

NOCAP: Near-Optimal Correlation-Aware Partitioning for Joins

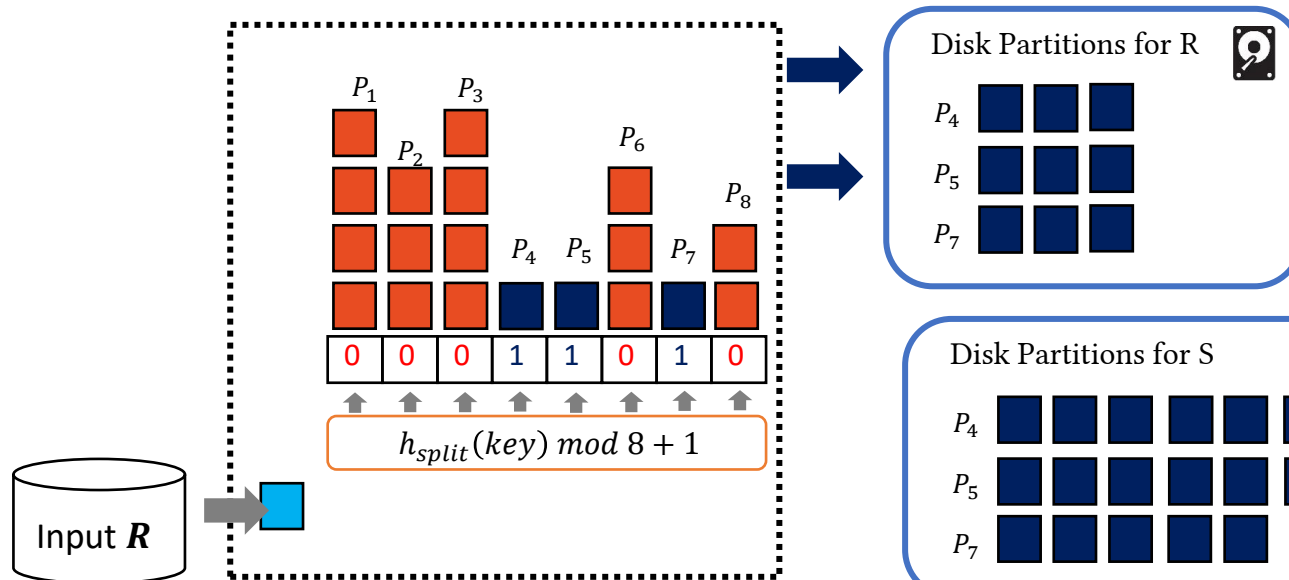
Zichen Zhu Xiao Hu Manos Athanassoulis

Dynamic Hybrid Hash Join (DHH)

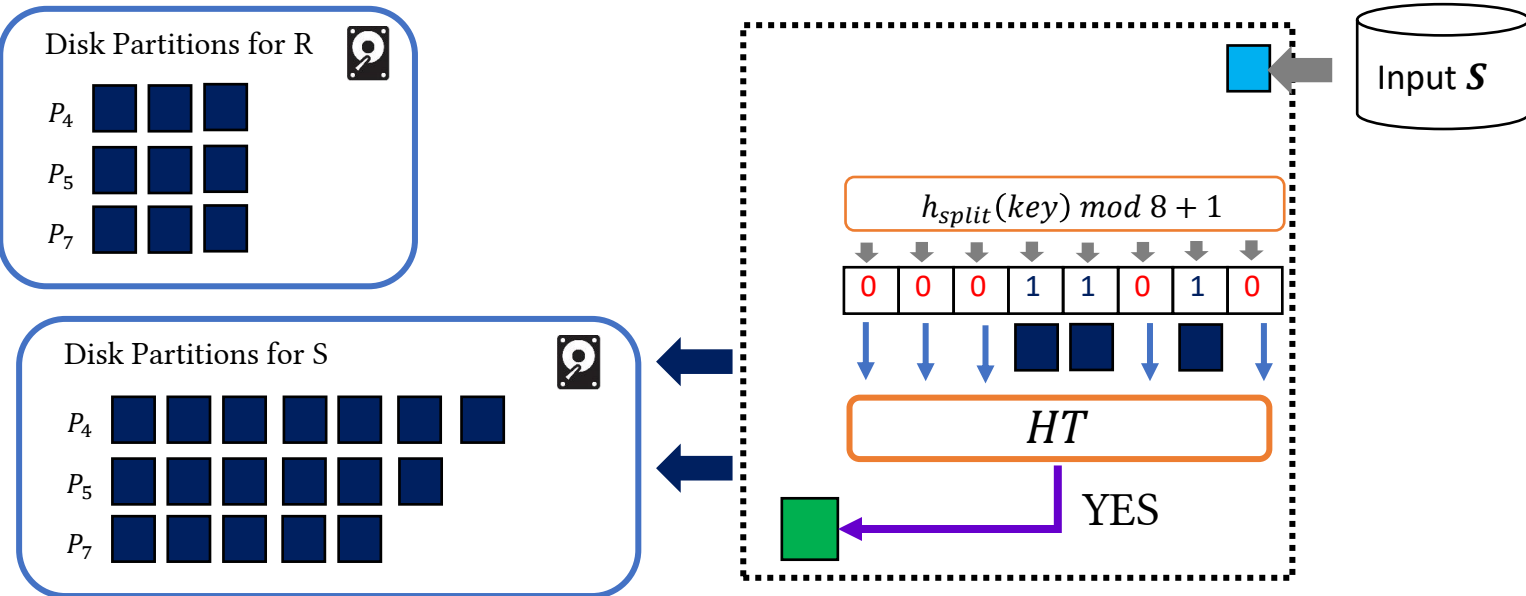


State of the art DBs (e.g., PostgreSQL and AsterixDB) employ DHH to implement storage-based equij-joins.

