

Efficient Parallel Partition based Algorithms for Similarity Search and Join with Edit Distance Constraints

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Similarity Search&Join Competition on EDBT/ICDT 2013

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- 1 Motivation
 - Problem Definition
 - Application
- 2 Our Approach
 - Pass Join Algorithm
 - Additional Filters
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- 3 Experiment
 - Evaluating Pruning Techniques
 - Evaluating Parallelism
 - Evaluating Scalability

Problem Definition

STRING SIMILARITY JOINS

Given a set of strings \mathcal{S} , the task is to find all pairs of τ -similar strings from \mathcal{S} . A program must output all matches with both string identifiers and distance τ . (Track II)

An Example

Table: A string dataset

| ID | Strings | Length |
|-------|-------------------|--------|
| s_1 | vankatesh | 9 |
| s_2 | avataresha | 10 |
| s_3 | kaushic chaduri | 15 |
| s_4 | kaushik chakrab | 15 |
| s_5 | kaushuk chadhui | 15 |
| s_6 | caushik chakrabar | 17 |

Consider the string dataset in Table 1.

Suppose $\tau = 3$. $\langle s_4, s_6 \rangle$ is a similar pair as $ED(s_4, s_6) \leq \tau$

Application

- Data cleaning
- Information Extraction
- Comparison of biological sequences
- ...

Basic Idea

Lemma

Given a string r with $\tau + 1$ segments and a string s , if s is similar to r within threshold τ , s must contain a segment of r .

Example

$\tau = 1$, $r = \text{"EDBT"}$ has two segments "ED" and "BT". $s = \text{"ICDT"}$ cannot be similar to r as s contains none of the two segments.

Even Partition Scheme

Definition

In even partition scheme, each segment has almost the same length. ($\lfloor \frac{|s|}{\tau+1} \rfloor$ or $\lceil \frac{|s|}{\tau+1} \rceil$)

Example

$\tau = 3$, we partition $s_1 = \text{"vankatesh"}$ into four segments "va", "nk", "at", "esh".

Substring Selection

Basic Methods

- Enumeration:
Enumerate all substrings for each of the segment.
- Length-based:
For each segment, only select substrings with same length.
- Shift-based:
For segment with start position p_i , select substrings with start position in $[p_i - \tau, p_i + \tau]$

Substring Selection

Position-aware Substring Selection

Observation

r_l r_r
 $r = \text{"vankatesh"}$

s_l s_r
 $s = \text{"avataresha"}$

$$||s_l| - |r_l|| + ||s_r| - |r_r|| = 2 + 3 > 3$$

Theorem (Position-aware Substring Selection)

For segment with start position p_i , select substrings with start position in $[p_i - \lfloor \frac{\tau - \Delta}{2} \rfloor, p_i + \lfloor \frac{\tau + \Delta}{2} \rfloor]$ where $\Delta = |s| - |r|$.

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Substring Selection

Position-aware Substring Selection

Example

$r = \text{"vankatesh"} \quad s = \text{"avataresha"}$

| | | |
|--------------|------------------|--|
| $p_1=1, va$ | \longleftarrow | $[1,3]: av \quad va \quad at$ |
| $p_2=3, nk$ | \longleftarrow | $[2,5]: va \quad at \quad ta \quad ar$ |
| $p_3=5, at$ | \longleftarrow | $[4,7]: ta \quad ar \quad re \quad es$ |
| $p_4=7, esh$ | \longleftarrow | $[6,8]: res \quad esh \quad sha$ |

$$\tau = 3, \Delta = 1, [p_i - \lfloor \frac{\tau - \Delta}{2} \rfloor, p_i + \lfloor \frac{\tau + \Delta}{2} \rfloor] = [p_i - 1, p_i + 2]$$

Substring Selection

Multi-match-aware Substring Selection

Observation

$r_l = ""$ r_r
 $r = \text{"vankatesh"} \rightarrow \{\text{va. nk, at, esh}\}$
 $s = \text{"avataresha"} \quad ||s_l| - |r_l|| = 1$
 s_l s_r

r_r has 3 segments to detect, 2 errors allowed

There must be another matching between r_r and s_r .

Theorem (Multi-match-aware Substring Selection)

For the i -th segment with start position p_i , select substrings within $[p_i - i, p_i + i] \cap [p_i + \Delta - (\tau + 1 - i), p_i + \Delta + (\tau + 1 - i)]$.

