Dynamic Fusion of Web Data: Beyond Mashups

Erhard Rahm Andreas Thor, David Aumüller

http://dbs.uni-leipzig.de



24th September, 2007

Top VLDB '97 Pubs: Google Scholar's Top-5



DataGuides: Enable query formulation and optimization in semistructured databases

R Goldman, J Widom - Proc. of VLDB, 1997 - citeseer.ist.psu.edu
The recent database difficulties have been resolved. Please let us know if you
encounter any data corruptions. ... Citation: Context R. Goldman and J. Widom.
DataGuides: Enable query formulation and optimization in semistructured ...
Cited by 56 - Related Articles - Cached - Web Search

Vertical Data Migration in Large Near-Line Document Archives Based on Markov-Chain Predictions -

AKG Weikum - 1997 - vldb.org

Abstract Large multimedia document archives hold most of their data in near-line tertiary storage libraries for cost reasons. This paper de- velops an integrated approach to the vertical data migration he- tween the tertiary and ... Cited by 24 - Related Articles - View as HTML - Web Search - BL Direct

[PS] Garbage Collection in Object Oriented Databases Using Transactional Cyclic Reference Counting

S Ashwin, P Roy, S Seshadri, A Silberschatz, S ... - VLDB, 1997 - cse.iitb.ac.in Garbage Collection in Object Oriented Databases Using ... S. Ashwin 1 Prasan Roy 1 S. Seshadri 1 Avi Silberschatz 2 ... 1 Indian Institute of Technology, Mumbai 400 076, India sashwin@cs.wisc.edu f prasan,seshadri,sudarsha g ... Cited by 9 - Related Articles - View as HTML - Web Search - BL Direct

іспатіон) Optimization in semi structured data Q Formulation - 1997 - VLDB Cited by 2 - Related Articles - Web Search

ıсıтатıом Logical and physical versioning in main memory databases PBDWL Abraham, SSSR Rastogi, S Seshadri - 1997 - VLDB <u>Cited by 1 - Related Articles</u> - <u>Web Search</u>

Google Scholar's Top-5 (2)



[сітатіом] ^aVisual Data Mining, ⁰tutorial

DA Keim - Proc. Conf. Very Large Databases, 1997

Cited by 1 - Related Articles - Web Search

[сітатіом] Don't scrap it, wrap it! an architecture for legacy data sources MT Roth, P Schwarz - International Conference on Very Large Databases, 1997

Cited by 6 - Related Articles - Web Search

[сітатіом] Visual Data Mining, Tutorial Notes, Int

DA Keim - Conference on Very Large Databases, Athens, 1997

Cited by 1 - Related Articles - Web Search

լспатион Geo/EnvironmentalandMedicalDataManagement in the RasDaMan System P Baumann, P Furtado, R Ritsch, N and Widmann - Proc. 23rd Conf. Very Large Databases, 1997 Cited by 1 - Related Articles - Web Search

[сітатіом] Visual Data Mining, Conf

D Keim - On Very Large Databases (VLDB97), Athens, Greece, 1997

Cited by 1 - Related Articles - Web Search

... more GS quality problems

DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases - all 31 versions »

R Goldman, J Widom Proceedings of the 23rd International Conference on Very ..., 1997. Page 1. 1 DataGuides. Enabling Query Formulation and Optimization in Semistructured 1997 - www-db.stanford.edu

Databases * Roy Goldman Stanford University royg@cs.stanford.edu ...

Cited by 732 - Related Articles - View as HTML - Web Search - BL Direct

DataGuides Enable buery formulation and optimization in semi R Goldman, J Widom Proc. of VLDB, 1997 - citeseer.ist.psu.edu

Heterogeneous venue names

• How to query for "VLDB '97"?

... Document: Details DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases (1997) Roy Goldman, Jennifer Widom Citation: Context R ...

Cited by 56 - Related Articles - Cached - Web Search

PataGuides: Enabling Ouery Formulation and Optimization in Semistructured Databases بستا

G Roy, W Jennifer Proc. 23rd VLDB, 1997

[статіон] Dataguides: Enabling query formulation and optimization in semistructure databases, VLDB'97

R Goldman, J Widom 23rd International Conference on Very Large DataBase, Athens.

Cited by 4 - Related Articles - Web Search

итатиом **Dataguides: Enabling query** formulation and optimization R Golman J Widom - Proceedings of the Twenty-Third Internation Conference

Cited by 1 - Related Articles - Web Search

[сітатіом] DataGuides: Enabling Query Formulation and Optimizat

R Goldman - VLDB 1997

Cited by 1 - Related Articles - Web Search

Duplicates due to

- Extraction errors (title, authors)
- Different titles
- Typos (author name)
- Heterogeneous venue names
- Missing / additional authors (!)

lan.

