



清華大學  
Tsinghua University

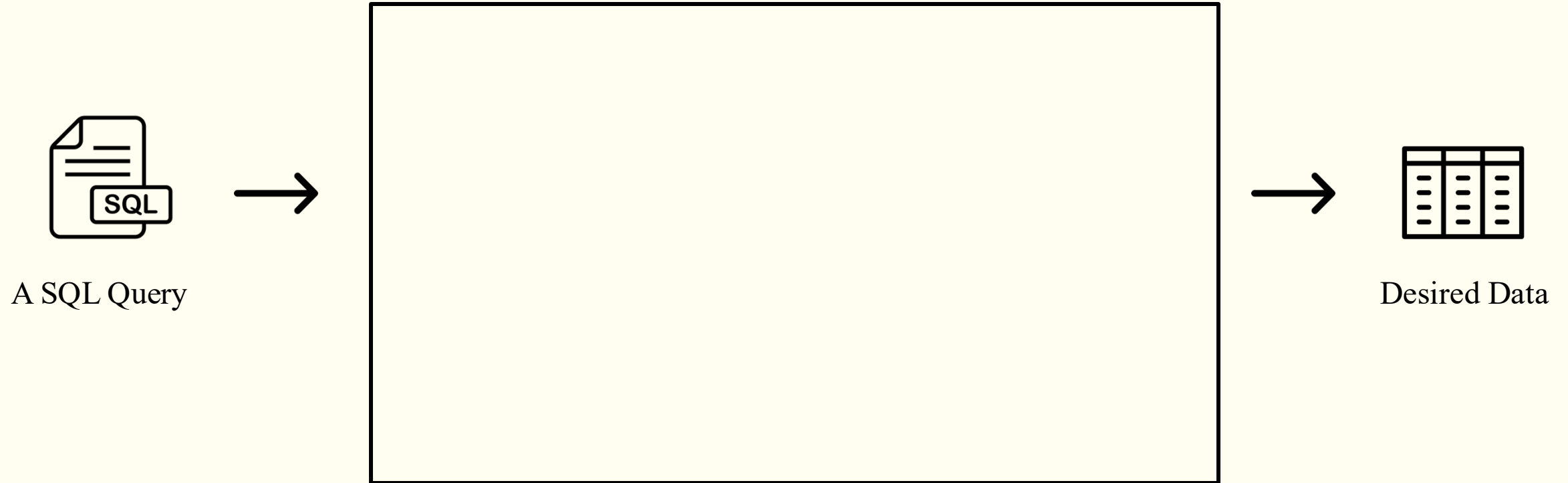
# Data Chunk Compaction in Vectorized Execution

Yiming Qiao, Huanchen Zhang

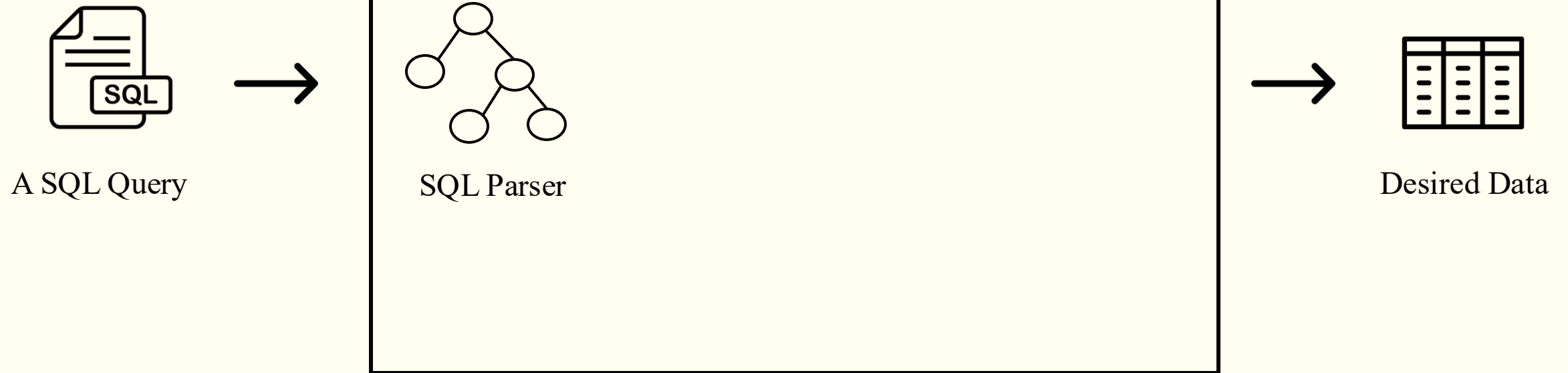
Institute for Interdisciplinary Information Sciences

Tsinghua University

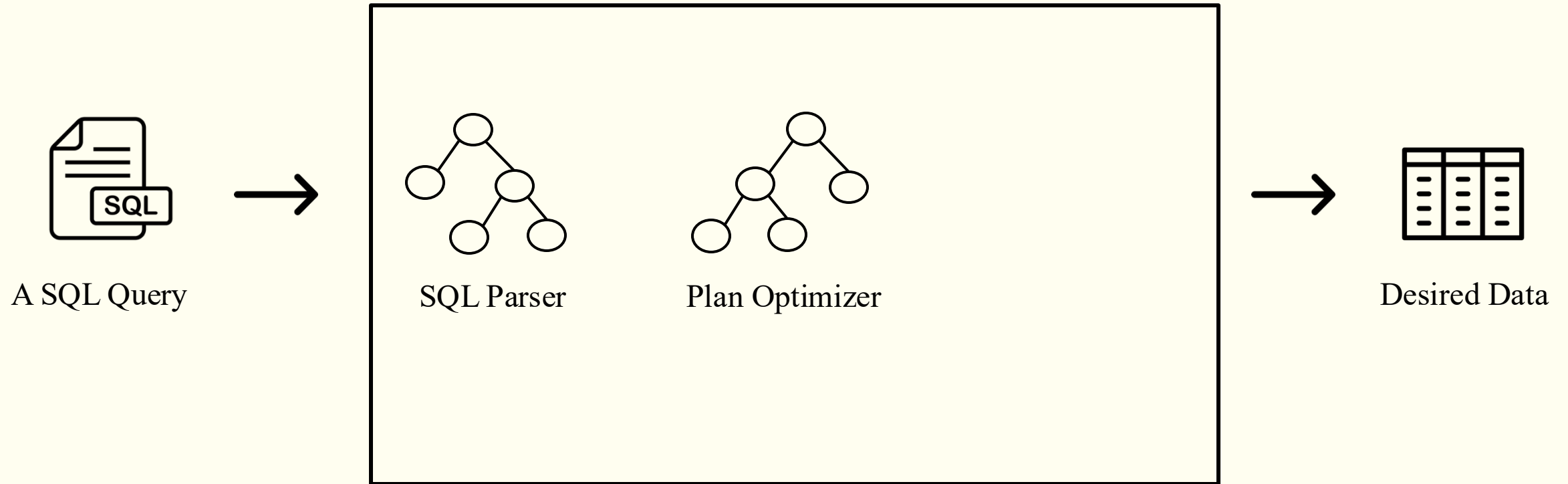
# Database System



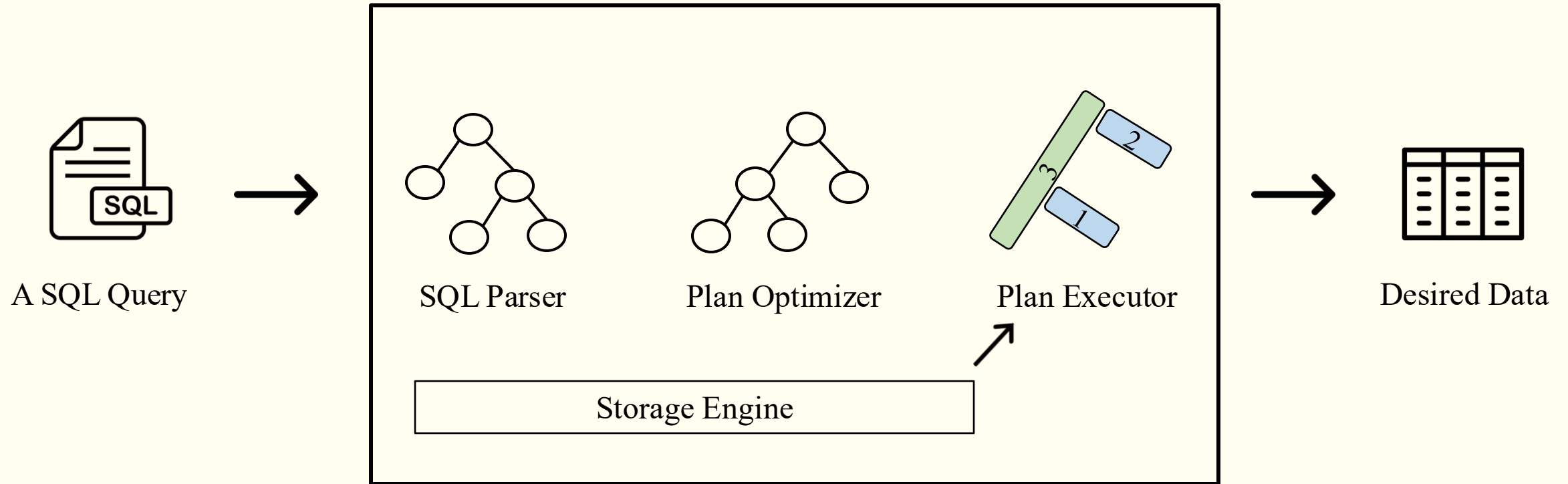
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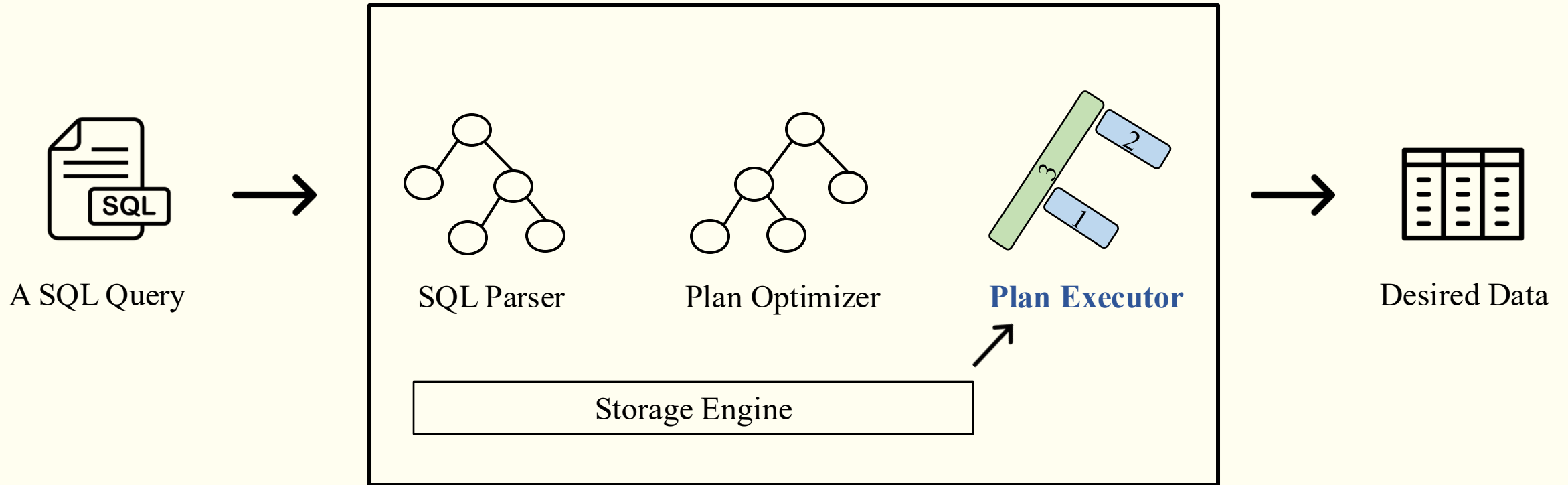
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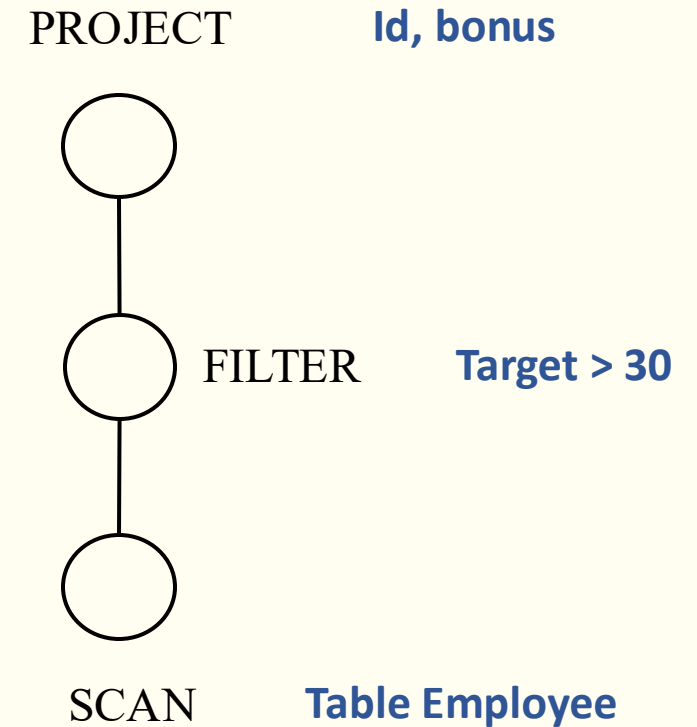
# Database System



We Focus on Improving the Query Plan Executor

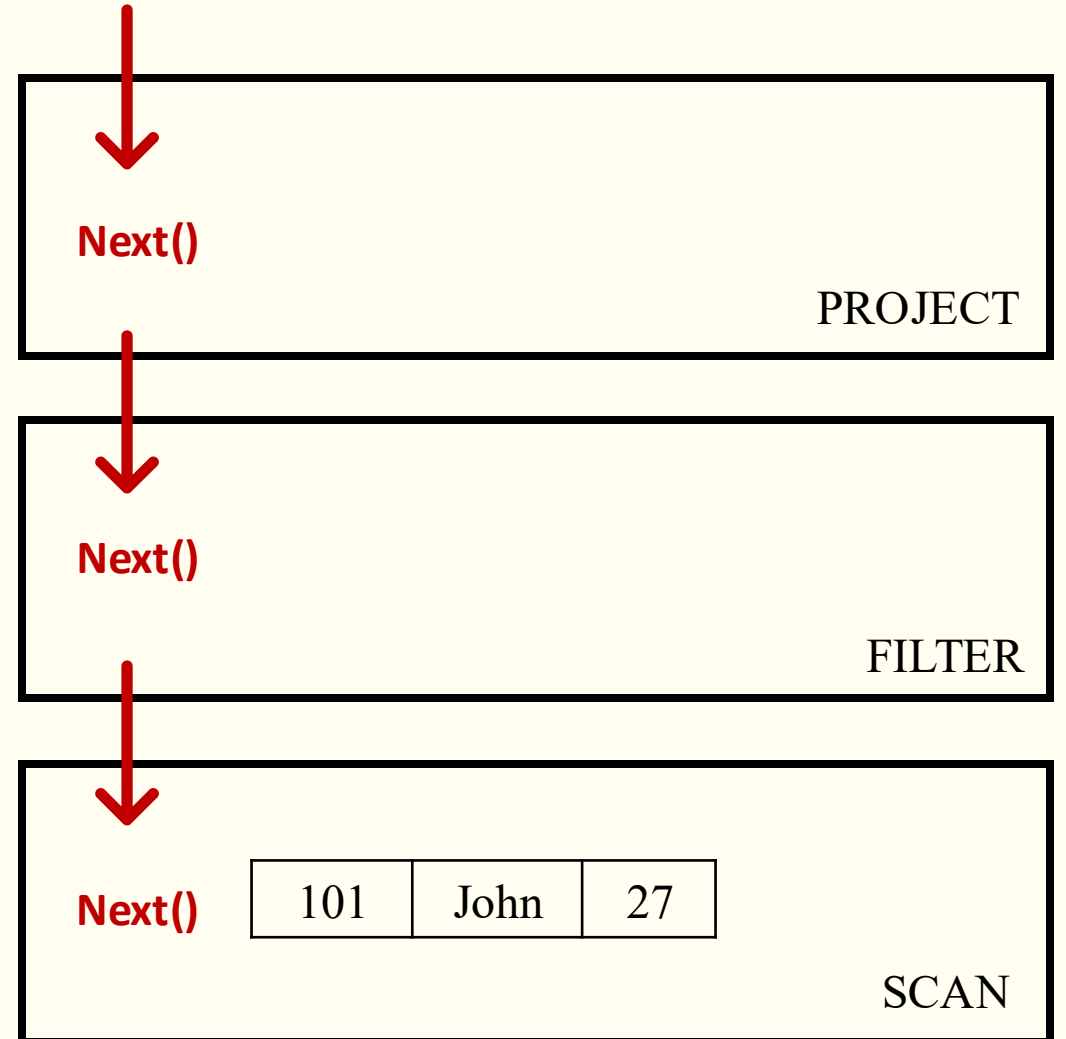
# Iterator Execution Model

```
SELECT id,  
       (target - 30) * 50 AS bonus  
FROM   employee  
WHERE  target > 30
```



# Iterator Execution Model

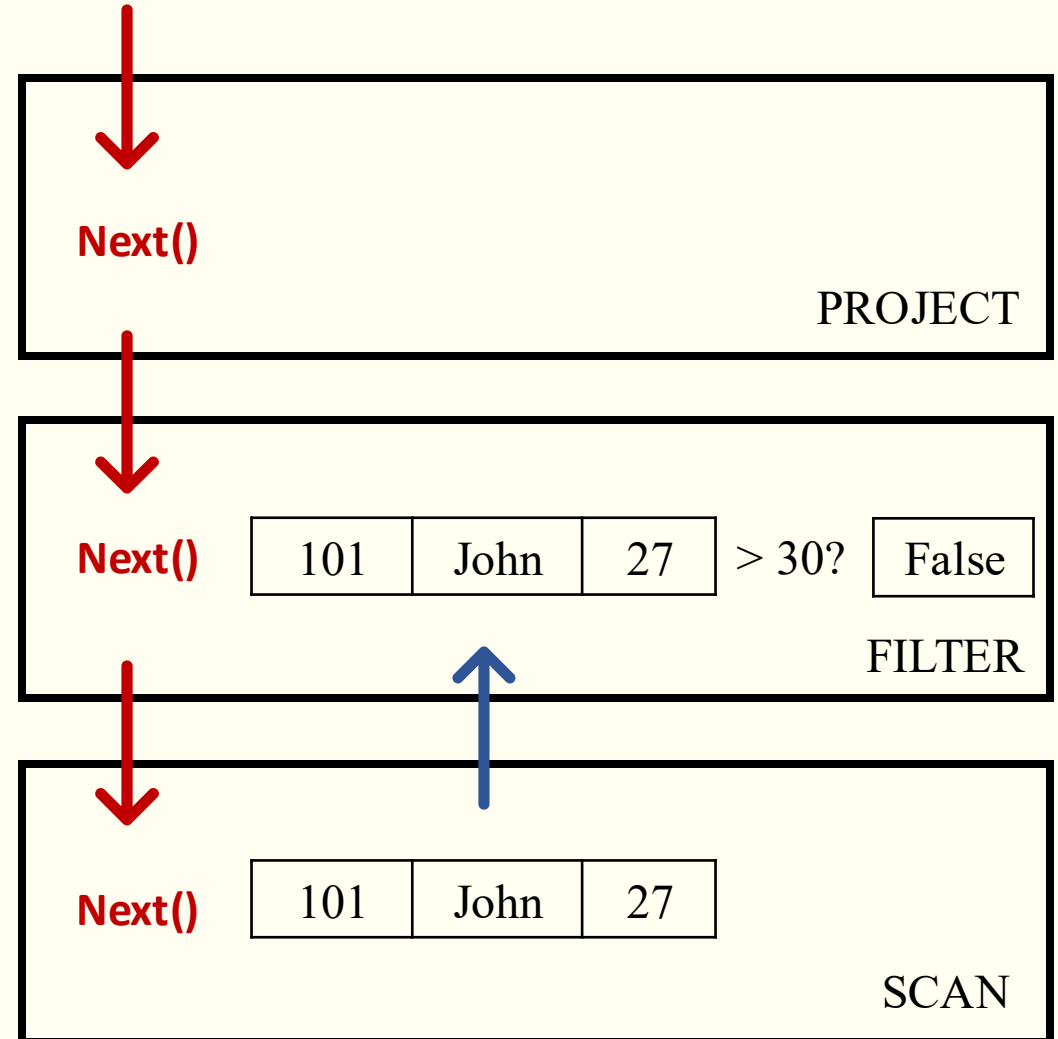