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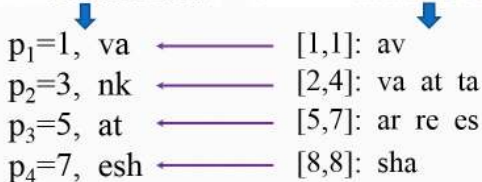
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Substring Selection

Multi-match-aware Substring Selection

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$r = \text{"vankatesh"} \quad s = \text{"avataresha"}$



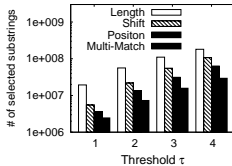
Substring Selection

Theoretical Results

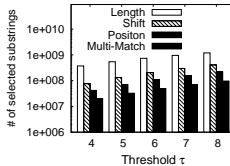
- 1 The number of selected substrings by the multi-match-aware method is minimum.
- 2 For strings longer than $2 * (\tau + 1)$, our selection method is the only way to select minimum number of substrings.

Substring Selection

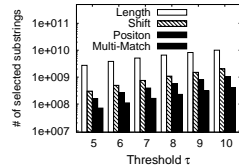
Experimental Results



(a) Author Name
 (Avg Len = 15)



(b) Query Log
 (Avg Len = 45)

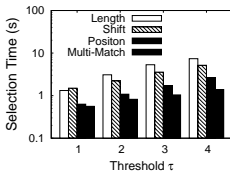


(c) Author+Title
 (Avg Len = 105)

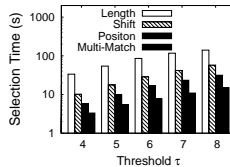
Figure: Numbers of selected substrings

Substring Selection

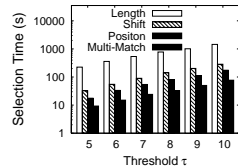
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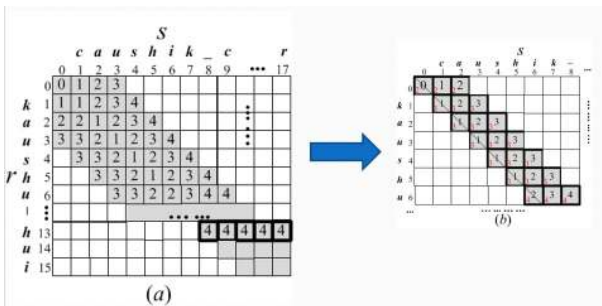


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Figure: Elapsed time for generating substrings

Verification

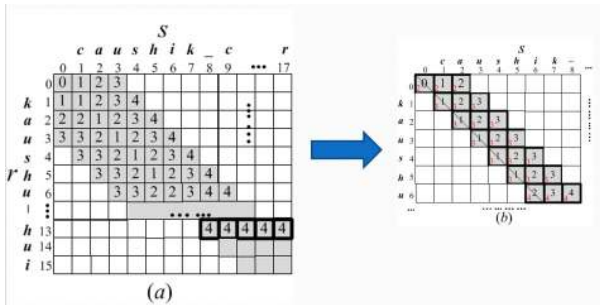
Length-aware Verification



- Inspired by the position-aware substring selection.
- Save at least half computation than traditional dynamic method.
- Save even more using improved early termination.

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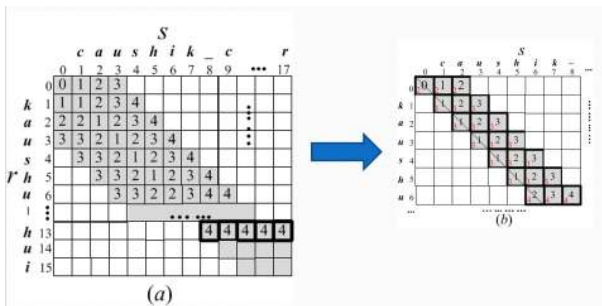
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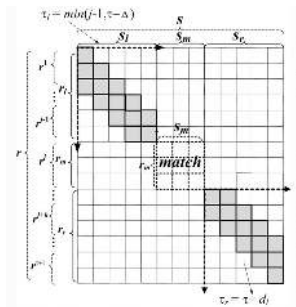
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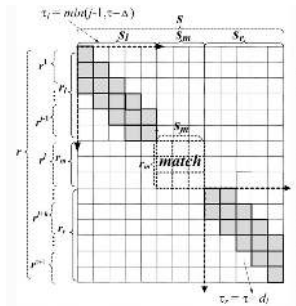
Extension-based Verification



- Inspired by the multi-match-aware substring selection.
- Using tighter thresholds to verify the candidate pairs.
- Verify if $ED(r_r, s_r) \leq \tau + 1 - i$ and $ED(r_l, s_l) \leq i - 1$.

