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FORTH-ICS



ACKNOWLEDGEMENTS

- Research done mainly with my students (more in the acknowledgements)
 - Pavlos Fafalios



Panagiotis Papadakos





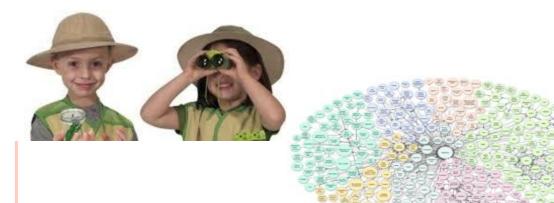
OUTLINE

- Introduction and Background (60')
 - Exploratory Search
 - Semantic Web and Linked Data
 - Faceted Exploration
- Bridging the Web of Documents with the Web of Data (50')
 - Possible Ways
 - Focus on doing this at search time
 - Case studies: presented as a series of 9 milestones
- Synopsis, Challenges, Discussion (10')
- References
- Acknowledgements



"Study without desire spoils the memory, and it retains nothing that it takes in." Leonardo da Vinci

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1. Introduction and Background

Exploratory Search
Semantic Web and Linked Data
Faceted Exploration







WHAT USERS USUALLY WANT/DO WHEN SEARCHING?

Kinds of information needs

- Precision-oriented
 - Locate one resource or/and its attributes
 e.g. Find the telephone of a store
- Recall-oriented
 - Locate a **set** of resources e.g. Medical information seeking, travel planning

Over 60% of web search queries are *recall-oriented*[Broder 02, Rose and Levinson 04]





RECALL-ORIENTED INFORMATION NEEDS

- In Recall-Oriented Information Needs:
 - the users require >1 hit
 - essentially such needs correspond to decision tasks
- Examples of Recall-oriented information needs
 - Booking
 - Product-buying
 - Bibliography search
 - Patent Search
 - Medical Search
 - • • •





QUESTIONS



How many of you (raise your hand):

- have bought your mobile phone by looking at the first hit(s) that google returned to your query?
- have ever booked the 1st hotel returned by google?
- have entirely read the first paper that Scholar Google returned to your query?
- have made at least once a search about a medical issue?
 - keep your hand raised if you had submitted to Google only one query
 - keep your hand raised if you have read only the first 5 hits of that single query that you submitted to Google, and eventually that was enough for fully satisfying your information need





EXPLORATORY SEARCH



Wikipedia:

- "Exploratory search is a specialization of information exploration which represents the activities carried out by searchers who are either:
- a) unfamiliar with the domain of their goal (i.e. need to learn about the topic in order to understand how to achieve their goal)
- b) unsure about the ways to achieve their goals (either the technology or the process)
- c) or even unsure about their goals in the first place.

Consequently, exploratory search covers a broader class of activities than typical information retrieval, such as *investigating*, *evaluating*, *comparing*, *and synthesizing*, where new information is sought in a defined conceptual area; exploratory data analysis is another example of an information exploration activity. Typically, therefore, such users generally combine querying and browsing strategies to foster learning and investigation."





THEREFORE...



•Ranking is not enough for exploratory search



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EXPLORATORY SEARCH: FROM FINDING TO UNDERSTANDING

Research tools critical for exploratory search success involve the creation of new interfaces that move the process beyond predictable fact retrieval.

rom the earliest days of computers, search has been a fundamental application that has driven research and development. For example, a paper published in the inaugural year of the IBM Journal 36 years ago outlined challenges of text retrieval that continue to the present [4]. Today's data storage and retrieval applications range from database systems that manage the bulk of the world's structured data to Web search engines that provide access to petabytes of text and multimedia data. As computers have become consumer products and the Internet has become a mass medium, searching the Web has become a daily activity for everyone from





children to research scientists.

SOME COMMON REQUIREMENTS FOR EFFECTIVE EXPLORATORY SEARCH



- Allow easy and fast access even to low ranked hits
- Allow browsing and inspecting the found hits in groups (according to various criteria)
- Offer **overviews** of the search results
 - Compute and show descriptions and **count** information for the various groups, or other **aggregated** values
- Allow **gradual** restriction/ranking of the search results

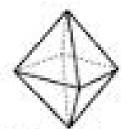


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FACETED SEARCH



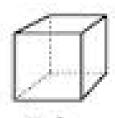
Tetrahedron



Octahedron



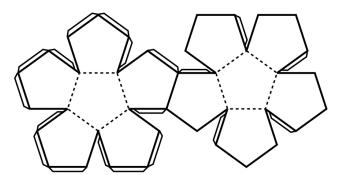
Icosahedron



Cube



Dodecahedro





FACETED SEARCH/EXPLORATION

Faceted Exploration is a **widely** used interaction scheme **for Exploratory Search**

A short (and rather informal) definition:

FE is a **session-based** interactive method for **query formulation** (commonly over a multidimentional information space) through simple clicks offering

- ✓ an overview of the result set (groups and count information)
- ✓ never leading to empty results sets
- The access paradigm supported is a *conceptual exploration*,
 - far more frequent in "search" tasks than the retrieval by exact specification supported by search engines and database queries.
- o Simple and easily understood by users.





FACETED SEARCH

Wikipedia:

- Faceted search, also called faceted navigation or faceted browsing, is a technique for accessing information organized according to a faceted classification system, allowing users to explore a collection of information by applying multiple filters. A faceted classification system classifies each information element along multiple explicit dimensions, called facets, enabling the classifications to be accessed and ordered in multiple ways rather than in a single, predetermined, taxonomic order.
- o Facets correspond to properties of the information elements. They are often derived by analysis of the text of an item using entity extraction techniques or from pre-existing fields in a database such as author, descriptor, language, and format. Thus, existing web-pages, product descriptions or online collections of articles can be augmented with navigational facets.
- Within the academic community, faceted search has attracted interest primarily among <u>library and information science</u> researchers, and to some extent among <u>computer science</u> researchers specializing in <u>information retrieval</u>





A PART OF MY INCOMPLETE FACEBOOK OF FACETED CLASSIFICATION/BROWSING/SEARCH



S. R. Ranganathan (1892-1972 Faceted Classification



Giovanni Maria Sacco Dynamic Taxnomies



Sébastien Ferré University Rennes 1



Marti A. Hearst University of California, Berkeley



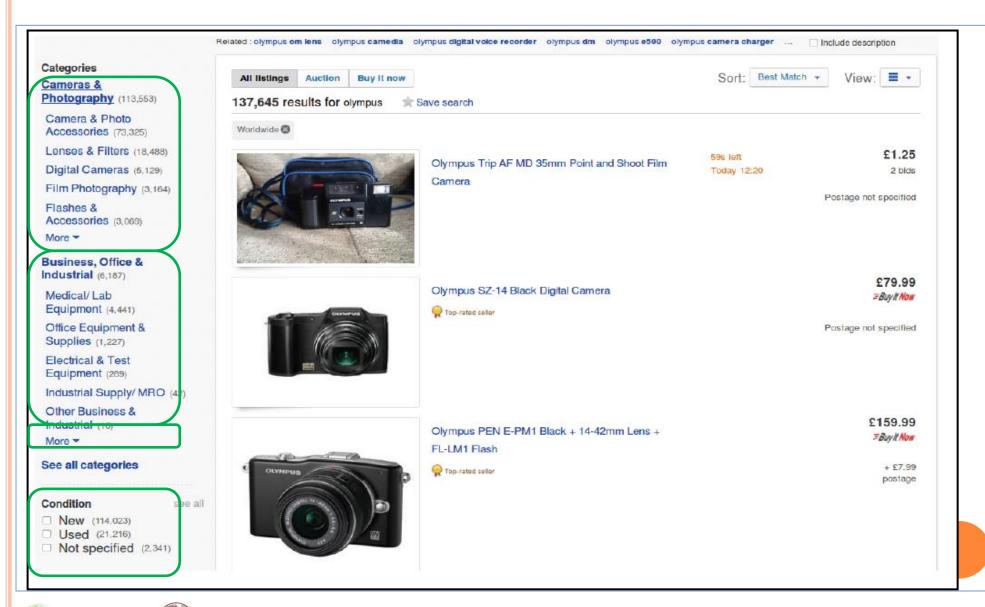


• Let's now see some examples from some widely used systems



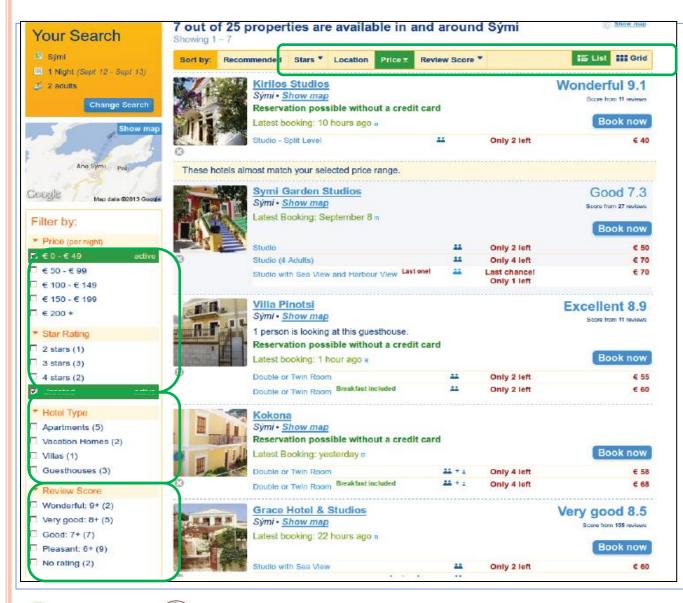


EXAMPLE: **EBAY**



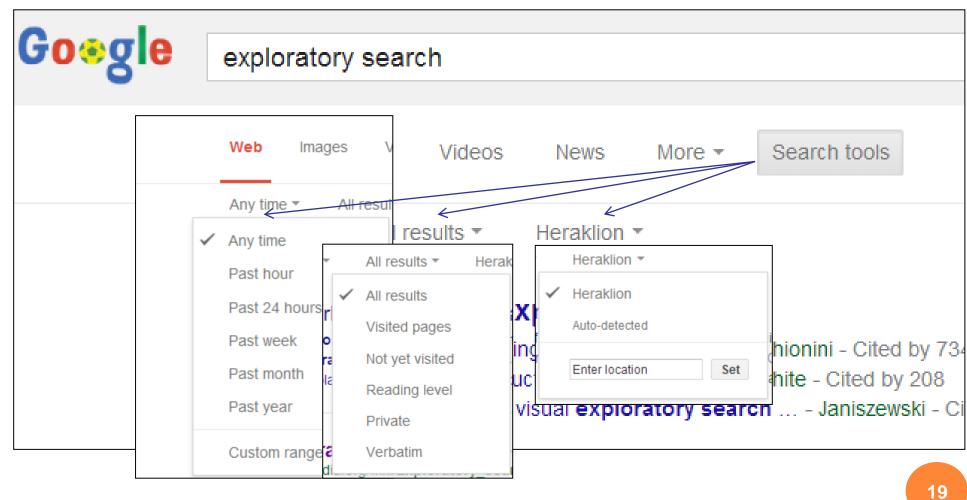


EXAMPLE OF FDT: BOOKING.COM



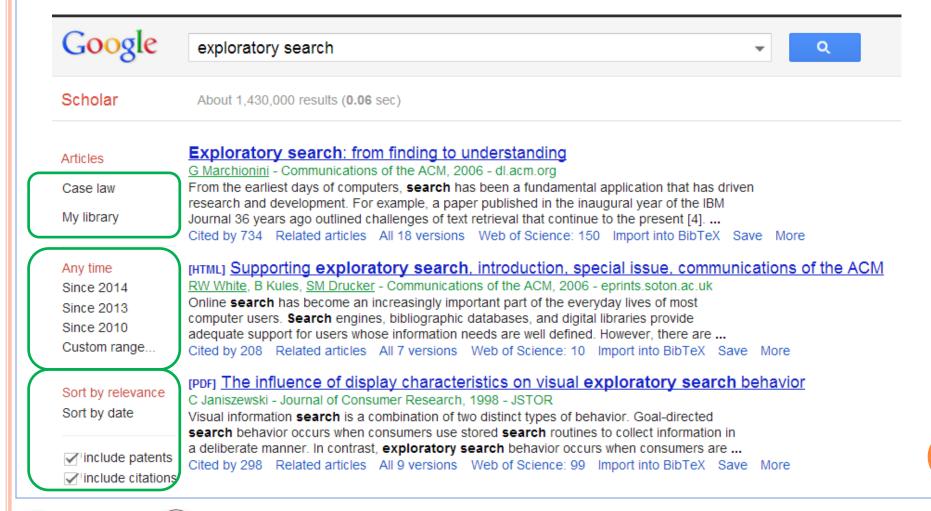


EXAMPLE: GOOGLE SEARCH (LIMITED FUNCTIONALITY: NO COUNT INFORMATION)





EXAMPLE: SCHOLAR GOOGLE





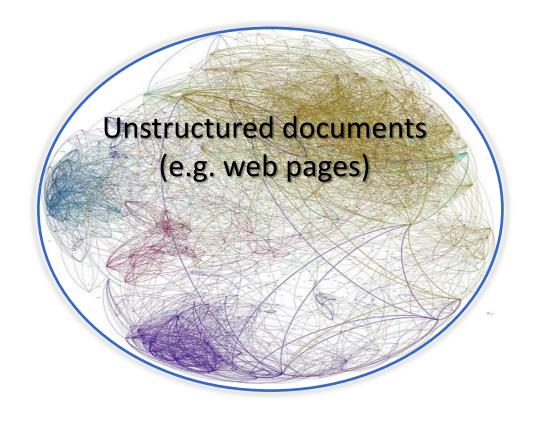
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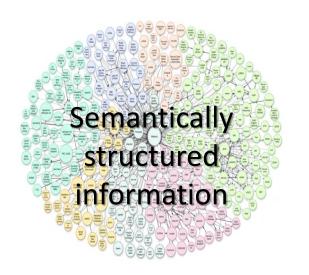
More on FE

- We will return to define more formally the interaction of faceted exploration, after first making a short introduction of the Semantic Web and Linked Data.
 - After that we will show how FE can be applied on such data



RELATIVELY RECENTLY A ... SECOND "WORLD" STARTS GROWING















THE SEMANTIC WEB VISION

- The Semantic Web is an evolving extension of the WWW where the content can be expressed not only in natural language but also in formal languages (e.g. RDF/S, OWL) that can be read and used by software agents, permitting them to find, share and integrate information more easily
- Imagine that the objective is the <u>collaborative creation</u> and <u>evolution</u> of a <u>world wide distributed graph</u> (about everything ②)
 - We could say that this graph resembles the structure of an EntityRelationship Diagram (recall Databases)



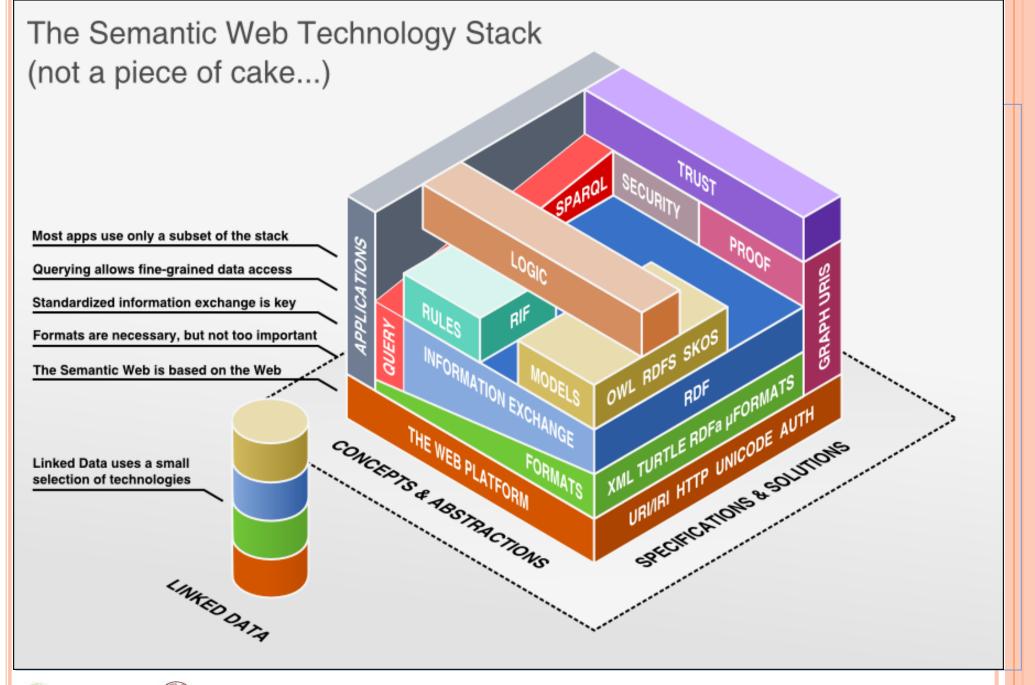


SEMANTIC WEB TECHNOLOGIES

- For achieving the Semantic Web vision, the recent years several technologies have been emerged, many of them are international (W3C) standards
- These technologies include:
 - knowledge representation languages (e.g. RDF/S, OWL) and formats for exchanging knowledge
 - query languages (π.χ. SPARQL),
 - rule languages and inference engines
 - Techniques for constructing mappings for integrating/harmonizing schemas and data
 - Technologies for mining structure knowledge from texts
 - Various APIs.
- Linked Open Data Are based on these technologies.