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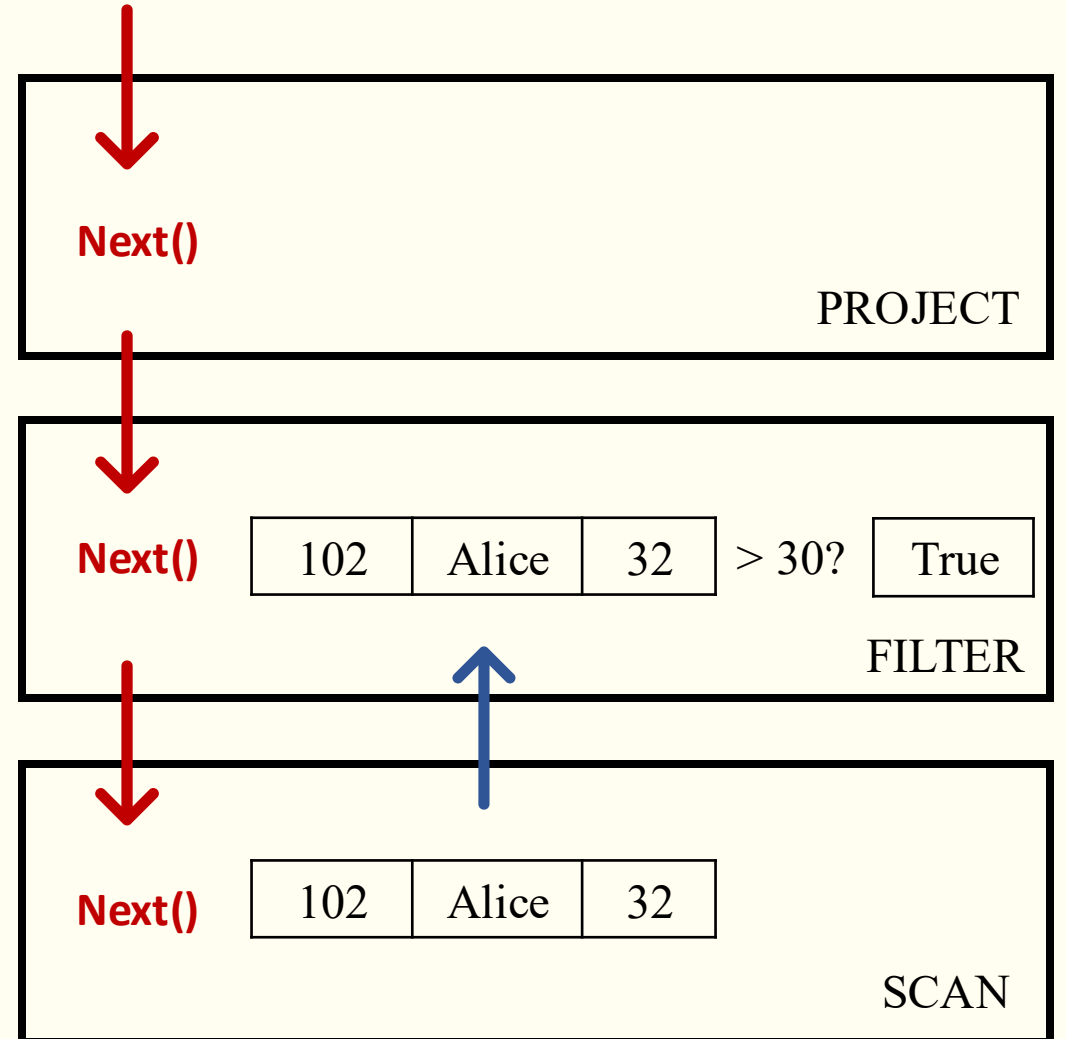
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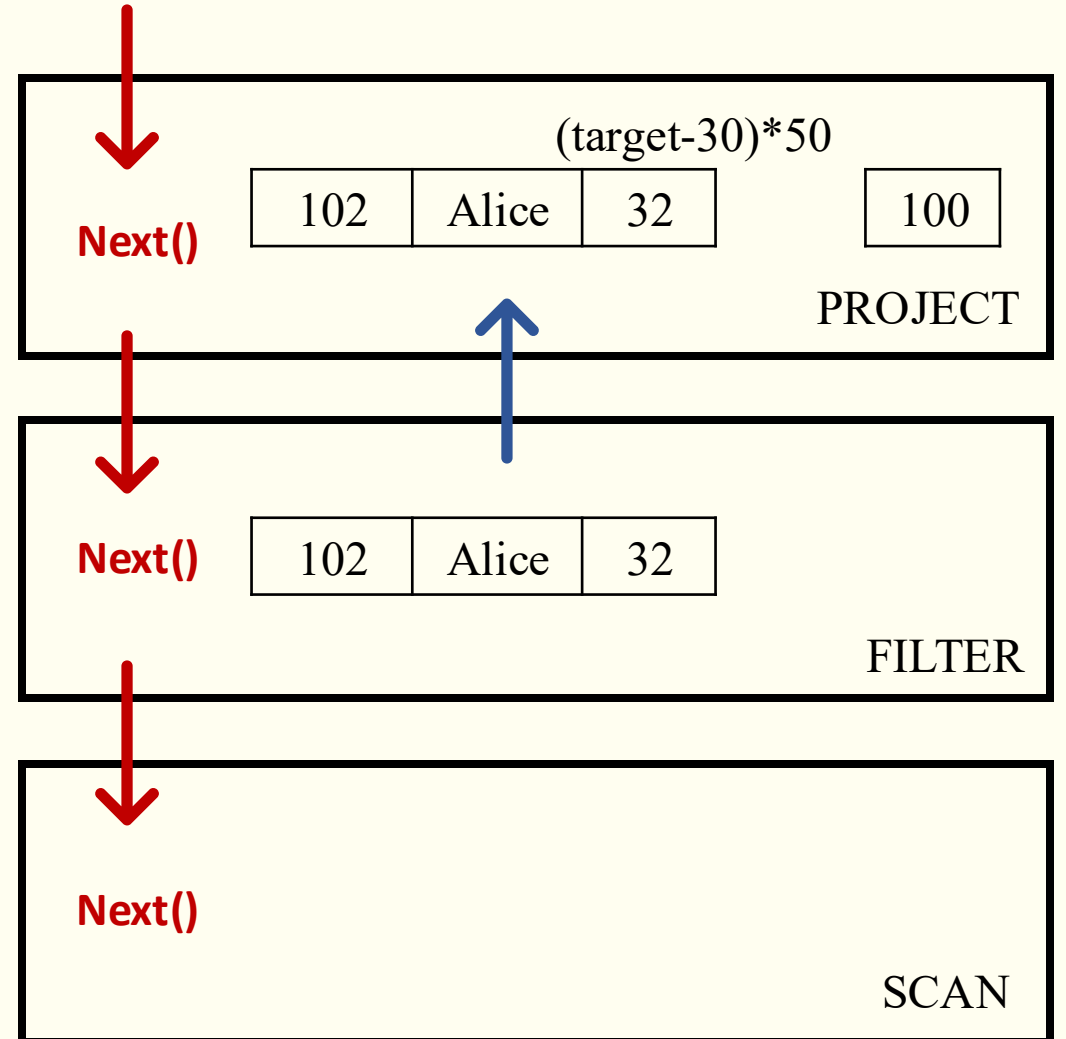
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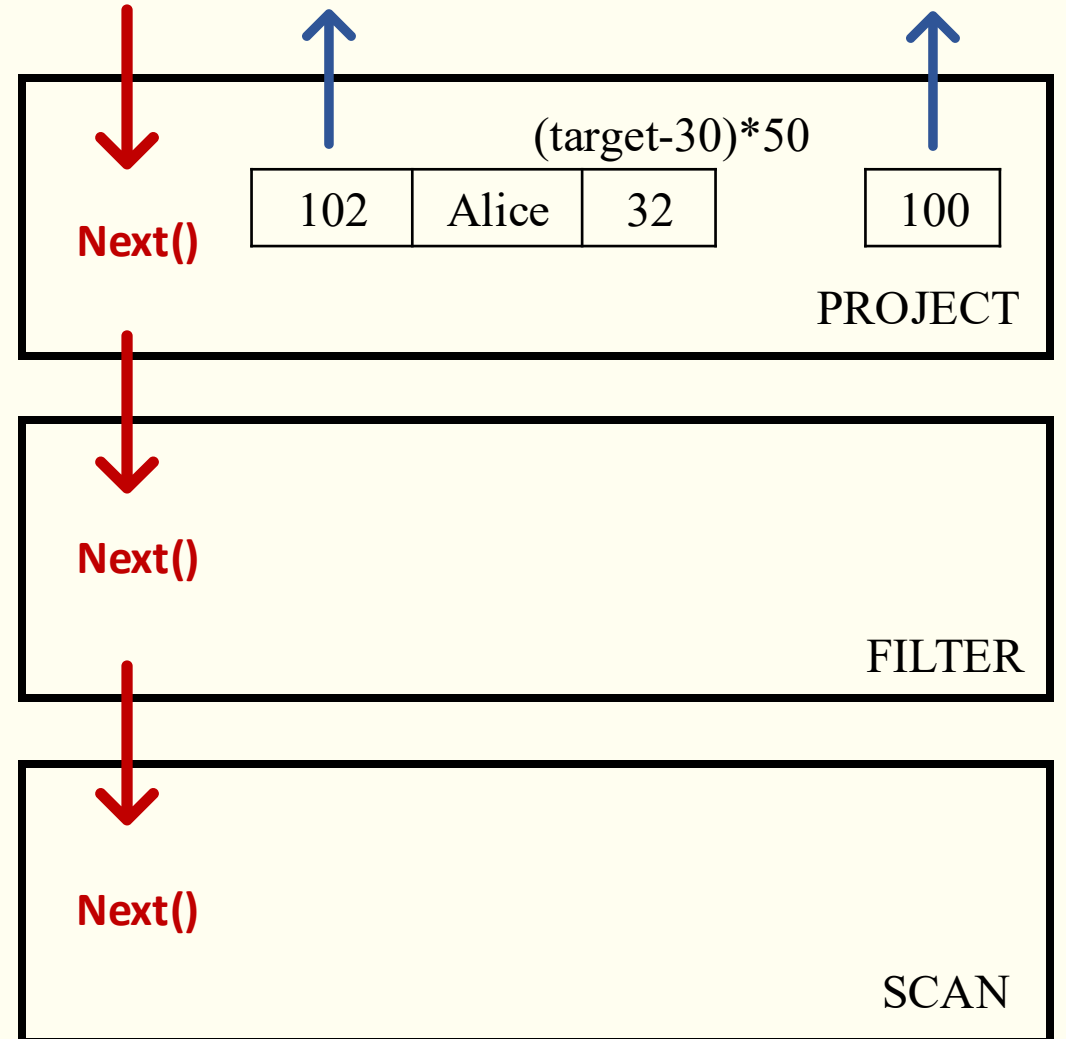
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# Iterator Execution Model

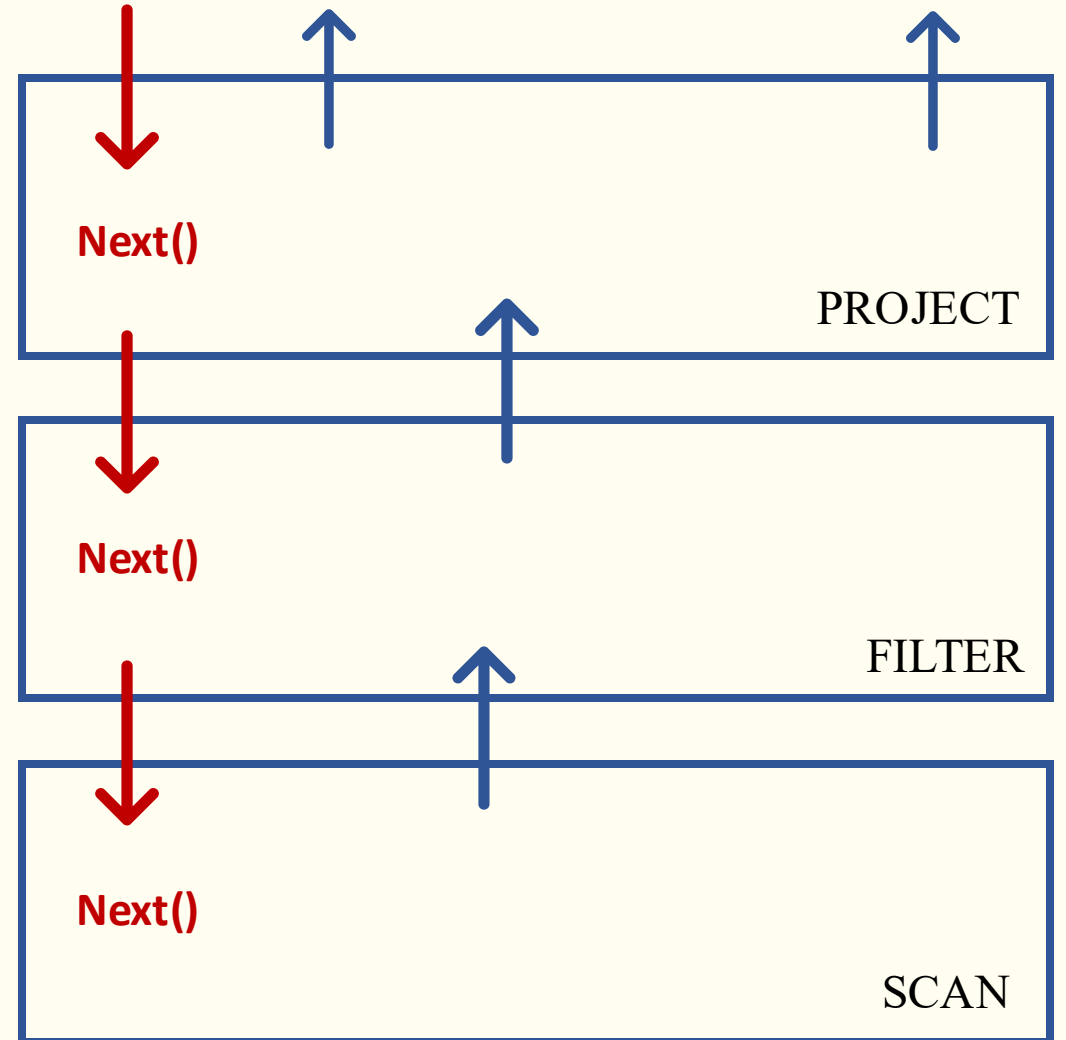
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SELECT id,  
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FROM   employee  
WHERE  target > 30
```



# Iterator Execution Model

- Operators: SCAN, FILTER, PROJECT
- Iterator Interface:
  - **Open()**
  - **Next()** # “Next Tuple Please”
  - **Close()**

Three Function Calls for Processing a Single Tuple!



# Buying Beer for a Party. The Silly Way.

An Intuitive Example From Peter Boncz

- Go to a store
