

# Implicit Order Join: Joining Log Data with Property Data by Discovering Implicit Order-oriented Keys with Human Assistance

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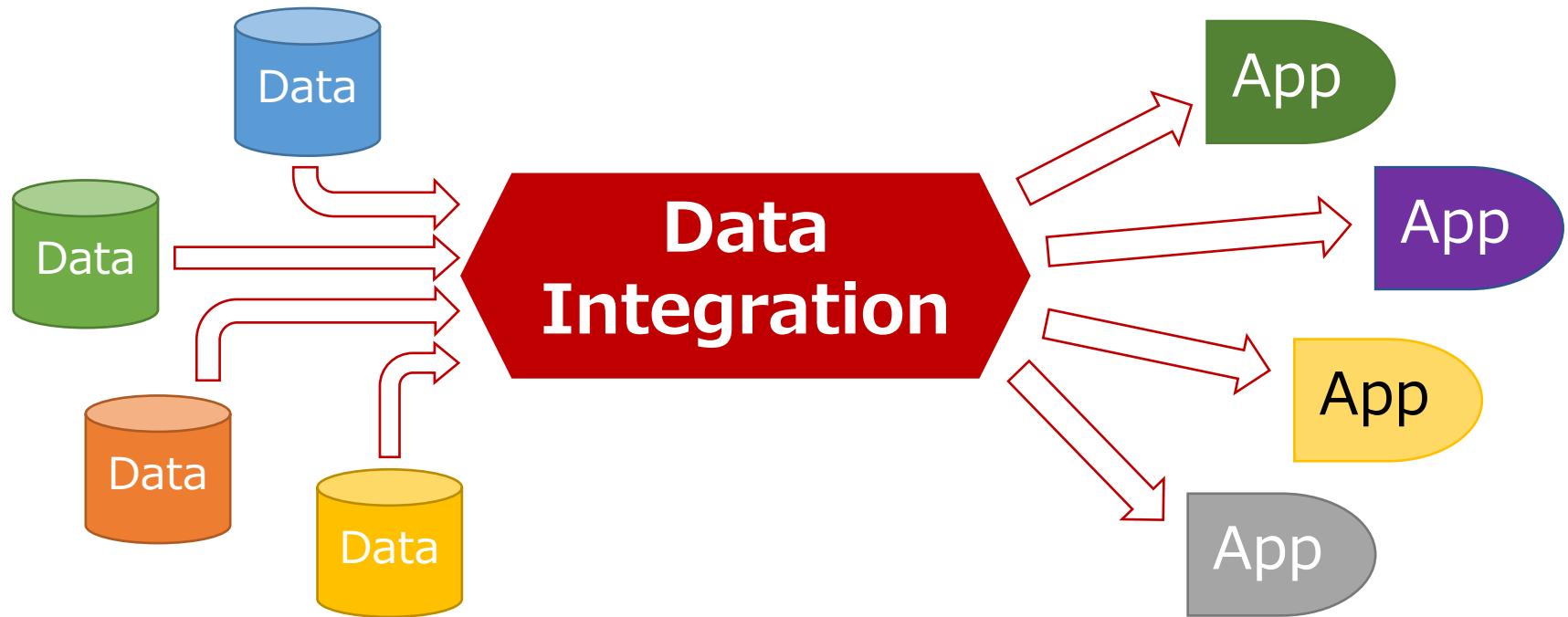
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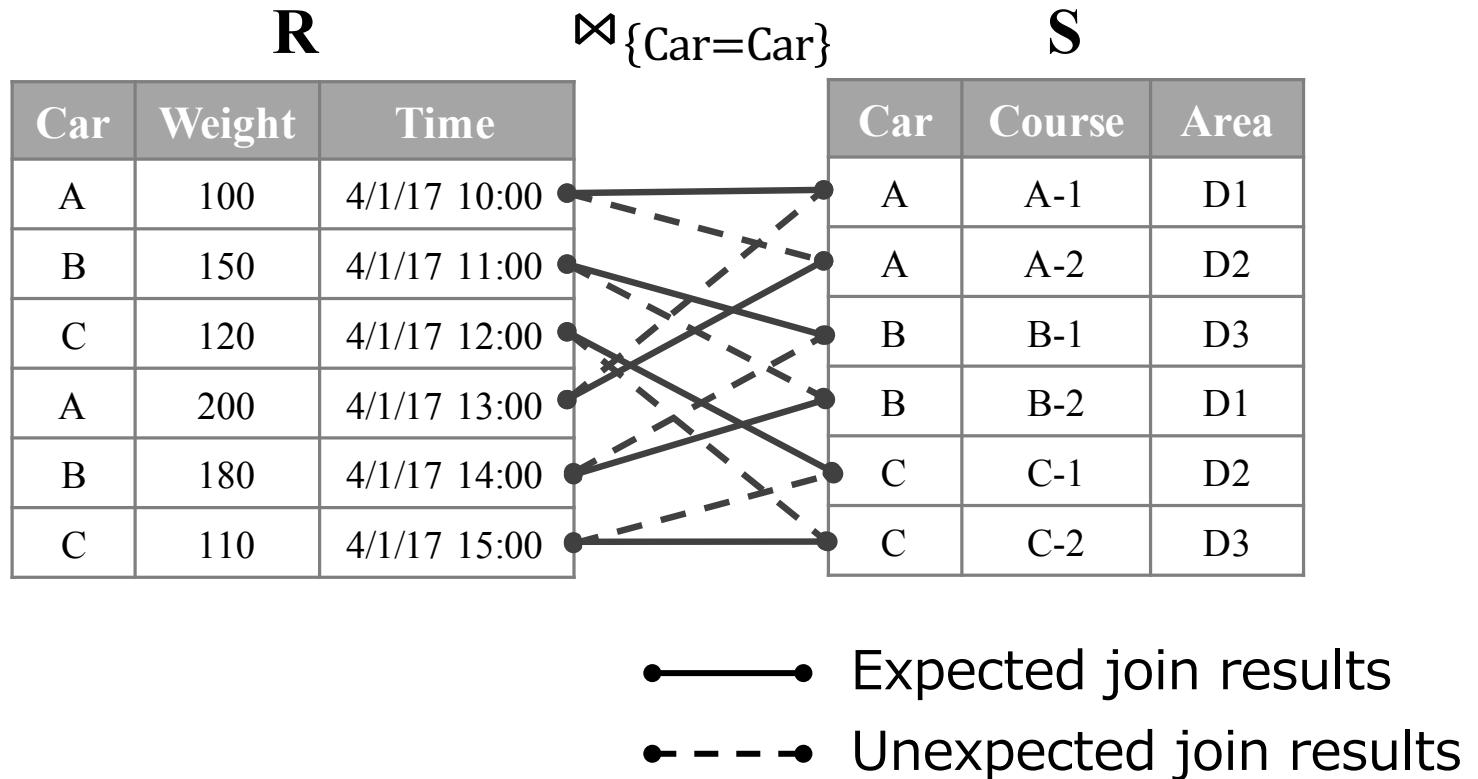
# Data Integration