Partition R



Partition S

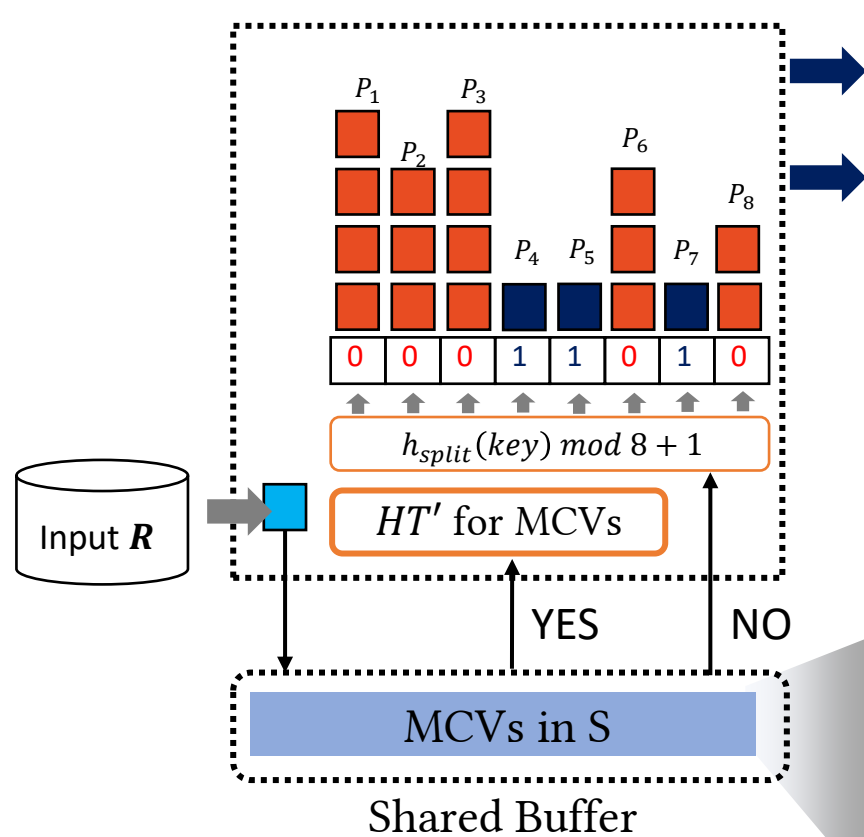


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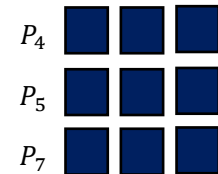


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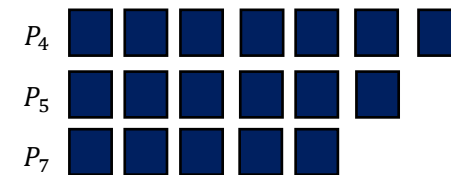
Partition R