- Take one beer bottle
- Pay at the register
- Return home
- Put the bottle in the fridge
- Repeat... 100 times!



= “tuple”

# Buying Beer for a Party. A Better Way.

An Intuitive Example From Peter Boncz

- Go to a store
- Take **two crates of beer** (2x24!)
- Pay at the register
- Return home
- Put crates in the fridge
- **Repeat 48x less!**



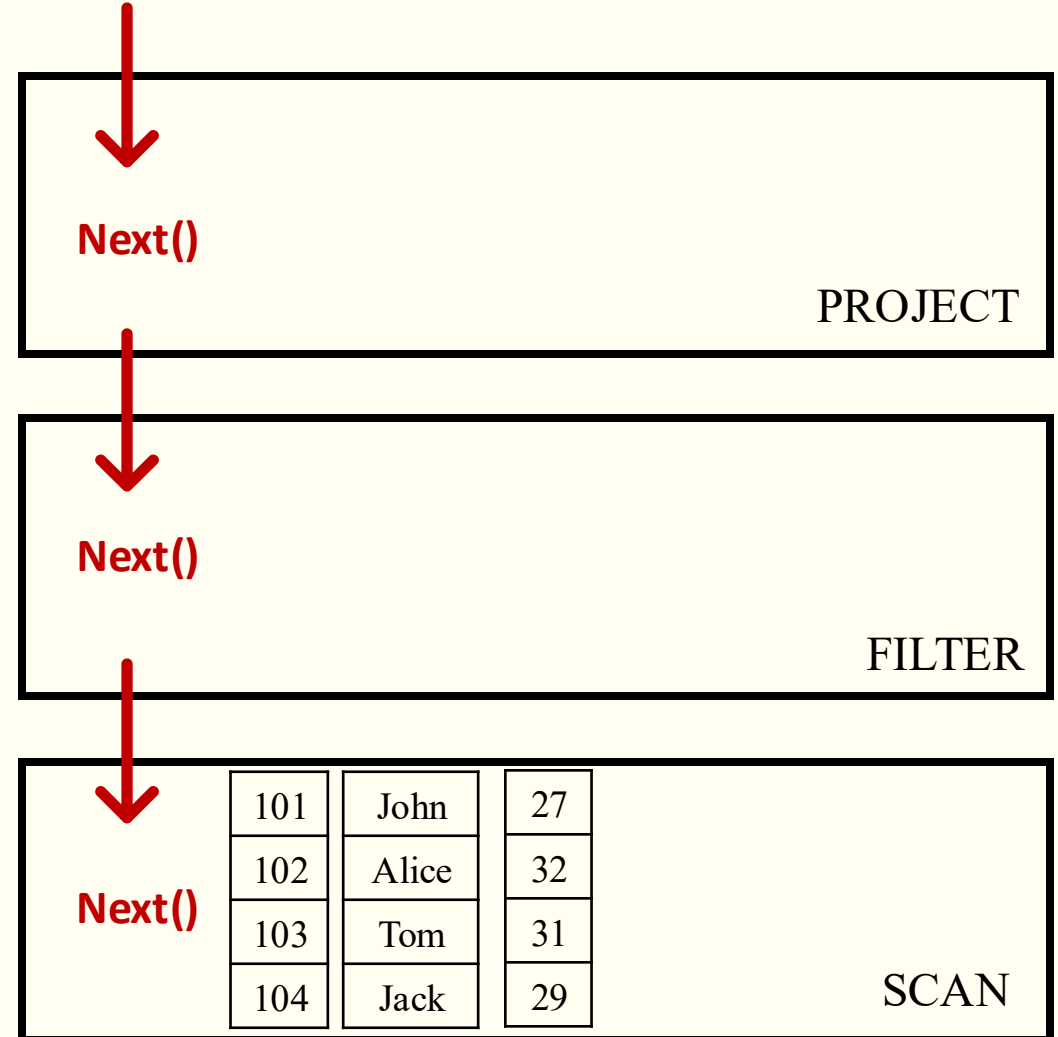
= “vector”

**48x less Function Calls for Processing a Single Tuple!**

# A Smart Way: Vectorized Execution

```
SELECT id,  
       (target - 30) * 50 AS bonus  
FROM   employee  
WHERE  target > 30
```

Process a Chunk of Tuples at  
a Time, with Columnar Store

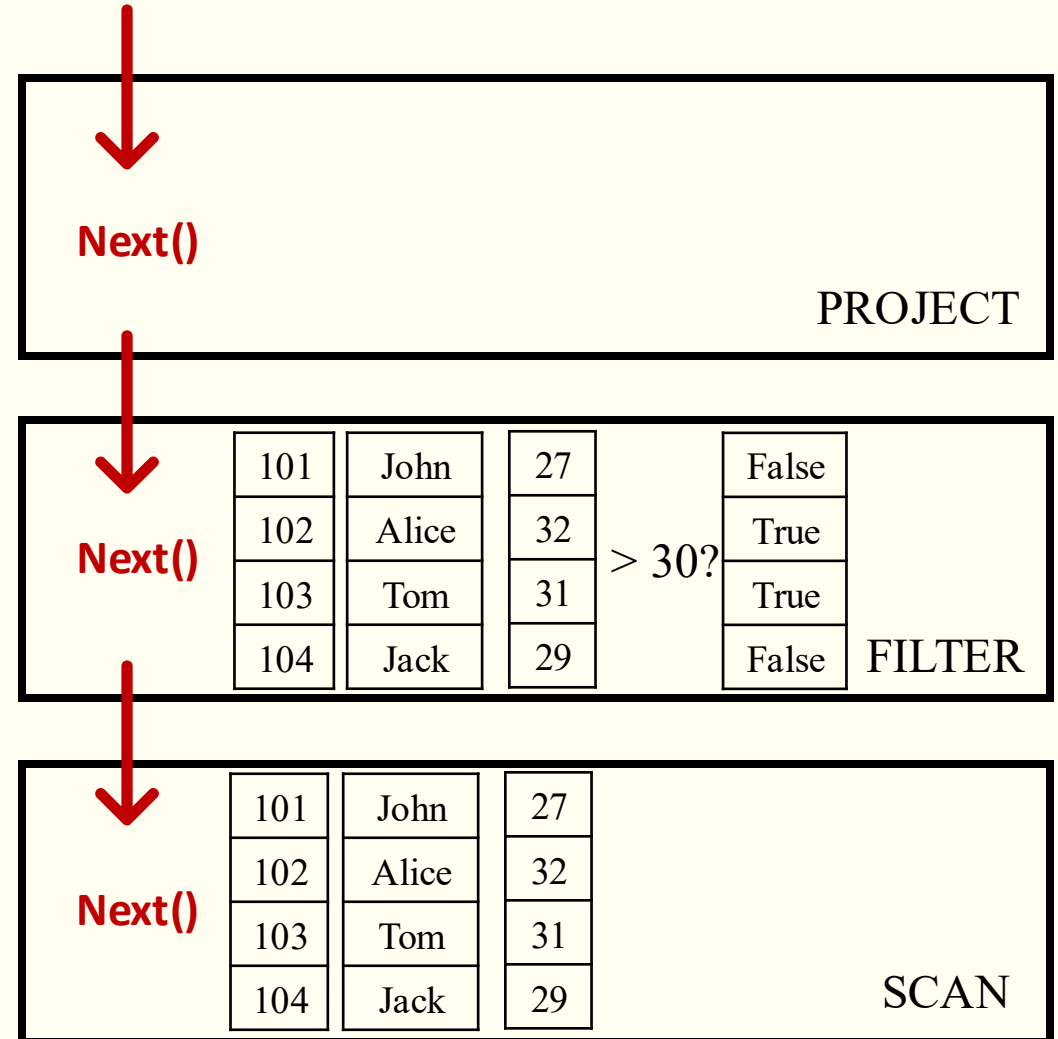




# A Smart Way: Vectorized Execution

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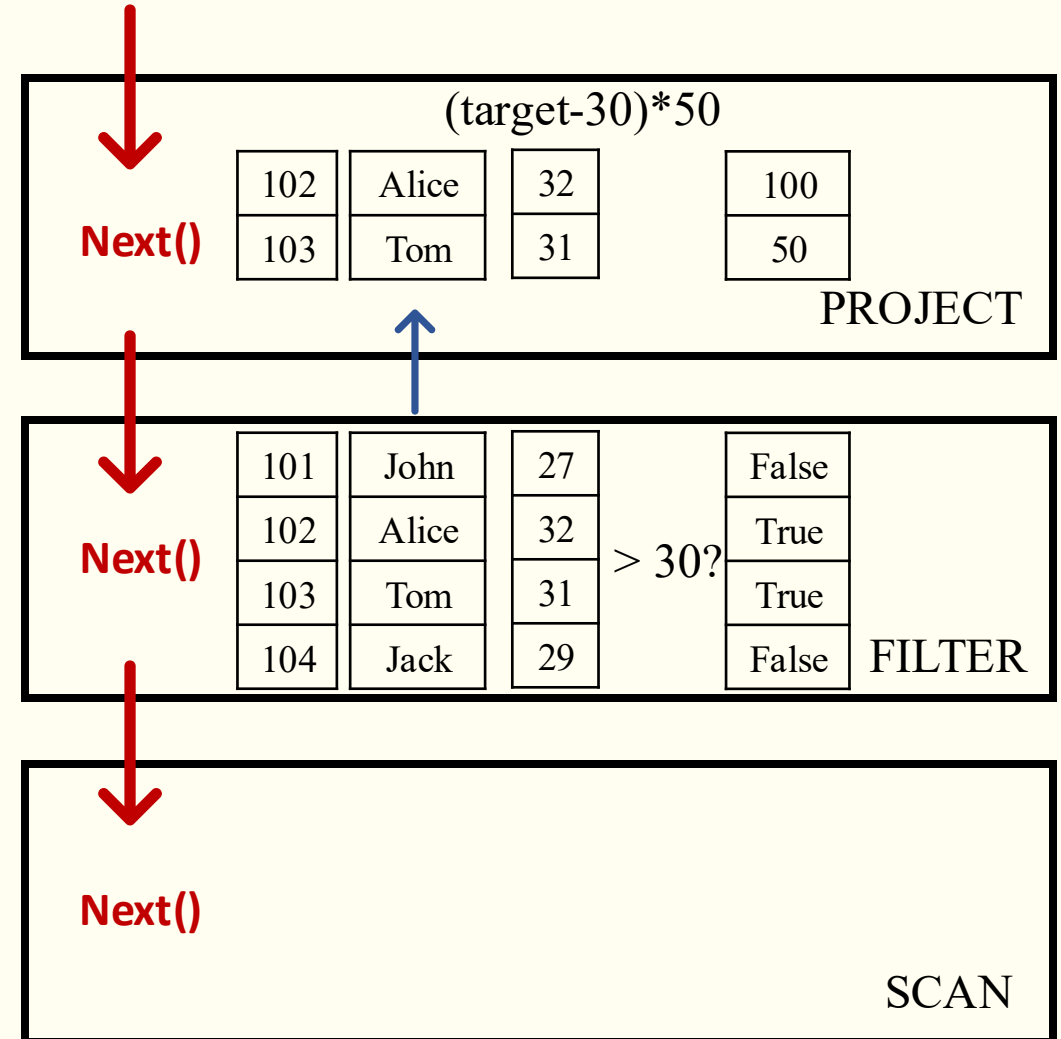
