

# Matching Large Schemas with COMA++

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# Database Group U Leipzig (Feb. 2005)

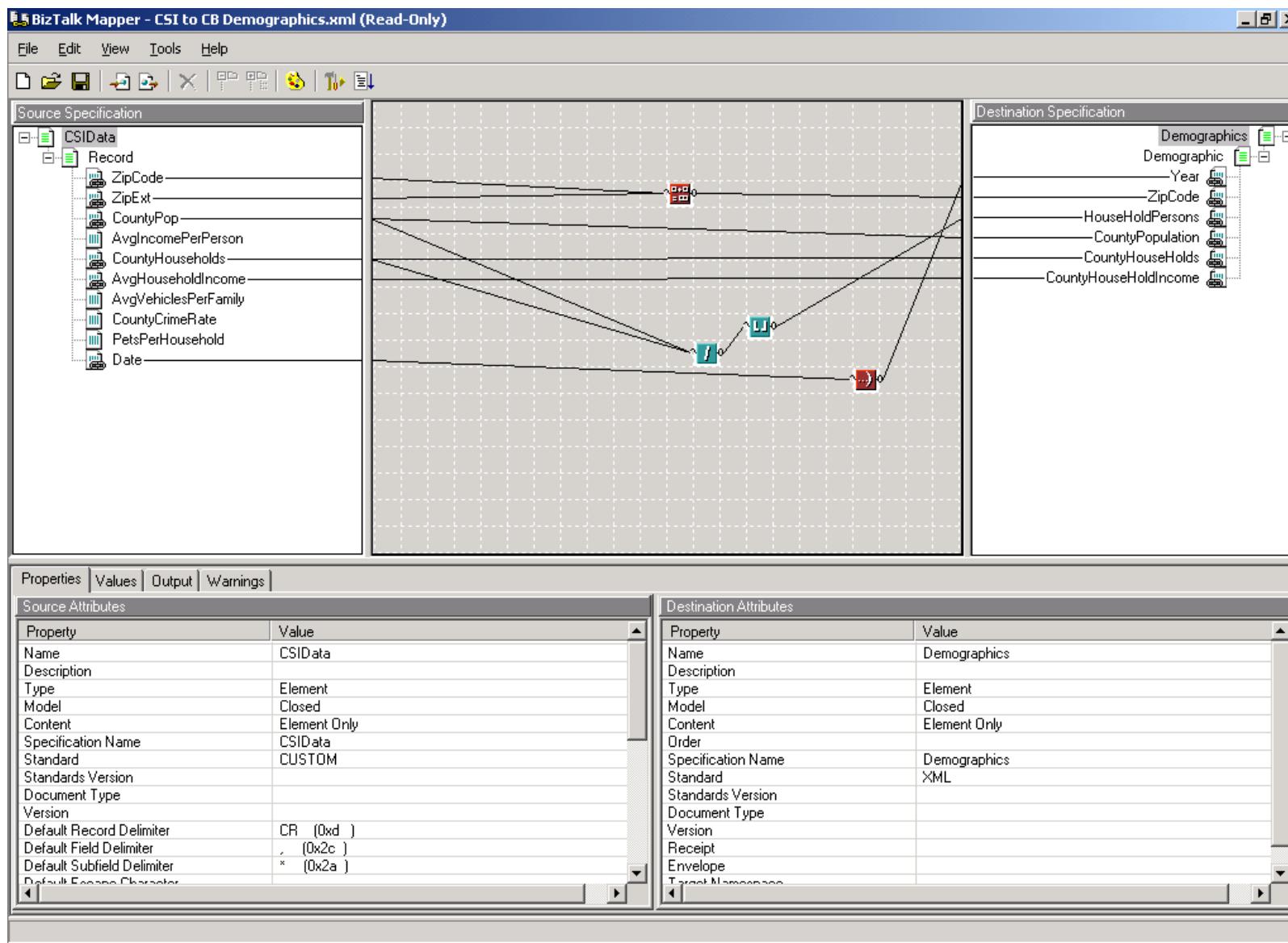


- **Projects:** Metadata Mgmt (COMA++),  
Data Integration (GeWare, GenMapper, *iFuice*),  
Adaptive web recommendations (AWESOME), E-Learning

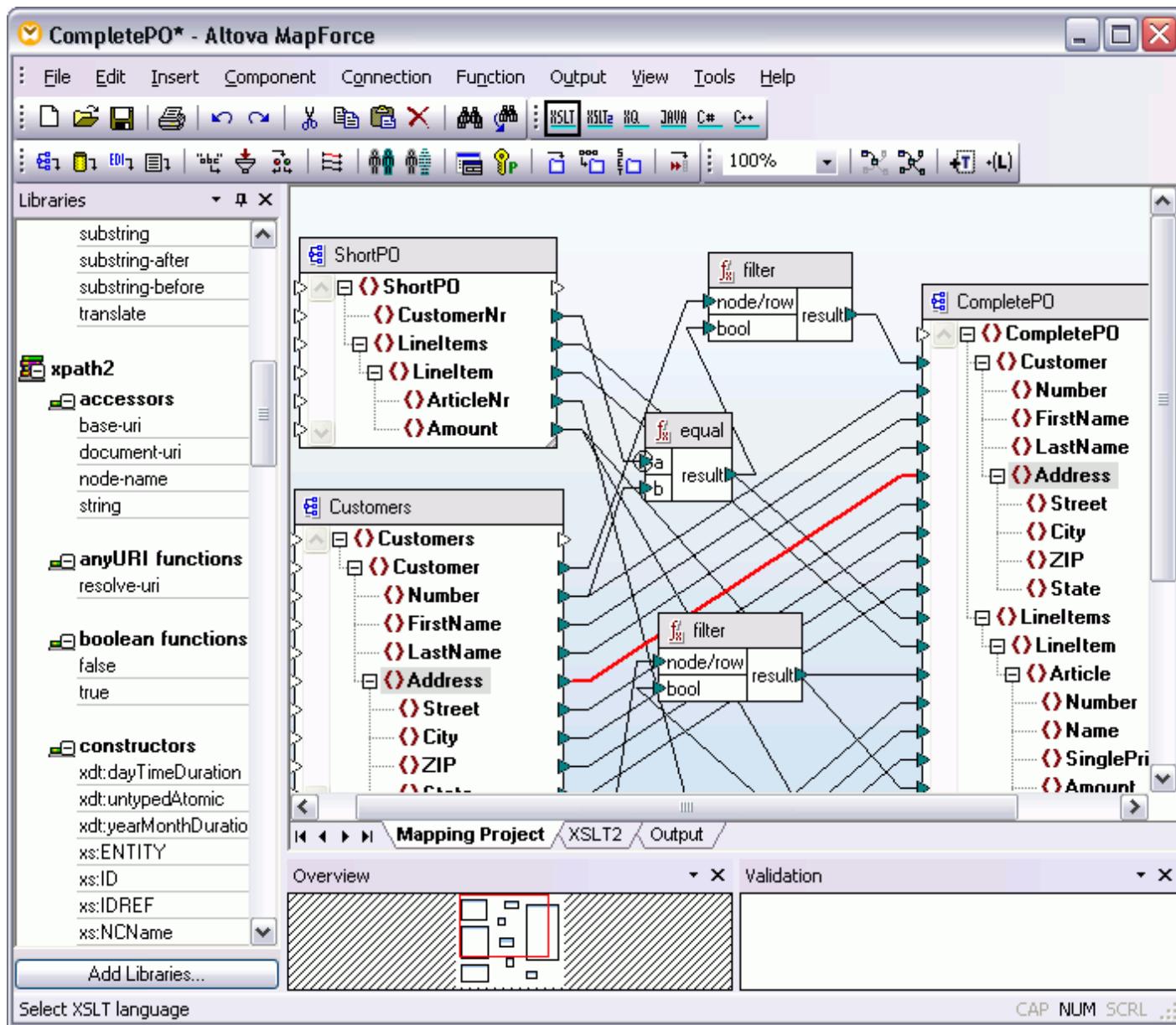
# Schema Matching

- Find semantic correspondences between 2 schemas
  - DB schemas, XML schemas, web service interfaces, ontologies, ...
- Key step in many metadata applications
  - Data integration: mediators, data warehouses
  - E-Business: XML message mapping; matching of product catalogs
  - Semantic Web: ontology matching (alignment)
- Input: 2 schemas  $S_1$  and  $S_2$ 
  - Possibly: instances of  $S_1$  and  $S_2$ , *background knowledge*
- Output: Mapping between  $S_1$  und  $S_2$ 
  - Correspondences between schema components
  - Expressions, e.g. for data transformation
- Need to automate (many schemas, large schemas, error-prone)
  - manual control still necessary, especially for business apps

