



Applied Ontologies

Industrial applications of knowledge graphs

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http://www.ugr.es

Second oldest university in Spain (1531) 60.000 graduate and post-graduate students 28 teaching centres 75 degrees, 68 MSc degrees, 116 PhD degrees

http://decsai.ugr.es

Computer Science and Artificial Intelligence (1988)
Ranked 42 in ARWU-2015 Computer Science
70 permanent professors and lecturers, 50 research associates

Soft computing: Fuzzy Logic, Genetic Algorithms, Probabilistic Models

PhD Programme in Data Science



Research Fellow @ DECSAI









1. Motivation

- 2. Context-aware computing
- 3. Knowledge-based systems & NLP
- 4. Current trends and opportunities



"An ontology is a formal, explicit specification of a shared conceptualization."

Ontologies == Knowledge models with special features

Formal

Mathematical underpinnings: unambiguous, automatic inference, etc.

Machine-processable

Well-defined representation languages: RDF(S), OWL

Information exchange (different serializations), query (SPARQL), storage (triplestores), etc.

Standard

W3C standardization

Interaction with other property-graph software: TinkerPop (+Gremlin), Neo4j (+Cypher), etc.

Tools

Editors: Protégé, TopBraid

APIs: Apache Jena, RDF4J (previously Sesame), OWL API, RDFLib, etc.

Triplestores: Virtuoso, Blazegraph, GraphDB, etc.

Reasoning engines: HermiT, RACER, Stardog, Pellet, ELK, etc.



Knowledge base development

Support knowledge-based systems

From simple (pizza recommender) to complex (galen, umls, gene ontology)
Pizza and (hasTopping some MozzarellaTopping) and ...

Publish open linked data

DBPedia, Wikidata

Geonames

YAGO2

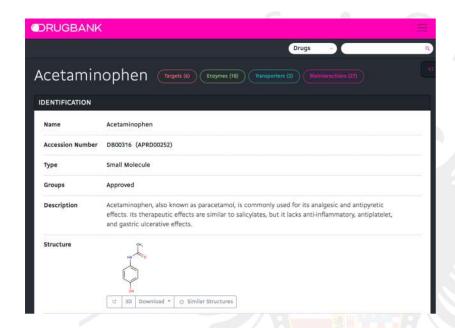
Drugbank

data.gov

Information exchange & annotation format

DCAT (datasets)

RDF Data Cube (statistical data)





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An example: video-surveillance systems

Objective

To achieve a high degree of understanding of the scene from multiple observations to barely require operator attention while cutting component costs

From PETS2002 ftp://ftp.pets.rdg.ac.uk/pub/PETS2002





Tracking moving objects with Kalman filter + identification

Issues



Trackloss



Bad adaptation to tracked entities (people)



322

Reflections



Totals STATE STATE

Tricking Control of the Control of t

Occlusions



5/.

Undetected tracks & Reflections



Tracking C D X

Groupings & Occlusions



645





Tracking

Track 008

pos ()

vel ()

Track 010 pos ()

vel()

Low level High level

Person

Entry

> Entering

Mirror

> Reflection

Column

Interpretation

Person 1 is

(Entering through Entry 2)

and

(Reflected by Mirror 1)

Context



Context-aware systems

Computational systems that use a massive amount of context knowledge The interpretation of the available information depends on context knowledge

Ambient Intelligence & Ubiquitous Computing



J. Gómez-Romero, M.A. Serrano, M.A. Patricio, J. García & J.M. Molina (2012). *Context-based scene recognition from visual data in smart homes: an Information Fusion approach*. Personal and Ubiquitous Computing 16(7), 835-857.



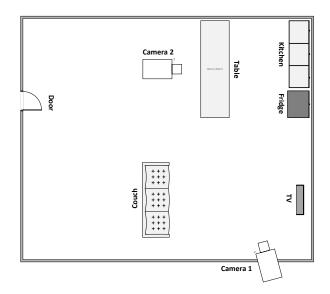
N. Díaz-Rodríguez, M.P. Cuéllar, J. Lilius & M. Delgado (2011). *A fuzzy ontology for semantic modelling and recognition of human behaviour*. Knowledge-Based Systems 66, 46-60.

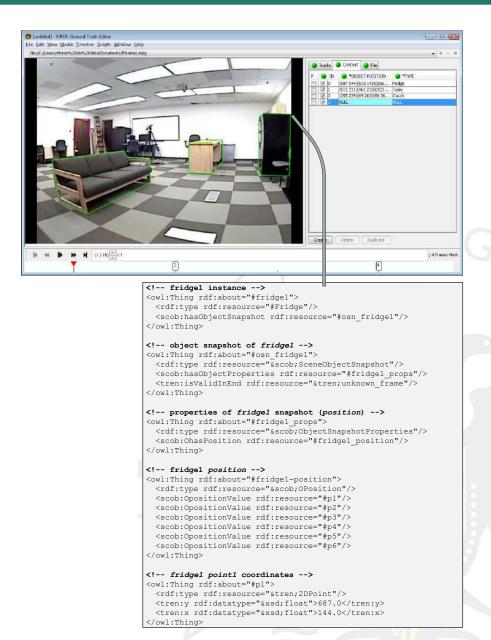


N. Díaz-Rodríguez, M.P. Cuéllar, J. Lilius & M. Delgado (2014). A survey on ontologies for human behavior recognition. ACM Computer Surveys 46(4), 43.