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# A Smart Way: Vectorized Execution

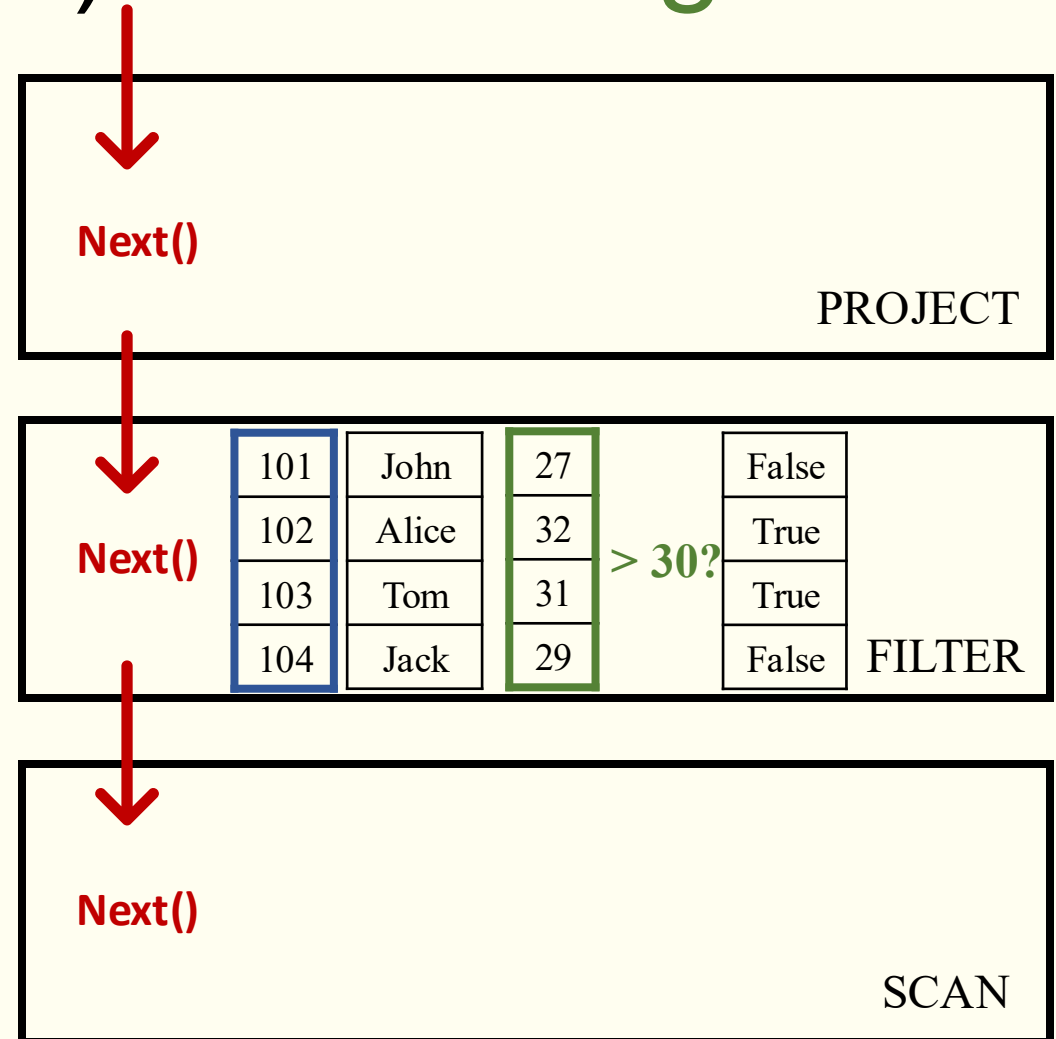
```
SELECT id,  
       (target - 30) * 50 AS bonus  
FROM   employee  
WHERE  target > 30
```

A Selection Vector/Bit Map is  
used to Mark Tuples



# Vectorized (In-Cache) Processing

- Vector-at-a-time:
  - About 1000 tuples
- Processed in a tight loop:
  - $>$ ,  $+$ ,  $*$
- Cache-friendly:
  - A vector is sequentially accessed

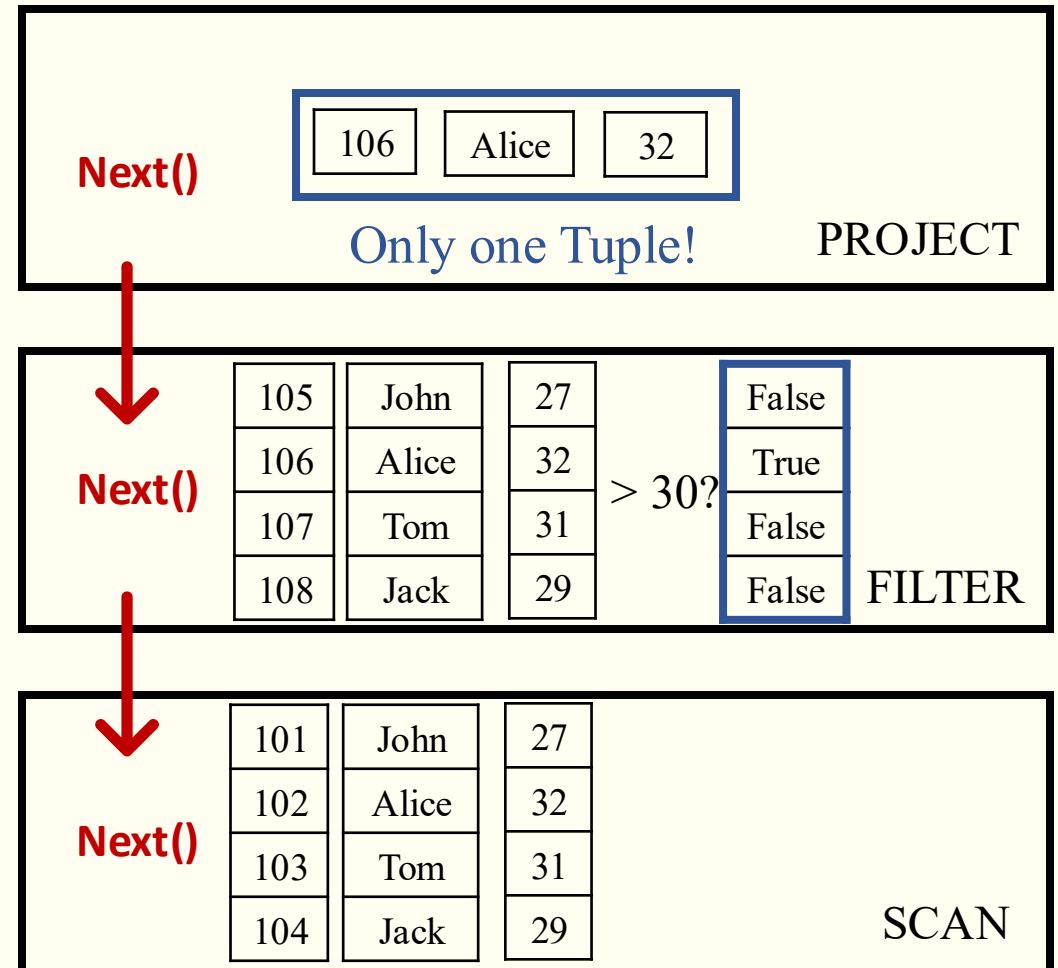


# Part 1

## Data Chunk Compaction in Vectorized Execution

# Small Chunk Problem

```
SELECT id,  
       (target - 30) * 50 AS bonus  
FROM   employee  
WHERE  target > 30
```

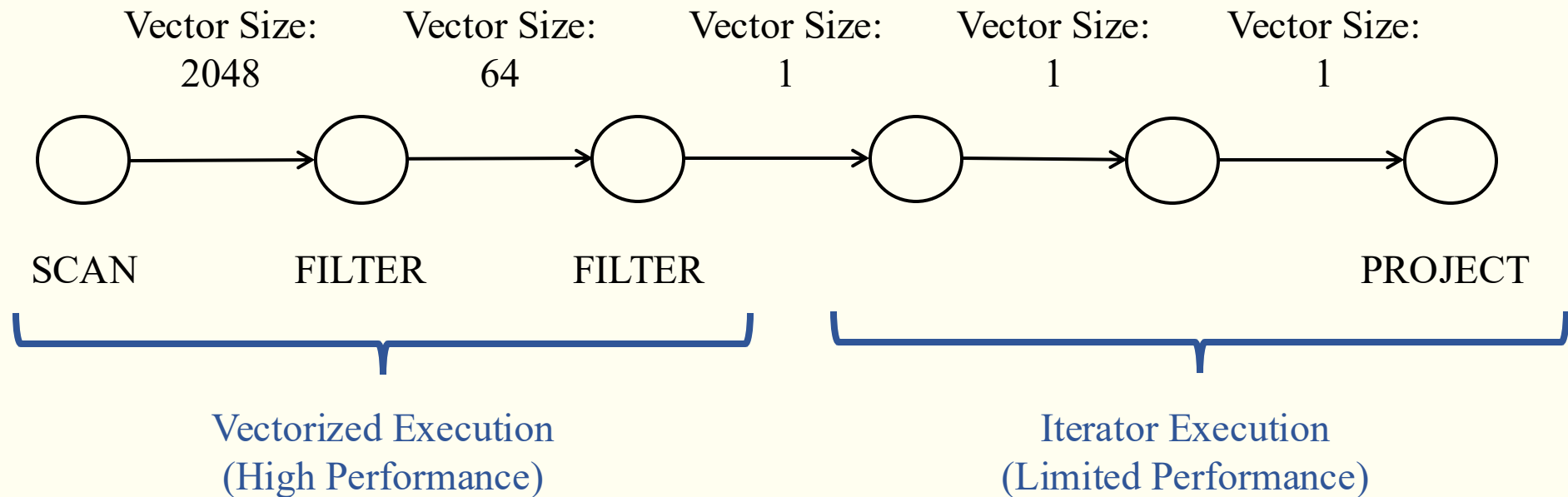


# Buying Beer for a Party. An Accident.



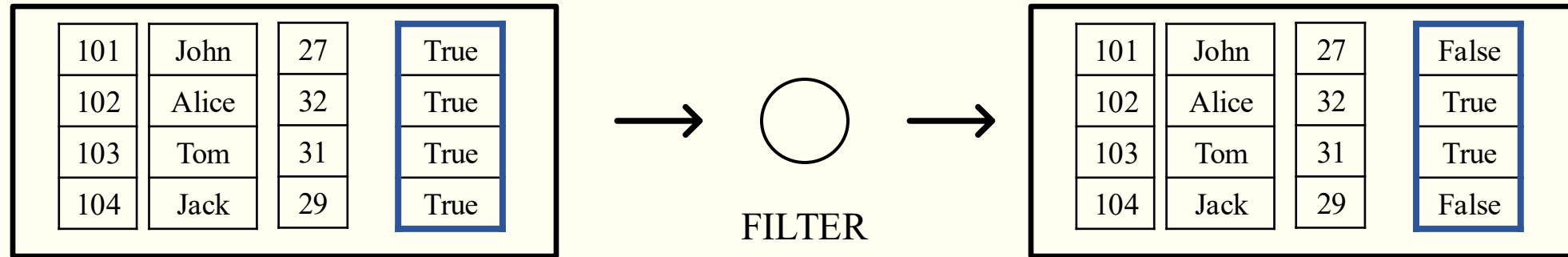
We take **two crates** from the shop  
but only put **one bottle** in the fridge!