- Fundamental task for data analysis
- Combining data from multiple sources



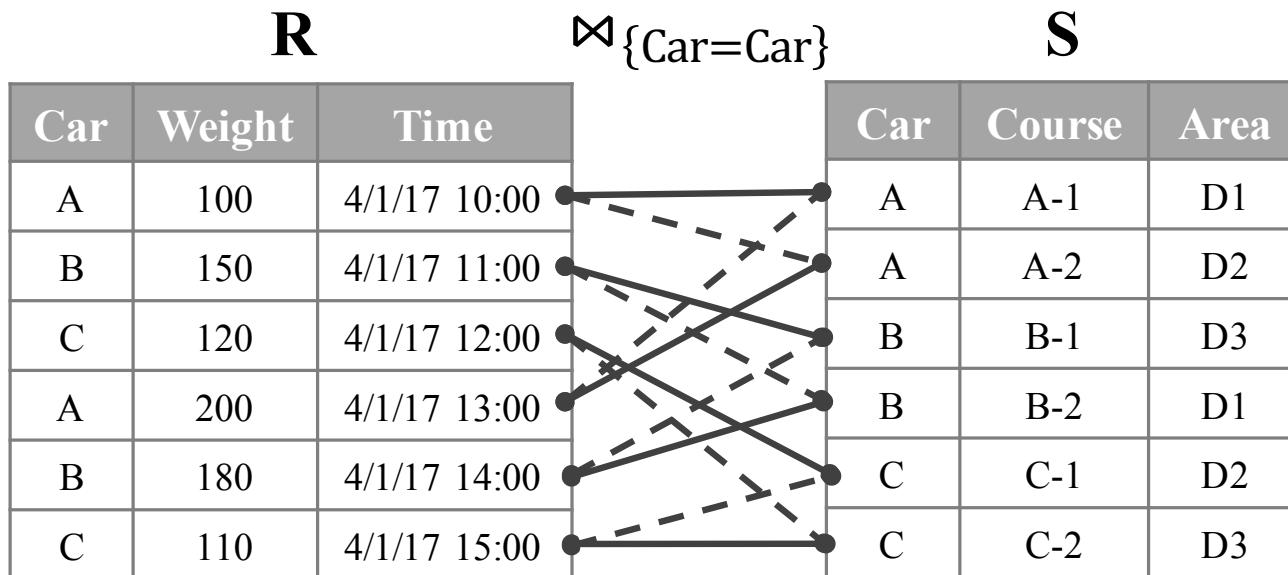
# Missing Key Problem

- Inconsistency of data



# Formal Definition

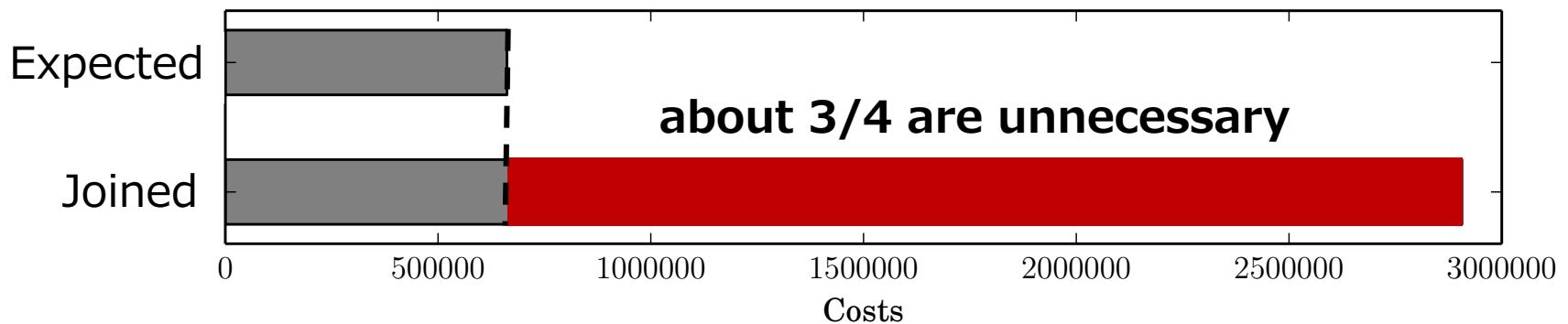
*Definition 1 (Missing Key Problem):* Given relations  $R$ ,  $S$ , join condition  $J$  and expected join results  $U^*$ , no query over  $R \bowtie_J S$  provides  $U^*$ , and there is no auxiliary relation which enables to join  $R$  and  $S$  to provide  $U^*$ .  $\square$



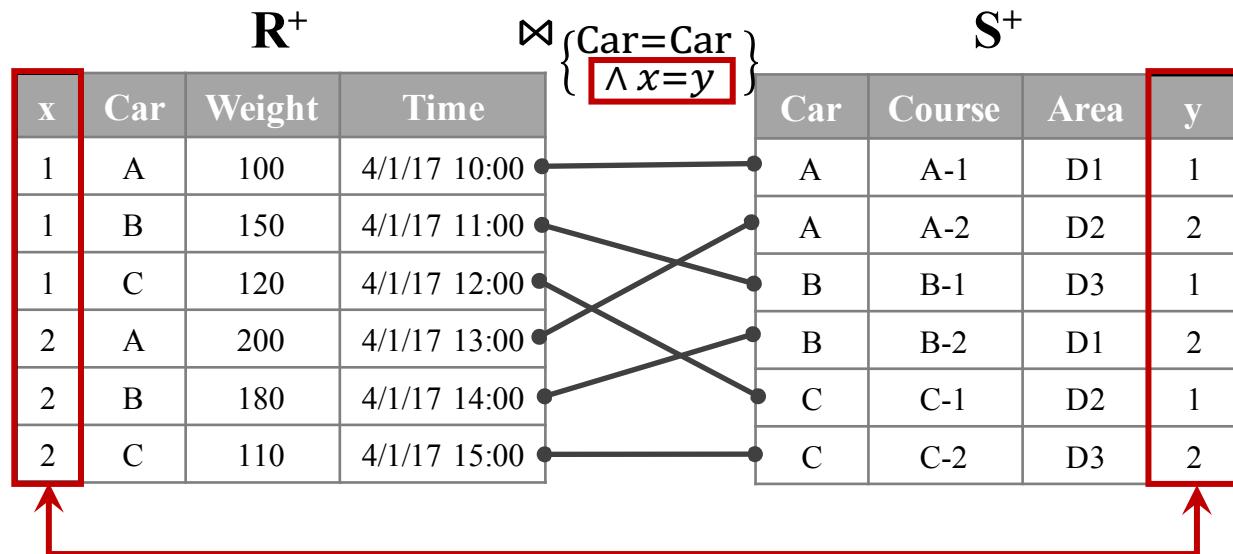
# Trouble from Missing Key Prob.

