Information Integration project

Integrating heterogeneous data sources into a unified global schema

Goal of the project

Accomplishing a domain-based integration of multiple and heterogenous data sources, merging them into a single global view that can be queried by the users.

The global schema should be able to satisfy queries not trivially satisfiable by accessing directly the data sources.





GAV – Mappings (1)

In GAV approach, the (sound) mappings are assertions as such:

 $\forall x. \ \varphi S(x) \rightarrow g(x)$

Where:

- > S is the source schema
- φS is a CQ over the source DB
- g is a predicate of arity n that indicates an element of G



GAV (Global As View): the global schema is a view of the set of sources



GAV – Retrieved global database (2)

Given $I = \langle G, S, M \rangle$ and source database C for S, we denote the retrieved global database as M(C), that is the database obtained by:

- Applying the CQs specified in the mappings
- Transferring the tuples from the sources to the global DB









Data integration using Pentaho (1)

The end result of the project is a **one-shot materialization** of the global schema, achieved by mappings in the form of Pentaho transformations.





Data integration using Pentaho (2)

Pentaho Kettle is a tool that allows loading input data in **heterogeneous** formats and from different sources, performing the transformations needed in order to map the information as requested.



Finally, the data is materialized in output as a SQL database in the example shown for this project.



Domain and data sources (1)







Actors



Movies



Users Ratings



Production studios



Oscar awards



Domain and data sources (2)

5 Data sources













> 11 Files (TSV and CSV)

+45.000 Correctly reconciled movies across datasets



Domain and data sources (3)





- title.principals.tsv
- title.basics.tsv
- name.basics.tsv





the_oscar_award.csv





- ratings.csv
- movies_metadata.csv
- links.csv



Domain and data sources (4)





- studios.csv
- movies.csv





- rotten_tomatoes_critic_reviews.csv
- rotten_tomatoes_movies.csv



Domain and data sources – Reconciliation issues (5)

Physical heterogeneity

Data stored in different formats: CSV and TSV Easily mitigated by Pentaho transformations that can handle multiple input formats

Conceptual heterogeneity

- Different datasets uniquely identify movies in diverse formats: IDs both in string and number data types
- Some datasets refer to movies with their original title, some with their English title, some have both fields



Domain and data sources – Reconciliation issues (6)

Contextual heterogeneity

Datasets can refer to the same field using different representation: both IMDB and Movielens sources contain the IMDB ID, but the latter lacks a prefix in the string.

```
import pandas as pd
import os

input_file = os.path.abspath("../datasets/links.csv")
output_file = os.path.abspath("../datasets/links_upd.csv")
column_name = "imdbId"

try:
    # Load the CSV file into a DataFrame
    df = pd.read_csv(input_file)

# Ensure the specified column exists
    if column_name not in df.columns:
        raise ValueError(f"Column '{column_name}' not found in the CSV file.")

# Update the column by prepending 'tt' to each value
    df[column_name] = 'tt' + df[column_name].astype(str)

# Save the updated DataFrame to a new CSV file
    df.to_csv(output_file, index=False)
    print(f"Updated CSV file saved to {output_file}")

except Exception as e:
    print(f"An error occurred: {e}")
```

Python preprocessing required to solve the heterogeneity between sources



Source schema (1)

 $IMDBTitleBasics_{/10}$ (tconst, titleType, primaryTitle, originalTitle, isAdult, startYear, endYear, runtimeMinutes, genres)

IMDBTitlePrincipals₆ (tconst, ordering, nconst, category, job, characters)

IMDBNameBasics₆ (nconst, primaryName, birthYear, deathYear, primaryProfession, knownForTitles)

MovielensRatings_{/4} (userId, movieId, rating, timestamp)

MovielensMoviesMetadata/24 (adult, belongsToCollection, budget, genres, homepage, id, imdb_id, original_language, originalTitle, overview, popularity, poster_path, production_companies, production_countries, release_date, revenue, runtime, spoken_languages, status, tagline, title, video, vote_average, vote_count)

