

# Optimizing Recursive Joins in Graph Database Management Systems

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# Outline

- Background on Graph DBMS & Recursive Joins
- Why are Recursive Joins challenging ?
- Query Processing for Recursive Joins
- Future Work

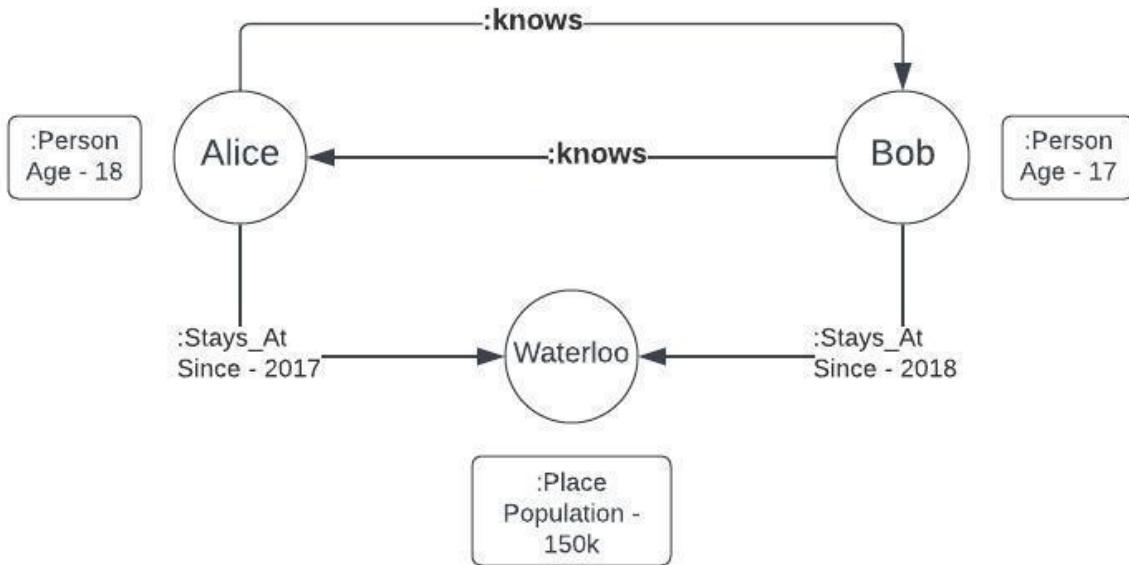
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# Background on Graph Databases

## Data Model

- Labeled Property Graph (LPG)



## Query Language

Neo4j  
Cypher Query  
Language



```
MATCH
(a:Person)-[r1:Stays_At] ->(b:Place)<-[r2:Stays_At] - (c:Person)
RETURN b.Population;
```

- Express subgraph pattern for Pattern Matching
- Express recursive queries for Graph Path Traversal

# Background on Recursive Joins

## Core competency of GDBMS compared to RDBMS

(1) Easier to express in the query language of GDBMS:

**Query:** *Return all people ‘Alice’ knows directly / **indirectly** and the path length between them*

Cypher:

```
MATCH p = (p1:Person)-[:knows* SHORTEST 1..30]->(p2:Person)  
WHERE p1.name = ‘Alice’ RETURN p2, length(p)
```

Harder to express in recursive SQL.

# Background on Recursive Joins

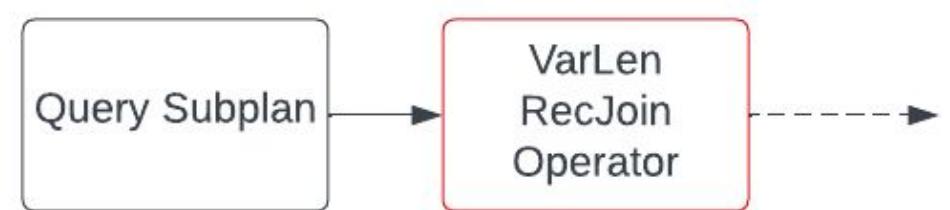
Core competency of GDBMS compared to RDBMS

(2) Also often faster to execute in GDBMS

*GDBMS have **specialized recursive join operators***

**Query:**

```
MATCH p = (a:Person)-[r:knows* 1..30]->(b:Person)  
WHERE a.name = "Alice"  
RETURN a.ID, b.ID, length(p)
```