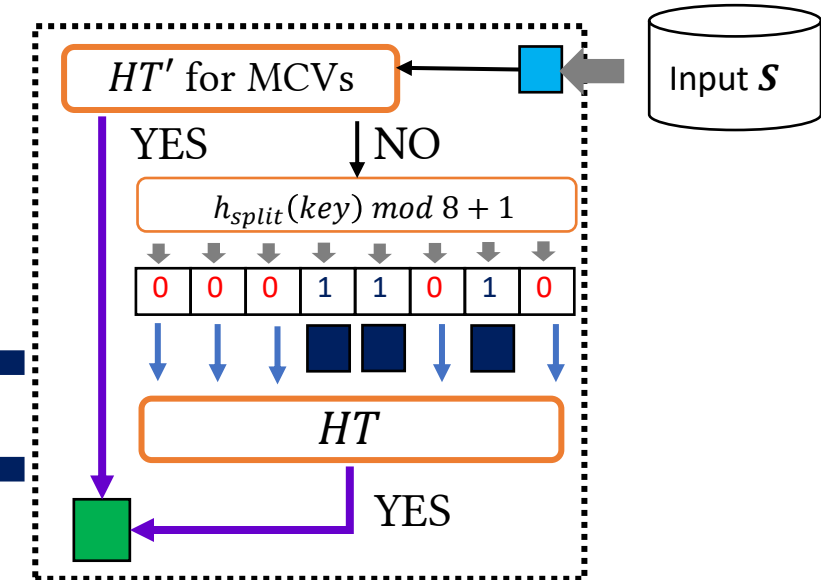
Disk Partitions for R



Disk Partitions for S



Partition S

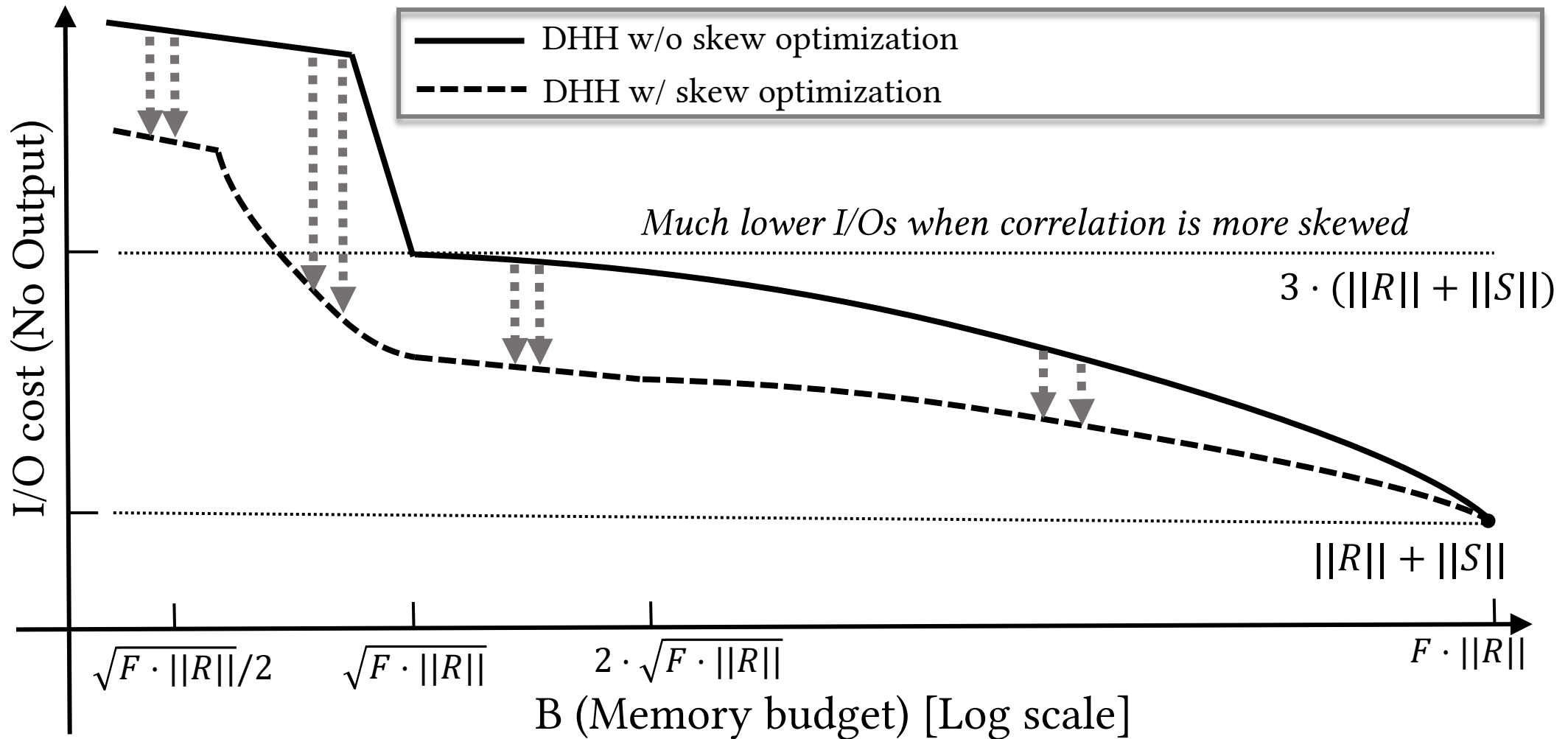


Key	Frequency
key1	1000
...	...
keyn	500

Correlation Table (CT)

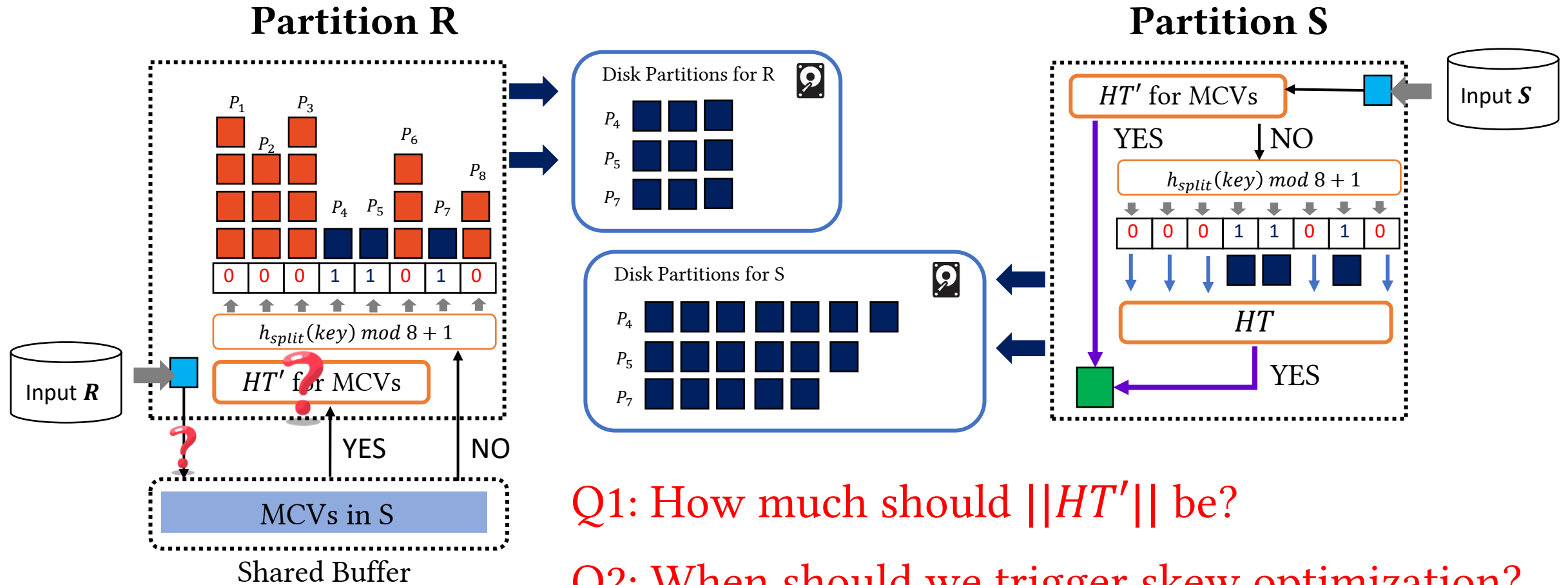
Skew Optimization

Skew Optimization in DHH



Skew optimization reduces the number of I/Os when the matching exhibits skew

Problems



OCCAP (Optimal Correlation-Aware Partitioning)