MovielensLinks/3 (movield, imdbld, tmdbld)



Source schema (2)

TheOscarAward/7 (year_film, year_ceremony, ceremony, category, name, film, winner)

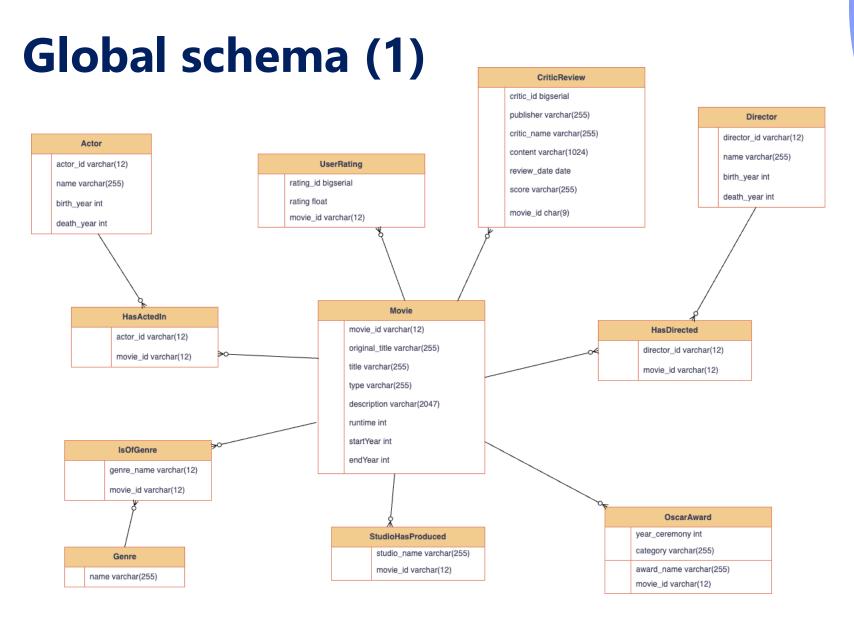
LetterboxStudios_{/2} (id, studio)

LetterboxMovies₇ (id, name, date, tagline, description, minute, rating)

RottenTomatoesCriticReviews_{/8} (rotten_tomatoes_link, critic_name, top_critic, publisher_name, review_type, review_score, review_date, review_content)

RottenTomatoesMovies_{/22} (rotten_tomatoes_link, movie_title, movie_info, critics_consensus, content_rating, genres, directors, authors, actors, original_release_date, streaming_release_date, runtime, production_company, tomatometer_status, tomatometer_rating, tomatometer_count, audience_status, audience_rating, audience_count, tomatometer_top_critics_count, tomatometer_fresh_critics_count, tomatometer_rotten_critics_count)







Global schema (2)

Movie_{/8} (movie_id, original_title, title, type, description, runtime, startYear, endYear)

UserRating_{/3} (rating_id, rating, movie_id)

CriticReview_{/7} (critic_id, publisher, critic_name, content, review_date, score, movie_id)

Actor_{/4} (actor_id, name, birth_year, death_year)

HasActedIn_{/2} (actor_id, movie_id)

StudioHasProduced_{/2} (studio_name, movie_id)



Global schema (3)

Director_{/4} (director_id, name, birth_year, death_year)

HasDirected_{/2} (director_id, movie_id)

Genre_{/1} (name)

IsOfGenre_{/2} (genre_name, movie_id)

OscarAward_{/4} (year_ceremony, category, award_name, movie_id)



Conjunctive GAV FOL Mappings (1)

Movie:

 \forall id, ot, ti, ty, dc, rt, sy, ey.(\exists ia, g, ad, btc, bg, hp, mid, ol, ott, pop, pp, pcp, pcn, rd, rev, rut, sl, st, tl, tit, va,vc.(IMDBTitleBasics (id, ty, ti, ot, ia, sy, ey, rt, g) \land MovielensMovieMetadata(ad, btc, bg, hp, mid, id, ol, ott, dc, pop, pp, pcp, pcn, rd, rev, rut, sl, st, tl, tit, va, vc))) \rightarrow Movie(id, ot, ti, ty, dc, rt, sy, ey)

