CompSci 516 Data Intensive Computing Systems

Lecture 23 Data Integration

Instructor: Sudeepa Roy

Duke CS, Fall 2016

Announcements

• No class next week

- thanksgiving recess!
- We meet again on 11/30 (Wed)
- Final report first draft due on 11/28 (Mon) night

but can update until Friday 12/2 night

- send me an email if you update
- I will post a message on piazza looking for three groups who will present on 11/30 (Wed)
 - the remaining seven groups present on 12/2 (Fri)
 - 10 minutes talk/demo for each group (8 mins talk + 2 mins questions)

Today's topic

• An overview of data integration

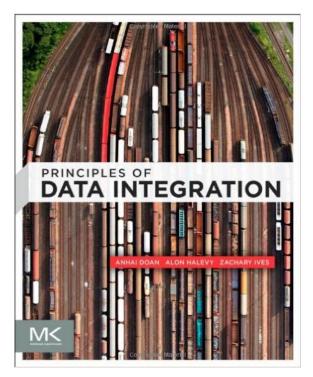
• Some optional additional slides at the end

Reading Material

Optional Reading:

• The "Principles of Data Integration" book by AnHai Doan, Alon Halevy, Zack Ives The lecture slides are based on Ch. 1, 3, 5 of this book

 Data integration PODS 2005 tutorial by Phokion Kolaitis (more on the theoretical aspects)



What is Data Integration? 1/2

- Internet and WWW have revolutionized people's access to digital data
- We take it for granted that a search query into a browser taps into millions on documents and databases and returns what we are looking for
- Systems on the Internet must efficiently and accurately process and serve a large amount pf data

What is Data Integration? 2/2

- Unlike traditional RDBMS, the new services need the ability to
 - Share data among multiple organizations
 - Integrate data on a flexible and efficient fashion
- Data integration:
 - A set of techniques that enable building systems geared for flexible sharing and integration of data across multiple autonomous data providers.

Why data integration? 1/2

- With issues like normalizations, and trade-offs in design choices, different people design different schemas for the same data
- Sometimes different needs as well
 - not all attributes are needed by all people
- Sometimes people want to share their data
 - collaborators
 - researchers who want to publish data for others' use

Why data integration? 2/2

• In the Web,

- Many websites posting job applications, hotel or flight deals, movie information
- To keep up with new information and for new need, you may have to look at all of them
- But now there are websites where you can access all
- e.g. TripAdvisor helps you see the price of the same hotel on the hotel website, hotels.com, booking.com, expedia,
- But this type of data integration has its challenges too

Why data integration? 2/2

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Challenges in Data Integration: 1. Query

 Offer uniform access to a set of autonomous and heterogeneous data sources

• Query:

query disparate data sources, sometimes update them

Challenges in Data Integration: 2. Number of sources

- Number of sources:
 - challenging even for 10 or 2 data sources
 - amplified for hundreds of sources say in Webscale

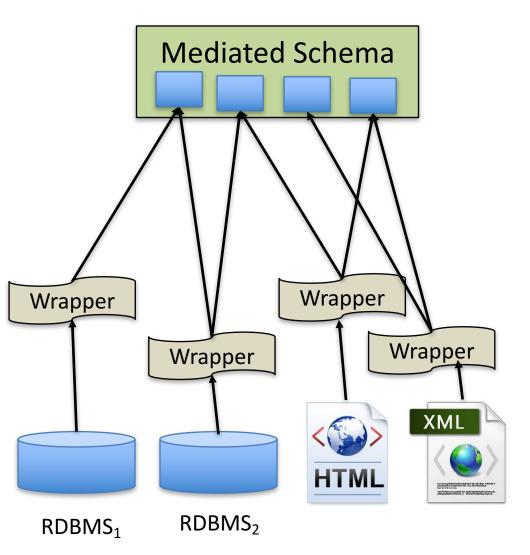
Challenges in Data Integration: 3. Heterogeneity

- Heterogeneity:
 - data sources were developed independently of each other
 - databases, files, html
 - different schema and references
 - some structured some unstructured

