

Exploring Knowledge Bases through Questions and Faceted Answers

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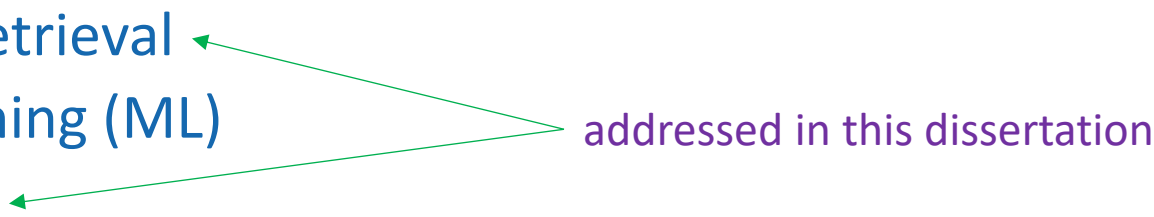
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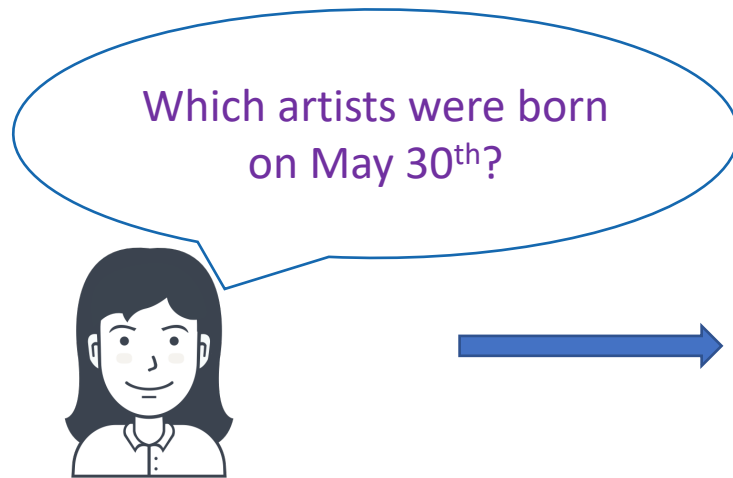
Introduction

- Question Answering (QA) systems combine techniques from multiple fields of computer science, among which:
 - Natural Language Processing (NLP)
 - Information Retrieval
 - Machine Learning (ML)
 - Semantic Web
 - A QA system may be split into two parts: *question*, which receives a user's input in natural language, transforms it into a structured query and searches the data; and *answer*, which displays consistent results in a human-readable format to the user
- 
- addressed in this dissertation

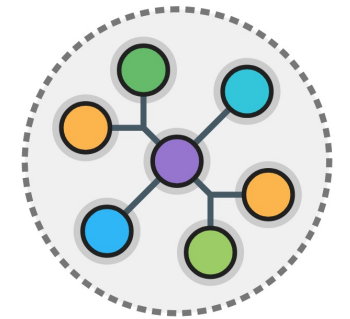
Introduction

- This study addresses the problem of *query answer modification* to improve the quality of the user's experience
 - in the context of an RDF knowledge base
- The dissertation proposes a *fully automated process* that reorganizes the original query answer by applying heuristics to summarize the results
- The heuristics together with a set of thresholds allow:
 - deciding which properties are interesting to apply aggregations (group by operations)
 - deciding if the answer is ready to be displayed to the user, or if the answer must be improved

Motivation



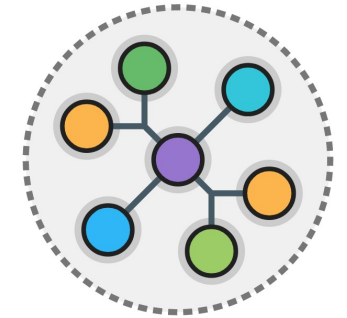
voice virtual assistant



RDF knowledge base

Motivation

#	Artist	Background	Birth Date	Death Date	Gender	Nationality
1	Goodman, Benny	Jazz	1909-05-30	1986-06-13	Male	American
2	Leonhardt, Gustav	Classical	1928-05-30	2012-01-16	Male	Dutch
3	Green, CeeLo	Pop	1974-05-30	-	Male	American
4	Biosphere	Electronic	1962-05-30	-	Male	Norwegian
5	Fredriksson, Marie	Pop	1958-05-30	2019-12-09	Female	Swedish
...
122	Banhart, Devendra	Folk	1981-05-30	-	Male	American



RDF knowledge base



voice virtual assistant

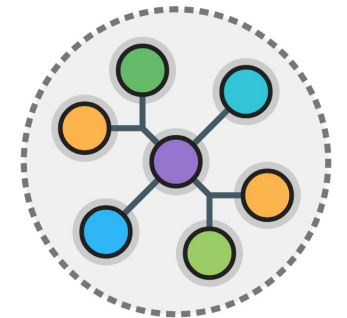
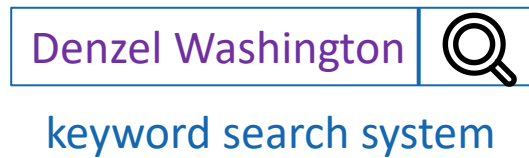
very long result set