UserRating:

 \forall rid, r, mid.(\exists uid, mid2, ts, tid.(MovielensRatings (uid, mid2, r, ts) \land MovielensLinks(mid2, mid, tid))) \rightarrow UserRating(rid, r, mid)

Genre:

 \forall g.(\exists id, ty, ti, ot, ia, sy, ey, rt.(IMDBTitleBasics (id, ty, ti, ot, ia, sy, ey, rt, g))) \rightarrow Genre(g)

IsOfGenre:

 \forall g, id.(\exists id, ty, ti, ot, ia, sy, ey, rt.(IMDBTitleBasics (id, ty, ti, ot, ia, sy, ey, rt, g))) \rightarrow IsOfGenre(g, id)



Conjunctive GAV FOL Mappings (2)

Actor:

```
∀aid, n, by, dy.(∃ pp, kft.(IMDBNameBasics(aid, n, by, dy, pp, kft)))

→ Actor(aid, n, by, dy)
```

HasActedIn:

```
\forallaid, mid.(\exists o, ca, j, ch.(IMDBTitlePrincipals(mid, o, aid, ca, j, ch) \land ca = 'actor')) \rightarrow HasActedIn(aid, mid)
```

Director:

```
\foralldid, n, by, dy.(\exists pp, kft.(IMDBNameBasics(did, n, by, dy, pp, kft))) \rightarrow Actor(did, n, by, dy)
```

HasDirected:

```
\foralldid, mid.(\exists o, ca, j, ch.(IMDBTitlePrincipals(did, o, aid, ca, j, ch) \land ca = 'director')) \rightarrow HasDirected(did, mid)
```

StudioHasProduced:

```
\forallsn, mid.(\exists mid2, n, d, tl, dc, m, ty, ot, ia, ey, rt, g, r.(LetterboxStudios(mid2, sn) \land LetterboxMovies(mid2, n, d, tl, dc, m, r) \land IMDBTitleBasics (mid, ty, n, ot, ia, d, ey, rt, g))) \rightarrow StudioHasProduced(sn, mid)
```



Conjunctive GAV FOL Mappings (3)

CriticReview:

Vcid, p, cn, ct, rd, sc, mid.
(∃ rtl, tc, rt, mt, mi, cc, cr, g, d, a, ord, srd, r, pc, ts, tr, tc, as, ar, ac, ttcc, tfcc, trcc, ty, ot, ia, ey, rt2, g2. (RottenTomatoesCriticReviews (rtl, cn, tc, p, rt, sc, rd, ct) Λ
RottenTomatoesMovies(rtl, mt, mi, cc, cr, g, d, a, ord, srd, r, pc, ts, tr, tc, as, ar, ac, ttcc, tfcc, trcc) Λ IMDBTitleBasics (mid, ty, mt, ot, ia, d, ey, rt2, g2))) → CriticReview(cid, p, cn, ct, rd,

OscarAward:

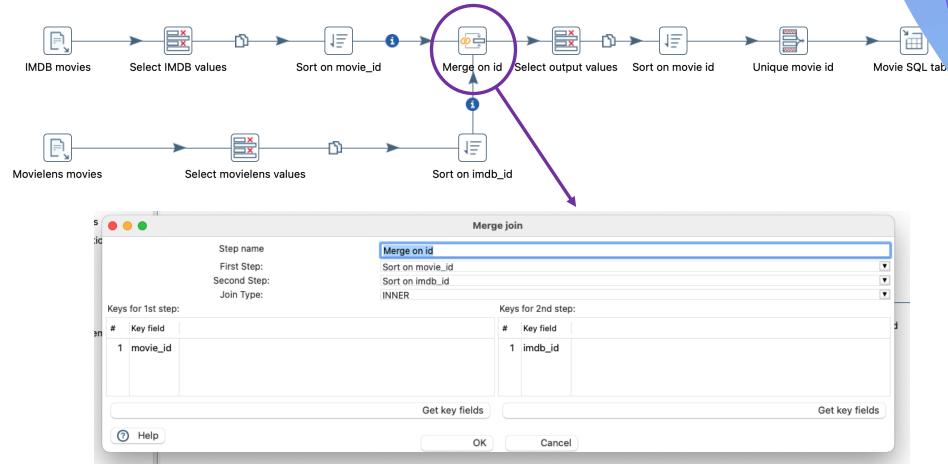
sc, mid)