Top VLDB'97 Pubs: MS Libra's Result



- 🔁 DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases (1997) (citation: 187)
 - 🦓 Roy Goldman Jennifer Widom
- 🔼 Optimizing Queries across Diverse Data Sources(1997) (citation:163)
 - 🦓 Laura M. Haasj Donald Kossmann, Edward L. Wimmers, Jun Yang
- 🔼 To Weave the Web(1997) (citation: 128)
 - 🦓 Paolo Atzeni <u>Giansalvatore Mecca Paolo Merialdo</u>
- 🔼 Selectivity Estimation Without the Attribute Value Independence Assumption(1997) (citation: 100)
 - 🔏 Viswanath Poosala | Yannis E. Ioannidis
- 🔁 STING : A Statistical Information Grid Approach to Spatial Data Mining(1997) (citation:88)
 - 🔏 Wei Wang Jiong Yang Richard R. Muntz

... similar problems

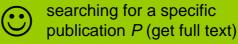
- 🟂 DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases (1997) (citation: 187)
 - 🤽 Roy Goldman Jennifer Widom
- 🗖 DataGuides: Enabling Query Formulation and Optimization in

Semistructured Databases (1977) (citation: 170)

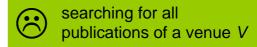
🔏 Jennifer Widom Roy Goldman

Jennifer Widom received her Bachelors degree in 1982 and her Computer Science Ph.D. in 1987.

- 🔼 dataguides: enable query formulation and optimizat semistructured databases(1997) (citation:9)
 - 🦥 r. Goldman 🛚 j. Widom
- 🔁 DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases (1998) (citation:6)
 - 🦓 r. Goldman j. Widom
- 🔼 DataGuides: Enabling Query Formulation and Optimiza-ti Semistructured Databases (1998) (citation: 2)
 - 🔏 Roy Goldman Jennifer Widom



searching for all publications of an author A



Top VLDB '97 Publications: Desired Result

	Title	Authors	Venue	Year	Citation ▼
4	DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases.	Roy Goldman, Jennifer Widom	VLDB	1997	795
4	M-tree: An Efficient Access Method for Similarity Search in Metric Spaces.			1997	598
4	STING: A Statistical Information Grid Approach to Spatial Data Mining.	Wei Wang, Jiong Yang, Richard R. Muntz	VLDB	1997	386
4	Optimizing Queries Across Diverse Data Sources.	Laura M. Haas, Donald Kossmann, Edward L. Wimmers, Jun Yang	VLDB	1997	366
4	To Weave the Web.	Paolo Atzeni, Giansalvatore Mecca, Paolo Merialdo	VLDB	1997	249
4	Selectivity Estimation Without the Attribute Value Independence Assumption.	Viswanath Poosala, Yannis E. Ioannidis	VLDB	1997	220
4	A Foundation for Multi-dimensional Databases.	Marc Gyssens, Laks V. S. Lakshmanan	VLDB	1997	204
4	Algorithms for Materialized View Design in Data Warehousing Environment.	Jian Yang, Kamalakar Karlapalem, Qing Li	VLDB	1997	178
4	Fast Computation of Sparse Datacubes.	Kenneth A. Ross, Divesh Srivastava	VLDB	1997	165
4	Data Warehouse Configuration.	Dimitri Theodoratos, Timos K. Sellis	VLDB	1997	155

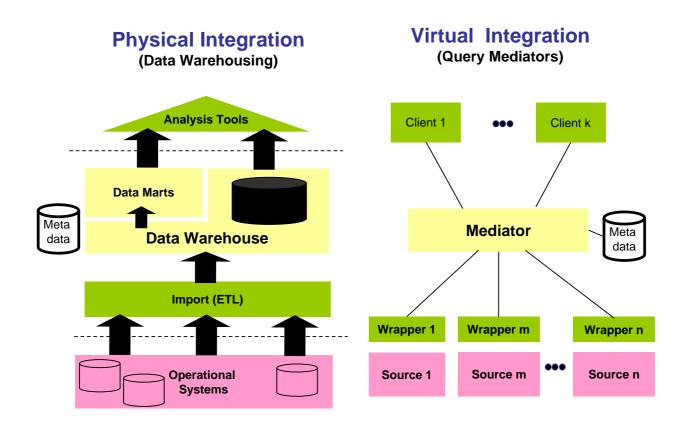
Top VLDB '97 Publications: Desired Result (2)

