

Indexing for Near-Sorted Data

Aneesh Raman

Subhadeep Sarkar

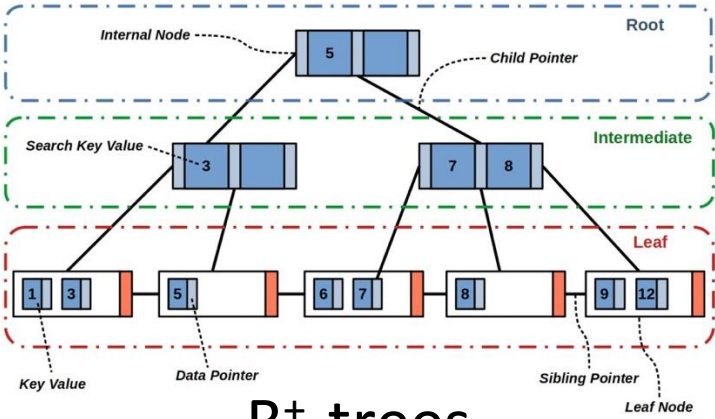
Matthaios Olma

Manos Athanassoulis

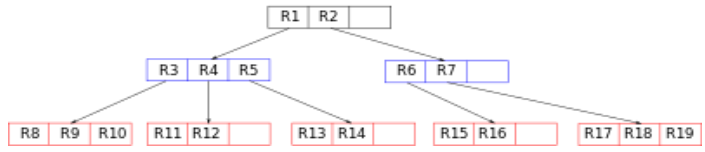
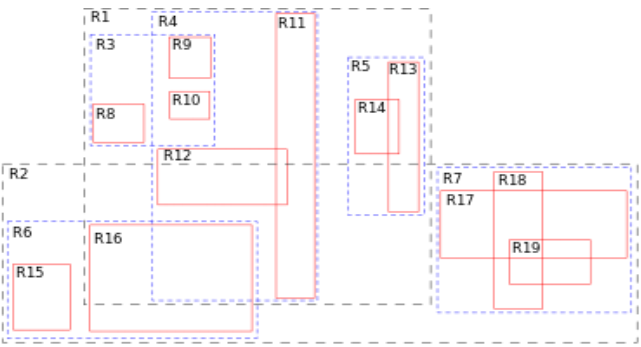
BOSTON
UNIVERSITY



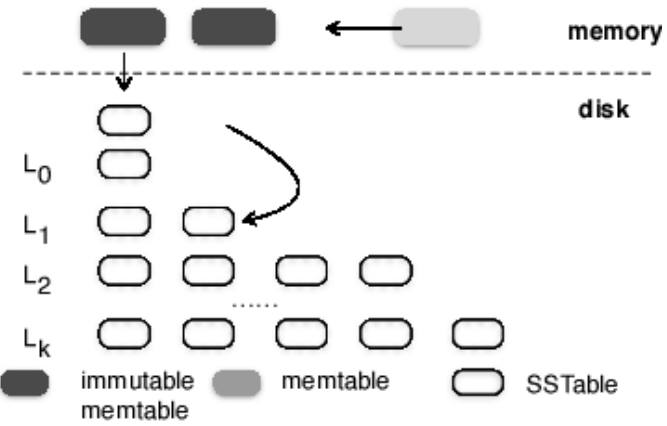
Indexes in Databases



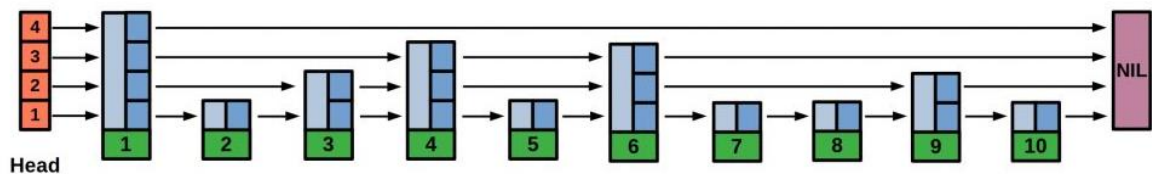
B⁺-trees



R-trees

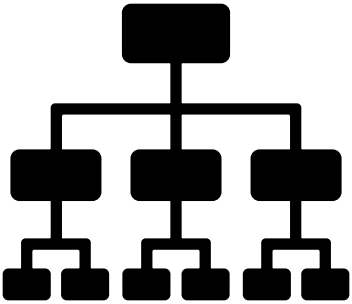


LSM-trees

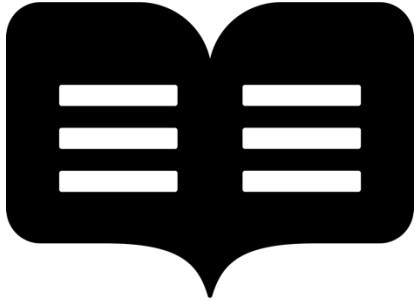


Skip lists

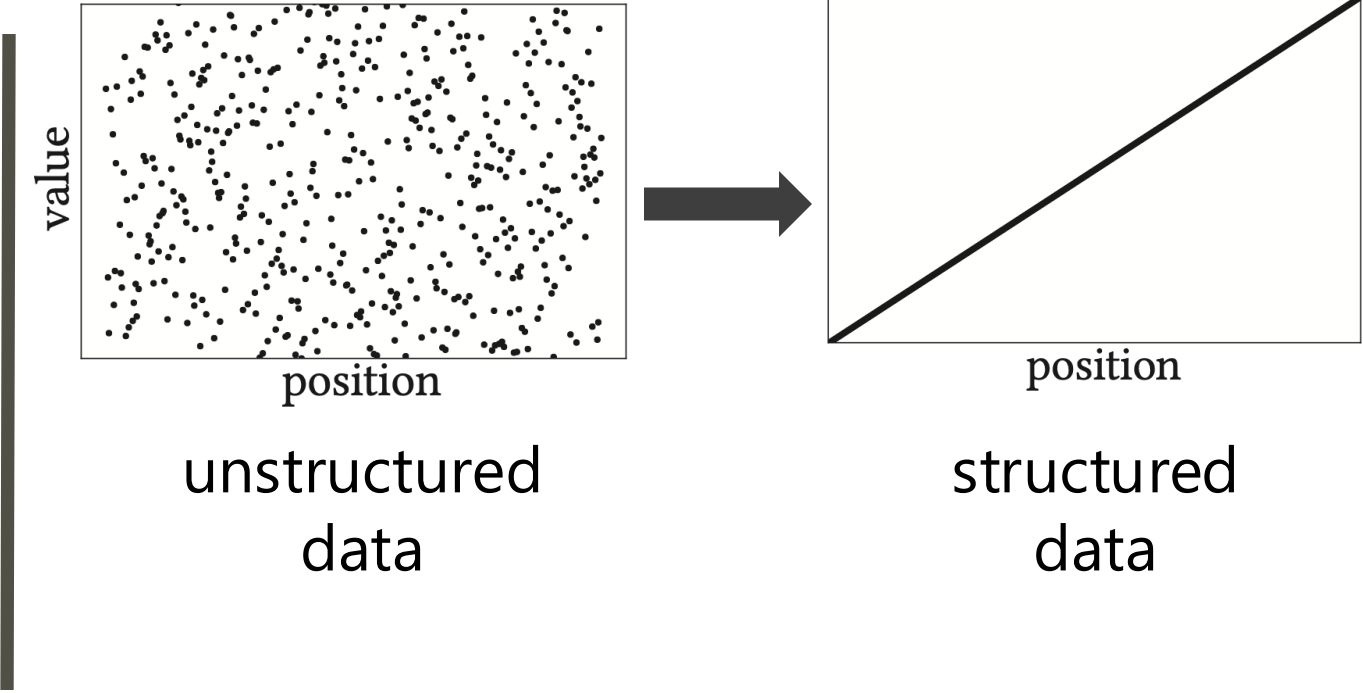
Indexes in Databases



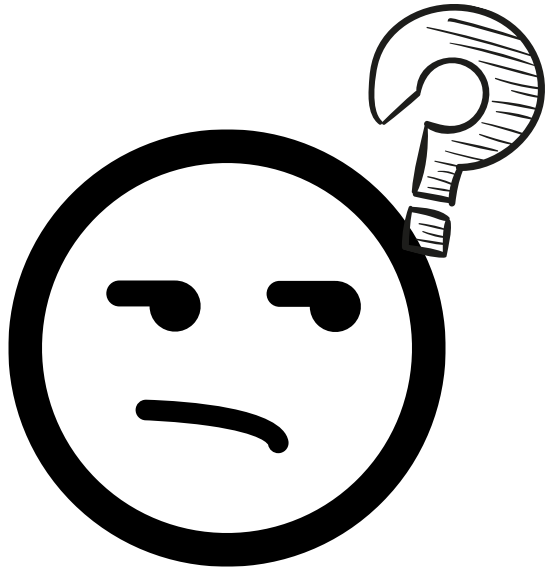
organize
data



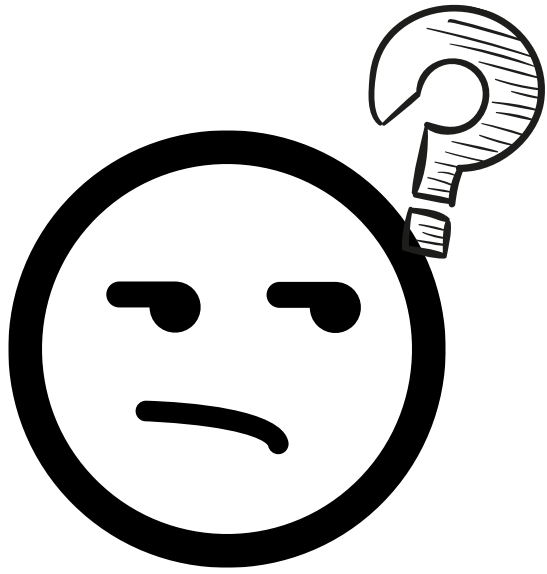
efficient
queries



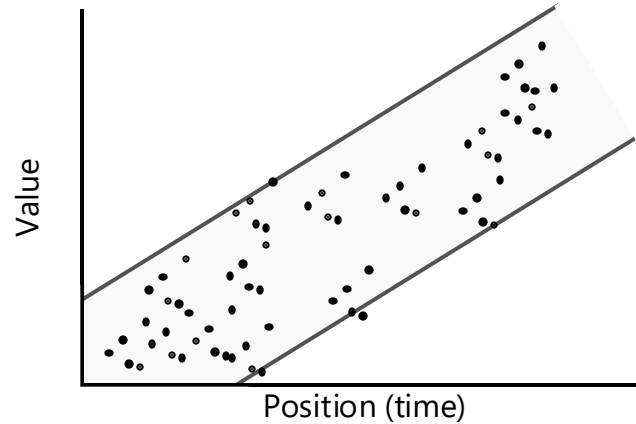
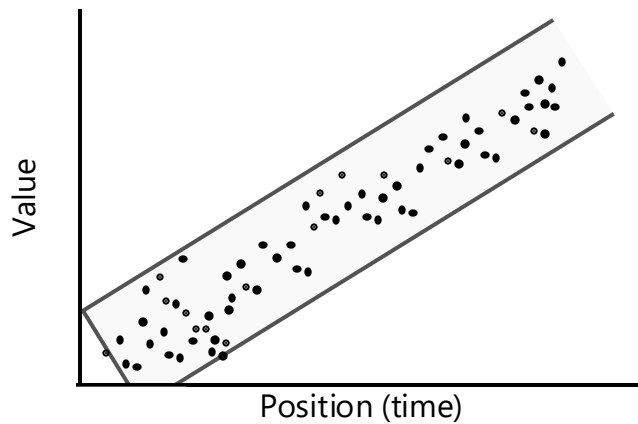
The process of inducing “sortedness” to an otherwise unsorted data collection



What if data already has some structure?



What if data already has some structure?

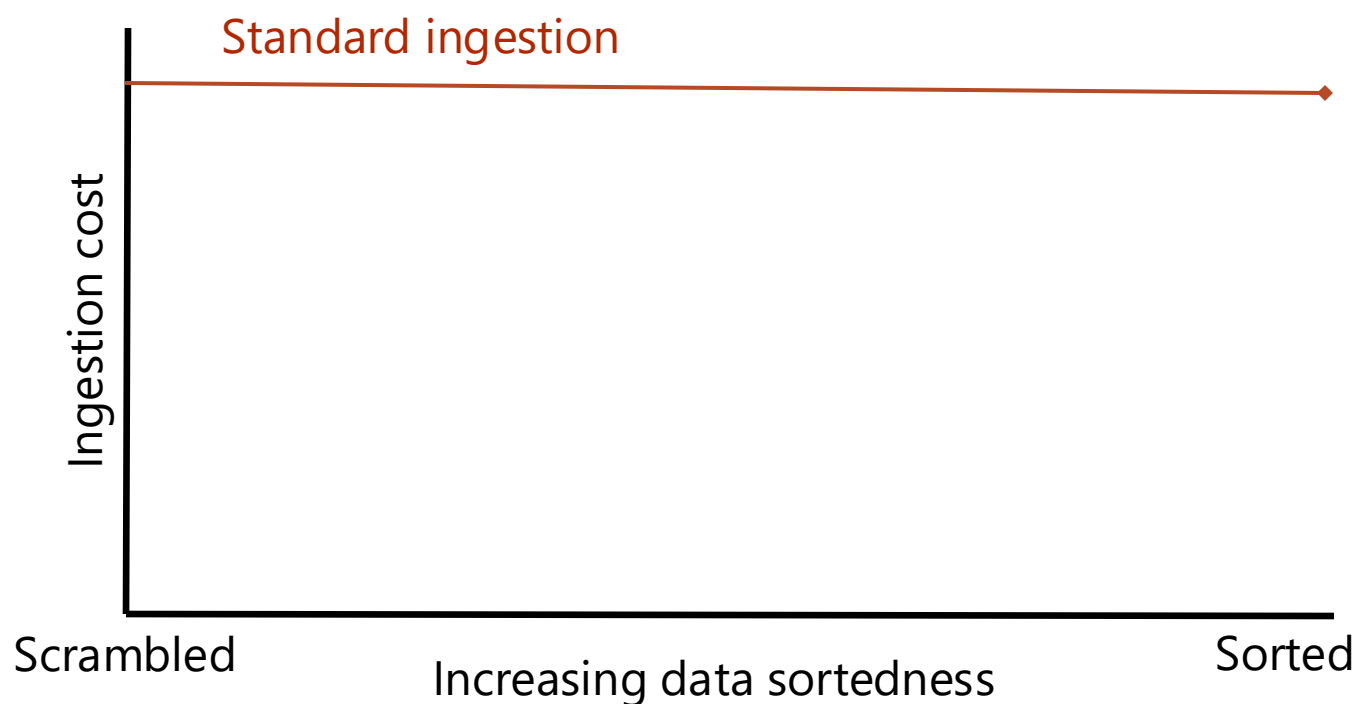


treated same as unstructured data!

