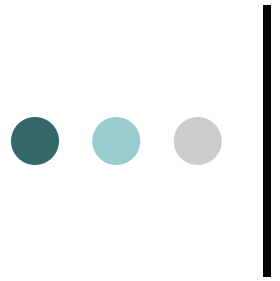


Inducing Source Definitions for Web Service Composition

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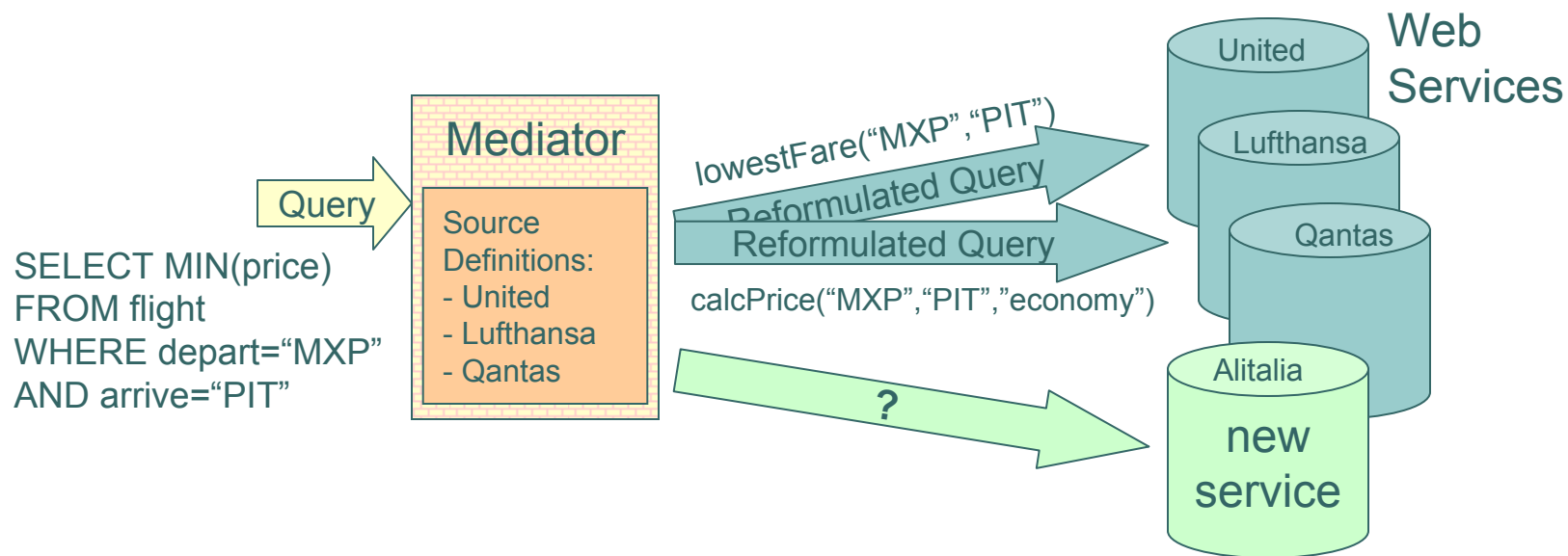


Overview of the Talk

- Mediators for composing services
- Inducing source definitions: A simple example
- Generation & test framework
- Case study & preliminary experiments
- Challenges & future work
- Related work

Mediators for Composing Web Services

- Provide uniform access to heterogeneous sources
- Source definitions are used to reformulate query
- New service, no source definition, no integration!
- Can we discover definitions automatically?



Inducing Source Definitions: A Simple Example

- Step 1: use metadata to classify input types (\$)
- Step 2: invoke service and classify output types

Mediator

Semantic Types:

currency \supseteq {USD, EUR, AUD}

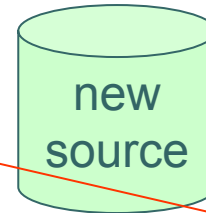
rate \supseteq {1936.2, 1.3058, 0.53177}

Predicates:

exchange(currency, currency, rate)



LatestRates(\$country1,\$country2,rate):-
exchange(country1,country2,rate)



