

# **Towards Rigorous Evaluation of Data Integration Systems**

# It's All About the Tools

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### Outline



- 1) Empirical Evaluation of Integration Systems
- 2) iBench
- 3) BART
- 4) Success Stories
- 5) Demo
- 6) Conclusions and Future Work





### Overview

- Challenges of evaluating integration systems
  - Diversity of tasks
    - Various types of **metadata** used by integration tasks
  - Quality is as important as performance
    - Often requires "gold standard" solution
- Goal: make empirical evaluations ...
  - ... more robust, repeatable, shareable, and broad
  - ... less **painful** and **time-consuming**
- This talk:
  - iBench a flexible metadata generator
  - BART generating data quality errors



### Overview



- Challenges of evaluating integration systems
  - Diversity of tasks
    - Various types of **metadata** used by integration tasks

Patterson [CACM 2012] "When a field has good benchmarks, we settle debates and the field makes rapid progress."

- iBench a flexible metadata generator
- BART generating data quality errors



# **Integration Tasks**



Many integration tasks work with metadata:

#### • Data Exchange

- Input: Schemas, Constraints, (Source Instance), Mappings
- *Output*: Executable Transformations, (Target Instance)
- Schema Mapping Generation
  - Input: Schemas, Constraints, Instance Data, Correspondences
  - Output: Mappings, Transformations
- Schema Matching
  - Input: Schemas, (Instance Data), (Constraints)
  - Output: Correspondences
- Constraint-based Data Cleaning
  - Input: Instance Data, Constraints
  - Output: Instance Data
- Constraint Discovery
  - Input: Schemas, Instance Data
  - Output: Constraints
- Virtual Data Integration
  - Input: Schemas, Instance Data, Mappings, Queries
  - Output: Rewritten Queries, Certain Query Results
- ... and many others (e.g., Mapping Operators, Schema Evolution, ...)



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  - Output: Mappings, Transformations
  - Schema Matching

#### Inputs/Outputs

#### Metadata: Schemas, Constraints, Correspondences, Mappings

**Data:** Source Instance, Target Instance

**Constraint Discovery** 

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- Input: Schemas, Instance Data
- Output: Constraints
- Virtual Data Integration
  - Input: Schemas, Instance Data, Mappings, Queries
  - Output: Rewritten Queries, Certain Query Results
  - ... and many others (e.g., Mapping Operators, Schema Evolution, ...)



## State-of-the-art



- How are integration systems typically evaluated?
- Small real-world integration scenarios
  - Advantages:
    - Realistic ;-)
  - Disadvantages:
    - Not possible to scale (schema-size, data-size, ...)
    - Not possible to vary parameters (e.g., mapping complexity)
- Ad-hoc synthetic scenarios
  - Advantages:
    - Can influence scale and characteristics
  - Disadvantages:
    - Often not very realistic metadata
    - Diversity requires huge effort



## Requirements



- We need tools to generate inputs/outputs
  - Scalability
    - Generate large integration scenarios efficiently
    - Requires low user effort
  - Control over metadata and data characteristics
    - Size
    - Structure
    - ...

6

- Generate inputs as well as gold standard outputs
- Promote reproducibility
  - Enable other researchers to regenerate metadata to repeat an experiment
  - Support researchers in understanding the generated metadata/data
  - Enable researchers to reuse generated integration scenarios





- **STBenchmark** [Alexe et al. PVLDB '08]
  - Pioneered the **primitive** approach:
    - Generate metadata by combining typical micro scenarios
- Data generators
  - PDGF, Myriad
  - Data generators are not enough



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8

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# iBench Overview



- **iBench** is a metadata and data generator
- Generates synthetic integration scenarios
  - Metadata
    - Schemas
    - Constraints
    - Mappings
    - Correspondences
  - Data
- "Realistic" metadata



**Integration Scenarios** 

10



- Integration Scenario
  - $-\mathbf{M} = (\mathbf{S}, \mathbf{T}, \boldsymbol{\Sigma}_{\mathbf{S}}, \boldsymbol{\Sigma}_{\mathbf{T}}, \boldsymbol{\Sigma}, \mathbf{I}, \mathbf{J}, \boldsymbol{A} \blacksquare)$





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- Integration Scenario
  - $-\mathbf{M} = (\mathbf{S}, \mathbf{T}, \boldsymbol{\Sigma}_{\mathbf{S}}, \boldsymbol{\Sigma}_{\mathbf{T}}, \boldsymbol{\Sigma}, \mathbf{I}, \mathbf{J}, \boldsymbol{A} \blacksquare)$
  - Source schema S with instance I
  - Target schema T with instance J
  - Source constraints  $\boldsymbol{\Sigma}_S$  and target constraints  $\boldsymbol{\Sigma}_T$ 
    - Instance I fulfills  $\boldsymbol{\Sigma}_S$  and instance J fulfills  $\boldsymbol{\Sigma}_T$
  - Schema mapping  $\Sigma$ 
    - Instances (I,J) fulfill  $\Sigma$
  - Transformations 🖪 🖪



# iBench Input/Output



- Inputs Configuration
  - Scenario parameters  $\Pi$  (min/max constraints)
    - Number of source relations
    - Number of attributes of target relations

• .

