Information Retrieval WS 2013 / 2014

Lecture 12, Tuesday January 28th, 2014 (Ontologies, SPARQL, Relation to SQL)

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Overview of this lecture

Organizational

- Your results + experiences with Ex. Sheet 11 (NB vs. SVM)

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Ontologies

– Ontologies = fact databases ... ask questions like:

Actors that are married and starred in the same movie

- The SPAROL ontology query language
- Translate SPARQL queries to SQL queries on a database
- Performance issues (join order)
- Exercise Sheet 12: Implement SPARQL → SQL translation and use to process SPARQL queries with the SQLite database

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- Summary / excerpts last checked January 28, 14:00
 - Exercise helped to understand concepts more deeply
 - The theoretical task wasn't as hard as it first looked
 - Ambiguous notation: the w₁, ..., w_m denoted both the words from the vocabulary and the m components of the w vector

Sorry, will be fixed next time I give the course

- Small mistake on slide 12 of last lecture: $Pr(b|A) \rightarrow Pr(a|B)$
- Vector representation was different from that for ES#10
 But the same as in all other lectures before that, and anyway, the two representations are equivalent (slide 9 of last lecture)
- Thanks again for the great feedback from the tutor

Summary

Dataset 1: actors and politicians, 18,499 documents
 Accuracy of 98% for both SVM and NB

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- **Dataset 2:** singers and songwriters, 8,913 documents
- Accuracy of 92% for SVM and of 89% for NB
- On both datasets zero outliers for the SVM and a much larger margin than for NB (which does not care about margin size)
 However, this does not seem to matter much for accuracy
- Experience shows that NB typically estimates badly the Prob(C=c | doc), but nevertheless often classifies well

In practice, one is often not interested in accurate probabilities, but just that the correct class gets the highest probability

Ontology = a database of facts on entities

- With **unique** names / identifiers for each entity
- Facts are expressed as subject predicate object triples

Brad Pitt acted in Mr. and Mrs. Smith Brad Pitt acted in Burn After Reading Angelina Jolie acted in Mr. and Mrs. Smith Joel Cohen directed Burn After Reading Ethan Cohen directed Burn After Reading Brad Pitt married to Angelina Jolie

Understand: we can always decompose more complex facts into triples, so triples is all we need

- Relation to the "Semantic Web"
 - The classical web contains a lot of facts hidden in text
 For example: infos about an actor or a movie on IMDB

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- The Semantic Web (SW) initiative is concerned with making ontology data **explicitly** available on the web
- The challenges of SW are really about standardization:
 - Unique identifiers ... use URIs + namespaces
 - Diff. identifiers meaning the same thing ... use owl:sameAs
 - Well-defined syntax ... RDF has become common
- This is **not** the topic of this lecture / course

- Example 1: the GeoNames ontology
 - Very complete database of geographical features:

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- Cities, countries, rivers, mountains, roads, ...
- Around 10M entities, 250MB compressed
- Download from http://www.geonames.org
- RDF endpoint: <u>http://www.geonames.org/ontology</u>

Great dataset, but for this lecture we want something more general-purpose ...

- Example 2: the YAGO ontology (Yet Another Great Ontology)
 - From Suchanek et al, WWW 2007 & J.Web.Sem 2008
 - General-purpose facts, extracted from Wikipedia + WordNet
 - Original dataset: about 120M facts on 10M entities
 - Of those, only about 10M are real "facts" that we as humans would find useful ... this is typical for ontologies
 - Download from <u>http://www.mpi-inf.mpg.de/yago</u>
 - Accuracy is good, but many popular facts are missing, e.g. only very few actors per movie are known

Nevertheless, small and simple and was hence quite popular with researchers (including us) for a while ...

- Example 3: the FreeBase Ontology
 - A general-purpose ontology, community-maintained
 - Developed by Metaweb, aquired by Google in 2010
 - Freely available: <u>https://developers.google.com/freebase/data</u>
 Currently 2500M facts on 50M entities, 25GB compressed
 Rather complex schema + some inconsistencies
 - Nicer version: <u>http://freebase-easy.cs.uni-freiburg.de</u>
 Around 250M facts on 50M entities, 2.5GB compressed

The currently most complete and most accurate generalpurpose ontology ... we extracted a nice subset for you !