# Small Chunk Problem



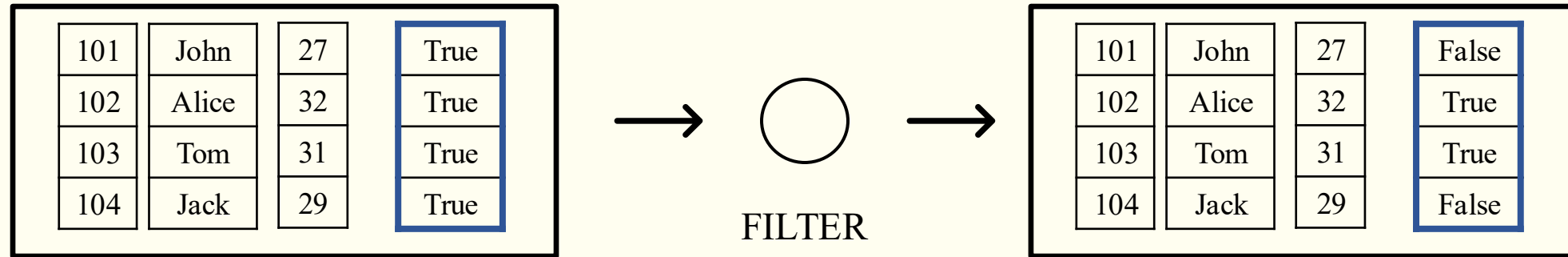
Chunk Size (and Data Volume) is Greatly Reduced During Execution

# Chunk-reducing Operators: Filter



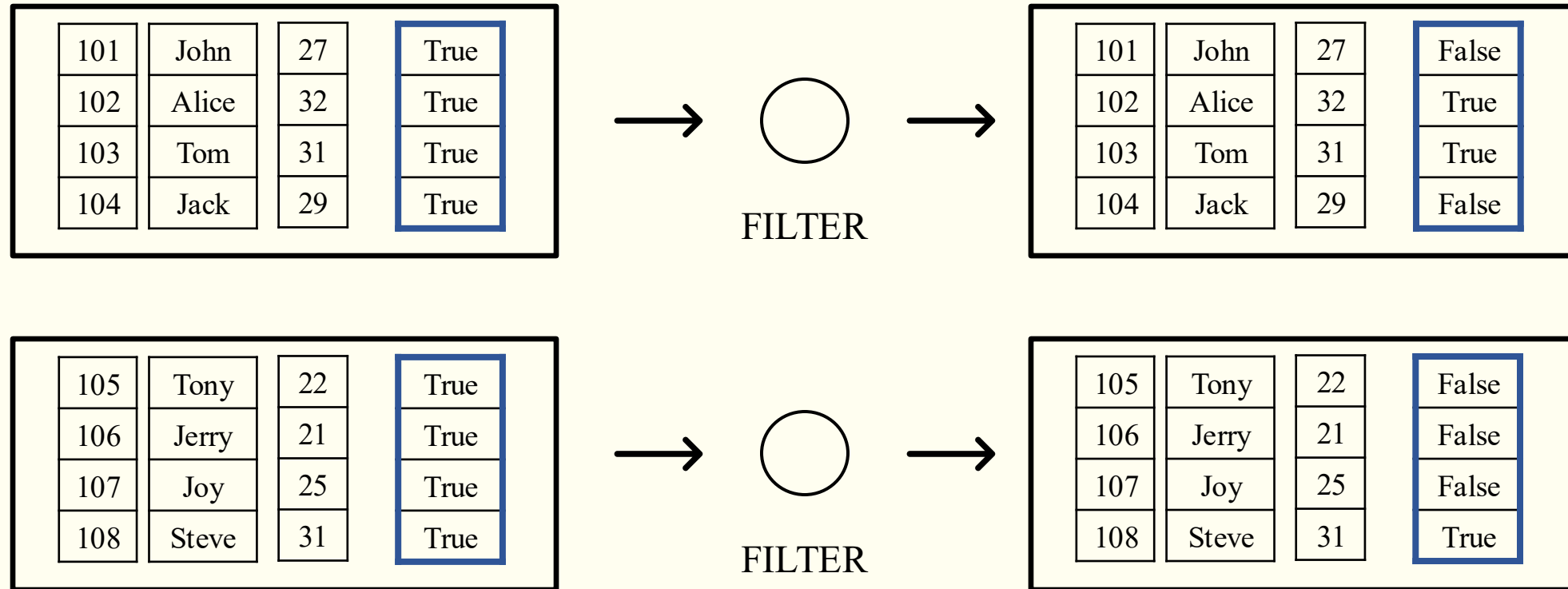


# Chunk-reducing Operators: Filter

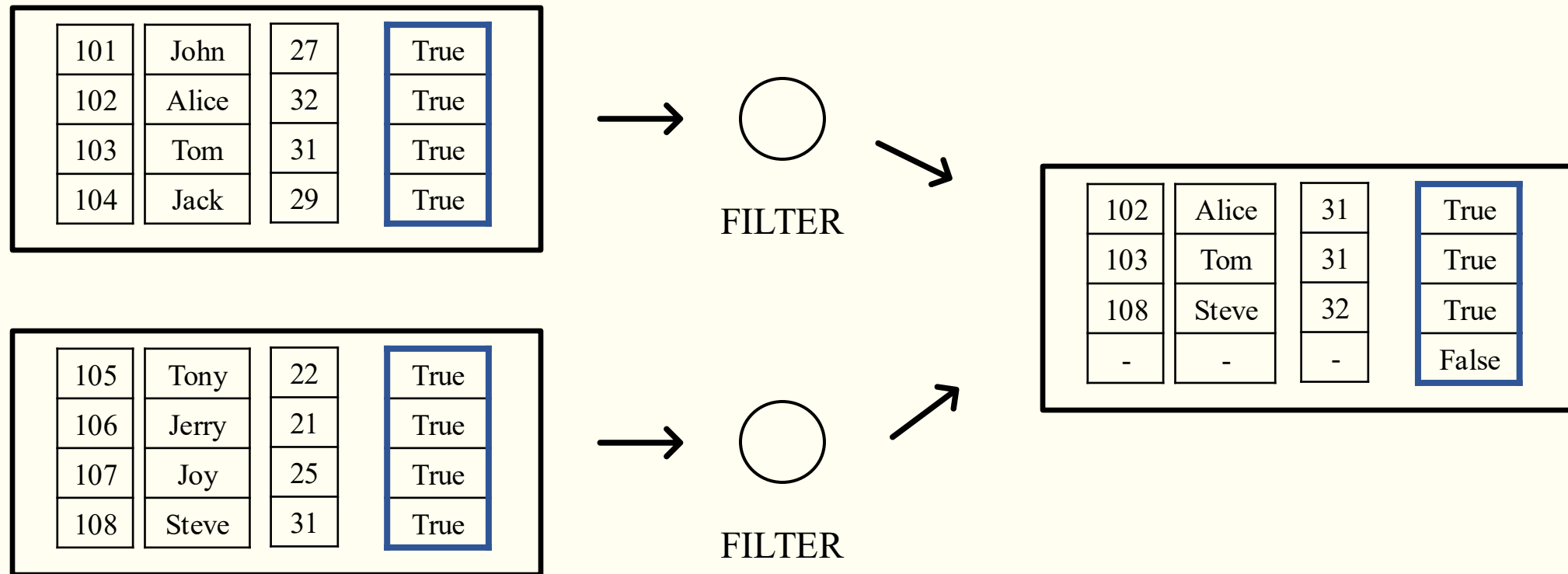


- Data Chunk Structure: Some Data Vectors + **One Selection Vector**
- **Zero-copy Benefit**

# Chunk-reducing Operators: Filter



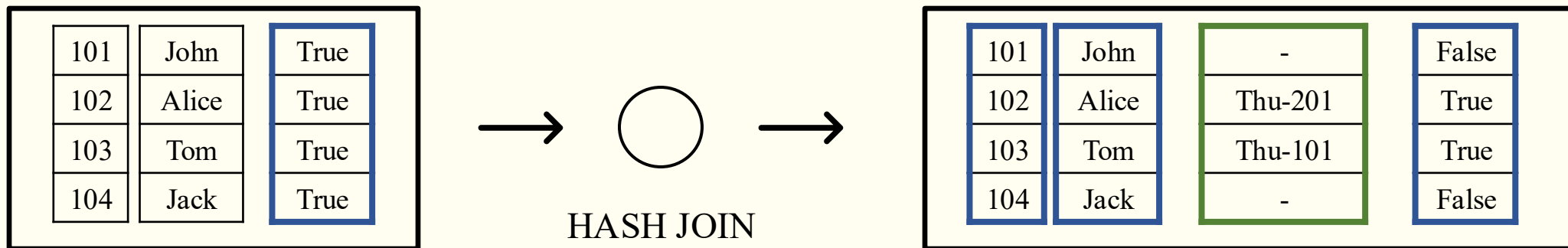
# Chunk-reducing Operators: Filter



If we compact them, we have additional **memory copies**;  
otherwise, the **vectorized execution suffers**.

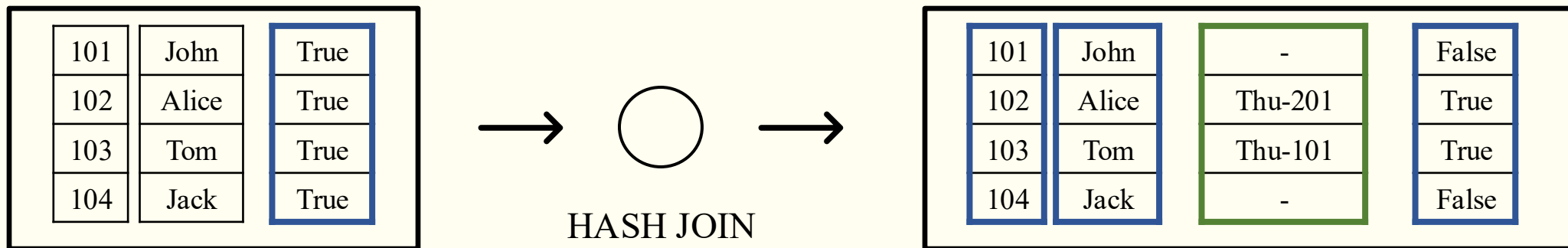
# Chunk-reducing Operators: Hash Join

**SELECT** id, name, course\_id,  
**FROM** students, courses



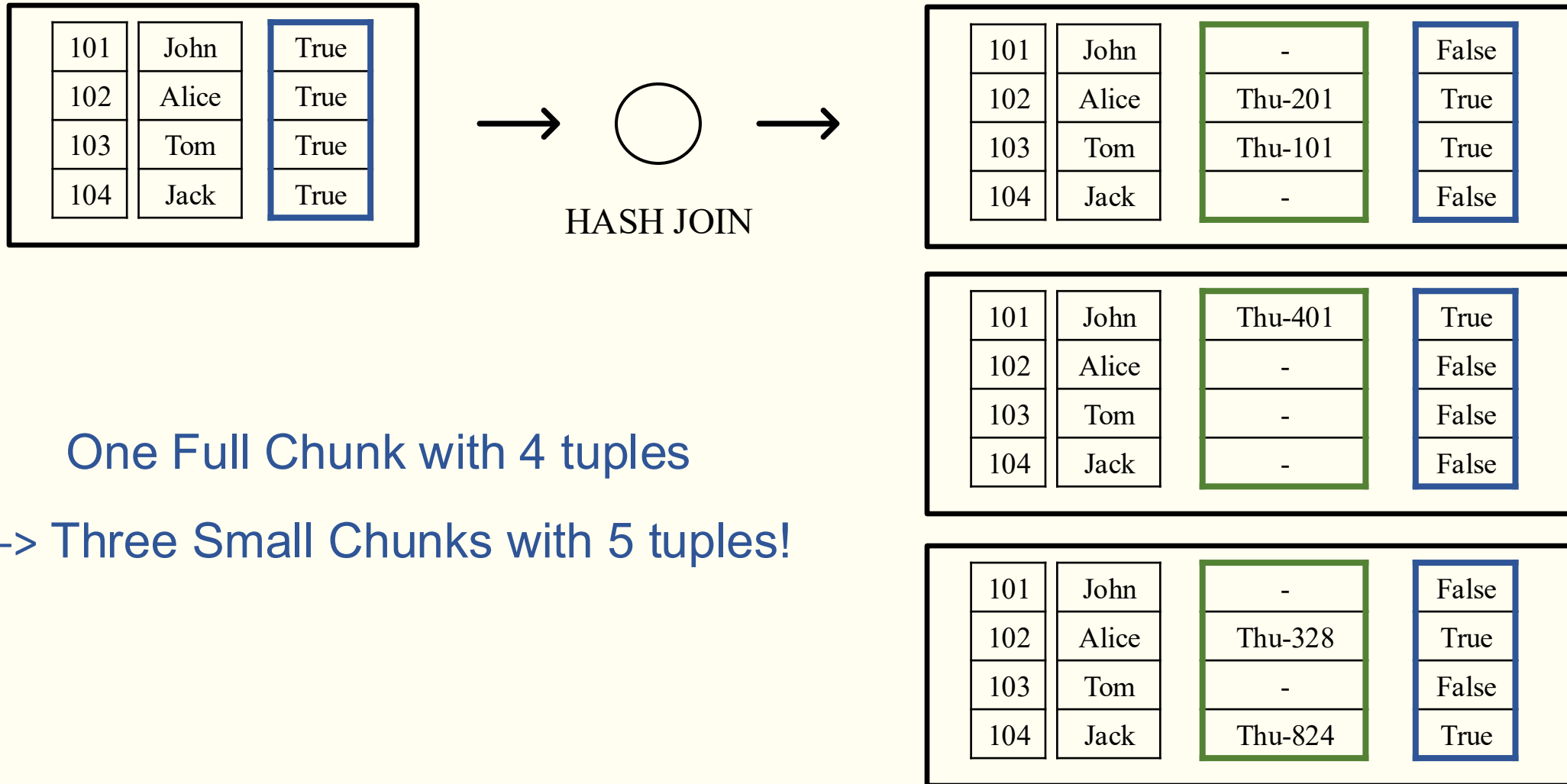
# Chunk-reducing Operators: Hash Join

**SELECT** id, name, course\_id,  
**FROM** students, courses



- **Zero-copy Benefit:** Reuse the data in the column **id** and **name**.