currency

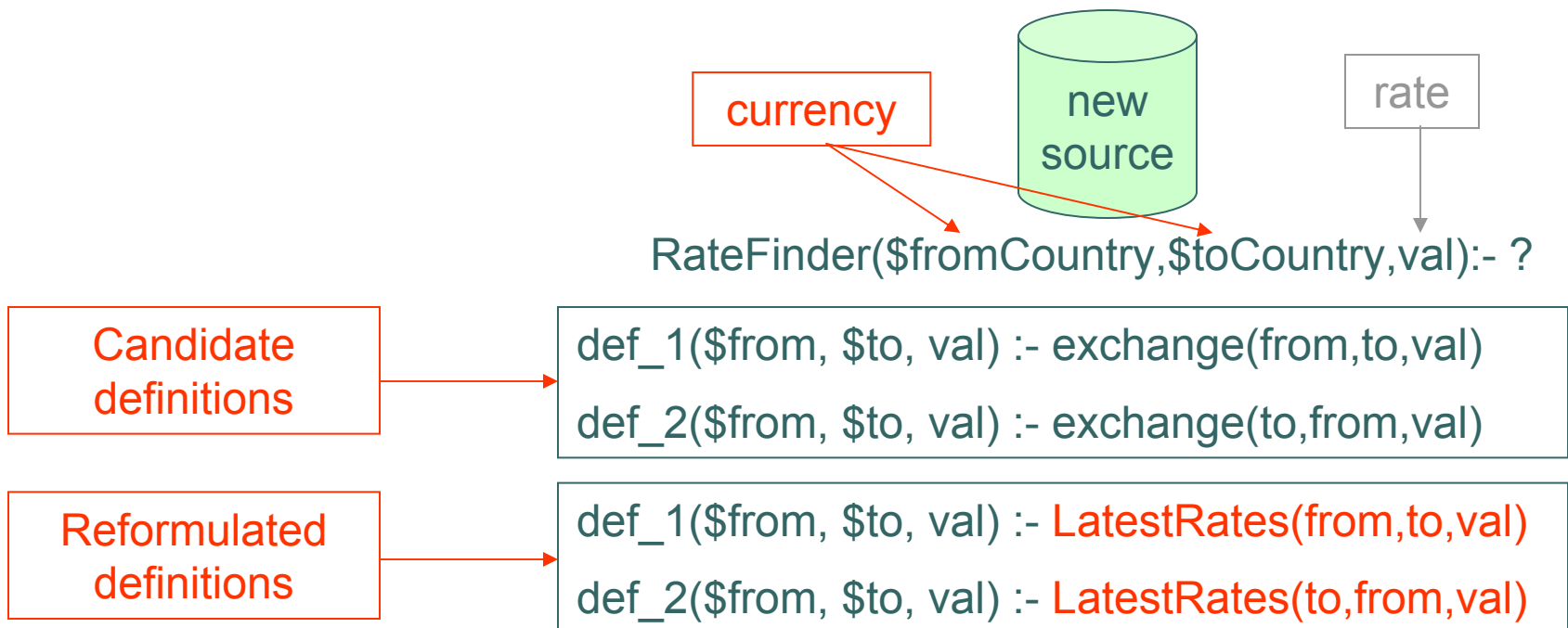
rate

RateFinder(\$fromCountry,\$toCountry,val):- ?

{<EUR,USD,1.30799>,<USD,EUR,0.764526>,...}

Inducing Source Definitions: A Simple Example

- Step 3: generate plausible source definitions
- Step 4: reformulate in terms of other sources



Inducing Source Definitions: A Simple Example

- Step 5: invoke services and compare output

```
def_1($from, $to, val) :- exchange(from,to,val)
def_2($from, $to, val) :- exchange(to,from,val)
```

```
def_1($from, $to, val) :- LatestRates(from,to,val)
def_2($from, $to, val) :- LatestRates(to,from,val)
```

match

Input	RateFinder	Def_1	Def_2
<EUR,USD>	1.30799	1.30772	0.764692
<USD,EUR>	0.764526	0.764692	1.30772
<EUR,AUD>	1.68665	1.68979	0.591789



The Framework

Intuition: *Services often have similar semantics, so we should be able to use what we know to induce that which we don't*

Two phase algorithm

For each operation provided by the new service:

1. Classify its input/output data types
 - *Classify* inputs based on metadata similarity
 - *Invoke* operation & classify outputs based on data
2. Induce a source definition
 - *Generate* candidates via Inductive Logic Programming
 - *Test* individual candidates by reformulating them



Comparing Candidate Definitions

Sources may return *multiple tuples* for each input

&

Sources may be *incomplete*

- Use Record Linkage to discover common tuples
- Compare candidate definitions using:

$$score(def) = \frac{|src \cap def|}{|src| + |def|}$$

- Approximate score through sampling
- Terminate search when highest score converges:

$$\frac{mean(score(def_1) - score(def_2))}{\sqrt{variance(score(def_1) - score(def_2)) / N}} \geq t_value(0.05, N)$$



Use Case: Zip Code Data

- Single *real* zip-code service with multiple operations
- The first operation is defined as:

```
getDistanceBetweenZipCodes($zip1, $zip2, distance) :-  
    centroid(zip1, lat1, long1),  
    centroid(zip2, lat2, long2),  
    distanceInMiles(lat1, long1, lat2, long2, distance).
```

- Goal is to induce definition for a second operation:

```
getZipCodesWithin($zip1, $distance1, zip2, distance2) :-  
    centroid(zip1, lat1, long1),  
    centroid(zip2, lat2, long2),  
    distanceInMiles(lat1, long1, lat2, long2, distance2),  
    (distance2 ≤ distance1),  
    (distance1 ≤ 300).
```

- Same service so no need to classify inputs/outputs or match constants!