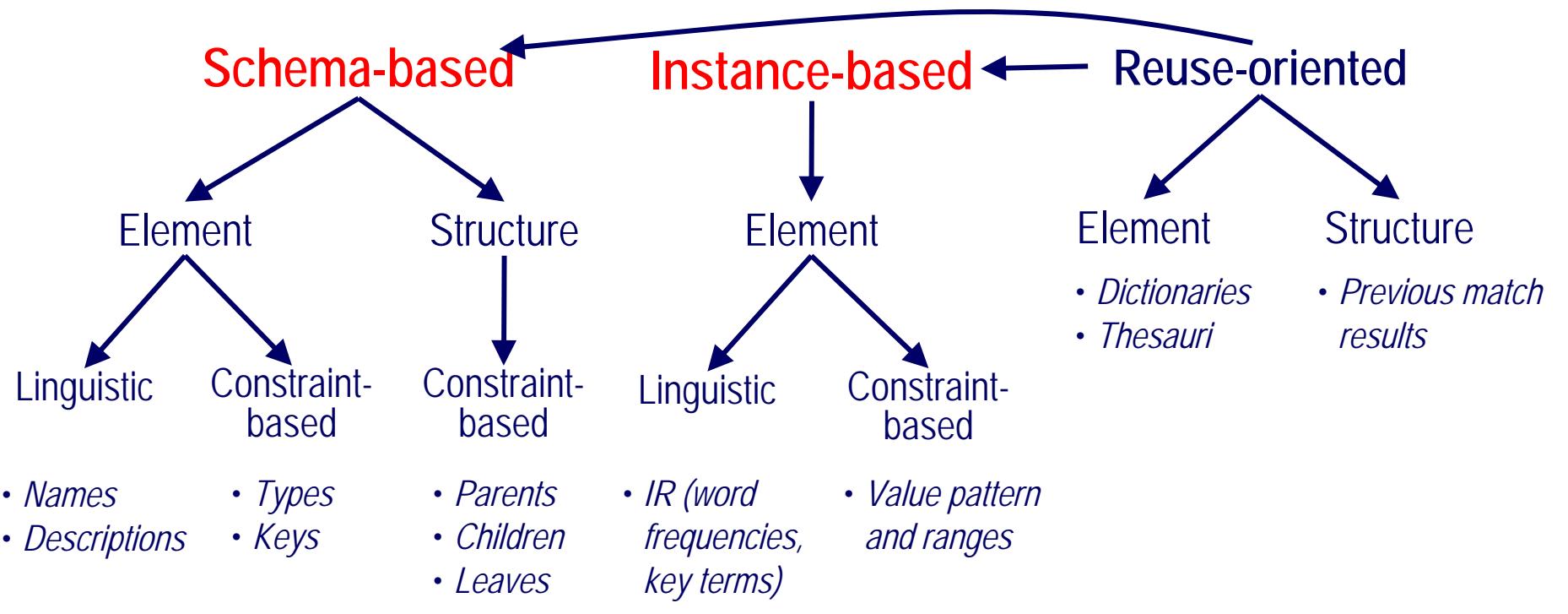
# Tool Example: Biztalk Mapper



# Tool Example: Altova MapForce



# Automatic Match Techniques\*



- Combined Approaches: Hybrid vs. Composite

\*Rahm, E., P.A. Bernstein: *A Survey of Approaches to Automatic Schema Matching*. VLDB Journal 10(4), 2001

# Current Situation

- High research interest in recent years
  - Many papers and several new prototypes:  
LSD (Sigmod01), Cupid (VLDB01), COMA (VLDB02), Clio (IBM), ...
  - Semantic web research on ontology matching
- Published results mostly for small and simple schemas  
(< 50 elements, simple types, low degrees of nesting)
- Challenges
  - Match quality for large schemas (Increased likelihood for false matches)
  - Execution time for large schemas (quadratic complexity in schema size)
  - Advanced modeling capabilities, e.g. of W3C XSD, such as distributed schemas / namespaces, user-defined types, alternative design styles, ...
  - Context-dependent matching for shared components
  - User interface for large schemas
  - Verification of proposed match results
  - Evaluation of matchers on large schemas

# Talk Outline

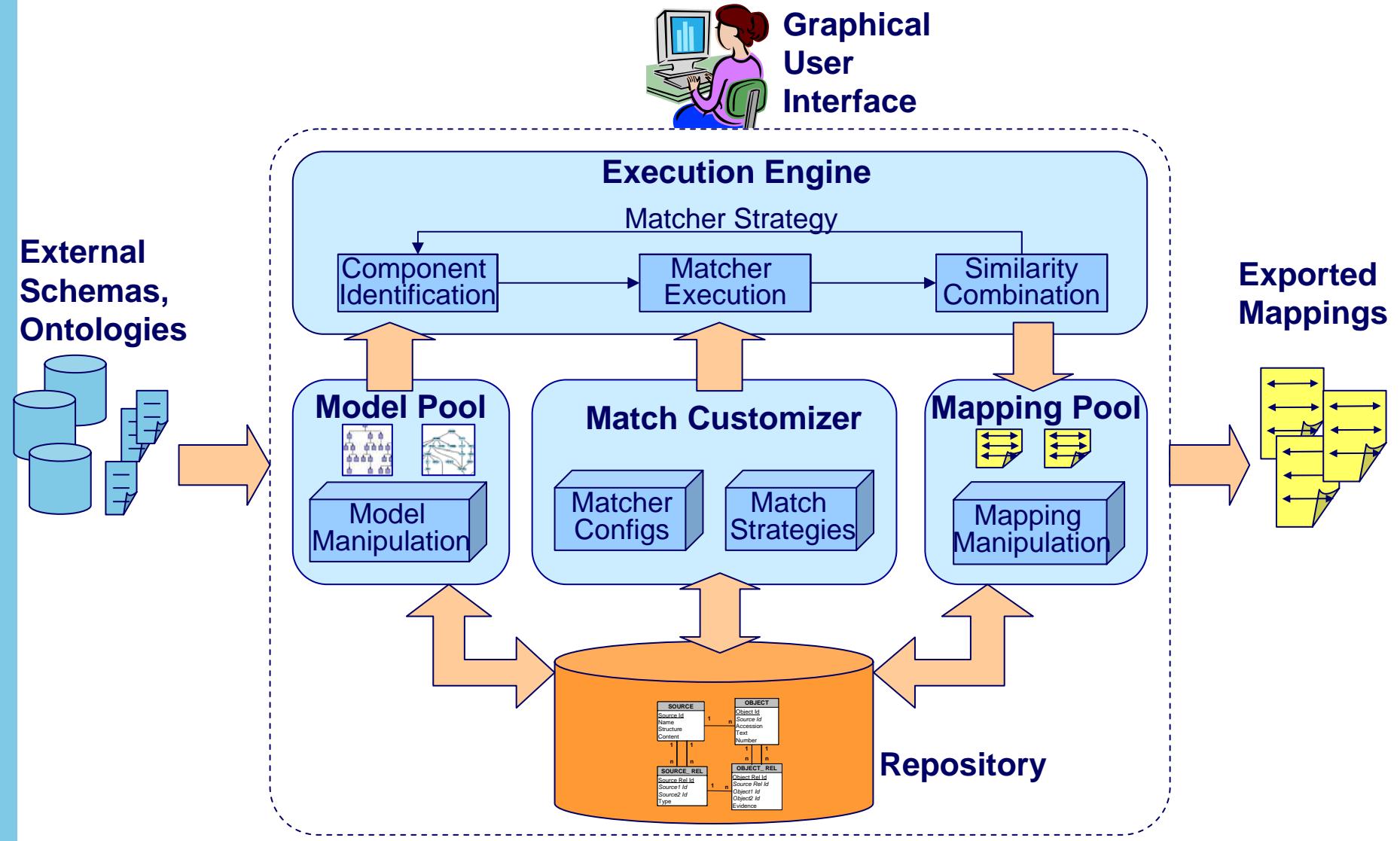
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- System architecture of COMA++
- Schema import
- Basic match processing
- Matcher construction
- Taxonomy matching / ontology support
- Context-dependent matching
- Reuse of previous match results
- Fragment-based matching
- Evaluation

# COMA++ Characteristics

- Extends previous COMA prototype (VLDB2002)
- Support of XSD, OWL and relational schemas
- Repository to store schemas and mappings (match results)
- Many matchers including a new taxonomy matcher
- Flexible construction and configuration of matchers and match strategies
- Context-dependent matching for schemas with shared components
- Fragment-based matching for large schemas
- Reuse of previous match results
- Supports comparative evaluation of matchers and match strategies
- GUI
- Much faster than COMA