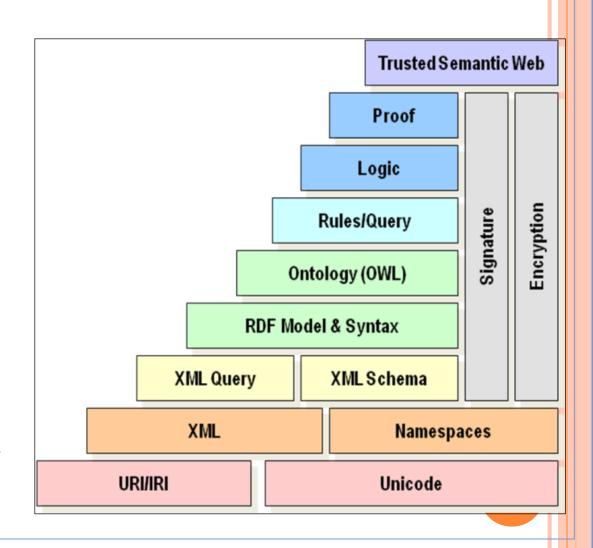




THE TECHNOLOGY STACK OF SEMANTIC WEB

By starting from the World Wide Web where the content is represented as hypertext (i.e. pages containing texts and links to other pages), let's now give an overview of the basic technologies on which the Semantic Web is based.

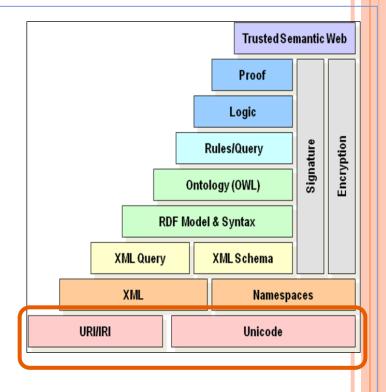
The current <u>technology</u> <u>stack</u> of <u>Semantic Web</u> can be illustrated as follows:





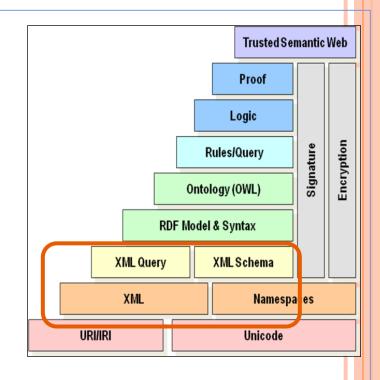
TECHNOLOGICAL STACK: URIS & UNICODE

• At the lowest level we have the URI (Uniform Resource Identifiers) for identifying and locating resources, and UNICODE for representing characters from various natural languages



TECHNOLOGICAL STACK: XML

- XML provides syntax for having structured documents, but does not impose any constraint regarding the semantics of these documents.
- XML Schema offers a method to restrict the structure that XML documents can have.
- For querying XML documents, we have XPath and XQuery.

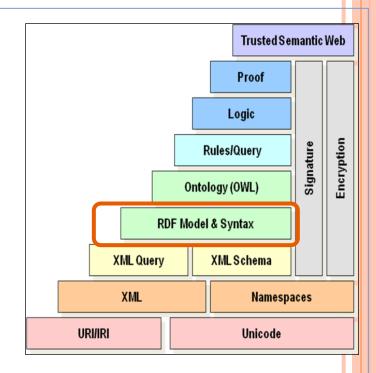






TECHNOLOGICAL STACK: RDF & RDFS

- RDF (Resource Description Framework) is a structurally objectoriented model for representing objects (resources) and associations between them.
- It allows expressing content in form of triples (*subject*, *predicate*, *object*), and sets of triples actually form a semantic network/graph.
- These triples can be expressed in various formats (TriG, N3 RDF/XML), some of them are based on XML (RDF/XML)
- RDFS allows defining the vocabulary to be used in RDF and semantic relationships between the elements of this vocabulary

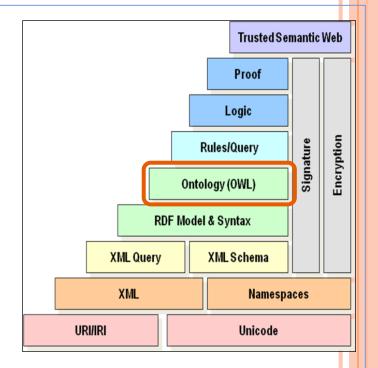






TECHNOLOGICAL STACK: OWL

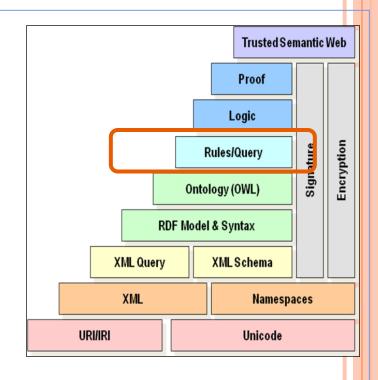
• The Web Ontology Language (OWL) is a family of knowledge representation languages or ontology languages for authoring ontologies or knowledge bases. The languages are characterized by formal semantics and RDF/XML-based serializations for the Semantic Web.





TECHNOLOGICAL STACK: SPARQL & SWRL

- For exploiting the structured content that has been represented using RDF/S, there are query languages and rule languages
- Specifically, SPARQL (SPARQL Protocol and RDF Query Language) is a query language for knowledge expressed in RDF/OWL.
- SWRL (Semantic Web Rule Language) allows expressing inference rules (essentially Horn rules).

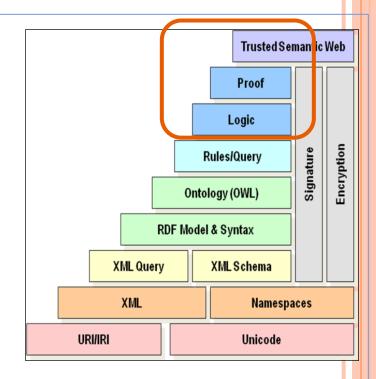






TECHNOLOGICAL STACK: LOGIC, PROOF, TRUST

- The layers *Logic* and *Proof* concern the enrichment of the expressiveness of the representation languages
- o Finally the *Trust* layer concerns trust issues, e.g. digital signatures (for proving that one particular person has written or agrees with a particular document or sentence, as well as trust networks allowing users to define who they trust (and so on), eventually yielding trust networks (Web of Trust)





STANDARDIZATION, PERSISTENT STORAGE

- o The part of the stack that has been implemented and standardized allows describing resources (of digital objects mainly) using classes and properties that have been defined in ontologies (expressed using RDF/S or OWL) which are accessible from the network which in turn can be connected (one ontology can extend classes and properties of other ontologies).
- For the persistent storage of such contents (independently of whether it corresponds to data or metadata) in the form of connected semantic network (semantic graph), there are tools, usually referred to as triplestores. We could call them *semantic data bases* since they receive as input RDF/S data, support integrity constraints, enable querying (SPARQL support), and concurrent access.

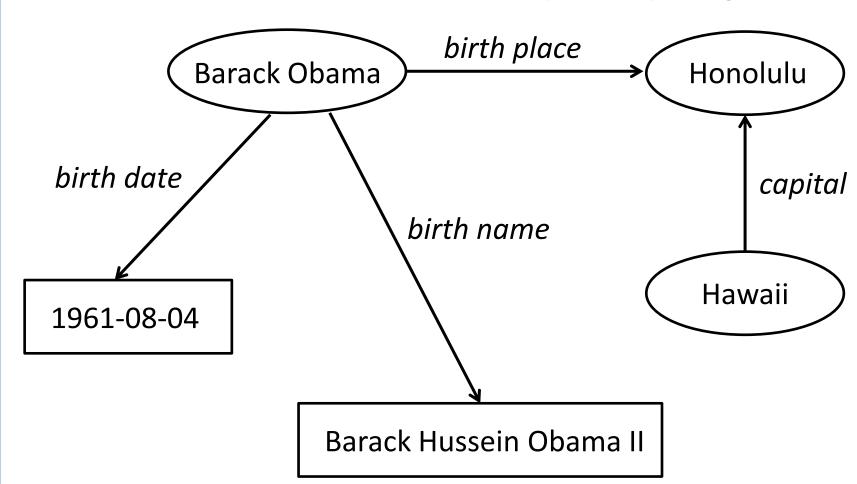






EXAMPLE OF AN RDF GRAPH

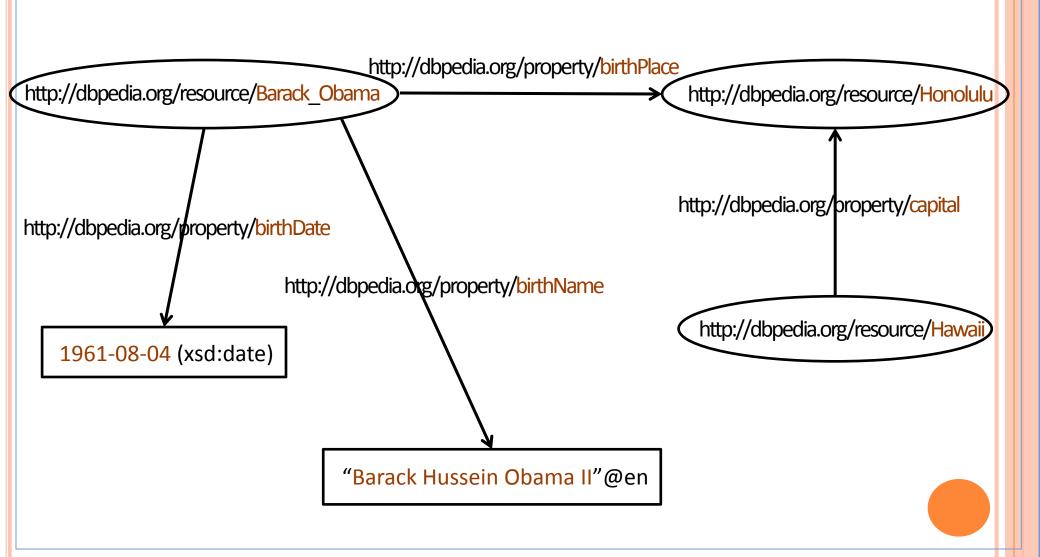
Directed Labeled (Multi)Graph







RDF GRAPH







RDF GRAPH REPRESENTATION

In XML (RDF/XML):

```
k?xml version="1.0" encoding="utf-8" ?>
krdf:RDF
   xmlns:rdf="http://www.w3.org/1999/02/22-rdf-syntax-ns#"
   xmlns:dbpprop="http://dbpedia.org/property/"
 <rdf:Description rdf:about="http://dbpedia.org/resource/Barack Obama">
   <dbpprop:birthDate rdf:datatype="http://www.w3.org/2001/XMLSchema#date">
        1961-08-04
   </dbpprop:birthDate>
   <dbpprop:birthPlace rdf:resource="http://dbpedia.org/resource/Honolulu" />
   <dbpprop:birthName xml:lang="en">
        Barack Hussein Obama II
   </dbpprop:birthName>
 </rdf:Description>
 <rdf:Description rdf:about="http://dbpedia.org/resource/Hawaii">
   <dbpprop:capital rdf:resource="http://dbpedia.org/resource/Honolulu" />
 </rdf:Description>
</rdf:RDF>
```

In N-Triples:

```
<http://dbpedia.org/resource/Barack Obama> <http://dbpedia.org/property/birthPlace> <http://dbpedia.org/resource/Honolulu> .

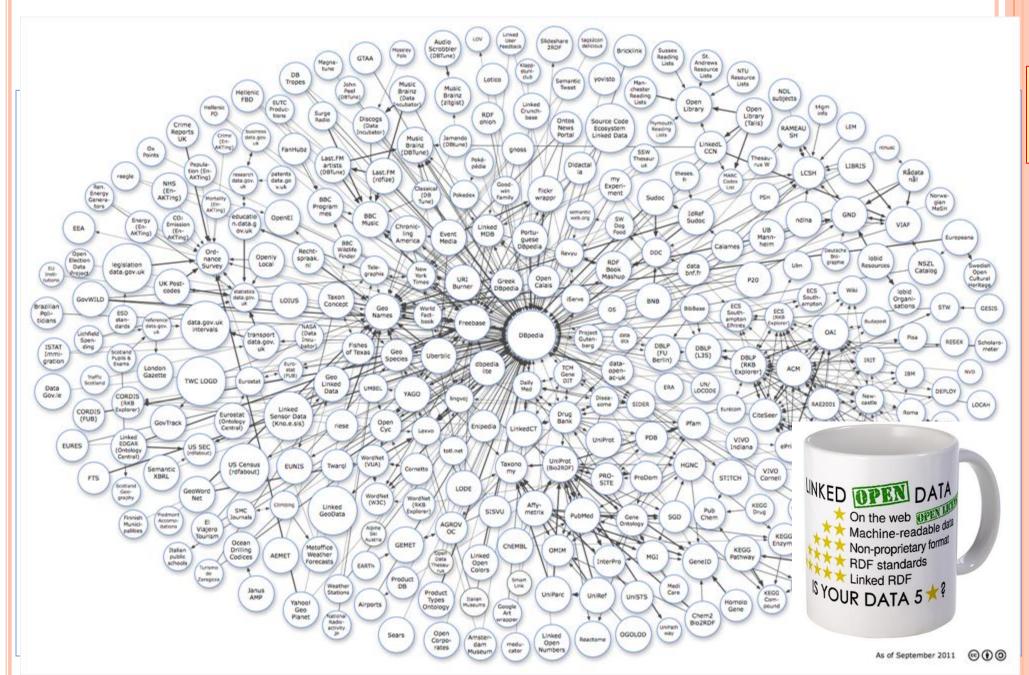
<http://dbpedia.org/resource/Barack_Obama> <http://dbpedia.org/property/birthDate> "1961-08-04"^^<http://www.w3.org/2001/XMLSchema#date> .

<http://dbpedia.org/resource/Barack_Obama> <http://dbpedia.org/property/birthName> "Barack Hussein Obama II"@en .

<http://dbpedia.org/resource/Honolulu> .
```



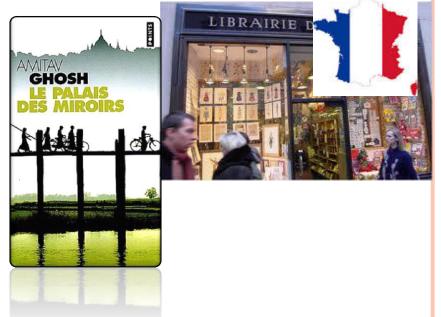






EXAMPLE: LINKED DATA SCENARIO





This example will show how data from two *bookstores* can be exposed and published in a <u>structured way</u>, how they can be integrated, how they can be queried, how they can be enriched with other knowledge (e.g. context or domain knowledge) from other sources (e.g. wikipedia).

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See acknowledgements for the source of this example





LINKED DATA SCENARIO A SIMPLIFIED BOOKSTORE DATA (DATASET "A")

Amitav Ghosh
THE GLASS PALACE
The magnificent, peignant, fascinating novel of three generations that starts in Mandalay...

Suppose the data of the bookstore A are stored in a relational database. The relational tuples related to the book of our running example could be

Books

ISBN	Author	Title	Publisher	Year
0006511409X	id_xyz	The Glass Palace	id_qpr	2000

Authors

ID	Name	Homepage
id_xyz	Ghosh, Amitav	http://www.amitavghosh.com

Publishers

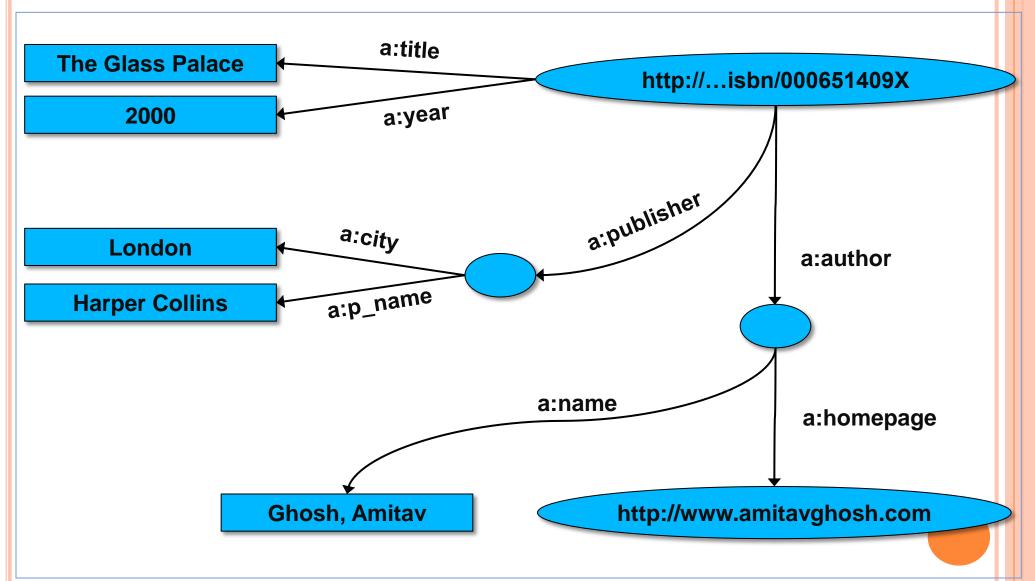
ID Publisher's nan		Publisher's name	ne City	
id_qpr		Harper Collins	London	





LINKED DATA SCENARIO

1ST: EXPORT YOUR DATA AS A *GRAPH*





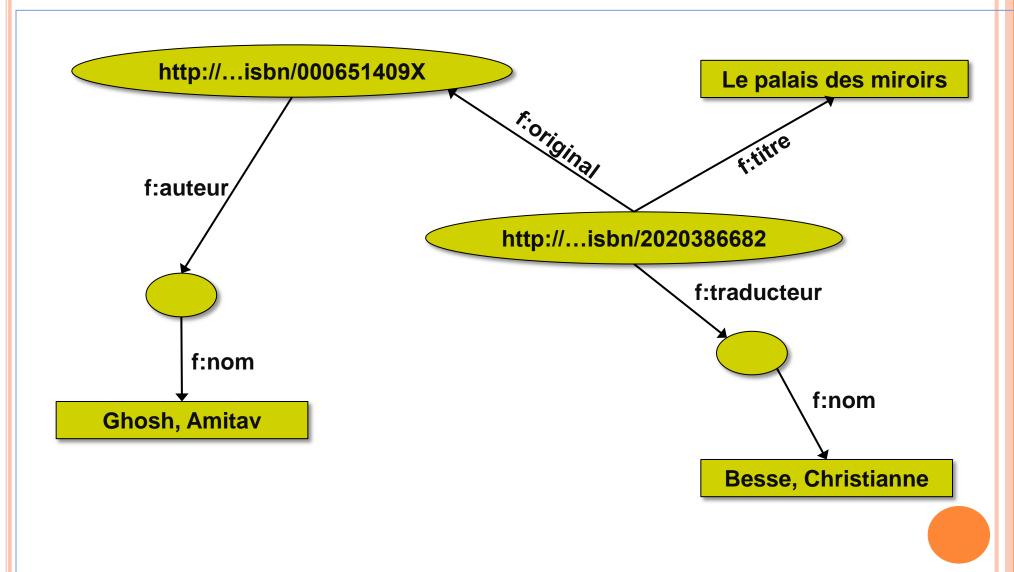


LINKED DATA SCENARIO ANOTHER BOOKSTORE DATA (DATASET "F" FROM FRANCE)