Generating definitions: ILP

- Want to induce source definition for:

```
getZipCodesWithin($zip1, $distance1, zip2, distance2)
```

- Predicates available for generating definitions:

{centroid, distanceInMiles, ≤, =}

- New type signature contains that of known source
 - Use known definition as starting point for local search:

```
getDistanceBetweenZipCodes($zip1, $zip2, distance) :-  
    centroid(zip1, lat1, long1),  
    centroid(zip2, lat2, long2),  
    distanceInMiles(lat1, long1, lat2, long2, distance).
```

Generating definitions: ILP

- Want to induce source definition for:

`getZipCodesWithin($zip1, $distance1, zip2, distance2)`

	Plausible Source Definition	
1	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d1), (d2 = d1)	INVALID d2 unbound!
2	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d1), (d2 ≤ d1)	#d is a constant
3	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d2 ≤ d1)	
4	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d1 ≤ d2)	UNCHECKABLE lt1 inaccessible!
5	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d1 ≤ #d)	
6	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (lt1 ≤ d1)	contained in defs 2 & 4
	...	
n	cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d2 ≤ d1), (d1 ≤ #d)	



Testing definitions

Checking definitions requires LOTS of queries to sources!

- Reformulation's binding constraints may be different:

```
def_1($zip1, $distance1, zip2, distance2) :-  
    centroid(zip1, lat1, long1),  
    centroid(zip2, lat2, long2),  
    distanceInMiles(lat1, long1, lat2, long2, distance1),  
    (distance2 = distance1).
```

```
def_1($zip1, distance1, $zip2, distance2) :-  
    getDistanceBetweenZipCodes($zip1, $zip2, distance1),  
    (distance2 = distance1).
```

- Should invoke operation with every possible zip code!
- Don't want to be banned from using the service!

Implementation:

1. Store output tuples for reuse across definitions & trials
2. Sample to estimate score for \forall -type queries



Preliminary Results

Settings:

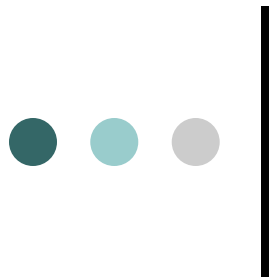
- Number of zip code constants initially available: 6
- Number of samples performed per trial: 20
- Number of candidate definitions in search space: 5

Results:

- Converged on “almost correct” definition!!!

```
getZipCodesWithin($zip1, $distance1, zip2, distance2) :-  
    centroid(zip1, lat1, long1),  
    centroid(zip2, lat2, long2),  
    distanceInMiles(lat1, long1, lat2, long2, distance2),  
    (distance2 ≤ distance1),  
    (distance1 ≤ 243).
```

- Number of iterations to convergence: 12, never, ...
- **Lesson learned:** Need strategy for selecting inputs!



Active Input Selection

- Idea: *Select input tuples which best differentiate the two best performing candidates*
- Sometimes it is possible to select inputs that are guaranteed not to return tuples for one definition:

`cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d2 ≤ d1)`

`cen(z1,lt1,lg1), cen(z2,lt2,lg2), dIM(lt1,lg1,lt2,lg2,d2), (d2 ≤ d1), (d1 ≤ 243)`

- Useful only if we can check this property *without* accessing any sources
 - the predicates involved must be interpreted



Challenges & Future Work

- Need methodology for selecting inputs
 - Random strategy results in very long convergence times
 - **Actively select inputs** to best differentiate candidates!
 - Take variable type into account (nominal or numeric?)
- Number of tuples needed for effective sampling
 - Depends on number of trials performed thus far
 - Possibly also on number of known constants
- Compare local and global ILP search
- Need methodology for assigning constants in definitions



Related Work

- **Classifying Web Services**
(Hess & Kushmerick 2003), (Johnston & Kushmerick 2004)
 - Classify input/output/services using metadata/data
 - We learn semantic relationships between inputs & outputs
- **Category Translation**
(Perkowitz & Etzioni 1995)
 - Learn functions describing operations available on internet
 - We concentrate on a relational modeling of services
- **CLIO**
(Yan et. al. 2001)
 - Helps users define complex mappings between schemas
 - They do not automate the process of discovering mappings
- **iMAP**
(Dhamanka et. al. 2004)
 - Automates discovery of certain complex mappings
 - Our approach is more general (ILP) & tailored to web sources
 - We must deal with problem of generating valid input tuples