# System Architecture



# Import / Design Unification

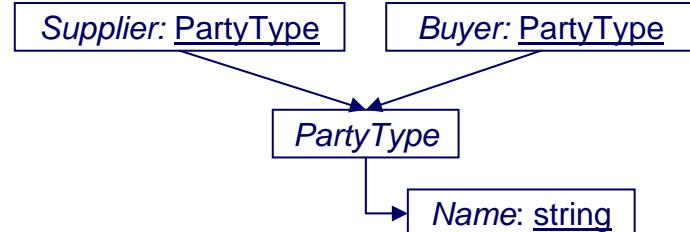
- Transform distributed schemas to monolithic

po.xsd

```
<include schemaLocation="PartyType.xsd" />
<element name="Supplier" type="PartyType" />
<element name="Buyer" type="PartyType" />
```

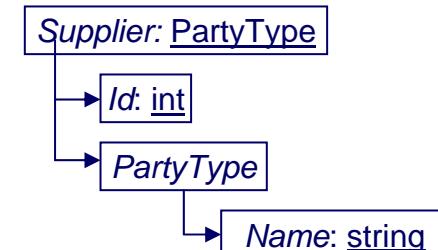
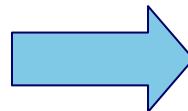
PartyType.xsd

```
<complexType name="PartyType">
    <element name="Name" type="string"/>
</complexType>
```



- Transform type derivation to composition

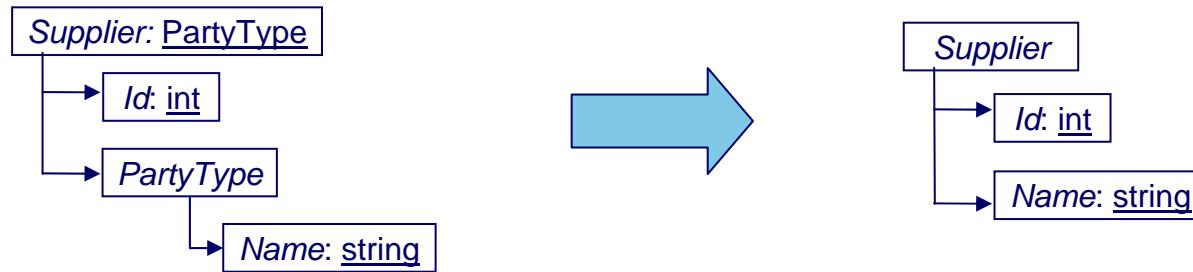
```
<complexType name="Supplier">
    <extension base='PartyType'>
        <element name="Id" type="int" />
    </extension>
</complexType>
<complexType name="PartyType">
    <element name="Name" type="string" />
</complexType>
```



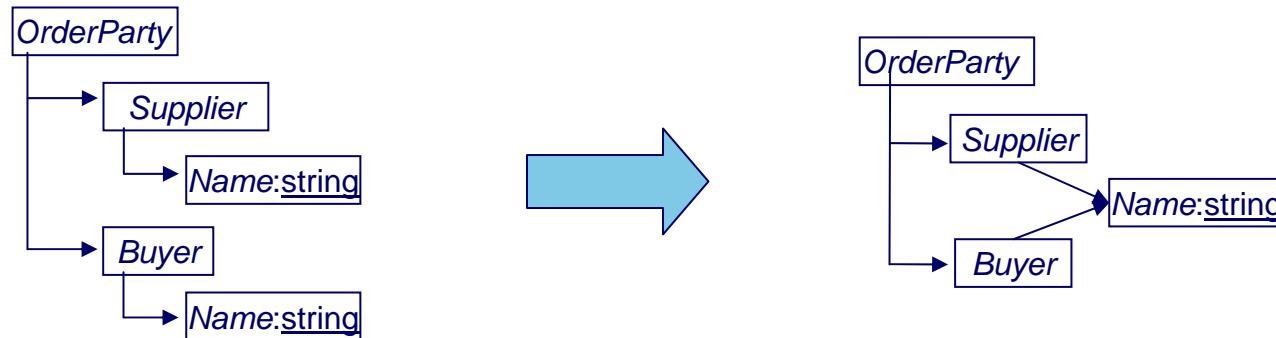
Legends: Name: Type → Containment

# Design Unification (Cont)

- Transform type reuse to element reuse



- Reducing inline declarations: Transform inline to shared

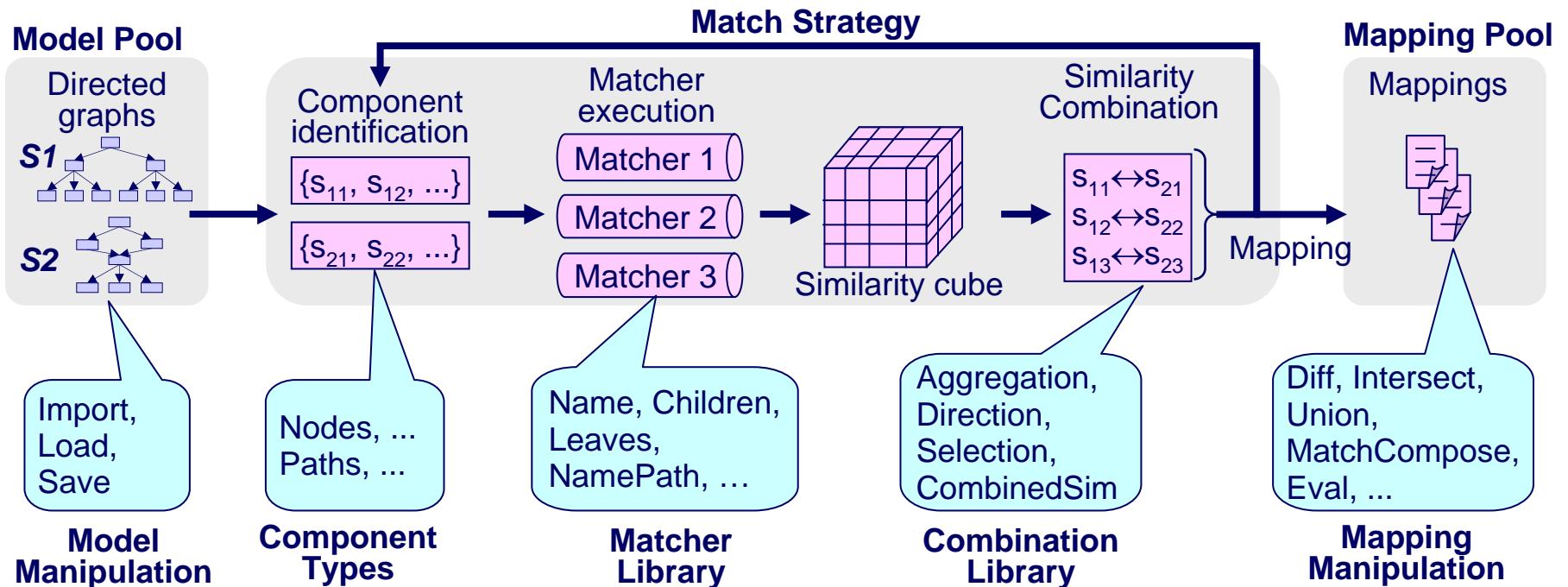


- Result:

- Connected graph of instantiable components (elements, attributes)
- Fewer schema components with max number of shared elements

# Match Processing

- *Combine*: Composite approach to combine independently executed matchers
- *Refine*: Successive refinement of previously identified match results
- Default and user configuration



# Example: Schema Elements

*match(S1, S2)*

1. Determine all paths from roots of S1, S2
2. Execute Name & NamePath matchers

S1 element	S2 element	Matcher	Sim
<i>ShipTo.shipToCity</i>	<i>DeliverTo.Address.City</i>	Name	0.6
		NamePath	0.8
<i>ShipTo.shipToStreet</i>	<i>DeliverTo.Address.City</i>	Name	0.5
		NamePath	0.7

A) Similarity Cube

3. Aggregation

Average

S1 element	S2 element	Sim
<i>ShipTo.shipToCity</i>	<i>DeliverTo.Address.City</i>	0.7
<i>ShipTo.shipToStreet</i>	<i>DeliverTo.Address.City</i>	0.6

B) Similarity Matrix

4. Direction

$|S1| < |S2|$  SmallLarge  
Ranking S1 elements  
for elements of larger schema S2

S1 element	S2 element	Sim
<i>ShipTo.shipToCity</i>	<i>DeliverTo.Address.City</i>	0.7
<i>ShipTo.shipToStreet</i>	<i>DeliverTo.Address.City</i>	0.6

C) Directional Ranking

5. Selection

Max1	S1 element	S2 element	Sim
	<i>ShipTo.shipToCity</i>	<i>DeliverTo.Address.City</i>	0.7