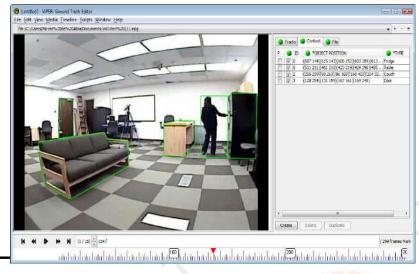


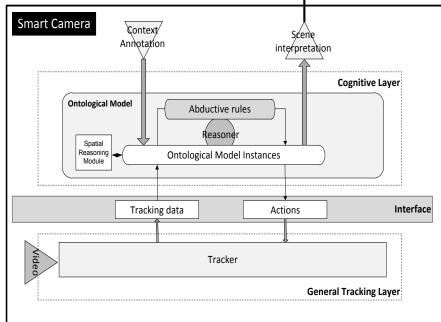












person touches fridge

fix track positions

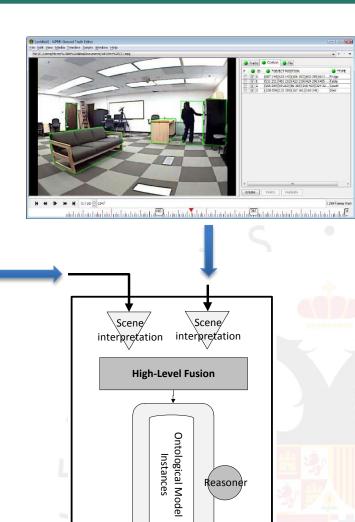












Reasoner

Fusion Node

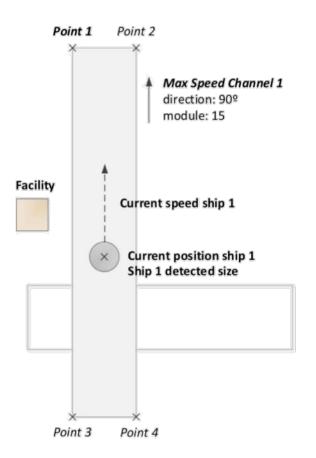
Ontological Model

person action breakfast





J. Gómez-Romero, M.A. Serrano, J. García, J.M. Molina, G. Rogova (2015). *Context-based multi-level information fusion for harbor surveillance*. Information Fusion 21, 173-186.



Representation Channel 1

Individual: channel 1

Types:

```
RestrictedArea
    Facts:
     delimitedBy p 01,
     delimitedBy p 02,
     delimitedBy p 03,
     delimitedBy p 04
     allowedSpeed speed channel 1
Individual: p 01
    Types:
        Point
    Facts:
    lat "54.6043",
    lng "19.1022"
Individual: speed channel 1
    Types:
        Speed
    Facts:
          "90"
     dir
          "15"
     mod
```

Representation Ship 1

```
Individual: shipl
    Types:
        Vessel
    Facts:
        hasSnapshot sn01
Individual: sn01
    Types:
        Snapshot
    Facts:
        speed speed 1
        length length 1
        position position 1
        insideOf channel 1
        alignedTo channel 1
        closeTo facility 1
Individual: length 1
    Types:
        Length
    Facts:
        len "10"
```



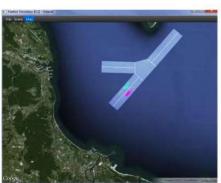


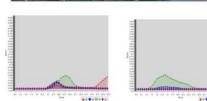
J. Gómez-Romero, M.A. Serrano, J. García, J.M. Molina, G. Rogova (2015). *Context-based multi-level information fusion for harbor surveillance*. Information Fusion 21, 173-186.

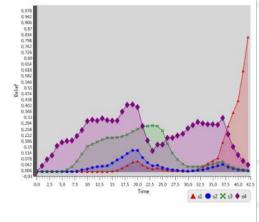
Fuzzy ontologies for situation representation

$$\langle (\mathbf{a},\mathbf{b}) : \mathsf{nearOf} \geq \alpha \rangle, \alpha = \begin{cases} 1 & dist(a,b) \leq d_1 \\ 0 & dist(a,b) > d_1 + d_2 \\ \frac{d_1 + d_2 - dist(a,b)}{d_2} & otherwise \ (d_2 \neq 0) \end{cases}$$

Fuzzy / belief-based aggregation for threat assessment









Limitations

Knowledge base must be manually created

Context description

Scene recognition

Solutions

Hybridize with Machine Learning

Automatic feature extraction



J. Wang, Y. Chen, S. Hao, X. Peng, L. Hu (2018). *Deep learning for sensor-based activity recognition: A Survey*. Pattern Recognition Letters, In Press (Corrected Proof).

7. Grand challenges

C. Flexible models to recognize high-level activities. More complex high-level activities need to be recognized other than only simple daily activities. It is difficult to determine the hierarchical structure of high-level activities because they contain more semantic and context information. Existing methods often ignore the correlation between signals, thus they cannot obtain good results.