	Title	Authors	Venue	Year	Citation ▼
	DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases. R Goldman, J Widom: DataGuides: Enabling 732 Query Formulation and Optimization in Semistructured Databases (1997) R Goldman, J Widom: Dataguides: Enabling query 4 formulation and optimization in semistructured databases. VLDB'97 (1997) R Goldman, J Widom: DataGuides: Enable query 56 formulation and optimization in semistructured databases (1997) R Golman, J Widom: Dataguides: Enabling query 1 formulation and optimization in semistructured databases R Goldman: DataGuides: Enabling Query 1 Formulation and Optimization in Semistructured Databases, R. Goldman, J. (1997) R Goldman, J Widom: Dataguides: enabling 1 querying formulation and optimization in semi-structured databases, VLDBâ??97 (1997)	Roy Goldman, Jennifer Widom	VLDB	1997	795
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÷	Optimizing Queries Across Diverse Data Sources.	Laura M. Haas, Donald Kossmann, Edward L.	VLDB	1997	366

Agenda

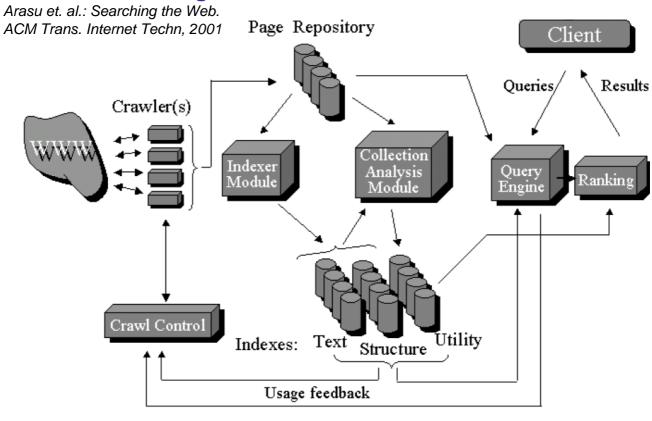
- Motivation
- Data Integration Systems
 - Overview
 - Workflow-based data integration
- Mashups
 - Overview
 - Tools: classification & examples
 - Open problems
- iFuice-based integration workflows
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 - Query strategies: Example: Online Citation Service
 - Dynamic object matching: The MOMA approach
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"DB World"

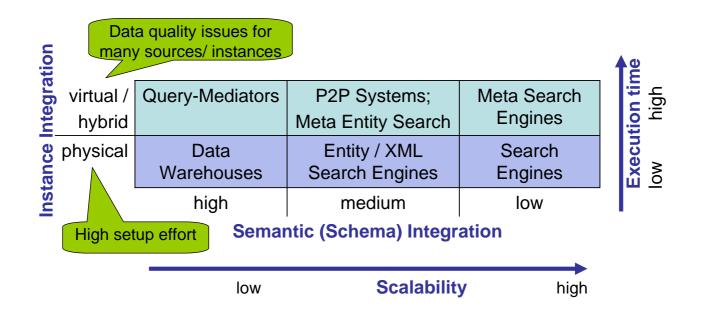


"IR World"

General search engine architecture



Integration "Sextant"



(Some) Problems of current data integration approaches

- Setup time too high: crawling; schema mapping / integration ...
- Current data integration approaches are query-focussed: search engine queries, query mediators, warehouse access
- Queries are not enough: complex data integration problems cannot easily be solved in 1 query / search
 - What is the most cited XSym paper so far?
 - Which famous scientists lived close to the VLDB 2007 venue
- Data quality for heterogeneous/dirty web data and query results
- Execution time for dynamic fusion of larger data sets

Workflow-like data integration

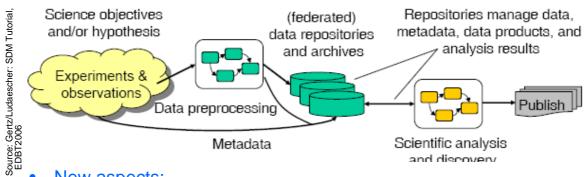
- "You only have three hours how far can you go to solve a data integration problem?"
- Reuse + Combine existing (data) services within data integration workflows
 - Reuse existing services
 - Reuse existing data integration systems, e.g. search engines, query mediators, warehouses
 - Combine query/service results within a workflow
 - Perform data cleaning and data transformation
 - Perform data analysis
- Must be supported by a flexible data integration framework
- Workflow-like data integration complements query-based data integration

Workflow-based Integration

Examples

Offline: ETL processes for Data Warehouses

- Online: Workflows for analyzing biological data



- New aspects:
 - Combine ETL and analysis workflows (on-demand information extraction)
 - Share and reuse existing data services and tools
 - Reuse existing (entity) search engines
 - Easy development and use of workflows (-> Mashups)