$$\arg \min_{P,m} \sum_{j=2}^{m+1} \text{Cost}(\|R_j\|, \|S_j\|)$$

$$\text{s.t. } \forall i \in [n], \sum_{j=1}^{m+1} P_{i,j} = 1$$

$$\|R_1\| + m + 2 \leq B$$

$$P_{i,j} \in \{0,1\}, \forall i \in [n], \forall j \in [m+1]$$

$$\|R_j\| = \sum_{i=1}^n P_{i,j} / b_R$$

$$\|S_j\| = \sum_{i=1}^n P_{i,j} \cdot CT[i] / b_S$$

Correlation Table (CT)

Index i	Frequency in S
1	1
...	...
n-1	77
n	100

Define an $n \times (m+1)$ Boolean matrix P to represent the partitioning assignment

Notation	Meaning
n (n_R)	The number of tuples in relation R
m	The number of partitions on disk
$P = \begin{bmatrix} 0 & \cdots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \cdots & 0 \end{bmatrix}_{n \times (m+1)}$	A Boolean matrix P where $P_{i,j} = 1$ represents the i^{th} record belongs to the j^{th} partition
R_1	A partition cached in memory
$\text{Cost}(\ R_j\ , \ S_j\)$	The write cost of R_j, S_j plus the join cost between R_j, S_j

Exponential searching space !



$O(n^2 \cdot \log B / B)$ for PK-FK joins

Practical Challenges for OCAP

1. *We cannot have the whole CT in practice*

Index i	Frequency in S
1	1
...	...
10M	1000

2. *Partitioning assignment also occupies memory*

$$P = \begin{bmatrix} 0 & \cdots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \cdots & 0 \end{bmatrix}_{n \times (m+1)}$$

Observations from OCAP

O1: MCVs can be prioritized in *two* ways: build an in-memory hash table (if B is large) or **assign them into a small partition on disk (if B is small)**

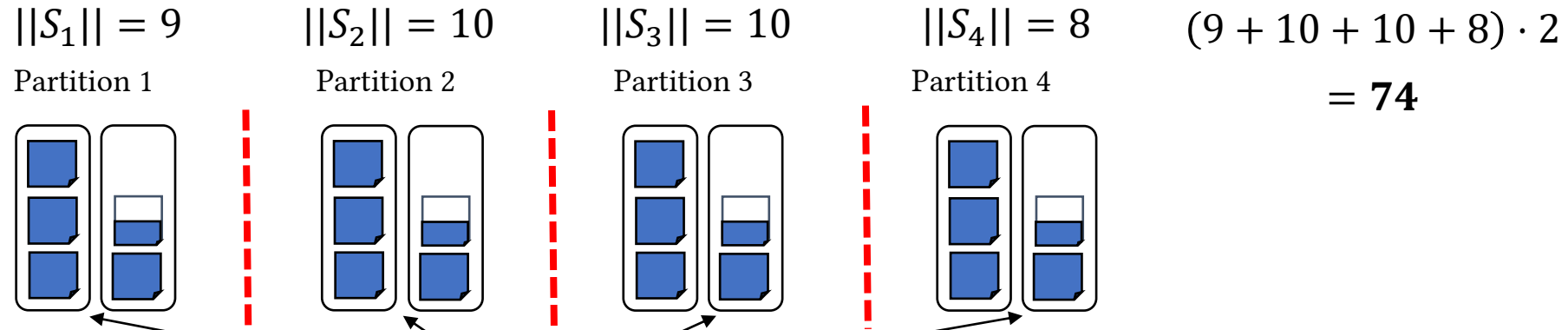
O2: We should ensure on-disk partitions fully fill $z \cdot (B - 2)$ pages ($z \in \mathbb{Z}^+$)