- Joined results include large number of unnecessary tuples.
- To use the results for applications, (automatic/manual) data cleansing is required.

An example situation

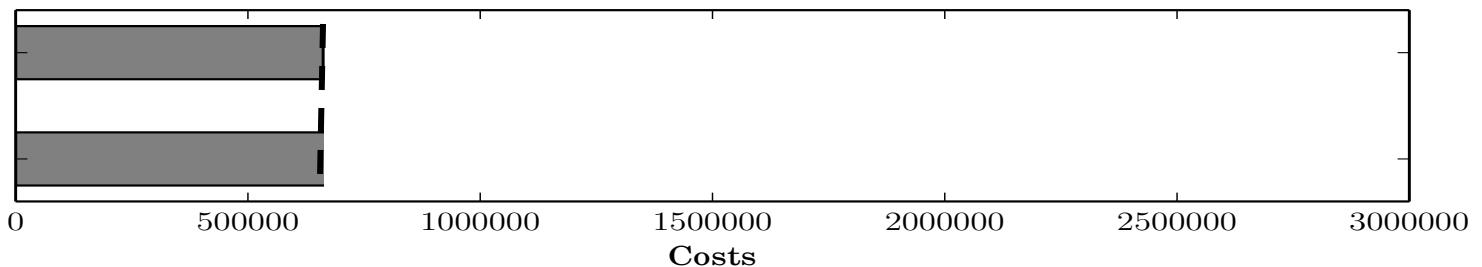


# Objective: Implicit Key Discovery



Expected

Joined



# Observation: Order-oriented Correlation

- Assumed real-world situation: Joining log records with supplemental information
  - e.g., garbage collection logs and collecting routes of garbage cars

Garbage collection log

Car	Weight	Time
A	100	4/1/17 10:00
B	150	4/1/17 11:00
C	120	4/1/17 12:00
A	200	4/1/17 13:00
B	180	4/1/17 14:00
C	110	4/1/17 15:00

Collecting routes in a day

Car	Course	Area
A	A-1	D1
A	A-2	D2
B	B-1	D3
B	B-2	D1
C	C-1	D2
C	C-2	D3

# Observation: Order-oriented Correlation

- Order-oriented correlation: an order of records in log data is corresponding with the of supplemental information.

Garbage collection log

Car	Weight	Time
A	100	4/1/17 10:00
B	150	4/1/17 11:00
C	120	4/1/17 12:00
A	200	4/1/17 13:00
B	180	4/1/17 14:00
C	110	4/1/17 15:00

Collecting routes

Car	Course	Area
A	A-1	D1
A	A-2	D2
B	B-1	D3
B	B-2	D1
C	C-1	D2
C	C-2	D3

Order-oriented correlation

# Tackling Issue

- Discovery of attribute set pair with order-oriented correlation with help of human judged samples

$\widehat{\mathbf{U}}^*$

Car	Weight	Time	Car	Course	Area
A	100	4/1/17 10:00	A	A-1	D1
B	150	4/1/17 11:00	A	A-2	D2
C	120	4/1/17 12:00	B	B-1	D3
A	200	4/1/17 13:00	B	B-2	D1
B	180	4/1/17 14:00	C	C-1	D2
C	110	4/1/17 15:00	C	C-2	D3