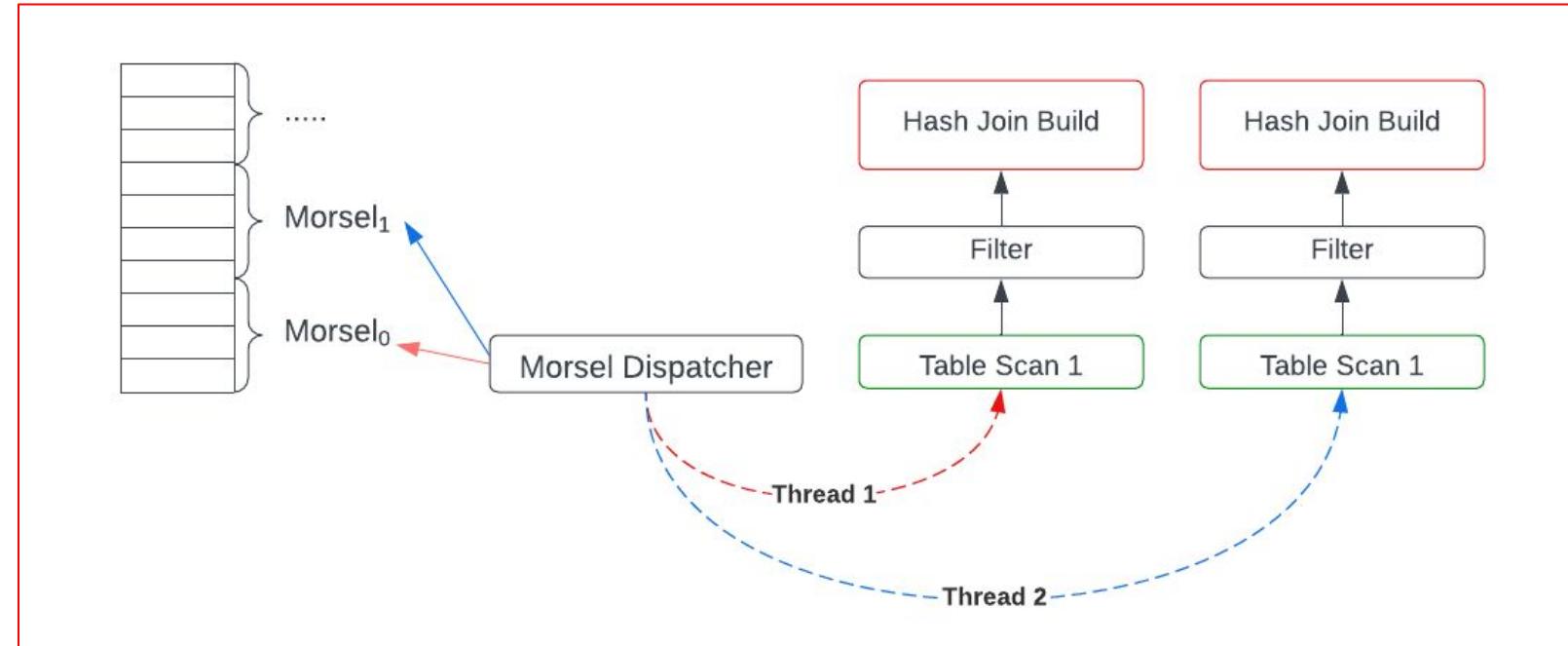


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# SoTA Approach in Analytical DBMS: Morsel-driven Parallelism

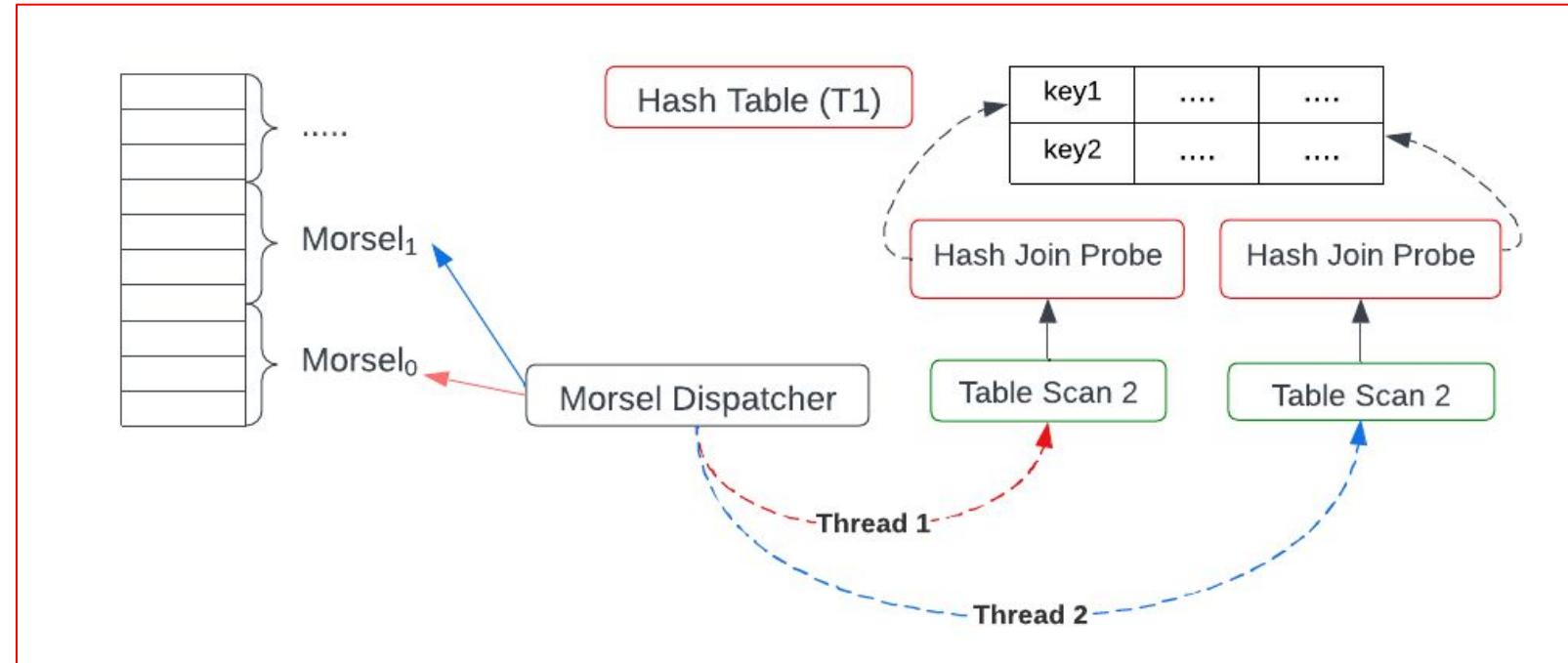
- allot fixed size “morsels” to threads (1024 - 2048 - 100,000 tuples)
- threads execute on their morsels for 1 pipeline until the pipeline breaker



