# A Flexible Approach for Planning Schema Matching Algorithms

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# Introduction



Figure: Discovering semantic correspondences between 2 schemas still a challenging issue in many applications

Semi automatic matchers combine several match algorithms to improve matching quality [Rahm and Bernstein, 2001, Euzenat et al., 2004]

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## Related Work

COMA++ [Aumueller et al., 2005]

- combination of many terminological measures and a user-defined synonym table
- a matrix is built for each couple of elements and for each measure
- a strategy is applied to select the mappings
- mappings are modified and/or validated by the user

Similarity Flooding [Melnik et al., 2002]

- a simple string matching algorithm to provide initial matchings
- structural rules and propagation to refine the matchings
- mappings are modified and/or validated by the user

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### Motivations

A brutal aggregation function entails drawbacks:

- $\bullet~$  quality  $\rightarrow~$  more weight to closely-related match algorithms can have a negative impact
- flexibility  $\rightarrow$  how to aggregate new match algorithms ?
- $\bullet~$  threshold  $\rightarrow~$  one threshold for each match algorithm instead of a global one
- **performance**  $\rightarrow$  useless measures are computed.

Recall vs precision:

- most matching tools promote precision
- easier to remove irrelevant discovered matches than finding relevant missed matches → recall seems a better choice

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# Contributions

Our approach: MatchPlanner:

- it is based on decision trees to combine match algorithms and avoid previous drawbacks.
- notion of planning in the schema matching process.
- a tool has been designed based on the planning approach.
- experiments demonstrate that our tool provides good performance and quality of matches w.r.t. the main matching tools.

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# MatchPlanner



**Input:** schemas to be matched a decision tree

**Algo:** for each pair of schema elements, match it with the decision tree.

**Output** : list of matches (optionnally validated by an expert)

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Figure: Examples of decision trees

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# Definitions

A decision tree contains plans (i.e ordered sequences) of match algorithms. More formally, it is a set of

- $\bullet\,$  internal nodes  $\rightarrow$  the match algorithms
- $\bullet\,$  edges between 2 nodes  $\rightarrow\,$  conditions on the result of match algorithms
- $\bullet~$  leaf nodes  $\rightarrow$  the relevance of the match

#### Features

- performance, in terms of discarded match algorithms
- **quality**, minimum F-measure obtained during training phase (for learned decision tres only)

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## Example of matching with a decision tree



Figure: How to match the pair of elements *(author, writer)* with this decision tree ?





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# Discussion

#### Advantages of the decision trees

- simple to understand or interpret (boolean logic).
- handles both numerical and categorical data.
- many related match algorithms cannot have a very strong impact on a similarity value, thus improving matching quality.
- threshold is specific for each match algorithm.
- applies only a subset of the match algorithms, thus improving performance.

#### Shortcoming

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### Experiments

Comparison with COMA++ and SF on two aspects:

- quality (precision, recall and F-measure)
- performance (time in seconds)

Seven scenarios:

- book and university (widely used in the literature)
- **thalia** (benchmark with the courses offered by some American universities)
- travel (airfare web forms)
- person (describing people)
- currency and sms (popular web services).

Quality Aspect Performance Aspec



Figure: COMA++ achieves the best precision in 5 scenarios

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Quality Aspect

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Figure: MP obtains the highest recalls (mostly above 60%) and it discovers all the relevant matches for 3 scenarios

Quality Aspect Performance Aspec



Figure: MP performs best on 6 scenarios

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Quality Aspect Performance Aspect



Figure: Time performance for matching each scenario

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MatchPlanner, a new schema matching approach

- based on decision trees to plan match algorithms
- flexible and it promotes recall
- outperforms the existing matching tools on the quality aspect
- provides an acceptable time performance

Ongoing work

- automatic generation of decision trees with machine learning techniques
- improving results with expert feedback
- comparing our approach with SMB ;-)



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