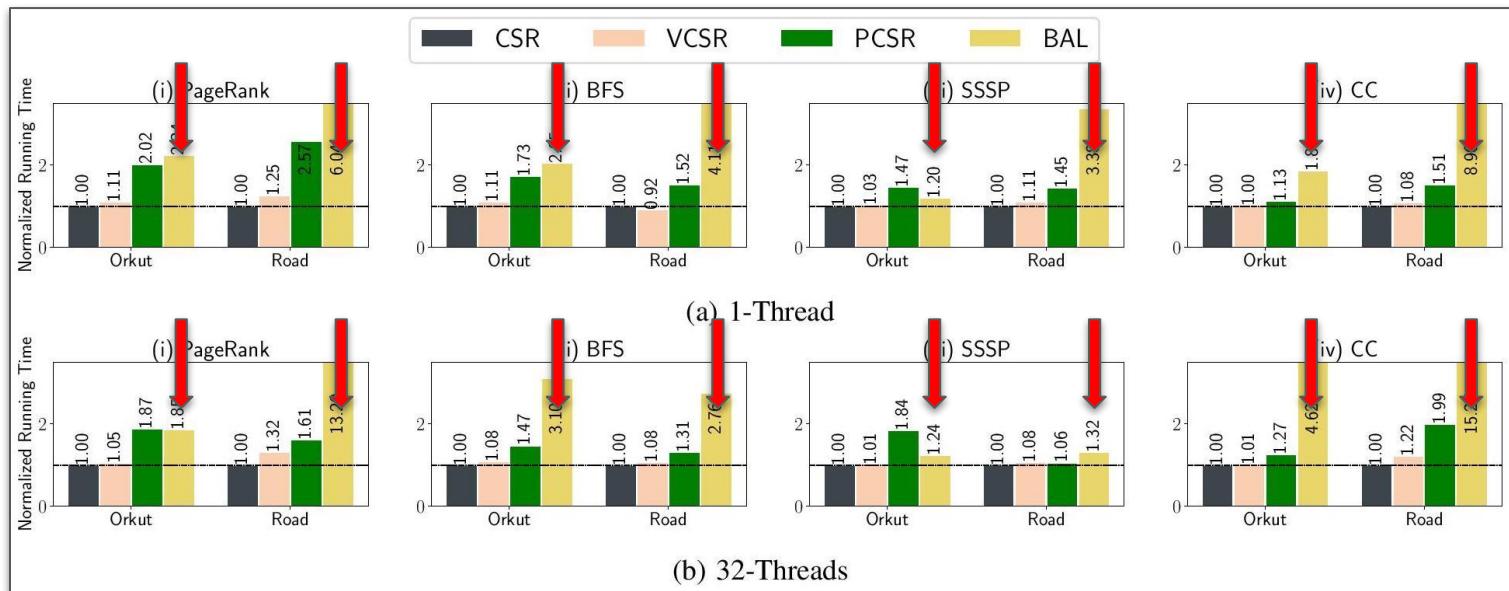


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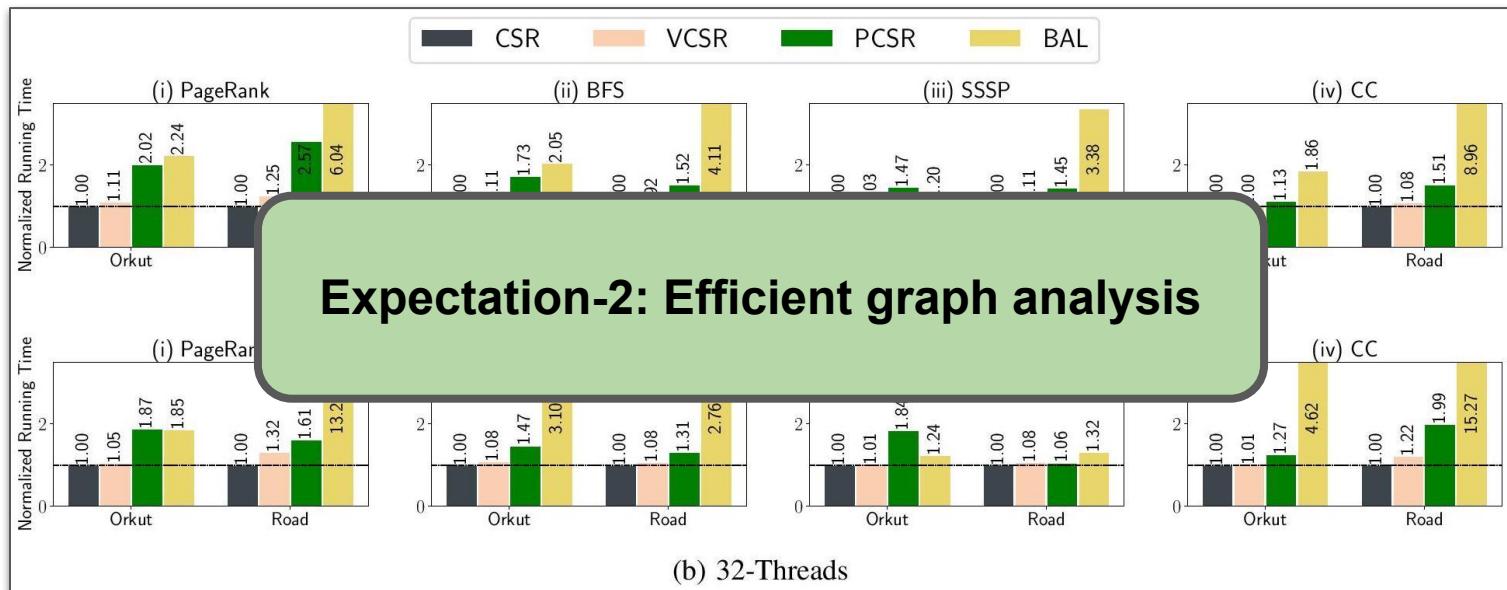


Figure: Comparing graph analysis runtime normalized to CSR.

Summary

- Identify fundamental limitations in existing PMA based mutable CSR extension
- Demonstrate graph's power-law also exist while it evolves
- Design a novel vertex-centric CSR extension: VCSR
 - Solves the fundamental limitations in handling graph imbalances
- Evaluation results
 - 1.41x-3.81x better performance in graph insertions
 - 1.22x-2.05x better performance in running typical graph analytic algorithms

Thank You

Question?



DIRLAB 84