Human judged samples

**R**

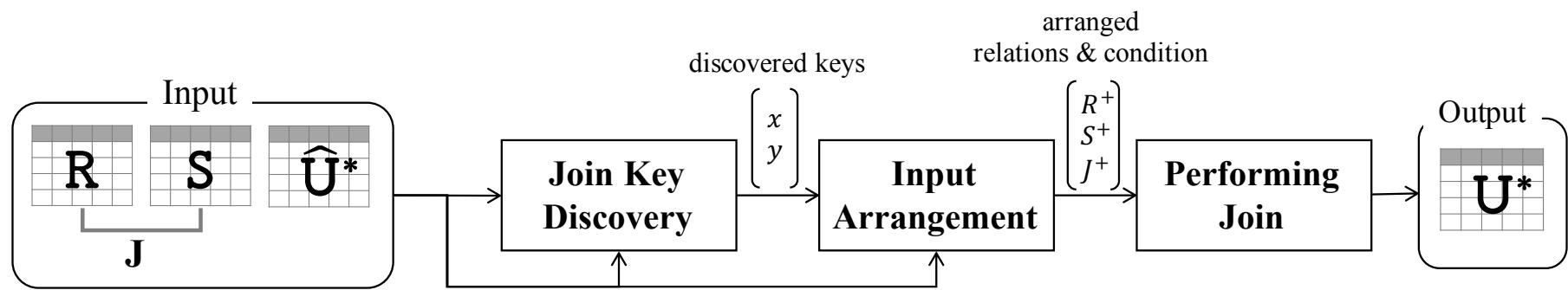
Car	Weight	Time
A	100	4/1/17 10:00
B	150	4/1/17 11:00
C	120	4/1/17 12:00
A	200	4/1/17 13:00
B	180	4/1/17 14:00
C	110	4/1/17 15:00

**S**

Car	Course	Area
A	A-1	D1
A	A-2	D2
B	B-1	D3
B	B-2	D1
C	C-1	D2
C	C-2	D3

Order-oriented correlation

# Implicit Order Join Framework



1. Discover order-oriented attribute pair.
2. Generate complemental attributes.
3. Arrange relations and join conditions.
4. Perform join operation.

# Combinatorial Problem

- Tremendous number of candidates of attribute set pairs.

$$\mathcal{O}(N_R!N_S!)$$

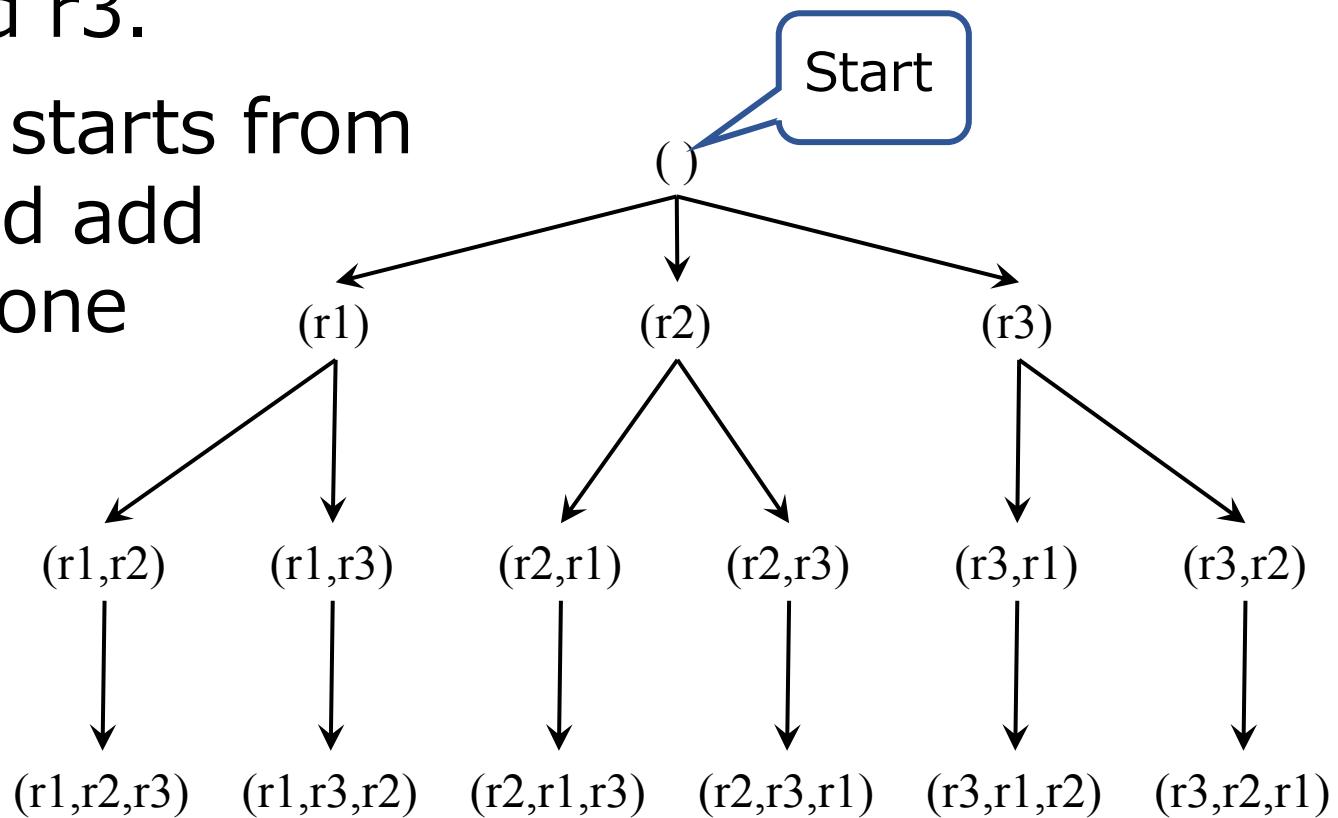
- where  $N_R$  (or  $N_S$ ) are the number of attributes of relation  $R$  (resp.  $S$ ).
- $N_X!$  is the number of enumerations of attributes in relation  $X$ .
- Taking subsequences into account, the number of each enumeration becomes  $\sum_{i=1}^N \binom{N}{i} i!$