	A	В	C	D
1	ID	Titre	Traducteur	Original
2	ISBN 2020286682	Le Palais des Miroirs	\$A12\$	ISBN 0-00-6511409-X
3				
4				
5				
6	ID	Auteur		
7	ISBN 0-00-6511409-X	\$A11\$		
8				AWITAV GHOSH
9				DES MIROIRS
10	Nom			The sto
11	Ghosh, Amitav			
12	Besse, Christianne			

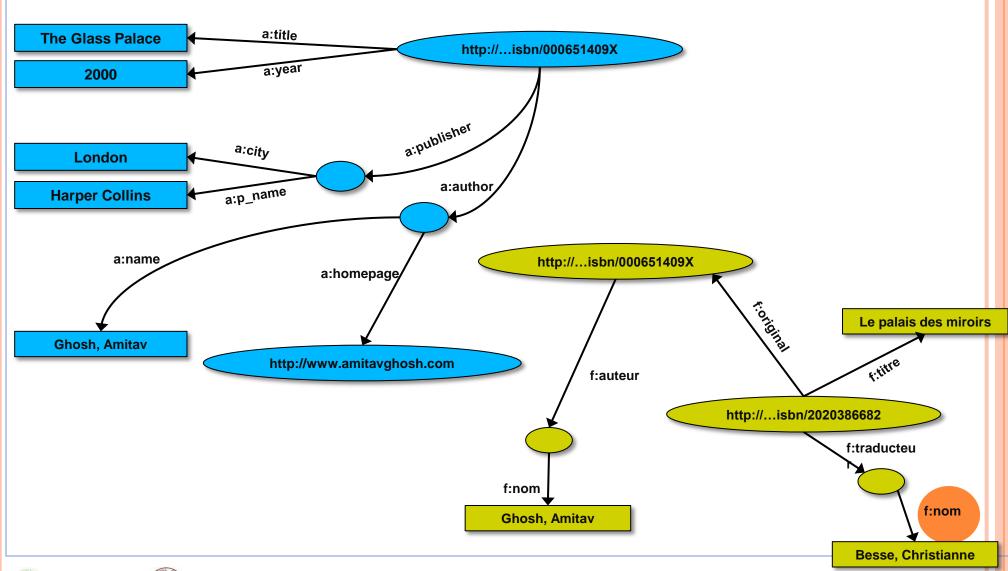


LINKED DATA SCENARIO 2ND: EXPORT YOUR SECOND SET OF DATA



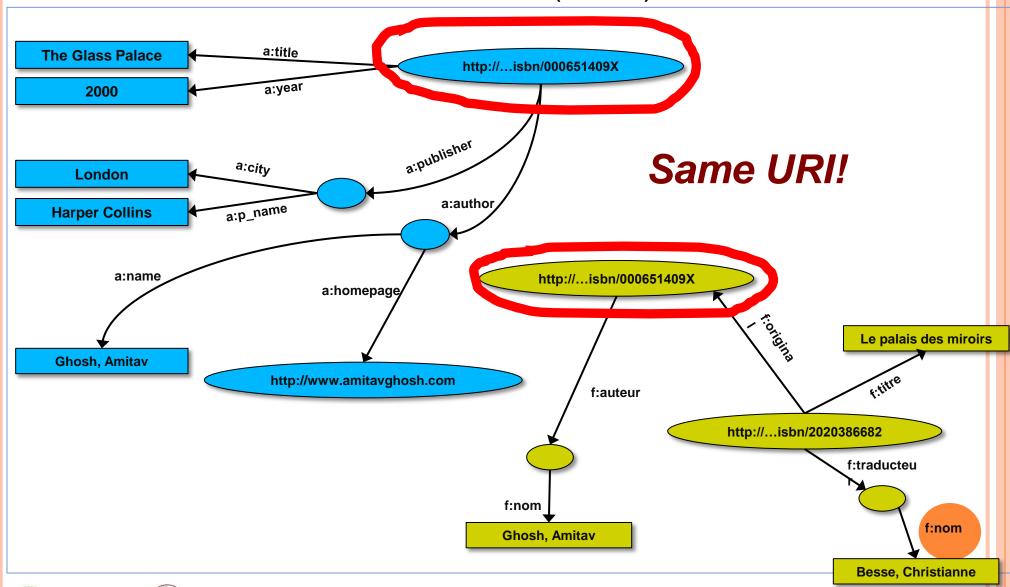


LINKED DATA SCENARIO 3RD: START MERGING YOUR DATA



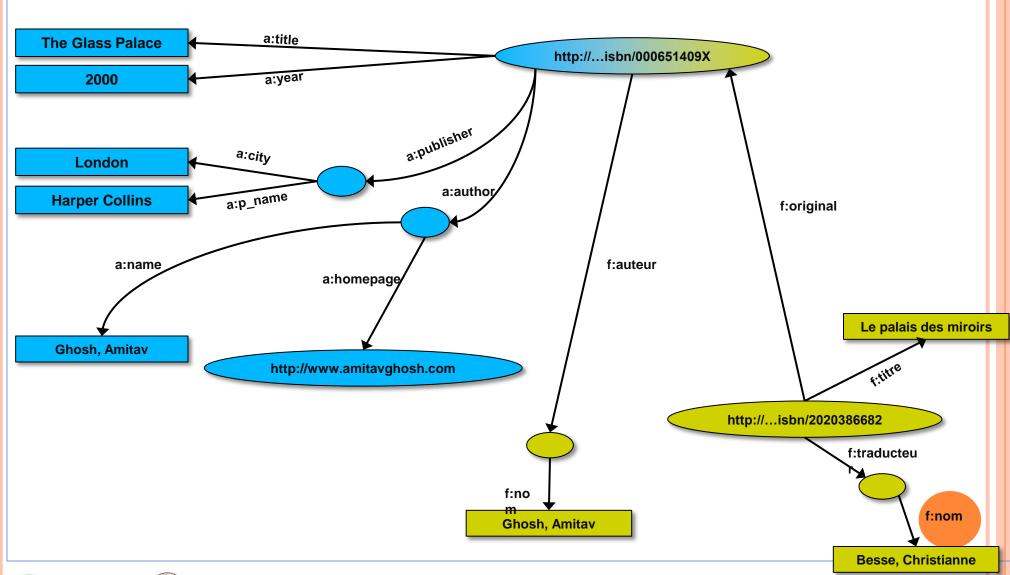


LINKED DATA SCENARIO 3RD: START MERGING YOUR DATA (CONT)





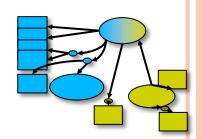
LINKED DATA SCENARIO 3RD: START MERGING YOUR DATA





LINKED DATA SCENARIO START MAKING QUERIES...

- User of data "F" can now ask <u>queries</u> like:
 - "give me the title of the original"
- This information is not in the dataset "F", but it can be retrieved by merging with dataset "A"

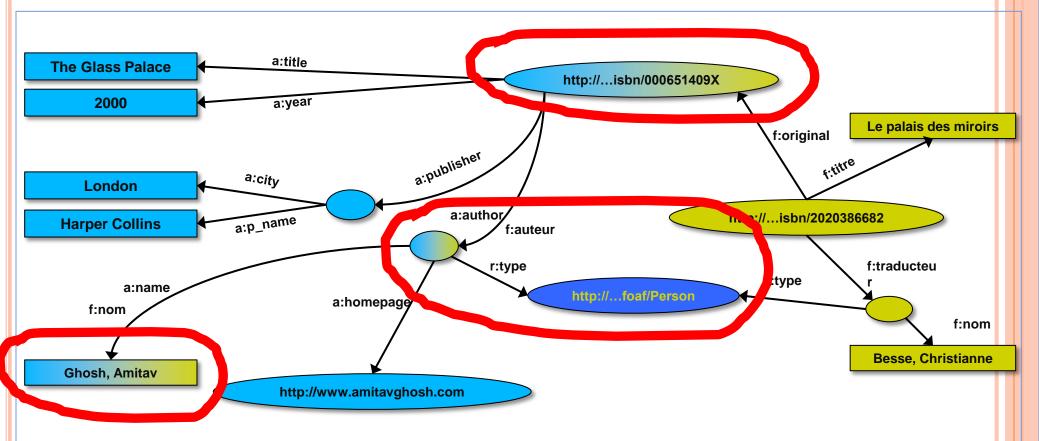


- We "feel" that a: author and f: auteur should be the same.
 But an automatic merge does not know that! We can add some extra information to the merged data:
 - a:author same as f:auteur
 - both identify a "Person" a term that a community may have already defined (a "Person" is uniquely identified by his/her name and, say, homepage (it can be used as a "category" for certain type of resources)





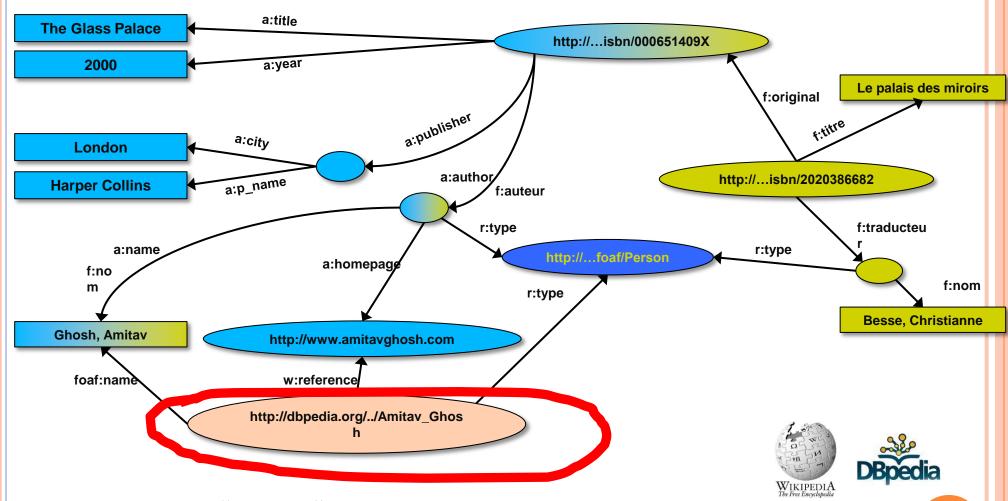
LINKED DATA SCENARIO 3RD REVISITED: USE THE EXTRA KNOWLEDGE



• User of dataset "F" can now query: "give me the home page of the original's 'auteur". The information is not in datasets "F" or "A", but was made available by merging datasets "A" and datasets "F" and adding three simple extra statements as an extra "glue"



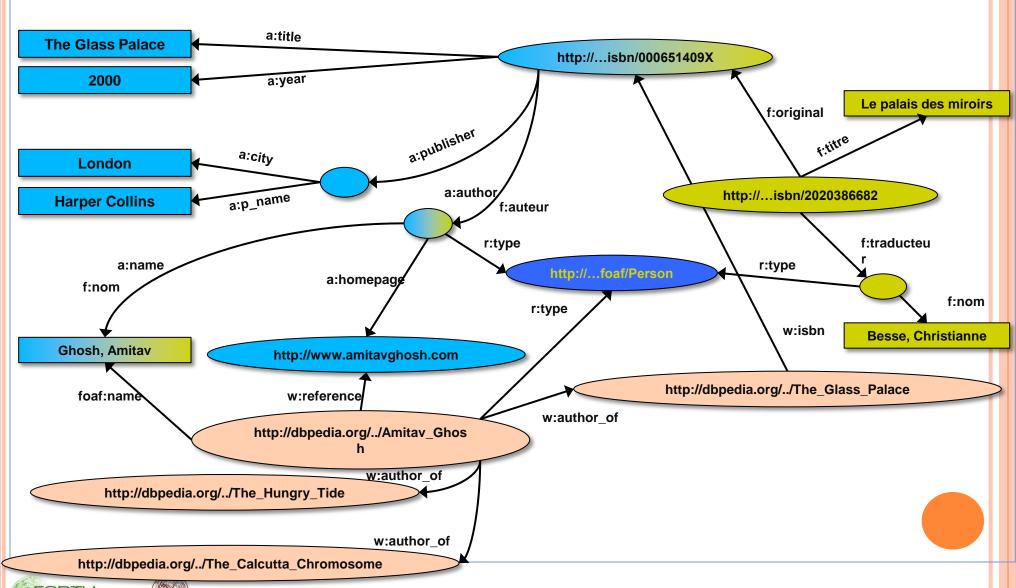
LINKED DATA SCENARIO MERGE WITH WIKIPEDIA DATA



Using, e.g., the "Person", the dataset can be combined with other sources. For example, data in Wikipedia can be extracted using dedicated tools

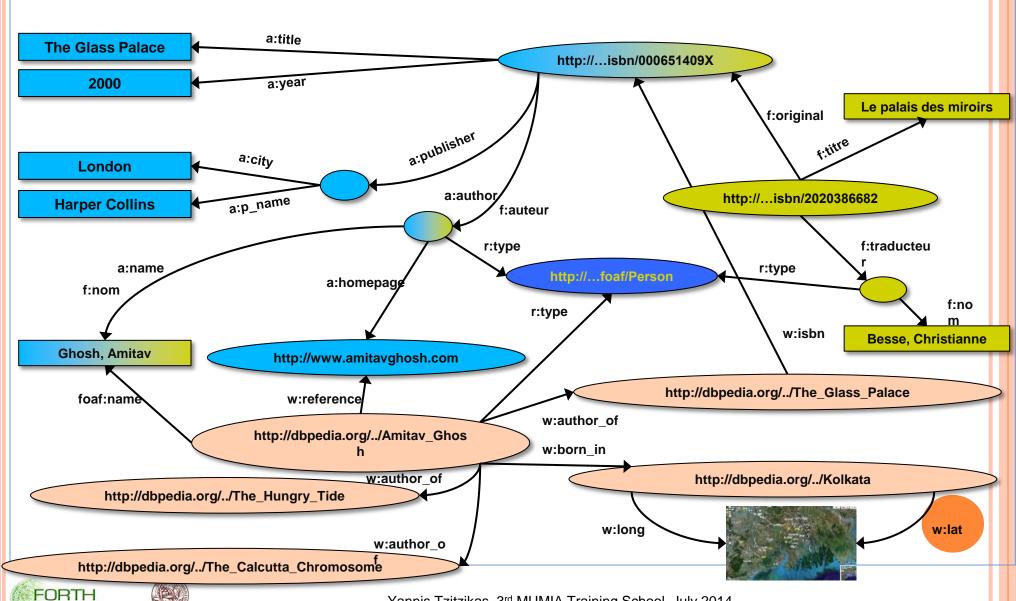


LINKED DATA SCENARIO MERGE WITH WIKIPEDIA DATA





LINKED DATA SCENARIO MERGE WITH WIKIPEDIA DATA



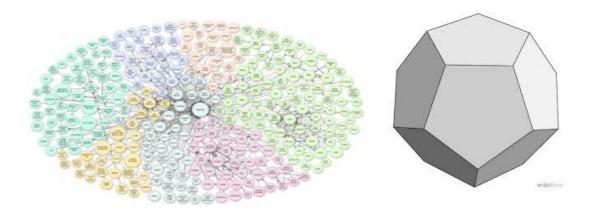
QUERIES AND THE DIFFICULTY OF QUERY FORMULATION

- The availability of Semantic Data allows formulating complex queries. E.g.:
 - scientists who have worked in an institute in Germany and are known for their work in fuzzy set theory
- however to formulate such queries one has to know not only SPARQL, but also the vocabulary used. E.g. (sketch):

```
SELECT ?x
WHERE {
    ?x type scientist .
    ?x ns:knownFor "fuzzySetTheory" .
    ?x ns:hasWorkedIn ?y .
    ?y type institution .
    ?y ns:locatedAt "Germany" .
}
```



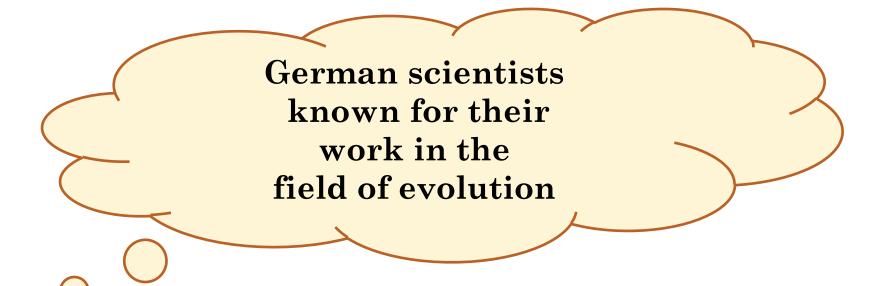
QUERYING THE SEMANTIC WEB / LOD



- Faceted exploration can allow users to satisfy their information needs without having to be aware of the **employed terminology**, **contents**, or **query language** of the sources
- An example from an application offering faceted exploration to DBpedia follows





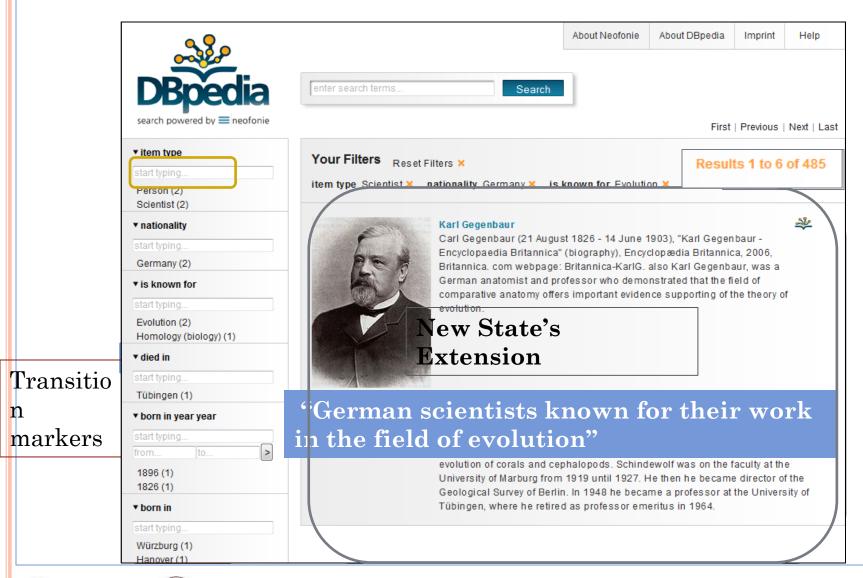








GUIDED EXPLORATION - AN EXAMPLE







SURVEYING FACETED EXPLORATION APPROACHES

We use the term BA to refer to a *browsing approach*. We can survey the works (BAs) that have been applied or proposed according to various aspects:

- 1. Characteristics of the underlying information space. The structuring of the underlying information base is an important aspect since each case requires tackling different difficulties.
- 2. Configuration. Some approaches can be applied without requiring any form of configuration or application design (regarding the browseable information space), while some others require configuration steps, e.g. specify the contents and structuring of the browsable part through the view-based approach over a DB or an RDF repository. Since the browsable part of the information source is defined by a query, its structure may be different from that of the original source.
- 3. State Space. In general we can view the interaction as a state space consisting of states and transitions, therefore we can characterize, or comparatively evaluate, two BAs by comparing their state spaces, e.g. by identifying properties which are satisfied by their state space.