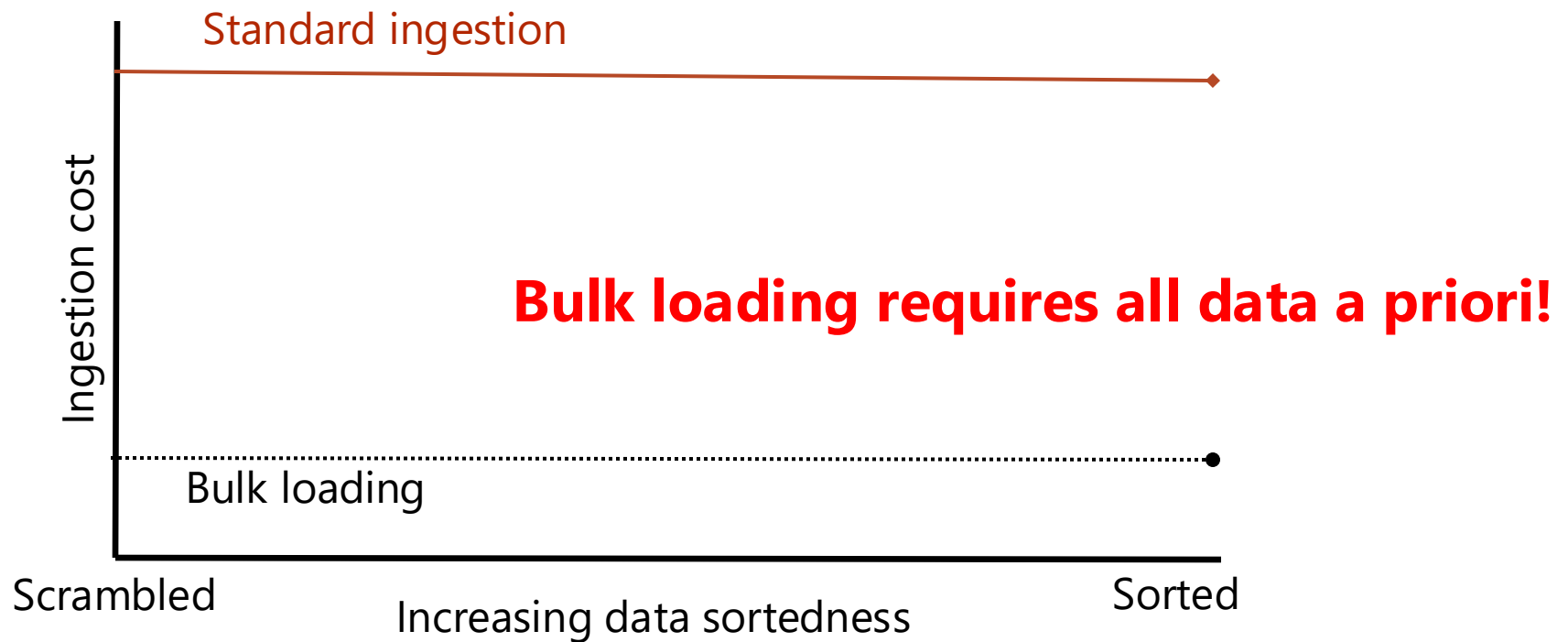


Near-sorted data

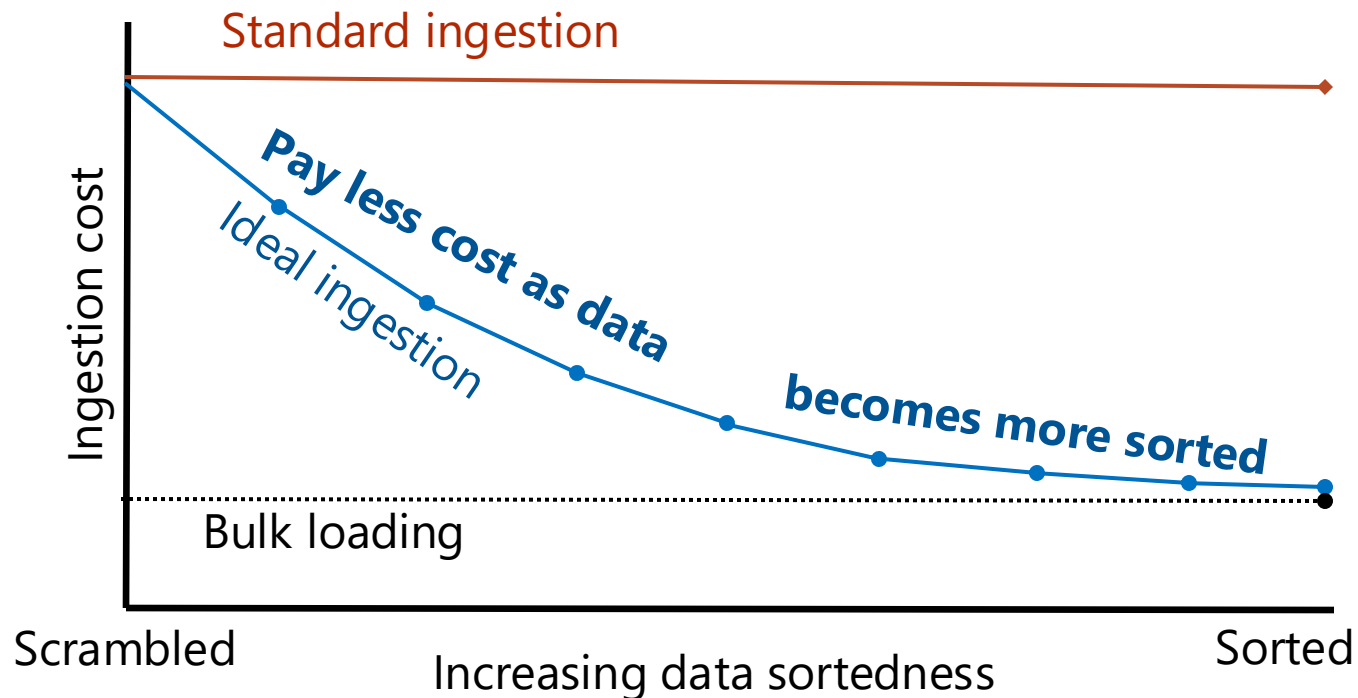
Irrespective of Sortedness, Same Ingestion Performance



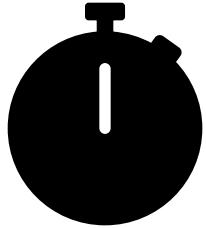
Are There Faster Alternatives?