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An example: building information model

BIM: representation of volumes, materials and equipment in a building

US National Building Information Model Standard Project Committee:

A digital representation of physical and functional characteristics of a facility.

Shared knowledge resource for information about a facility that provides support for decision-making during its life-cycle



lcon	REVIT Element	IFC Type Entity	IFC Occurrence Entity
	Duct	IFCFLOWSEGMENT	IFCDUCTSEGMENTTYPE
\mathbb{M}	Flexible Duct		
	Duct Fitting	IFCFLOWFITTING	IFCDUCTFITTINGTYPE
P	Duct Accessory	IFCBUILDINGELEMENTPROXY	-
	Air Terminal	IFCFLOWTERMINAL	IFCAIRTERMINALTYPE
8	Mechanical Equipment	IFCFLOWTERMINAL	IFCAIRTERMINALTYPE

IFC (Industry Foundation Classes) specification

Object-based data model (EXPRESS) + text-based file interchange format (STEP)

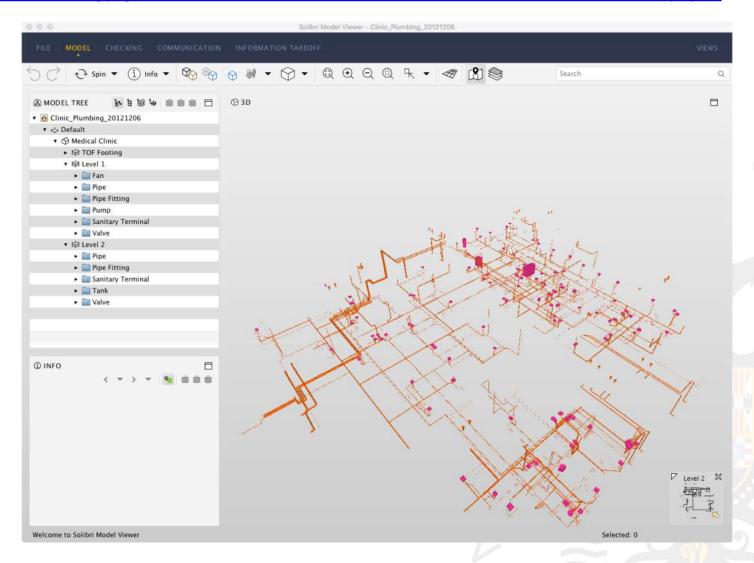
Allows creating readable models and data validation rules

Lacks a mathematical characterization of the semantics of its representation primitives



Clinic_Plumbing_20121206

https://www.nibs.org/page/bsa_commonbimfiles?&hhsearchterms=%22common+and+bim+and+file%22%3E#project3





2012-03-23-Duplex-02-Design-COBie

https://www.nibs.org/page/bsa_commonbimfiles?&hhsearchterms=%22common+and+bim+and+file%22%3E#project1

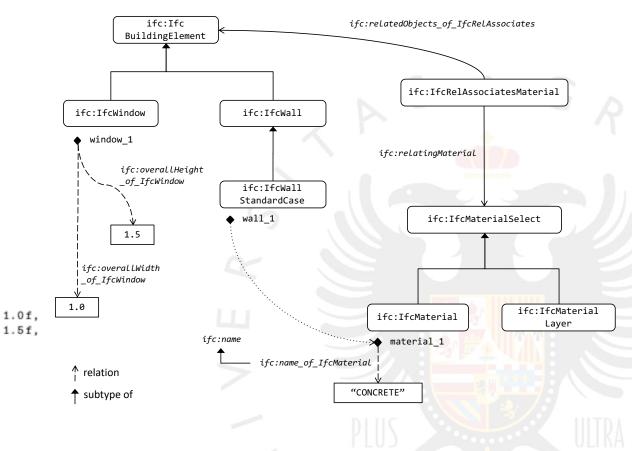
```
ISO-10303-21;
HEADER;
FILE DESCRIPTION ((''), '2;1');
FILE NAME ('', '2012-03-26T07:44:57', (''), (''), '', '', '');
FILE SCHEMA (('IFC2X3'));
ENDSEC:
DATA;
#528817= IFCRELDEFINESBYPROPERTIES('3jRe8Qj014LexP6MAAaocL', #521411, $, $, (#521705), #528819);
#528818= IFCPROPERTYSINGLEVALUE('Perimeter', 'Perimeter', IFCREAL(21.422000885009766),$);
#528819= IFCPROPERTYSET('0BTfgrhSzE7A4ylBNY0c08',#521411,'PSet Revit Dimensions',$,(#528818,#528772));
#528823 = IFCPROPERTYSINGLEVALUE('Volume', 'Volume', IFCREAL(12.239999771118164), $);
#528825= IFCRELDEFINESBYPROPERTIES('0SxgxlR9HBTv4S80Oy8qKz',#521411,$,$,(#521767),#528827);
#528826= IFCPROPERTYSINGLEVALUE('Perimeter', 'Perimeter', IFCREAL(15.319000244140625),$);
#528827= IFCPROPERTYSET('0s2gvnbuHFsPHqUBVmekO5', #521411, 'PSet Revit Dimensions', $, (#528826, #528815));
#528828= IFCRELDEFINESBYPROPERTIES('1DT1FrbgbAzBJW7JxLFHG0', #521411, $, $, (#521829), #528830);
#528830= IFCPROPERTYSET('3Kej1LMmLFFv1q5cOROun2',#521411,'PSet Revit Dimensions',$,(#528831,#528842));
#528831= IFCPROPERTYSINGLEVALUE('Perimeter', 'Perimeter', IFCREAL(5.434999942779541),$);
#528800= IFCRELDEFINESBYPROPERTIES('1YTeCslg99wBKwvk5n7MVq',#521411,$,$,(#521668),#528802);
#528803= IFCPROPERTYSINGLEVALUE('Perimeter', 'Perimeter', IFCREAL(9.840999603271484),$);
#528802= IFCPROPERTYSET('1c9QrLEi51DAOC5wSkN0jT',#521411,'PSet Revit Dimensions',$,(#528803<mark>,#528823</mark>));
```