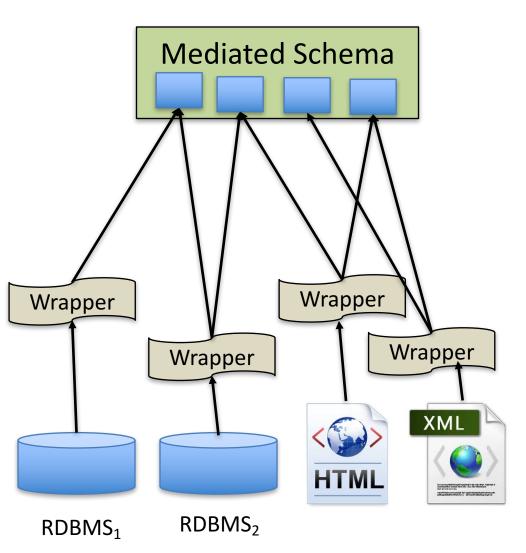
Challenges in Data Integration: 4. Autonomy

• Autonomy:

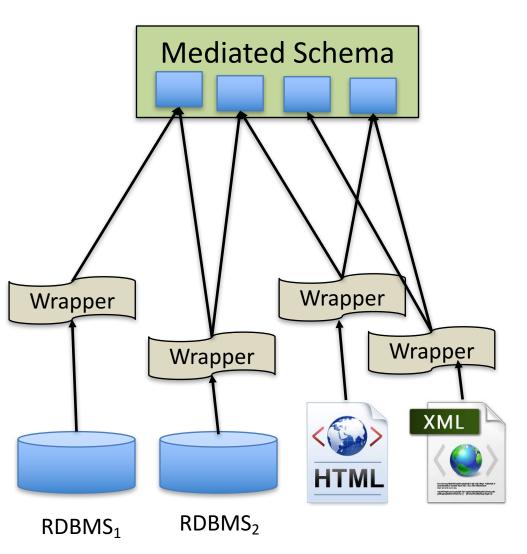
- the sources may not belong to the same administrative entity
- even then may be run by different organizations
- may not have full access to the data
- there may be privacy concerns
- the sources may change their formats and access patterns at any time without notifying



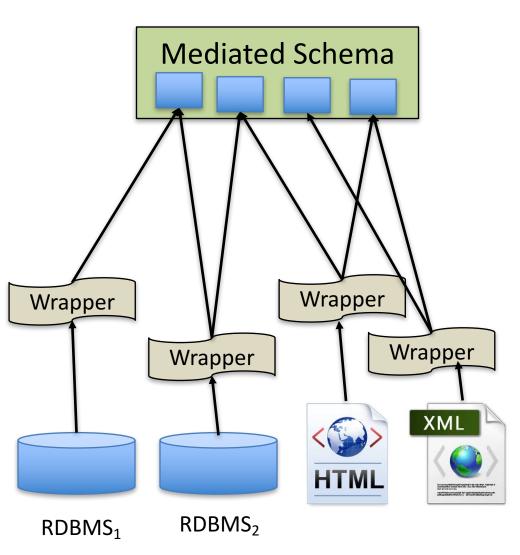
- Three components
 - Data sources
 - Wrappers
 - Mediated Schema



- Data sources
 - can be any data
 model like
 relational dbms
 with SQL interface
 - XML with Xquery interface
 - HTML

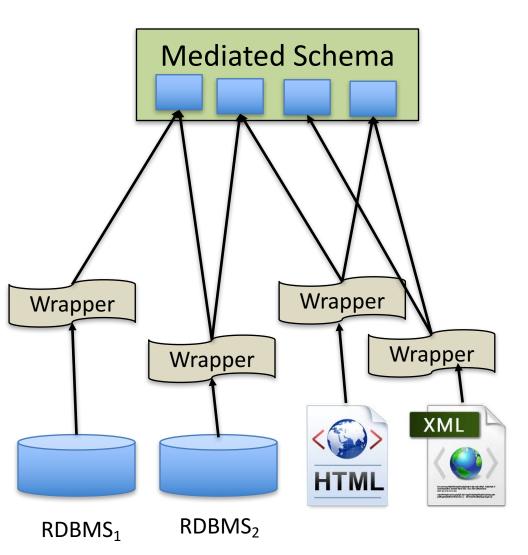


- Wrappers
 - programs that send queries to a data source
 - receives answers
 - apply some basic transformations
- e.g. to a web form source
 - translate query to a http request with a url
 - when the answer comes back as an html file, extract tuples