Divisibility when Partitioning Data on Disk



uniform

$$PartID = h(k) \bmod 4$$

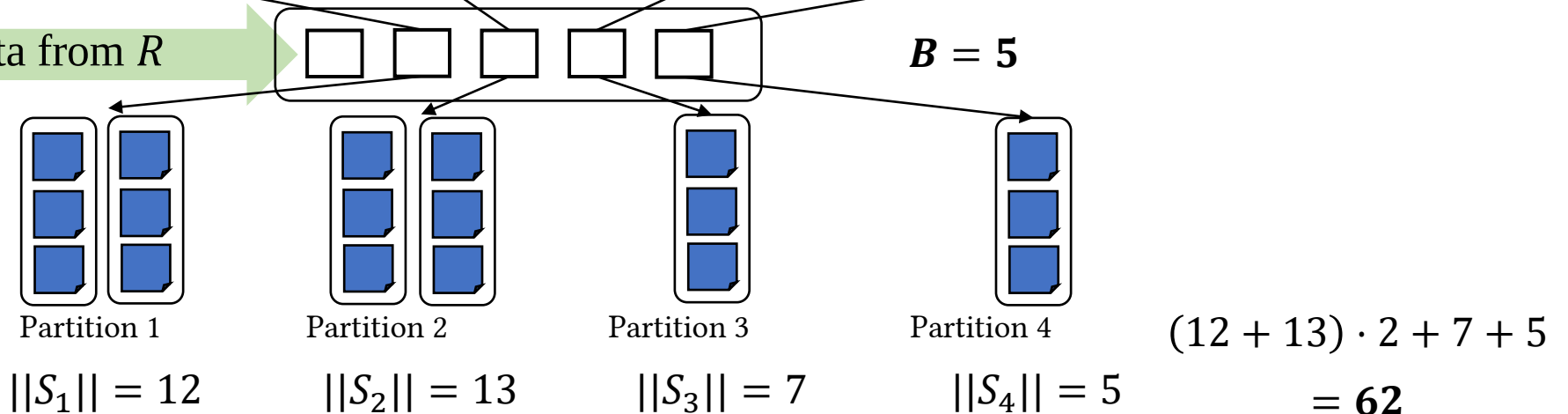


18-page input data from R

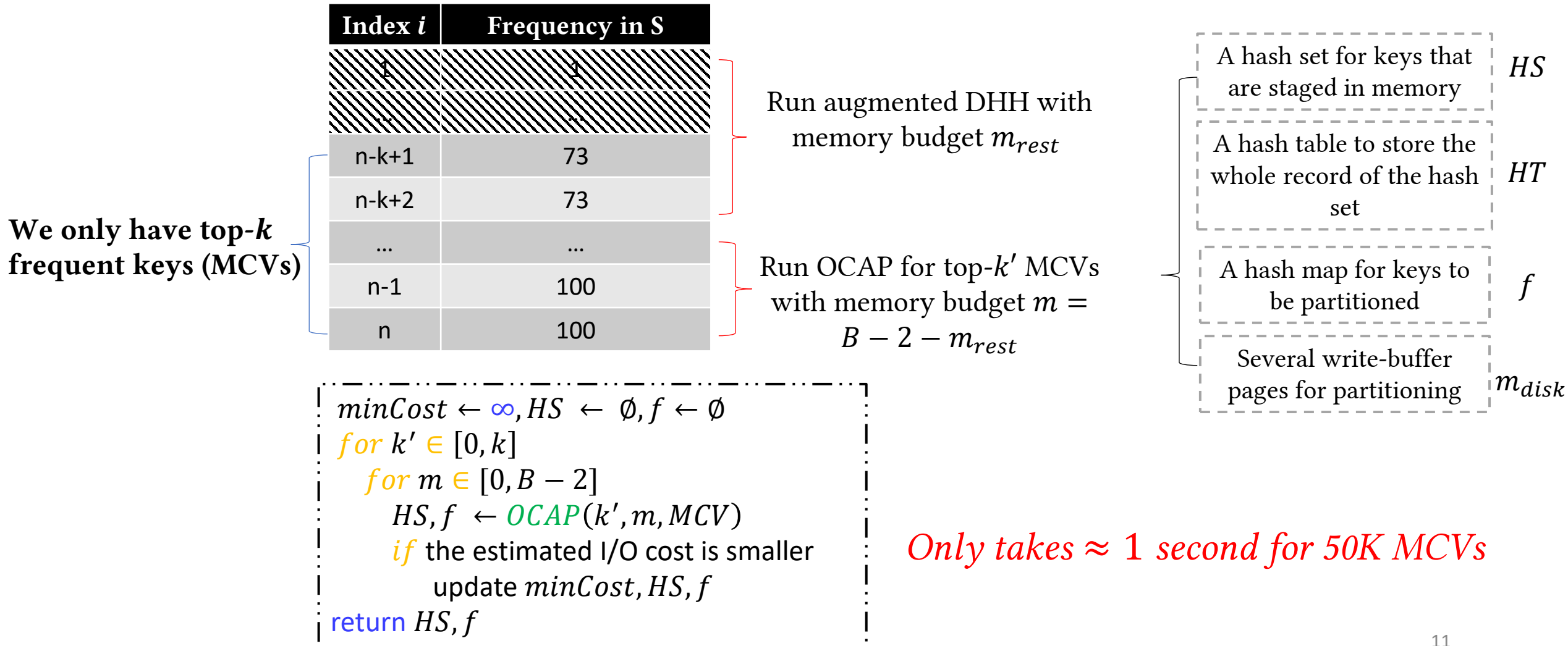
$$PartID = (h(k) \bmod 6) \bmod 4$$

non-uniform

$$6 = ||R|| / (B - 2)$$

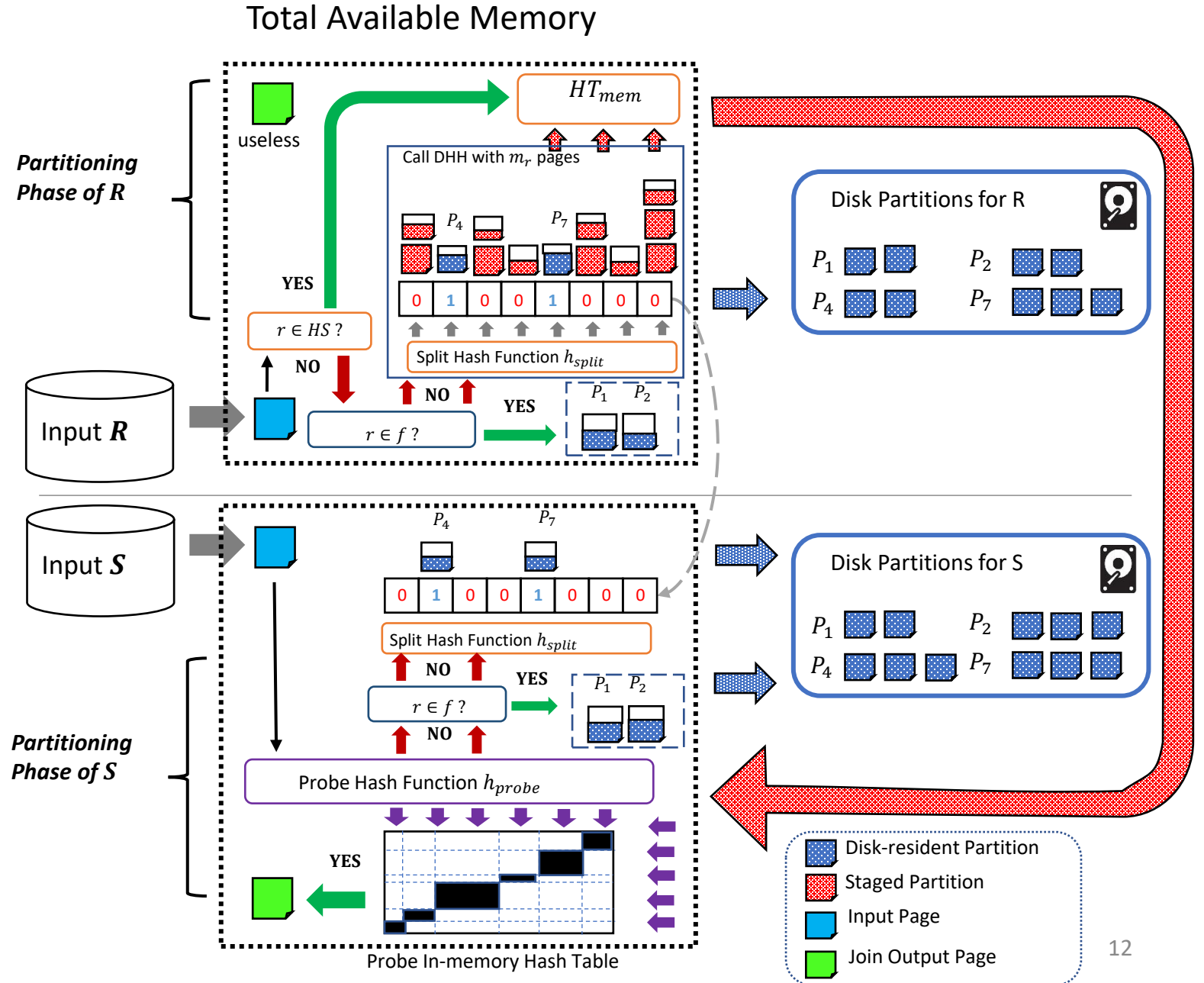


Prioritizing MCVs with Constrained Memory



NOCAP

Partitioning Workflow:
OCAP for top- k' frequent keys