Mapping from IFC to OWL > ifcOWL ontology IFC-to-RDF tool

```
Class: ifc:IfcWindow
SubClassOf:
    ifc:IfcBuildingElement,
    ifc:overallHeight_of_IfcWindow
        only xsd:float,
    ifc:overallHeight_of_IfcWindow
        max 1 xsd:float,

Individual: window_1
Types:
    ifc:IfcWindow
Facts:
    ifc:overallWidth_of_IfcWindow
    ifc:overallHeight_of_IfcWindow
```





Querying IFC RDF

"All the building elements built from concrete"

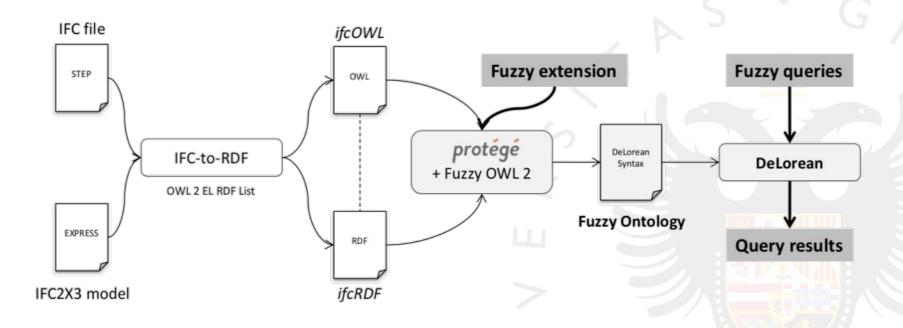
(solved by a reasoning engine)

- + More complex expressions
- + User-defined concepts
- + Detection of inconsistencies





J. Gomez-Romero, F. Bobillo, M. Ros, M. Molina-Solana, M.D. Ruiz, M.J. Martin-Bautista (2015). *A fuzzy extension of the semantic Building Information Model*. Automation in Construction 57, 202-212.





Example 1. If we consider the class IfcMaterial as a fuzzy concept, we can add a new instance representing "paper" that can be only partially considered a material. (Note that IfcMaterial already has an individual, material_1, as depicted in Figure 2.)

```
(instance :material_2 ifc:IfcMaterial >= 0.8)
(related :material_2 "PAPER" ifc:name_of_IfcMaterial)
```



Example 2. A new fuzzy role has been defined in the ontology to relate the similarity degree between two building materials, namely the similar_to_IfcMaterial object property. This property can be defined as symmetric (R9), because it holds in both directions (with the same degree), and transitive (R8). By extension, it would be possible to define other features of the property with the axioms R3-R14: reflexive, irreflexive, functional, etc. Let us also suppose that we have in the fuzzy ontology additional instances of IfcMaterial representing 'mortar' and 'ecologic mortar' materials. We can now assert that 'concrete' is quite similar to 'mortar', but 'mortar' is only moderately similar to 'ecologic mortar'.

```
Queries
(instance
             :material_3 ifc:IfcMaterial)
                                              (some :similar_to_IfcMaterial
             :material_3 "MORTAR"
(related
                                                  (value ifc:name "CONCRETE"))
             ifc:name_of_IfcMaterial)
(instance
             :material_4 ifc:IfcMaterial)
                                              (and
(related
             :material_4 "ECOLOGIC MORTAR"
                                                  ifc:IfcBuildingElement
                                                  (some inv ifc:relatedObjects_of_IfcRelAssociates
             ifc:name_of_IfcMaterial)
                                                      (some ifc:relatingMaterial
                                                         (and
(symmetric
             :similar_to_IfcMaterial)
                                                             ifc: If cMaterial
(transitive :similar_to_IfcMaterial)
                                                             (some :similar_to_IfcMaterial
                                                                 (value ifc:name "MORTAR"))))))
(related
             :material_1 :material_3
             :similar_to_IfcMaterial >= 0.8)
(related
             :material_3 :material_4
             :similar_to_IfcMaterial >= 0.6)
```



more...