\* *Morsel-Driven Parallelism: A NUMA-Aware Query Evaluation Framework for the Many-Core Age*, Leis et al.

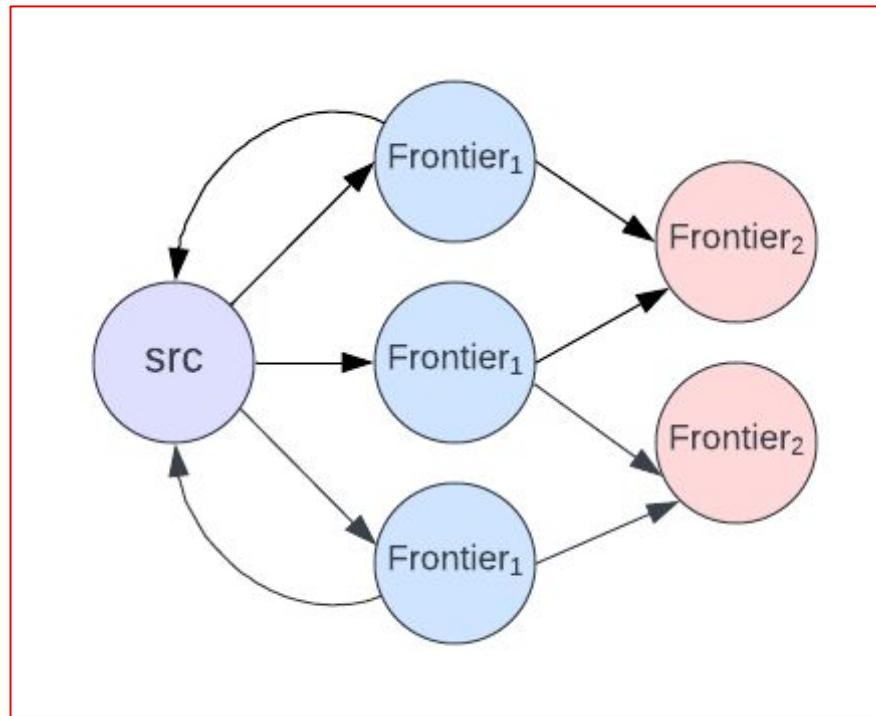
# SoTA Approach in Analytical DBMS: Morsel-driven Parallelism

- allot fixed size “morsels” to threads (1024 - 2048 - 100,000 tuples)
- threads execute on their morsels for 1 pipeline until the pipeline breaker
- morsel dispatcher allots other morsels to threads, after completion of previous pipeline



# Problem with Morsel-Driven Parallelism for Recursive Joins

1. Recursive Join operators find variable length / shortest path from a single source node. Usually involve some form of BFS style traversal.