 \forall yc, ca, an, mid. (\exists yf, ce, n, f, w, ty, ot, ia, ey, rt, g. TheOscarAward(yf, yc, ce, ca, n, f, w) \land IMDBTitleBasics (mid, ty, f, ot, ia, yf, ey, rt, g) \land w='true') \rightarrow OscarAward(yc, ca, an, mid)



Pentaho implementation (1)

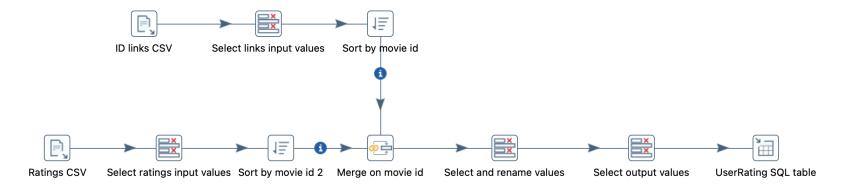
Movie transformation



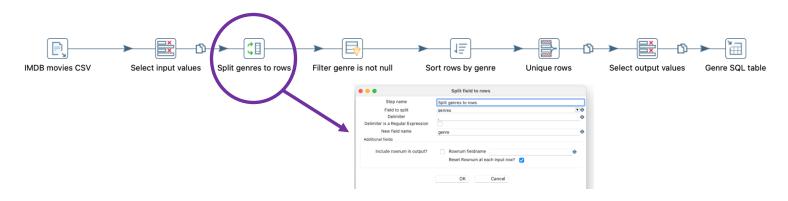


Pentaho implementation (2)

UserRating transformation



Genre transformation

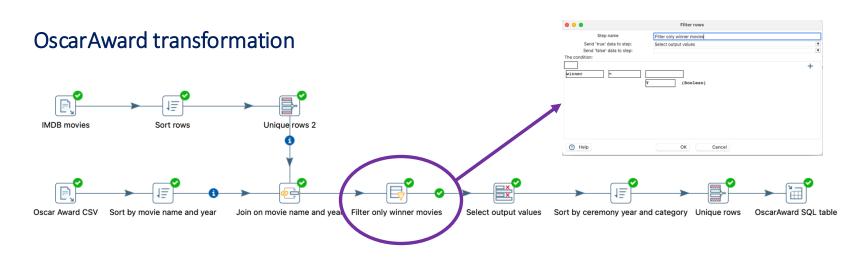




Pentaho implementation (3)

IsOfGenre transformation

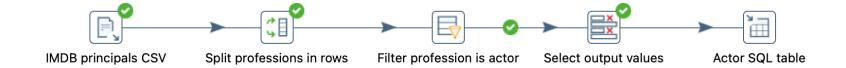






Pentaho implementation (4)

Actor transformation



HasActedIn transformation





Pentaho implementation (5)