# Pruning of Candidates

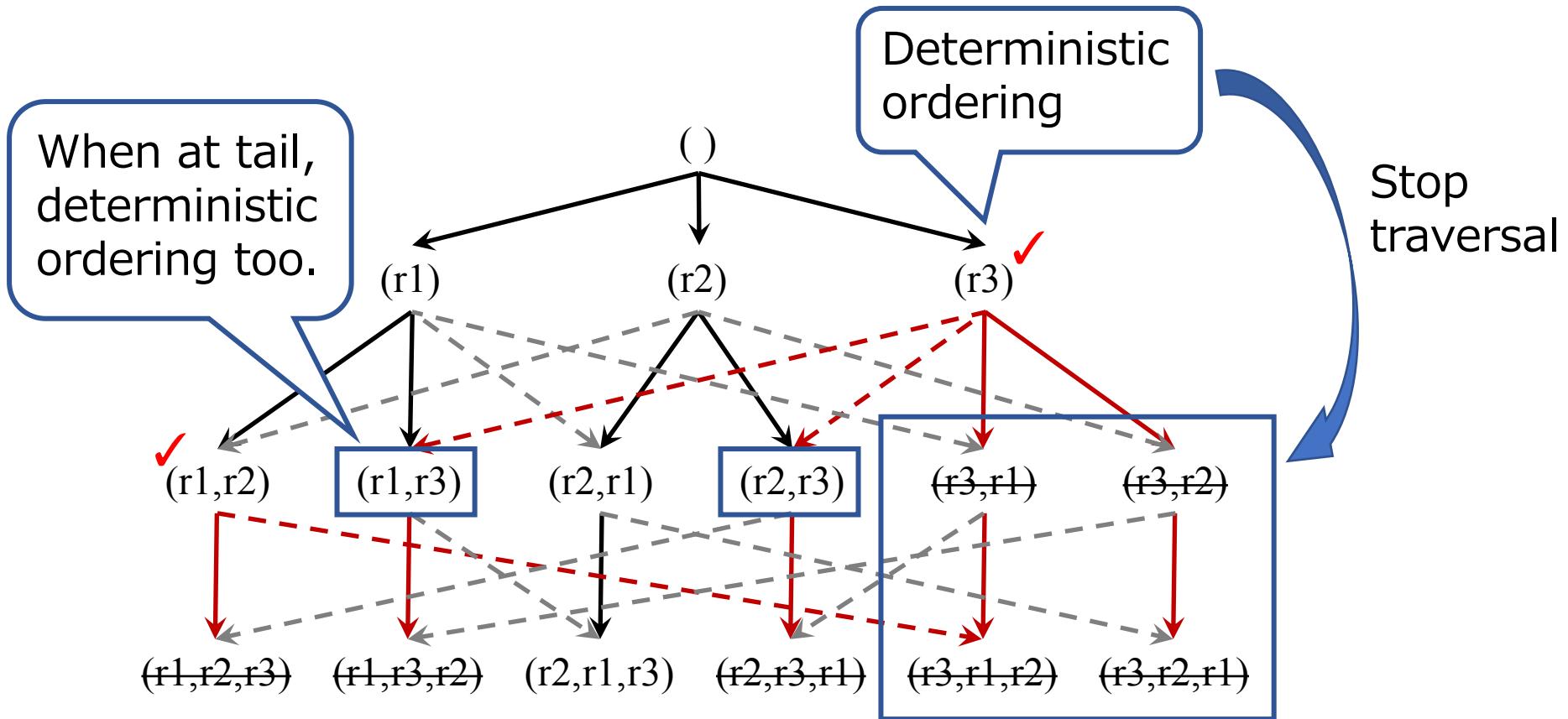
- Idea: a sequence of attribute gives deterministic ordering of records, super-sequences of it give the same ordering.
  - e.g., if  $(r_1, r_2) \rightarrow (d_1, d_2, d_3)$ ,  
then  $(r_1, r_2, r_3) \rightarrow (d_1, d_2, d_3)$
- Strategy
  - Bottom-up traversal
  - Stopping enumeration by the idea.

