

Performance from Experience

Data Quality and Reconciliation

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Data Reconciliation

- Matching records
 - Detect duplication
 - Correct data inconsistencies
- Multi-disciplinary problem
 - Record linkage
 - Matching algorithms
 - Clustering algorithms
 - Joins in databases



Major Telephone Operating Company

Problem:

- Mergers and acquisitions of wireless companies resulted in the RBOC's inability to determine common customers among wireline and wireless businesses.
- Customer databases for each business unit use different schema and contain many quality problems
- RBOC's experience with commercial vendor's data reconciliation tool was unsatisfactory
- Solution:
 - Use small manually-verified data samples (~100 records) to determine appropriate matching rules
 - Use machine learning to prune the rules for efficient analysis of the large dataset
 - Resulted in 30% more correct matches than the commercial tool



Large Media Conglomorate

Problem:

- Company provides magazines to wholesalers who in turn provide magazines to resalers for distribution. Company loses money because of inconsistencies among wholesaler and retailer databases regarding number of sales.
- Reconciliation of wholesaler and retailer databases would make it easier to track where gaps in reporting are occurring.
- Identify 'bad' retailers.
- Solution:
 - Group by primary keys
 - Match by secondary keys
 - e.g. 3000 C.V.S. Pharmacies are grouped and compared by zip code and street address – identify 'bad' pharamacies



International Government

Problem:

- Reconcile vital taxpayer data from several different sources.
- Known problems include record duplication, address mismatches, address obsolescence, distributed responsibility for database accuracy and updates.
- Identify causes for mistakes
- Solution:
 - Improve the process flows and architecture to allow for rapid modification of pre-processing rules and matching rules.
 - Detection and classification of likey causes of duplication
 - Analysis and improvements reduced number of records that required manual verification.



ILEC-CLEC Billing Reconciliation

- Problem
 - ILECs charge CLECs for use of network resources
 - Verfication of actual usage vs. charging
 - E.g customer changing providers
 - Identify actual usage and send verification to ILEC
 - Resources identification in ILEC and CLEC are different
- Solution
 - Check charges in bill against actual usage
 - Common identification of resources (matching table)
 - Solution has only been implemented for access line charge



Framework

- Rules and solutions differ from domain to domain.
- Flexibility in resolution rules is essential
- Preprocessing and cleaning varies with the nature of the data.
- However, in most cases a common pattern emerges:
 - data is handled in stages,
 - each stage is responsible for one step of transformation (matching, merging, cleaning, etc.) and
 - Stages can be chained together to produce an overall data flow.
- We define a framework that allows
 - any number of custom-made processing blocks
 - Combination of those blocks in various ways



Framework

- data-flow architecture.
 - the framework enforces a simple producer-consumer model between pairs of functional blocks
- flexiblity and extensibility:
 - late binding of functional blocks to the framework
- blocks independence
 - component-like: they meet on common interfaces
 - introspection for run-time loading of new blocks.
- XML-based representation of the data flow.
 - Persistence of the flow state
- Data flow compilation:
 - the front-end can be thought of as a way for experts to configure the tool on a specific application.
 - Once configured, the data flow can be compiled into a "black-box" application that can be deployed to end users.



Demo

Demo of Tool



Framework Issues

- Other Architectures
- Enhancement of the simple producer-consumer model
 - pipelining where possible,
 - parallelize.
- Resource sharing vs blocks independence
- Separating data from algorithms
 - should the matching/cleaning algs be pushed close to the data e.g. DBMS
- Generating code from XML specifications
- Automation of rule generation, pruning



Conclusion

- Framework allows user-defined algorithms and rules
 - multiple clustering algorithms can be incorporated/chosen
 - type specific matching can be designed (e.g address)
- Extend tool to other domains
 - auto-discovery of networks and populating network databases
- Frequency of data reconciliation
- Metrics for data quality
- Performance of reconciliation process (real-time requirements)