Fuzzy taxonomies

A concept is partially included into other concept GlassMaterial is a MineralMaterial with degree 0.8

Fuzzy datatypes

Imprecise statements over a concrete domain

A *HighWindow* is a window with *height* defined by the trapezoid (1.2, 1.7, 10, 10)

Fuzzy modifiers

Change the meaning of a fuzzy concept by modulating its membership function A *VeryHighWindow* is a *Highwindow* modulated by the triangle function (0.4 1 1)



Applications

Cross-domain knowledge linking

A concept is partially included into other concept; graded relationships

Imprecise BIM query

Retrieve instances of fuzzy concepts; e.g. big room, breezeway

Fuzzy parametric modeling

Define soft constraints & use fuzzy constraint satisfaction

Pros & Cons

- + Inferencing
- + Available tools
- Expressiveness is computationally expensive
- +/- Ontology modeling knowledge is required

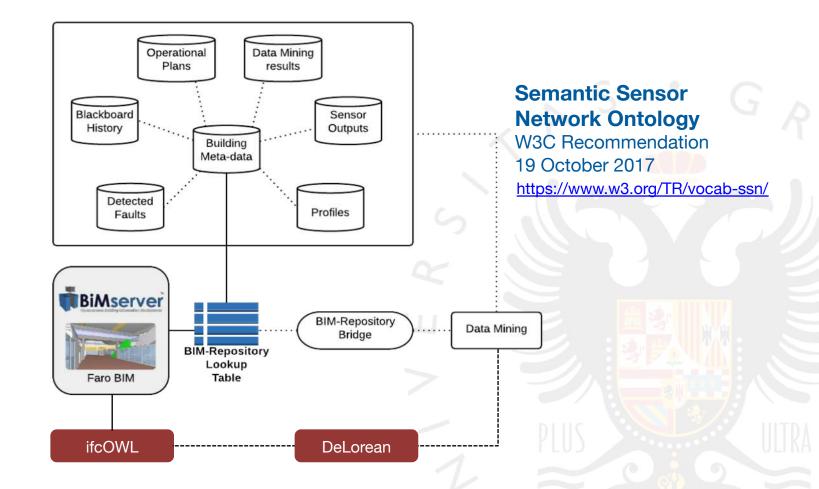


Energy IN TIME

Simulation-based control for energy efficiency building operation and maintenance









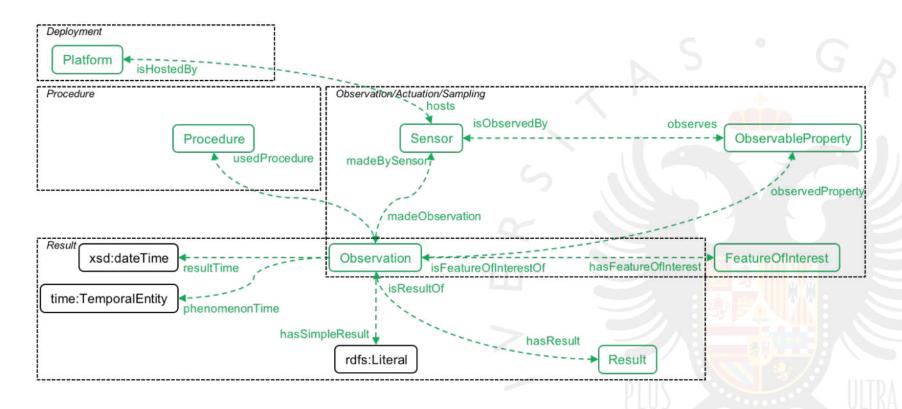
Energy IN TIME

Simulation-based control for energy efficiency building operation and maintenance



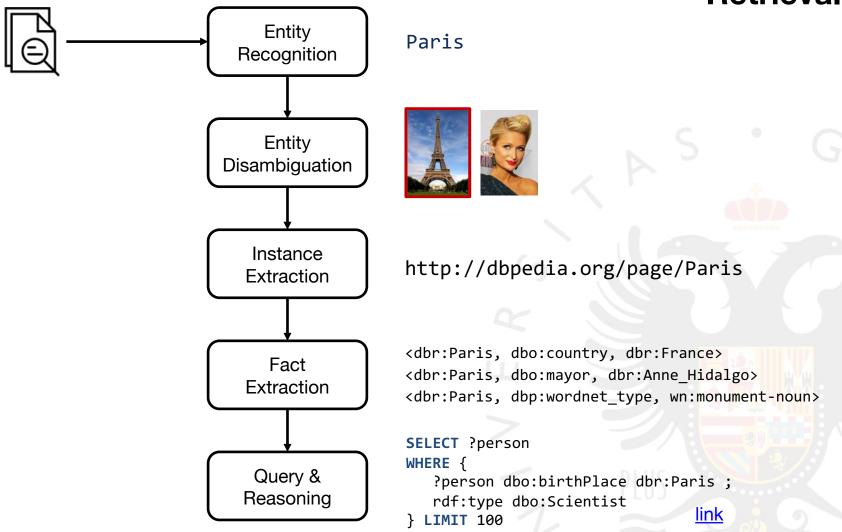


Semantic Sensor Network Ontology





An example (II): Natural Language Processing & Information Retrieval





ePOOLICE

Early pursuit against organized crime using environmental scanning, the law and intelligence systems





Extracting & processing open data to provide support to strategic analysis by means of an integrated indicator dashboard