- Primitive parameters
  - Template micro-scenarios that are instantiated to create part of the output

#### • Output

- A integration scenario **M** that fulfills the constraints of specified in the configuration
  - XML file with metadata
  - CSV files for data



# Example - MD Task



#### • Input

Parameter 🖪 🕟	Source	Target
Number Relations	2-4	1-3
Number Attributes	2-10	2-10
Number of Join Attr	1-2	1-2
Number of Existentials		0-3

- Example solution (mappings)
- S1(A,B,C),S2(C,D,E) -> T(A,E)
- S3(A,B,C,D),S4(E,A,B) → ∃X,Y,Z T1(A,X,X), T2(A,Y,C),T3(C,B,Y,Z)



# Example - MD Task



#### • Input

Parameter 🖪 🕟	Source	Target
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- Example solution (mappings)
- S1(A,B,C),S2(C,D,E) -> T(A,E)
- $S3(A,B,C,D),S4(E,A,B) \rightarrow \exists X,Y,Z T1(A,X,X),$ T2(A,Y,C),T3(C,B,Y,Z)
- Limited usefulness in practice

- Can we generate "realistic" scenarios?

# Mapping Primitives



### Mapping Primitives

- Template micro-scenarios that encode a typical schema mapping/evolution operations
  - Vertical partitioning a source relation
- Used as building blocks for generating scenarios
- Comprehensive Set of Primitives
  - Schema Evolution Primitives
    - Mapping Adaptation [Yu, Popa VLDB05]
    - Mapping Composition [Bernstein et al. VLDBJ08]
  - Schema Mapping Primitives
    - STBenchmark [Alexe, Tan, Velegrakis PVLDB08]
      - First to propose parameterized primitives



# Scenario Primitives



#### **Example Mapping Primitives**





- Parameterize primitives
  - Number of relations for partitioning
  - Number of attributes for invention



14

...

# Integration Scenario Generation



### • Approach

- Start with empty integration scenario
- Repeatedly add instances of primitives according to specs
- If necessary add additional random mappings and schema elements





# **Primitive Generation**



- Example Configuration
  - I want 1 copy and 1 vertical partitioning



16

# **Primitive Generation**



- Example Configuration
  - I want 1 copy and 1 vertical partitioning

Source	ource Target	
$\mathbf{Cust}$	Customer	
Name ——	→ Name	
$\_$ Addr —	→ Addr	



# **Primitive Generation**



- Example Configuration
  - I want 1 copy and 1 vertical partitioning







- Sharing across primitives
  - Primitives cover many patterns that occur in the real world
  - however in the real world these primitives do not occur in isolation
- Enable primitives to share parts of the schema
  - Scenario parameters: *source reuse, target reuse*
  - Probabilistically determine whether to reuse previously generated relations



# Sharing Schema Elements



• Example





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# **User-defined Primitives**



- Large number of integration scenarios have been shared by the community
  - Amalgam Test Suite (Bibliographic Schemas)
    - Four schemas 12 possible mapping scenarios
  - Bio schemas originally used in Clio
    - Genomics Unified Schema GUS and BioSQL
  - Many others (see Bogdan Alexe's archive)
- User defined primitive (UDP)
  - User encodes scenario as iBench XML file
  - Such scenarios can then be declared as UDPs
    - Can be instantiated just like any build-in primitive



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## Motivation



- Evaluating constraint-based data cleaning algorithms
  - Need dirty data (and gold standard)
  - Algorithms are sensitive to type of errors
- Need a tool that
  - Given a clean DB and set of constraints
  - Introduces errors that are detectable by the constraints
  - Provides control over how hard the errors are to repair (repairability)



#### Overview







• Benchmarking Algorithms for data Repairing and Translation





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  - open-source error-generation system with an high level of control over the errors





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- Input: a clean database wrt a set of data-quality rules and a set of configuration parameters





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- Output: a dirty database (using a set of cell changes) and an estimate of how hard it will be to restore the original values

#### 22

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- Benchmarking Algorithms for data Repairing and Translation
  - open-source error-generation system with an high level of control over the errors
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#### 22

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# **Creating Violations**



Stadium

Juventus Stadium

**BMO** Field

Juventus Stadium

Yankee St.