Comparison: Query- vs. Workflow-based DI

	Query-based	Workflow-based
User defines	Query	Workflow
Building blocks	Sources	Services for ETL, Query/Search and Analysis
Access Flexibility	High	Restricted
Development Time	High ("End-user System")	Medium (Framework + Workflow)
Specific to	Source(s) / Domain	Task / Problem

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Mashups - a light-weight data integration approach

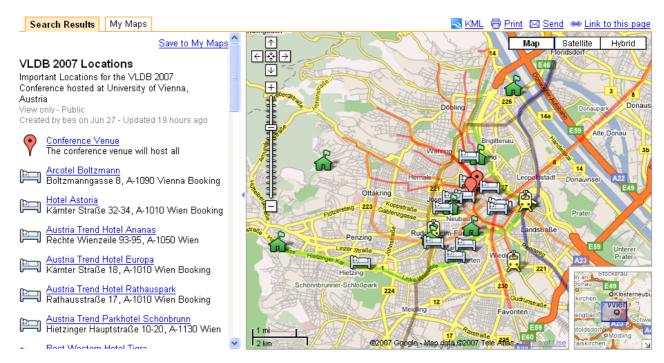
- "A web mashup is a web page or application that combines data from two or more external online sources." (ProgrammableWeb)
- "A mashup is a web application that combines data from more than one source into an integrated experience." (Wikipedia)
- "Mashups are an exciting genre of interactive Web applications that draw upon content retrieved from external data sources to create entirely new and innovative services." (Merrill: Mashups: The new breed of Web app)

Mashup Example: Forbes List



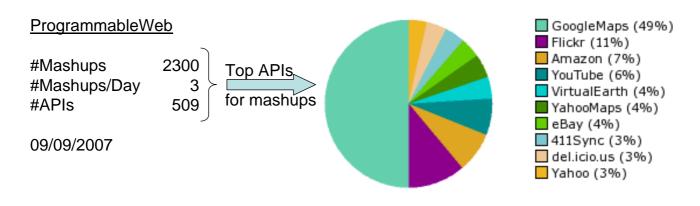
VLDB Locations

 Displays conference venue of VLDB 2007 as well as nearby hotels, train stations, airports, ...

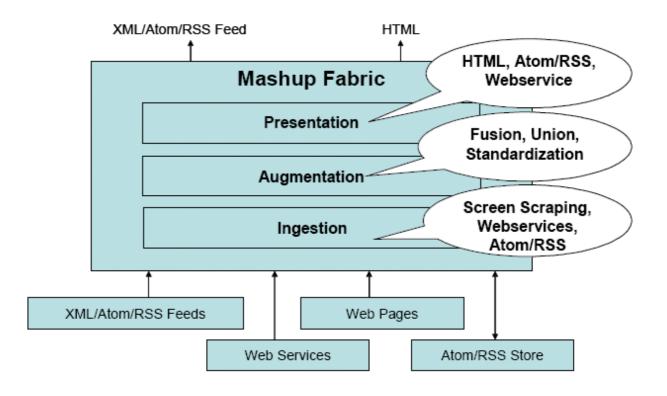


Mashups: Driving forces

- AJAX (Asynchronous Javascript and XML)
 - Desktop-like look-and-feel of Web applications
- Development tools, e.g. Google Web toolkit
- Visual development tools without programming need
- Increasing number of Web services (APIs)
 - Easy access to "interesting" content and services
 - 50% of mashups use Google Maps

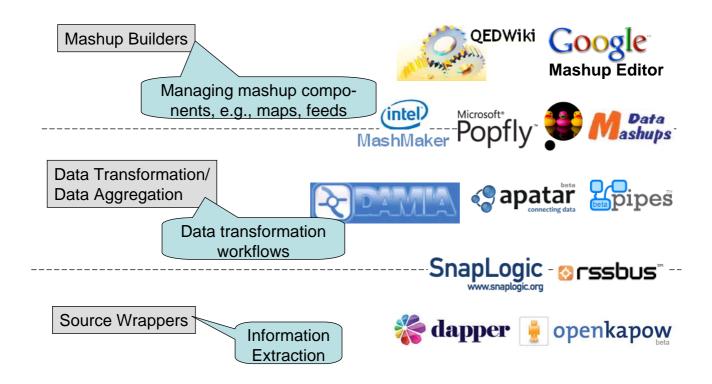


The Big Picture: Mashup Fabric*

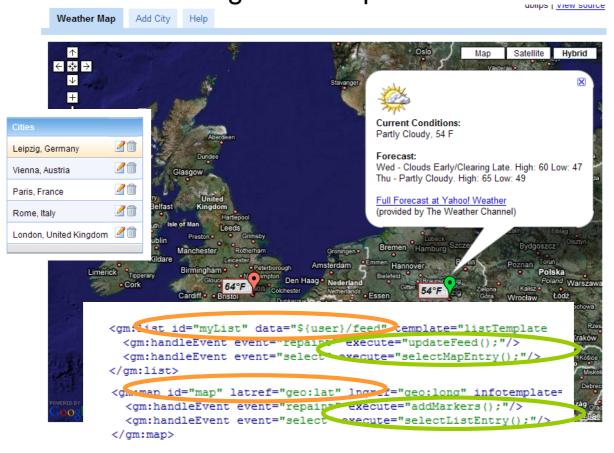


^{*} Jhingran: Enterprise Information Mashups: Integrating Information, Simply. Keynote at VLDB'06