Verification

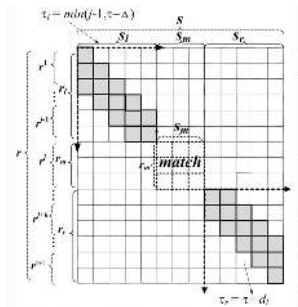
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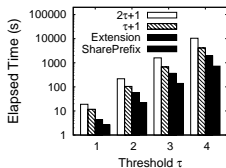
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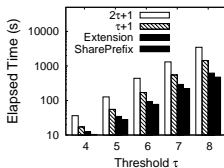
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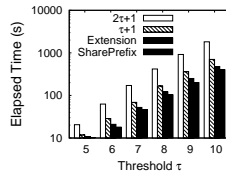
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Figure: Elapsed time for verification

Additional Filters

Effective Indexing Strategy

- Partition longer strings into segments.
- Select substrings from shorter strings.
- Longer segments decrease the possibility of matching.
- Thus decrease the number of candidates.

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Additional Filters

Content Filter

Observation

- Let \mathcal{H}_r denote the character frequency vector of r .
- $r = \text{"abyyyy"}, s = \text{"axxyyyxy"}$.
 $\mathcal{H}_r = \{\{a, 1\}, \{b, 1\}, \{y, 4\}\}, \mathcal{H}_s = \{\{a, 1\}, \{x, 3\}, \{y, 4\}\}$
- Let $\mathcal{H}_\Delta = |\mathcal{H}_r - \mathcal{H}_s|$.
- $\mathcal{H}_\Delta = |\mathcal{H}_r - \mathcal{H}_s| = ||1| + |-3|| = 4$.
- A deletion or insertion changes \mathcal{H}_Δ by 1 at most.
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Additional Filters

Content Filter

Observation

- *At most τ edit operations, $\mathcal{H}_\Delta \leq 2\tau$.*
- *At most $\tau - ||r| - |s||$ substitutions, $\mathcal{H}_\Delta \leq 2\tau - ||r| - |s||$.*
- *Group symbols to improve the content-filter running time.*
- *Integrate the content filter with the extension-based verification.*

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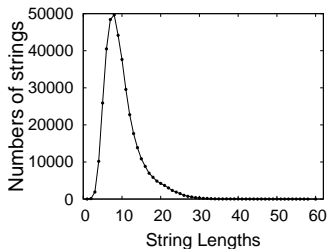
- 1 Parallel Sorting. Group strings by lengths using existing parallel algorithm.
- 2 Parallel Building Indexes. Parallel building indexes for each group.
- 3 Parallel Joins. Parallel perform similarity joins on each groups.

Experiment Setup

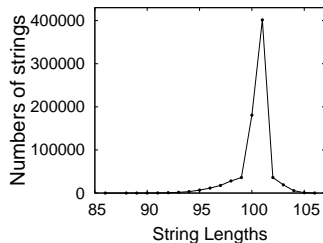
Table: Datasets

| Datasets | cardinality | average len | max len | min len |
|----------------|-------------|-------------|---------|---------|
| GeoNames | 400,000 | 11.106 | 1 | 60 |
| GeoNames Query | 100,000 | 10.7 | 2 | 43 |
| Reads | 750,000 | 101.388 | 86 | 106 |
| Reads Query | 100,000 | 101.2 | 88 | 116 |

Experiment Setup



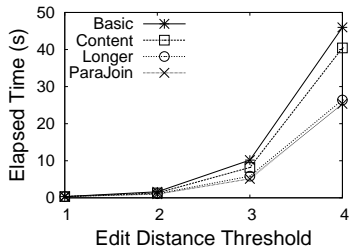
(a) GeoNames



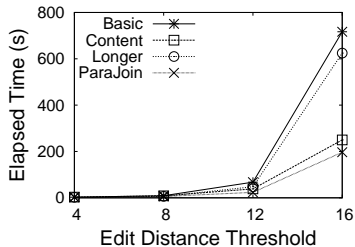
(b) Reads

Figure: Length Distribution.

Evaluating Pruning Techniques



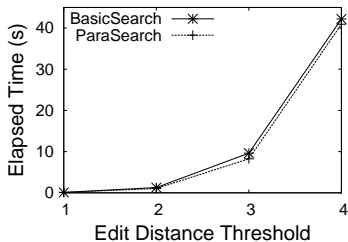
(a) GeoNames



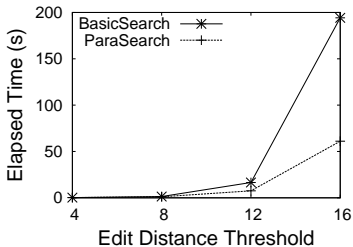
(b) Reads

Figure: Evaluating pruning techniques for similarity joins(8 threads).