DHH to partition the rest



Experiment Setup

Storage: PCIe NVM SSD (15 μs for reading a 4KB page)

Measured read/write symmetry:

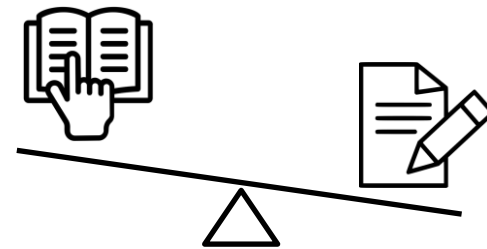
$$\text{random_write_latency/sequential_read_latency} = 3.3$$

$$\text{sequential_write_latency/sequential_read_latency} = 3.2$$

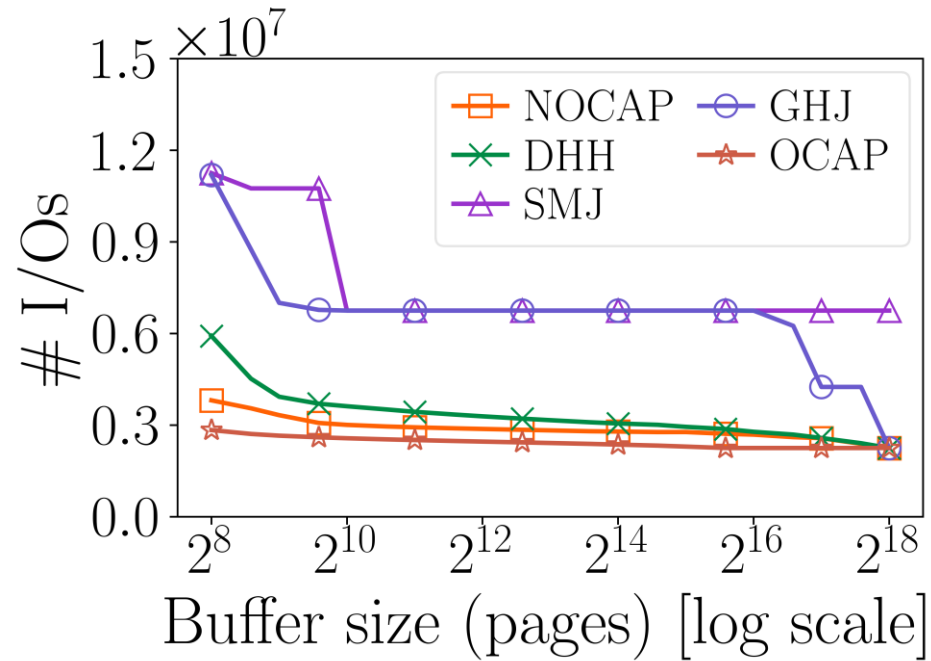
PK-FK join input size: 1M #records join with 8M #records