D) Match Results

# CombineMatcher

- Interactive construction of new matchers from existing ones
- Combining results of independently executed matchers

```
1  CombineMatcher(oType, defMatchers, agg, dir, sel)

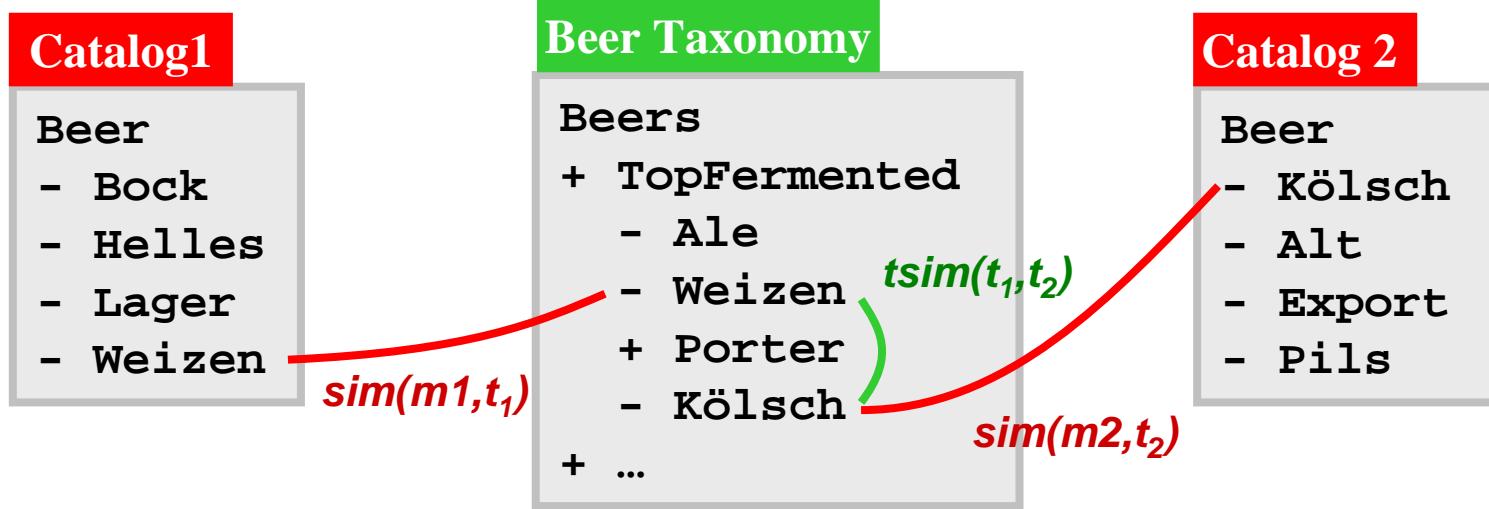
2  match(s1, s2) {
3      //Step1: Determine elements/constituents to match
4      s1.objects = determineObjects(oType, s1)
5      s2.objects = determineObjects(oType, s2)
6      //Step2: Compute similarity cube
7      allocate simCube[compMatchers][s1.objects][s2.objects]
8      for each m in defMatchers
9          for each o1 in s1.objects
10             for each o2 in s2.objects
11                 simCube[m][o1][o2] = m.sim(o1, o2)
12     //Step 3, 4, 5: Similarity combination
13     matchResult = selection(
14                     direction(
15                         aggregation(simCube, agg), dir), sel)
16     return matchResult
17 }
```

# Matcher Library

- Simple matchers:
  - String matchers: EditDistance, Trigram, ...
  - Type matcher, Synonym matcher
  - Statistics (Euclidean distance between structural statistics captured by a feature vector)
  - Taxonomy matcher, Reuse matcher
- Predefined combined matchers

Name	Elements / Constituents	Default Matchers	Combination Scheme
Name	Name tokens	Synonym, Trigram	Avg, Both, Max1, Avg
NameType	Self	Name, Type	Wgt(0.7,03), Both, Max1, Avg
NameStat	Self	Name, Statistics	
Children	Children	NameType	
Leaves	Leaves	NameType	
Parents	Parents	Leaves	Avg, Both, Max1, Avg
Siblings	Siblings	Leaves	
NamePath	Ascendants	Name	

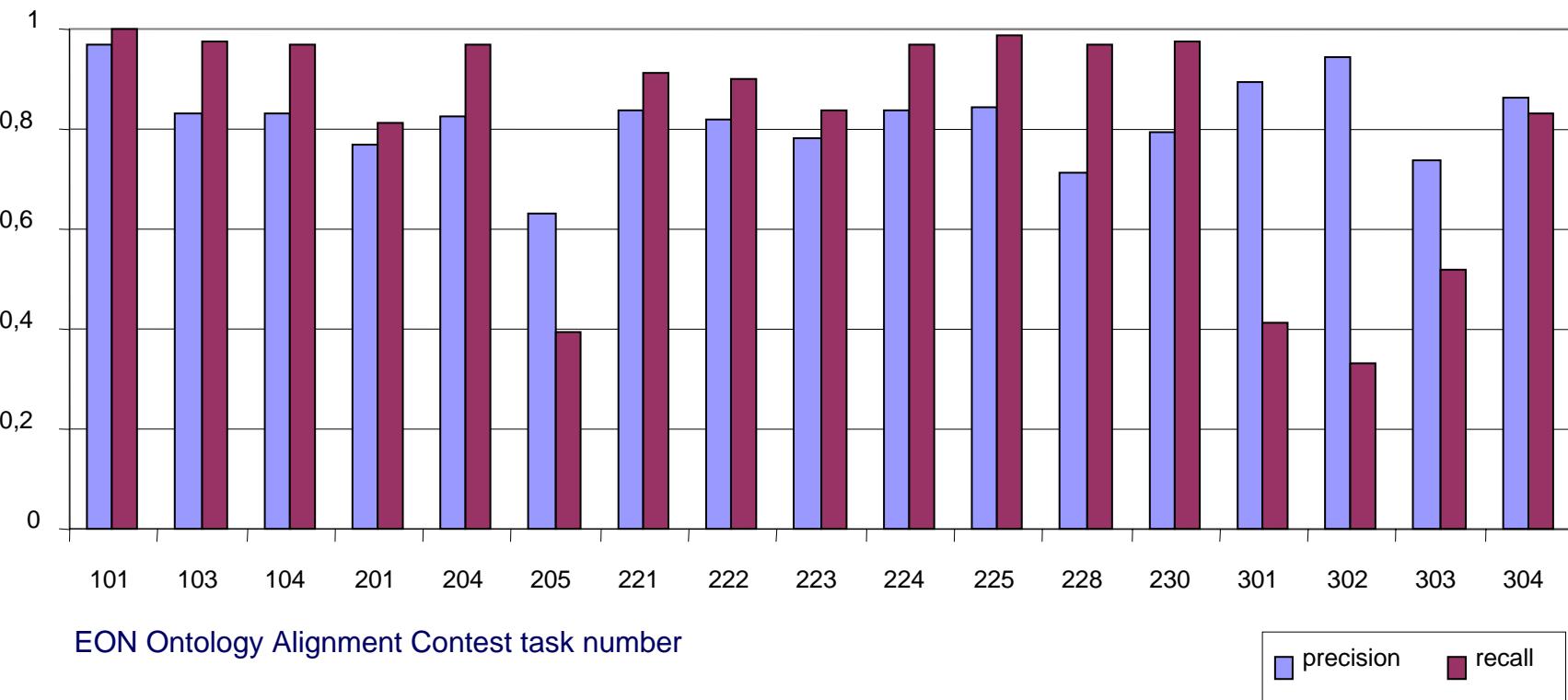
# Taxonomy Matcher



- Reference taxonomy helps find correspondences  
 $\text{sim}(\text{Weizen}, \text{Kölsch}) = 0.8$
- Similarity of schema elements → combination of  $\text{sim}(m_1, t_1)$   $\text{tsim}(t_1, t_2)$  and  $\text{sim}(m_2, t_2)$
- $\text{tsim}$ : measures semantic distance between concepts *within* taxonomy
  - different approaches possible
  - precomputation of tsim for most related concepts