- **Copy the New Vector Values** into the Result Chunk.

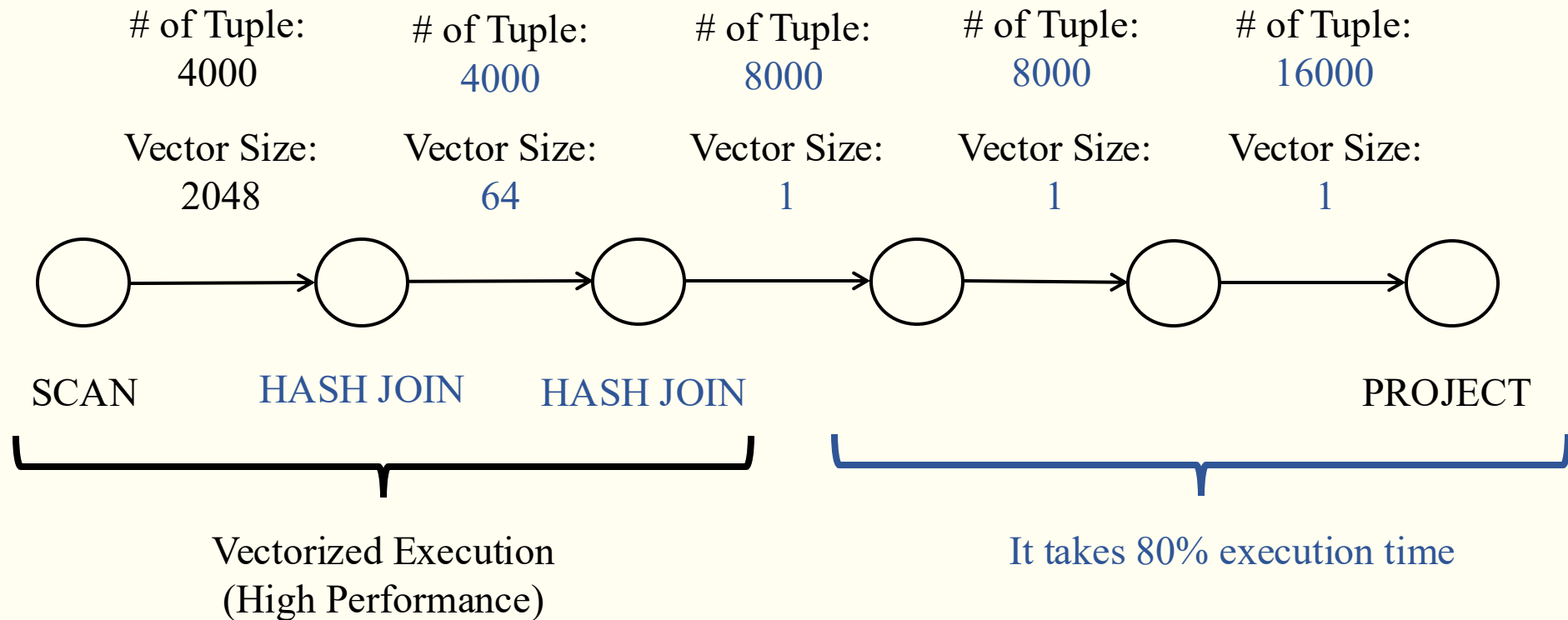
# Chunk-reducing Operators: Hash Join



One Full Chunk with 4 tuples

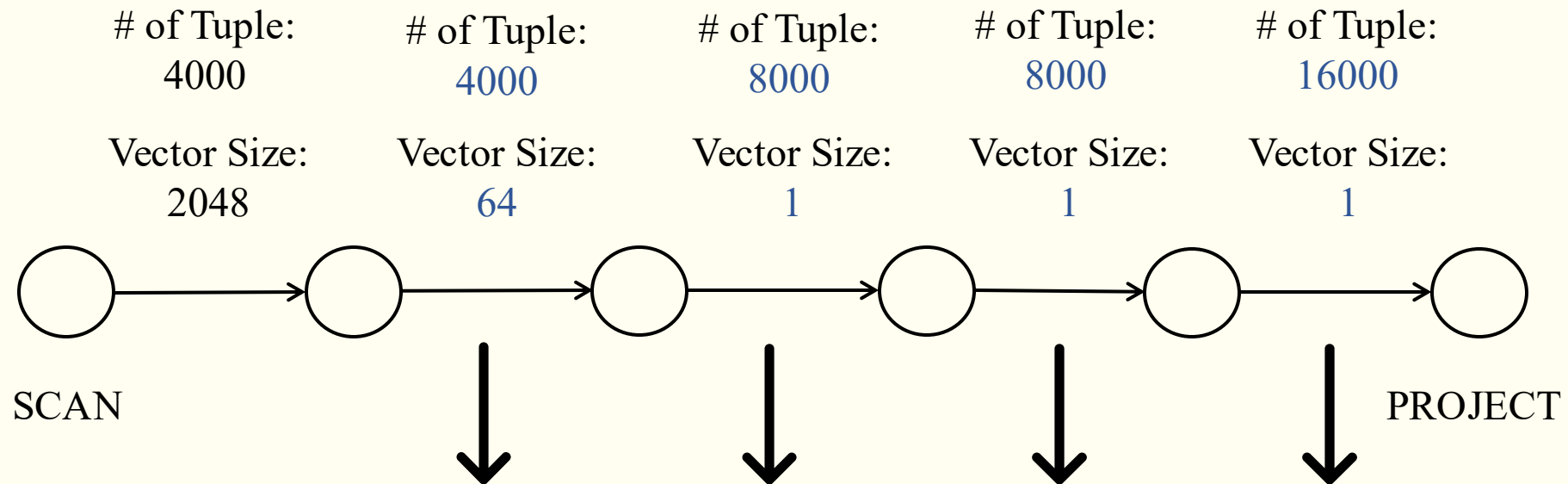
-> Three Small Chunks with 5 tuples!

# Small Chunk Problem



Chunk Size is Greatly Reduced During Execution, but **Data Volume is not Necessarily Reduced.**

# Small Chunk Problem

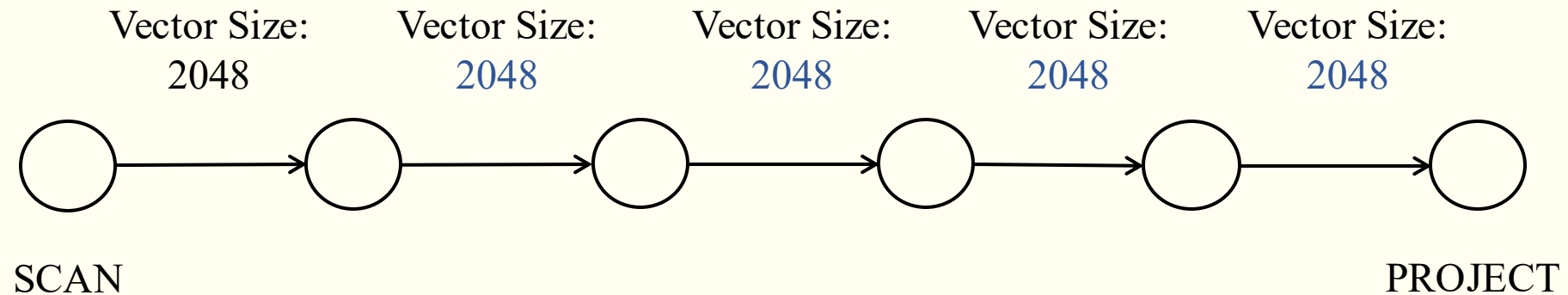


Given a Set of Data Chunks, Determine How to Compact Them to  
Minimize the Total Execution Time (= compaction time + compute time).



# Compact as More as Possible

- **Full Compaction:** compacts all chunks containing less than 2048 tuples.

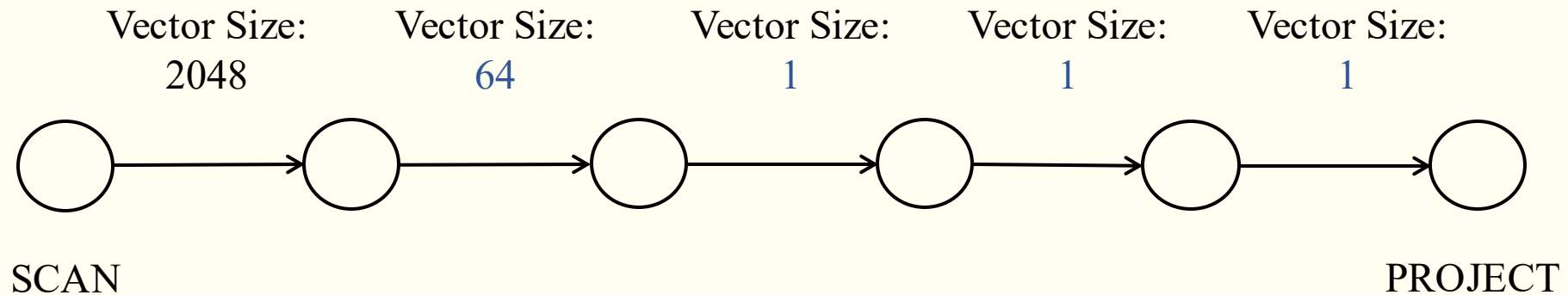


Too Many Memory Copies!

Consider the Case Where a Tuple is in the Length of 1 MB...

# Compact as Less as Possible

- **No Compaction**: no chunks are compacted



**Too Many Small Chunks!**

Consider the Case Where a Tuple is in the Length of **10 Bytes**...

# Compact Small Chunks Only

- **Binary Compaction**: chunks smaller than a predefined threshold are compacted.
- A Trade-off Between **Interpretation Overhead** vs. **Memory Copy**
  - The smaller the chunk, the greater the benefit from such compaction.

Choosing a Predefined Threshold is Difficult

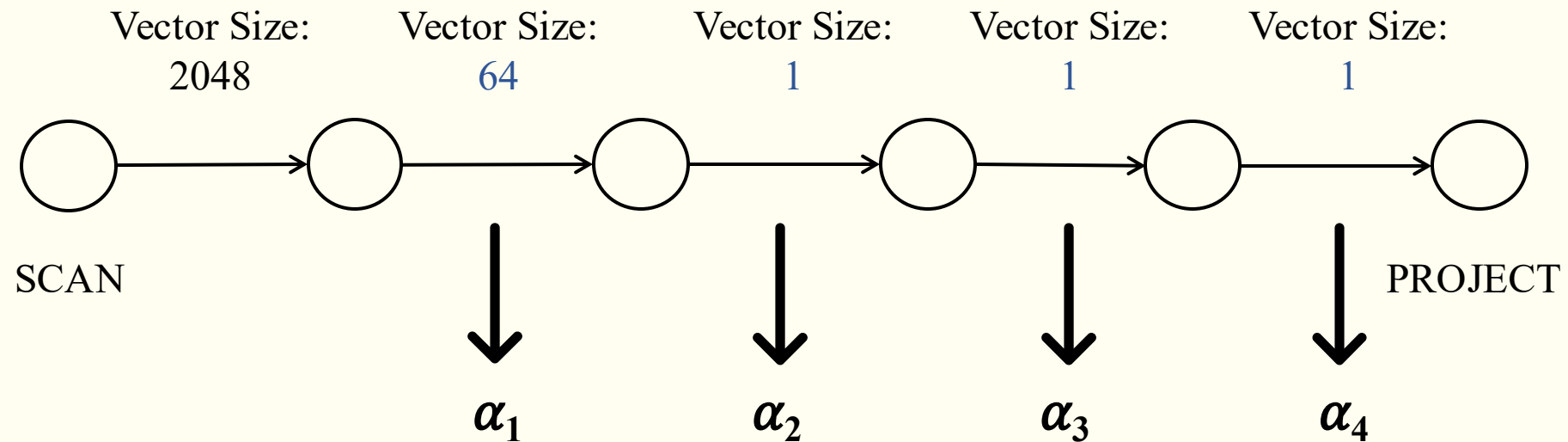
Because it involves Database Design and Workload Characteristics.

# Our Solution 1: Learning Compaction

- **Learning Compaction:** chunks smaller than a **Learned** threshold are compacted.
- **Morsel-driven Parallelism:** data is divided into chunks, with each thread responsible for fetching and processing a chunk through the entire pipeline before moving on to the next.
  - Then, each chunk can **serve as a sample** for a learning algorithm
- **Multi-armed Bandit Problem:** select the optimal thresholds

# Learning Compaction

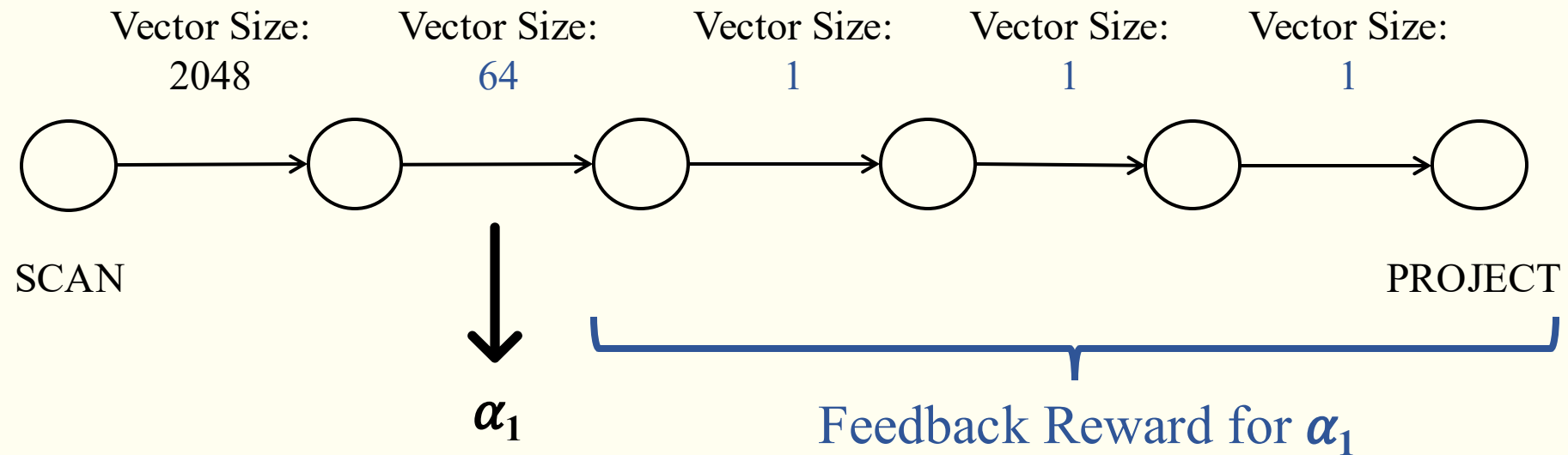
- **Learning Compaction**: chunks smaller than a **learned** threshold are compacted.