Motivation

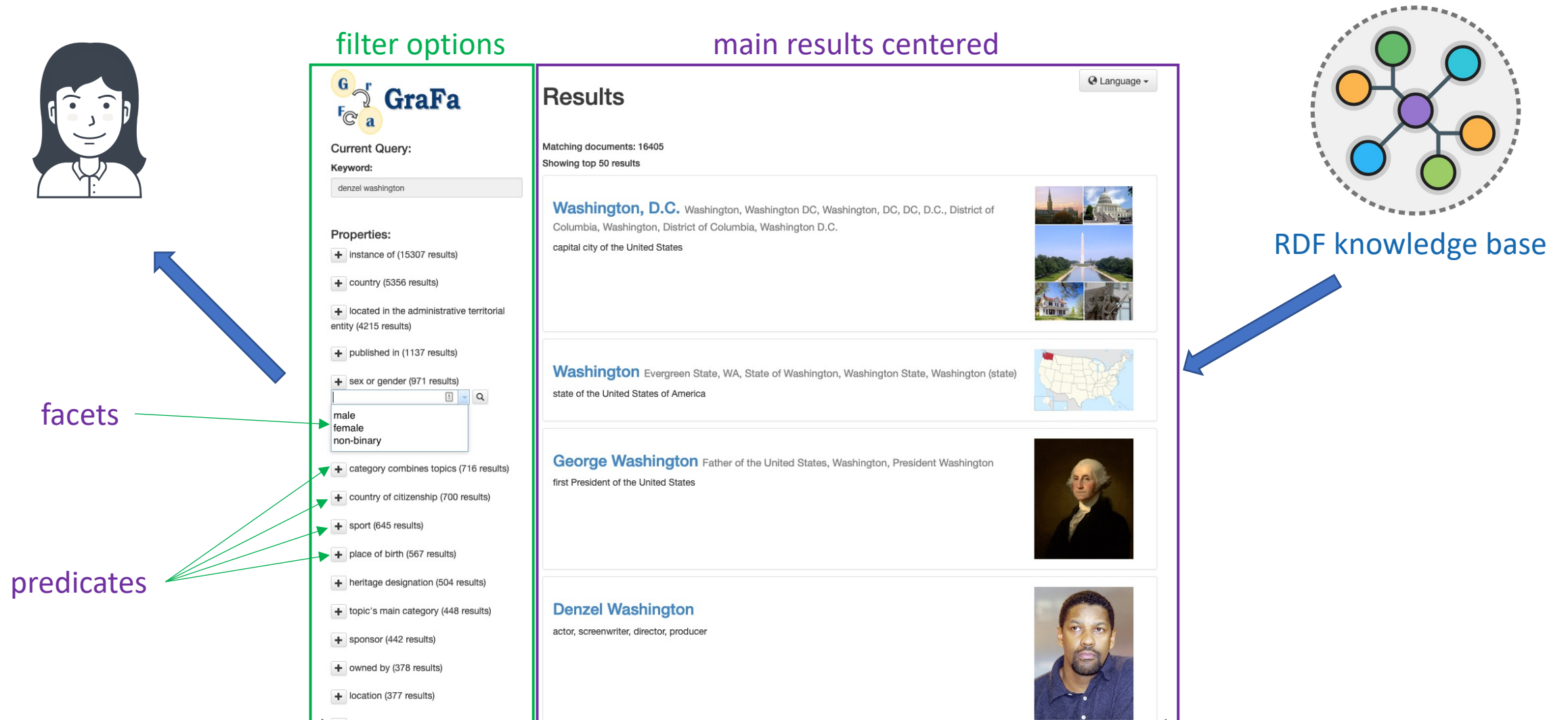
- Instead of listing the results, the virtual assistant may formulate questions to the user based on the prior result set, such as:
 - *“Do you want to list American or European artists?”*
 - *“Do you prefer Jazz, Pop, or Classical music?”*
 - *“Do you want to filter by active artists?”*

Motivation



RDF knowledge base

Motivation



Background and Related Work



Berners-Lee, T., Hendler, J., and Lassila, O. The semantic web; a new form of web content that is meaningful to computers will unleash a revolution of new possibilities. *Scientific American* (May 2001).

Linked Data (Semantic Web, RDF, SPARQL)



Webber, B. L. Questions, answers and responses: Interacting with knowledge-base systems. In *Topics in Information Systems*. Springer New York, 1986, pp. 365–402.

Question Answering (Humans, Machines)



Dalianis, H., and Hovy, E. Aggregation in natural language generation. In *Trends in Natural Language Generation An Artificial Intelligence Perspective*, J. G. Carbonell, J. Siekmann, G. Goos, J. Hartmanis, J. Leeuwen, G. Adorni, and M. Zock, Eds., vol. 1036. Springer Berlin Heidelberg, Berlin, Heidelberg, 1996, pp. 88–105.