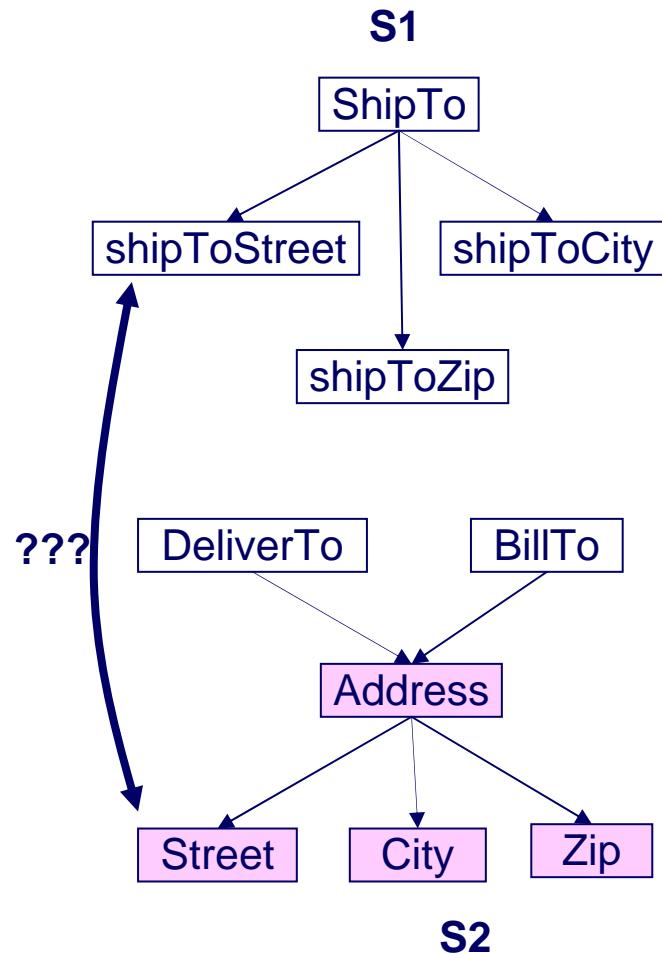
# OWL Support

- Ontology access using OWL API
- Ontology Matching can use all existing matchers, e.g. Name and structural matchers
- High effectiveness for EON Ontology Alignment test tasks even without tuning (no synonyms, etc.)



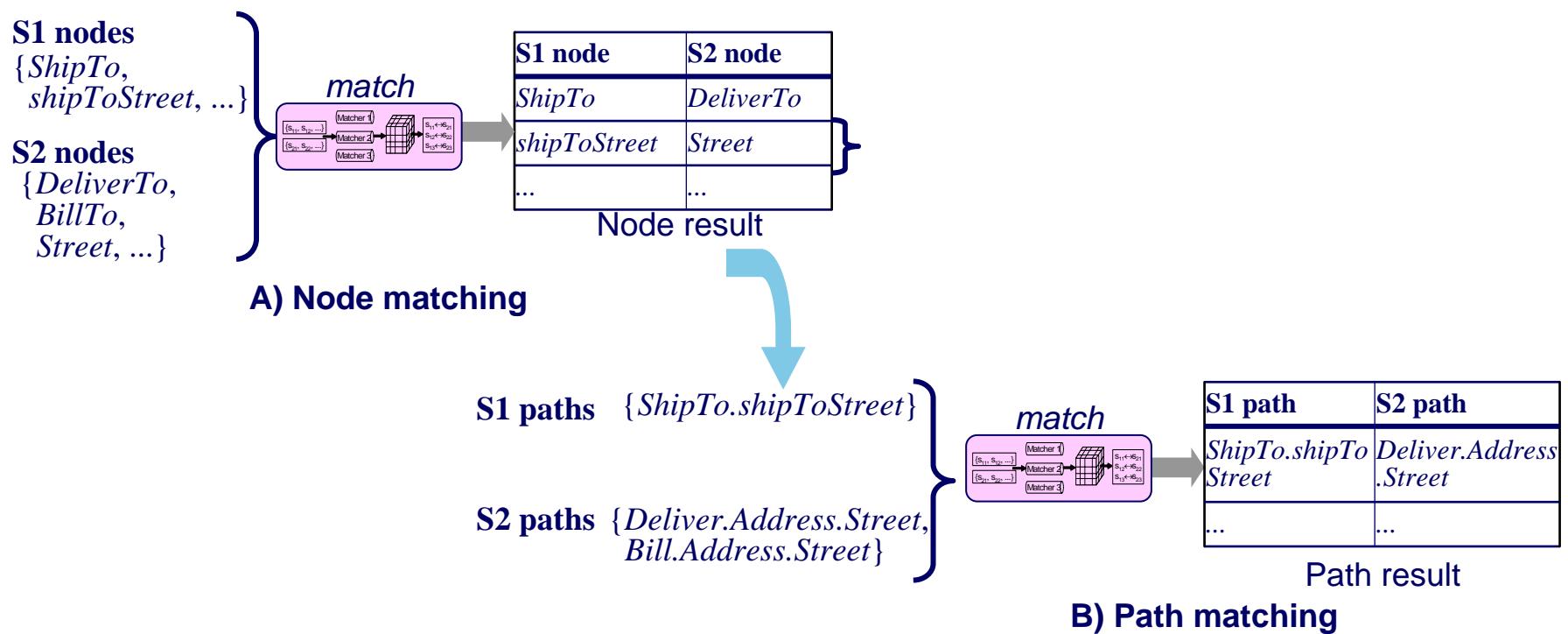
# Context-dependent Matching

- Problem: Shared components with context-dependent semantics
  - e.g. user-defined types
- **NoContext** strategy: most previous systems
  - only node correspondences
  - efficient but poor match quality:  
e.g. *shipToStreet*  $\leftrightarrow$  *Street*
- **AllContext** strategy: Cupid, COMA
  - match between all unique contexts, e.g. paths
  - scalability problems for large schemas with many shared components
  - explosion of search space ( $|S1 \text{ paths}|^*|S2 \text{ paths}|$ )



# FilteredContext Strategy

- Two-phase matching
  1. Node matching with multiple matchers
  2. Context (path) matching only for most similar node pairs
- Complexity: mainly in node matching (comparable to NoContext)
  - Significant reduction of complexity (if  $|nodes| \ll |paths|$ )



# Reuse of Previous Match Results

- Example: Use result for S1—S2 to match S1'—S2

## Schema S1'

Purchase-order2  
  Product  
  BillTo  
    Name  
    Address  
  ShipTo  
    Name  
    Address  
  ContactPhone

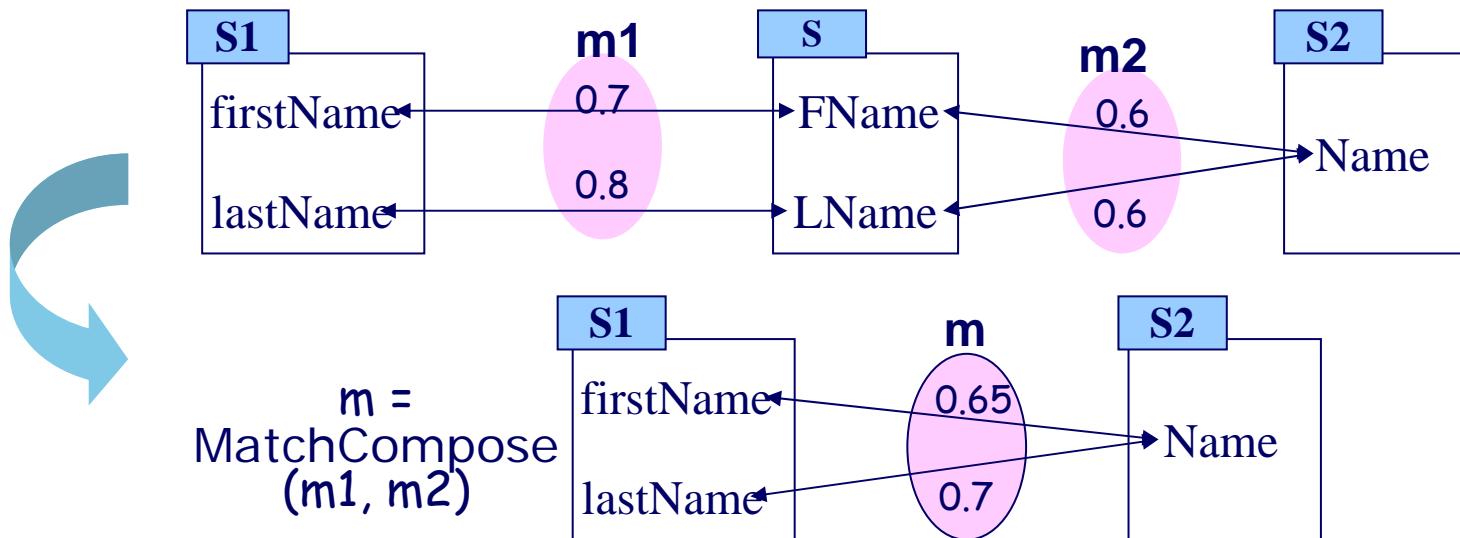
## Schema S1

Purchase-order ← → POrder  
Product ← → Article  
BillTo ← → Payee  
  Name ← → BillAddress  
  Address ← → Recipient  
ShipTo ← → ShipAddress  
  Name ← →  
  Address ← →  
Contact  
  Name  
  Address