SURVEYING FACETED EXPLORATION APPROACHES

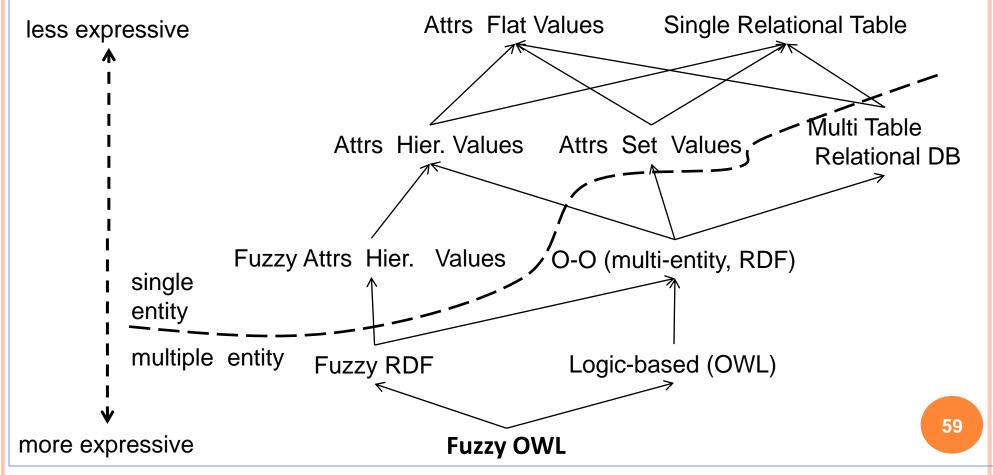
1. THE UNDERLYING INFORMATION SPACE

- Some browsing approaches are applicable to simple structures, while others to complex information structures (e.g. OWL-based KBs). There are several options, some of them follow:
 - attribute-value pairs with flat values (e.g. name=Yannis),
 - attribute-value pairs with hierarchically organized values (e.g. location=Crete),
 - **set-valued attributes** (either flat or hierarchical) (e.g. accessories={ABS, ESP}),
 - **multi-entity** (or object-oriented) (e.g. RDF, linked open data), and relational databases.
 - Furthermore, we could have **fuzziness** and we can consider this as an independent aspect (e.g. there are fuzzy extensions of the RDF model).
- Therefore one important aspect is how the underlying information is structured.



CONT.

Here we can see the above categories **organized hierarchically** where an option X is a (direct or indirect) child of an option Y if whatever information can be expressed in Y can also be expressed in X.

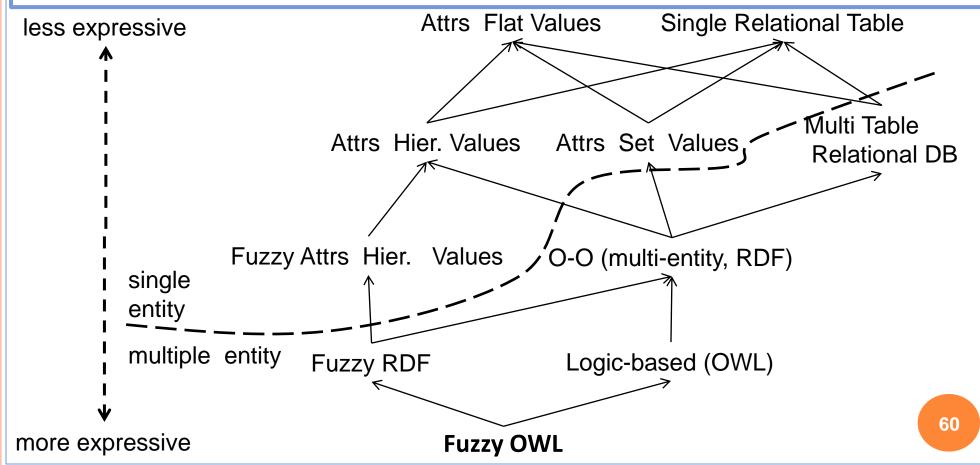






CONT.

The **value of this diagram** is that it depicts the fact that if a browsing approach is appropriate for an option X, then certainly it is appropriate for all options which are parents of X.

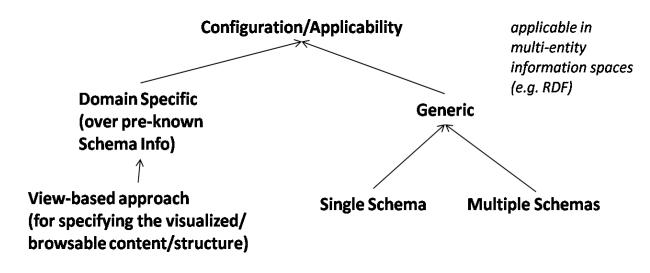




SURVEYING FACETED EXPLORATION APPROACHES

2. Configuration

- View-based approach
 - The structure of the browsable part has to be explicitly specified through configuration steps (*e.g.* by using a query language or logic rules)
- Generic approach
 - No configuration requirements w.r.t the underlying information space







SURVEYING FACETED EXPLORATION APPROACHES 3. STATE SPACE (STATES, TRANSITIONS AND TRANSITION MARKERS)

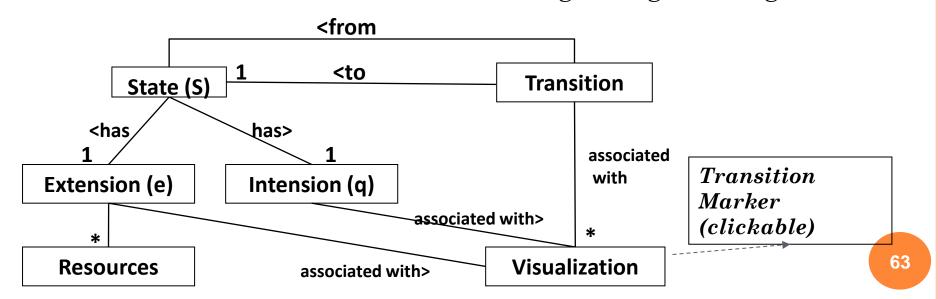
• State Space. In general we can view the interaction as a state space consisting of states and transitions, therefore we can characterize, or comparatively evaluate, two BAs by comparing their state spaces, e.g. by identifying properties which are satisfied by their state space.





CONT. STATES

- A **state** has an **extension** (a set of items displayed, e.g. WS results), an **intension** (i.e. condition or query; satisfied by the extension), probably a name, and a number of **transitions** each leading to a different state.
- In addition each state has a visualization format for its (a) extension, (b) intention, as well as (c) its transitions (e.g. a tree-control, a list, a table). In any case, each transition has a clickable **transition marker** signifying the existence of the transition. Usually these markers are enriched with information regarding the target state.





CONT. STATES

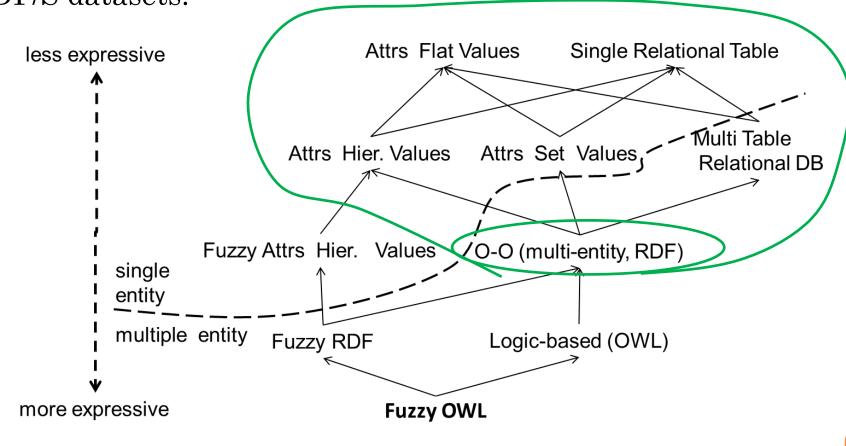






A "GENERIC" INTERACTION MODEL FOR EXPLORING RDF/S DATASETS

• Now we will see an interaction model for faceted exploration over RDF/S datasets.

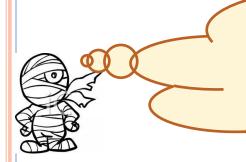




A GENERIC INTERACTION MODEL FOR EXPLORING RDF/S DATASETS

• Objective: Through simple clicks the user can reach states whose extension corresponds to the answer of complex queries.

• Example:



Japanese cars for sale which are driven by persons who work at FORTH and know a person who knows Bob

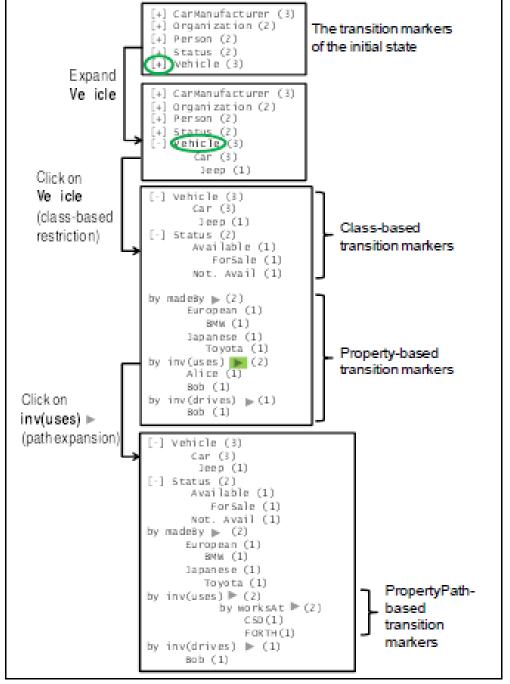
- Source: It is a simplified version of the model for browsing Fuzzy RDF that is described in
 - Nikos Manolis, Yannis Tzitzikas: Interactive Exploration of Fuzzy RDF Knowledge Bases. ESWC (1) 2011: 1-16





SUPPORTED KINDS OF TRANSITIONS

- Supported transitions
- Class-based
- Property-based
- Property-path based
- Entity Type Switch



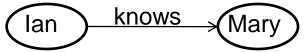




NOTATIONS (FOR RDF/S)

• A RDF/S KB is defined by a set of **RDF triples** of the form:

(subject, predicate, object)

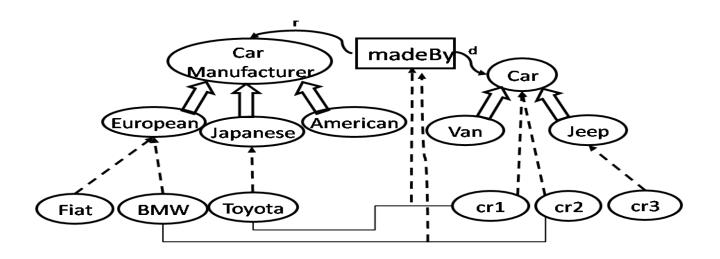


- The *closure* C(K) of a KB K contains all the triples explicitly asserted or <u>inferred</u> (based on the RDF/S semantics) from a KB
- The **Schema** of a RDF/S KB K is a 6-tuple $\Gamma = < C$, Pr, domain, range, \leq^*_{cl} , $\leq^*_{pr} > C$
 - C: Classes, Pr: Propertis, domain and range of properties, subclassOf (among C), subPropertyOf (among Pr)
- *Instance notations* for a KB:
 - Instances of a class $c \in C$: inst(c) = { $\mathbf{o} \mid (\mathbf{o}, \text{type}, \mathbf{c}) \in C(K) }$
 - Class instance triples: (o type c)
 - Instance of a property $p \in Pr$: $inst(p) = \{ (o, p, o') \mid (o, p, o') \in C(K) \}$
 - Property instance triples: (o p o')





CONT.



- inst(Jeep)={cr3}
- inst(Car)={cr1, cr2, cr3}
- Inst((Car,madeBy,CarManufacturer)) = {(cr1, madeBy,Toyota)(cr2, madeBy,BMW)





AUXILIARY DEFINITIONS: RESTRICTIONS AND JOINS

Assuming an initial state we need to tackle the following:

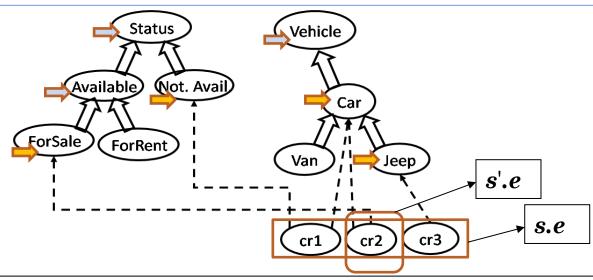
- * How the **transitions markers** (tms) available to that state are computed
- * How the **new state's extension** is computed after selecting a *tm*
- To define the above we need to define the notion of **restriction** and **join** $\mathbf{p} \in \mathrm{PR} \cup \mathrm{PR}^{-1}$, **E**: set of resources, **vset**: a set of resources or literals
- Restrictions: Given a set E
 - 1. $Restrict (E, p : v) = \{ e \in E \mid (e, p, v) \in inst(p) \}$
 - 2. Restrict (E, p : vset) = { $e \in E \mid v' \in vset \text{ and } (e, p, v') \in inst(p)}$
 - 3. Restrict (E, c) = { $e \in E \mid e \in inst(c)$ }
- Joining values: Compute values which are linked with the elements of E
 - $Joins (E, p) = \{ v \mid \exists e \in E \text{ and } (e, p, v) \in inst(p) \}$







CLASS-BASED TRANSITIONS



Candidate tms: $TM_{cl}(s.e) = \{ c \in C \mid Restrict (s.e, c) \neq \emptyset \}$

Clicking on a $c \in TM_{cl}(s.e)$ then s'.e = Restrict (s.e, c)

[-]Vehicle(3)

Car(3)

Jeep(1)

[-] Status(2)

Available(1)

ForSale(1)

Not Available(1)



[-] Vehicle(1)

Car(1)

[-] Status(1)

Available(1)

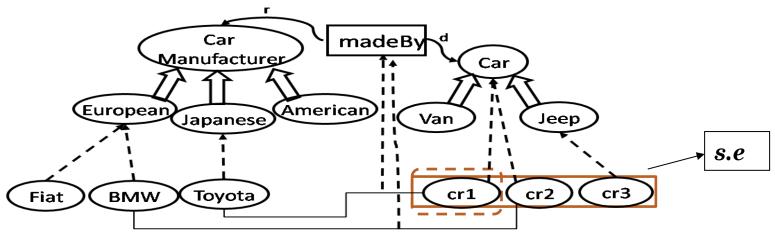
ForSale(1)





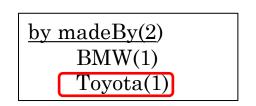
PROPERTY-BASED TRANSITIONS

• The goal: Restrict the current state's extension according to a property value **v** (resource or literal). E.g. All cars made by Toyota



• Candidate properties: Props(s) = { $p \in Pr \cup PR^{-1} \mid Joins(s.e, p) \neq \emptyset$ }

Candidate tms for $a p \in Props(s) : Joins(s.e, p) = \{v \mid e \in s.e \ and \ (e, p, v) \in inst(p)\}$



OR

by madeBy(2)

European

BMW(1)

