Chapter 1 – Introduction (planning)

[Flow: Communication Process->Vocabularies->Controlled Vocabularies->Goals of CV->Ontology->Ontology problems->Ontology learning->Pattern Extraction/Knowledge discovery->Approach to enrich an ontology based in data mining techniques.]

1P - Communication process/Languages/Vocabularies

2P - Controlled vocabularies, what are they?

3P - What problems CVs address?

* words with similar meaning and different spelling (Synonym),
* words with the same spelling and different meaning (Homograph)
* Standardization of the vocabulary of a community, one term represent only one concept

4P - Ontology as a form of CV, what is an ontology?

Why use an ontology?

5P - Problems of ontologies.

Static

Require previous agreement on the vocabulary

Require high maintenance

6P - (Automatic maintenance) Ontology learning, to help on maintenance of an ontology

What processes/techniques exist?

7P - Pattern Extraction and Data Mining Techniques to help on ontology learning and knowledge discovery

8P - What will I propose? Approach to discover knowledge in unstructured documents.

Section 1.1 – Challenges

* Lack of existence of a pure approach to quantify relations discovered from unstructured information in documents, without help of an ontology.

Ontology learning is a problem because there are no pure automatic mechanisms. (Explain ontology learning??)

* What can be done to measure a relation and find its meaning?
* This document presents an approach to help discover relations in unstructured information in documents, knowing that there are no real methods to help measure a relation between two or more concepts.

Research question:

How to quantify semantic relations between concepts in a domain ontology, using external sources of non-structured information.

Hypothesis:

Semantic relations between concepts from a domain ontology, can be quantified by applying data mining techniques for pattern extraction into non-structured sources of information.

* Having a set of documents with unstructured information, how could meaning be discovered, in the way of relations between its concepts?
* How to discover the domain of a set of words?

Section 1.2 – Expected outcomes

Present the way that I will propose solutions to research questions.

* How to address the problems?
* What techniques to use?
* Why are these techniques used to solve the problems, and not others?
* Develop a system, proof of concept, to present the results to domain experts.

Section 1.3 – Context of work

* Falar onde foi desenvolvido o trabalho
* A sua ligação com os projetos europeus (e-Cognos e CoSPaces)
* Enquadramento da tese de doutoramento do Ruben e a minha contribuição para a mesma.

Section 1.4 – Document Structure

Chapter 2 – Controlled Vocabularies

* (What are they? What do they represent?)
* What forms of representation of information exist?
* Ontologies (Definition, Construction, relations, concepts)

What is an ontology? What is it utility? How to construct one? Languages to represent it.

* E-cognos (European project for the creation of an ontology in B&C domain)
* Ontology learning
* Relations (meaning)
* Concepts
* Application domain. (Practical cases in building and construction domain)

Chapter 3 – Pattern Extraction from unstructured information sources

* Data mining / Knowledge Discovery. (What is DM/KD? Techniques used today?)
* Association Rules (Definition, Rules)
  + Algorithms to discover [ECLAT, APRIORI, FP-GROWTH]
  + Weaknesses/Strengths between them
  + Why FP-Growth?
* Application domain. (Practical cases where association rules are used)

Chapter 4 – Concept Model

- Explain conceptual model/solution

- Describe an application example

From unstructured information to knowledge representation and ontology structure

- Dimensions included in the model???

- Enrichment process

FP-Growth how to build and FP-Tree

Association rule evaluation

- DER / MVC / UML Diagrams

Chapter 5 – Model Design and Development (Proof of concept)

- Method proposal to address the question.

- What were the technologies used for the solution.

Technologies used,

- Implementation description.

(Present the server / front end solution)

- Include use cases (Relations discovered, new concepts discovered, etc.)

(Discover a relation between two concepts, update a relation between two concepts, and discover new concepts)

- Front end

Brief explanation of the functionality of the front end. Explain in a form of manual??

Chapter 6 – Assessment

* Present list of relations discovered and discuss them
* Present new concepts discovered

Chapter 7 – Conclusion and Future directions

- Evaluate if the goals reached success.