## Schema S2

# Reuse (2)

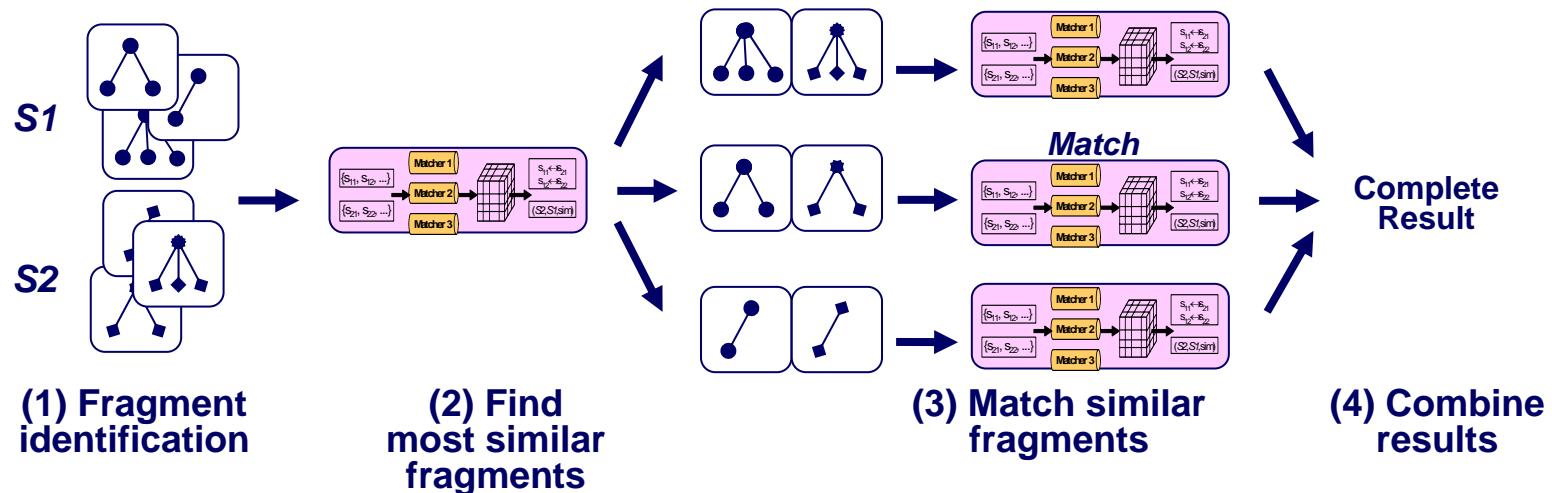
- MatchCompose operation



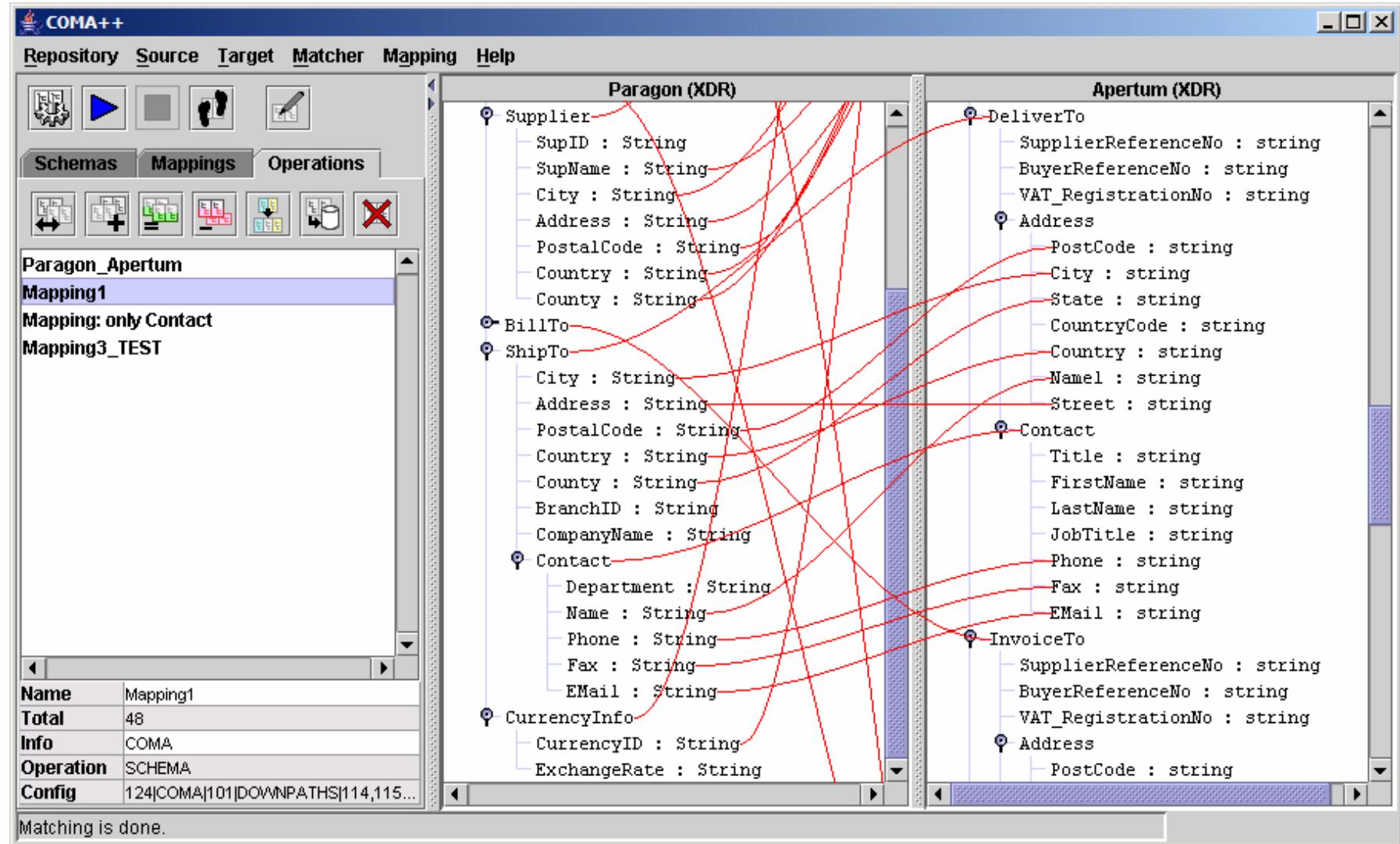
- COMA++ reuse options using mapping repository (match problem S1-S2 )
  - Use already existing direct mapping(s) S1-S2
  - Look for *mapping paths* (S1-S3-S2, S2-S4-S5-S1, ...)
  - Use of a *pivot schema* (global schema) P: Compose S1-P with P-S2
  - Look for similar mappings, e.g. for different schema versions

# Fragment-based Matching

- Decompose large match problems into smaller ones
  - Reduce search space and potential for false matches
  - Simplify user control
- Fragments: Rooted subgraphs down to leaf level
  - Special case: complete schema
  - Instantiable subschemas, e.g. tables, message formats
  - Shared schema components
  - user-selected fragments



# Graphical User Interface



# Real-world Evaluation

- PO schemas (XDR) & E-Business standards (XSD)