- Mediated schema
 - built for the data integration application
 - contains only the aspects that are relevant
 - may not contain all attrbutes
 - does not store any data typically
 - logical schema for posing queries by the users

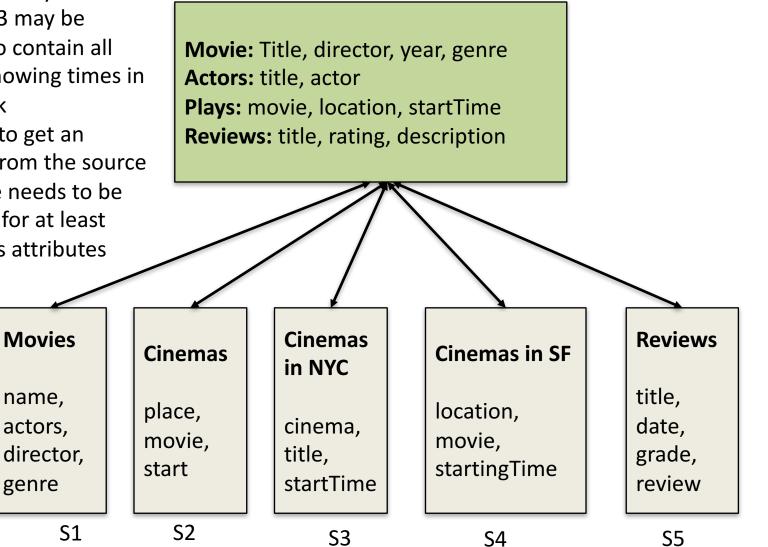
Source Descriptions



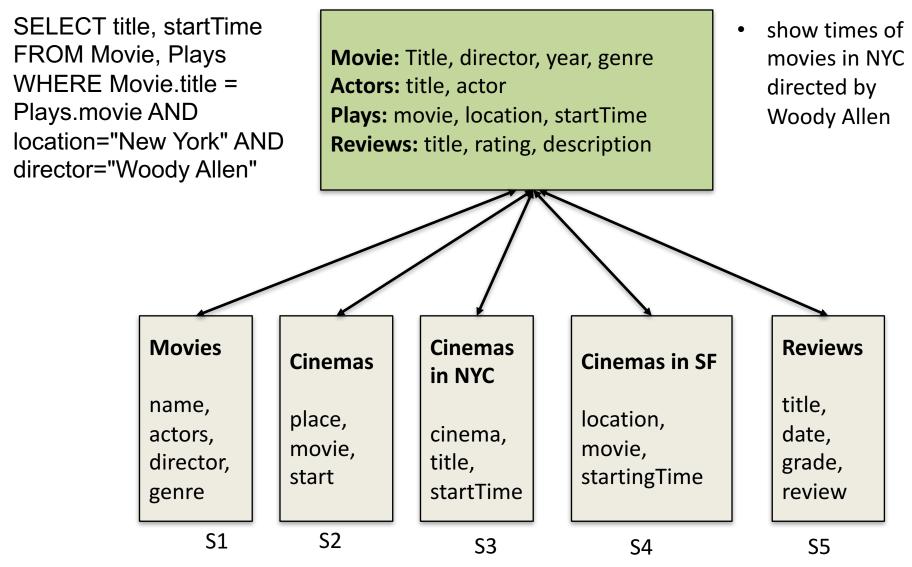
- Specify the property of the sources that the system needs to know
- main components are semantic mappings
 - relate the schema of the sources to the attributes in the mediated schema
- specified declaratively
- between data sources and mediated schema
 - not between two sources
- also specifies
 - whether sources are complete or not
 - limited access patterns to sources

- source S2 may not contain all the movie showing times in the entire country
- source S3 may be known to contain all movie showing times in New York
- in order to get an answer from the source S1, there needs to be an input for at least one of its attributes

Example



Example: Query on the Mediated Schema



- S2: may not contain all the movie showing times in the entire country
- S3: known to contain all • movie times in NYC
- S1: to get an answer ٠ there needs to be an input for at least one of its attributes

SELECT title, startTime **FROM Movie**, Plays WHERE Movie.title = Plays.movie AND location="New York" AND director="Woody Allen"