Deutch, D., Frost, N., and Gilad, A. Provenance for natural language queries. *Proceedings of the VLDB Endowment* 10, 5 (Jan. 2017), 577–588.

Aggregation and Summarization



Moreno-Vega, J., and Hogan, A. Grafa: Faceted search & browsing for the wikidata knowledge graph. In *International Semantic Web Conference* (2018).

Faceted Browsing (or Faceted Search)

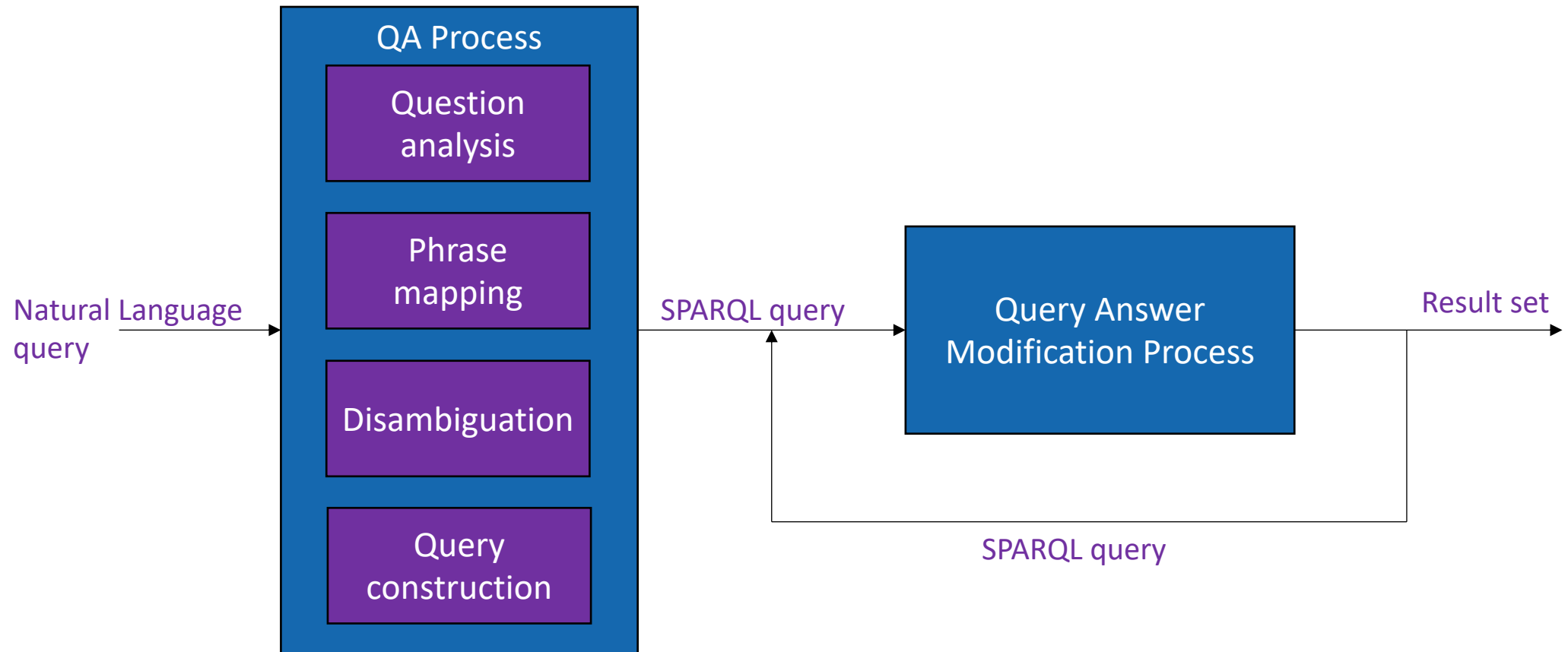
The Query Answer Modification Process

Overview

- Several studies addressed the problem of creating a question-answering (QA) interface to databases
- Usually, the proposed QA process has four steps:
 - Question Analysis
 - Phrase Mapping
 - Disambiguation
 - Query Construction
- The *query answer modification* process we propose starts after the query is executed

The Query Answer Modification Process

Overview



The Query Answer Modification Process

Heuristics and Thresholds

- The process main goal is to reduce the size of the original result set using a set of heuristics with parameterized thresholds
- Heuristics responsibilities:
 - user's navigation thru data
 - automate the predicate/facet decision
- Thresholds responsibilities:
 - cut undesirable branches
 - reduce result set
 - control the stop condition

The Query Answer Modification Process

Heuristics and Thresholds

- The words **restrictive** and **embracing** are used to characterize predicates and facets
- In this context, their meaning are:
 - **restrictive**: reduce the total number of rows of the result set
 - **embracing**: keep or barely reduce the total number of rows of the result set

Initial Result Set: 49 rows

Predicate/Facet selected:

`imdb:label => Columbia/Tristar`

Final Result Set: 5 rows

Compression Rate: 89,80%

Initial Result Set: 91 rows

Predicate/Facet selected:

`imdb:color_info => Color`

Final Result Set: 74 rows

Compression Rate: 18,68%

The Query Answer Modification Process

Heuristics and Thresholds

- The heuristics tested were:
 - Σ : from the most **embracing** predicate, select the most **embracing** facet
 - \prod : from the most **restrictive** predicate, select the most **embracing** facet
 - Ω : select the most **embracing** facet, regardless the predicate
- The thresholds applied were:
 - α : max number of predicates' distinct values
 - β : max number of unique subjects to be returned to the user
 - δ : min and max range of predicates' rate presence over unique subjects

The Query Answer Modification Process

Transforming single-column into three-column result sets

- A simple approach for enriching the answers is to add to the instances returned their property values (denormalization)

```
prefix mo: <http://purl.org/ontology/mo/>  
prefix foaf: <http://xmlns.com/foaf/0.1/>
```

```
select distinct ?artist  
where {  
    ?artist a mo:MusicArtist .  
    ?artist dbo:birthDate ?date .  
    filter(regex(?date, "5-30$", "i")) .  
}
```



artist
mo:MusicArtist/1
mo:MusicArtist/2
mo:MusicArtist/3
mo:MusicArtist/4
mo:MusicArtist/5
mo:MusicArtist/6
mo:MusicArtist/7
mo:MusicArtist/8
mo:MusicArtist/9
mo:MusicArtist/10

The Query Answer Modification Process

Transforming single-column into three-column result sets

```
select distinct ?artist ?predicate ?object
where {
  {
    select ?artist
    where {
      ?artist a mo:MusicArtist .
      ?artist dbo:birthDate ?date .
      filter(regex(?date, "5-30$", "i")) .
    }
  } # graph pattern 1 – prior query
  . # conjunction (inner join)
  {
    select ?artist ?predicate ?object
    where {
      ?artist ?predicate ?object .
      filter(isLiteral(?object)) .
    }
  } # graph pattern 2
}
```



artist	predicate	object
mo:MusicArtist/1	foaf:name	"Green, CeeLo"
mo:MusicArtist/1	mo:genre	"pop"
mo:MusicArtist/1	dbo:BirthDate	"1974-05-30"
mo:MusicArtist/1	dbo:DeathDate	" "
mo:MusicArtist/1	foaf:gender	"Male"
mo:MusicArtist/1	dbp:nationality	"American"
mo:MusicArtist/2	foaf:name	"Leonhardt, Gustav"
mo:MusicArtist/2	mo:genre	"Classical"
mo:MusicArtist/2	dbo:BirthDate	"1928-05-30"
mo:MusicArtist/2	dbo:DeathDate	"2012-01-16"
mo:MusicArtist/2	foaf:gender	"Male"
mo:MusicArtist/2	dbp:nationality	"Dutch"
mo:MusicArtist/3	foaf:name	"Goodman, Benny"
mo:MusicArtist/3	mo:genre	"Jazz"
mo:MusicArtist/3	dbo:BirthDate	"1909-05-30"
mo:MusicArtist/3	dbo:DeathDate	"1986-06-13"
mo:MusicArtist/3	foaf:gender	"Male"
mo:MusicArtist/3	dbp:nationality	"American"

The Query Answer Modification Process

Frequency analysis based on computed metadata

- There are two types of frequencies used in the process
 - *global frequency*: defined over full graph and computed once
 - *local frequency*: defined over sub-graph and computed at run time
 - both *frequencies* are computed over predicates pointing to literals only
- Entity ranking is based on *InfoRank*, a family of importance measures proposed in *Menendez et al. (2019)*
 - approach inspired by *PageRank* algorithm to propagate the importance scores from entity to entity
 - helps our process prioritize the most relevant triples of the result set