Director transformation

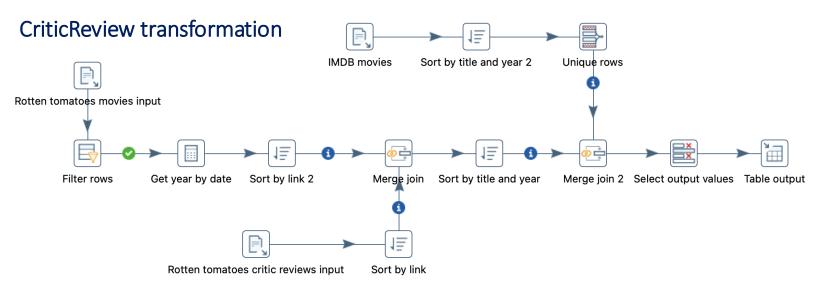


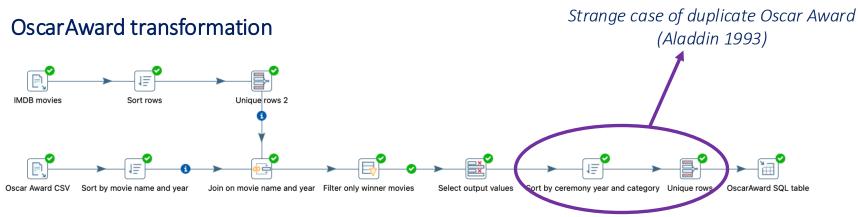
HasDirected transformation





Pentaho implementation (6)







Some statistics...

- ➤ In the RGBD (Retrieved Global Database)
 - >11.000.000 movies
 - >26.000.000 reconciled user ratings
 - >20.000.000 actor appearances
- ➤ 17 minutes each for CriticRating, OscarAward, StudioHasProduced transformations (ran on M1 Pro)
- > 7,09 GBs of source data



Query answering

Given the source database C, now that we have obtained the **RGBD** (Retrieved Global Database) M(C) by performing a one-shot materialization, we can query it

We have proved that, given GAV mappings with sound sources and conjunctive queries, we can compute **certain answers** in polynomial time complexity



Name of actors born after 1960 participating in a "Best Picture" Oscar-winning movie, criticized by the Independent (UK) publisher

```
select distinct a."name"
from actor a, hasactedin h, movie m, oscaraward o, criticreview c
where a.actor_id = h.actor_id
    and h.movie_id = m.movie_id
    and m.movie_id = o.movie_id
    and m.movie_id = c.movie_id
    and birth_year > 1960
    and o.category = 'BEST PICTURE'
    and c.publisher = 'Independent (UK)'
```

```
Az name

1 Azharuddin Mohammed Ismail
2 Danny Dorland
3 Dev Patel
4 Dominic West
5 Irrfan Khan
6 Javier Bardem
7 John C. Reilly
8 Joseph Fiennes
9 Josh Brolin
10 Mark Ivanir
11 Martin Clunes
```

```
q = \{(name)\}
```

∃ aid, mid, name, by, dy, ot, ti, ty, dc, r, sy, ey, yc, cat, aw, cid, pub, cn, co, rd,sc. Actor(aid, name, by, dy) ∧ HasActedIn(aid, mid)

 \bigwedge Movie(mid, ot, ti, ty, dc, r, sy, ey) \bigwedge OscarAward(yc, 'BEST PICTURE', aw, mid)

 Λ CriticReview(cid, 'Independent (UK)', cn, co, rd, sc, mid) Λ by > 1960}



Name of movies of genre Drama produced by "Paramount Pictures" where Cillian Murphy acted

```
select m.title
from genre g, isofgenre i, movie m, hasactedin h, studiohasproduced s, actor a
where g."name" = i.genre_name
    and i.movie_id = m.movie_id
    and m.movie_id = s.movie_id
    and i.movie_id = h.movie_id
    and h.actor_id = a.actor_id
    and s.studio_name = 'Paramount'
    and a."name" = 'Cillian Murphy'
    and g."name" = 'Drama'
```



```
q = {(title) | \exists aid, mid, by, dy, ot, ty, dc, r, sy, ey, . Actor(aid, 'Cillian Murphy', by, dy) \land HasActedIn(aid, mid) \land Movie(mid, ot, title, ty, dc, r, sy, ey) \land IsOfGenre('Drama', mid) \land StudioHasProduced('Paramount', 'Cillian Murphy')}
```