Mashup Tools: Overview



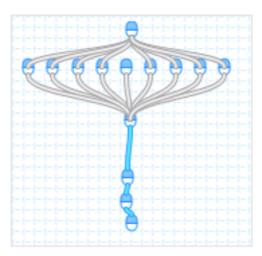




Yahoo Pipes

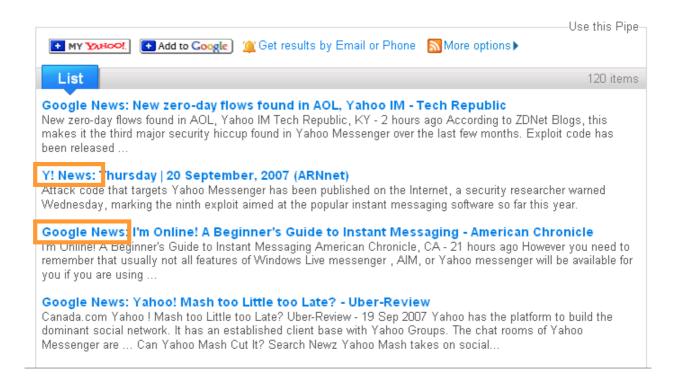


- Composition tool to aggregate, manipulate, and mashup web content, especially RSS feeds
 - Pipe = data transformation workflow
 - Visual specification
- Example: Aggregated News Alert



- User input: keyword(s)
- Parallel search at
 - Yahoo! News
 - MSN Live News ...
- Merge
- Sort by date
- Deduplication (unique title)
- Output

Yahoo Pipes: Aggregated News Alert (2)



IBM DAMIA



- Similar to Yahoo! Pipes
- Modules
 - Sources: URLs, Excel files
 - Operators (see right)
- Example: Aggregated News





Dapper: Google News as RSS feed

- "Get any content from the web"
 - Information extraction from any website into XML, RSS ...
- Automatic detection of parts ("blocks") of the same structure
 - e.g., several news entries, news headline, ...
- → User selects "blocks" for extraction



Mashups: Characteristics

- Easy and fast development
 - Visual programming (drag & drop) or integration with power development environments (e.g., Eclipse)
- Service-oriented paradigm
 - Sharing and reuse of web services
- Web2.0 interfaces
- Standardized XML-based data formats, e.g. RSS, SOAP/REST (data exchange)
- Simple processing workflows
- Simple instance-based data integration
 - Geographical coordinates
 - Keywords (e.g. names)
- Simple keyword queries dominate (no query transformation)
- Limited result postprocessing (primarily merge instead of match)

Mashups: Query example

- Keyword queries
- "Merge instead of match"



Mashups: Open Problems

- More complex queries, e.g. for heterogeneous entity search engines (-> query transformation)
- Data quality
 - Precision and recall depend on developer's choices (source selection, query formulation)
 - Typos, missing/wrong attribute values (e.g., due to extraction errors)
 - Duplicates, i.e., sources contain multiple instances for the same (real world) object
- Performance for large data volume (automatic optimization)
- Semantic repository of services
 - Service description & service disvovery
- Support for business applications, e.g. security restrictions

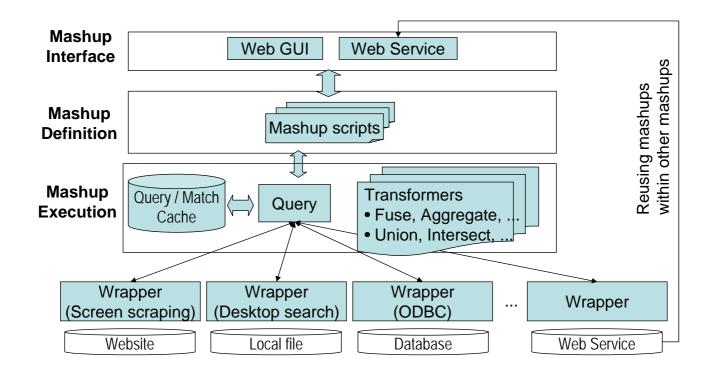
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Information Fusion with iFuice [RTA+05]