- Evaluate the achievement of the hypothesis

- Present the paper

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# Symbols and Notation

|  |  |  |
| --- | --- | --- |
|  | **API** | **A**pplication **P**rogramming **I**nterface |
|  | **AR** | **A**ssociation **R**ules |
|  | **B&C** | **B**uilding and **C**onstruction |
|  | **CV** | **C**ontrolled **V**ocabulary |
|  | **DB** | **D**ata**b**ase |
|  | **DM** | **D**ata **M**ining |
|  | **DOKS** | **D**ynamic **O**ntology learning with **K**nowledge sources from unstructured data **S**ystem |
|  | **ECLAT** | **E**quivalent **CLA**ss **T**ranformation |
|  | **FI** | **F**requent **I**tem |
|  | **FIM** | **F**requent **I**temset **M**apping |
|  | **FP** | **F**requent **P**atterns |
|  | **HTTP** | **H**yper**T**ext **T**ransfer **P**rotocol |
|  | **IT** | **I**nformation **T**echnology |
|  | **KDD** | **K**nowledge **D**iscovery in **D**atabases |
|  | **OL** | **O**ntology **L**earning |
|  | **OWL** | **W**eb **O**ntology **L**anguage |
|  | **PHP** | **H**ypertext **P**re**P**rocessor |
|  | **RDBMS** | **R**elational **D**ata**b**ase **M**anagement **S**ystem |
|  | **SEKS** | **S**emantic **E**nrichment of **K**nowledge **S**ources |
|  | **XML** | e**X**tended **M**arkup **L**anguage |
|  | **PDF** | Portable Document File |
|  |  |  |
|  |  |  |
|  |  |  |
|  |  |  |

1

# Introduction

The exponential growth of available information in digital format created the need to discover ways to organize it, in order to be easily accessible. First search engines were essentially word-based, meaning that the results provided by the search process could only be achieved if documents had in their bodies exactly the same words being searched for (Lei et al., 2006). The evolution of search engines motivated by the fact that a simple search by term for the information could not be enough, as the set of terms, or vocabulary available in information being searched could be different from the vocabulary being used. Therefore, it was of great importance to discover approaches for the representation of ideas (concepts), and not just the representation of terms, aiming at getting better results for queries (Almeida and Souza, 2011).

Nowadays, computers systems can represent sets of terms or words (also referred to as vocabularies). However, vocabularies themselves, do not represent ideas or concepts, they just represent words. In order to represent concepts and ideas, one approach can be considered. This approach is the use of mechanisms to represent more than pure words, to represent concepts. These mechanisms are referred to as Controlled Vocabularies (CV) (Lima et al., 2007). CVs are defined subsets of terms from a natural language (e.g. Esperanto), or can be pure symbols of any sort (e.g. sequence of digits) used to represent concepts, with some sort of organization. CVs represent the concepts by assigning to each, one or more words, or phrases and some describing properties that both translates its meaning. CVs also describe if or how a concept is related to other concept.

Natural languages are very rich in their vocabulary properties. They can have different meanings represented by the same word (Homograph words), in several contexts. Also, there are words that can be pronounced in the same way, however have different spelling and meaning (Homophone words). Homograph and Homophone words can lead to ambiguity and confusion when using the terms by people. CVs address the problems of Homograph and Homophone words solving them by assigning each term to just one concept, and adding properties to explain and provide a better meaning to each concept. For instance, the word “board” can represent a base used in a classroom to write with chalk, or can represent a platform to use in snow sports to ride on top of a mountain hill covered with snow. The way CVs deal with this, is by adding some properties that will increase the precision of the meaning of each term, reducing the ambiguity when these words are used. (N.I.S.O. (US) and others, 2005)

An Ontology is a type of CV that addresses problems like the consistent representation or word ambiguity in information. According to Gruber (Gruber, 1993) an ontology is “*(...) a formal specification of a shared conceptualization of a domain of interest.*” In other words, an ontology represents a formal agreement, where *formal* implies that it has to be machine readable, and *agreement* implies a shared understanding of meaning on the ontological concepts. An Ontology is used when there is the need to share or exchange knowledge within a given domain. Ontologies can be represented as a hierarchically structured set of concepts describing a specific domain of knowledge.