Ideally, Higher Sortedness Should Lead to Faster Ingestion



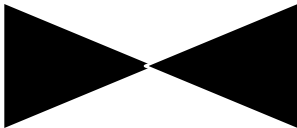
Near-Sorted Data is Frequently Found



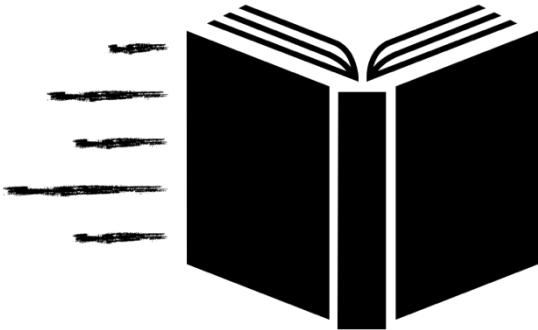
Time Series



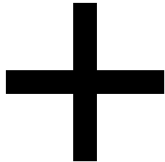
Stock market



Join/query



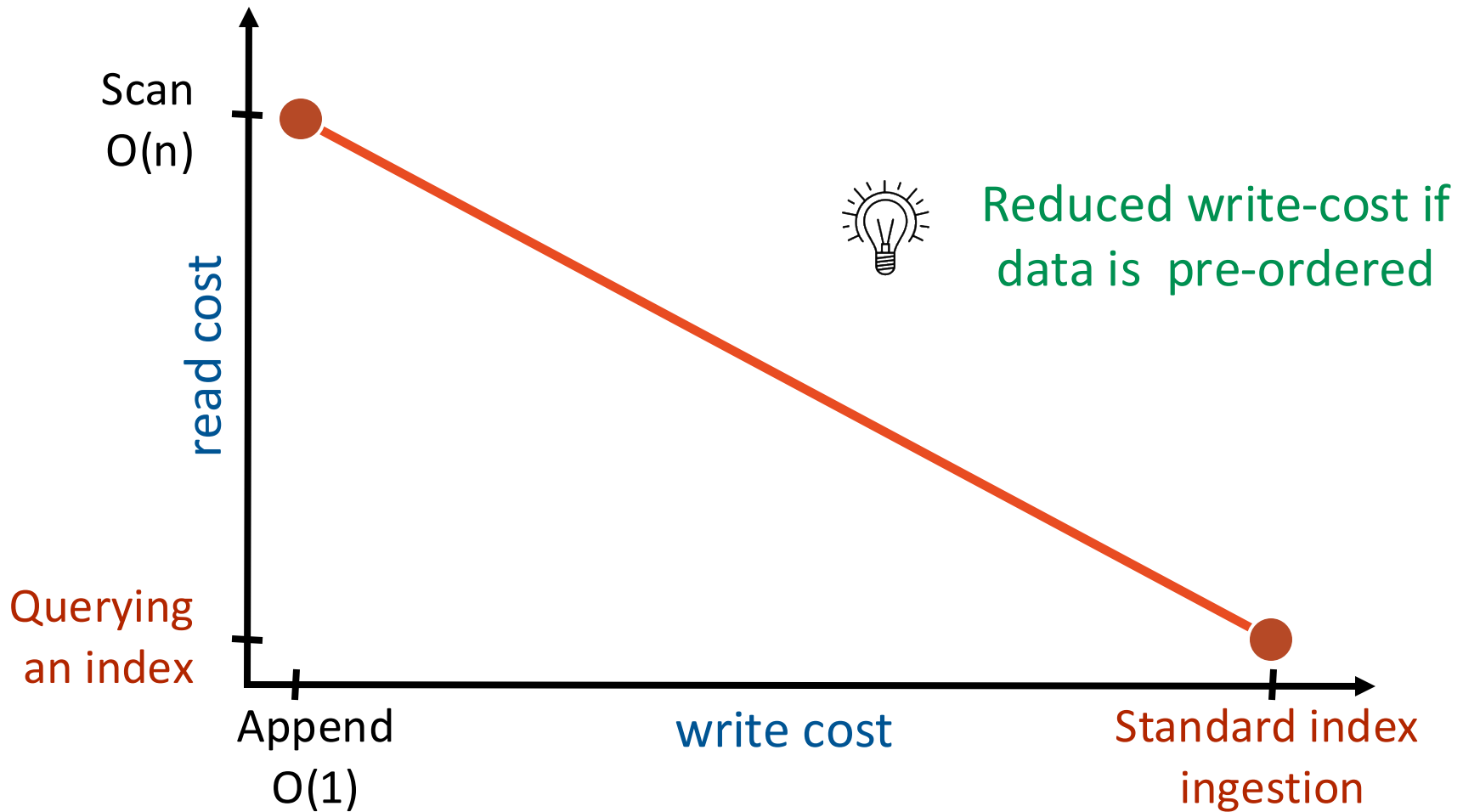
efficient reads



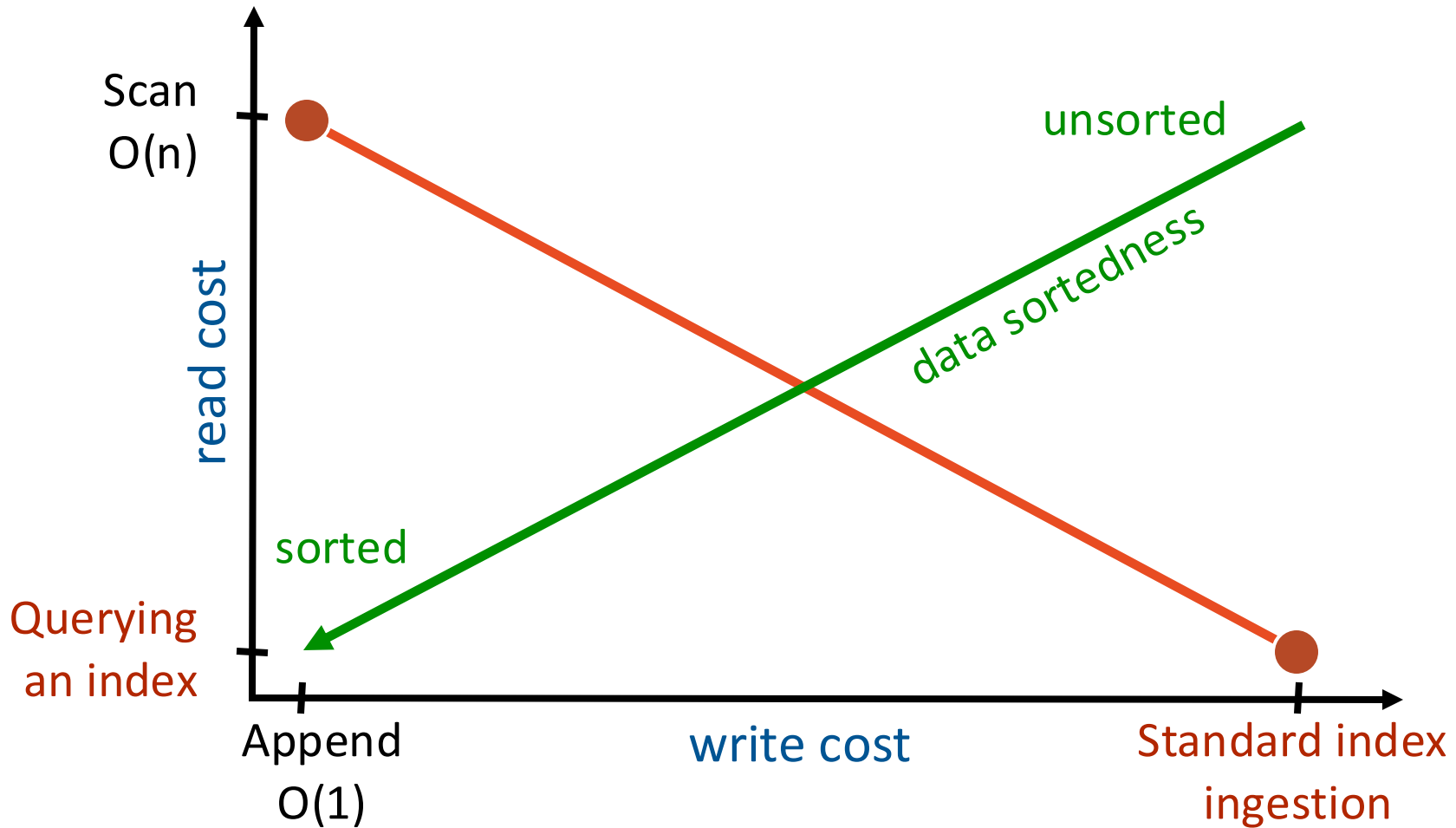
fast writes

classical indexes carry ***redundant effort!***

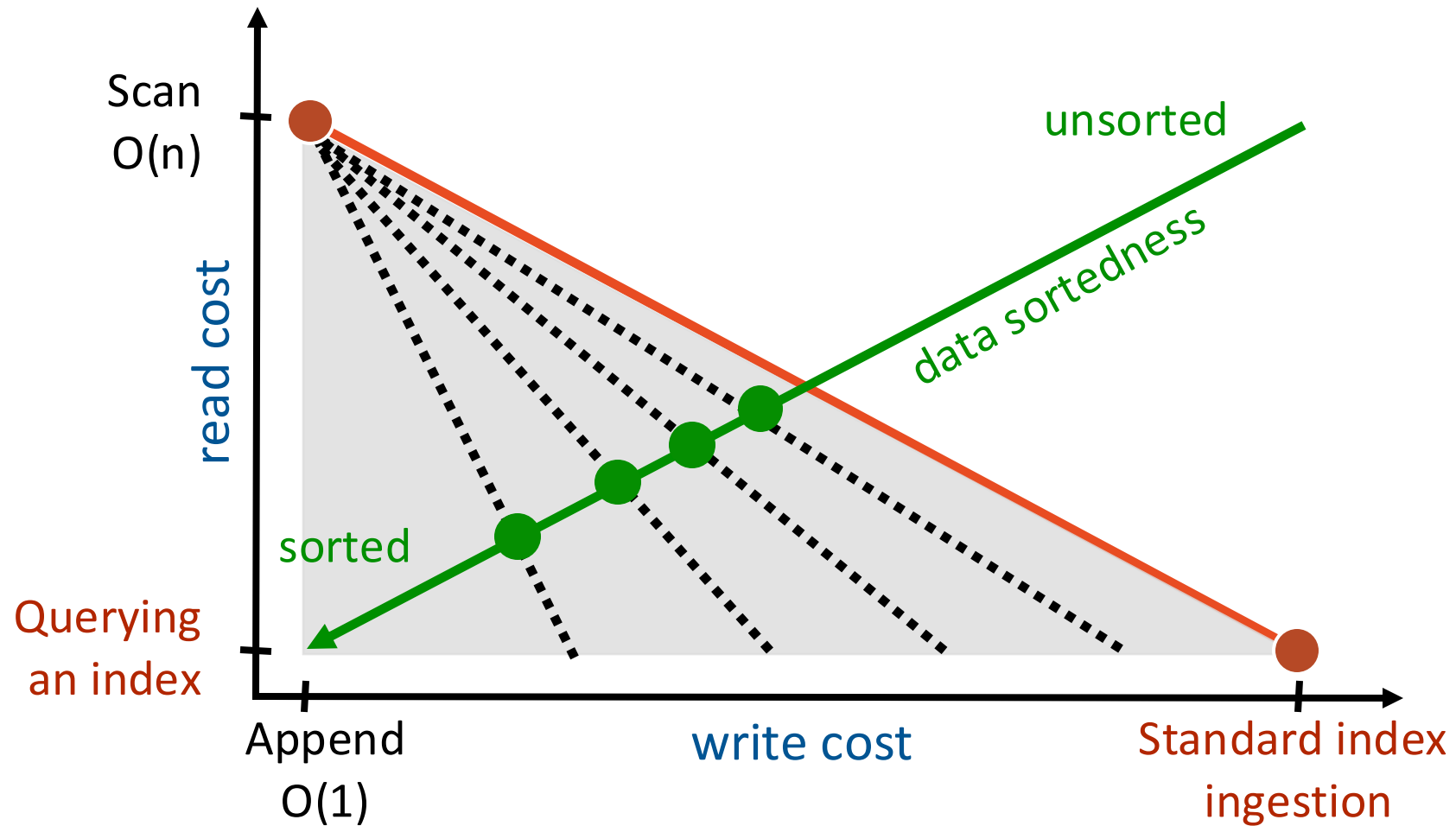
Vision for Sortedness-Aware Indexing



Vision: Sortedness-Aware Indexes



Vision: Sortedness-Aware Indexes



Agenda

Introduction

Vision

Sortedness Metrics

Sortedness Aware (SWARE) Indexing

A Simpler Design

Open Questions

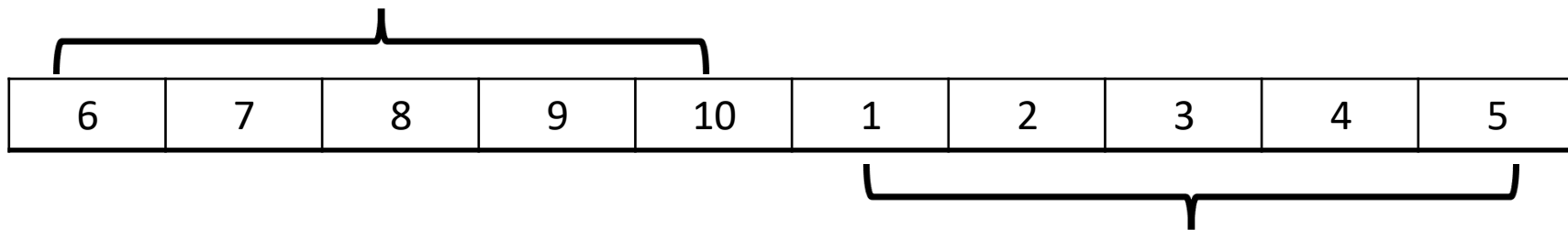
Quantifying Data Sortedness

Metric	Description
Inversions	# pairs in incorrect order
Runs	# increasing contiguous subsequences
Exchanges	least # swaps needed to establish total order

Any downsides of the
"simple" metrics?

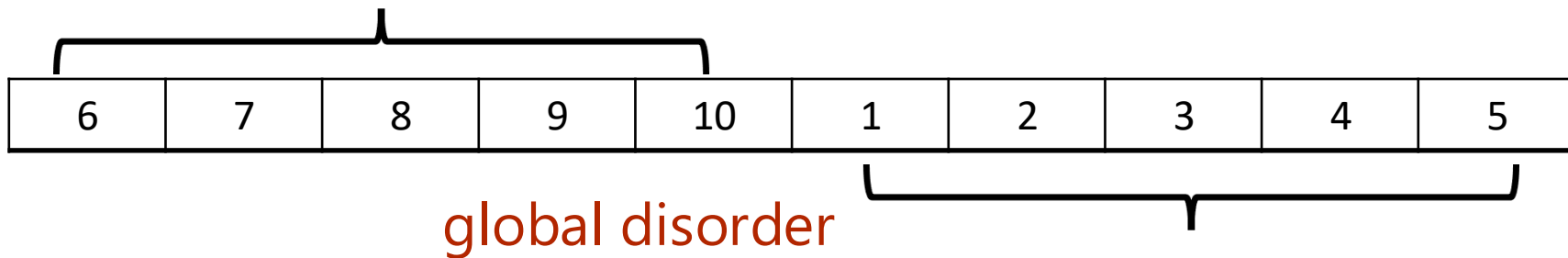
Quantifying Data Sortedness

Metric	Description
Inversions	# pairs in incorrect order
Runs	# increasing contiguous subsequences
Exchanges	least # swaps needed to establish total order



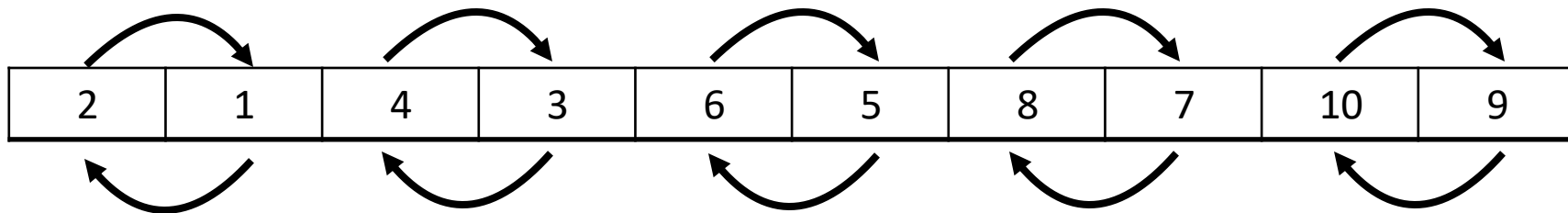
Quantifying Data Sortedness

Metric		Description
Inversions	⊖	# pairs in incorrect order
Runs	✓	# increasing contiguous subsequences
Exchanges	⊖	least # swaps needed to establish total order






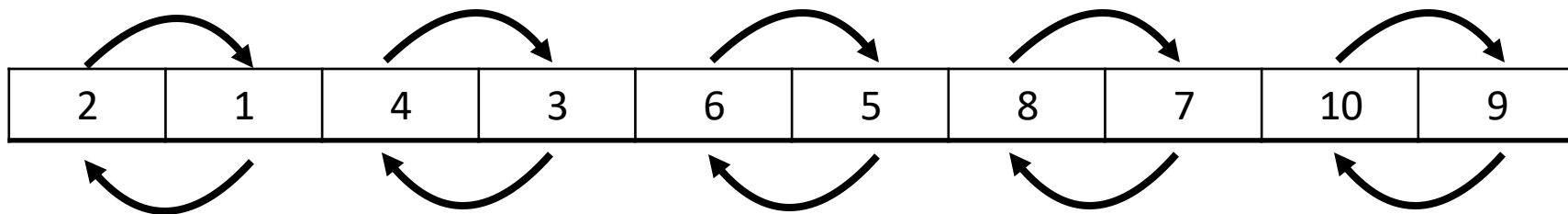
Quantifying Data Sortedness

Metric	Description
Inversions	# pairs in incorrect order
Runs	# increasing contiguous subsequences
Exchanges	least # swaps needed to establish total order



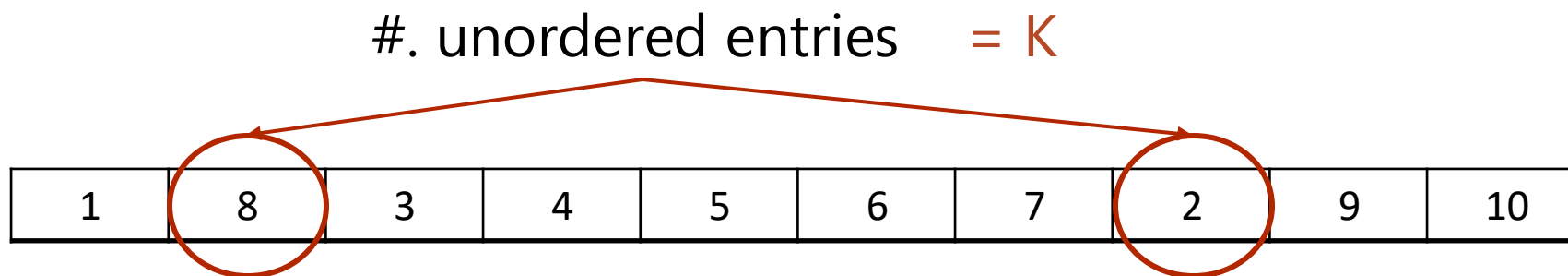
Quantifying Data Sortedness

Metric	Description
Inversions 	# pairs in incorrect order
Runs 	# increasing contiguous subsequences
Exchanges 	least # swaps needed to establish total order