# Problem with Morsel-Driven Parallelism for Recursive Joins

1. Recursive Join operators find variable length / shortest path from a single source node.

- Most real world graphs display small world network property
- 5 or 6 steps may “traverse” the entire database
- *This makes recursive joins, even from 1 source very expensive*

## Example:

```
MATCH p = (p1:Person)-[:knows* SHORTEST 1..30]->(p2:Person)
WHERE p1.name = 'Alice' RETURN p2, length(p)
```

# Problem with Morsel-Driven Parallelism for Recursive Joins

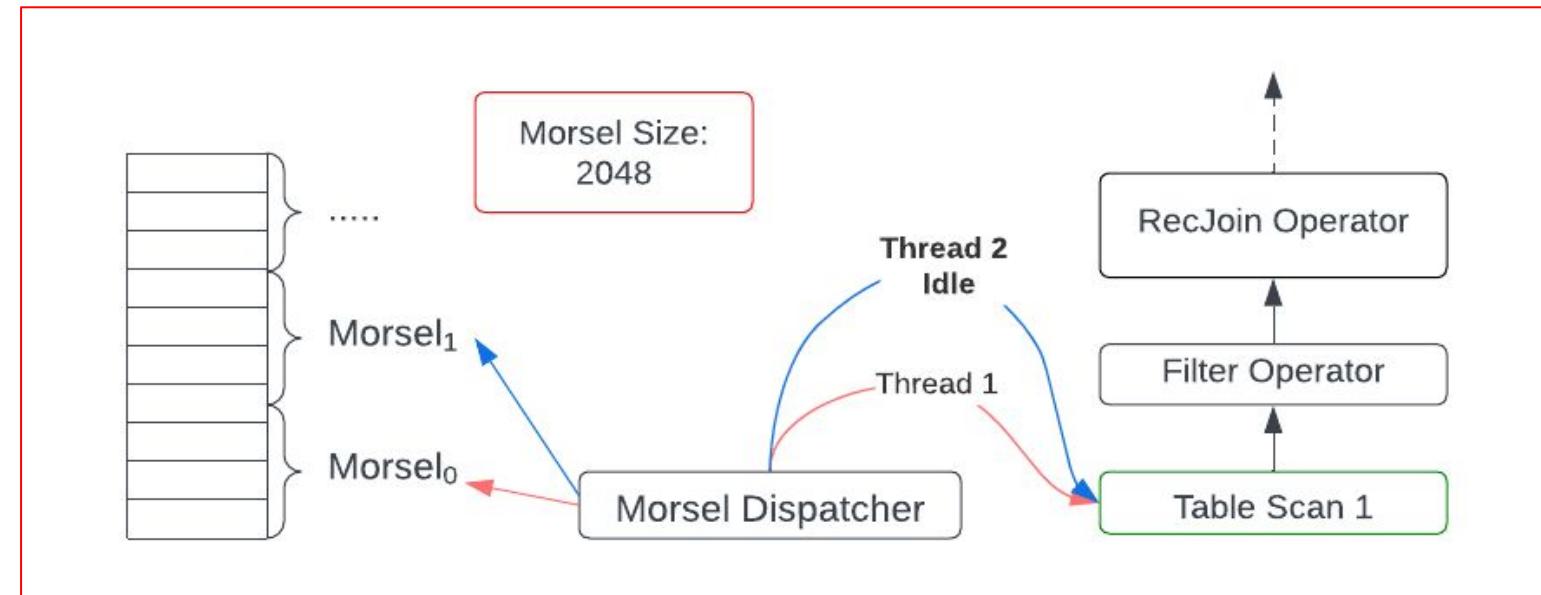
2. Dispatcher may allot a morsel with disproportionate no. of sources to a single thread

# Problem with Morsel-Driven Parallelism for Recursive Joins

2. Dispatcher may allot a morsel with disproportionate no. of sources to a single thread

## Example:

```
MATCH
p = (p1:Person)-[:knows* SHORTEST
1..30]->(p2:Person)
WHERE p1.ID < 50 RETURN length(p)
```



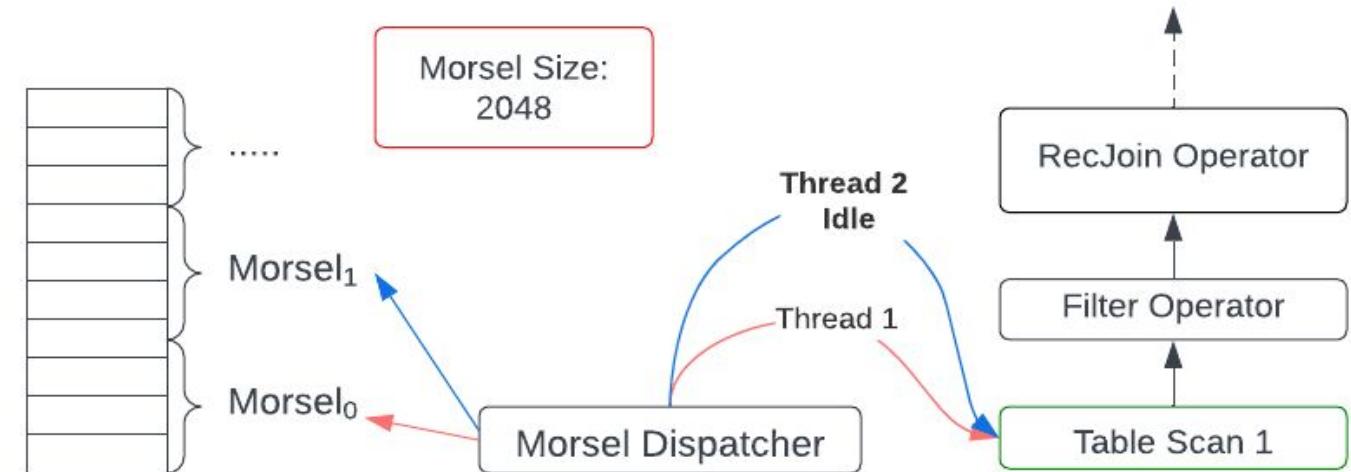
# Problem with Morsel-Driven Parallelism for Recursive Joins