Although ontologies provide structures for concept representation, they face some challenges (Uschold and Gruninger, 1996). So why use an ontology? Inside an organization people from different domains can have different points of view and different words to communicate. In this sense the benefits of using an ontology is to be able to provide a common ground that can lead to a shared understanding for the same concepts. Additionally, when two IT systems need to exchange knowledge, ontologies provides them inter-operability features in order to ease the integration between them (Pouchard et al., 2000). Furthermore, ontologies are useful when there is the need to reuse its contents and features. There is no need to re-invent the wheel (Gangemi and Presutti, 2009).

Ontology Learning (OL) deals with the creation and maintenance of an ontology, and studies the mechanisms and processes to transform heavy tasks like creation and maintenance of Ontologies, into a semi or complete automatic process. IT is worth noting that relevant literature already presents first results on automatic maintenance of ontologies, but still in a very early stage. Human-based processes are still the current way to update and maintain ontology growth (Zhou, 2007).

One of the motors that drive OL itself is the recognition of patterns in the data that could originate new knowledge to further evaluation. For instance, this could be learned from some information not yet known or unpredictable in a specific domain. A pattern, in the area of information retrieval and text mining, can be defined as a predictable occurrence that repeats itself along some text data. Furthermore, Knowledge is defined as “*awareness, familiarity, or understanding of someone or something (e.g. facts, information, descriptions or skills), acquired through experience or education by perceiving, discovering or learning*.” (Oxford University Press, 2006) Therefore, OL provides techniques to discover knowledge.

Several processes can be used for a system be able to recognize patterns and further extract knowledge from data and information. Data Mining (also referred in literature as Knowledge Discovery in Databases or KDD) is one of them (Hand et al., 2001). Data mining allows experts to find knowledge in new data or data they already have. Additionally, by adopting data mining techniques, it is expected that decision makers can use new knowledge that otherwise could be unknown, unavailable or difficult to discover, to make better decisions. (Witten et al., 2011)

Having settled the context, urge to say that this dissertation aims at proposing an approach to support part of the process of ontology learning. Specifically, the proposed approach adopts a mechanism suitable for the use of data mining techniques for pattern discovery and extraction, and knowledge discovery from unstructured sources of information from a document corpus. Additionally, it is also proposed an approach to help maintain and update CVs, namely domain ontologies, with the previous discovered knowledge. This means: (i) to discover concepts and relations between them; (ii) to propose an approach to quantify these relations; (iii) to discover new concepts; And finally, (iv) to take advantage of (i), (ii) and (iii) results to update a domain ontology. Furthermore, a proof of concept to characterize this approach, referred as DOKS (Dynamic Ontology learning with Knowledge sources from unstructured text System), is also part of the results produced.

## Challenges

One of the biggest challenges in information systems when constructing a CV is to find both meaning and relations among concepts and ideas. Furthermore, how to say that a concept is more related to one, than it is to other concept? How to quantify this relation? Similarly, other challenge is to discover knowledge in sources of information that could be later used, for instance, to update a CV. Moreover, is it possible to fully automate this process? Still, other challenge identified relates to the limited amount of information that is inside a single document. This dissertation proposes an approach to help solving these challenges based in the following guiding question:

**How to formally discover and quantify semantic relations between concepts in a domain ontology, using external sources of non-structured information?**

That question highlights the research path leading the development of this work, as follows:

**Semantic relations between concepts from a domain ontology, can be quantified by applying data mining techniques for pattern extraction and knowledge discovery into unstructured sources of information.**

## Expected Outcomes

When a study is made, there is a need to consider its contribution and applicability that can arise from it. In this sense, the expected outputs to be provided by this work are the following ones:

* To develop a method to describe how to extract concepts and recognize relations between them from a data document corpus, and to find new knowledge sources in order to update a domain ontology.
* To develop a proof of concept, a software platform, based in the previous method in order to reflect the application of the studied techniques.
* Present results of the semi-automatic OL process. Results composed by patterns discovered in the documents, their relations and the new concepts discovered. They should be presented in an understandable way to the user.
* Finally, publication of scientific documents about the work, to be assessed by the academic community.