Movie: Title, director, yea genre Actors: title, actor Plays: movie, location, sta Reviews: title, rating, des

Example: Reformulation on source databases: 1/5

- Tuples for Movie can be obtained from source **S1**
 - but the attribute title needs to be reformulated to name

	Movies	Cinemas	Cinemas in NYC	Cinemas in SF	Reviews
ear, tartTime	name, actors, director, genre	place, movie, start	cinema, title, startTime	location, movie, startingTime	title, date, grade, review
escription	S1 CompSci 5	S2 516: Data Intensive	S3 Computing System	S4	S5

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Movie: Title, director, genre Actors: title, actor Plays: movie, location **Reviews:** title, rating,

Example: Reformulation on source databases: 2/5

- Tuples for Plays can be obtained from ٠ either source S2 or S3
 - Since the latter is complete for showings in NYC, we choose it over S2

en"	Movies	Cinemas	Cinemas in NYC	Cinemas in SF	Reviews
r, year, n, startTin , descripti	name, actors, director, genre	place, movie, start	cinema, title, startTime	location, movie, startingTime	title, date, grade, review
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Movie: Title, director, year, genre Actors: title, actor Plays: movie, location, startTime Reviews: title, rating, descriptio

Example: Reformulation on source databases: 3/5

- Source S3 requires the title of a movie as input
 - but such a title is not specified in the query
 - the query plan must first access source S1
 - then feed the movie titles returned from S1 as inputs to S3

	Movies	Cinemas	Cinemas in NYC	Cinemas in SF	Reviews
artTime cription	name, actors, director, genre	place, movie, start	cinema, title, startTime	location, movie, startingTime	title, date, grade, review
cription	S1 CompSci 5	S2 516: Data Intensive	S3 Computing System	S4	S5 23

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Movie: Title, director, year, genre Actors: title, actor Plays: movie, location, startTim Reviews: title, rating, description

Example: Reformulation on source databases: 4/5

- Options of logical query plan:
 - access S1, S3
 - could access S1 then S2 as well (possibly not complete)

Then query optimization

- as in traditional database system
- take a logical plan output a physical plan

	Movies	Cinemas	Cinemas in NYC	Cinemas in SF	Reviews
ne	name, actors, director, genre	place, movie, start	cinema, title, startTime	location, movie, startingTime	title, date, grade, review
	S1		S3 Computing System	S4	S5

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Example: Reformulation on source databases: 5/5

- Then query execution
 - execute the physical query plan
 - May ask the optimizer to reconsider the plan (unlike RDBMS), e.g. if S3 is too slow
- sometimes contingencies are included in original plan
 - tradeoff between complexity of plan and ability to respond to unexpected events

	Movies	Cinemas	Cinemas in NYC	Cinemas in SF	Reviews	
me	name, actors, director, genre	place, movie, start	cinema, title, startTime	location, movie, startingTime	title, date, grade, review	
.1011	S1		S3	S4	S5	
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Schema Mapping should handle the discrepancies between source and the mediated schema: 1/4

- Relation and attribute names
 - "description" in the mediated schema (MS) the same as "review" in S5
 - "name" of Actors in MS and is S3 in NYCCinemas are not the same

Movie: Title, director, year, genre Actor: title, name	S1: Actor (AID, firstName, lastName, nationality, yearof Birth) Movie (MID, title), AcrtorPlays(AID, MID) MovieDetail (MID, directorm genre, year)		
Plays: movie, location, startTime	S2:	S5:	
Reviews: title, rating, description	Cinemas(place, movie, start)	MovieGenres(title, genre)	
Mediated schema	S3:	:	
S7:	NYCCinemas(name, title, startT	MovieDirectors(title, dir)	
MovieYears(title, year)	S4: Reviews (title, date, grade, revi		
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Schema Mapping should handle the discrepancies between source and the mediated schema: 2/4

- Tabular organization
 - In MS, Actor stores movie title and actor name
 - In S1, a join is needed that has to be specified by the mapping