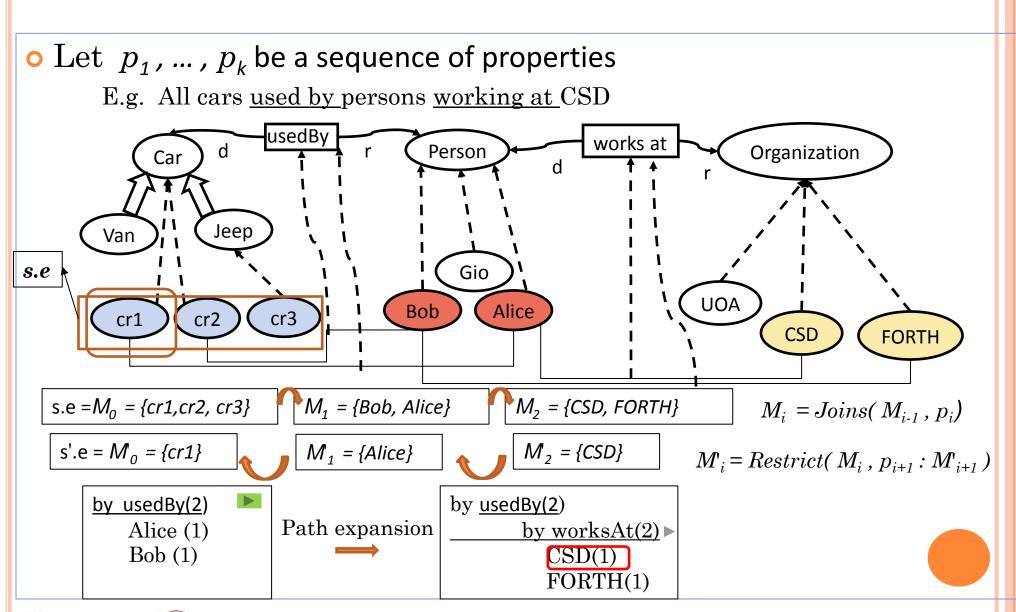
Japanese (1)

Toyota(1)





PROPERTY PATH-BASED TRANSITIONS

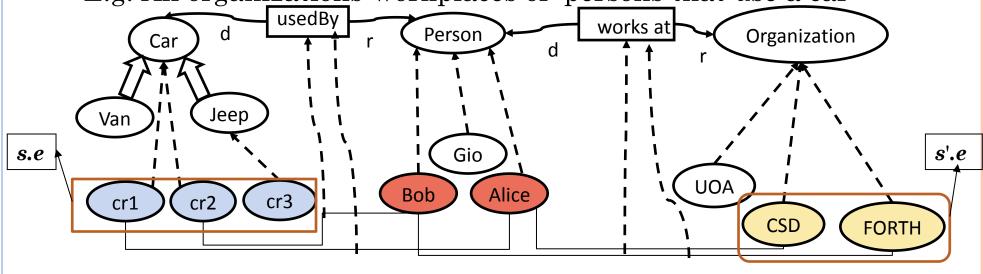




ENTITY TYPE SWITCH TRANSITIONS

Allow users to move to a state whose extension is the current set of tms

E.g. All organizations-workplaces of persons that use a car



s.e =
$$M_0$$
 = {cr1,cr2, cr3}

$$M_1 = \{Bob, Alice\}$$

 $M_2 = \{CSD, FORTH\}$

s'.e = M_2 = {CSD, FORTH}

by inv(uses)(2)

all

Alice (1)

Bob (1)

Path expansion

by worksAt(2)

all

CSD(1)

FORTH(1)

by inv(uses)(2)





APPLICABILITY

- To apply it, one has to implement Restrict and Join using the technology used for the underlying dataset
 - (e.g. over SPARQL, over SQL, over WhatEverQL, ...)
- For more see the related publication
 - It also captures the fuzzy aspect





Synopsis of the first Part



- A significant percentage of information needs are recall-oriented and this justifies the need for exploratory search services
- Faceted Search/Exploration is an effective model for exploratory search (now de facto standard)
- There are many variation of FE according to: structural complexity of the information space, configurability, supported state space
- We have seen a model for FE that captures the majority of datasets that exist





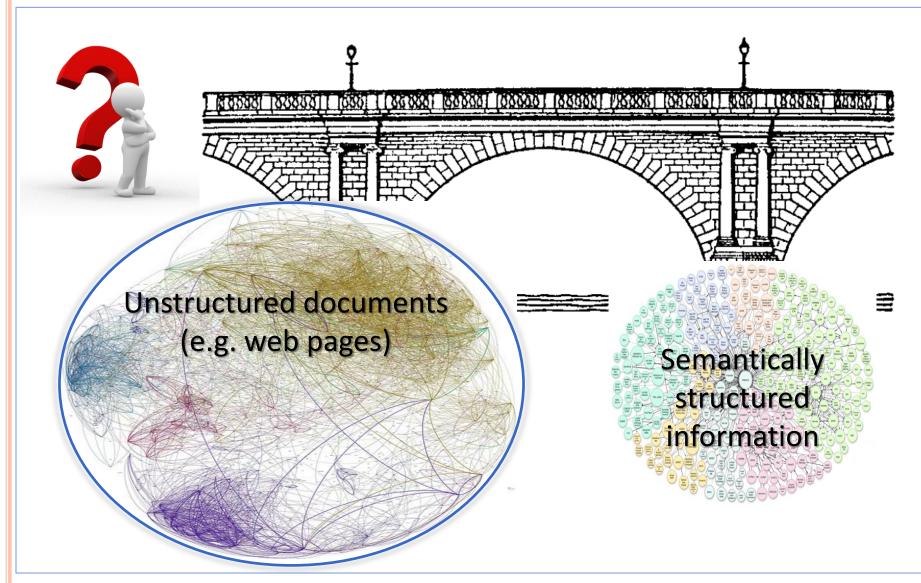
3. BRIDGING THE WEB OF DOCUMENTS WITH THE WEB OF DATA AT SEARCH TIME

Possible Ways
Focus on doing this at Search Time





AN INTERESTING QUESTION



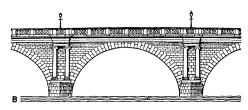




This question is instance of a Bigger Question

 How to integrate the results and/or foster collaboration between different communities







AI Community











Bridging the Web of Documents with the Web of Data

- Why
 - Both exist (why not exploit both?)
- When and Where
 - There is plethora of options. Let's consider the direction: How the searching over the Web of Documents can benefit from the existence of the Web of Data. A few options follow:
 - At indexing time
 - E.g. for the disambiguation of words
 - At query formulation time
 - E.g. autocompletion/query expansion/term_recommendation services that exploit the web of data, the various vocabularies, ontologies, etc.
 - At query evaluation time
 - Exploitation of how words/concepts are connected in the ranking formula
 - After query evaluation
 - For semantically post-processing of search results
 - o ...





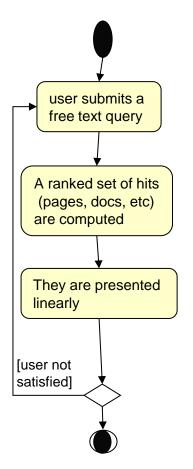
OUR FOCUS IN THIS TALK

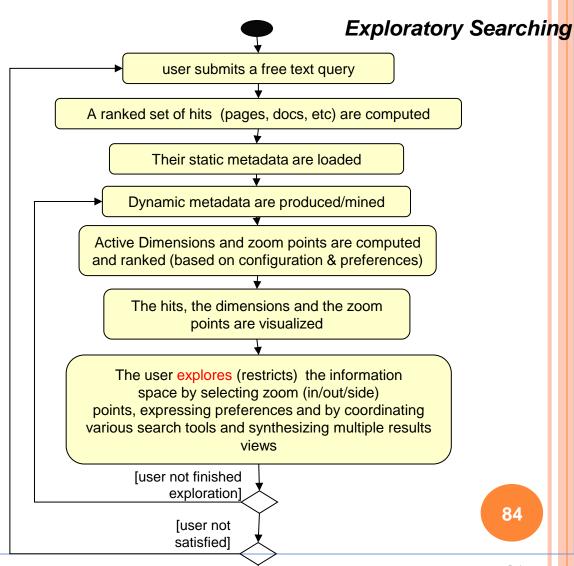
- Don't change the way users search for information (keyword queries in Web Search Engines).
- Try to exploit LOD to offer value-adding services



A Process for Exploratory Search

Web searching today



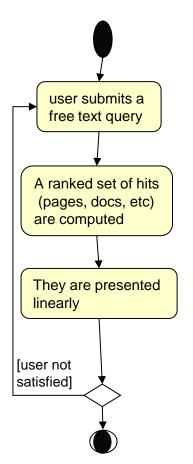


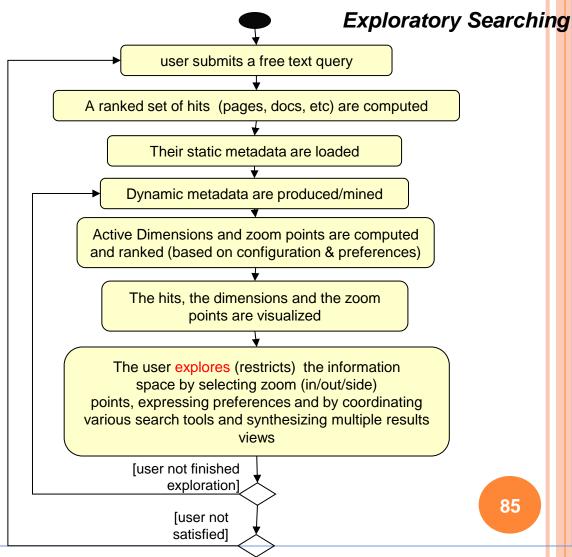




A Process for Exploratory Search

Web searching today







(2)

BENEFITS (OF THIS PROCESS)

- Does not change the way users search for information
- Can be applied over existing search systems/interfaces
- Provides overviews of the results (not only of the top hits)
- By clicking on a facet term (metadata value, mined entity, etc) the user can see the related hits even if they are low ranked
- Allows restricting the answer gradually

Next we will see various applications of this process (these applications concern Web Search, Exploratory Search and LOD)





4. CASE STUDIES



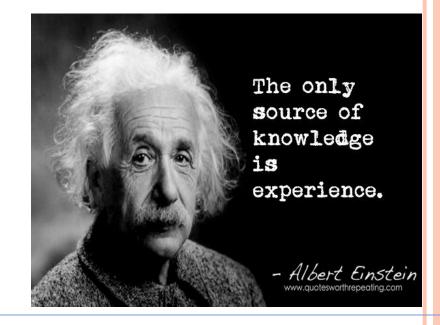


CASE STUDIES OUTLINE

• As Case Studies I will present you a kind of "story" organized in 9 milestones



- These 9 milestones correspond to activities of ISL (Information Systems Laboratory) of FORTH-ICS.
 - They correspond to the period 2009-2014
 - They are presented in (almost) chronological order.







MILESTONE 1. THE MITOS WSE (2009)



• MITOS is WSE build from scratch and collaboratively with the students of the Computer Science Department of the University of Crete.

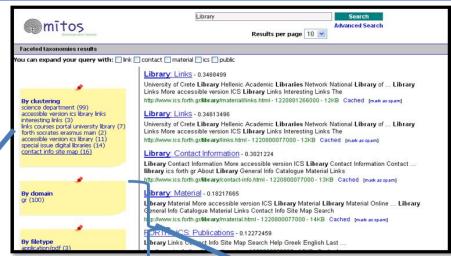


- Apart from the classical WSE functionality, Mitos offers <u>faceted search over the results</u> of the submitted queries.
 - It supports facets corresponding to <u>metadata attributes</u> of the web pages (static metadata), as well as facets corresponding to the outcome of <u>snippet-based clustering algorithms</u> (a kind of dynamic metadata).
 - The user can then restrict his/her focus <u>gradually</u>, by interacting with the resulting multidimensional structure through <u>simple clicks</u>.

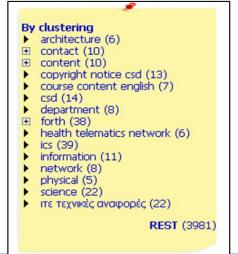




THE MITOS WSE (2009)

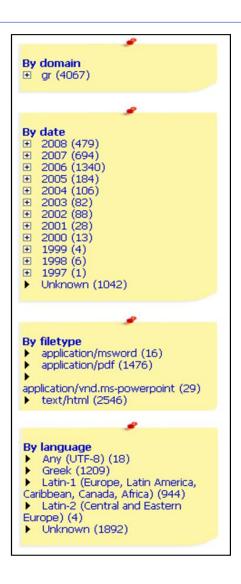


Dimension based on dynamic (query-dependent) metadata (of the top ranked hits)



Dimensions based on static metadata

N. Manolis and Y. Tzitzikas (ESWC'11)



90

Tzitzikas,

Panel@ExploreDB,

Athens



information systems laboratory

Search

Results per page | 10 ▼

Advanced Search

A user wants to get information about Information **Systems** Laboratory

Faceted taxonomies with on-demand clustering results

8558 Results 1 - 10 from 8558 for information

By clustering

- activities (39)
- biomedical informatics laboratory (10)

mītos

- decision support systems (2)
- dimitris (7)
- events (43)
- forth (25)
- history (35)
- ics (85)
- informatics (64)
- laboratories (2)
- projects (53)
- publications (47)
- seminars (26) support (5)
- yannis (3)

By domain

2009 (606) 2008 (668) 2007 (3113) **REST** (8458)

Information Systems Laboratory: - 0.916549

information systems laboratory ics isl img isl src images buttons isl ... http://www.ics.forth.gr/isl/publications/by_name.jsp?Person_ID=7 - 0 - 8KB Cached - Similar pages [mark as spam]

FORTH - ICS: Information Systems Laboratory - 0.8703994

information systems information retrieval systems database workflow management systems semantically rich ... forth ics information systems laboratory information systems laboratory head laboratory prof

http://www.ics.forth.gr/isl/index.html - 1173087253000 - 17KB Cached - Similar pages [mark as spam]

Information Systems Laboratory: - 0.8702642

information systems laboratory ics isl panos constantopoulos muse multimedia accessible version ... ics isl isl centre cultural informatics history events activities projects publications

http://www.ics.forth.gr/isl/publications/by_year_jsp?Year_of_publication=1987 - 0 -17KB Cached - Similar pages [mark as spam]

Information Systems Laboratory: - 0.8692224

information systems laboratory null ics isl accessible version ics isl isl ... centre cultural informatics history events activities projects publications seminars people links

We can focus on "By date"

publication=null - 0 -

facet, clicking the "2009" label.

document retrieval multos ... accessible version ics isl isl centre cultural informatics history events activities

http://www.ics.forth.gr/isl/publications/by_year.jsp?Year_of_publication=1986 - 0 -

8558 initial results

2014 Y. Tzitzikas, Panel@ExploreDB, Athens



(CONT.)

By domain gr (606)

By date

2009 (606) June (71)

May (80)

April (212)



information systems laboratory

Search

Advanced Search



Results per page | 10 ▼

Faceted taxonomies with on-demand clustering results

606 Results 1 - 10 from 606 for information systems laboratory. (15729 ms)

By clustering athanasios mouchtaris (3) communication (15) distributed (13) dynamic (13) forth (11) home page (3) networks (13) news (5) oikonomou (2) page (5) presentation (3) publications (6) spring (3) tziritas (2) πε (3) **REST** (558)

Information Systems Laboratory: Seminars - 0.28782406

challenge succeed transition traditional information systems information retrieval systems database workflow ... management systems semantically rich large scale adaptive information systems systems characterized http://www.ics.forth.gr/isl/services.html - 1244639664000 - 21KB Cached -Similar pages [mark as spam]

We can further limit the results, by selecting one of the clusters (they were recomputed for the new focus)

atory: Seminars - 0.276 eminars seminars ics isl in matics subjects developed including greek

ml - 1244123830000 - 16KB

Il Systems - 0.19771457

The results of the selected group are loaded in the results' panel and all facets are updated.

ems CS 463 Information Retrieval aching Material Lectures and Program Links

/grades.html - 1241012788000 - 2KB

Cached - Similar pages [mark as spam]

CS-463 Information Retrieval Systems - 0.18768528

CS 463 Information Retrieval Systems CS 463 Information Retrieval Systems ... Course Information Teaching Material Lectures and Program Exercises and Assignments Grades

http://www.csd.uoc.gr/~hy463/2007/en/announcements.html - 1241012778000 -2KB Cached - Similar pages [mark as spam]

CS-463 Information Retrieval Systems - 0.18636355

CS 463 Information Retrieval Systems CS 463 Information Retrieval Systems Spring ... Information Teaching Material Lectures and Program

(CONT.)



information systems laboratory

Search

Advanced Search

Results per page 10 ▼



Faceted taxonomies with on-demand clustering results

5 Results 1 - 5 from 5 for information systems laboratory. (86 ms)

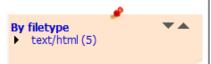


By date ▼▲

∃ 2009 (5) ⊞ June (2)

⊞ May (2)

∃ January (1)



By language

Latin-1 (Europe, Latin America, Caribbean, Canada, Africa) (5)

FORTH - ICS: News - 0.053278793

information greek information greek

http://www.ics.forth.gr/news/news-prev.html - 1241420849000 - 54KB Cached - Similar pages [mark as spam]

FORTH - ICS: News - 0.040669773

laboratories publications services library links contact info site map search help ...

http://www.ics.forth.gr/news.html - 1244102246000 - 23KB Cached - Similar pages [mark as spam]

FORTH - ICS: Welcome Note by the Director of ICS-FORTH - 0.028257346

zoomin ics announcements news press releases **laboratories** publications services library links ...

http://www.ics.forth.gr - 1232022089000 - 19KB Cached - Similar pages [mark as spam]

FORTH - ICS: News - 0.02504004

technical aspects multimodal **systems** tams department informatics university hamburg germany university ... **information** science university pennsylvania professor head cognitive informatics **laboratory** laval university http://www.ics.forth.gr/news/lectures-prev.html - 1244102254000 - 100KB Cached - Similar pages [mark as spam]

FORTH - ICS: Lectures - 0.01583067

announcements news **laboratories** publications services library links contact info site map ...

http://www.ics.forth.gr/news/ian_cernocky_lecture.htm! - 1241420696000 - 20KB

With only 2 clicks, we have limited the results to 5 hits.





CONT.