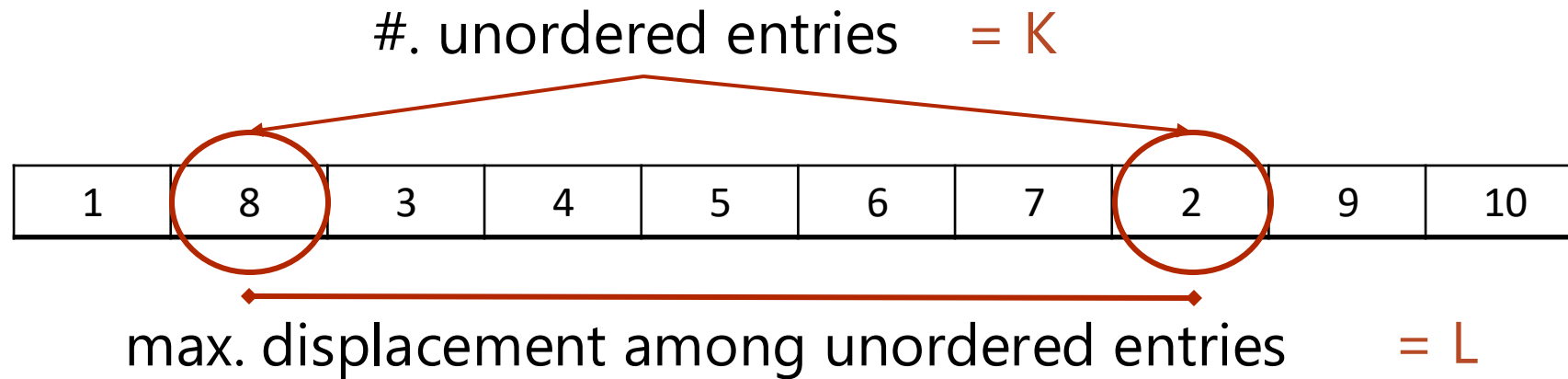


local disorder

(K, L)-Sortedness Metric



(K, L)-Sortedness Metric



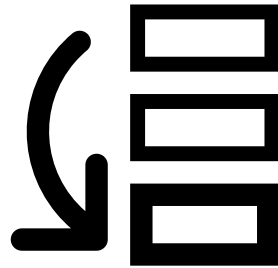
The Sortedness-Aware (SWARE) Paradigm

Sortedness-Aware (SWARE) Paradigm



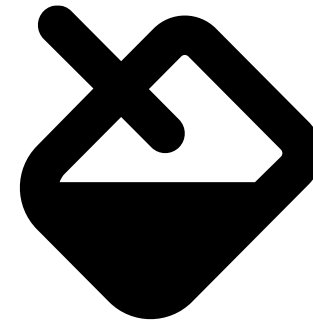
intelligent
buffering

+



opportunistic
bulk loading

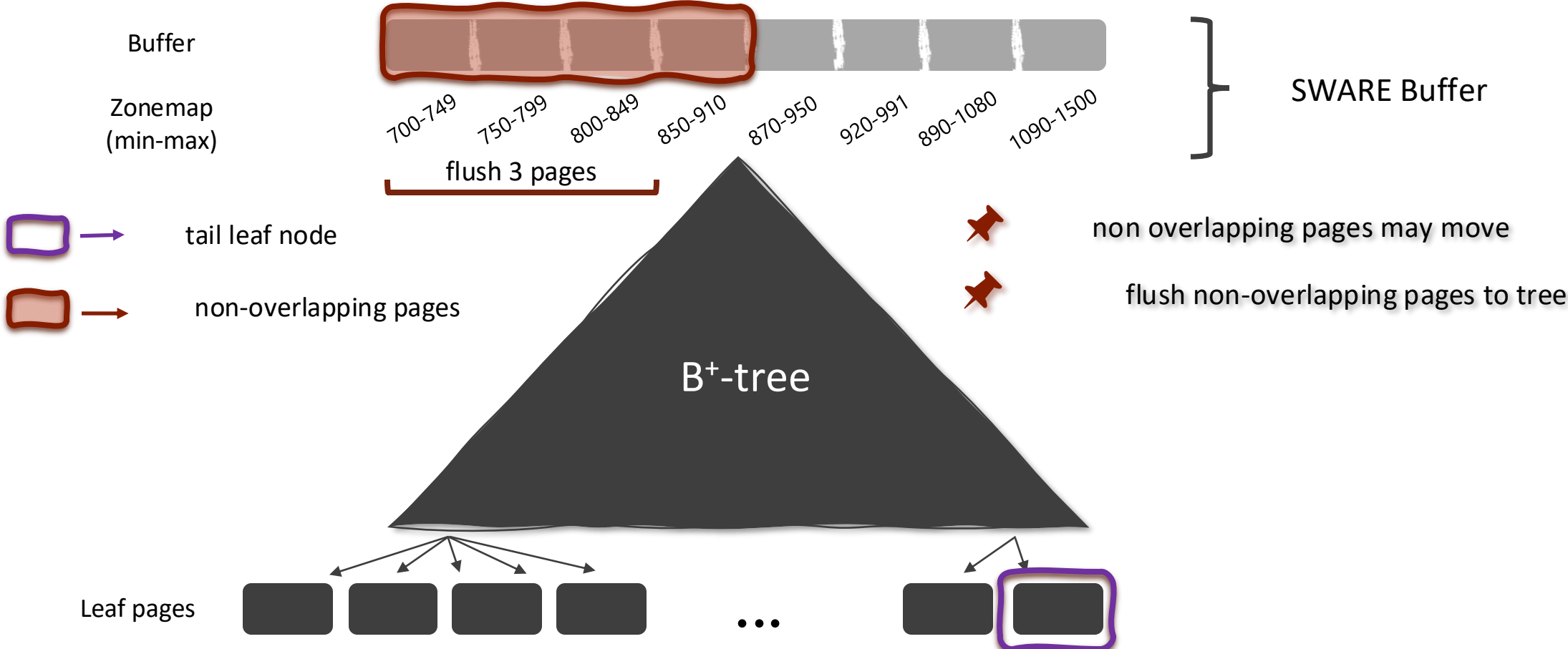
+



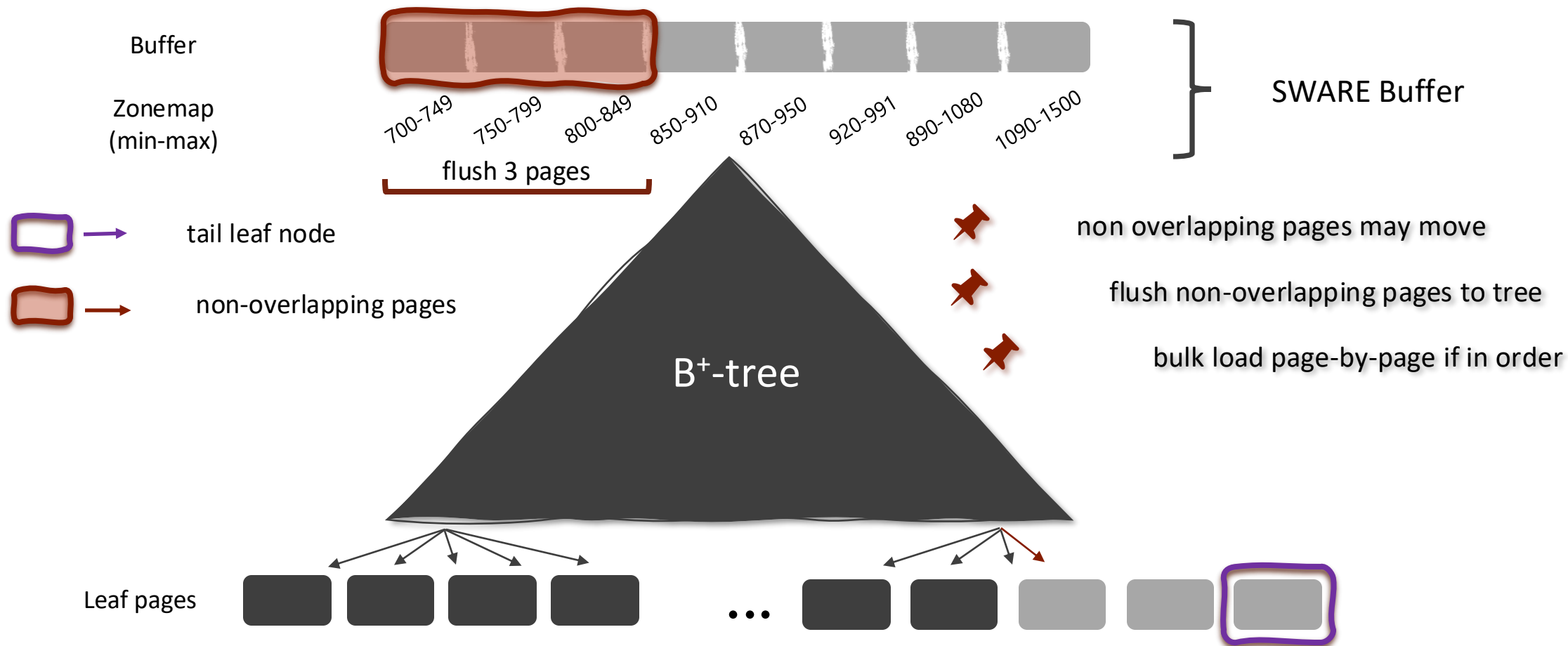
increased fill
and split factor

SWARE framework can be applied to any tree-index!