Movie: Title, director, year, genre Actor: title, name	S1: Actor (AID, firstName, lastName, nationality, yearof Birth) Movie (MID, title), AcrtorPlays(AID, MID) MovieDetail (MID, directorm genre, year)	
Plays: movie, location, startTime Reviews: title, rating, description	S2:	S5:
	Cinemas (place, movie, start)	MovieGenres(title, genre)
Mediated schema	NYCCinemas(name, title, startTi	ime) S6:
S7: MovieYears(title, year)	S4:	MovieDirectors(title, dir)
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Schema Mapping should handle the discrepancies between source and the mediated schema: 3/4

- Domain coverage
 - the coverage and level of detail may differ
 - S1 stores more info about actors than in MS

S1: Actor (AID, firstName, lastName, nationality, yearof Birth) Movie (MID, title), AcrtorPlays(AID, MID) MovieDetail (MID, directorm genre, year)		
S2: Cinemas (place movie start)	S5:	
S3:	MovieGenres(title, genre)	
NYCCinemas(name, title, start]	- <u>ime)</u> S6: MovieDirectors(title, dir)	
	ew)	
	Actor (AID, firstName, lastNa Movie (MID, title), Acrt MovieDetail (MID, directorm S2: Cinemas(place, movie, start) S3: NYCCinemas(name, title, start]	

Schema Mapping should handle the discrepancies between source and the mediated schema: 3/4

- Data level variations
 - GPA as a letter grade vs. a numeric score of 4.0 scale
 - S1 stores actor names in two columns, MS stores in one

Movie: Title, director, year, genre Actor: title, name	S1: Actor (AID, firstName, lastNam Movie (MID, title), Acrto MovieDetail (MID, directorm)	orPlays(AID, MID)
Plays: movie, location, startTime	S2:	S5:
Reviews: title, rating, description	Cinemas(place, movie, start)	MovieGenres(title, genre)
Mediated schema	S3:	
S7:	NYCCinemas(name, title, startTi	ime) S6: MovieDirectors(title, dir)
MovieYears(title, year)	S4: Pavian e/title data grada ravis	
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Three Desired Properties of Schema Mapping Languages 1/3

- Flexibility
 - significant differences between disparate schemas
 - the languages should be very flexible
 - should be able to express a wide variety of relationships between schemas

Three Desired Properties of Schema Mapping Languages 2/3

- Efficient reformulation
 - our goal is to use the schema mapping to reformulate queries
 - we should be able to develop reformulation algorithms whose properties are well understood and are efficient in practice
 - often competes with flexibility, because more expressive languages are typically harder to reason about

Three Desired Properties of Schema Mapping Languages 3/3

• Easy update

- for a formalism to be useful in practice, it needs to be easy to add and remove sources
- If adding a new data source potentially requires inspecting all other sources, the resulting system will be hard to manage for a large number of sources

Three standard schema mapping languages

- 1. Global-as-View (GAV)
- 2. Local-as-View (LAV)
- 3. Global-Local-as-View (GLAV)

Global-as-View (GAV)

- GAV defines the mediated schema (MS) as a set of views over the data sources
 - Mediated Schema = Global schema
 - An intuitive approach
- Mediated schema (MS) G
 - $-G_i$ = some relation in G
 - X_i denotes attributes in G_i
- Source schema S₁, S₂, ..., S_n

GAV Definition

• A GAV schema mapping M is a set of expressions of the form:

- $G_i(X_i) \supseteq Q(S_1, S_2, ..., S_n)$
 - open world assumption
 - instances computed for MS are assumed to be incomplete

- or, $G_i(X_i) = Q(S_1, S_2, ..., S_n)$
 - closed world assumption
 - instances computed for MS are assumed to be complete

GAV Example

- Movie(title, director, year, genre) ⊇ S1.Movie(MID, title), S1.MovieDetail(MID, director, genre, year)
 Movie(title, director, year, genre) ⊇ S5.MovieGenres(title, genre), S6.MovieDirectors(title, director), S7.MovieYears(title, year)
 Plays(movie, location, startTime) ⊇ S2.Cinemas(location, movie, startTime)
 Plays(movie, location, startTime) ⊇ S3.NYCCinemas(location, movie, startTime)
- S1: **Actor** (AID, firstName, lastName, nationality, yearof Birth) Movie: Title, director, year, AcrtorPlays(AID, MID) **Movie** (MID, title), genre **MovieDetail** (MID, directorm genre, year) Actor: title, name **Plays:** movie, location, startTime S2: S5: **Reviews:** title, rating, description Cinemas(place, movie, start) **MovieGenres**(title, genre) S3: Mediated schema S6: NYCCinemas(name, title, startTime) MovieDirectors(title, dir) S7: S4: MovieYears(title, year) **Reviews**(title, date, grade, review) Duke CS, Fall 2016 CompSci 516: Data Intensive Computing Systems 36