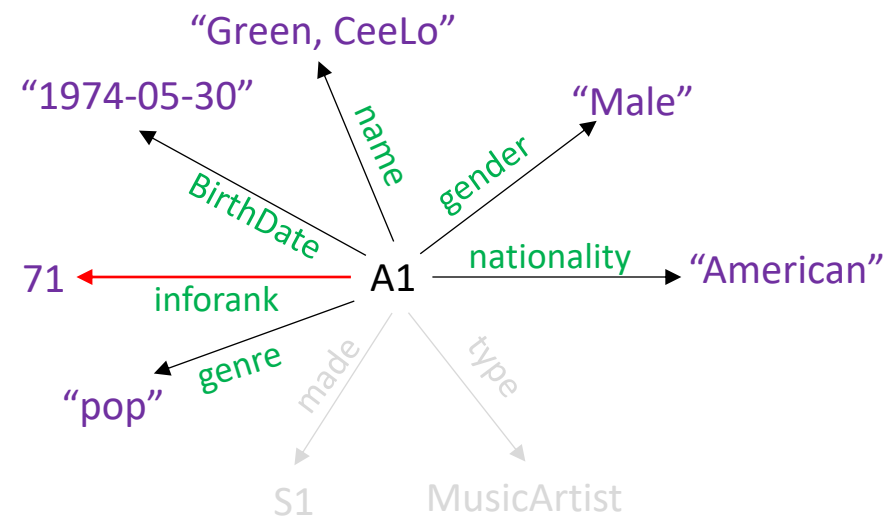
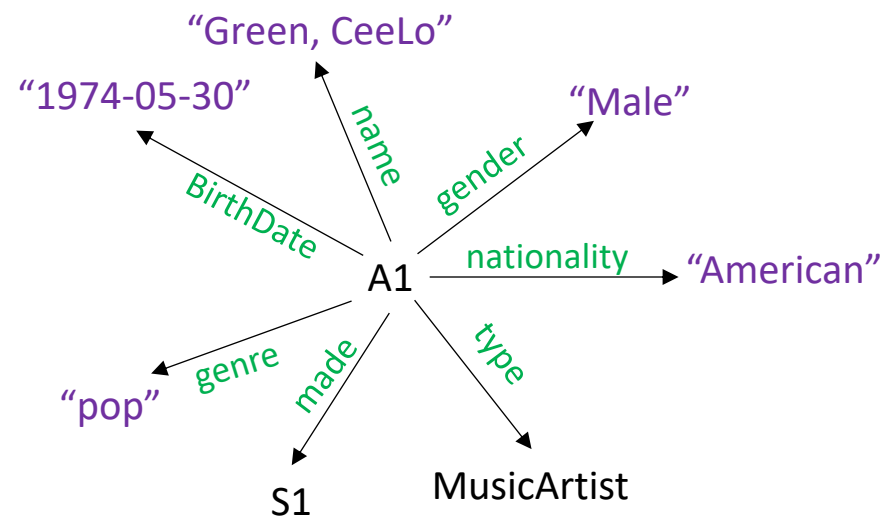


Menendez, E. S., Casanova, M. A., Leme, L. A. P., and Boughanem, M. (2019). Novel node importance measures to improve keyword search over rdf graphs. In International Conference on Database and Expert Systems Applications, pages 143–158. Springer. DOI:https://doi.org/10.1007/978-3-030-27618-8_11

The Query Answer Modification Process

Frequency analysis based on computed metadata

- Example with instance **A1**
 - left figure is the initial state: pointing to literals and other instances
 - right figure is the final state: pointing to literals only and *inforank* score
- The parameterized threshold δ is used to filter predicates that are candidates to be used in a group by operation



The Query Answer Modification Process

Verifying stop condition

- The process stops when:
 - the number of unique subjects is **less or equal than** the threshold β ; **or**
 - the previous result set is **the same as** the current result set; **and**
 - there is an empty set of selectable predicates
- Also, the final result set is sorted by *InfoRank*

Experiments

Benchmarks

- We performed initial experiments using the RDF versions of MusicBrainz and IMDb datasets (~200MM tuples each)
 - MusicBrainz dump enriched with DBpedia data
 - IMDb relational data transformed into RDF through R2RML
- We used sample queries from the QALD challenge and Coffman's benchmark
- To simplify the visualization, we decided to illustrate the example flows as K-D trees
 - the examples were generated applying heuristic Σ

Experiments

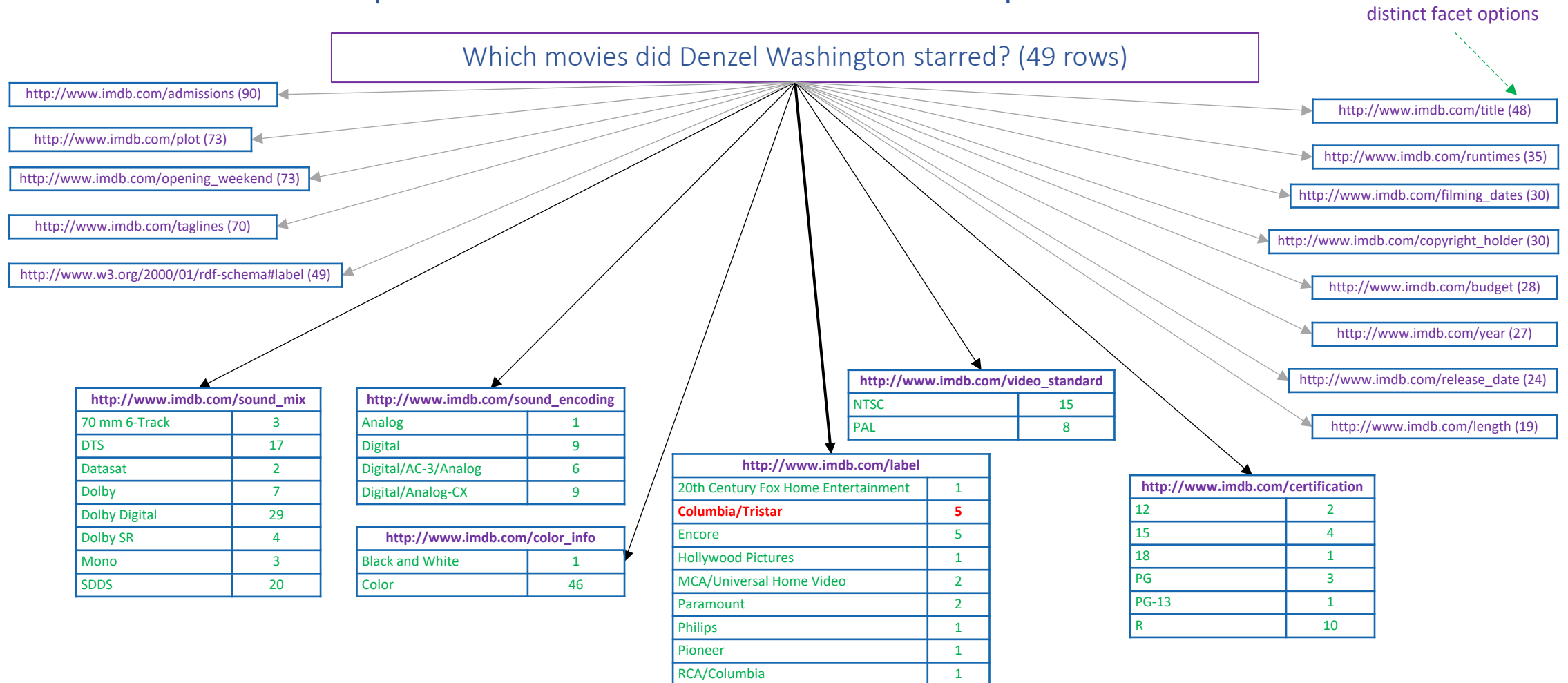
K-D Three Example with IMDb dataset – 1st step