## Context of work

The context of the present work arisen from three MSc. Dissertations (Antunes, 2010; Figueiras, 2012; Parada, 2010) in the area of Data Mining and Knowledge Sources. These studies provided the background and inspiration for the reasoned path choice of the present work. The setting made through these studies was provided by CoSPaces. CoSPaces was an European Research project aiming to provide digital solutions in a collaborative workspace between individuals, teams and enterprises. The project expected to achieve the former by improving collaboration methods, like human communication and knowledge sharing support, taking advantage and improving existing IT systems.

EU research project E-Cognos was an inspiration in CV domain. Specifically, it provided the insight and methodology needed to build a domain ontology. Also, provided the ground for the structure representation of the semantics in an ontology applied in the B&C sector.

This work takes advantage of the application domain background based in the Building & Construction sector, which provided the knowledge sources, specifically technical documents (e.g. reports and papers) to be used. They were adopted from (Costa, 2014), a PhD Thesis, that also received a contribution from this study. Namely, “*Semantic enrichment of knowledge sources supported by domain ontologies*”, whose main goal was to “*introduce a novel conceptual framework to support the creation of knowledge representations based on enriched Semantic Vectors, using the classical vector space model approach extended with ontological support*”. The respective contribution was the proposal of an ontology learning method based in knowledge discovery techniques.

SEKS (Figueiras, 2012) also provided some resources which were adopted in this work, namely the domain ontology manipulation libraries.

The applicability context of the present work relied in B&C sector, as it was the domain that provided the resources and inspiration. However, in a more abstract sense, the contribution made here can be further used wherever there is the need of a shared communication and understanding of concepts, and in all the fields where knowledge and domain ontologies can be used.

## Document Structure

Following this brief introduction in Chapter 1 with the setting of the problem, the expected outcomes to achieve and the contextualization of the work by the author of the present document, this dissertation will be guided by the following structure.

In Chapter 2, Controlled Vocabularies are the domain of study. Ontology will be the selected CV discussed. It will be explained in more detail what is an Ontology and how to build one. Additionally, it will be presented some existent formalisms to represent them and where are they used.

Chapter 3 will explain what is data mining and knowledge discovery, and describe techniques to discover patterns from non-structured data. One of them, Association Rules will be explained in more detail. FP-Growth, and the concurrent algorithms to discover patterns will be compared, and explained why the former was chosen.

In the following chapter, can be observed the explanation for the solution proposed. Thus, Chapter 4 will present the concept model, an application example describing how to reach from non-structured information to knowledge representation and ontology learning. This chapter also includes the methodology behind FP-Growth and the evaluation of an Association Rule.

With Chapter 5, one can expect to read about the development of a proof of concept. The design and development of a model, with the proposed method to address the question. This will be described with the technologies used, following a description of the implementation and use cases. The framework developed will also be presented in this chapter.

Chapter 6 will be the assessment of the solution proposal, and the evaluation of the results. Chapter 7 will present some conclusions from the author, namely an overview of this dissertation, the achievement of the proposed outcomes, some possible future directions in this area and some scenarios where this work could be an asset.

2

# Controlled Vocabularies

In this chapter it will be presented an introduction to some concepts and definitions about Controlled Vocabularies. Moreover, it will be described forms of information representation. In particular it will be given special attention to the “appealing” and “scary” Ontologies. Furthermore, it will be explained how to represent a concept (or idea) and the relations between them, into an information system and how ontologies use them. Additionally, a more in depth overview of Ontology Learning will be explained in order to better understand what is it and how does it works. Lastly, a brief insight to the project that inspired the idea of Ontology use in the present work, the E-Cognos European project, applied in the Building and Construction sector.