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## Example:

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**Q.) How can we parallelize pipelines with recursive join operators robustly ?**



# Solution

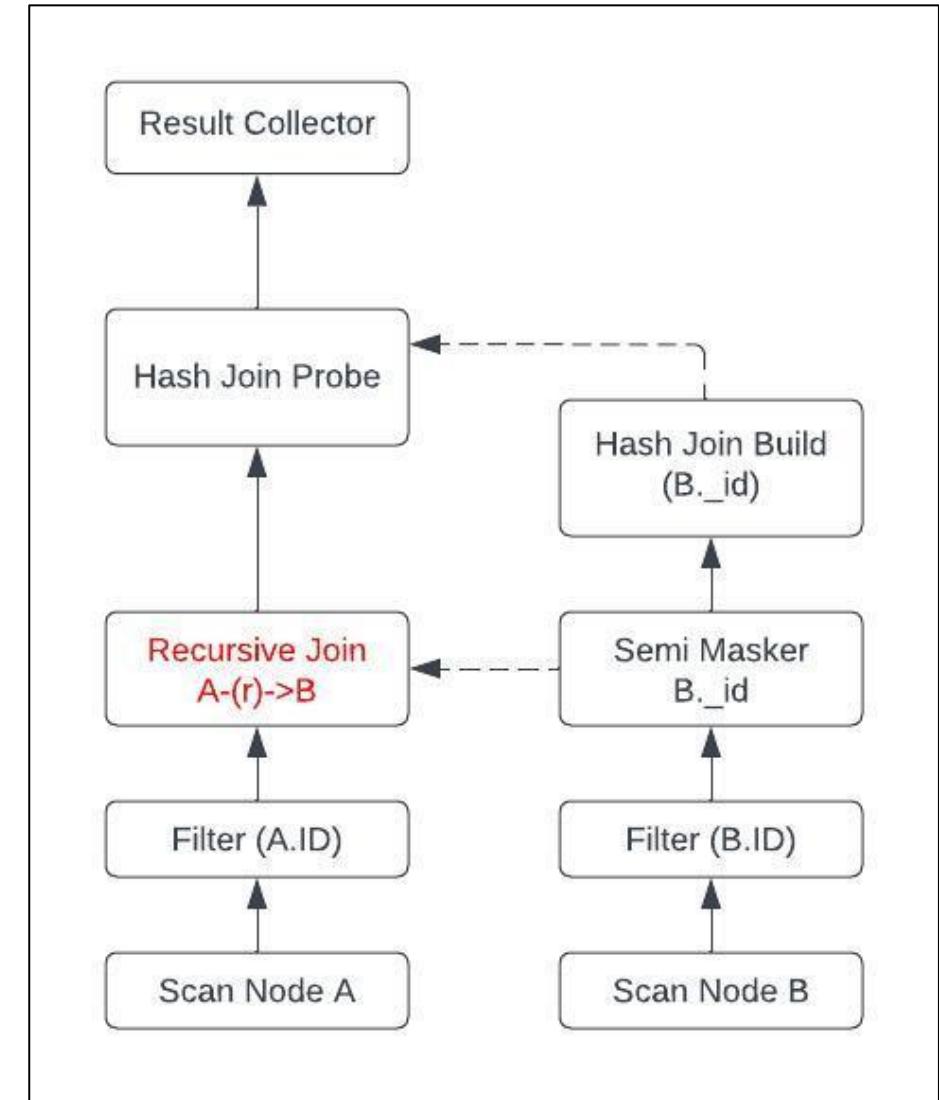
- (1) Make Recursive Join operator into a source operator to start the query pipeline. Threads should start BFS with ***a single source*** from this operator.