Σ : most **embracing** predicate, most **embracing** facet

$\alpha = 10$: max number of predicates' distinct values (distinct facet options)

$\beta = 15$: max number of unique subjects in final result set

$\delta = (0.4, 2)$: min and max range of predicates' rate presence over unique subjects



Experiments

K-D Three Example with MusicBrainz dataset – 1st step

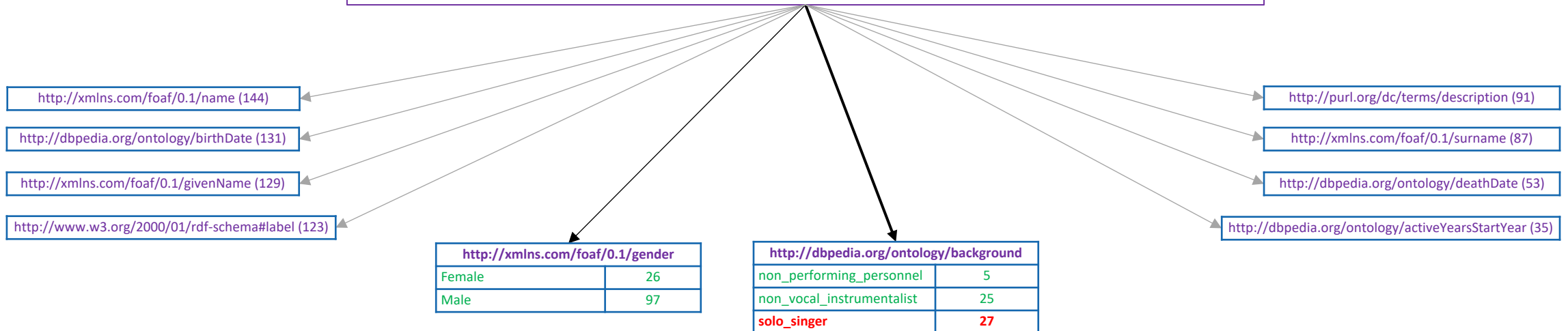
Σ : most **embracing** predicate, most **embracing** facet

$\alpha = 10$: max number of predicates' distinct values (distinct facet options)

$\beta = 15$: max number of unique subjects in final result set

$\delta = (0.4, 2)$: min and max range of predicates' rate presence over unique subjects

Which artists were born on May 30th? (123 rows)