Juventus Stadium

Allianz Arena

Goals

3

23

5

0

5

3

Player

Team

Juventus

Toronto

Juventus

N.Y. City

Juventus

Bayern

- Constraint language: denial constraints
  - Subsumes FDs, CFDs, editing rules, ...
- Update values of a cell to create a violation of a constraint
  - -t2.Team ='Juventus'

dc: -(	Player(n,	s, t, st,	g),	Player(n', s'	, t′,	st', g'),	t=t', s	st≠st'
		-, -, -,	011		/ - /	/ 0 //	, -	



15			
t, g), Player(n', s	s', t', st', g')	, t=t', st ≠ st	_/

Name

Giovinco

Giovinco

Pirlo

Pirlo

Vidal

Vidal

t1

t2

t3

t4

t5

t6

Season

2013-14

2014-15

2014-15

2015-16

2014-15

2015-16





- Error generation is an NP-complete problem
  in the size of the DB
- How to identify cells to change efficiently?
- How to avoid interactions among introduced constraint violations?



# **Error Generation**



• Our approach

#### - Sound, but not complete

#### - Avoid interactions among cell changes

- Once we decide on a cell change to introduce a violation we exclude other cells involved in the violation from future changes
- Vio-Gen queries
  - Derived from detection queries for denial constraints
  - Find cell to update such that the update is guaranteed to introduce a violation
  - Tuples that are almost in violation

 $\label{eq:player} \begin{array}{l} dq: \mbox{Player}(n, \, s, \, t, \, st, \, g), \mbox{Player}(n', \, s', \, t', \, st', \, g'), \mbox{t=t', st $\neq$ st'} \\ vg: \mbox{Player}(n, \, s, \, t, \, st, \, g), \mbox{Player}(n', \, s', \, t', \, st', \, g'), \mbox{t=t', st $\neq$ st'} \end{array}$ 



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- iBench has already been applied successfully by several diverse integration projects
- We have used iBench numerous times for our own evaluations
  - Our initial motivation for building iBench stemmed from our own evaluation needs



# Value Invention



- Translate mappings
  - from expressive, less well-behaved language (SO tgds)
  - into less expressive, more well-behaved language (st-tgds)
- Input: schemas, integrity constraints, mappings
- **Output**: translated mappings (if possible)
- Evaluation Goal: how often do we succeed
- Why iBench: need a large number of diverse mappings to get meaningful results
- Evaluation Approach: generated 12.5 million integration scenarios based on randomly generated configuration file



# Vagabond



- Vagabond
  - Finding explanations for data exchange errors
    - User marks attribute values in generated data as incorrect
    - System enumerates and ranks potential causes
- Input: schemas, integrity constraints, mappings, schema matches, data, errors
- Output: enumeration of causes or incremental ranking
- Evaluation Goal: evaluate scalability, quality
- Why iBench:
  - Control characteristics for scalability evaluation
  - Scale real-world examples



# Mapping Discovery



- Learning mappings between schemas using statistical techniques
- Input: schemas, data, constraints
- Output: mappings
  - University of California, Santa-Cruz
    - Lise Getoor, Alex Memory
    - Reneé Miller
    - https://linqs.soe.ucsc.edu/people



### And more ...



- Functional Dependencies Unleashed for Scalable Data Exchange
  - [Bonifati, Ileana, Linardi arXiv preprint arXiv:1602.00563, 2016]
  - Used iBench to compare a new chase-based data exchange algorithm to SQL-based exchange algorithm of ++Spicy
- Approximation Algorithms for Schema-Mapping Discovery from Data
  - [ten Cate, Kolaitis, Qian, Tan AMW 2015]
  - Approximate the Gottlob-Senellart notion
  - Kun Qian currently using iBench to evaluate effectiveness of approximation
- Comparative Evaluation of Chase engines
  - [Università della Basilicata, University of Oxford]
  - Using iBench to generate schemas, constraints



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## Conclusions



- Empirical Evaluations of Integration Systems
  - Need automated tools for robust, scalable, broad, repeatable evaluations
- BART
  - Controlled error generation
  - Detectable errors, measure repairability
- iBench
  - Comprehensive metadata generator
  - Produces inputs and outputs (gold standards) for a variety of integration tasks



## Future Work



- Data quality measures
  - Implement complex quality measures
- iBench
  - More control over data generation
  - Orchestrating multiple mappings
    - Sequential: e.g., schema evolution
    - Parallel: e.g., virtual integration
- BART
  - Support combined mapping/cleaning scenarios
  - How to efficiently generate clean data (without having to run full cleaning algorithm)
  - Similarity measure for instances with labelled nulls/ variables







### • iBench

Webpage: <u>http://dblab.cs.toronto.edu/project/iBench/</u>

Code: <u>https://bitbucket.org/ibencher/ibench/</u>

Public Scenario Repo: <u>https://bitbucket.org/ibencher/</u> <u>ibenchconfigurationsandscenarios</u>

### • BART

Webpage: <u>http://www.db.unibas.it/projects/bart/</u> Code: <u>https://github.com/dbunibas/BART</u> Example Datasets: <u>http://www.db.unibas.it/projects/bart/files/</u> <u>BART-Datasets.zip</u>









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