# Bottom-up Traversal

- Relation R has three attributes  $r_1$ ,  $r_2$  and  $r_3$ .
- Traversal starts from empty and add attribute one by one.



# Pruning

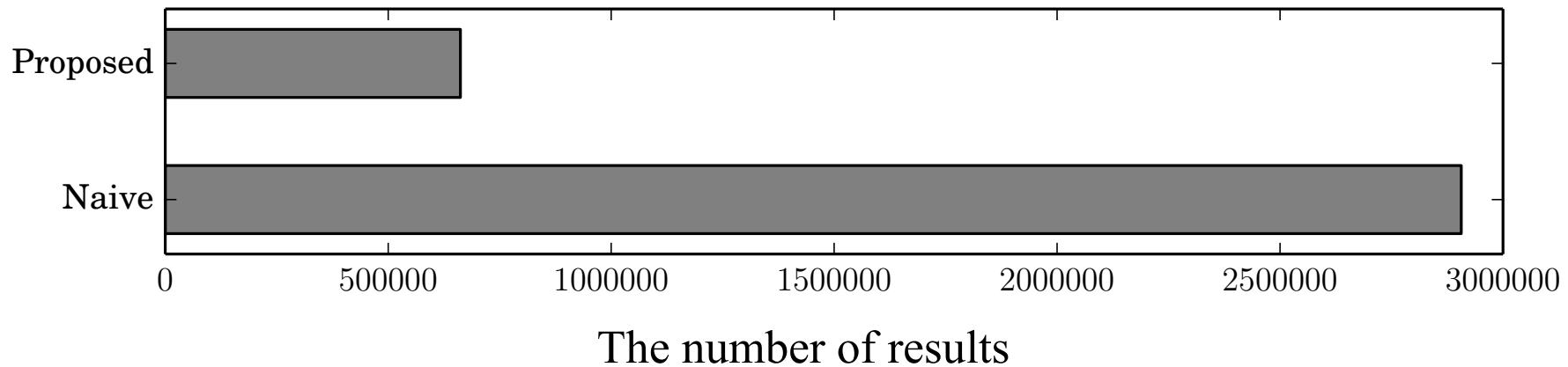


# Experimental Evaluation

- Objective
  1. Check effectiveness of the implicit order join.
  2. Check efficiency of the pruning.
- Datasets
  1. Real-world data from Fujisawa city, Japan.
    - Garbage collection logs and routing info.
  2. Synthetic data\*
    - Tunable parameters
      - #attributes: total number of attributes
      - #oo-attributes: size of order-oriented attribute set