- Evaluation with Users (main results):
 - Faceted search, combining dynamically and statically mined metadata
 - olead to much <u>improved task completeness</u> with much less user interactions
 - owas more preferred by the users (advanced and plain ones) and lead to greater satisfaction, than plain clustering or faceted interfaces
- Most Important Related Publications
 - [ECDL'09] P. Papadakos, S. Kopidaki, N. Armenatzoglou and Y. Tzitzikas. Exploratory Web Searching with Dynamic Taxonomies and Results Clustering. In ECDL 2009
 - [WISE'09] S. Kopidaki, P. Papadakos, and Y. Tzitzikas. STC+ and NM-STC: Two novel online results clustering methods for web searching. In WISE 2009
 - [J. KAIS 2012] P.Papadakos, S.Kopidaki, Nikos Armenatzoglou and Y. Tzitzikas On exploiting Static and Dynamically mined Metadata for Exploratory Web Searching, KAIS Journal, 2012





MILESTONE 2. DURING TYPING?

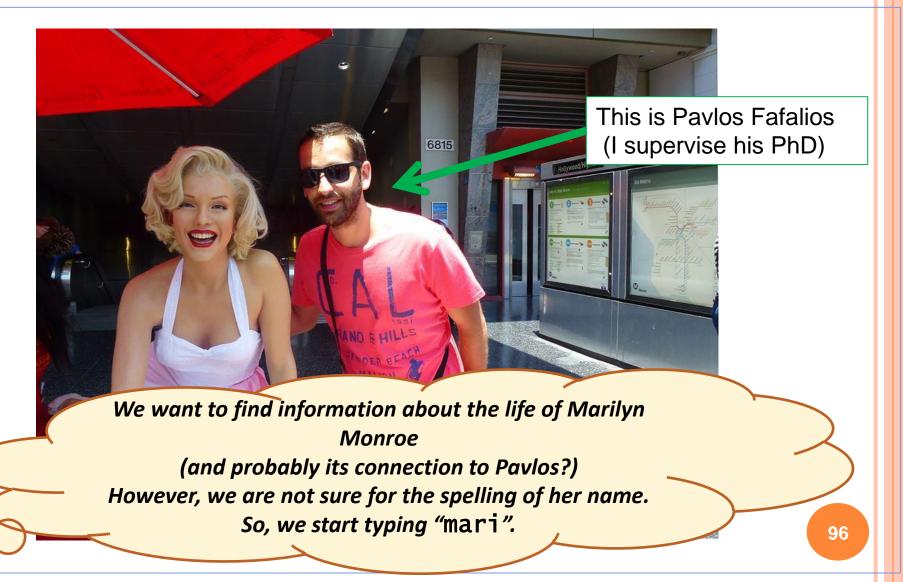


- Then we questioned ourselves:
 - why not offering this functionality during query typing, i.e. a kind of richer autocompletion service?
- This resulted to what we called **Instant Overview Search** (IOS).
- The idea:
 - For the frequent queries, pre-compute and store not only the first page of results, but also the analysis of these hits
- Technical challenge
 - Since the amount of information that has to be stored for each query is higher (and obviously does not fit in main memory) we devised a partitioned trie-like index for efficiency (plus a dedicated cache)





IOS (INSTANCE OVERVIEW SEARCH), 2011-2012







(CONT.)

home page . visit uoc-csd . visit ics-forth

» clusters overview:

- marilyn(99)
 - monroe(33)
 - quotes(4)
 - biography(3)
 - manson(8)
 - home(8)
 - free(7)
 - encyclopedia(3)
 - quotes(5)
 - photo(5)
 - collection(5)
 - news(5)
 - encyclopedia(4)
 - biography(4)
 - inew(10)
 - york(4)
 - gavin rossdale(3)
 - images(5)
 - andy(3)
 - online(3)



marilyn
marilyn monroe
mario games

• SET (default)

• PET
• STIE
• PTIE

search

search

ch engine

» over linked open data

List of query's suggestions.

We can continue typing

the query. Instantly new

suggestions are shown

» first page results overview:

First page of results of the top suggestion "marilyn"

Marilyn (singer) - Wikipedia, the free encyclopedia - 0

Peter Robinson (born 3 November 1962), better known as **Marilyn**, is a British pop singer who achieved international fa me in the 1980s with his hit song ...Early life - Blitz years - Career - Recent activity

Manthan Managa - Millian dia da fan anna dia dia

Marilyn Monroe - Wikipedia, the free encyclopedia - 1

Marilyn Monroe born Norma Jeane Mortenson, but baptized Norma Jea Miller - Somethings Got to Give - Some Like It Hot

en.wikipedia.org/wiki/Marilyn_Monroe

en.wikipedia.org/wiki/Marilyn (singer)

Marilyn (hill) - Wikipedia, the free encyclopedia - 2

A **Marilyn** is a mountain or hill in the United Kingdom, Ireland or Isle of ... en.wikipedia.org/wiki/Marilyn_(hill)

Cluster Label Tree of the top suggestion "marilyn"

Agency - 3

rmodel agency in Paris. Marilyn Agency est la plus importante agence de mannequin Γ





(CONT.)

home page . visit uoc-csd . visit ics-forth

» clusters overview:

- marilyn monroe(100)
 - quotes(8)
 - news(10)
 - online(5)
 - photos(7)
 - collection(6)
 - images(7)
 - death(6)
 - gallery(5)
 - photo(6)
 - video(4)
 - free(3)
 - pictures(5)
 - encyclopedia(3)

 - biography(5)





» first page results overview:

Marilyn Monroe Biography from Who2.com

Marilyn Monroe s sex appeal talent and untimely death combined to make her an enduring star and one of Hollywood s most recognizable icons. Early in.

marilyn monroe

We selected the suggestion

"marilyn monroe".

The results' first page and

cluster label tree for this

suggestion were loaded

immediately.

search

http://www.who2.com/marilynmonroe.html

Marilyn Monroe: Biography from Answers.com

Marilyn Monroe Actor Born: 1 June 1926 Birthplace: Los Angeles California Died: 4 August 1962 (drug overdose) Best Known As: Hollywood s most.

http://www.answers.com/topic/marilyn-monroe

The Marilyn Pages-Marilyn Monroe biography and images

life of Marilyn Monroe Norma Jean. ... Marilyn Monroe. The Marilyn Pages have moved to ellensplace.net/marilyn.ht (If you are not taken to the new ...

//www.ionet.net/~jellenc/marilyn.html

Marilyn Pages-Marilyn Monroe biography and images

life of Marilyn Monroe Norma Jean. ... NOTE FOR AOL USERS 🏚 Site Awards for The Marilyn Pages 🏟 to ellen s

http://www.ellensplace.net/marilyn.html

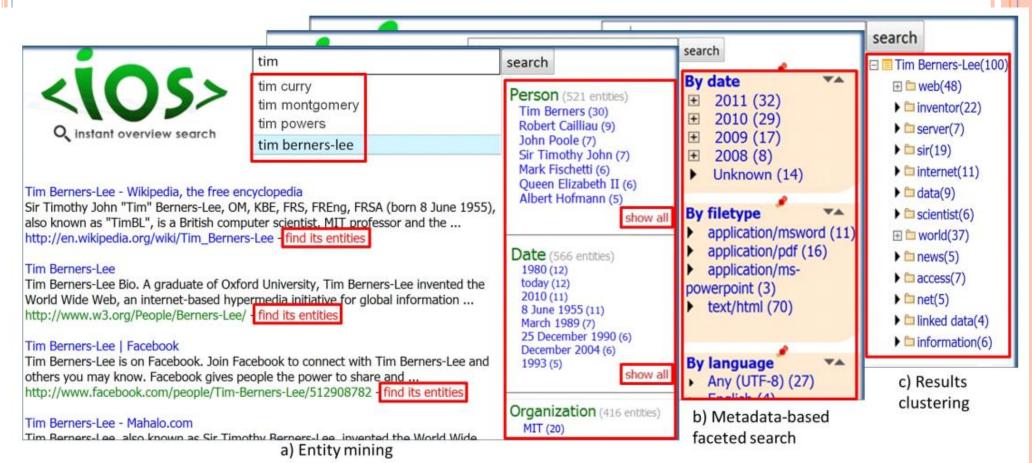
By clicking a label, the results of the specific cluster are loaded in the results panel.





IOS (INSTANCE OVERVIEW SEARCH), 2011-2012

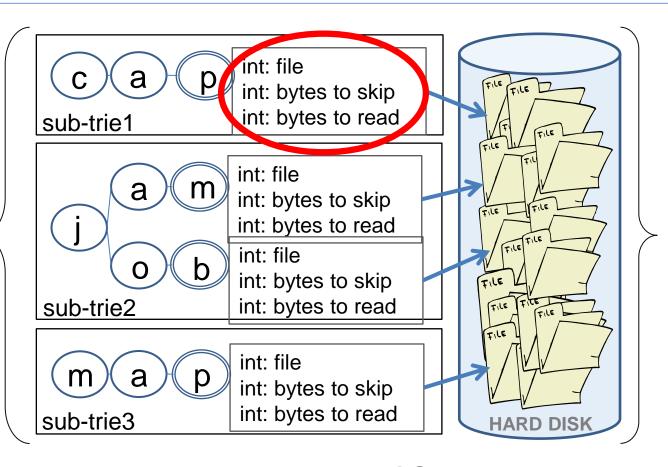
We can exploit this technique <u>for any king of pre-processing of search</u> <u>results</u> (e.g. metadata-based faceted search, snippet-based clustering, entity mining, etc)





IOS INDEXES

Hosted in main memory (based on < requests and space)



Precomputed information (always in hard disk)

Average Retrieval Time ≈ 135 ms

Experiments over a server running on a **modest personal computer**, with a synthetic query log of **1 million** distinct queries and synthetic precomputed information of **1 Terabyte**





CONT

Key results

- A <u>partitioned trie-based index</u> structure that can efficiently support recommendations for millions of distinct queries even with modest hardware
 - One can provide instant access to large amount of data, utilizing the existing resources, without requiring more hardware
- A hybrid <u>caching policy</u> (70% static and 30% dynamic) seems to be the more appropriate choice yielding a throughput increment of around 80% and a 25% speedup

Demo

- http://www.ics.forth.gr/isl/ios
 - Select the system "Instant Entity Mining + Clustering (over Bing)"

• Related Publications

- [WISE'11] P. Fafalios and Y. Tzitzikas, Exploiting Available Memory and Disk for Scalable Instant Overview Search, 12th International Conference on Web Information System Engineering (WISE 2011), Sydney, Australia, October 2011
- [WWW'12] P. Fafalios, I. Kitsos and Y. Tzitzikas, Scalable, Flexible and Generic Instant Overview Search, 21st International Conference on World Wide Web, (WWW 2012), Demo Paper, Lyon, France, April 2012





MILESTONE 3. ENTITY MINING AND LOD?

- Then we questioned ourselves:
 - why not exploiting LOD in the context of entity mining of the search results?
- Motivation
 - LOD contains plenty of information about Named Entities (their names, attributes, relationships with other entities, etc)
- Output
 - IOS Entity Mining
 - LOD is used as source for Named Entity Recognition
 - LOD is used for providing more information about the identified entities





http://www.ics.forth.gr/isl/ios







barack obama

Search

100 - results to mine

mine only snippets

Results of selected entities: reset

Barack Obama

BarackObama.com is the official re-election campaign website of PresidentBarack Obama. Visit the site for the latest updates from the Obama campaign, ...

http://www.barackobama.com/ - find its entities

About Barack Obama — Barack Obama

Barack Obama is the 44th President of the United States of America. PresidentObama speaking. President Obama was born in Hawaii on August 4th 1961 to a

Exploitation for restricting the focus

entities

Record - Barack Obama my was losing more than d quickly to pass the

American Recovery ...

http://www.barackobama.com/record - find its entities

News for barack+obamaBarack Obama - Wikipedia, the free encyclopedia Barack Hussein Obama II is the 44th and current President of the United

Person (1427 entities) Barack Obama (16) 4 Michelle Obama (19) George W. Bush (16) Ann Dunham (15) Craig Pobinson (15) Joe Biden (13) <</p> 🗖 John McCain (8) 🔩 Kennedy (9) Sarkozy (8) 4 Clinton (6) < show all

Organization (842 entities)

Harvard (14)

■ White House (22) <</p>

Congress (14)

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Yannis Tzitzikas, 3rd MUMIA Training School, July 2014

FORTH

CONT.

• Some results

- Real-time NEM <u>over snippets</u> is feasible and yields about 1.2 entities per snippet
- NEM <u>over contents</u> is more time consuming, but mines much more entities
- String similarity between the query and the entity name does <u>not</u> improve entity ranking (in our setting)
- The top-10 entities derived from snippet mining are quite different from those derived from contents mining (< 30% Jaccard similarity)

Related Publications

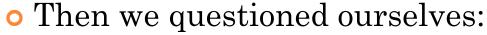
• [IRFC'12] P. Fafalios, I. Kitsos, Y. Marketakis, C. Baldassarre, M. Salampasis and Y. Tzitzikas, Web Searching with Entity Mining at Query Time, 5th Information Retrieval Facility Conference (IRFC 2012), Vienna, Austria, July 2012







MILESTONE 4. CONFIGURABILITY (AND LOD)



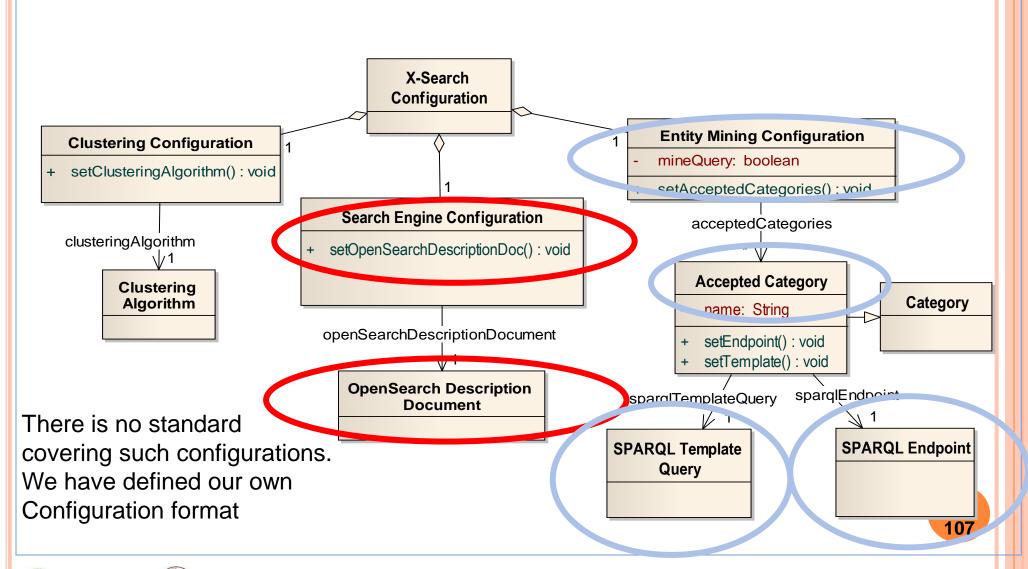


- why not allowing the user to configure himself the entities of interest by exploiting LOD (again in the context of entity mining of the search results)?
- Outcome
 - X-ENS (eXplore ENtities in Search)
- Related Publications
 - [SIGIR'13] P. Fafalios and Y. Tzitzikas, X-ENS: Semantic Enrichment of Web Search Results at Real-Time, 36th International ACM SIGIR Conference, Demo Paper, Dublin, Ireland, 28 July - 1 August 2013





XSEARCH-CONFIGURABILITY: THE CONCEPTUAL MODEL







top tennis players

about x-ENS | admin configuration

Search

200 ▼ results to mine

Tennis Player (39 entities)

Roger Federer (14) 4

Rafael Nadal (7) <

Novak Djokovic (5) ◀

Andy Roddick (4) 4

serena williams (4)

Maria Sharapova (3)

Andy Murray (2) <

Tsvetana Pironkova (2

Urszula Radwanska (2

Vania King (2) **4**

Tennis - ATP World Tour - Home

... photos, video, behind-the-scenes footage to hits sive tennis player and tennis tournament statis to hits ic opens with ...

http://www.atpworldtour.com/ - find its entities

Semantic Entity Enrichment (close)

Properties of: Andy Roddick

Description

Andrew Stephen "Andy" Roddick (b orn Aug 30, 1982) is an American p rofessional tennis player and a for mer World No. 1. He is...

Depiction



ite Rulebook; Careers; ATP Ac

TP World Tour 2013 Season .

les.aspx - find its entities

ree encyclopedia

players, both past and prese it includes only players who ha

nnis_players - find its entities

free encyclopedia

Country (11 antition) India (8) Canada (:

BirthPlace

Omaha, Nebraska

BirthDate 1982-08-30

MILESTONE 5. PROFESSIONAL SEARCH SYSTEMS?