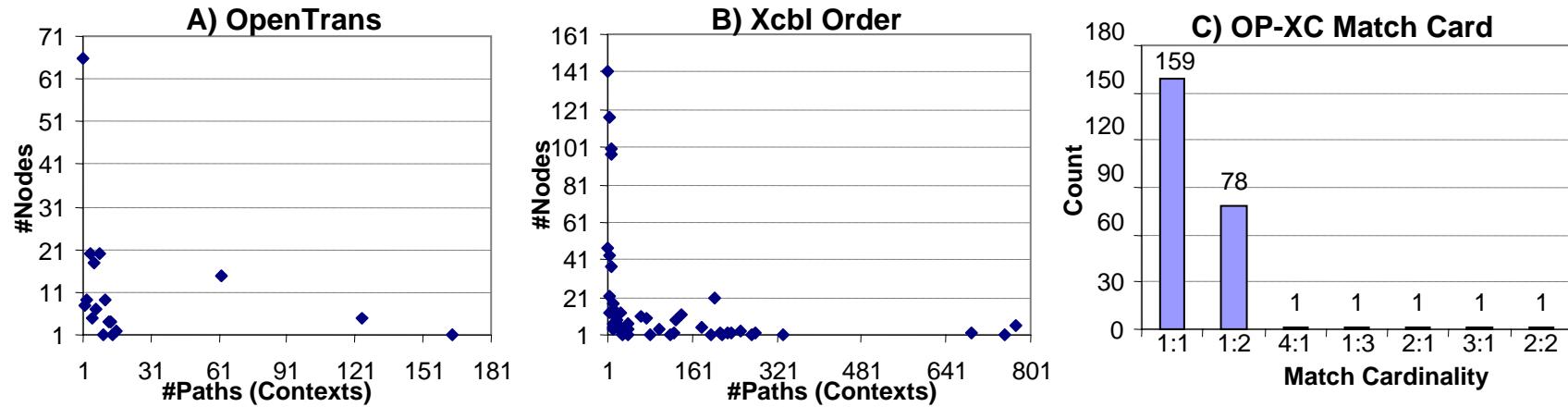
#	Schema	#Nodes	#Roots / Inners/ #Leaves / Shared	#Paths	Max/Avg Path Len
1	CIDX	27	1 / 7 / 20 / 7	34	4 / 2.9
2	Excel	32	1 / 9 / 23 / 11	48	4 / 3.5
3	Noris	46	1 / 8 / 38 / 18	65	4 / 3.2
4	Paragon	59	1 / 11 / 48 / 13	77	6 / 3.6
5	Apertum	74	1 / 22 / 52 / 24	136	5 / 3.6
6	OpenTrans	195	8 / 85 / 110 / 129	2,500	11 / 7
7	XCBLOrder	843	10 / 382 / 461 / 702	26,228	18 / 8.8

# Match Tasks

- 16 match tasks in 3 series

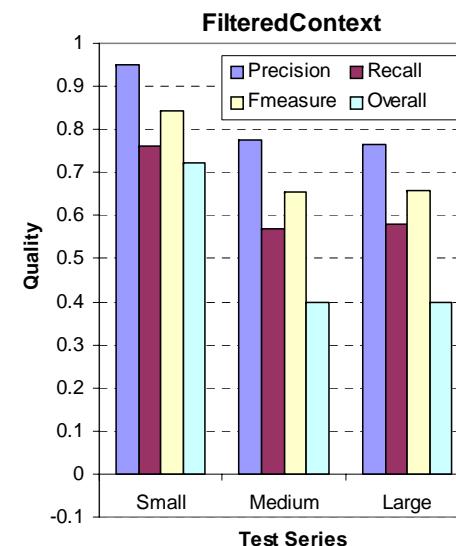
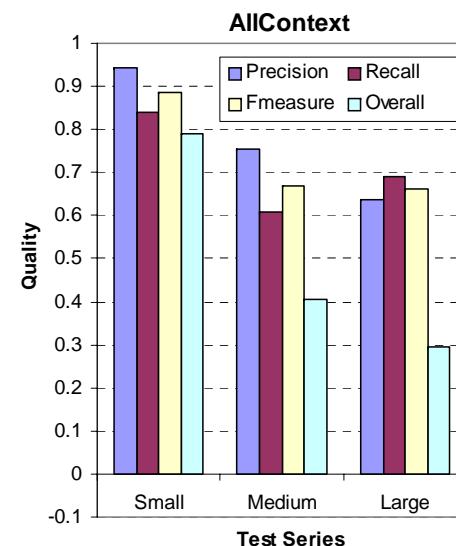
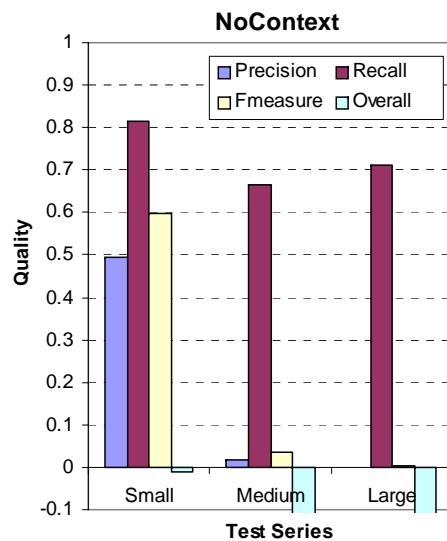
Series	Match tasks	#Tasks	Avg Source paths	Avg Target paths	Avg Corresp	Avg Schema Sim
Small	PO-PO	10	49	95	48	0.57
Medium	PO-OP	5	72	2,500	55	0.04
Large	OP-XC	1	2,500	26,228	331	0.02

- Largest match task: OpenTrans – XcbIOrder
  - # contexts per node up to 160 in OpenTrans and 800 in XcbI Order
  - many m:n match relationships
  - traditional approaches, i.e. AllContext, MaxN or Threshold, unlikely successful



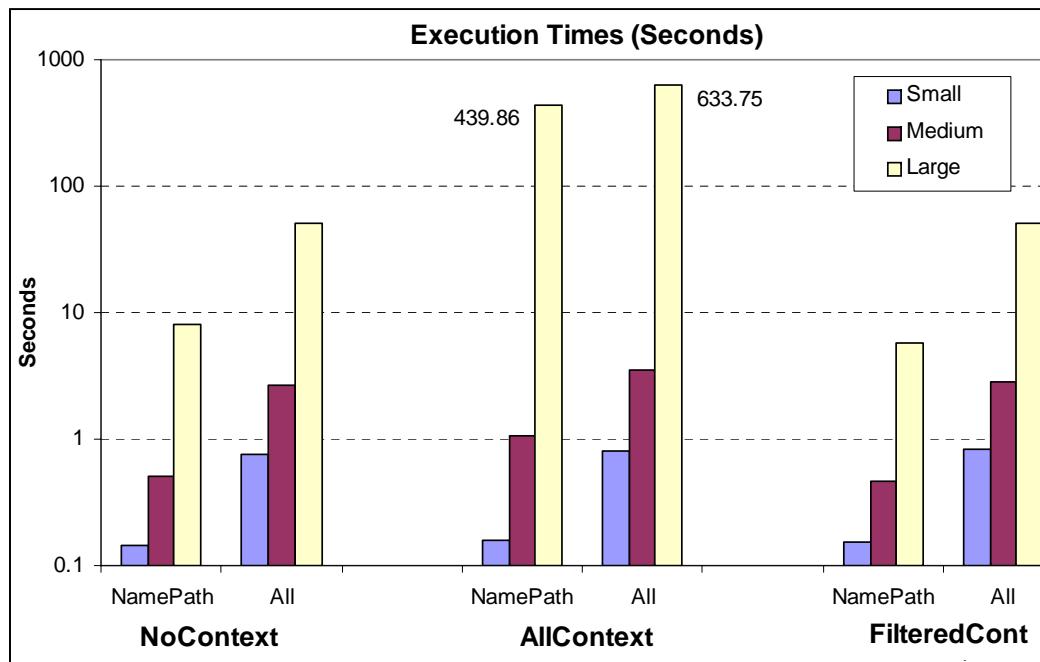
# Context Matching: Quality

- Quality decreases with increasing schema size, i.e. harder problem
- NoContext: Low quality even in small series
  - high recall but low precision, i.e. many false matches, fmeasure ~0.0
  - Unacceptable with many shared components
- FilteredContext slightly worse than AllContext in Small, but equal in Medium, Large