Name of critics that have criticized a movie directed by Cristopher Nolan where Al Pacino acted

query elapsed time to join actors, movies and directors: ~5 minutes

```
select c.critic_name
from actor a, hasactedin h, movie m, hasdirected h2, criticreview c, director d
where a.actor_id = h.actor_id
    and h.movie_id = m.movie_id
    and m.movie_id = h2.movie_id
    and h2.director_id = d.director_id
    and m.movie_id = c.movie_id
    and d.name = 'Christopher Nolan'
    and a."name" = 'Al Pacino'
```

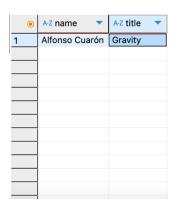
```
A-Z critic_name
Rob Blackwelder
Harvey S. Karten
Richard Roeper
Robin Clifford
Jeanne Aufmuth
Ed Gonzalez
Kevin N. Laforest
Frank Swietek
Andy Weil
Steve Rhodes
Kirk Honeycutt
```

```
q = {(name) |
∃ aid, mid, by, dy, ot, ti, ty, dc, r, sy, ey, did, by2, dy2, cid, pub, co, rd, sc.
Actor(aid, 'Al Pacino', by, dy) ∧ HasActedIn(aid, mid)
∧ Movie(mid, ot, ti, ty, dc, r, sy, ey) ∧ HasDirected(did, mid)
∧ Director(did, 'Cristopher Nolan', by2, dy2)
∧ CriticReview(cid, pub, name, co, rd, sc, mid)}
```



Name of directors and movie titles of oscar-winning movies in the DIRECTING category, produced by Warner Bros, with at least one user rating less than 0.5

```
select distinct d."name", m.title
from movie m, oscaraward o, studiohasproduced s, userrating u, hasdirected h, director d
where m.movie_id = o.movie_id
    and m.movie_id = u.movie_id
    and s.movie_id = m.movie_id
    and h.movie_id = m.movie_id
    and h.movie_id = d.director_id
    and u.rating <= 0.5
    and s.studio_name = 'Warner Bros. Pictures'
    and o.category = 'DIRECTING'</pre>
```

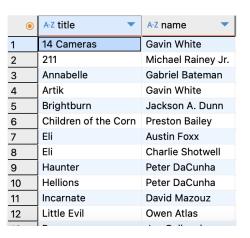


```
q = {(name, title) |
∃ did, by, dy, mid, ot, ty, dc, r, sy, ey, yc, aw, rid, rating.
Director(did, name, by, dy) Λ HasDirected(did, mid)
Λ Movie(mid, ot, title, ty, dc, r, sy, ey)
Λ StudioHasProduced('Warner Bros. Pictures', mid)
Λ OscarAward(yc, 'DIRECTING', aw, mid)
Λ UserRating(rid, rating, mid) Λ rating <= 0,5}
```



Title of horror movies that last less than 100 minutes, and name of actors participating, born after the year 2000, with a critic review on rotten tomato

```
select distinct m.title, a."name"
from movie m, genre q, isofgenre i, criticreview c, hasactedin h, actor a
where m movie id = i movie id
     and g."name" = i.genre_name
     and h.movie_id = m.movie_id
     and h.actor_id = a.actor_id
     and c.movie_id = m.movie_id
     and a.birth_year >= 2000
     and q."name" = 'Horror'
     and m.runtime < 100
 q = {(title, name)
 ∃ aid, by, dy, mid, ot, ty, dc, rt, sy, ey, cid, pub, cname, co, rd, sc.
 Actor(aid, name, by, dy) \Lambda HasActedIn(aid, mid)
 \Lambda Movie(mid, ot, title, ty, dc, rt, sy, ey)
 ↑ IsOfGenre('Horror', mid)
 ↑ CriticReview(cid, pub, cname, co, rd, sc, mid)
  \Lambda by >= 2000 \Lambda rt < 100}
```





...Thanks!

References

- > Theoretical notions: LSDM course taught by prof. Antonella Poggi
- ➤ Github: <u>Github repository</u>
- ➤ Datasets: see README file in repository
- Guide link to install Pentaho Kettle on Apple Silicon