Experiments

K-D Three Example with MusicBrainz dataset – 2nd step

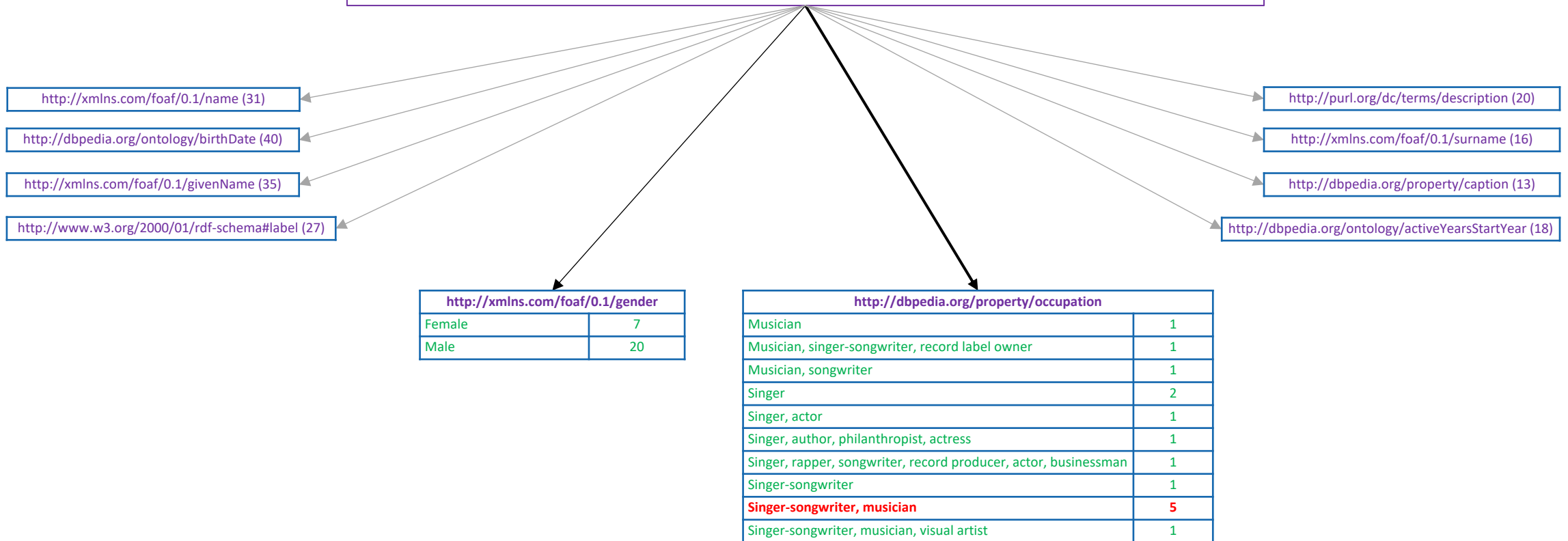
Σ : most **embracing** predicate, most **embracing** facet

$\alpha = 10$: max number of predicates' distinct values (distinct facet options)

$\beta = 15$: max number of unique subjects in final result set

$\delta = (0.4, 2)$: min and max range of predicates' rate presence over unique subjects

Which artists were born on May 30th and are **solo singers**? (27 rows)



Experiments

Compression Rate and final analysis

- A *Compression Rate* metric was defined and used to compare the obtained results

1. γ = compression rate
 2. η = # lines of the initial result set
 3. κ = # lines of the final result set
 4. $\gamma = 1 - \kappa / \eta$
- when $\kappa \sim \eta$ the compression is low
- when $\kappa \ll \eta$ the compression is high

- Recall that variable β sets an upper bound on the final result set number of lines

Experiments

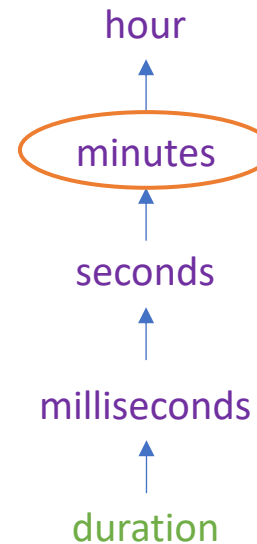
Compression Rate and final analysis - MusicBrainz

- Ten questions were originally tested over dataset, but three of them had not a single selectable predicate available
- Question: *“What are the songs performed by Aretha Franklin?”*
- Predicates:
 - `mo:track_number` (54 distinct facet options)
 - `rdfs:label` (1120 distinct facet options)
 - `mo:duration` (1881 distinct facet options)
- All predicates had the number of facets greater than α

Experiments

Compression Rate and final analysis - MusicBrainz

- Suppose the process have taxonomy information about specific predicates
 - the aggregations would be applied over multi-level
 - mo:duration is represented in milliseconds



#	Duration	Total Songs
1	11.0	9
2	10.0	6
3	9.0	21
4	8.0	28
5	7.0	69
6	6.0	157
7	5.0	337
8	4.0	781
9	3.0	1110
10	2.0	399

Experiments

Compression Rate and final analysis - MusicBrainz

Question	1	Which artists were born on May 30th?	IRS	FRS	Compression Rate
Heuristic	Steps	Facets			
Σ	2	dbo:background solo_singer (27); dbp:occupation Singer-songwriter (5);	123	5	95,93%
Π	3	foaf:gender male (97); dbo:background non_vocal_instrumentalist (23); dbo:alias Beyond The Wizards Sleeve (2);	123	2	98,37%
Ω	3	foaf:gender male (97); dbo:background non_vocal_instrumentalist (23); dbp:occupation Musician (4);	123	4	96,75%

- The first question presented excellent *Compression Rate* and selected meaningful predicates/facets
- Only the Π heuristic chose a questionable predicate in its 3rd step
 - `dbo:alias` should be almost unique by artist
 - the process is selecting a specific artist instead of selecting a property that few artists have in common