## Controlled Vocabularies – Definition

The exponential growth of available information in digital format created the need to discover ways to organize it, in order to be easily accessible. First search engines were essentially word-based, meaning that the results provided by the search process could only be achieved if documents had in their bodies exactly the same words being searched for (Lei et al., 2006). For instance, if one wanted to do a query on a common search engine (e.g. Yahoo, Google, Bing) for the word “*car*”, each result would need to explicitly contain the word searched for. (Figure 1)

The evolution of search engines was motivated by the fact that a simple search by term for the information could not be enough, as the set of terms, or vocabulary available in information being searched for could be different from the vocabulary being used. Referring to the example from Figure 1, it is shown that if a user could query a search engine for the concept “*road vehicle, typically with four wheels, powered by an internal-combustion engine and able to carry a small number of people”*(Oxford University Press, 2006) represented by the word “*car*”, consequently the results would include the documents containing this search term. Although, the terms “*automobile*” and “*vehicle*” could also describe the same concept. Alternatively, if the term used to search the same concept was “*automobile*”, the results would be a different set of documents. Therefore, it was of great importance to discover approaches for the representation of concepts (ideas), and not just the representation of terms, aiming at getting better results for queries (Almeida and Souza, 2011). In this sense, the results provided by the query of Figure 1 example would be a sum of the results provided by the terms *“car”*, *“automobile”* and *“vehicle”*.



Figure 1 - Word search example (Yahoo, 2015)

Nowadays, sets of terms or words (also referred to as vocabularies) can be represented in computers systems. However, vocabularies themselves, do not represent ideas or concepts, they just represent words. Vocabularies are just word lists with no specific organization. Also, words *per se* are just units of a language, they have the responsibility to be the carriers of meaning. One can only understand an idea represented by a word when a meaning is associated to that word, as the meaning is itself the idea that a person wants to express when using that word.

One approach can be considered, in order to represent concepts and ideas. This approach is the use of mechanisms to represent more than pure words, to represent concepts. These mechanisms are referred to as Controlled Vocabularies (CV) (Lima et al., 2007). CVs are defined subsets of terms from a natural language (e.g. Esperanto), or can be pure symbols of any sort (e.g. sequence of digits) used to represent concepts, with some sort of organization. CVs represent the concepts by assigning to each, one or more words, or phrases and some describing properties that both translates its meaning. CVs can also describe if or how a concept is related to other concept.

*Controlled* refers to (i) the fact that the vocabulary used needs to be organized based in some logic structure (e.g. Alphabetically, Geographically, Hierarchically, etc.) and defined based in the association of each term to just one meaning, reducing the ambiguity and improving the consistency of a CV; (ii) the fact that the constraints in its use are bigger than in a natural language; a CV can only have one meaning per word. (iii) the fact that the access control to the maintenance of the terms of a CV by the people is restricted. It can have different restrictions to different users (e.g. Normal user, Domain Expert user, Admin user). For example, just domain expert users can propose new words for a CV and just admin users can add new words.

### Problems Addressed by a CV

Natural languages are very rich in their vocabulary properties. They can have different meanings represented by the same word, in several contexts, referred to as **Homograph** words. Also, there are words that can be pronounced in the same way, however have different spelling and meaning. These kind of words are referred to as **Homophone** words. Homograph and Homophone words can lead to ambiguity and confusion when using the terms by people. (Refer to Table 1 a) and b) for examples of Homophone and Homograph words respectively)

CVs address the ambiguity problems of Homograph and Homophone words solving them by assigning each term to just one concept, and adding properties to explain and provide a better meaning to each concept. For instance, the homograph word “*board*” (Table 1-b)) can represent a base used in a classroom to write with chalk, or can represent a platform to use in snow sports to ride on top of a mountain hill covered with snow or can even represent a group of directors from a company. Likewise, the homophone words “*made*” and “*maid*” have the same pronunciation, although the first word refers to the conjugation of the verb “to make” in the simple past tense and past participle, and the second word refers to a female servant. How can a search system (e.g. Yahoo!, Bing from Microsoft or Google) deal with homograph words by being able to distinguish two different meanings that can be represented by the same word? How can a voice recognition system (e.g. Cortana from Microsoft, Siri from Apple or Google Voice Search) deal with homophone words by being able to recognize accordingly two words that sounds the same? CVs deal with this by adding some properties that will increase the precision of the meaning of each term, reducing the ambiguity when these words are used. (N.I.S.O. (US) and others, 2005).