Evaluating Pruning Techniques



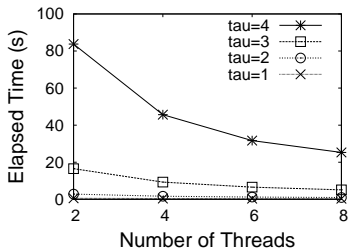
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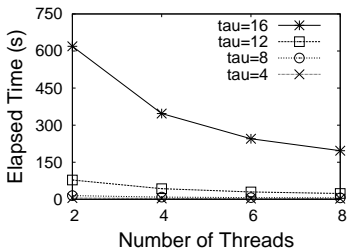
(b) Reads

Figure: Evaluating pruning techniques for similarity search(8 threads).

Evaluating Parallelism



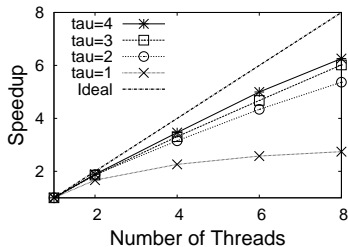
(a) GeoNames



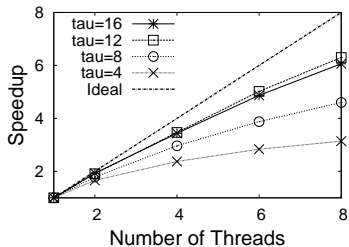
(b) Reads

Figure: Evaluating running time of similarity join by varying number of threads.

Evaluating Speedup



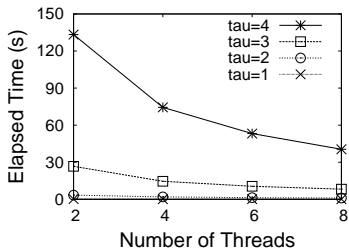
(a) GeoNames



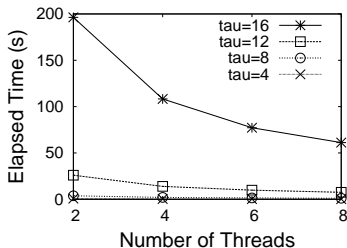
(b) Reads

Figure: Evaluating speedup of similarity join.

Evaluating Parallelism



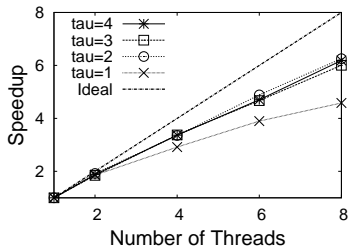
(a) GeoNames



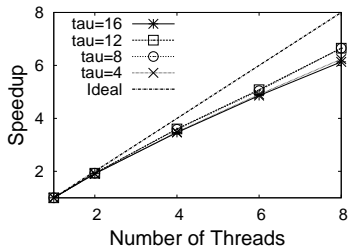
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Figure: Evaluating running time of similarity search by varying number of threads.

Evaluating Speedup



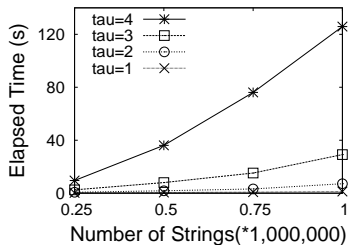
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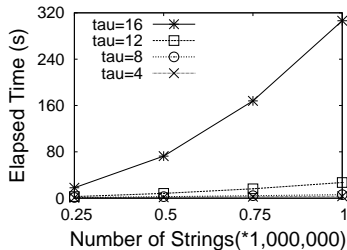
(b) Reads

Figure: Evaluating speedup of similarity search.

Evaluating Scalability



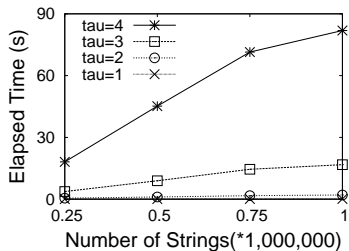
(a) GeoNames



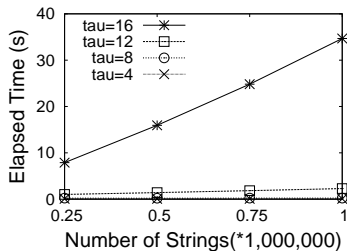
(b) Reads

Figure: Evaluating the scalability of the similarity join algorithm(8 threads).

Evaluating Scalability



(a) GeoNames



(b) Reads

Figure: Evaluating the scalability of the similarity search algorithm(8 threads).

About our team I

- We are from Tsinghua University, Beijing, China.
- Yu Jiang, Jiannan Wang, Guoliang Li, Jianhua Feng and Dong Deng.

About our team II



Thank You
Q & A

<http://dbgroup.cs.tsinghua.edu.cn/dd>

Pass-Join: A Partition based Method for Similarity Joins.
Guoliang Li, Dong Deng, Jiannan Wang, Jianhua Feng. VLDB
2012.