# Solution (Query Plan)

*Cypher query:*

```
MATCH p = (a:Person)-[r:knows*  
          SHORTEST 1..30]->(b:Person)  
WHERE a.ID < 1000 AND b.ID < 1000  
RETURN a.ID, b.ID, length(p)
```

*Recursive Join (RecJoin) operator  
must be the start of a pipeline*



# Solution

- (2) Morselize as before among threads with effective morsel size as 1 BFS source node (*Inter-RecJoin* parallelism)
- (3) When threads are idle, morselize a single RecJoin's BFS Level (frontier) into granular morsels among these threads (*Intra-RecJoin* parallelism)

# Solution

- (2) Morselize as before among threads with effective morsel size as 1 BFS source node (*Inter-RecJoin* parallelism)
- (3) When threads are idle, morselize a single RecJoin's BFS Level (frontier) into granular morsels among these threads (*Intra-RecJoin* parallelism)
- Define two types of morsels: (i) **BFSMorsel** (single source recursive join)  
(ii) **BFSLevelMorsel** (subset of *BFSMorsel's join*)
- Introduce a *Recursive Join scheduler* that distributes this work

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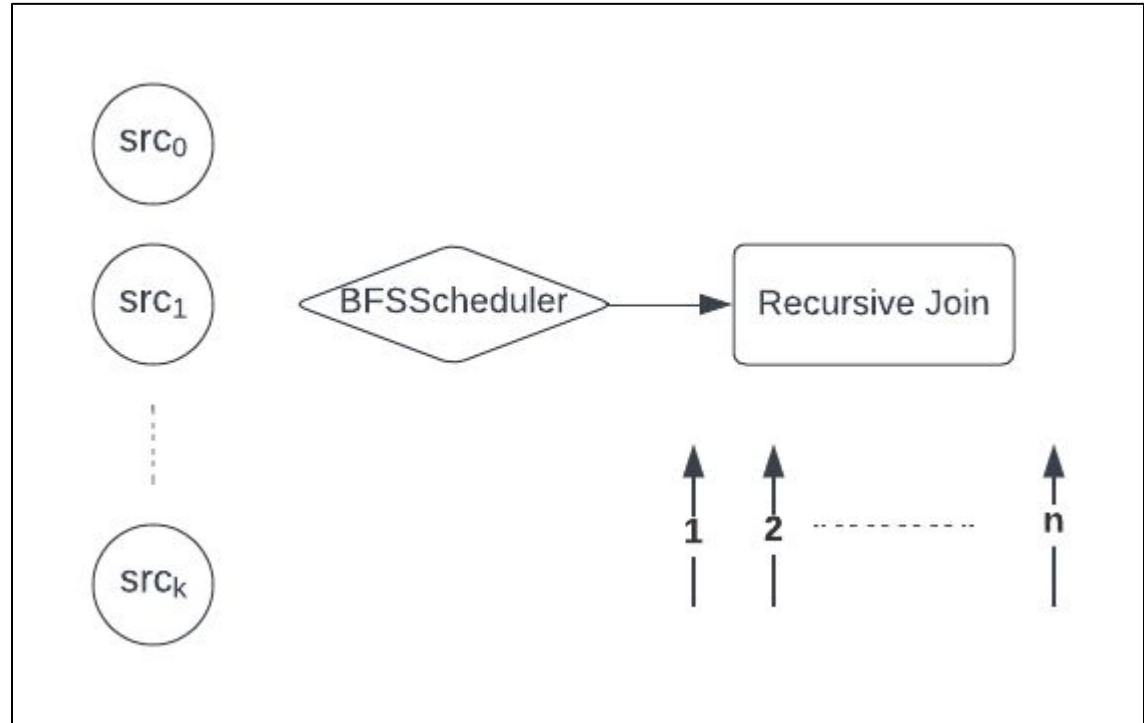
# Recursive Join Physical Operator

- BFSScheduler: Operators own scheduler that distributes work to the threads.

BFSScheduler controls total no. of concurrent **BFSMorsel** to at most  $k$ . Max limit is set to  $n$  (total threads).