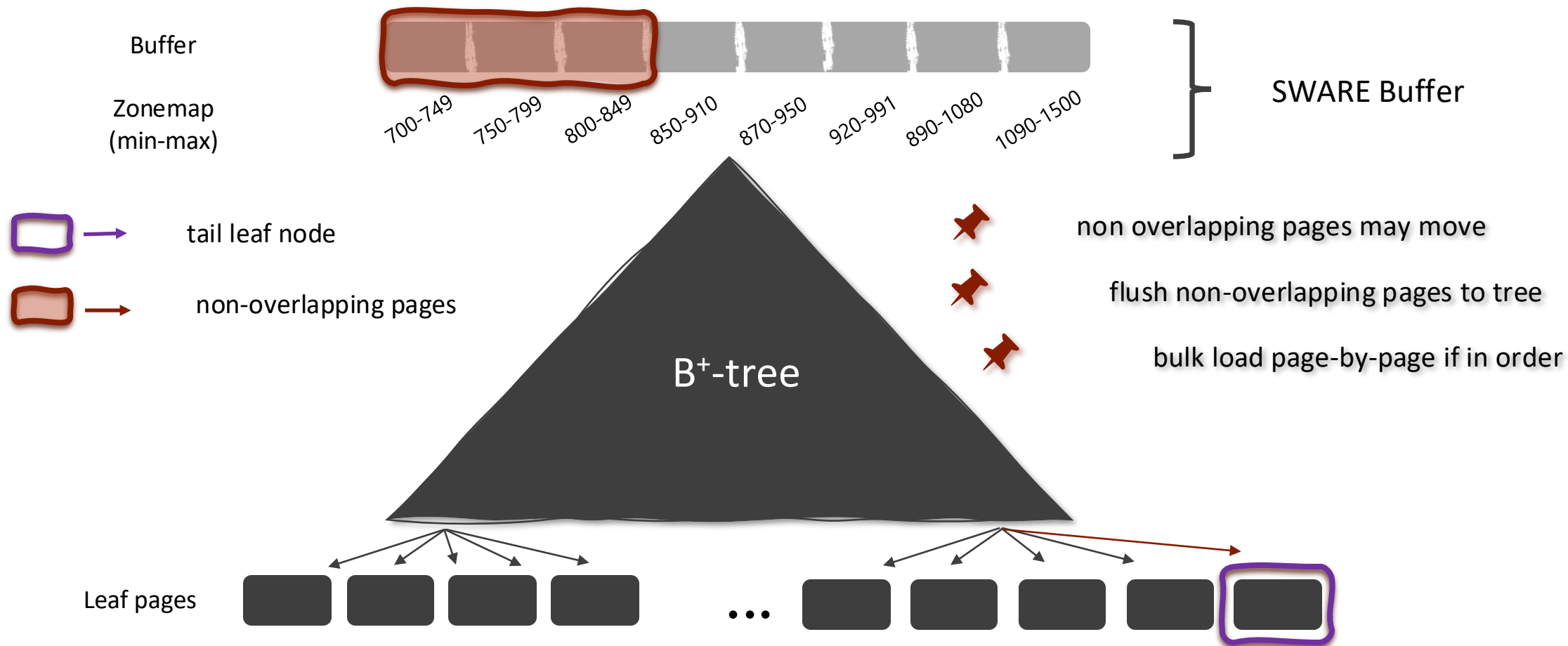
SWARE Ingestions



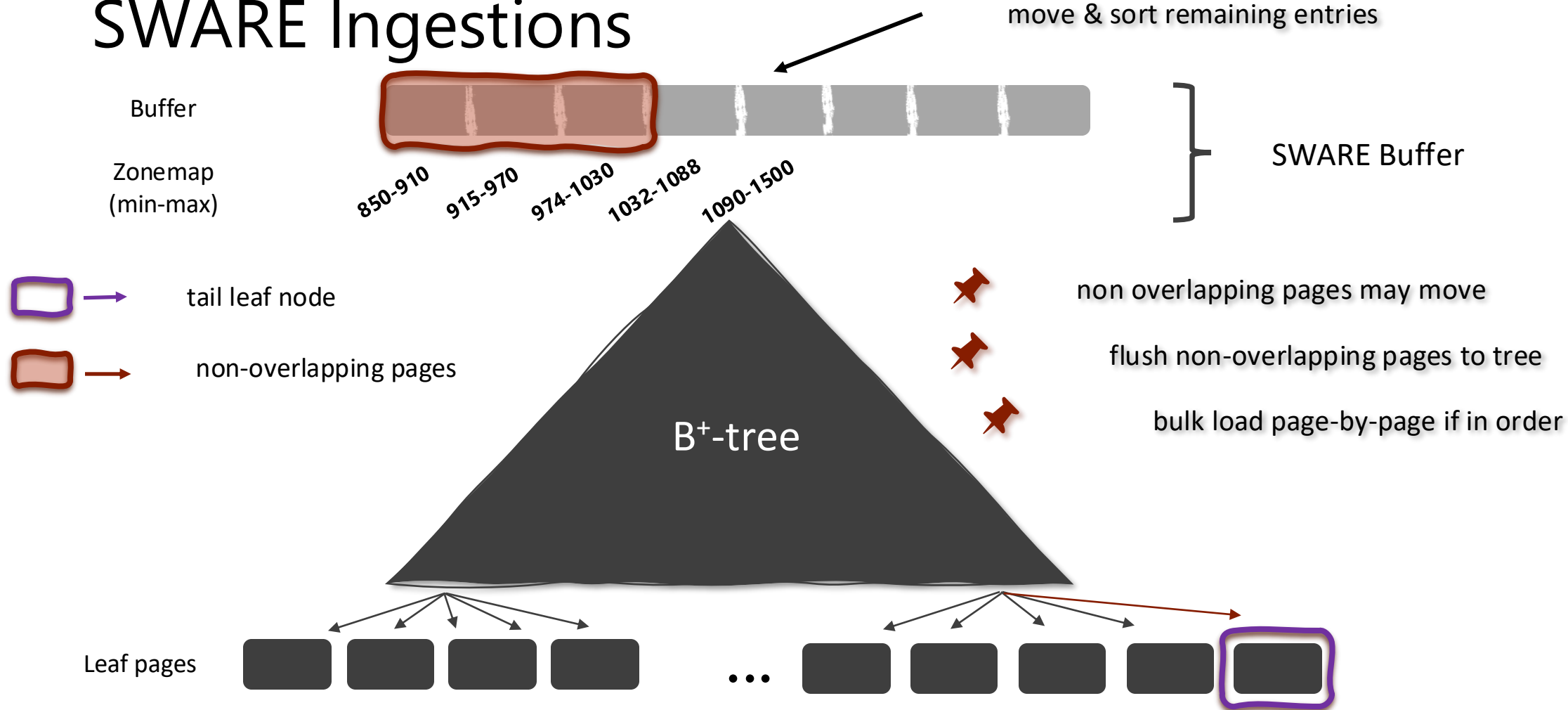
SWARE Ingestions



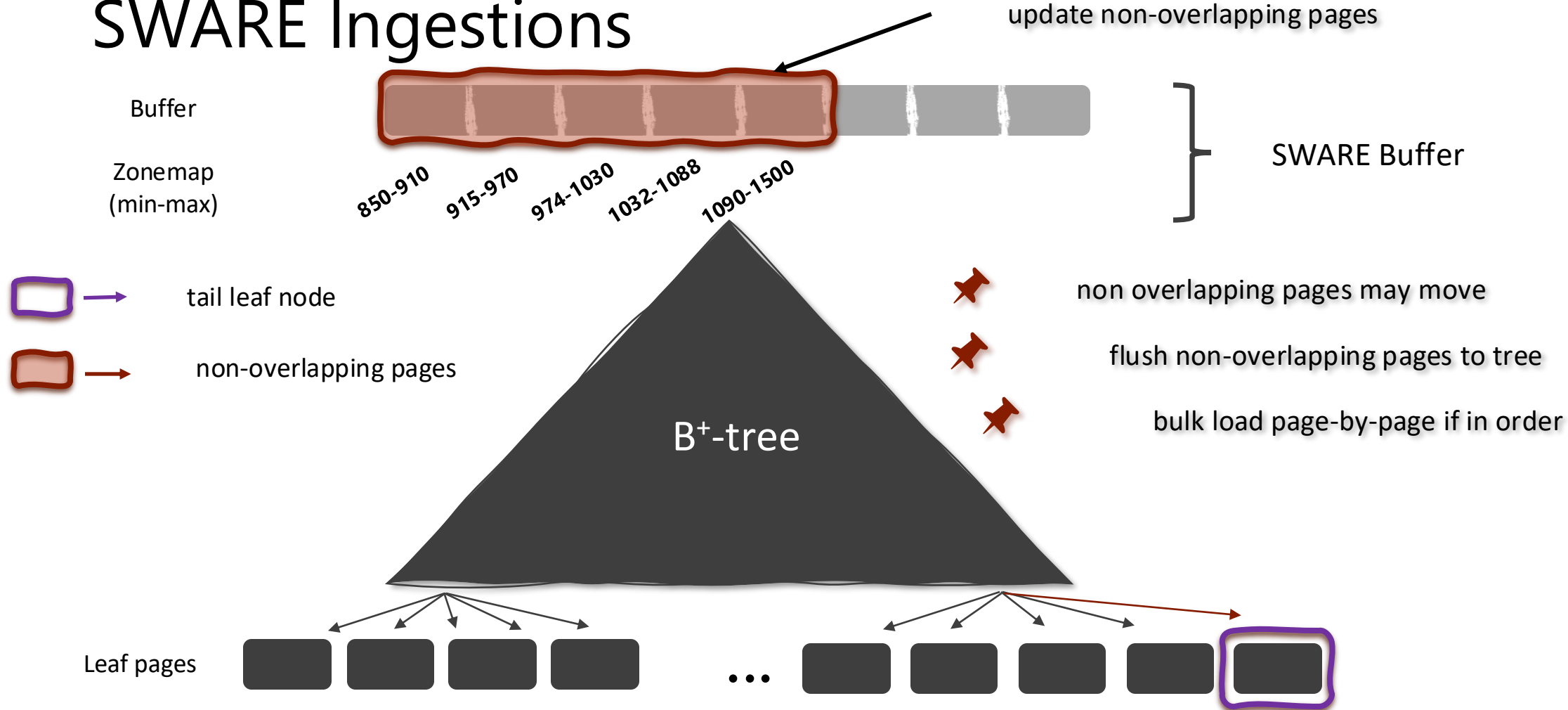
SWARE Ingestions



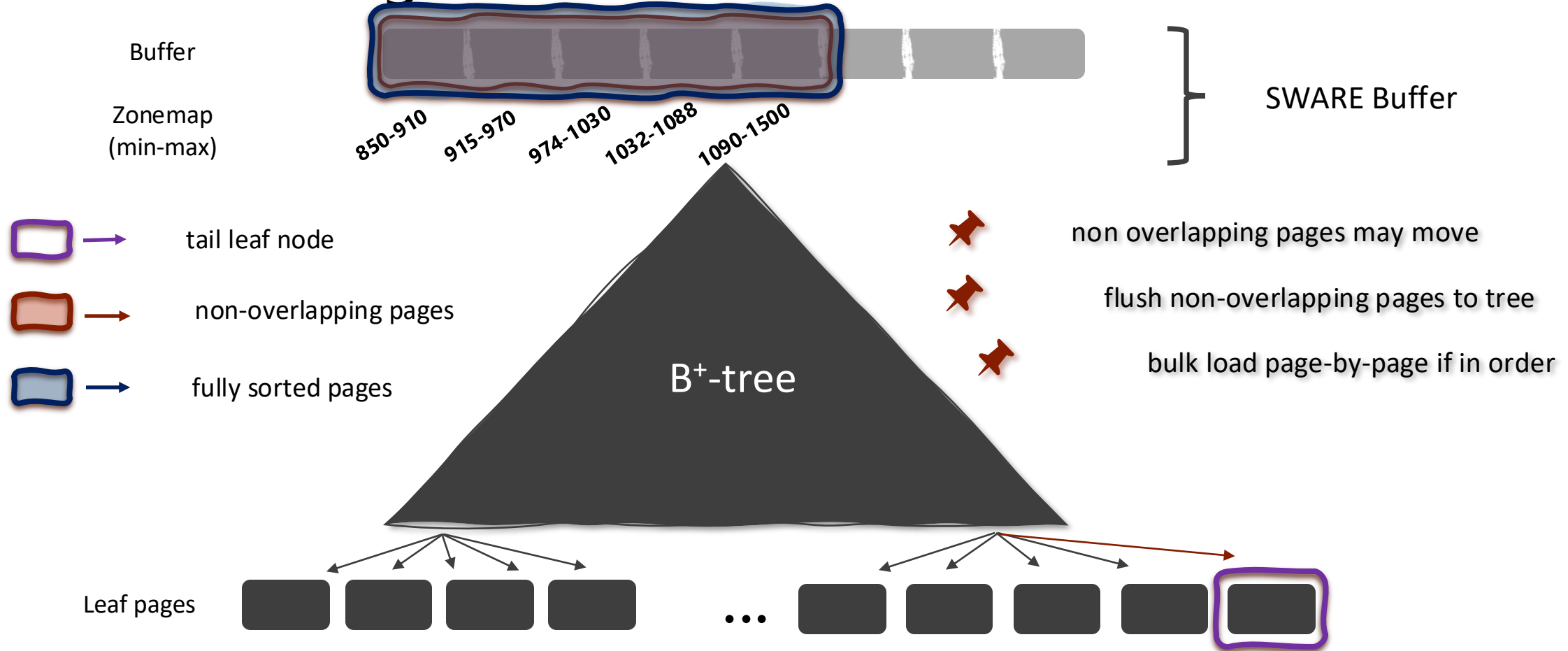
SWARE Ingestions



SWARE Ingestions

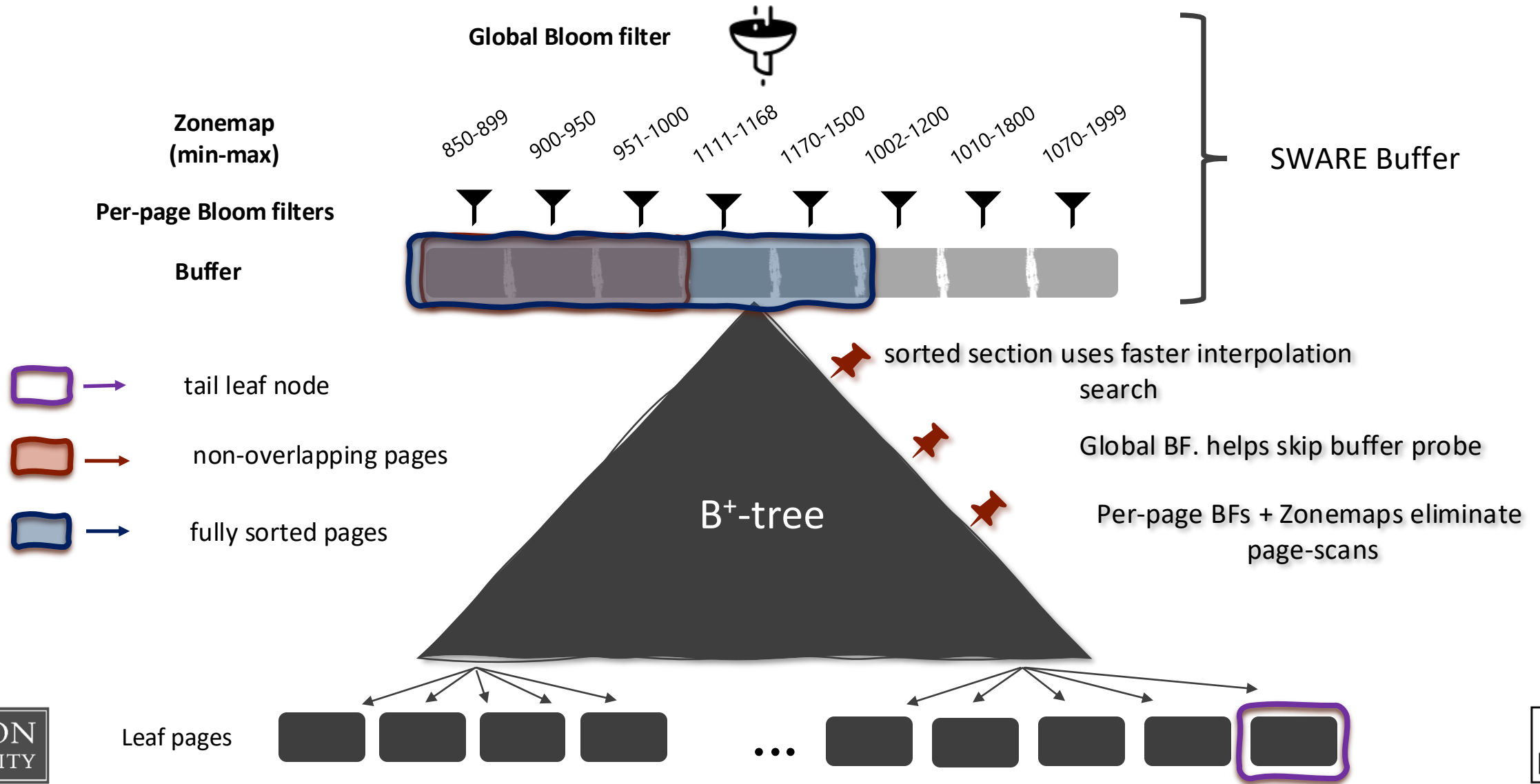


SWARE Ingestions



How do lookups work?

SWARE Lookups



Experimental Evaluation

System Setup:

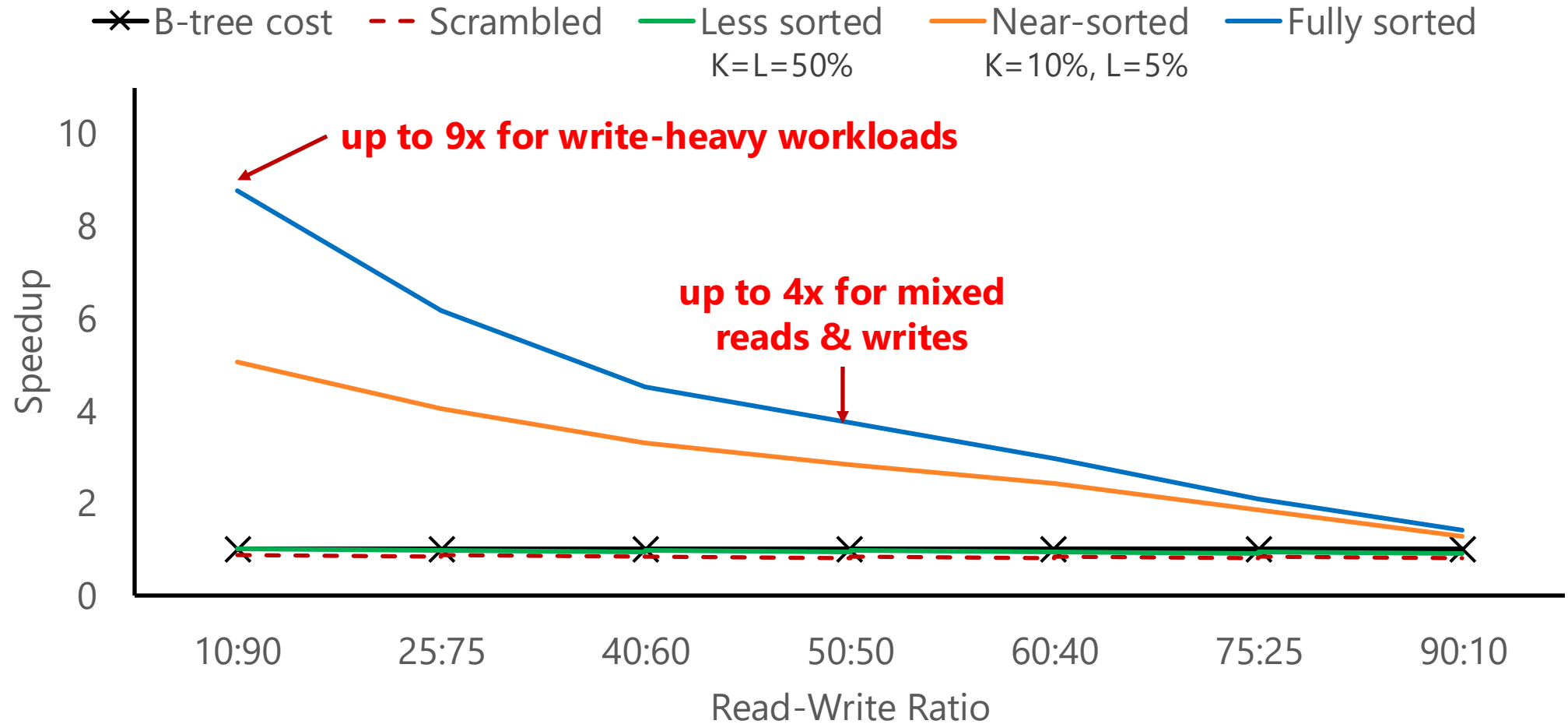
- Intel Xeon Gold 5230
- 2.1GHZ processor w. 20 cores
- 384GB RAM, 28MB L3 cache

Index Setup:

- Buffer = 40MB; flush \leq 50%
- BFs = 10 BPK; Murmur Hash
- Split at 80%

B^+ -tree design inspired by STX::B-tree can also work as B^ϵ -tree

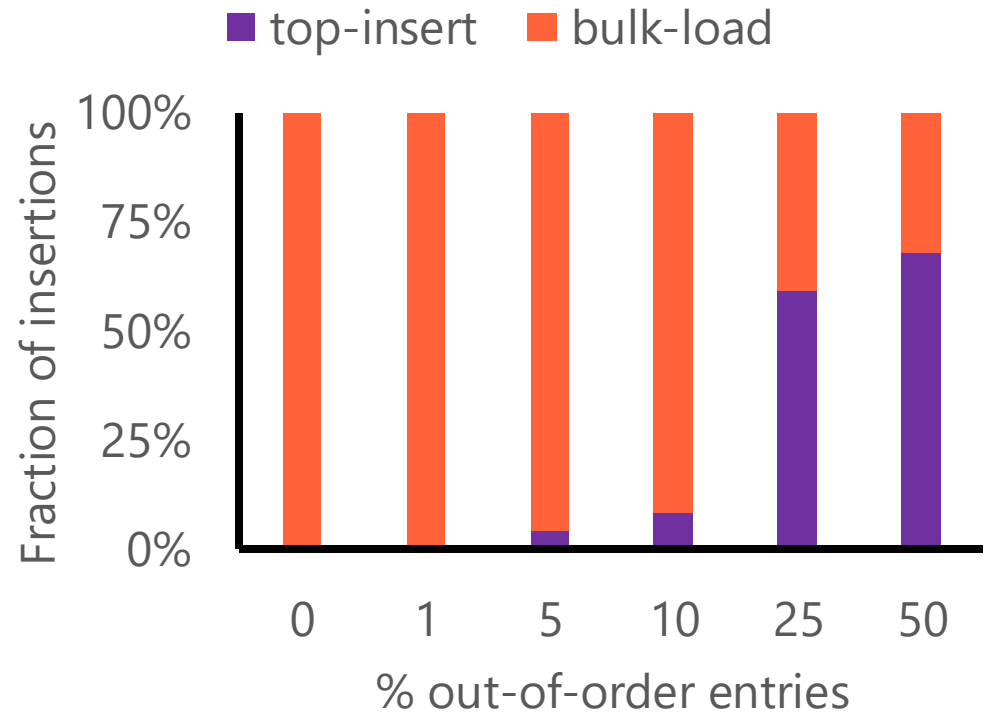
Evaluating SWARE Under Varying Sortedness