- Generic data integration platform for structured and unstructured data sources
 - Query / search / id-based data access
- Workflow-like data integration with operator-based programming model
 - Generic high-level operators for use within script programs
 - Example: query traverse, map, union, aggregate,
- Utilization of instance-level mappings
 - Correspondences between object instances
 - Represent semantic relationship ("is same", "is associated to")
- Metadata repository for data sources and services
 - Semantic object (e.g., Author, Publication) and mapping types
- Iterative query strategies
- On-the-fly object matching

Mashup Framework: Architecture



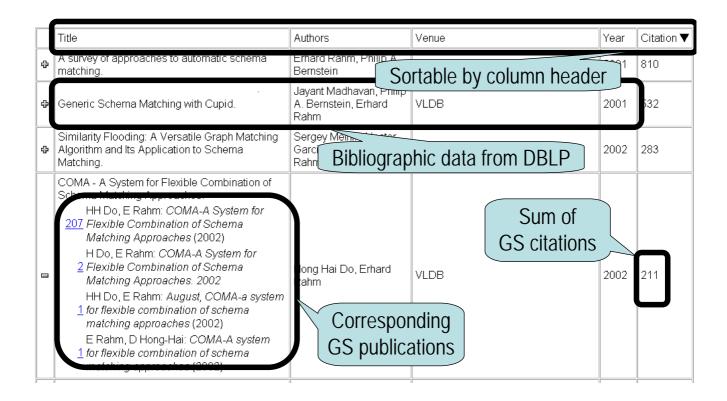
Iterative query strategies

- Problem: Simple queries may result in poor quality (precision/recall)
 - Search engines return top-k "best effort" results
 - Are the first 50-100 hits sufficient?
 - Does the ranking reflect the applications needs?
- Query Strategy
 - Set of subsequently executed queries
 - Goal: find a relevant set of instances with a minimal number of queries
 - Balance between query costs (#queries) & result quality (#relevant instances)
- Users expect immediate results
 - Start with query strategy with only 1 or a few queries
- AJAX allows result refinement
 - Asynchronous execution of more sophisticated query strategies and (once finished) result update
- User interaction for additional queries if needed

Example: Online Citation Service* [TAR07]

- On-demand citation analysis
 - What are the most cited papers of conference X?
 - What is the average citation number of publications from author Y?
 - Frequent changes, i.e., new publications & new citations
- Idea: Combine publication lists, e.g. from DBLP or Pubmed, with citation counts, e.g from Google Scholar, Citeseer or Scopus
 - DBLP, Pubmed: high bibliographic data quality
 - GS: large coverage of citations counts
- Query problem: Given a set of DBLP publications → How to find the corresponding GS publications?
 - Query GS and match DBLP-GS
 - * http://labs.dbs.uni-leipzig.de/ocs

Online Citation Service: Result overview



OCS: Query strategies overview

- 1st query strategy: name (#queries = 1)
 - Query: author's/venue's name, e.g., 'author:E-Rahm'
 - Goal: find as many relevant pub's with only one query
- 2nd query strategy: title pattern (#qu.=#pubs/10)
 - Query: Disjunction of title patterns, e.g., 'intitle:"survey * approaches *
 * schema matching" OR ... '
 - Goal: Precise search for a limited set of publications
- 3rd query strategy: title keywords (#queries = #pubs)
 - Query: publication title, e.g., 'A survey of approaches to automatic schema matching'
 - Goal: Find a certain publication at the cost of many irrelevant search results

OCS: 1st query strategy "name"

Query strategy	Name
#Queries	1
#GS pubs	40
#DBLP pubs (matched)	35
#DBLP pubs (unmatched)	31
#GS citations (overall)	2395

	Title	Authors	Venue	Year	Citation ▼
ф	Optimizing Queries Across Diverse Data Sources.	Laura M. Haas, Donald Kossmann, Edward L. Wimmers, Jun Yang	VLDB	1997	359
ф	To Weave the Web.	Paolo Atzeni, Giansalvatore Mecca, Paolo Merialdo	VLDB	1997	244
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4	Algorithms for Materialized View Design in Data Warehousing Environment.	Jian Yang, Kamalakar Karlapalem, Qing Li	VLDB	1997	178

OCS: 2nd query strategy "title pattern"

Query strategy	Name	Title Pattern
#Queries	1	8
#GS pubs	40	125
#DBLP pubs (matched)	35	59
#DBLP pubs (unmatched)	31	7
#GS citations (overall)	2395	5421

	Title	Authors	Venue	Year	Citation \	▼
÷	DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases.	Roy Goldman, Jennifer Widom	VLDB	1997	795	
4	M-tree: An Efficient Access Method for Similarity Search in Metric Spaces.	Paolo Ciaccia, Marco Patella, Pavel Zezula	VLDB	1997	598	
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G.	To Weave the Web.	Paolo Atzeni, Giansalvatore Mecca, Paolo Merialdo	VLDB	1997	249	