$n$  Threads,  $k$  BFSMorsel

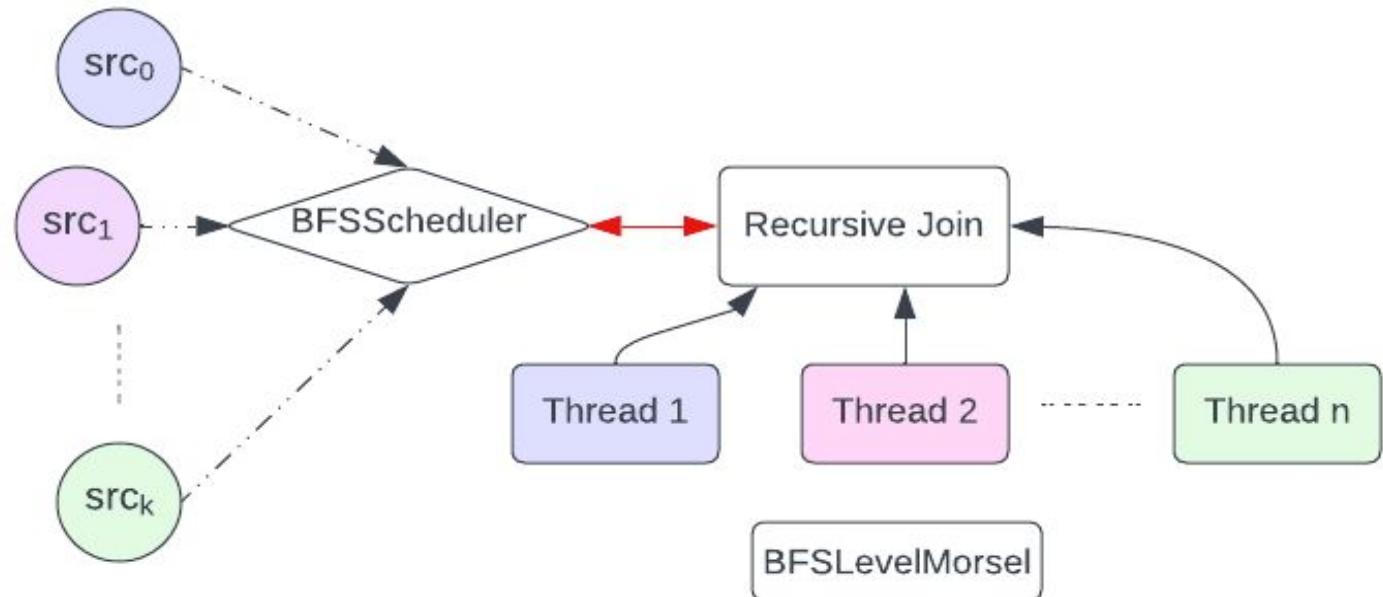
$nTkS$  scheduler



# Recursive Join Physical Operator

- BFSScheduler: Operators own scheduler that distributes work to the threads.

(1) BFS Scheduler launches a new BFS recursive join from a source if total (**active BFSMorsel** <  $k$ )

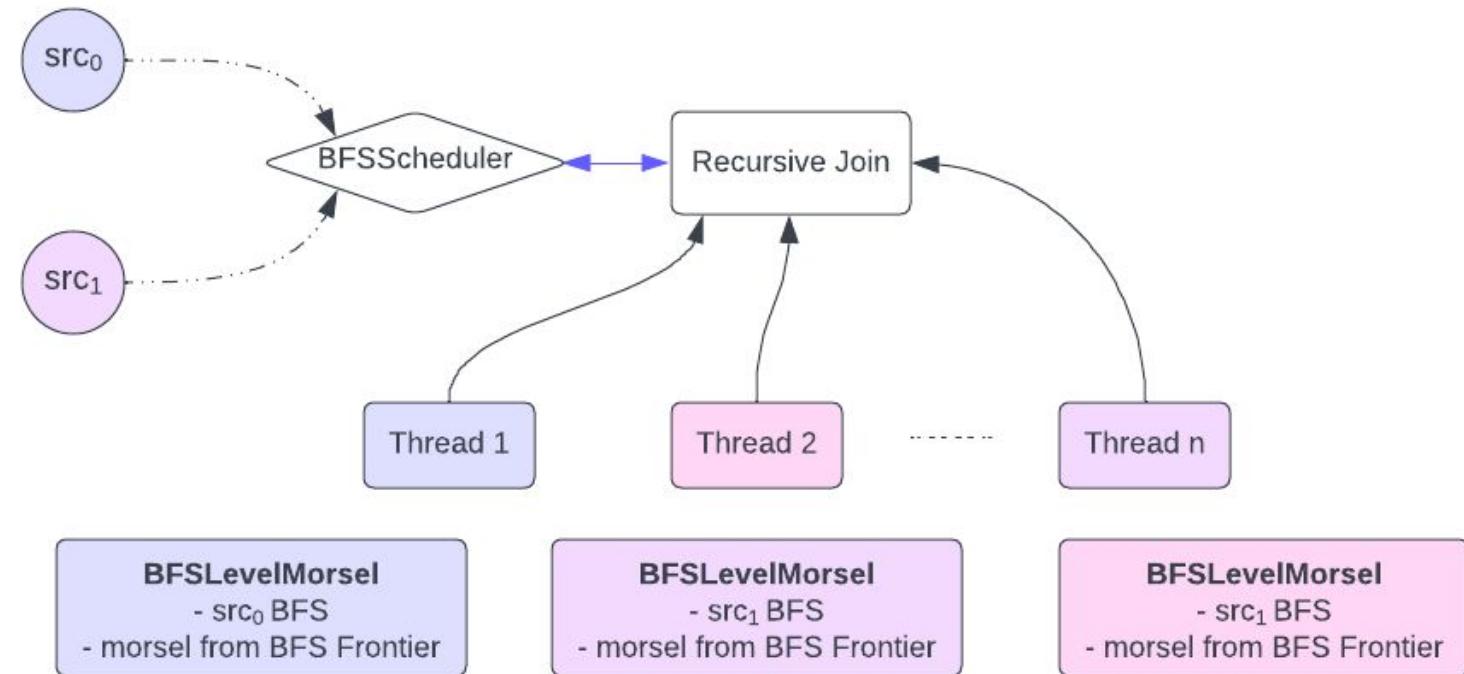


# Recursive Join Physical Operator

- BFSScheduler: Operators own scheduler that distributes work to the threads.