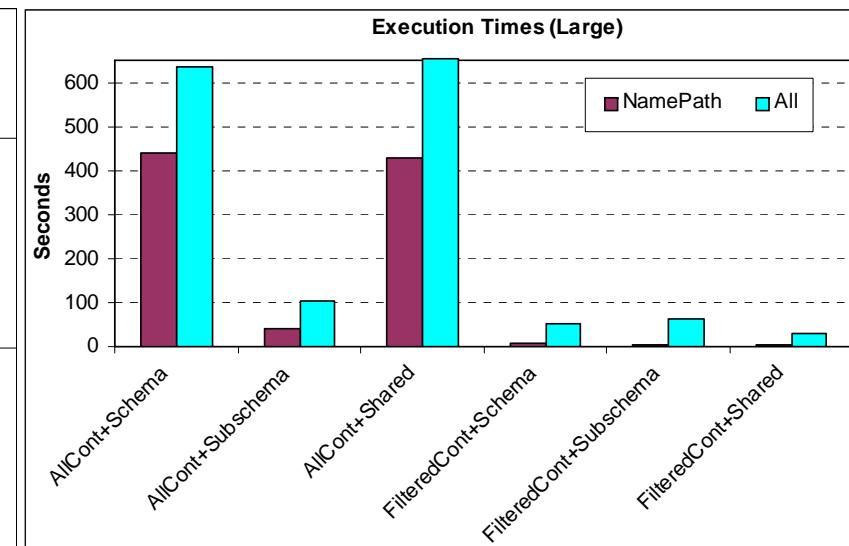
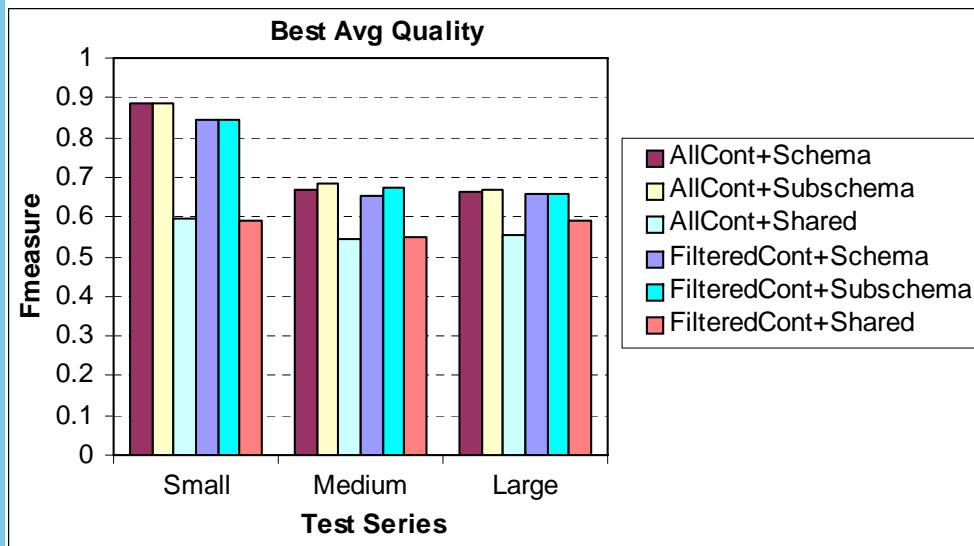
# Context Matching: Time

- 1 matcher (NamePath) vs. 8 matchers (All)
- Time increases with schema size & number of matchers
- Similar, fast execution times in Small, Medium, but differences in Large
- NoContext: fast execution times, max ~50 secs for Large
- AllContext: not scalable, 7-10 min for Large
- FilteredContext: ~ NoContext indicating negligible effort for path matching -> best context-dependent strategy for large schemas



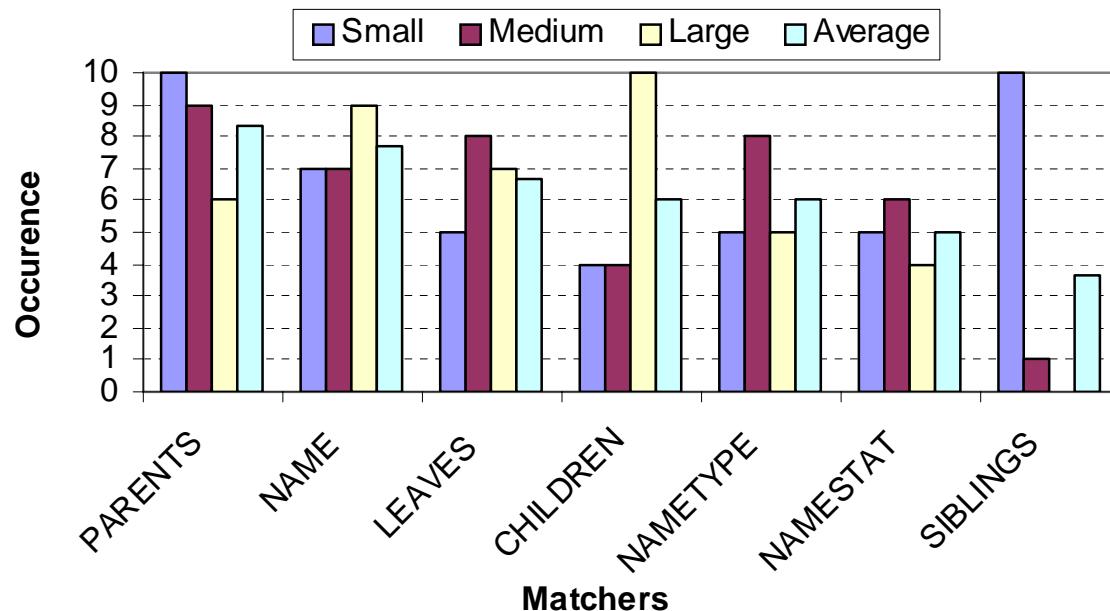
# Fragment Matching: Quality and Time

- Quality:
  - Subschema equal or better than Schema due to reduced search space
  - Shared worst due to incomplete schema coverage, still attractive for large schemas
- Time (Large series):
  - Significant improvement of Subschema over Schema in AllContext
  - Shared ~ Schema in AllContext: large number of Shared elements (i.e. fragments) to be compared
  - Similar, fast times between fragment types in FilteredContext



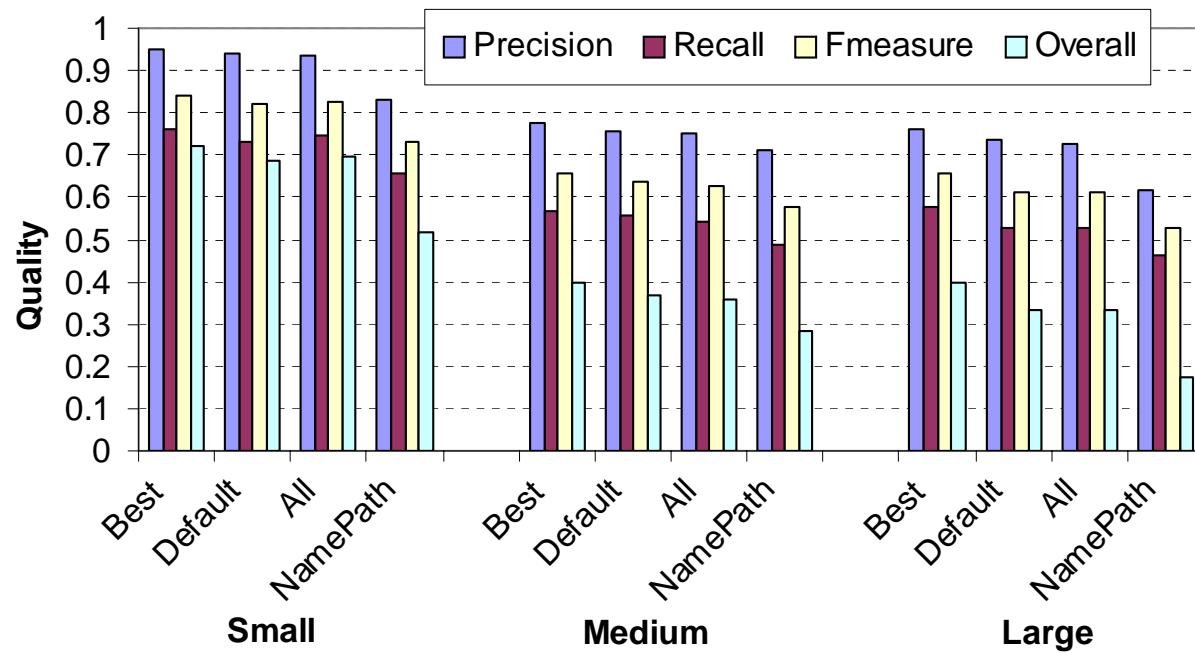
# Combination of Matchers (1)

- *Method:*
  - 128 matcher combinations (NamePath + selection from 7 other matchers)
  - For each matcher, determine #occurrences within the top-10 matcher combinations
- FilteredContext: more stable matcher occurrence than AllContext
  - Parents, Names, Leaves with high occurrence ( $>=5$ ) in all series
  - -> Default configuration: NamePath + Parents, Name, Leaves



# Combination of Matchers (2)

- *Method*: Best average Fmeasure of Best, Default, All and NamePath for FilteredContext
- NamePath:
  - worse quality than others
  - combining several matchers for better quality
- Default ~ Best ~ All in all series
  - All: very expensive, Default: good combination with only 4 matchers



# Conclusions and Future Work

- Comprehensive platform for schema and ontology matching
  - Flexible construction of new matchers, match strategies
  - Strategies for context-dependent matching, fragment-based matching, reuse
  - Repository of schemas and mappings
  - GUI
- Evaluation results
  - New match strategies with high quality and fast execution times for large schemas
  - Fmeasure ~0.9 for small tasks, ~0.7 for large tasks
  - FilteredContext combines good quality and fast execution
  - Fragment matching: Subschema better than Schema in both quality and time
  - Default combination of 4 matchers (NamePath, Parents, Name, Leaves) delivers good results
- Future work
  - Evaluation of reuse
  - Additional approaches for ontology matching
  - Instance matching

# References

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