OCS: 3rd query strategy "title keywords"

Query strategy	Name	Title Pattern	Title Keywords
#Queries	1	8	7 7
#GS pubs	40	125	125
#DBLP pubs (matched)	35	59	59
#DBLP pubs (unmatched)	31	7	7
#GS citations (overall)	2395	5421	5421

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Illustration of OCS mashup execution

User selects author from list

01: \$DBLPPubs := query (DBLP, "author=[name]");

02: \$GSPubs1 := query (GS, "author:[name]");

03: \$Result1 := fuse (\$DBLPPubs/Pub/, \$GSPubs1/Entry/);

04: \$Result1 := aggregate (\$Result1, "[DBLP/Pub/NoOfCit]", "sum([./Entry/Citations])");

Result1 is displayed to the user

Start next script automatically

05: \$GSPubs2 := query (GS, \$DBLPPubs, "intitle:"[DBLP/Pub/Titlepattern]"");

06: \$Result2 := union (fuse (\$DBLPPubs/Pub/, \$GSPubs2/Entry/), \$Result1);

07: \$Result2 := aggregate (\$Result2, "[DBLP/Pub/NoOfCit]", "sum([./Entry/Citations])");

Result1 is replaced by Result2

If user wants exhaustive search (e.g., by button click)→ start next script

08: \$DBLPPubs3:= query (\$Result2, "count([./Entry/Citations])=0");

09: \$GSPubs3:= query (GS, \$DBLPPubs3, "intitle:[DBLP/Pub/Title]");

10: \$ Result3 := union (fuse (\$DBLPPubs/Pub/, \$GSPubs3/Entry/), \$ Result2);

11: \$ Result3 := aggregate (\$Result3, "[DBLP/Pub/NoOfCit]", "sum([./Entry/Citations])");

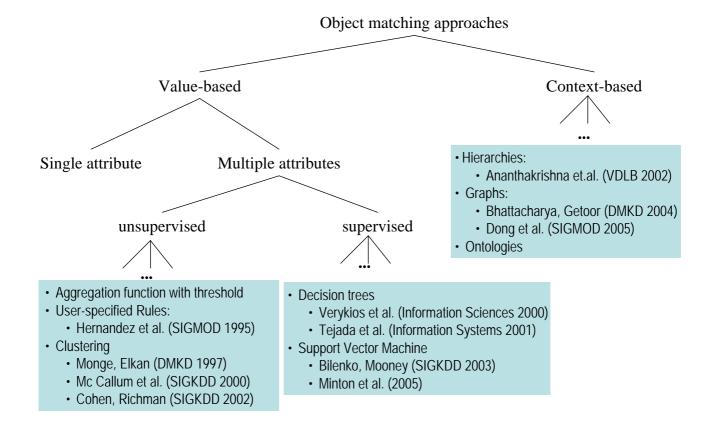
Query strategies: Summary & Future Work

- Iterative query strategies are a flexible approach for querying search engines
 - Heuristics for getting a maximum number of relevant object instances with a minimum number of queries
 - Allow approximated results as well as result refinement
- Evaluation and optimization of query strategies
 - Example: 'name' strategy often good for authors but not for venues
 - Dependency on instance values "E-Rahm" [⊕] vs. "J-Smith" [⊕]
- Automatic generation of query strategies
 - Can we automatically determine the relevant attributes (and their transformations) that should appear in the queries?
 - Generic approaches desirable

On-the-fly object matching

- Object matching is important part of data integration
 - prerequisite for information fusion
 - Example: group together multiple Google Scholar entries
- Goals
 - seamless integration in data integration workflows
 - effective & efficient

Many object matching approaches ...



Many data cleaning frameworks ...

Research prototypes

- AJAX (Galhardas et al., VLDB 2001)
- IntelliClean (Lee et al., SIGKDD 2000)
- Potter's Wheel (Raman et al., VLDB 2001)
- Febrl (Christen, Churches, PAKDD 2004)
- TAILOR (Elfeky et al., Data Eng. 2002)
- MOMA (Thor, Rahm, CIDR 2007)
- **–** ...

Commercial solutions

- DataCleanser (EDD), Merge/Purge Library (Sagent/QM Software), MasterMerge (Pitnew Bowes) ...
- MS SQL Server 2005: Data Cleaning Operators (Fuzzy Join / Lookup)

– ...