(1) BFS Scheduler launches a new BFS recursive join from a source if total (active BFS  $< k$ )

(2) If not, scheduler iterates over all active BFSMorsel to find BFS with most work and allot subset of the join as a BFSLevelMorsel.



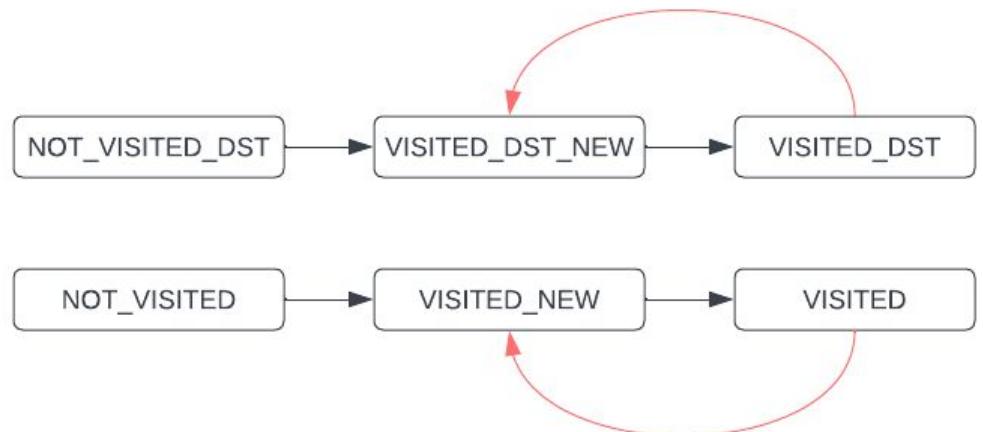
# Returning Path (Shortest / All Shortest)

- maintain a global ***visited array***
- update node states as they are encountered
- use lightweight ***lock-free*** synchronization
- additionally maintain ***multiplicity*** (all shortest path)
- use ***atomic CAS*** operations to update states and track {source node, edge} of nodes
- use ***atomic fetch and add (faa)*** operations to update multiplicity (for path length)



# Returning Path (Variable Length)

- maintain a global ***visited array***
- update node states as they are encountered
- use lightweight ***lock-free*** synchronization
- additionally maintain ***multiplicity*** at ***different levels***
- use ***atomic CAS*** operations to update states and track {source node, edge} of nodes at ***different levels***
- use ***atomic fetch and add (faa)*** operations to update multiplicity (for path length)



# Results

## Microbenchmark: (LDBC-100)

MATCH (a:Person)-[r:knows\* **SHORTEST** 1..30]->(b:Person) WHERE a.ID = 94 return b.ID, length(r);  
Total tuples: **407,396**

Kùzu (Baseline MDP - 32 threads)	Kùzu (nTkS - 32 threads)
728.6 ms	61 ms ( <i>12x faster</i> )

## Microbenchmark: (LiveJournal)

MATCH (a:lj\_node)-[r:lj\_rel\* **SHORTEST** 1..30]->(b:lj\_node) WHERE a.id < 1000 return b.ID, length(r);  
Total tuples: **4,237,533,225**

Kùzu (Baseline MDP - 32 threads)	Kùzu (nTkS - 32 threads)
158 s	105 s ( <i>1.5x faster</i> )

# Results

## Microbenchmark: (graph500-23)

MATCH (a:nodes)-[r:rels\* **ALL SHORTEST** 1..30]->(b:nodes) WHERE a.id = 307 RETURN r;  
Total tuples: **105,576,064**

Kùzu (Baseline MDP - 32 threads)	Kùzu (nTkS - 32 threads)
511s	35s ( <b>14.6x faster</b> )

## Microbenchmark: (graph500-24)

MATCH (a:nodes)-[r:rels\* **1..4**]->(b:nodes) WHERE a.id = 0 RETURN r;  
Total tuples: **126,749,073**

Kùzu (Baseline MDP - 32 threads)	Kùzu (nTkS - 32 threads)
634s	37.6s ( <b>16.9x faster</b> )

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# Future Work

- Integrating other techniques [Multi source BFS (MS-BFS), Bidirectional BFS]
- Storing paths in a compressed manner for vectorized execution ?
- Weighted Shortest Path (Dijkstra, Bellman Ford, ...) ?

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Thank You