\*<https://github.com/Taka-Coma/OOJBench>

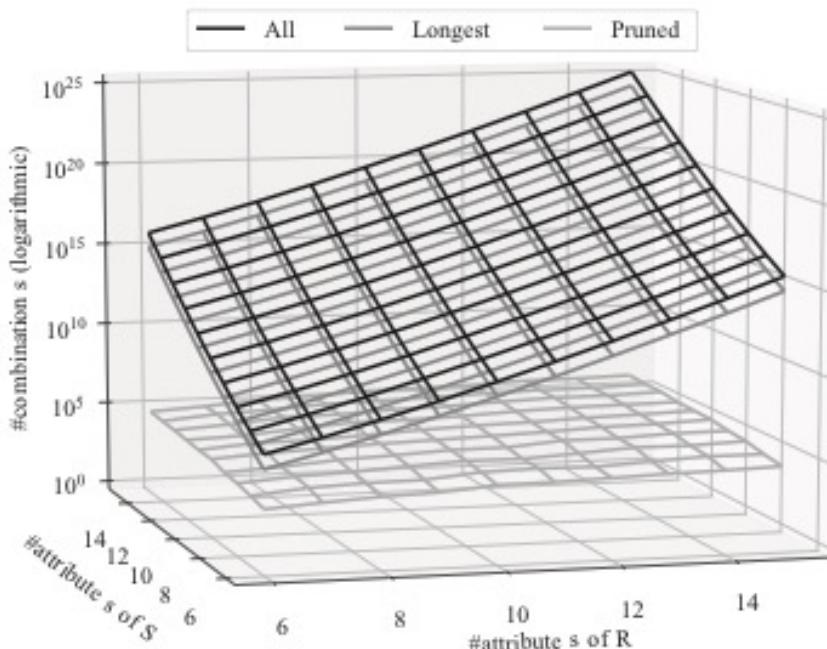
# Implicit Order Join is Effective.



- 77% reduction of joined results.
- Carefully checked by human judges that the results are correct.

# Efficiently prune for large # attrs.

Processing time in logarithmic scale



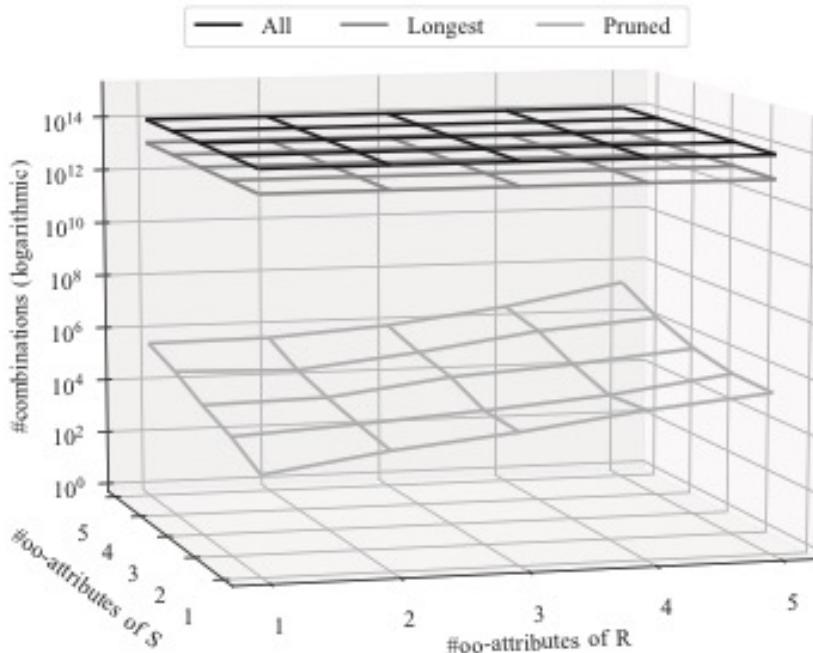
- The larger # attrs, the more # candidates in enumeration.
- Pruning effects big reduction of #candidates esp. when # attrs is large.

- **Baselines**

- all: enumeration of subsequences of attributes
- longest: enumeration of all attributes

# #oo-attrs affects performance.

## Processing time in logarithmic scale



- Baselines
  - all: enumeration of subsequences of attributes
  - longest: enumeration of all attributes
- The larger #oo-attrs, the more processing time.
- Still far better than baselines.

# Conclusion and Future Work

- Conclusion
  - Definition: Missing key problem
  - Proposal: Implicit order join framework
    - Order-oriented correlation assumption
  - Experiment: Effectiveness and Efficiency
- Future Work
  - General approach for implicit join
    - Removal of the assumption