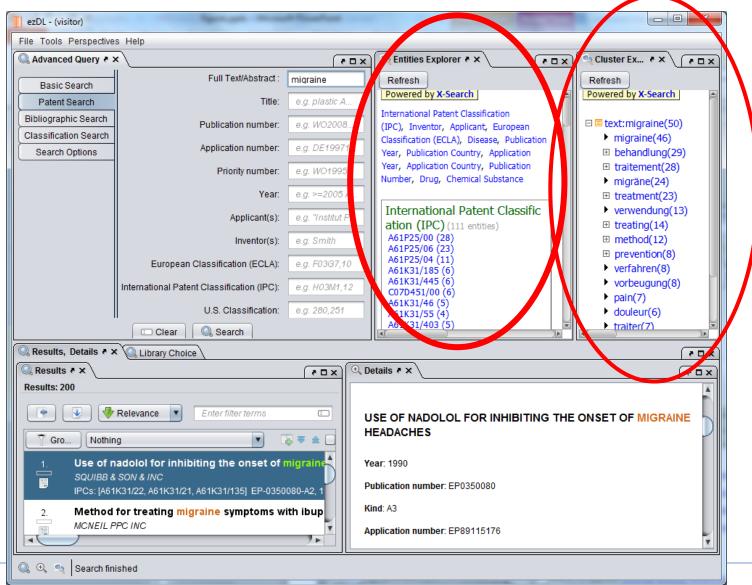
- Then we questioned ourselves:
 - why not applying and testing this in the context of a professional search system?
- Outcome
 - Application in **patent search**. Missing relevant documents is unacceptable in patent search (recall oriented search procedure). Retrieval of all relevant documents is usually necessary
 - Patents contain plenty of named entities of various kinds
 - Companies, Countries, Persons, Product types, Laws, etc
 - Inclusion of PerFedPat System
 - In collaboration with Mike Salampasis





PERFEDPAT

http://www.perfedpat.eu/



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Yannis Tzitzikas, 3rd MUMIA Training School, July 2014

PERFEDPAT (CONT)

- The proposed functionality:
 - offers a tight integration of different search tools with the main retrieval engine,
 - connects the search results (i.e. patents) with data and knowledge,
 - can be exploited by any patent search system (i.e. it acts as a service over a ranked list of results)
 - The time that we have to pay is proportional to the number of the top results that we want to "explore" ($\approx 1.5 \text{ sec} / 100 \text{ results}$)

• Related Publications

• P. Fafalios, M. Salampasis and Y. Tzitzikas, Exploratory Patent Search with Faceted Search and Configurable Entity Mining, 1st International Workshop on Integrating IR technologies for Professional Search, in conjunction with ECIR'13, Moscow, Russia, March 2013





MILESTONE 6

APPLYING IN THE CONTEXT OF AN INFRASTRUCTURE

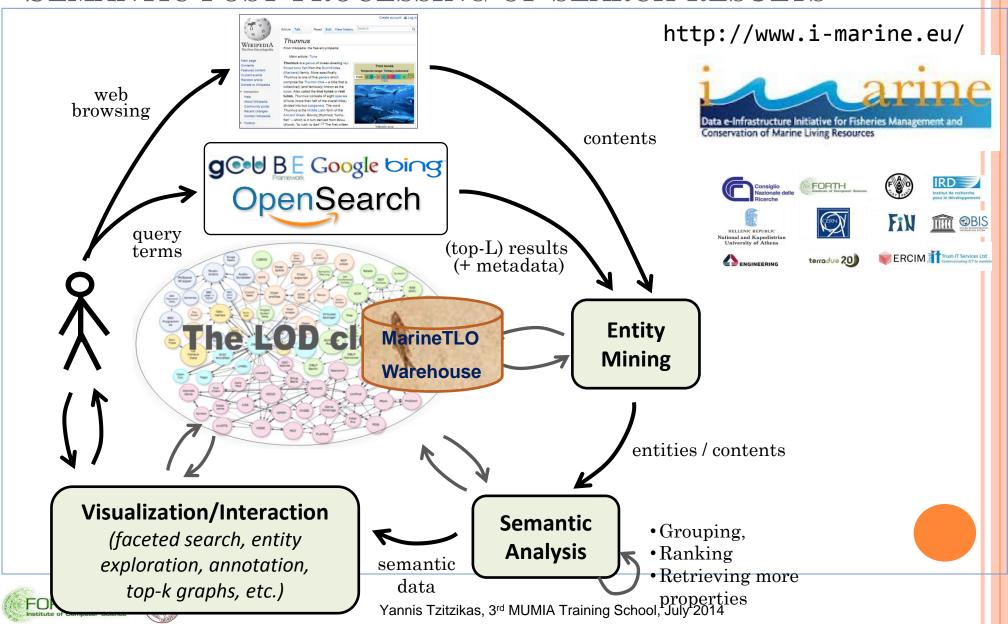
- Then we questioned ourselves:
 - why not applying this in another domain of professional search in the context of a real and operating EU research infrastructure?
- Outcome
 - X-Search in the context of the ongoing iMarine Research Infrastructure project



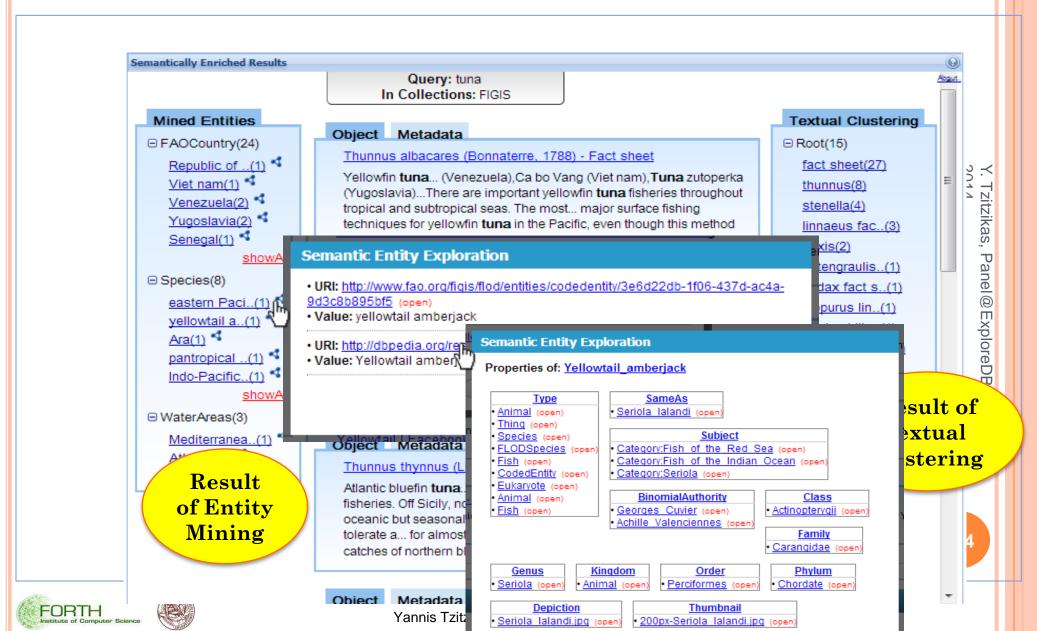
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XSEARCH:

SEMANTIC POST-PROCESSING OF SEARCH RESULTS



EXAMPLE: X-SEARCH DEPLOYED IN AN OPERATIONAL RESEARCH INFRASTRUCTURE (2012-NOW)



MILESTONE 7 PARALLELIZATION

- Then we questioned ourselves:
 - Can we do the same over the full contents of the search hits? Downloading takes time. The processing also is expensive
- Outcome
 - Investigation of how the task can be partitioned to several machines
 - A MapReduce-based parallelization of downloading and entity mining task for exploiting the resources of cloud
- Key points
 - How to tackle the uncertainty (of hits size) for achieving load balancing and reaching the ideal speedup
 - Two processes for carrying out this task





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LEVELS OF FUNCTIONALITY

• <u>L0</u>: Categories + Entities

e.g.

Location

Athens Greece

• <u>L1</u>: <u>L0</u> + Count Info for Entities e.g.

Location

Athens (3)

Greece (5)

• **<u>L2</u>**: <u>L1</u> + Rank Entities

e.g.

Location

Greece (5) Athens (3)

Ranked Entities

• <u>L3</u>: <u>L2</u> + Doc List for each entity

• <u>L4</u>: <u>L3</u> + Semantic Enrichment

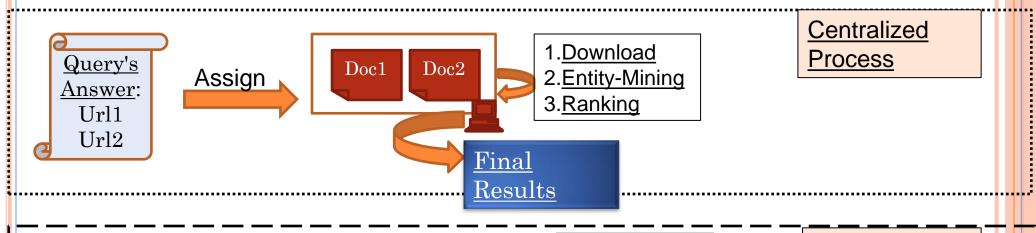
e.g.

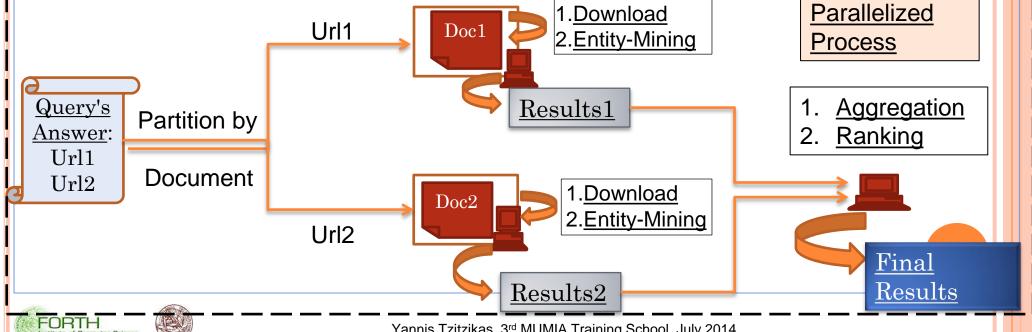
Location

Greece (5) LOD_URL Athens (3) LOD_URL

THE PARALLELIZATION PROCEDURE

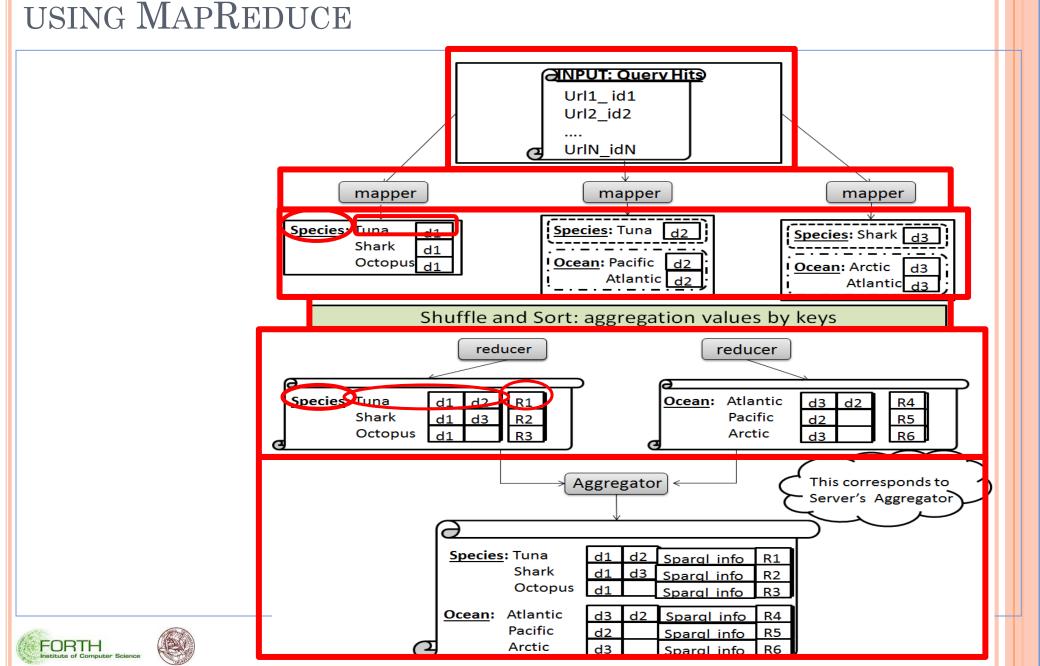
• Main idea: partition by documents (hit)



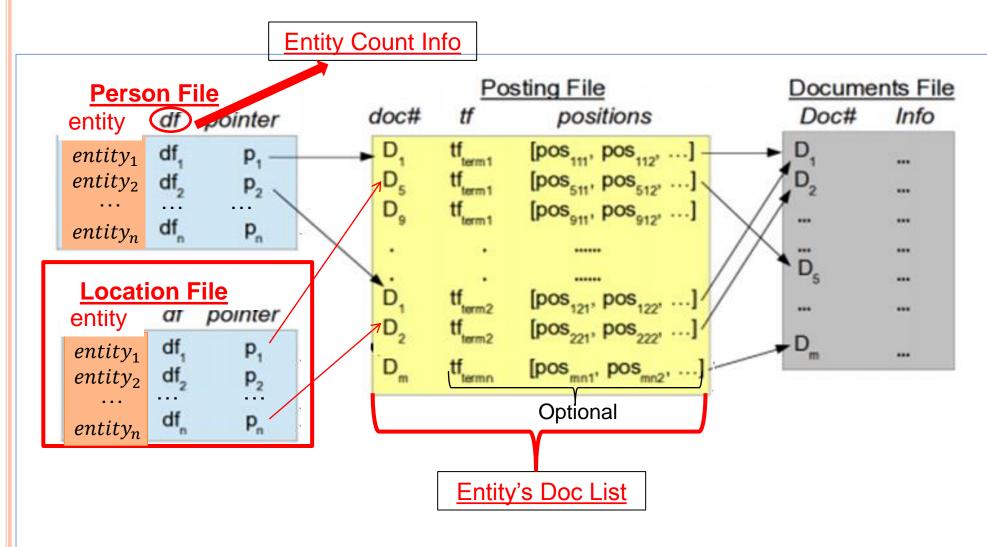


<u>Parallelization</u>

EXAMPLE OF DISTRIBUTED NEM PROCESSING



AN ANALOGY TO INVERTED FILES



- Note that our task is much more CPU and memory intensive
- Requires document's downloading

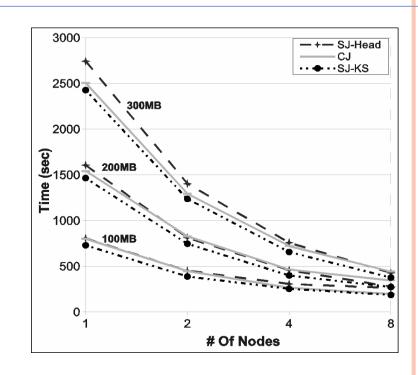






EXPLOITING MORE THAN ONE MACHINES [J. DAPD 2014]

- Two models of evaluation were investigated
- A thorough evaluation of the parameters that affect performance were conducted
- We reached a speedup close to the ideal (according to Amdahl's law)!



• Related Publications

• I. Kitsos, K. Magoutis and Y. Tzitzikas, Scalable Entity-based Summarization of Web Search Results using MapReduce, Journal on Distributed and Parallel Databases (DAPD), 32(3), 2014













MILESTONE 8 PREFERENCES

- Then we questioned ourselves:
 - What about the ordering of facets, terms and objects? Should the user only restrict the focus? Why not allowing the user to change the order based on his/her preferences?
- Outcome
 - A framework for preferences over <u>multi-dimensional</u> and <u>hierarchical</u> information spaces
 - An <u>extension of the interaction model of faceted search</u> with preferences
 - The <u>Hippalus</u> system that realizes it

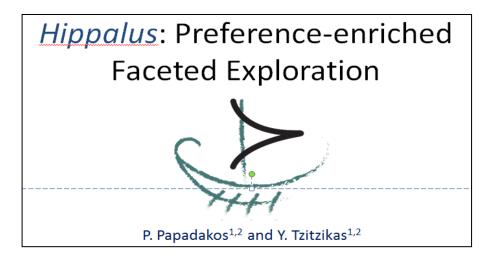




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System: Hippalus (2013)



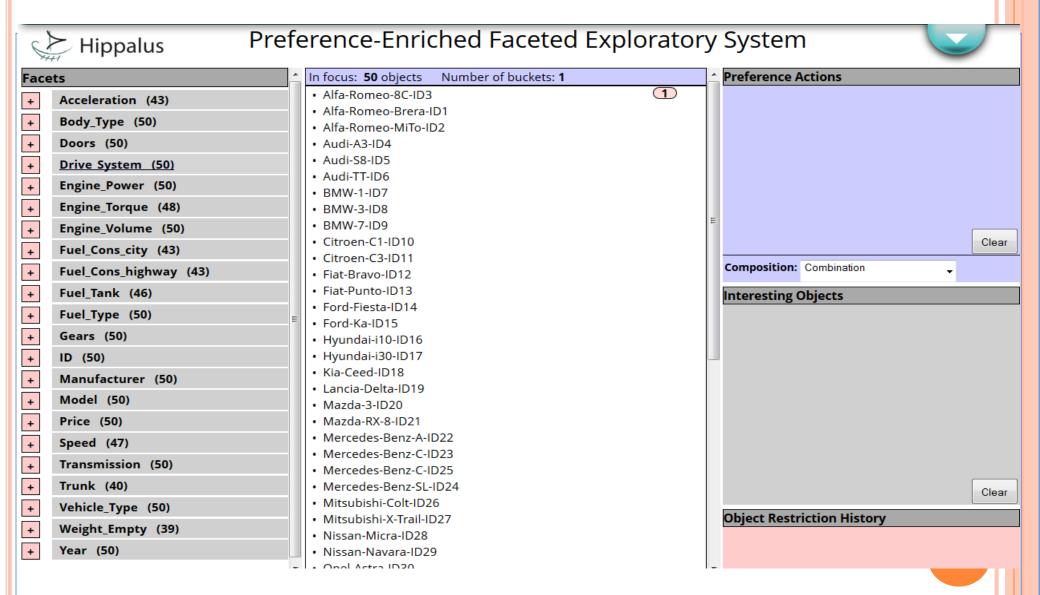


- Allows faceted browsing and also supports Preferences
 - User actions specify the ranking of the information space
 - Gradual preference specification
 - Automatic resolution of conflicts
 - Different preference composition modes
 - E.g. if the user defines the desired ordering wrt each dimension, then the first block of the ranked objects is the skyline