Experiments

Compression Rate and final analysis - MusicBrainz

Question	3	Which artists played on the same groups that David Bowie was member of?			
Heuristic	Steps	Facets	IRS	FRS	Compression Rate
Σ	1	dbo:wikiPageID 1515176 (1);	17	1	94,12%
Π	1	foaf:gender male (14);	17	14	17,65%
Ω	1	foaf:gender male (14);	17	14	17,65%

- In this case, the initial result set was already very close to variable β
- The *Compression Rate* from heuristics Π and Ω were the worst
- Analyzing the predicate's meaning, heuristic Σ made a bad choice
 - foaf:gender seems much more interesting than dbo:wikiPageID
 - the *Compression Rate* metric by itself does not mean all

Experiments

Compression Rate and final analysis - MusicBrainz

Question	7	Which bands broke up in 2010?	IRS	FRS	Compression Rate
Heuristic	Steps	Facets			
Σ	2	dbo:background group_or_band (236); dbo:activeYearsEndYear 2010 (236);	238	15	93,70%
Π	2	dbo:background group_or_band (236); dbo:activeYearsEndYear 2010 (236);	238	15	93,70%
Ω	2	dbo:activeYearsEndYear 2010 (238); dbo:background group_or_band (236);			

- In all cases, the process selected meaningful predicates but useless
 - selected predicates were related to the original question
 - group_or_band and 2010 compressed almost nothing
- The *Compression Rate* was good because of β threshold

Experiments

Compression Rate and final analysis - IMDb

Question	1	Which movies did Denzel Washington starred?			
Heuristic	Steps	Facets	IRS	FRS	Compression Rate
Σ	1	imdb:label Columbia/Tristar (5);	49	5	89,80%
Π	2	imdb:color_info Color (46); imdb:video_standard NTSC (14);	49	14	71,43%
Ω	3	imdb:color_info Color (46); imdb:sound_mix Dolby Digital (28); imdb:video_standard NTSC (9);	49	9	81,63%

- Except the heuristic Σ , the other heuristics in this first question selected some unhelpful predicates/facets
- Analyzing the other *Compression Rates*, all results were good because of β threshold
 - all other questions had no restrict facet and the original result set was long
 - $\kappa \ll \eta$ since $\kappa = \beta = 15$

Experiments

Compression Rate and final analysis - IMDb

- We decided to investigate further the lack of predicates with restrictive facets in this dataset
- The questions evaluated were related to resources:
 - `imdb:Movie`
 - `imdb:Actor` and `imdb:Actress` – called *Artists*

Experiments

Compression Rate and final analysis - IMDb

#	Predicate	Distinct Values
1	rdfs:label	512491
2	imdb:title	512469
3	imdb:release_dates	274127
...
53	imdb:subtitles	12
54	imdb:analog_left	8
55	imdb:picture_format	8
56	imdb:sound_encoding	8
57	imdb:digital_sound	7
58	imdb:analog_right	6
59	imdb:category	4
60	imdb:color_information	4
61	imdb:status_of_availability	4
62	imdb:video_standard	4
63	imdb:close_captions-teletext-ld-g	3
64	imdb:disc_format	3
65	imdb:disc_size	3
66	imdb:master_format	3
67	imdb:color_info	2
68	imdb:quality_program	1

imdb:Movie results

#	Predicates	Distinct Values
1	imdb:name	2192378
2	rdfs:label	2191943
3	imdb:trivia	495432
4	imdb:aka	470130
5	imdb:other_works	305969
...
22	imdb:salary_history	5384
23	imdb:biographical_movies	4648
24	imdb:portrayed_in	3585
25	imdb:height	444
26	imdb:gender	2

Artists' results

- The set of selectable predicates is small, and their meaningfulness is also low
- Except `imdb:category` and `imdb:gender` all predicates are related to technical information about the movies

Conclusion and Future Work

- The main contribution of this study was the definition of a process - called *Query Answer Modification Process*
 - addressed the problem related to open-ended questions (long result sets)
 - simplicity is key (combined heuristics with set of thresholds)
- Results compared and discussed over known baselines
 - QALD challenge and Coffman's benchmark
 - *Compression Rate* metric define to compare results

Conclusion and Future Work

- The initial research involving the *Query Answer Modification Process* **was published** in the Proceedings of the XXXV Brazilian Symposium on Databases – SBBD
- An extended version of this paper was submitted to the Journal of Information and Data Management (JIDM) and **was approved**
- Suggestions for improvements - part 1:
 - **allow users to provide a list of ignored predicates**
 - **or enable dynamically exclusion of undesired predicates in each step**
 - **develop and apply a questionnaire to users interested in using the system**
 - **we would be able to segment users by preferences and create user profiles**
 - **the system would be able to adapt heuristics based on associated profile**

Conclusion and Future Work

- Suggestions for improvements - part 2:
 - allow users to inform multi-level of specific predicates (taxonomy)
 - or automatically infer the related taxonomy during initial exploration of the RDF graph
 - develop a user interface similar to GraFa
 - navigation through predicates and facets
 - users of all types (beginners or experienced) would be able to use and evaluate our process
 - allow users to select more than one option at each interaction
 - disjunctive filtering would be possible, together with conjunctive filtering already developed

Thank you!

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