Raw Ingestion Performance

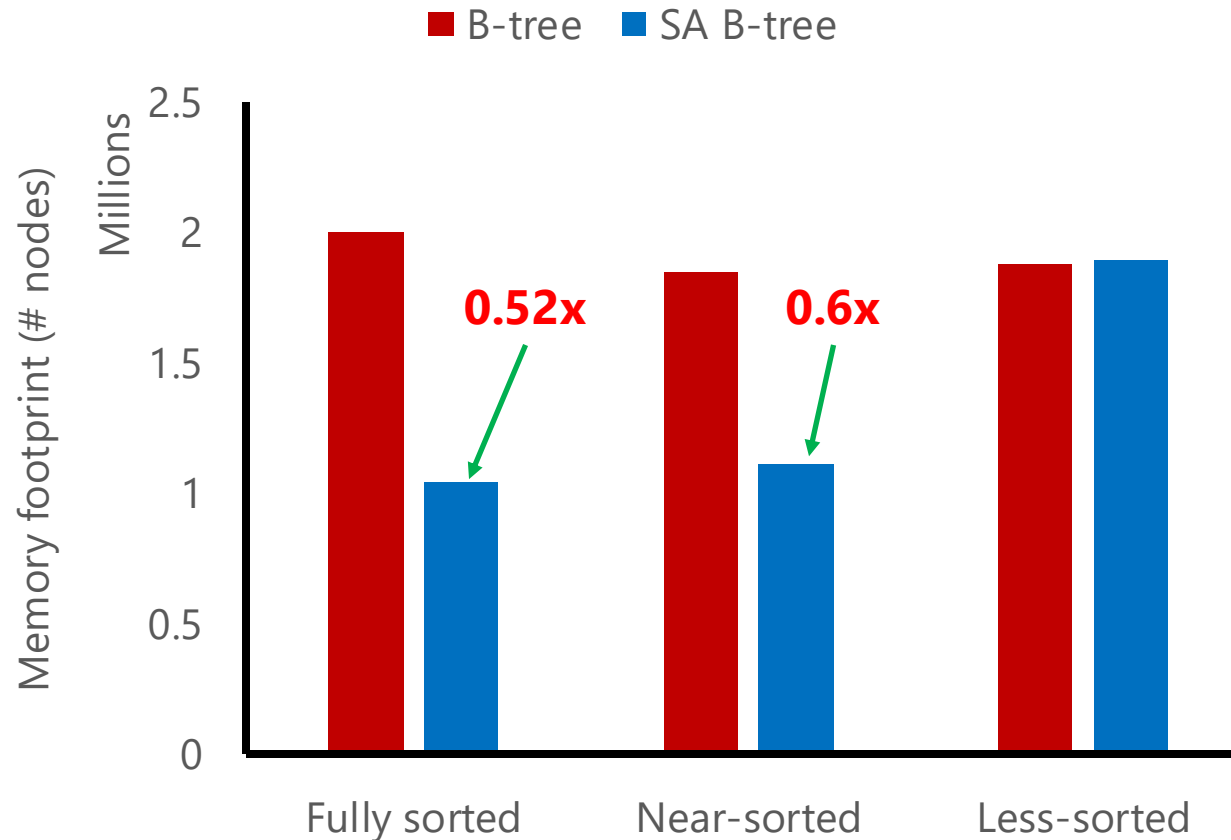


ingestion latency reduced between 27-90%



bulk loading maximized with high data sortedness

SWARE Improves Space Utilization



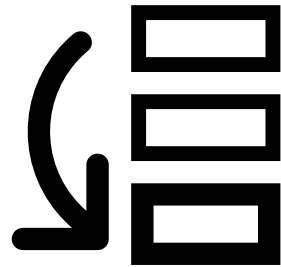
increased fill/split factor helps reduce memory footprint

Summarizing SWARE [ICDE 2023]

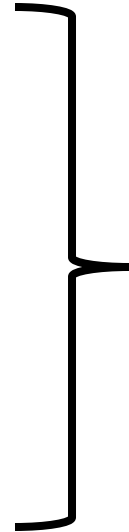


intelligent
buffering

+



opportunistic
bulk loading



Improves performance by
exploiting sortedness

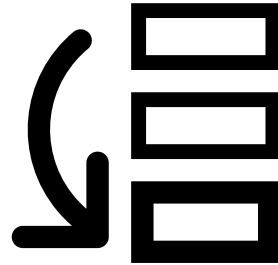
Any downsides to wider applicability?

Summarizing SWARE [ICDE 2023]

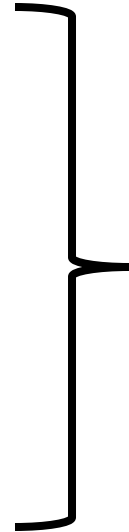


intelligent
buffering

+



opportunistic
bulk loading



Improves performance by
exploiting sortedness

Increases Complexity in Design!

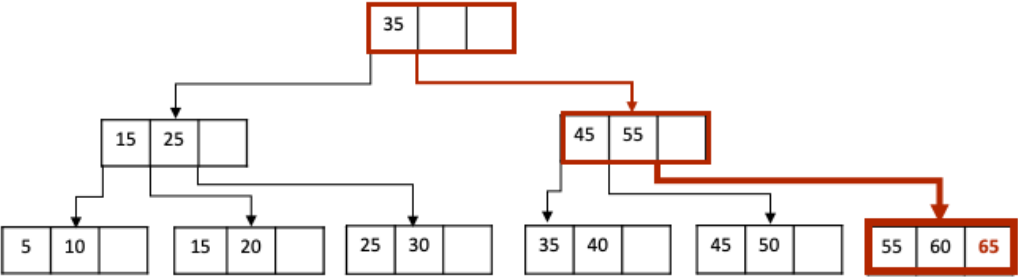
**Can we achieve fast ingestions without
buffering?**

Inserting to the Tail-leaf (PostgreSQL & MySQL)

Normal Insertion (top-insert)



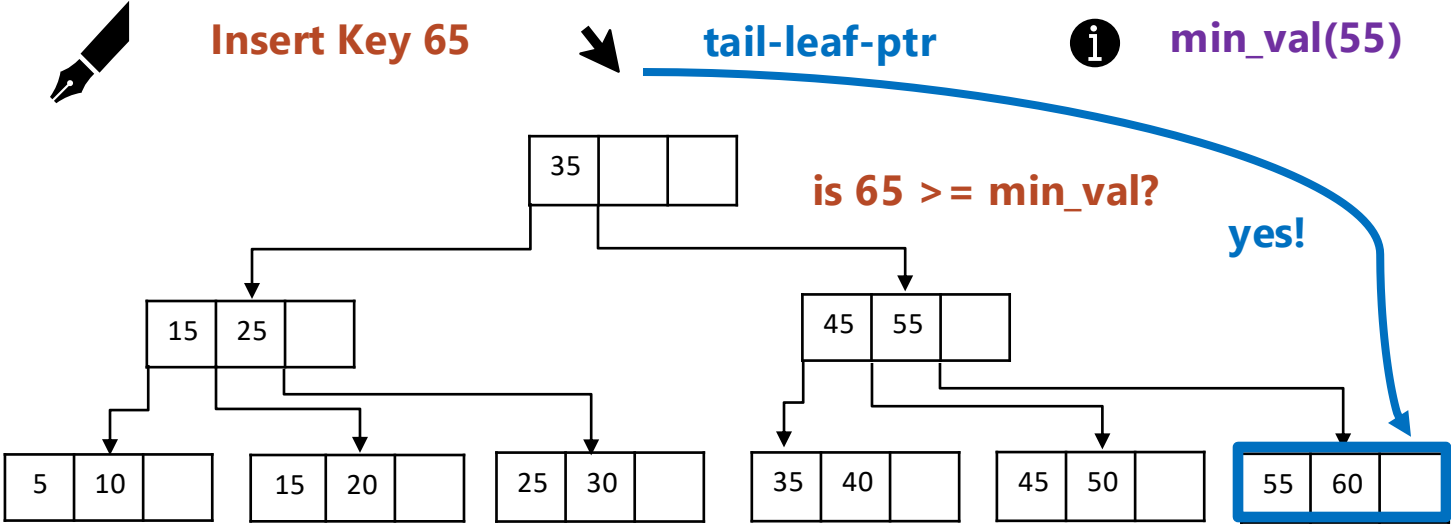
Insert Key 65



Tail-leaf Insertion



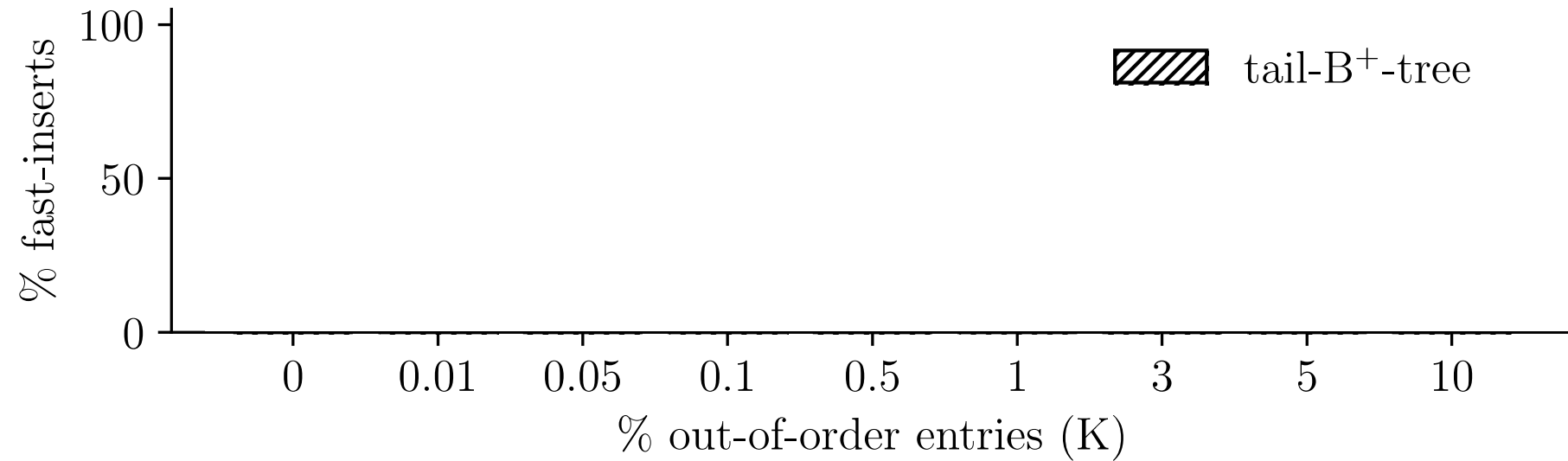
Insert Key 65



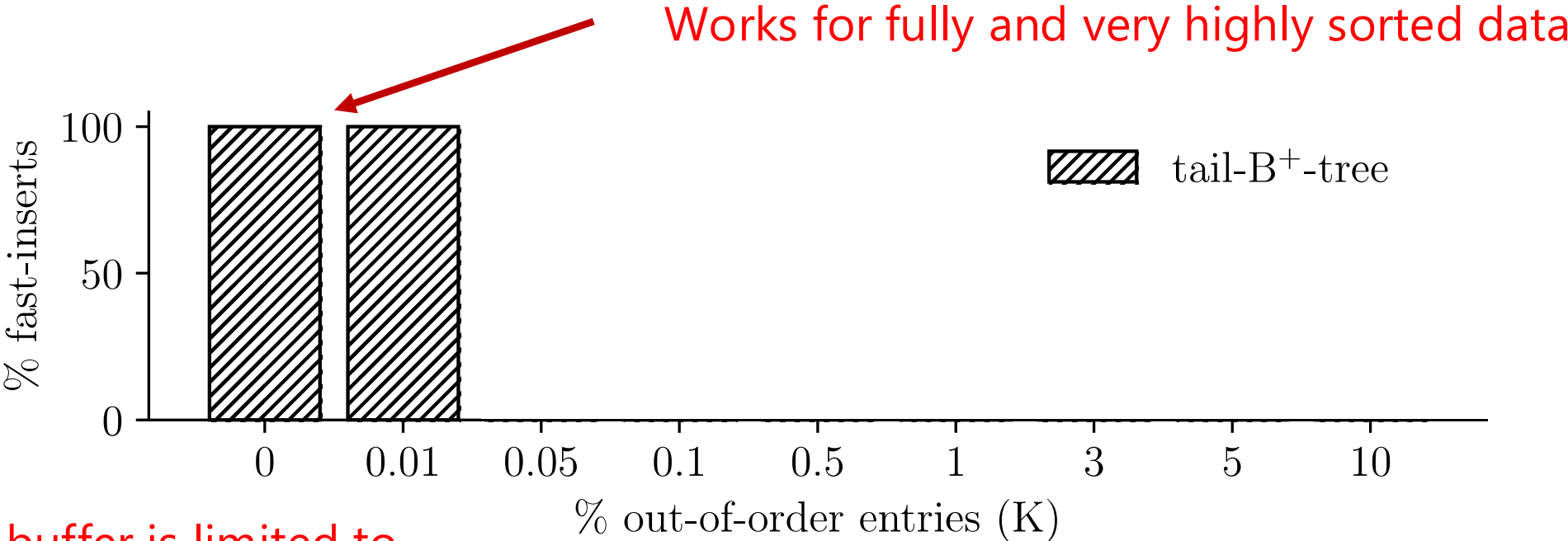
add key to tail leaf directly!

**Is the tail-leaf optimization
the solution?**

Does Tail-leaf Insertion Work?

