ACM SIGMOD Programming Contest 2022

Team April



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1. Task Overview

Task: Build a blocking system for Entity Resolution.

Dataset:

#	Name	# of Rows	Blocking Requirement	
Dataset 1	Notebook	1 million	1 million	

3. Preprocessing

Using around 10 bash commands to normalize the data:

Data cleaning: Fix typos, drop punctuations, normalization.

A few substitution rules: replace synonyms in different language. (e.g., "mémoire" \rightarrow "memory")

Formatting: Adjust the text to the same format, lowercase

Dataset 2

Altosight 1 million

2 millions

Measurement:

 $Recall = \frac{\# of true \ matches \ in \ candidate \ set}{Total \ \# \ of \ true \ matches \ in \ ground \ truth}$

Evaluation Environment: Azure standard F16s v2 (16 CPU x 2.7 Ghz Processors, 32 Gb Main Memory, 32GB Storage)

2. System Architecture



letters, replace punctuations. (e.g., "USB 3" \rightarrow "usb3")

Tokenization: Tokenize records by whitespace.

4. Deduplication

Observation: many duplicate records in the datasets.

Deduplication:



Goal : make the following similarity join more time efficient

5. Acceleration



Our approach leverages similarity grouping, Jaccard

- 1. **Deduplicating** near-duplicate records
 - Total number of records reduced.

2. Parallelization

- Run the two datasets simultaneously.
- Run the two similarity joins simultaneously.
- 3. Efficient similarity join algorithms
 - Adopt the SOTA overlap set join algorithm [1].
 - Adopt the SOTA Jaccard similarity join algorithm [2].

6. Result

Average Recall: 0.520

Runtime: 1679 seconds

join and overlap join to efficiently find a pool of candidate record pairs.

Then, we rank all the candidate record pairs by their weighted similarities and output a fixed number of record pairs as the blocking result.

#	T ₁	T ₂	T ₃	Recall
Dataset 1	0.8	0.57	9	0.743
Dataset 2	0.9	0.57	7	0.297

Hindsight:

Could use the training data ground-truth to boost the recall

6. Conclusion & References

• Similarity join is a robust model for blocking that doesn't rely on heavy data preprocessing or fancy manual rule.

• Parameter setting (adjusting similarity join thresholds) is vital to improve recall, best parameters varies for different dataset.

[1]: Dong Deng, Guoliang Li, He Wen, Jianhua Feng: An Efficient Partition Based Method for Exact Set Similarity Joins. VLDB 2016, Proc. VLDB Endow. 9(4): 360–371 (2015)
[2]: Dong Deng, Yufei Tao, Guoliang Li: Overlap Set Similarity Joins with Theoretical Guarantees. SIGMOD Conference 2018: 905–920