Table 1 - Examples of a) Homophone, b) Homograph and c) Synonym words

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Homophone |  | Homograph | |  | Synonym | |
| **Words** |  | **Word** | **Meaning** |  | **Word** | **Synonyms** |
| Board, Bored |  | Advocate | Speak or write in support of/Person who supports cause of another person |  | Car | Vehicle, Automobile |
| Dual, Duel |  | Board | Base to write/Platform to ride in snowboard |  | Couch | Sofa, Divan, Chair |
| Loan, Lone |  | Desert | A hot, arid region/To leave |  | Garbage | Trash, Junk, Waste |
| Made, Maid |  | Evening | Late afternoon/Making more even |  | Honest | Honourable, Fair, Sincere, Trustworthy |
| Sloe, Slow |  | Match | Wood stick to ignite fire/Sporting event |  | Intelligent | Smart, Bright, Brilliant, Sharp |
| Peak, Peek, Peke, Pique, Pick |  | Object | Thing to see or touch/Goal |  | Vocabulary | Dictionary, Terminology, Glossary |
| Rain, Reign, Rein |  | Tear | To rip/A drop of water from the eye |  | Woman | Lady, female, girl |
| **a)** |  | **b)** | |  | **c)** | |

Additionally, natural languages have more properties that must be addressed by a CV. In particular, there are words that have the same or similar meaning and have different spelling. These words are referred to as **Synonym** words (refer to Table 1c)). As a result, a concept can be represented by more than one word. Referring to Figure 1 example, one can infer from it that the concept represented by the word “*car*” have also some other words that can represent them, as any of its synonyms “*vehicle*” or “*automobile*”. A CV must allow the use of synonym words, as different people can use different vocabulary for the same concept. And in this case, a query for any of the words from the same concept must return the same or similar results.

### Advantages / Disadvantages of CV

When looking at the advantages from the use of CVs, the following can be enumerated:

* **CVs improve the efficiency and precision of retrieval systems;**

By providing more than one possible term to search for a concept, they allow for results that do not explicitly contain the search term and still are somehow related to the concept. For instance, a search by cars would provide results containing the word *“cars”* and also the results containing its synonyms words, like *“automobile”* and *“vehicle”*. Likewise, by limiting the terms that can be used, providing a more objective search through the terms used. For example, if one searches documents about *“football”* would not get documents about *“cars”*.

* **CVs remove the ambiguities from natural languages;**

Natural languages associate a word to more than one meaning. Consequently, it is hard for an information system to know what the user wants to search. As a result, each term is associated to a specific and unambiguous meaning.

* **CVs activate semantic search**, meaning that the search will be made by idea and not by word. Through the use of each term associated to the concept. In other words, means that the terms used do not need to explicitly be in the data searched for.
* **CVs improve communication through peers in a community or organization** (eg. Between two civil engineers from the same company, working on the same project). When everybody that uses a CV knows the terms to use when referring to a concept, it allows better communication through all people involved in CV use.
* **CVs provide its reusability in long-term.** To build a CV requires some time until it is ready to be used.

In the Disadvantages side, the following can be found:

* **CVs cost time & money to build for the first time**;

Building a CV takes time. First to gather all concepts and vocabulary related to a specific domain that will be necessary to include in a CV; second, to find and associate each term to a specific concept is time-consuming. As a result, companies are reluctant to adopt CVs if they want fast revenue from its investments in short term.

* **CVs allow Human/Domain Expert error**;

The concepts are gathered by humans, which should be experts in the CV applicable domain. Although expertise is an asset, the expert is still a human, and humans are prone to errors, even experts. Therefore these errors can lead to imprecise and badly formed CVs.