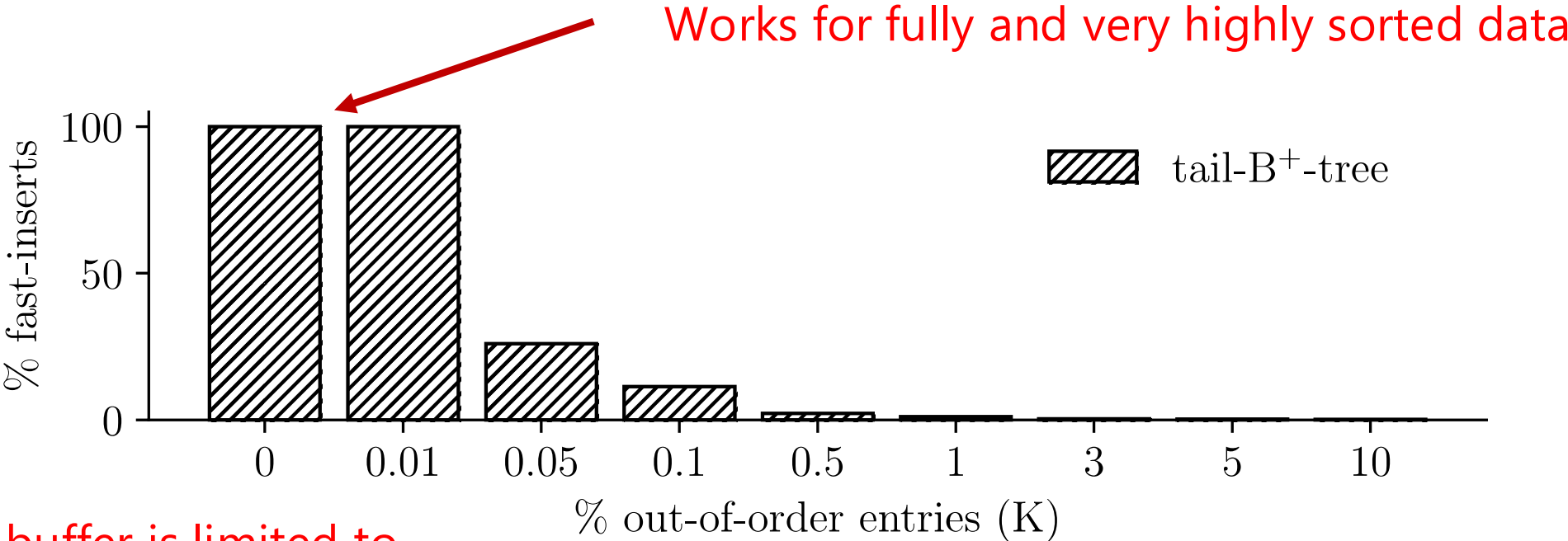


Does Tail-leaf Insertion Always Work?



Tail-leaf's buffer is limited to leaf node!

Does Tail-leaf Insertion Always Work?



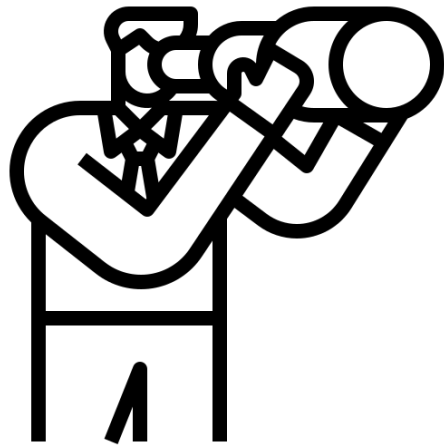
Tail-leaf's buffer is limited to leaf node!



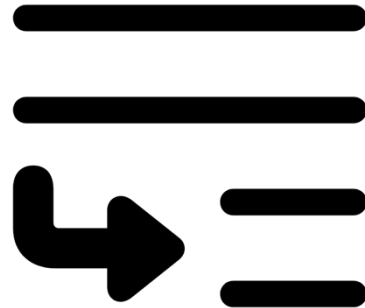
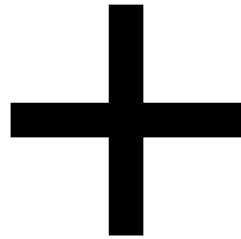
Degrades very quickly

**However, tail-leaf points us to
the right direction...**

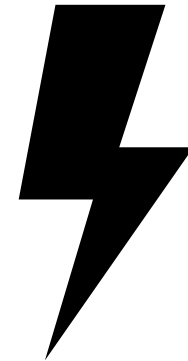
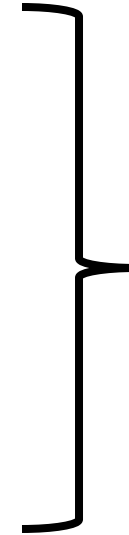
Key Idea – Predicting the Ordered LEaf (POLE)



Sortedness-aware predictor



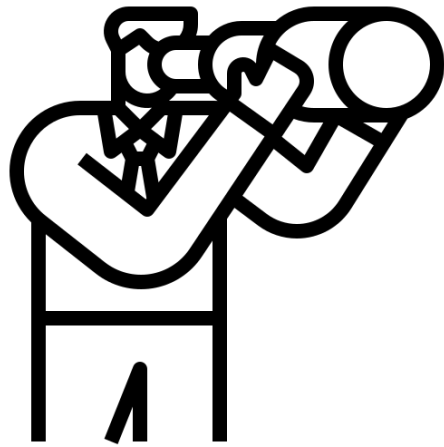
Leaf appends



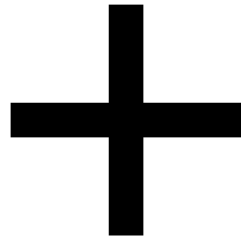
Fast ingestion

Key Idea – Predicting the Ordered LEaf (POLE)

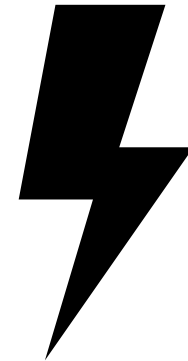
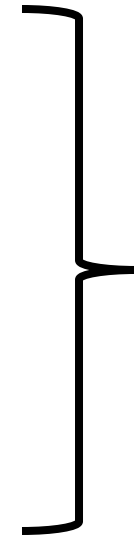
it could be any node!



Sortedness-aware predictor

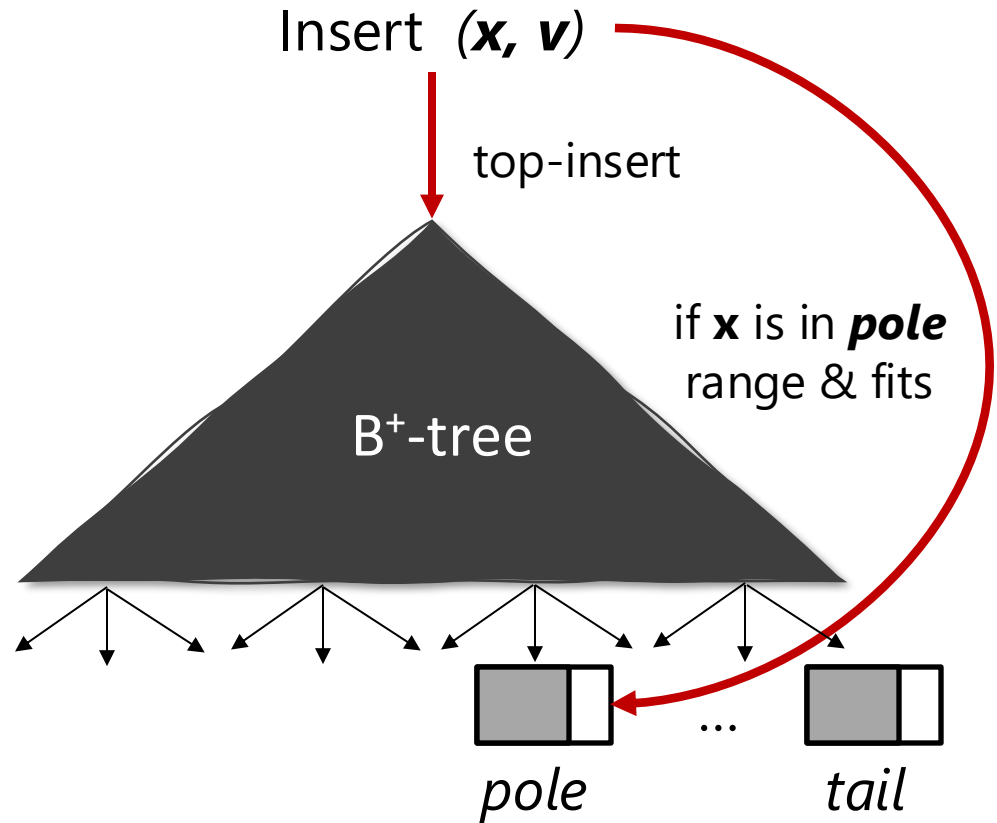


Leaf appends



Fast ingestion

Insertions in Steady-State



When Pole Splits

Legend

p = smallest entry in node previous to $pole$;

q = smallest entry in $pole$

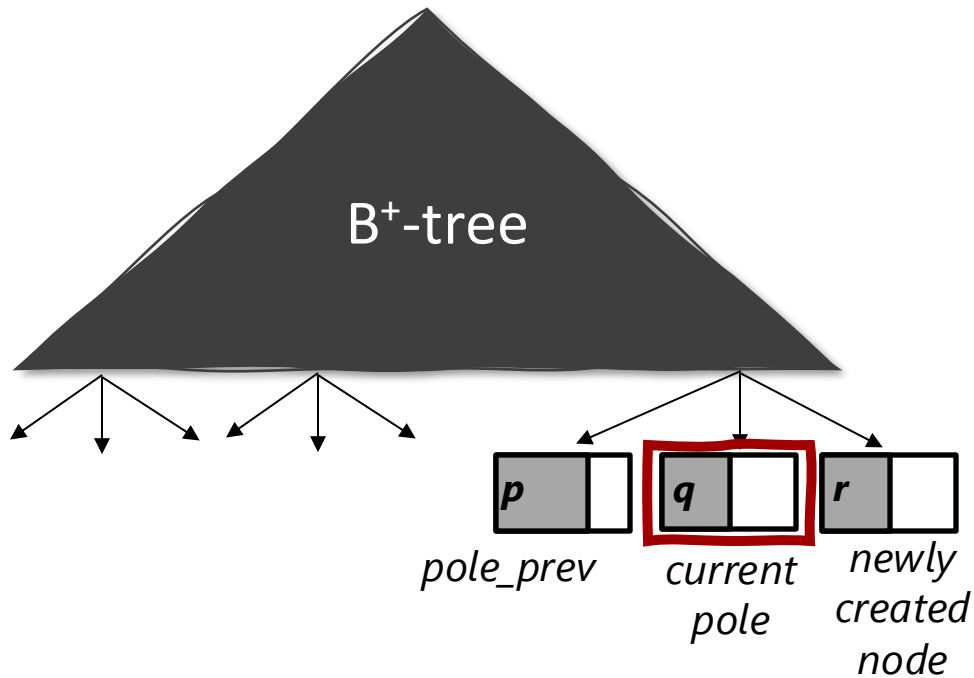
r = smallest entry in newly created node

□ = pointer to $pole$ node

Predict using **IKR (In-order Key estimator)**

$$x = q + \underbrace{\left(\frac{q - p}{pole_prev_size} \right)}_{\text{density between two non-outliers}} \cdot pole_size \cdot (1.5)$$

density between two non-outliers



When Pole Splits

if $r > x$, new node has outliers

Legend

p = smallest entry in node previous to $pole$;

q = smallest entry in $pole$

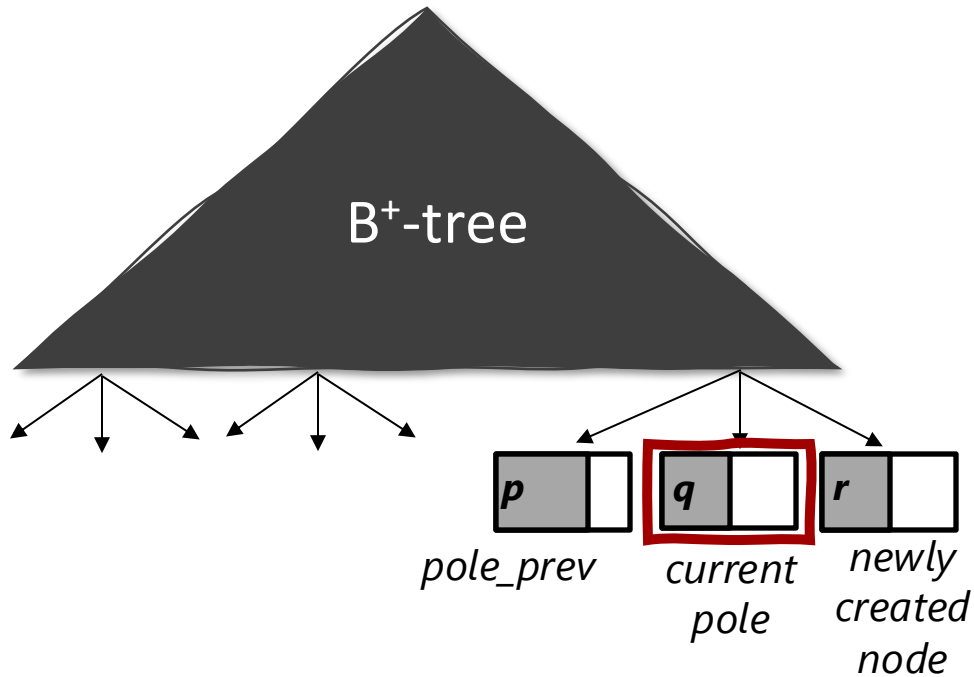
r = smallest entry in newly created node

 = pointer to $pole$ node

Predict using **IKR (In-order key estimator)**

$$x = q + \underbrace{\left(\frac{q - p}{pole_prev_size} \right)}_{\text{density between two non-outliers}} \cdot pole_size \cdot (1.5)$$

density between two non-outliers



pole stays as is

When Pole Splits

if $r \leq x$, new node has at least one non-outlier value

Legend

p = smallest entry in node previous to *pole*;

q = smallest entry in *pole*

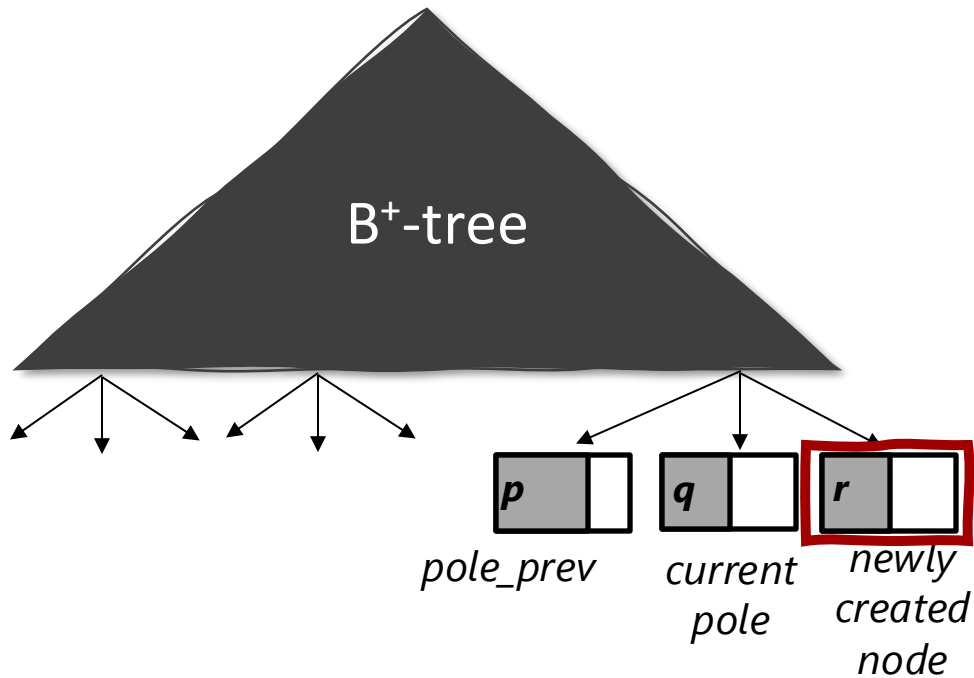
r = smallest entry in newly created node