- Structured queries on ontologies
 - Example query in natural language: actors who are married and starred together in at least one movie
 - Difference between ontology search and text search
 There is a well-defined result set ... no fuzzy "relevance"

pronounced Sport "

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- SPARQL = SPARQL Protocol And RDF Query Language
- The standard query language for ontology queries

SELECT ?person1 ?person2 ?movie WHERE {
 ?person1 acted_in ?movie .
 ?person2 acted_in ?movie .
 ?person1 married_to ?person2
}

acted

m

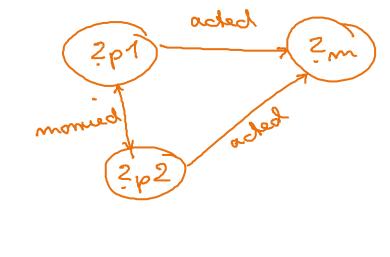
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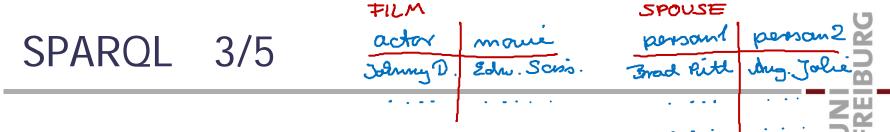
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- Viewing SPARQL queries as subgraphs
 - On can view a (triple) ontology as a graph, where the nodes are the entities, and the edges are the facts
 - A SPARQL query is then a sub-graph with variables at some or all of the nodes
 - The goal is to find all matches in the ontology graph





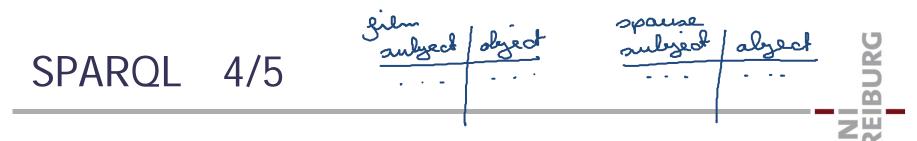
SPARQL looks very much like SQL

- Indeed, ontology data is naturally stored in databases
- The standard query language for **databases** is SQL
- Assume we have two tables film (with columns actor and movie) and spouse (with columns person1 and person2)

SELECT spouse.person1, spouse.person2

- FROM spouse, film as film1, film as film2
- WHERE spouse.person1 = film1.actor
- AND spouse.person2 = film2.actor
- AND film1.movie = film2.movie;

Let's play around a bit with **SQLite** ... see slides 15 - 17



SPARQL to SQL: generic translation

 In the following example, we use one table per relation, each with two columns, just named subject and object

For ES#11, use **one big table** for all the data, with three columns named **subject**, **predicate**, **object**

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SPARQL to SQL: implementation advice for ES#11

 If there are k query triples in the SPARQL query, have k entries in the FROM clause of the SQL query

FROM freebase as f1, freebase as f2, \dots , freebase as fk

 In your code, for each variable from the SPAROL query, build an array of all its occurrences in the query, e.g.

?x: f1.subject, f2.object, f5.object

 Then, when building the SQL query, add the corresponding equalities to the WHERE clause, e.g.

WHERE f1.subject = f2.object AND f2.object = f5.object

Note: if ?x occurs m times, m – 1 equalities are enough

- A full-fledged database, easy to install and use
 - Download from http://www.sqlite.org
 - On Debian/Ubuntu install with: sudo apt-get install sqlite3
 - Two types of commands ... examples on next slides
 SQL commands: must end with a semicolon
 SQLite commands: start with a dot, no semicolon at end
 - Two modes to start SQLite:
 - sqlite3will work on an in-memory databasesqlite3 <name>.dbcreate database in that file, and if file
exists, use database from that file

SQLite 2/3

- Some useful SQLite commands by example
 - Specifies the column separator used for input and output
 .separator " use Ctrl+V TAB for TAB !