Discussions: GAV 1/2

- Suppose we have a data source S8 that stored pairs of (actor, director) who worked together on movies.
- The only way to model this source in GAV is with the following two descriptions that use NULL:
 - Actors(NULL, actor) \supseteq S8(actor, director)
 - Movie(NULL, director, NULL, NULL) \supseteq S8(actor, director)
- These descriptions create tuples in the mediated schema that include NULLs in all columns except one

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S8: ActTogether(actor, director)

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genre Actor: title, name Plays: movie, location, startTime Reviews: title, rating, description

Movie: Title, director, year,

Discussions: GAV 2/2

- If the source S8 includes the tuples (Keaton, Allen) and (Pacino, Coppolag), then the tuples computed for the mediated schema would be:
 - Actors(NULL, Keaton), Actors(NULL, Pacino)
 - Movie(NULL, Allen, NULL, NULL), Movie(NULL, Coppola, NULL, NULL)
- Now suppose we have the following query that recreates S8:
 - Q(actor, director) :- Actors(title, actor), Movie(title, director, genre, year)
- We would not be able to retrieve the tuples from S8 because the source descriptions lost the relationship between actor and director.

S8: ActTogether(actor, director)

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Movie: Title, director, year, genre Actor: title, name Plays: movie, location, startTime Reviews: title, rating, description

Local As View (LAV)

- describes each data source as precisely as possible and independently of any other sources
 - opposite approach to GAV
- Mediated schema (MS) G
- Source schema S₁, S₂, ..., S_n
 X_i denotes attributes in S_i

LAV Definition

• A LAV schema mapping M is a set of expressions of the form:

- S_i(X_i) ⊆ Q_i(G)
 − open world assumption
- or, $S_i(X_i) = Q_i(G)$
 - closed world assumption
 - but completeness about data sources, not about the MS

LAV Example

- S5.MovieGenres(title, genre)
 Genres(title, director, year, genre)
- S6.MovieDirectors(title, director) \subseteq Movie(title, director, year, genre)
- S7.MovieYears(title, year)
 Govie(title, director, year, genre)
- Can also specify constraints on the contents
- S9(title, year, "comedy") ⊆ Movie(title, director, year, "comedy"), year ≥ 1970

Movie: Title, director, year, genre Actor: title, name	ctor: title, name Movie (MID, title), AcrtorPlays(AID, MovieDetail (MID, directorm genre, year	
Plays: movie, location, startTime	S2:	S5:
Reviews: title, rating, description	Cinemas(place, movie, start)	MovieGenres(title, genre)
Mediated schema	S3:	
	NYCCinemas(name, title, startTir	me) S6:
S7:	S4:	MovieDirectors(title, dir)
MovieYears(title, year) Duke CS, Fall 2016	Reviews(title, date, grade, review CompSci 516: Data Intensive Computing S	

Global-Local-As-View (GLAV)

- GAV and LAV can be combined into GLAV
- Has the expressive power of both
- The expressions in the schema mapping include
 - a query over the data sources on the left hand side
 - a query on the mediated schema on the right-hand side
- Mediated schema (MS) G
- Source schema S₁, S₂, ..., S_n

GLAV Definition

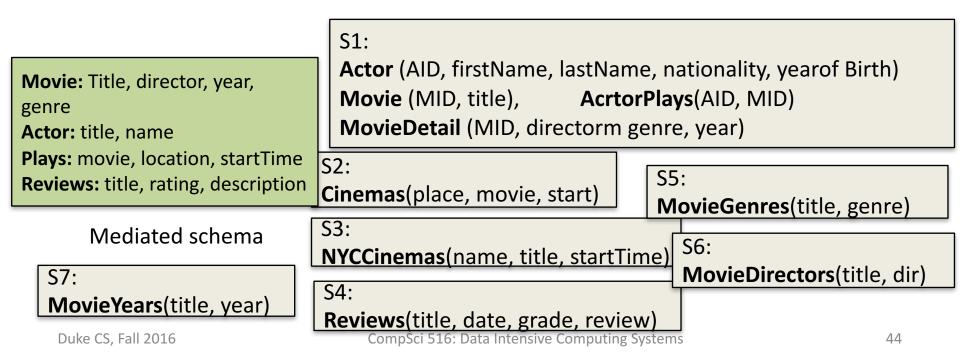
- A GLAV schema mapping M is a set of expressions of the form:
- $Q^{S}(X) \subseteq Q^{G}(X)$