- **Multi-armed Bandit Problem**: select each alpha from 0, 32, 64, 128, ..., 2048

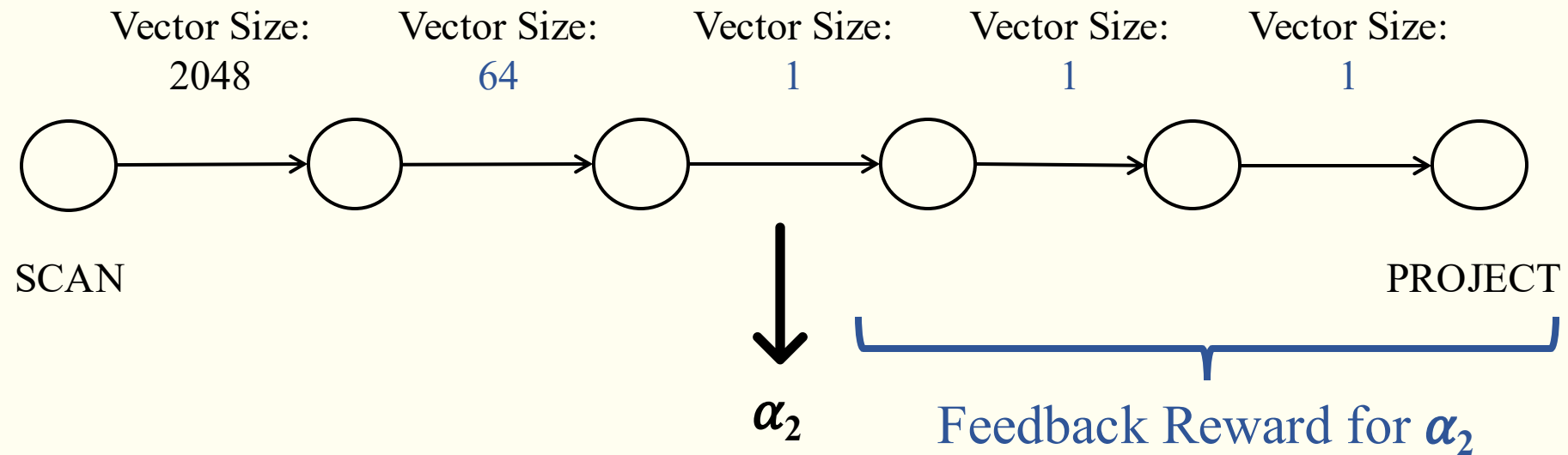
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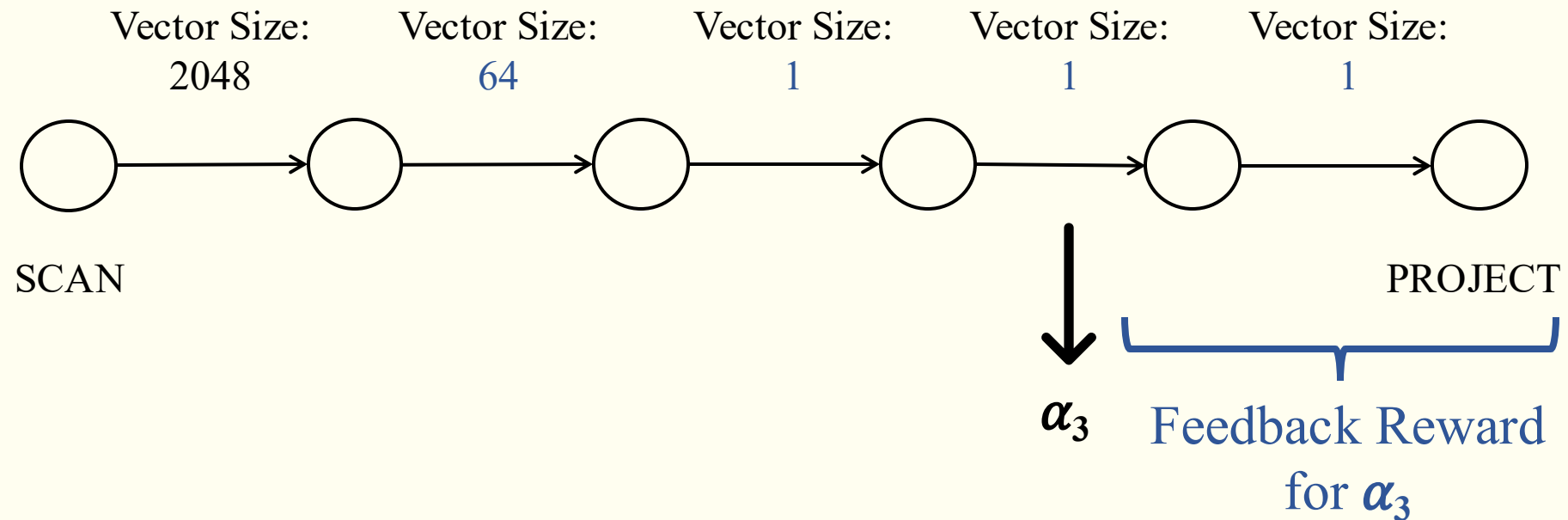
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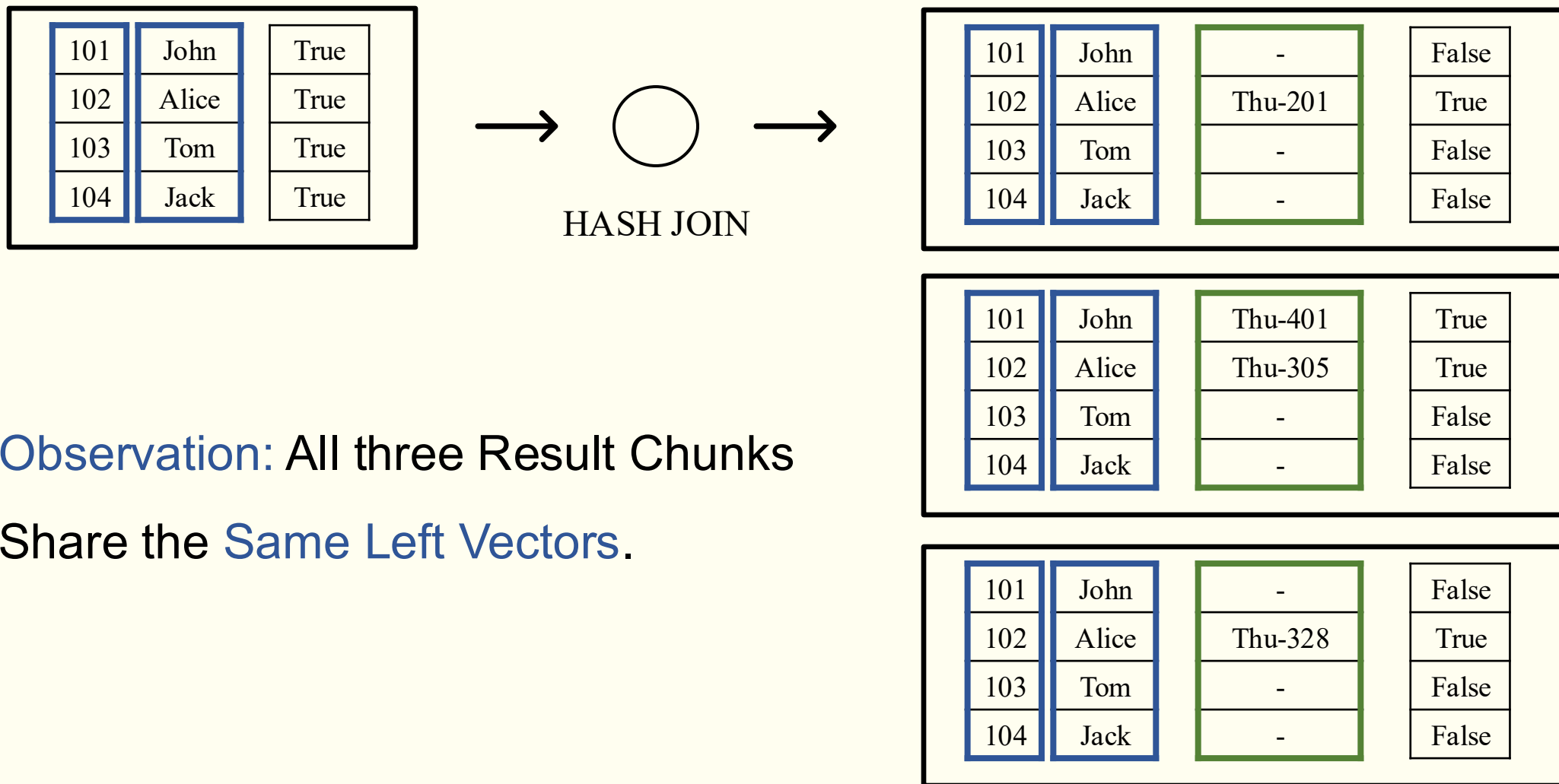
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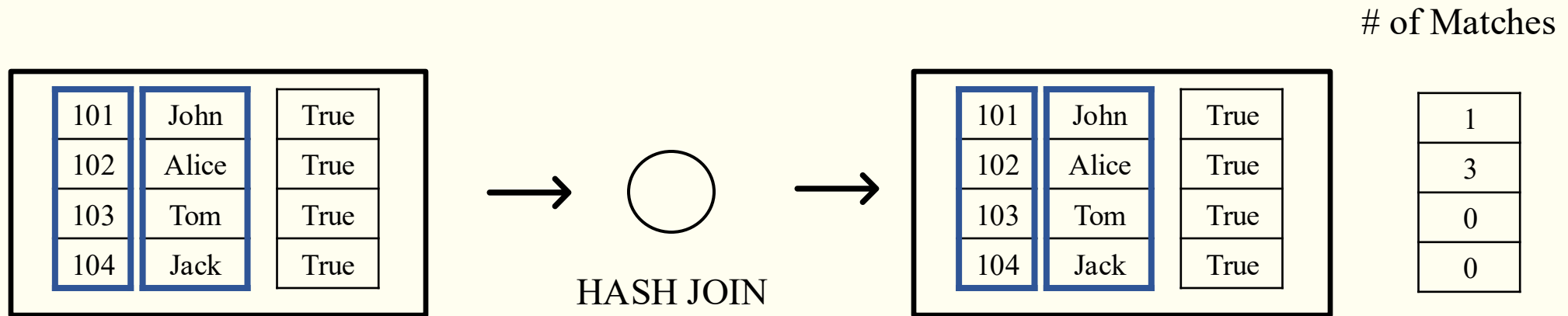




# Our Solution 2: Logical Compaction

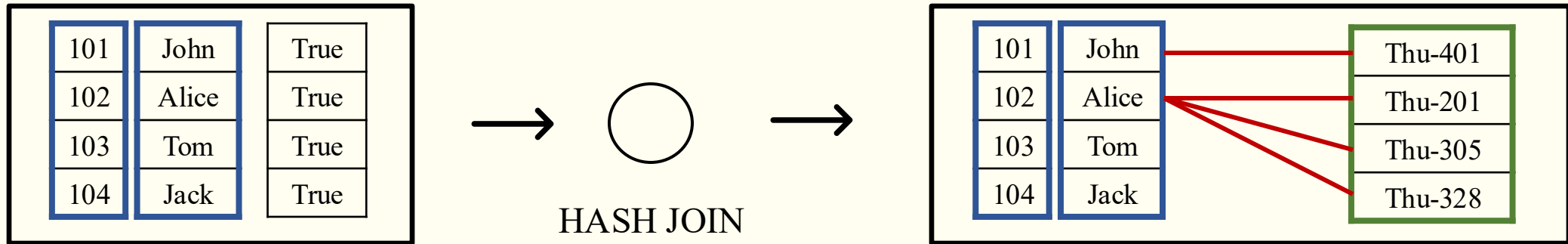


# Logical Compaction



Putting all generated tuples in one chunk seems to be a good idea, but  
how to keep the zero-copy property of the Left Side Vectors...

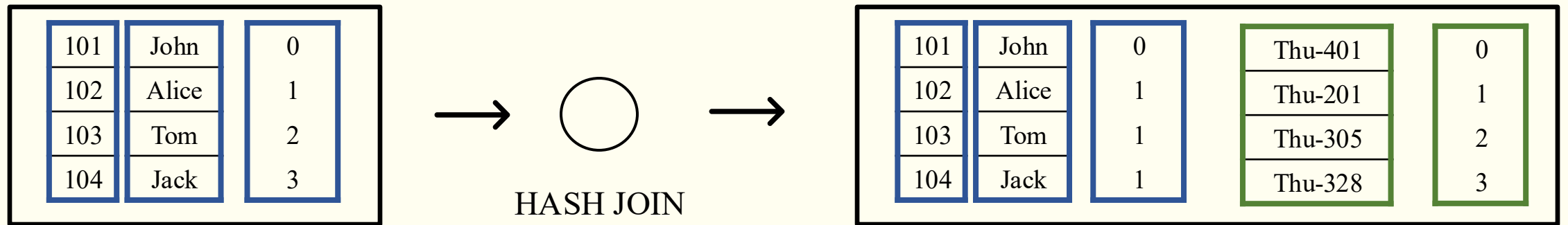
# Logical Compaction



Requirements:

- Left-side vectors should be **zero-copied**.
- The **mapping relationship** between the left-side and right-side vectors should be maintained.
- The resulting chunk can be an input chunk to other operators.