Data acquisition

Web, External databases, Internal knowledge repository

Text processing

Entity recognition, Document categorization and filtering

Pattern discovery

Mining of relationships between entities, Discovery of trends correlations

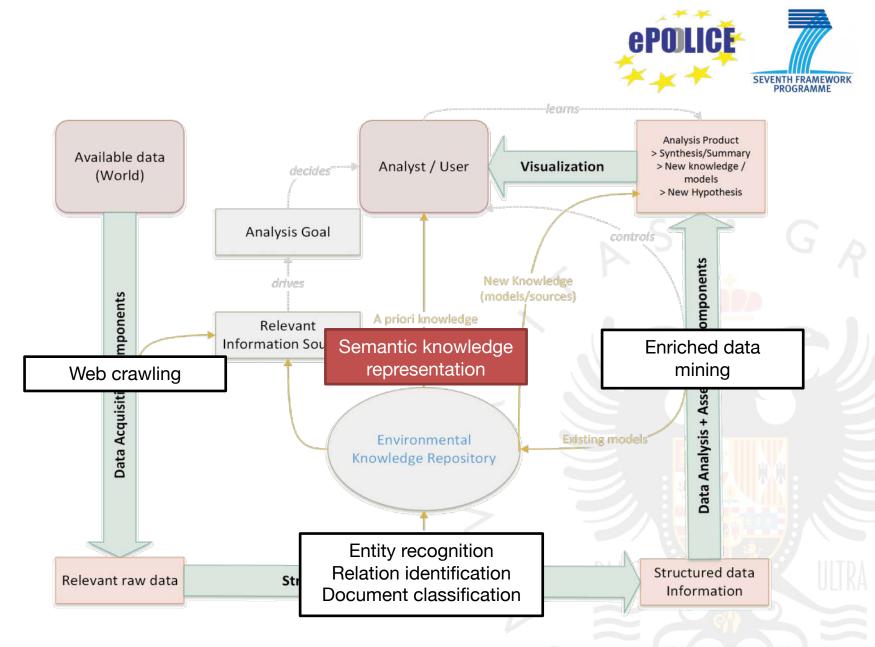
Situation and threat assessment

Threat models, Information Fusion, Alarms

Visualizing, interpreting, discovering

Map-based dashboard

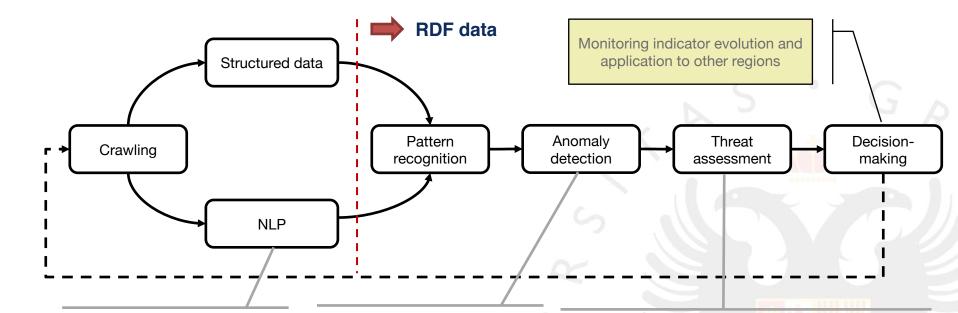






Monitoring indicators of Traffic of Human Beings in the UK





Newspaper data about investigations on fraudulent admissions in private colleges + QAA reports

Correlation between presence of non-legit private colleges and organized crime in some regions, but not in other

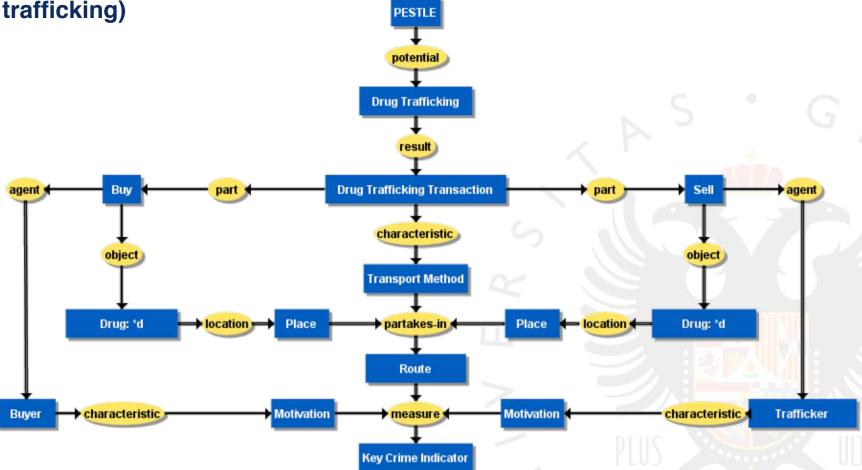
Threat assessment model:
Fraud + Conditions favoring traffic of
human beings (e.g. low wages) + events of
interest (e.g. changes in visa laws) =>
Opportunities for organized crime groups



Organized Crime taxonomy used for crawling (drug trafficking)