open world assumption

- or, $Q^{S}(X) = Q^{G}(X)$
 - closed world assumption
- Q^G is a query over G whose head variables are X
- Q^S is a query over data sources S₁, S₂, ..., S_n where the head variables are also S

GLAV Example

- Suppose S1 is known to have comedies produced after 1970 only
- S1.Movie(MID, title), S1.MovieDetail(MID, director, genre, year) ⊆ Movie(title, director, "comedy", year), year ≥ 1970



Optional/Additional Slides

Schema Matchings and Mappings

- Specify "matches", e.g.
 - attribute "name" in one source corresponds to attribute "title" in another
 - "location" is a concatenation of "city, state, zipcode"
- Elaborate matches into semantic "mappings"

 using queries like SQL

Challenges

- The tasks of creating the matches and mappings are often difficult
 - they require a deep understanding of the semantics of the schemas of the data sources and of the mediated schema
 - This knowledge is typically distributed among multiple people
 - these people are not necessarily database experts and may need help
- There is no algorithm that will take two arbitrary database schemas and flawlessly produce correct matches and mappings
 - goal is to create tools that educe the time by giving suggestions to the designer

optional slide Two database schemas

DVD-VENDOR

Movies(id, title, year) Products(mid, releaseDate, releaseCompany, basePrice, rating, saleLocID) Locations(lid, name, taxRate)

AGGREGATOR

Items(name, releaseInfo, classification, price)

- Attributes and tables in a schema are called its elements
- The aggregator is not interested in all the details of the product, but only in the attributes that are shown to its customers
- Schema DVD-VENDOR has 14 elements
 - 11 attributes (e.g., id, title, and year) and three tables (e.g., Movies)
- Schema AGGREGATOR has five elements
 - four attributes and one table.

Semantic mapping

DVD-VENDOR

Movies(id, title, year) Products(mid, releaseDate, releaseCompany, basePrice, rating, saleLocID) Locations(lid, name, taxRate)

AGGREGATOR

Items(name, releaseInfo, classification, price)

 A query expression that relates a schema S with a schema T

– recall GAV, LAV, GLAV

Semantic mapping - Example 1

DVD-VENDOR

Movies(id, title, year) Products(mid, releaseDate, releaseCompany, basePrice, rating, saleLocID) Locations(lid, name, taxRate)

AGGREGATOR

Items(name, releaseInfo, classification, price)

- "the title of Movies in the DVD-VENDOR schema is the name attribute in Items in the AGGREGATOR schema."
- SELECT name as title
- FROM Items

DVD-VENDOR FROM AGGREGATOR

Semantic mapping - Example 2

DVD-VENDOR

Movies(id, title, year) Products(mid, releaseDate, releaseCompany, basePrice, rating, saleLocID) Locations(lid, name, taxRate)

AGGREGATOR

Items(name, releaseInfo, classification, price)

- "get the price attribute of the Items relation in the AGGREGATOR schema by joining the Products and Locations tables in the DVD-VENDOR schema.."
- SELECT (basePrice * (1 + taxRate)) AS price
- FROM Products, Locations
- WHERE Products . saleLocID = Locations .lid

AGGREGATOR FROM DVD-VENDOR

Semantic mapping - Example 3

DVD-VENDOR

Movies(id, title, year) Products(mid, releaseDate, releaseCompany, basePrice, rating, saleLocID) Locations(lid, name, taxRate)

AGGREGATOR

Items(name, releaseInfo, classification, price)

- "Get the entire tuple in Items table from DVD-VENDOR"
- SELECT title AS name , releaseDate AS releaseInfo , rating AS
- classification , basePrice * (1 + taxRate) AS price
- FROM Movies, Products, Locations
- WHERE Movies .id = Products .mid AND Products . saleLocID =
- Locations .lid

AGGREGATOR FROM DVD-VENDOR

Semantic Matches

- Relates a set of elements in schema S to a set of elements in schema T
 - without specifying the details of the nature of relationship (as SQL queries)

Why are matching and optional slide mapping difficult?