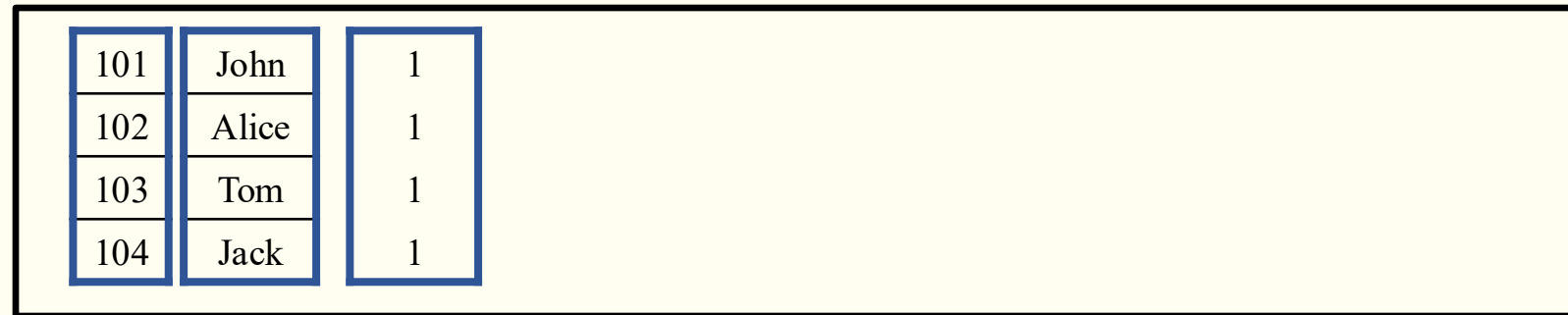
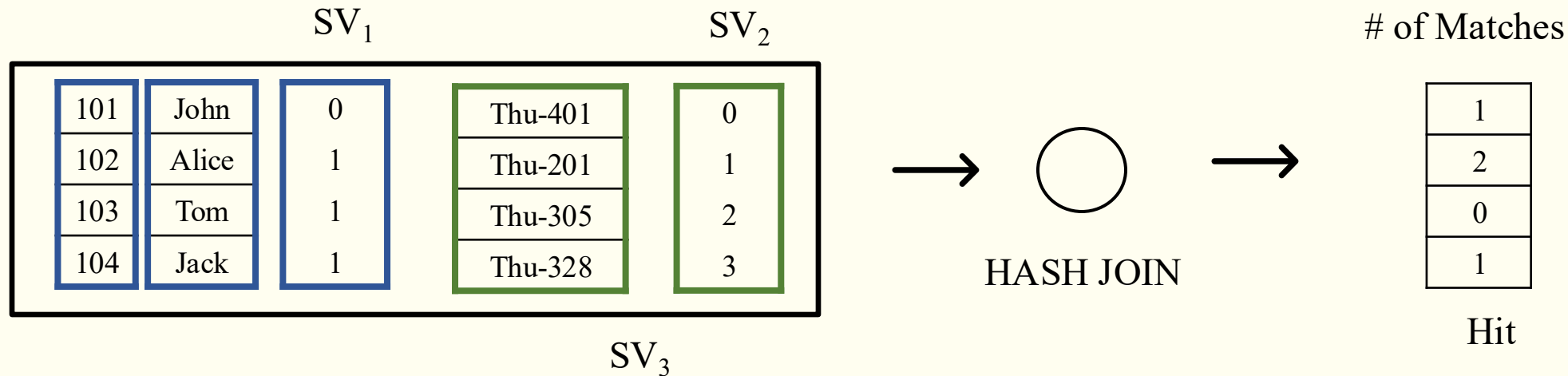
# Our Solution 2: Logical Compaction



Our Solution: **Selection Vector (SV) with Repeated Values:**

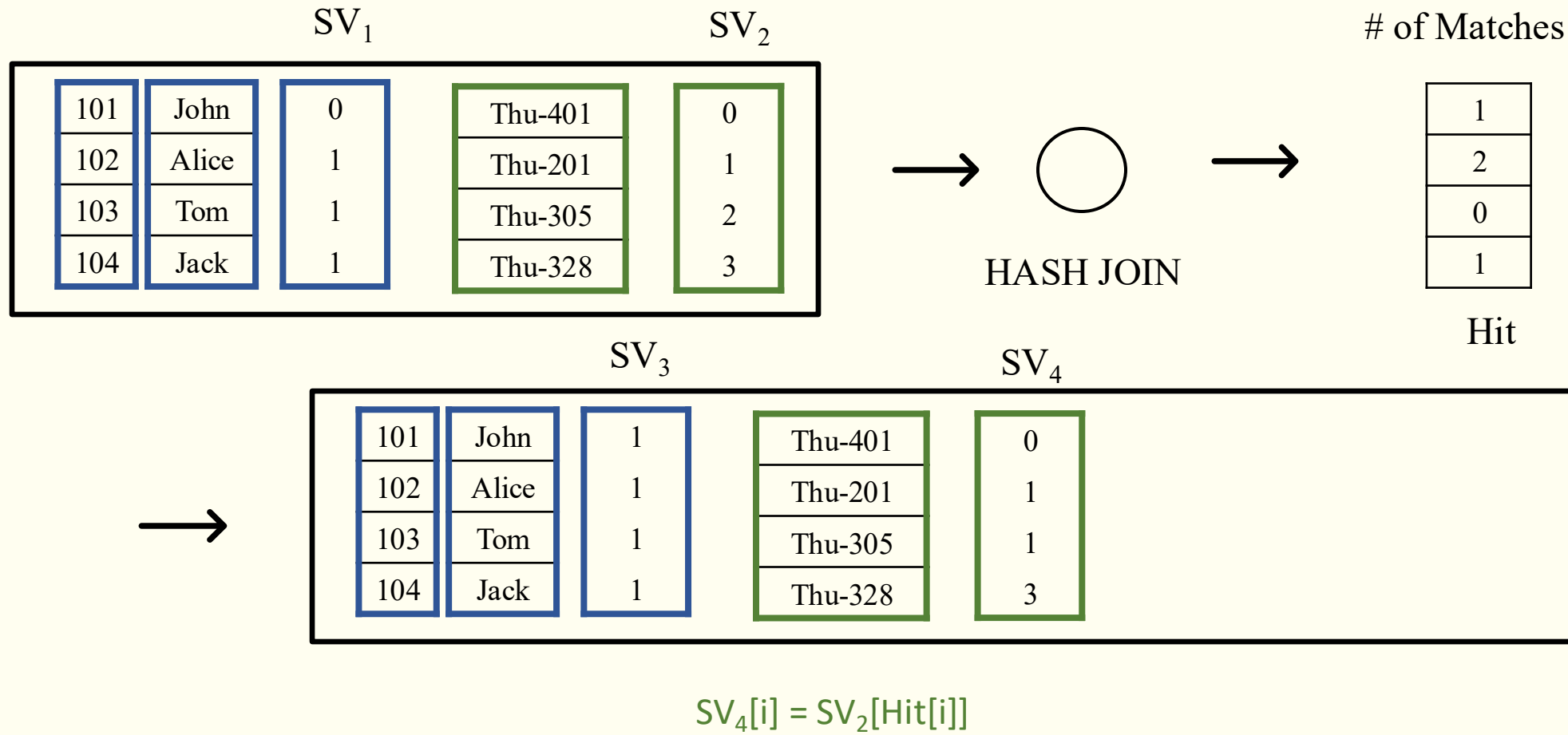
- Left-side vectors have an SV.
- Right-side vectors have another SV.

# Hash Probe the Result Chunk Again

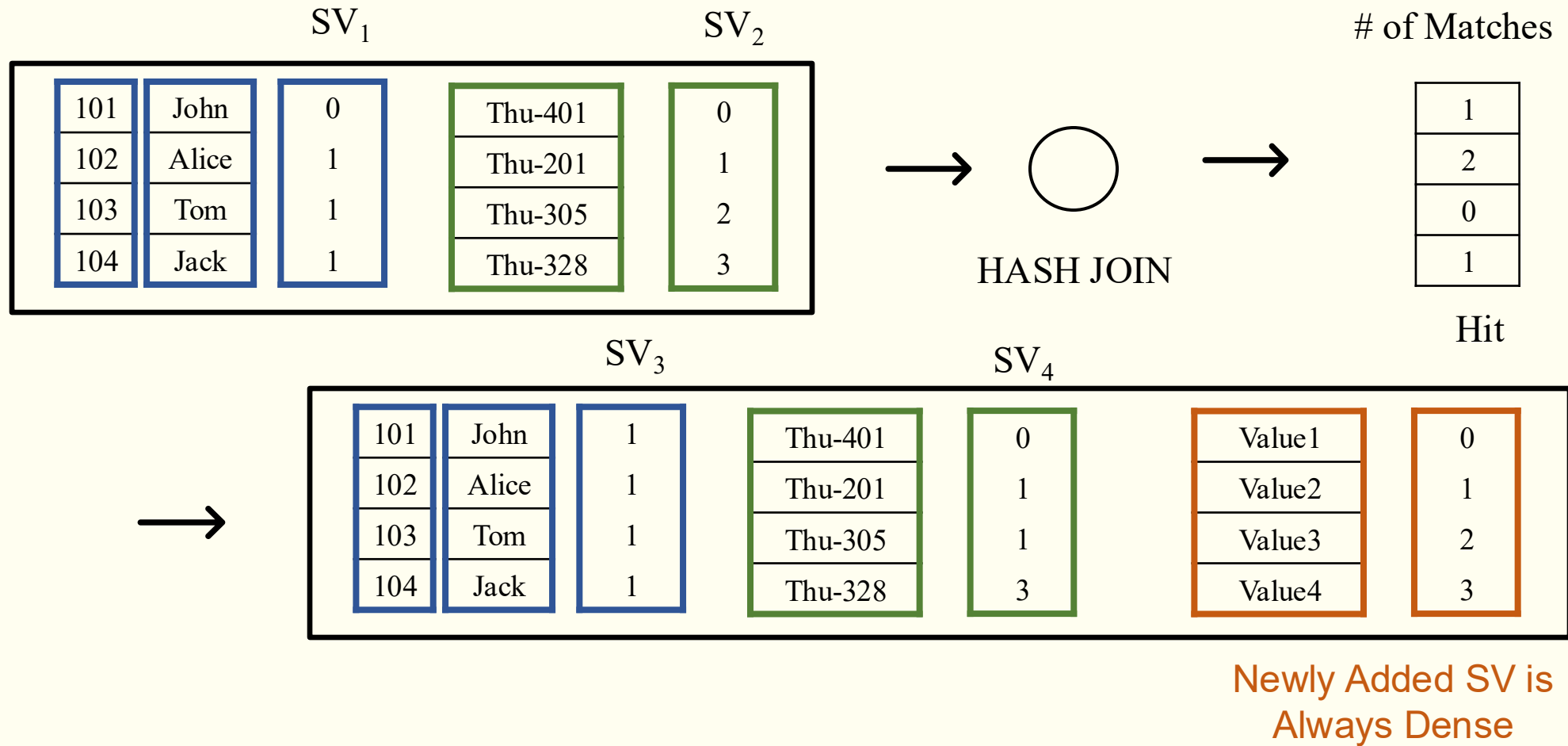


$$SV_3[i] = SV_1[Hit[i]]$$

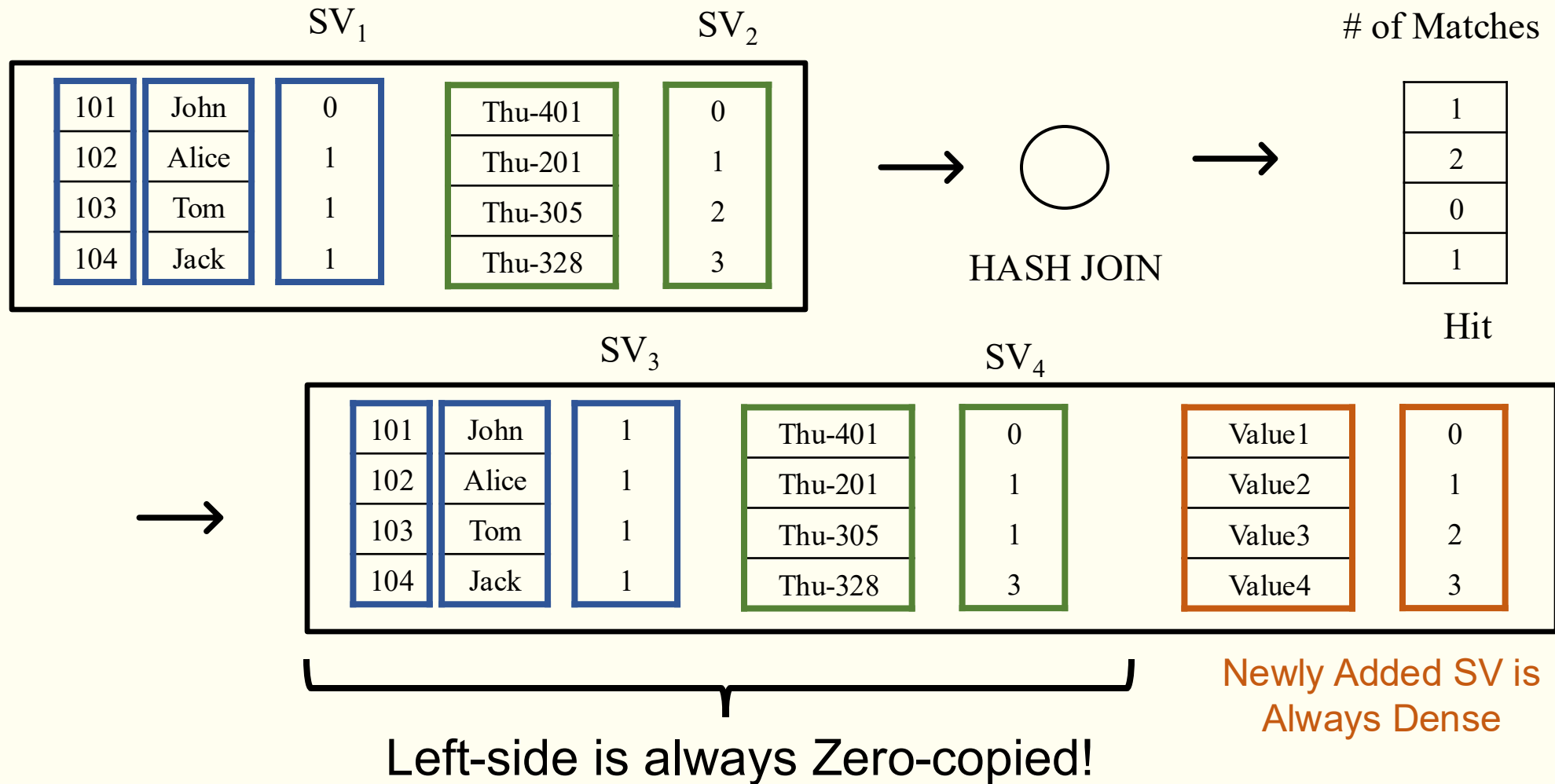
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# End-to-End Performance in DuckDB

- We integrate our solutions into DuckDB and measure the end-to-end performance.
- **Smart Compaction**: Learning Compaction + Logical Compaction

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