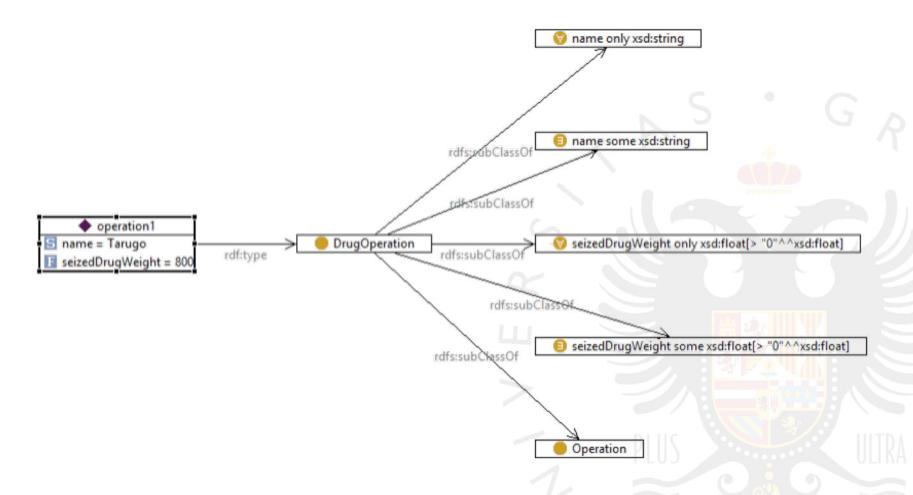




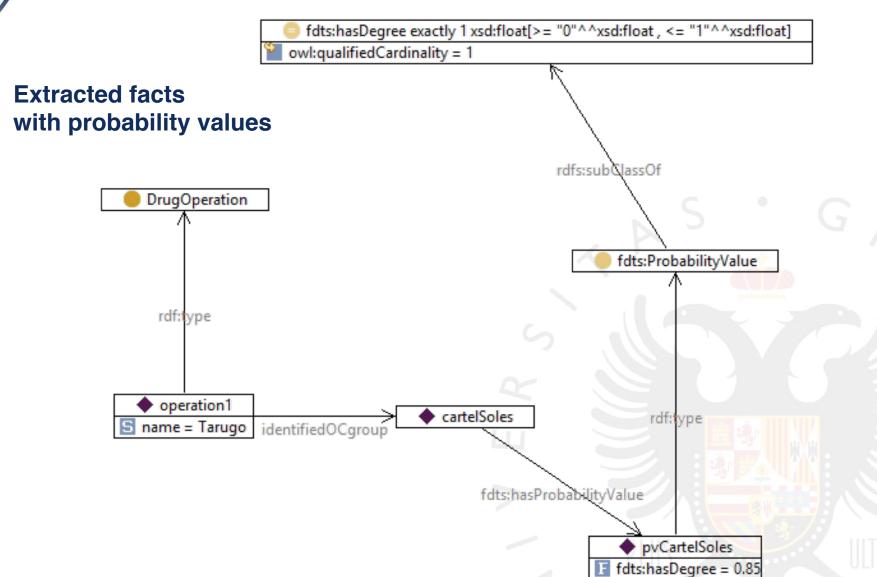
Extracted facts



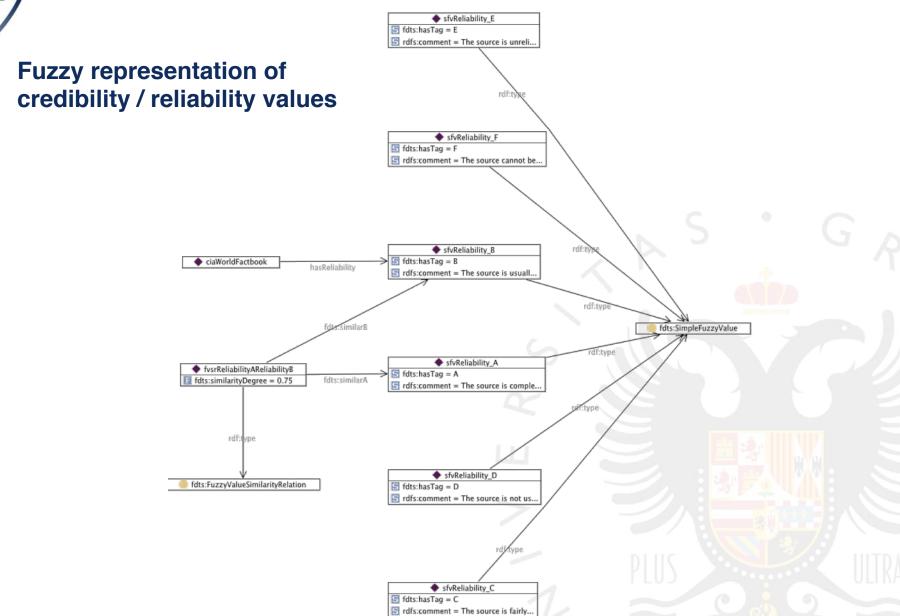






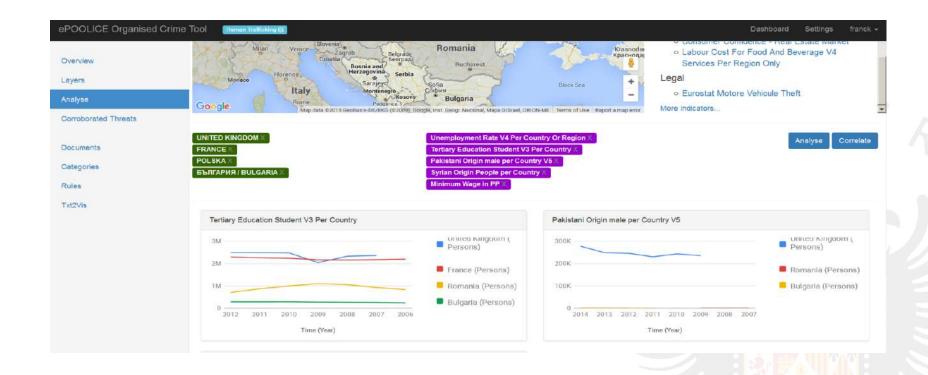








Dashboard indicator





COPKIT





Horizon 2020 European Union funding for Research & Innovation

https://copkit.eu/copkit-project-presentation-video/

Analyzing, investigating, mitigating and preventing the use of new information and communication technologies by organized crime and terrorist groups. For this purpose, COPKIT proposes an intelligence-led **Early Warning (EW) / Early Action (EA) system for both strategic and operational levels**.

Improvements

Federated knowledge base
API for read/write knowledge base
Crowdsourced expert knowledge
Enhanced support for NLP







Horizon 2020
European Union funding

Blazegraph



https://www.blazegraph.com

"ultra-scalable, high-performance graph database with support for the Blueprints and RDF/SPARQL APIs"

- 1. High Performance Native graph database
- 2. Apache TinkerPop™ API or RDF/SPARQL
- 3. Single machine data storage to ~50B triples/quads
- 4. REST API with embedded and/or webapp deployment

Virtuoso



https://virtuoso.openlinksw.com

"solution for data access, virtualization, integration and multi-model relational database management (SQL Tables and/or RDF Statement Graphs)"

- 1. Not-Only-SQL (NoSQL) data management
- 2. Web application deployment
- 3. Data privacy & security
- 4. Maximizing investments in legacy system







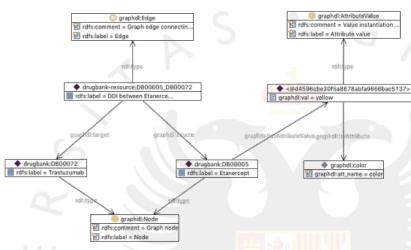
Horizon 2020 European Union funding for Research & Innovation

GraphDL

https://github.com/jgromero/graphdl

"OWL ontology that allows describing graphs with a simple vocabulary denoting nodes, edges, and properties that can be easily translated into other formats"







- J. Gomez-Romero, M. Molina-Solana (2018). *GraphDL: An Ontology for Linked Data Visualization*. 18th Conference of the Spanish Association for Artificial Intelligence (CAEPIA 2018)
- J. Gómez-Romero, M. Molina-Solana, A. Oehmichen, Y. Guo (2018). *Visualizing large knowledge graphs: A performance analysis.* Future Generation Computer Systems 89, 224-238.