- The semantics may not be fully captured in the schemas
 - "rating" may imply movie rating, customer rating, etc
 - sometimes accompanied by English text, hard for systems to parse and understand
- Schema clues can be unreliable
 - two elements may have the same name but different meaning, like "name" or "title"
- Semantics can be subjective
 - what "plot-summary" means
 - sometimes a committee of experts vote
- Combining data may be difficult
 - need to figure out a join path
 - full/left/right outer join or inner join
 - may need filter conditions
 - the designer has figure these out inspecting a large amount of data
 - erroneous and labor prone

Components in a schema optional slide matching system

- Matchers
- Combiners
- Constraint Enforcers
- Match Selectors

1. Matchers

- schemas \rightarrow similarity matrix
- takes two schemas S and T as input
- outputs a similarity matrix
- assigns to each element pair s of S and t of T a number between 0 and 1

– higher the number, s and t are more similar

• e.g.

- name ≈ <name: 1, title : 0:5>
- releaseInfo ≈ <releaseDate : 0:6, releaseCompany: 0.4>
- price \approx <basePrice : 0:5>

Types of Matchers

- Name-based matchers
 - compares the names of elements
 - but almost never written the same way
 - uses techniques for string matching as edit distance, Jaccard measure etc.; synonyms; normalization (capital letters); hyphens; etc
- Instance-based matchers
 - Look at data instances, builds recognizers (dictionaries), computes overlaps, classification

2. Combiners

• matrix $\times \dots \times$ matrix \rightarrow matrix

 merges the similarity matrices output by the matchers into a single one

 can take the average, minimum, maximum, or a weighted sum of the similarity scores

Types of Combiners

• Average combiners:

- Suppose k matchers to predict the scores between the element s_i of schema S and the element t_i of schema T
- then an average combiner will compute the score between these two elements as the average from these k matchers
- Hand-crafted scripts
 - e.g. if s_i = address, return score of naïve-bayes classifer, else average
- Weighted combiner
 - gives weights to each matcher

3. Constraint Enforcers

- matrix \times constraints \rightarrow matrix
- In addition to clues and heuristics, domain knowledge plays an important role in pruning candidate matches
 - e.g. knowing that many movie titles contain four words or more, but most location names do not, can help us guess that Items.name is more likely to match Movies.title than Locations.name
- an enforcer enforces such constraints on the candidate matches
 - it transforms the similarity matrix produced by the combiner into another one that better reflects the true similarities

optional slide Types of Constraint Enforcers : 1/2

- There may be hard or soft domain integrity constraints
- Hard constraints must be applied
 - The enforcer will not output any match combination that violates them
- Soft constraints are of more heuristic nature, and may actually be violated in correct match combinations
 - the enforcer will try to minimize the number (and weight) of the soft constraints being violated
 - but may still output a match combination that violates one or more of them
- Formally, we attach a cost to each constraint
 - For hard constraints, the cost is 1
 - for soft constraints, the cost can be any positive number.

Types of Constraint Enforcers : 2/2

• e.g.

- c1: If A Items.code, then A is a key (weight = infinity)
- c2 If A Items.desc, then any random sample of 100 data instances of A must have an average length of at least 20 words (weight = 1.5)
- c3: If more than half of the attributes of Table U matches those of Table V , then U is similar to V (weight = 1)

4. Match Selectors

- matrix \rightarrow matches
- Produces matches from the similarity matrix output by the constraint enforcer
- name ≈ <title : 0:5>
- releaseInfo ≈ <releaseDate : 0:6>
- classication ≈ <rating : 0:3>
- price ≈ <basePrice : 0:5>
- Given the threshold 0.5, the match selector produces matches:
 - name \approx title, releaseInfo \approx releaseDate, and price \approx basePrice

Types of Match Selectors

- The simplest selection strategy is thresholding
 - all pairs of schema elements with similarity score exceeding a given threshold are returned as matches
- More complex strategies include formulating the selection as an optimization problem over a weighted bipartite graph