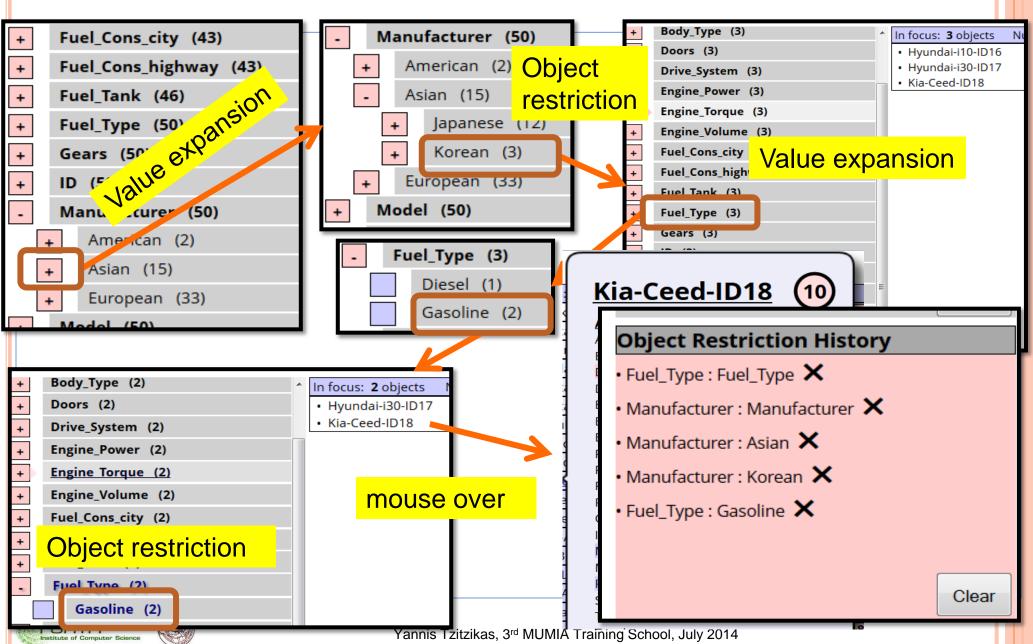


HIPPALUS: INTERACTION OVER A KB OF 50 CARS



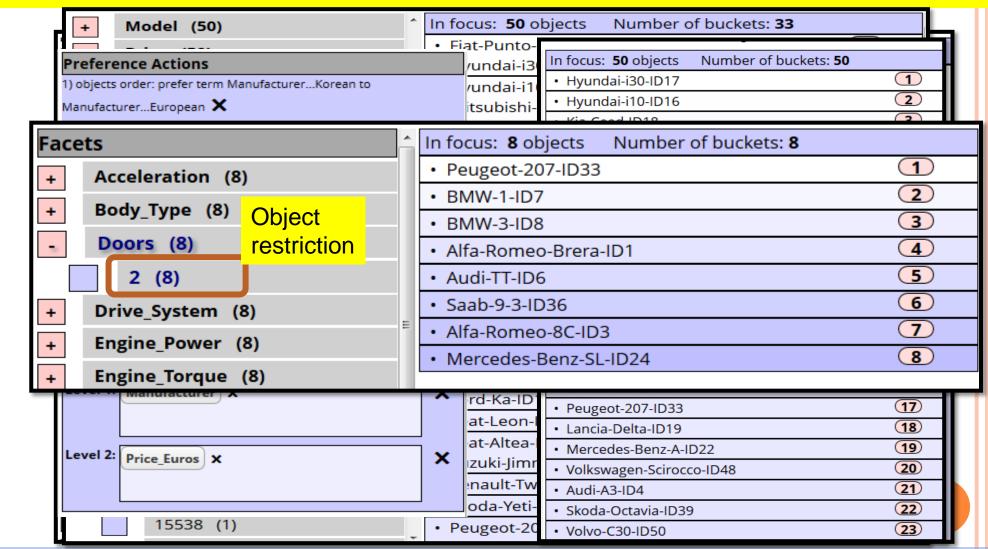


HIPPALUS: FDT INTERACTIONS



HIPPALUS: PREFERENCE ACTIONS

Cars ordered with priority on manufacturer





CONT.

Evaluation

- Over a collection of 50 cars
- With the preference-enriched, <u>all users completed successfully all tasks</u> leading to
 - ideal scores for Precision and Recall!
 - on average in 1/3 of the time!
 - on average with 1/3 of the actions!
- None of the users completed successfully all tasks with the plain interface
- <u>All users</u> (either plain or experts) <u>preferred the preference-enriched</u> interface

More information in the publications

- [J. FI 13] Yannis Tzitzikas and Panagiotis Papadakos. Interactive Exploration of Multidimensional and Hierarchical Information Spaces with Real-Time Preference Elicitation. In *Journal FUNDAMENTA INFORMATICAE*, 2013
- o [ExploreDB'14] Panagiotis Papadakos, Yannis Tzitzikas: Hippalus: Preference enriched Faceted Exploration. In EDBT/ICDT Workshops 2014



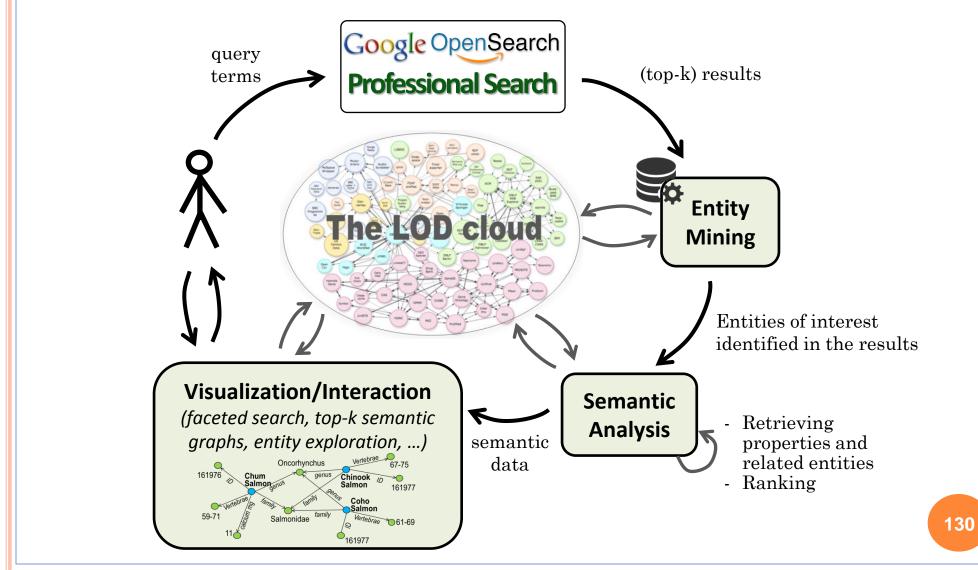
MILESTONE 9 FROM DIMENSIONS TO GRAPHS

- Then we questioned ourselves:
 - So far we have seen services for getting and exploiting multidimensional spaces over the search results. But what if the notion of dimension cannot be defined, or in case there are too many? What can be done without having to configure entity types?
- Outcome
 - A semantic post-processing of results that does not yield a multidimensional space but a graph.
- Challenges
 - Graph construction and exploitation for identifying the important (useful for the user) nodes and relationships





TOP-K SEMANTIC GRAPHS

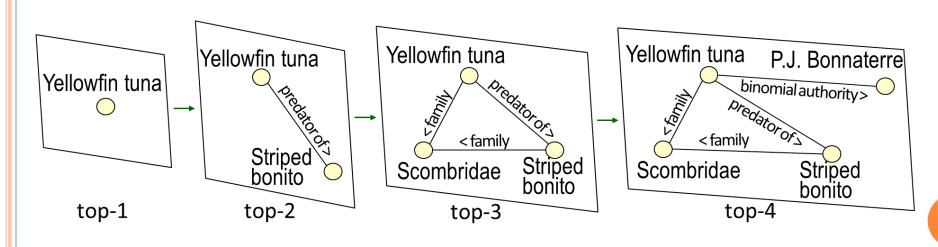






TOP-K SEMANTIC GRAPHS (CONT.)

- The system can return the top-K graph for any K from 1 to number of nodes produced
 - <u>Vertices</u>: the K most highly ranked nodes
 - Edges: the edges that connect the K most highly ranked nodes
- The user is free to increase or reduce the value of K
- Example (from a real domain):





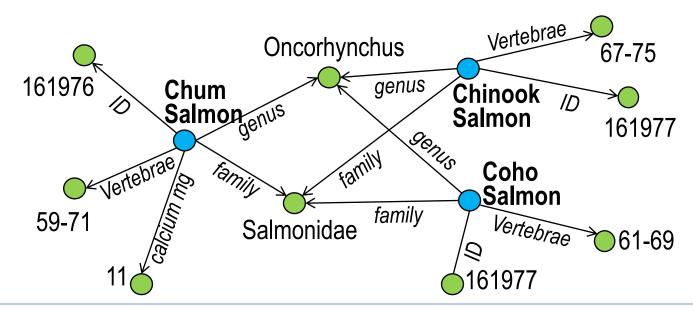


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TOP-K SEMANTIC GRAPH

This graph

- can complement the query answer with useful information regarding the connectivity of the identified entities
- o allows users to **instantly inspect** information that may lie in different places and that may be laborious and time consuming to locate
- o provides useful information about the **context** of the identified entities
- allows the users to get a **more sophisticated overview** and to make better sense of the results





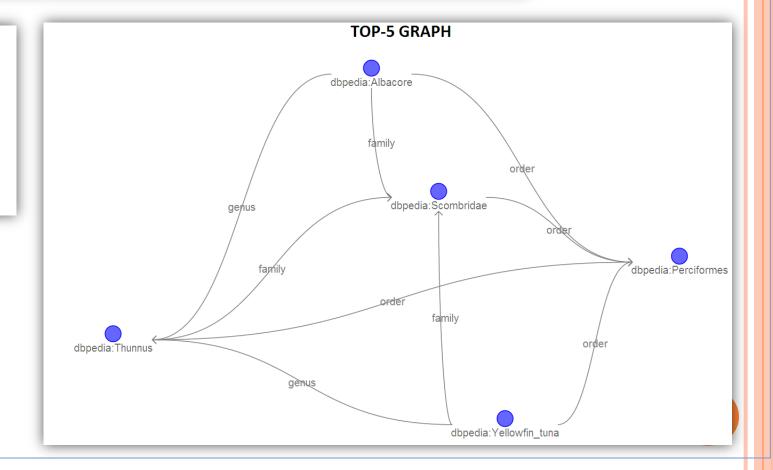


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yellowfin tuna Search

TOP-5 LIST

- 1. dbpedia:Yellowfin_tuna
- 2. dbpedia:Perciformes
- 3. dbpedia:Scombridae
- 4. dbpedia:Albacore
- 5. dbpedia:Thunnus





CONT.

- Evaluation (main results)
 - Usefulness Survey for the marine domain
 - The majority of participants believe that the appearance of a graph of semantic information related to the search results can help them during an exploratory search process
 - Effectiveness Comparative evaluation of ranking schemes:
 - The proposed PageRank-based ranking scheme produces more preferred ranking compared to other link analysis-based algorithms
 - Efficiency Case study over online DBpedia
 - The exploitation of LOD can be supported at query-time
 - For up to 100 detected entities we can offer the proposed functionality at real-time, even if we query an online KB (like DBpedia)
 - The major bottleneck is the reliability and performance of online SPARQL endpoints
 - We expect this limitation to get overcome in the near future
 - In the meanwhile, we can use caching / indexing / dedicated warehouses / distributed infrastructure
- Related Publications:
 - P. Fafalios and Y. Tzitzikas, Post-Analysis of Keyword-based Search Results using Entity Mining, Linked Data and Link Analysis at Query Time, IEEE 8th International Conference on Semantic Computing (ICSC'14), Newport Beach, California, USA, June 2014





5. SYNOPSIS AND DISCUSSION (30')

Synopsis Challenges





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SYNOPSIS

- We have discussed information needs of exploratory nature
- We have seen the basics of faceted exploration
- We have seen an overview of semantic technologies
- We have seen ways to exploit semantic datasets during searching (exploratory search) with emphasis on doing this at search time





So what's next?





DIRECTIONS & CHALLENGES

oUbiquity.

- Faceted browsing of search results and gradual restriction should be possible for <u>any kind of query</u>, for <u>any domain</u> and <u>with no predetermined facets</u>.
- In other words, methods that bypass the need for explicit configuration (regarding facets, entities types, LOD sources) are required.
 - This is why we currently study M9
 - Then we also have to define the interaction model over such graphs





DIRECTIONS & CHALLENGES (2)

oFusion of Structured and Unstructured Content.

- The exploitation of LOD in the exploratory search process is promising, e.g. for Named Entity Recognition and disambiguation. However, the fusion of structured and unstructured content requires more work.
 - E.g. should (and if yes how) the results of the semantic post processing should affect the search hits (a kind of semantic feedback)





DIRECTIONS & CHALLENGES (3)

oUser Control.

- Explicit, user-provided and controllable preference management is beneficial for supporting a transparent decision making process.
- We believe that the framework supported by the Hippalus system is a first step towards this direction.

DIRECTIONS & CHALLENGES (4)

o Evaluation.

- We need easy to follow methods for evaluating the <u>effectiveness</u> of exploratory search methods, and <u>easily reproducible</u> evaluation results.
- Although the classical IR has well established methodologies for evaluation, things are not so clear and straightforward in interactive IR (IIR).



QUESTIONS AND EXERCISES





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REFERENCES AND LINKS

Organized on the basis of the case studies



FORTH



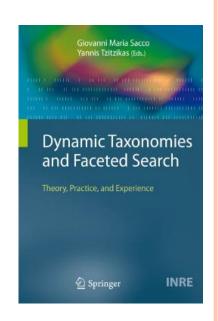
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References and Links

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 - Sacco, Giovanni Maria; Tzitzikas, Yannis (Eds.), Dynamic Taxonomies and Faceted Search: Theory, Practice, and Experience, Series: The Information Retrieval Series, Vol. 25, 2009
 - Giovanni Maria Sacco: Dynamic Taxonomies: A Model for Large Information Bases. <u>IEEE</u> <u>Trans. Knowl. Data Eng. 12</u>(3): 468-479 (2000)



- Sébastien Ferré's Publications
- Nikos Manolis, Yannis Tzitzikas: Interactive Exploration of Fuzzy RDF Knowledge Bases. ESWC (1) 2011: 1-16









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- [ECDL'09] Panagiotis Papadakos, Stella Kopidaki, Nikos Armenatzoglou and Yannis Tzitzikas. Exploratory Web Searching with Dynamic Taxonomies and Results Clustering. In ECDL 2009.
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- [J. KAIS 2012] Panagiotis Papadakos, Stella Kopidaki, Nikos Armenatzoglou and Yannis Tzitzikas. On exploiting Static and Dynamically mined Metadata for Exploratory Web Searching. In KAIS Journal 2012.





REFERENCES (M2)

o Instance Overview Search-related

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- P. Fafalios, I. Kitsos and Y. Tzitzikas, Scalable, Flexible and Generic Instant Overview Search, Proceedings of the 21st International Conference on World Wide Web (demo paper), WWW 2012, Lyon, France, April 2012
- Links to Online Prototypes
 - http://www.ics.forth.gr/ios





REFERENCES (M3+M4)

- Semantic-Post Processing of Search Results with Entity Mining and LOD
 - P. Fafalios, I. Kitsos, Y. Marketakis, C. Baldassarre, M. Salampasis and Y. Tzitzikas, Web Searching with Entity Mining at Query Time, Proceedings of the 5th Information Retrieval Facility Conference, IRF 2012, Vienna, July 2012
 - Pavlos Fafalios, Yannis Tzitzikas: X-ENS: semantic enrichment of web search results at real-time. SIGIR 2013, Dublin, Ireland, 28 July - 1 August 2013
- Links to Online Prototypes
 - http://62.217.127.118/x-ens/





REFERENCES (M5)

- Semantic-Post Processing of Search Results with Entity Mining in Patent Search
 - P. Fafalios and Y. Tzitzikas, "Exploratory Professional Search through Semantic Post-Analysis of Search Results", In "Professional Search in the Modern World", Lecture Notes in Computer Science (LNCS), Springer, 2014 (accepted for publication as a State-of-the-Art volume in LNCS)
 - P. Fafalios, M. Salampasis and Y. Tzitzikas, *Exploratory Patent Search with Faceted Search and Configurable Entity Mining*, 1st International Workshop on Integrating IR technologies for Professional Search, in conjunction with ECIR'13, Moscow, Russia, March 2013
- Links to Online Prototypes
 - http://www.perfedpat.eu/index.php/download-perfedpat
 - http://139.91.183.72/x-search-metadata-groupings/





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- Semantic-Post Processing of Search Results with Entity Mining in Marine Search
 - P. Fafalios and Y. Tzitzikas, "Exploratory Professional Search through Semantic Post-Analysis of Search Results", In "Professional Search in the Modern World", Lecture Notes in Computer Science (LNCS), Springer, 2014 (accepted for publication as a State-of-the-Art volume in LNCS).
- Links to Online Prototypes
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 - [J. FI 13] Yannis Tzitzikas and Panagiotis Papadakos. Interactive Exploration of Multidimensional and Hierarchical Information Spaces with Real-Time Preference Elicitation. In *Journal FUNDAMENTA INFORMATICAE*, Volume 122, Issue 4, pp 357-399, 2013.
 - [ExploreDB'14] Panagiotis Papadakos, Yannis Tzitzikas: Hippalus: Preference-enriched Faceted Exploration. EDBT/ICDT Workshops 2014: 167-172
- Video Demonstration available at <u>http://www.youtube.com/watch?v=Cah-z7KmlXc</u>
- Links to Online Prototypes
 http://www.ics.forth.gr/isl/Hippalus





REFERENCES (M9)

- o Graph-resulting Semantic Post-Processing of Search
 - P. Fafalios and Y. Tzitzikas, Post-Analysis of Keyword-based Search Results using Entity Mining, Linked Data and Link Analysis at Query Time, IEEE 8th International Conference on Semantic Computing (ICSC'14), Newport Beach, California, USA, June 2014
- Prototype (configured for the marine domain)
 - http://139.91.183.72/x-ens-2/





ACKNOWLEDGEMENTS

- Apart from Panagiotis Papadakos and Pavlos Fafalios, other students have also contributed to the "story" that was presented:
 - Nikos Armenatzoglou
 - Stella Kopidani
 - Nikos Manolis
- These slides include material from
 - The LOD integration scenario was taken from
 - http://www.csee.umbc.edu/courses/graduate/691/spring14/01/notes/01_in troduction/01motivating_example.pptx





Thanks for you attention