Record size: 1KB per record

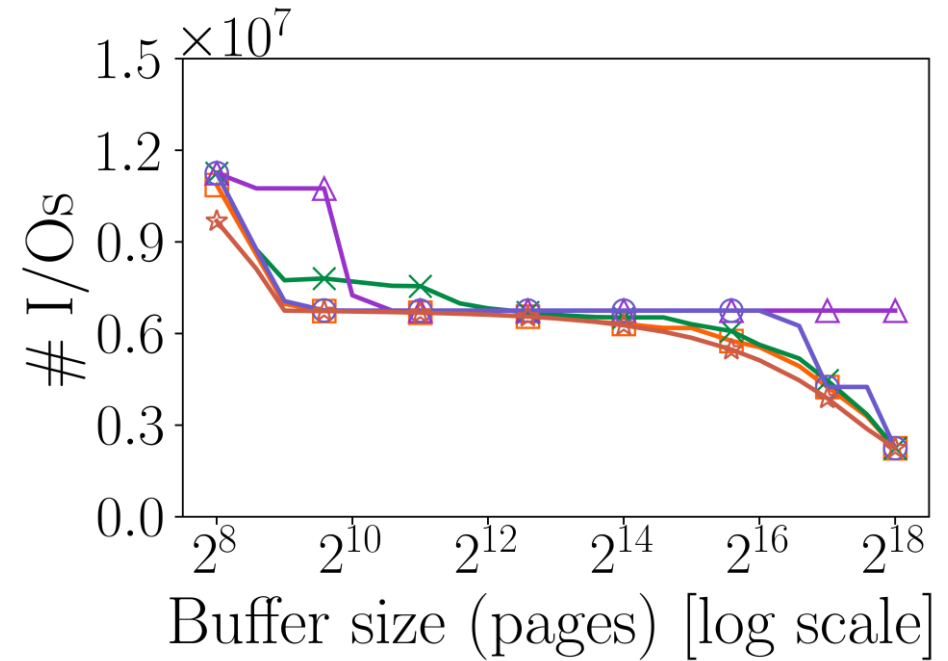
Page size: 4KB



Selected Experimental Results



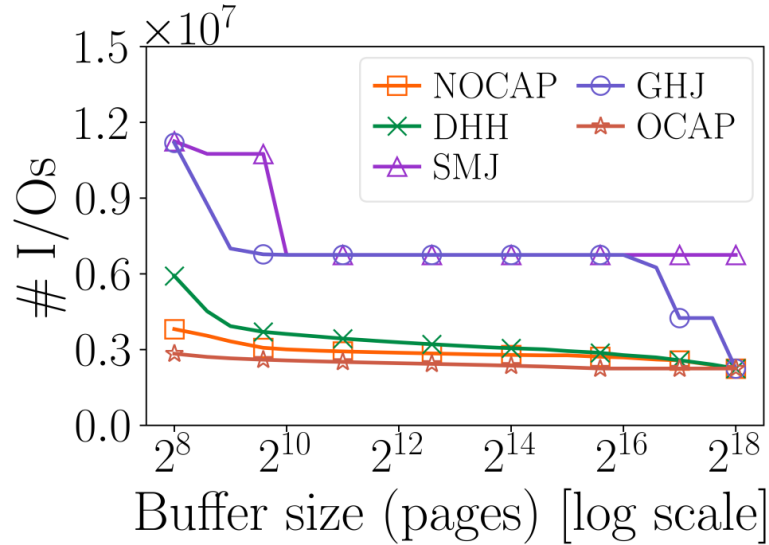
Zipfian ($\alpha = 1.3$)



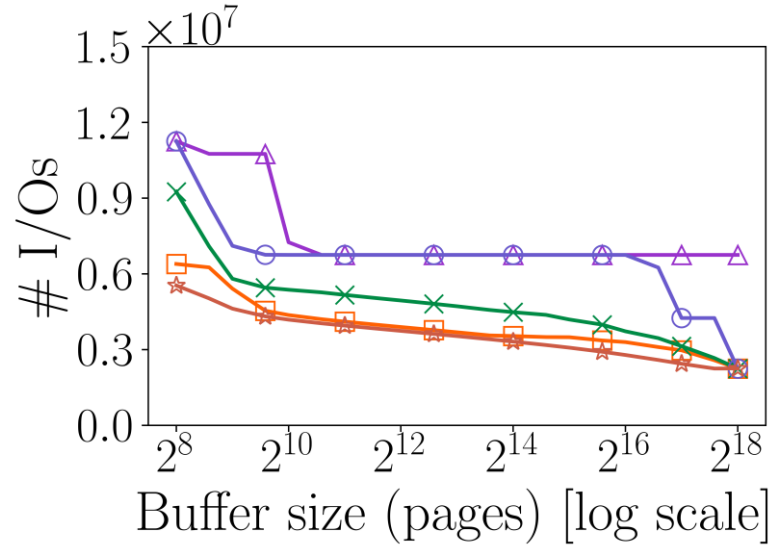
Uniform

Correlation-aware joins (**DHH and NOCAP**) can **adaptively** reduce I/O cost when it comes to a **skew** distribution.