European Union funding for Research & Innovation

Topbraid Composer

https://www.topguadrant.com/tools/ide-topbraid-composer-maestro-edition/



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Collaborative editor TopBraid EVN

Web server TopBraid Live

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Edition

Application Development

Tools

Data shapes SHACL

Rules SPIN **ETL** SPARQLMotion

Info pages SPARQL Web Pages SPARQL Web Application **Endpoint**







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SHACL

https://www.w3.org/TR/shacl/

Shapes Constraint Language: Language for validating RDF graphs against a set of conditions (*shapes*), which are as well expressed in RDF.

```
Example shapes graph
ex:PersonShape
   a sh:NodeShape;
   sh:targetClass ex:Person; # Applies to all persons
   sh:property [
                              # :b1
                       # constrains the values of ex:ssn
      sh:path ex:ssn;
     sh:maxCount 1;
     sh:datatype xsd:string;
      sh:pattern "^\\d{3}-\\d{2}-\\d{4}$";
  1:
   sh:property [
                               # _:b2
      sh:path ex:worksFor;
     sh:class ex:Company;
      sh:nodeKind sh:IRI;
   1:
   sh:closed true ;
   sh:ignoredProperties ( rdf:type ) .
```







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SHACL to GraphQL

https://www.topquadrant.com/graphql/shacl-graphql.html

"GraphQL schemas are automatically generated using data shape definitions in the Shapes Constraint Language (SHACL)"

GraphQL

https://graphql.org

"Query language for APIs and a runtime for fulfilling those queries"

```
Describe your data

Ask for what you want

type Project {
    name: String
    tagline: String
    contributors: [User]
}

GraphQL

Get predictable results

{
    project(name: "GraphQL") {
        tagline
    }
    }

#project": {
        "project": {
        "tagline": "A query language for APIs"
    }
}
```



"A little semantics goes a long way"

James Hendler, co-creator of the Semantic Web

https://www.cs.rpi.edu/~hendler/LittleSemanticsWeb.html



Thanks!



Juan Gómez Romero

Research Fellow

http://decsai.ugr.es/~jgomez