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- Read table from TSV (tab-separated values) file
 .import film.tsv film
- Show execution time of every command

.timer on

- Output to file (use stdout for output to console again) .output <file name>
- Execute commands from script file (typical suffix is .sql) .read <file with commands>

SQLite 3/3

- Some useful SQL commands by example
 - Create a table with a given schema
 CREATE TABLE film(actor TEXT, movie TEXT);
 - Create an index for a column of a table
 CREATE INDEX file_index ON film(actor);
 - Extract / combine data from tables
 SELECT * FROM film WHERE ... LIMIT 100;
 - Delete table / index (without error msg if it's not there)
 DROP TABLE IF EXISTS film;

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DROP INDEX IF EXISTS film_index;

Cross product of tables

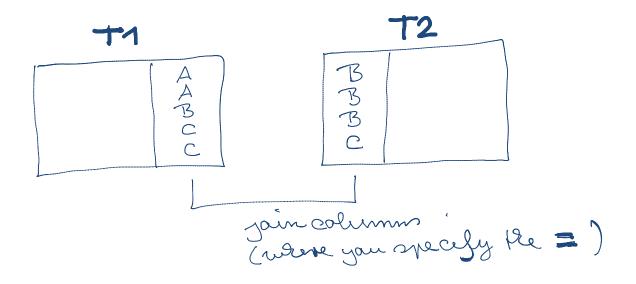
- Understand that, conceptually, an SQL statement like FROM $T_1, T_2, ..., T_k$ WHERE ... = ... AND ... = ... AND ... selects elements from the cross-product $T_1 \times \cdots \times T_k$ (which has $|T_1| \cdot \cdots \cdot |T_k|$ elements) (where some or all of the T_i can be the same table) UNI FREIBURG

Performance 2/4

Joining of tables

- The WHERE ... = ... effectively ask for a JOIN
- This JOIN effectively asks for a list intersection
- If we CREATE an index for the respective tables on the respective join attributes, this list intersection gets fast

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Join ordering

– Typical SQL-from-SPARQL queries require multiple joins

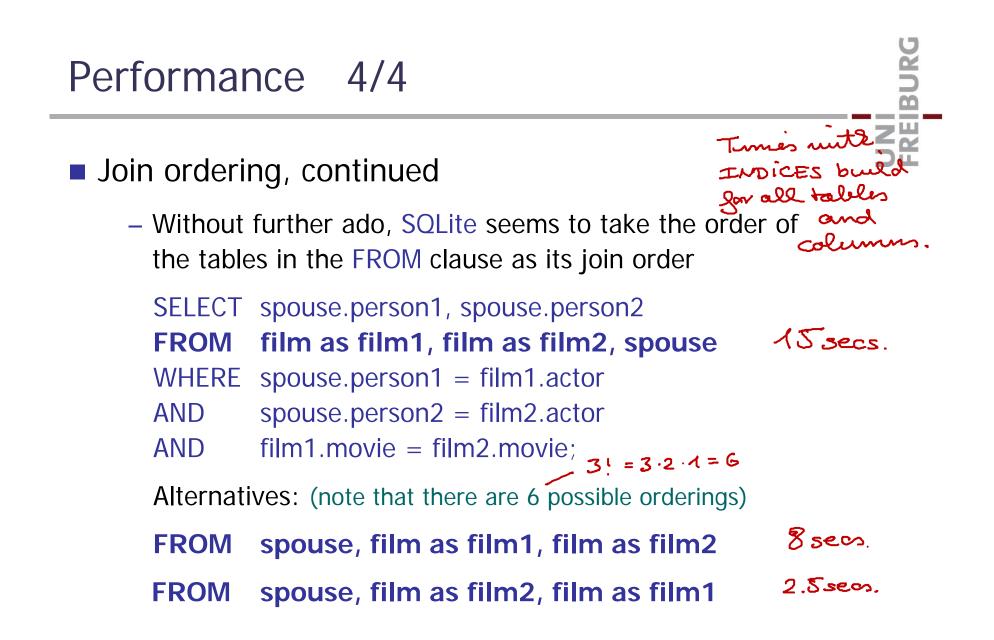
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- Order of joins can make a **huge** performance difference
- For our example query, the film table (actors movies) is more than ten times larger than the spouse table
- Join order 1: look at all married couples and for each get their movies and check whether they overlap

materializes list of movies of all married people (small)

 Join order 2: look at all pairs of actors who played in the same movie, and for each check whether they are married materialized all pairs of actors from same movie (large)



References

Textbook

 Nothing about this topic in the text book by Manning, Raghavan, and Schütze

Wikipedia

- http://en.wikipedia.org/wiki/Ontology_(information_science)

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- <u>http://en.wikipedia.org/wiki/SPARQL</u>
- http://en.wikipedia.org/wiki/SQL
- <u>http://en.wikipedia.org/wiki/SQLite</u>
- <u>http://en.wikipedia.org/wiki/Freebase</u>