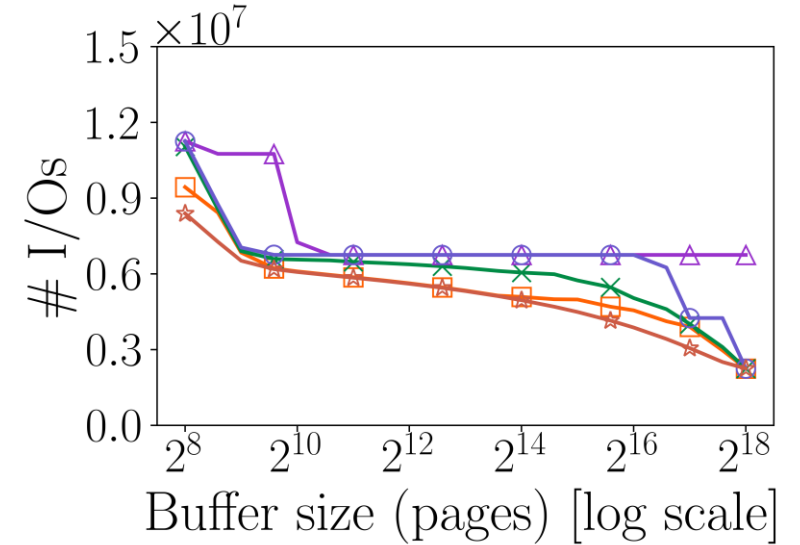
Varying skew



Zipfian ($\alpha = 1.3$)



Zipfian ($\alpha = 1.0$)



Zipfian ($\alpha = 0.7$)

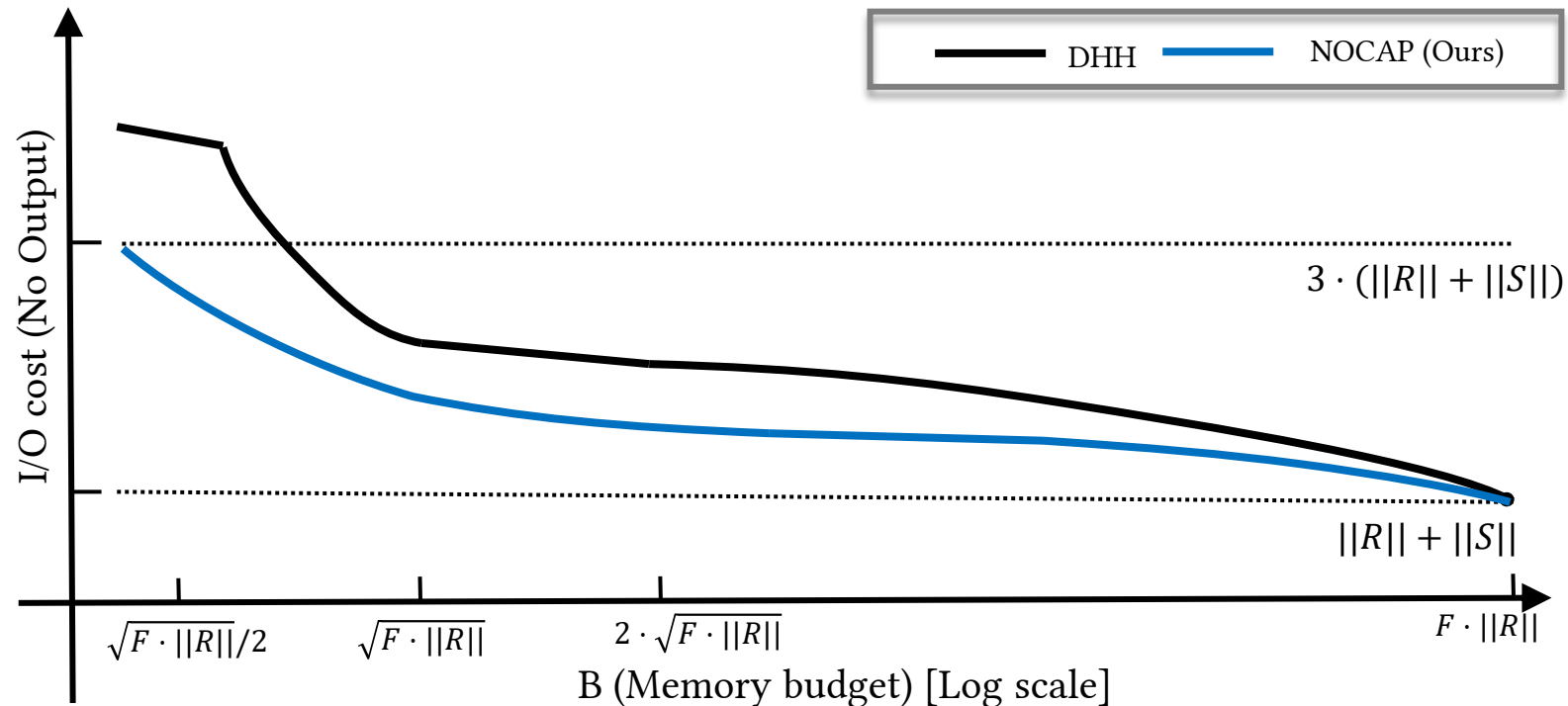
High skew

Low skew

While DHH helps reduce #I/Os, **NOCAP** can better exploit the correlation skew to **achieve even lower I/O cost**.

Summary of NOCAP

NOCAP join outperforms DHH by up to 30%, and the textbook GHJ by up to 4X. Even for uniform distribution, NOCAP outperforms DHH by up to 10%!



Thanks

Q&A