□ = pointer to *pole* node

Predict using *IKR (In-order key estimator)*

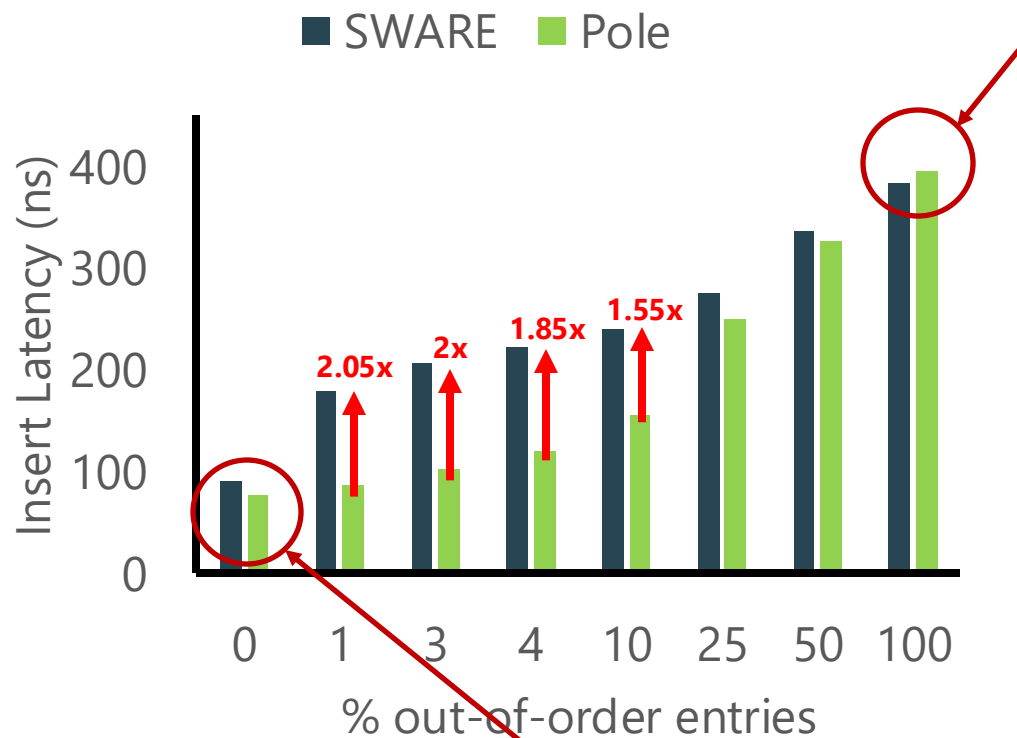
$$x = q + \underbrace{\left(\frac{q - p}{pole_prev_size} \right)}_{\text{density between two non-outliers}} \cdot pole_size \cdot (1.5)$$

density between two non-outliers



Update ***pole*** to newly created node from split

Comparing with SWARE



buffer helps: full bulk loading
Pole is still faster!

Buffer pays off: some vs. none fast ingestion

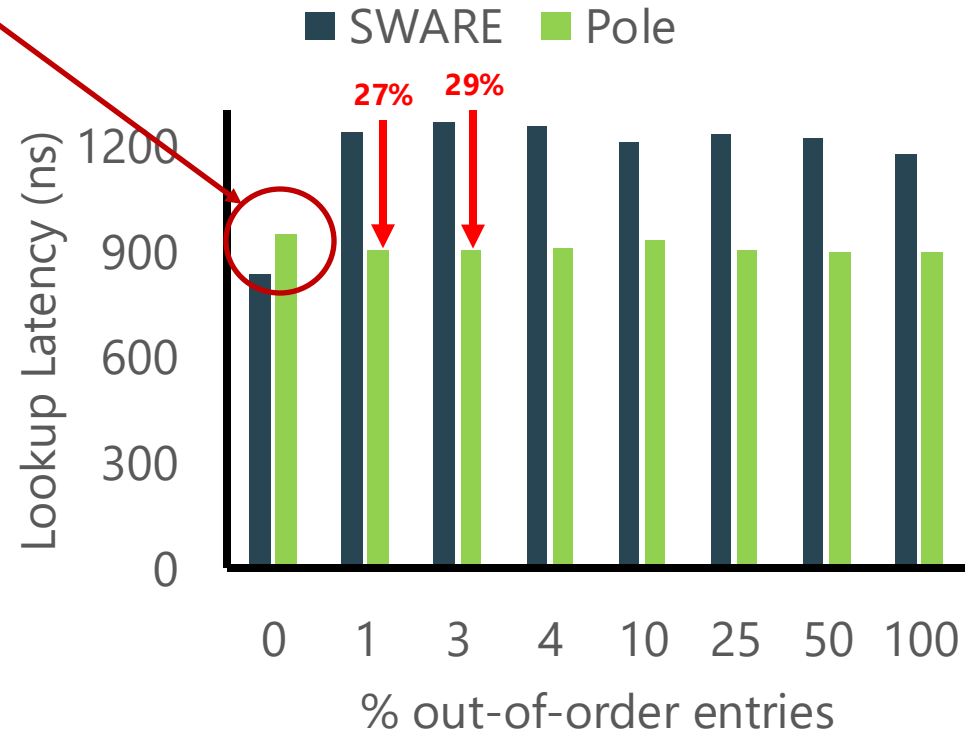
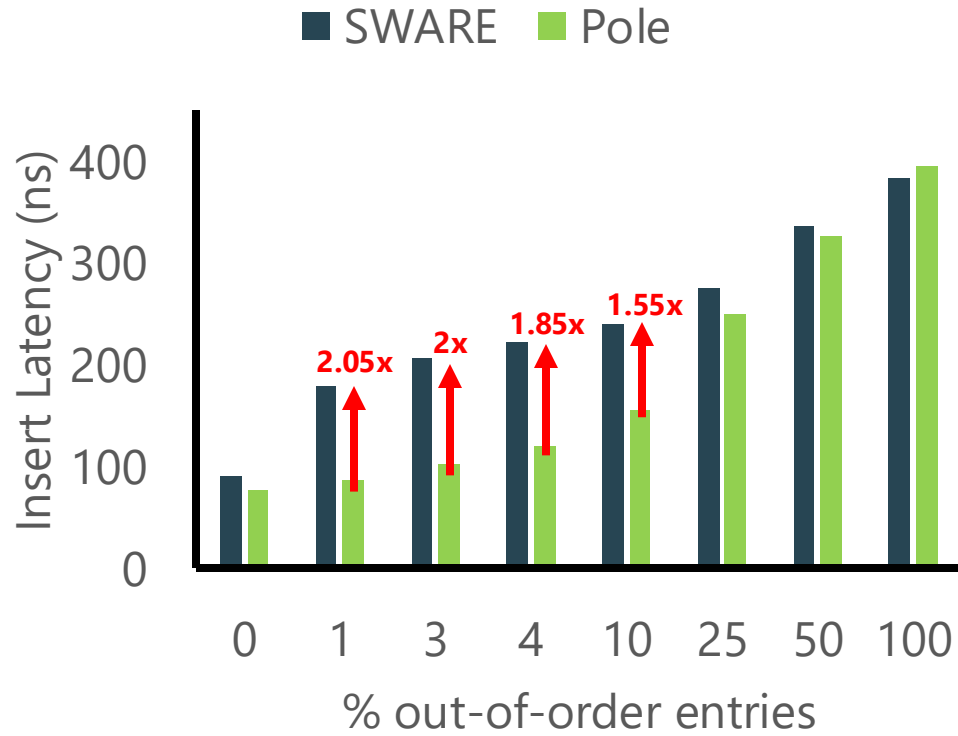
up to 2.05x faster

minimal metadata ✓

avoids SWARE buffer management ✓

Comparing with SWARE

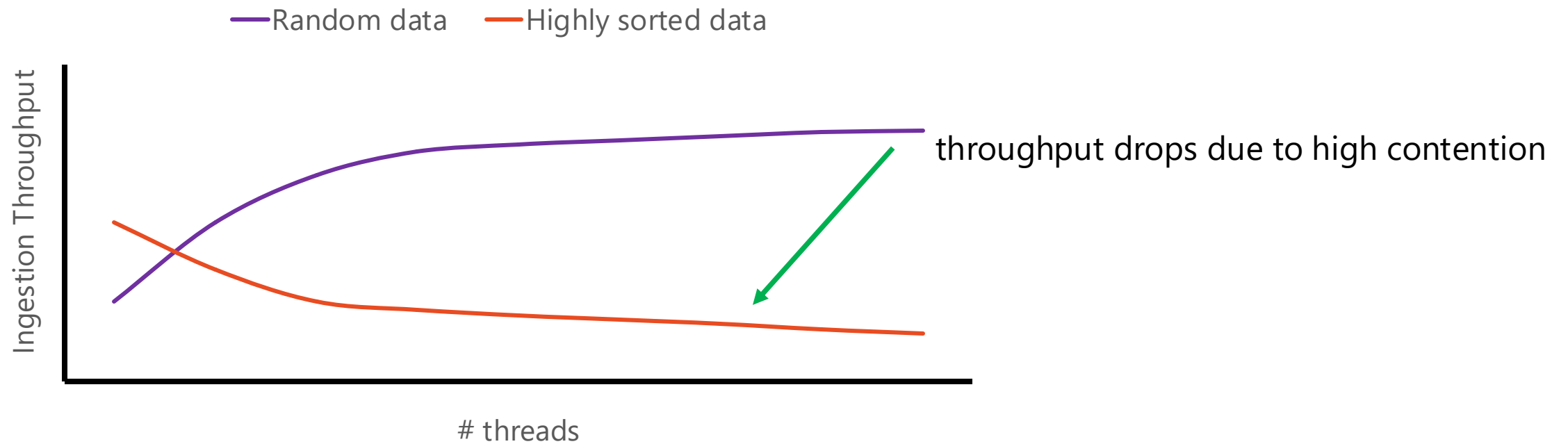
full bulk loading \Rightarrow smaller tree



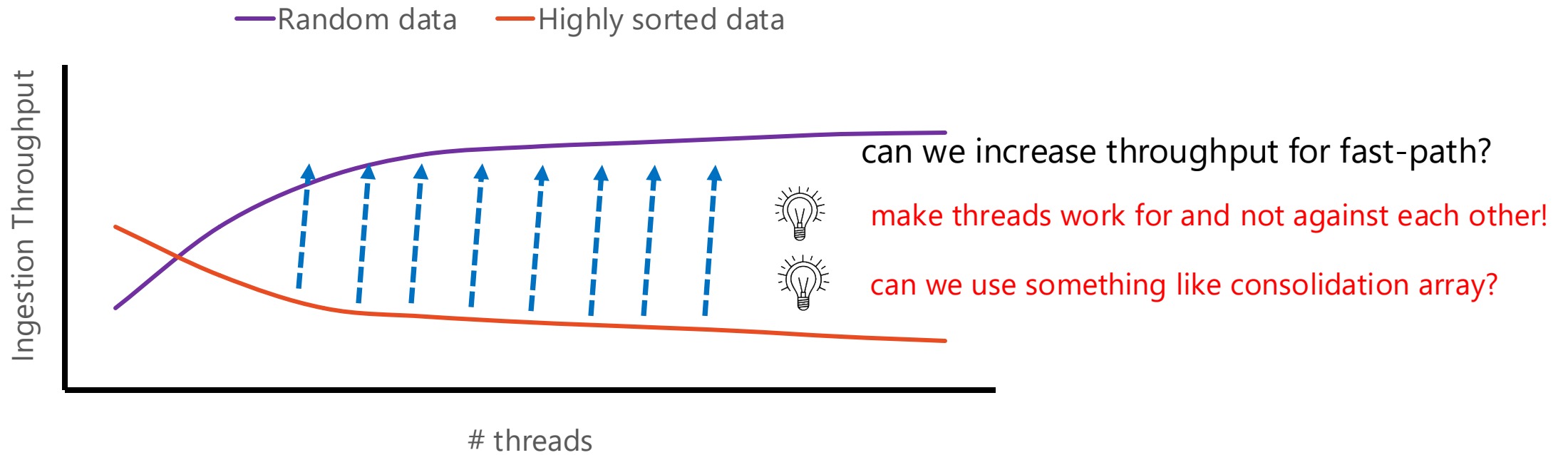
up to 29% faster for point lookups

No buffering \Rightarrow no read overhead!

Future Work - Concurrency in Fast Path



Future Work - Concurrency in Fast Path



Summary

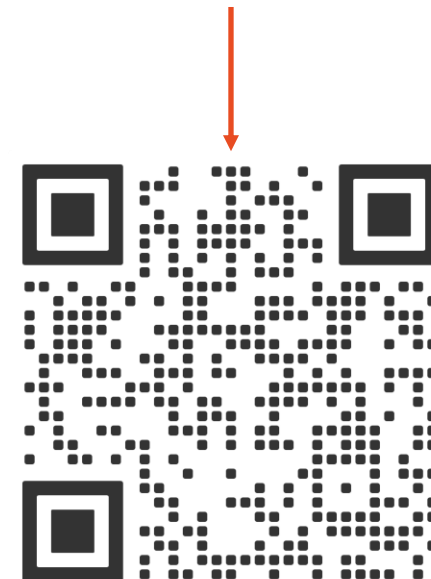
Identify “sortedness” as a resource

Classical indexes do not exploit sortedness by design!

SWARE paradigm & Pole optimization optimize for sortedness

Further research required for learned indexes + joins

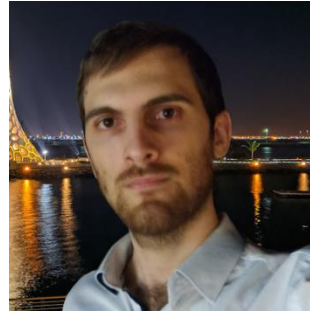
Scan here to learn
more about our work



The Team



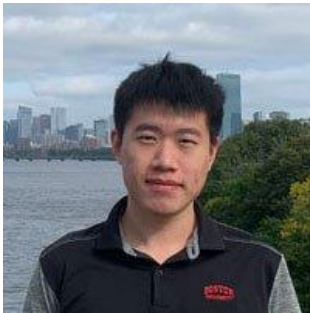
Aneesh Raman



Konstantinos
Karatsenidis



Andy Huynh



Jinqi Lu



Shaolin Xie



Subhadeep Sarkar



Matthaios Olma