MOMA Overview [TR07]

- MOMA = <u>Mapping based</u> <u>Object <u>Ma</u>tching
 </u>
- Object consolidation framework
 - Matching objects from 2 sources
 - Generation of instance mappings (correspondences)

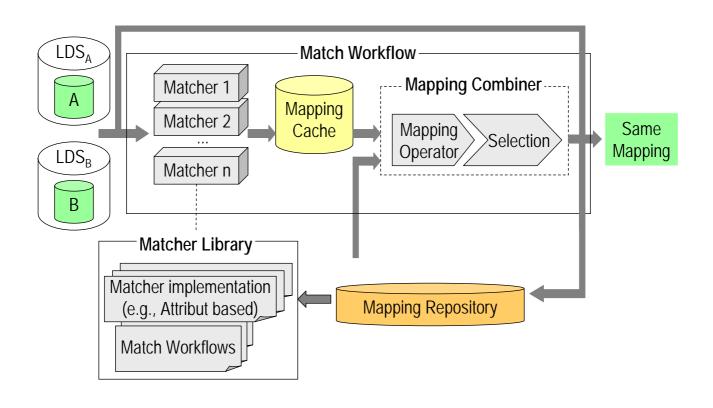
Special case: duplicate detection within 1 source (generation of self-mapping)

- Key features
 - Extensible matcher library
 - Mapping combination
 - Construction of match workflows
 - Storage of mappings for reuse in other match problems

same-mapping for authors

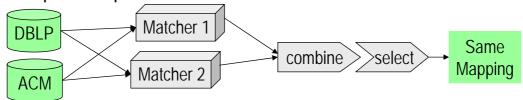
 Note: similar objectives than in schema matching, e.g. in COMA / COMA++

MOMA Architecture



Match Workflows

- Coordinated execution of matchers and combination of mappings
 - single-attribute matcher (e.g. based on specific string similarity function)
 - multi-attribute matcher (hybrid matcher)
 - context matcher ...
- Example: Independent matcher execution



■ Implemented as *iFuice* scripts

```
$M1 := attrMatch ($DBLP, $ACM, "[title]", TFIDF, 0.9);

$M2 := attrMatch ($DBLP, $ACM, "[year]", EditDistance, 0.7);

$Union := union ($M1, $M2, avg);

$Result := select ($Union, 0.8);
```

Match Strategies

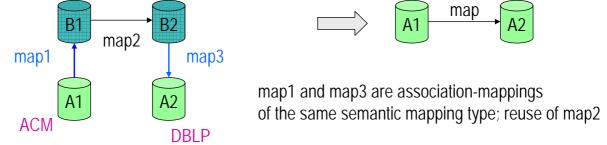
Merge same-mappings



Compose same-mappings



 Compose same- and association-mappings (Neighborhood matcher)



OCS Match Strategies

Interactive approach, i.e., user selects match thresholds

Title	Year	Authors	relaxed
<u>80%</u>	+/- two years	<u>50%</u>	†
<u>85%</u>	+/- one year	<u>60%</u>	
90%	equal year	<u>70%</u>	
<u>95%</u>		<u>80%</u>	
<u>100%</u>		<u>90%</u>	₩
	-	<u>100%</u>	restrictive

 Aggregated result is adjusted automatically based on match definition

Match strategies: Summary & Future Work

- Match strategies are a flexible approach for matching instances of different sources
 - Interactive result adjustment
 - Reuse of existing mappings (efficient matching)
 - Match refinement by applying multiple match strategies
- · Optimization of match strategies
 - Which matchers?
 - Which attributes?
 - What are the best thresholds?

Agenda

- Motivation
- Data Integration Systems
 - Overview
 - Workflow-based data integration
- Mashups
 - Overview
 - Tools: classification & examples
 - Open problems
- iFuice-based integration workflows
 - Overview
 - Query strategies: Example: Online Citation Service
 - Dynamic object matching: The MOMA approach
- Summary

Summary

- Mashups are a light-weight approach for dynamic workflowbased data integration
 - Fast development times
 - Sharing and massive reuse of existing services
 - Proliferation of tools
 - Business potential
- Mashups need more sophisticated data integration support
 - Generic data integration framework for mashups
- Challenging problems
 - Performance quality tradeoffs (recall, precision, data cleaning)
 - Automatic generation of queries → iterative query strategies
 - On-the fly object matching → mapping-based match strategies
 - Automatic identification of suitable (query) services

– ...

References

- [RTA+05] Rahm, Thor, Aumueller, Do, Golovin, Kirsten: iFuice - Information Fusion utilizing Instance Correspondences and Peer Mappings. Proc. WebDB, 2005
- [TAR07] Thor, Aumueller, Rahm: *Data Integration Support for Mashups*. Proc. of IIWeb, 2007
- [TR07] Thor, Rahm: MOMA A Mapping-based Object Matching System. Proc. of CIDR, 2007