### Types of CVs – Differences, strengths and characteristics

Nowadays there are several ways to represent information in retrieval systems. One of them are CVs. CVs can be divided by complexity, use needs and level of control.



Figure 2 - Vocabulary Example

The simplest form of CV is a **Vocabulary**, a list of words or terms without any specific organization logic that gives names to things (Figure 2). Although a vocabulary can have some uses, in a retrieval system, most of the times just the words are not enough for semantic retrieval purposes and are the starting point of a CV.



Figure 3 - Page from a Dictionary (Oxford University Press, 2006)

When a definition is added to each word from a vocabulary, this vocabulary becomes a **Dictionary** (Figure 3). Dictionary is a vocabulary, or a list of words alphabetically ordered which contains the source of all meaning. Each word has its own meaning described. There are several types of dictionaries, in which one of them is a Language Dictionary, which contains all the words that can be used in a particular language (e.g. English Dictionary (Oxford University Press, 2006)). Other type of dictionaries exist which provides the translation of the meaning of every word from one source language to one or more target languages (e.g. Essential Portuguese Dictionary (Oxford University Press, 2012)).

Figure 4 - Example of a Species Taxonomy for b) Dog, c) Human and d) Parrot. a) Class Name Hierarchy.

CLASS

SPECIES

GENUS

FAMILY

ORDER

PHYLUM

KINGDOM

HOMO SAPIENS

HOMO

HOMINIDAE

PRIMATES

CHORDATA

ANIMALIA

AVES

PSITTACUS ERITHACUS

PSITTACUS

PSITTACIDAE

PSITTACIFORMES

MAMMALIA

CANIS FAMILIARIS

CANIS

CANIDAE

CARNIVORA



b)

c)

d)

a)

A **Taxonomy** (Figure 4) is a structured vocabulary that adds a hierarchical layer to a dictionary. Each term is in a parent-child-based structure. It is through a taxonomy that the association between words can be introduced through the parent-child hierarchy. It provides a structured classification mechanism for each term from a vocabulary.

Adding another type of layer between terms, a sibling-based (on the same hierarchically level) structure, to a taxonomy, results in a **Thesaurus (**Figure 5**)**. A Thesaurus takes advantage of a taxonomic structure and associative relations, or semantic relations to its terms. These semantic relations are in the form of synonyms.



Figure 5 - Page from Oxford Mini School Dictionary & Thesaurus (Allen and Mannion, 2007)

**Ontology** (Figure 6) is the most complex form of a CV. An Ontology includes a vocabulary, hierarchical and associative relations, and properties for each concept. (Please refer to section 2.2 for in depth overview about ontologies)



Figure 6 - Domain Ontology example (Innovation Ontology adapted from (Stick-iSchool, 2013) )

In brief, one can find some arguments that are common to every type of CV presented above:

CVs are mechanisms to structure, classify and represent terms or concepts;

CVs allow a community to agree and use the same terms in the same way;

#### Uncontrolled Vocabularies

It is worth mentioning another kind of representation structure that is not a controlled vocabulary, however is still a managed vocabulary. It is referred to as **Folksonomy**, and is considered as **uncontrolled vocabulary** (Aquino, 2007). Folksonomy provides a user the possibility to associate any word he considers adequate to any information element (e.g. documents). This could be understood as the possibility to customize the information from each entity (e.g. person, company, etc) and adapt to the context of such entity. One of differences between a CV and a folksonomy is the control factor. For a folksonomy there is less control over the vocabulary, meaning that the control is not made by experts as in a CV; on the contrary, the control is made with the people that uses it every day. This gives the possibility to words that could have been forgotten by the expert could be added by the final users that access the information. Although, a folksonomy can be customized by each user, the terms used in the information are not validated by someone who has the expertise knowledge.

Summing up, a folksonomy is a nouvelle representation mechanism that takes advantage of users and social networks to help classify words and build a vocabulary. This new form of representation is more user-oriented in contrast to the CVs which are more standard-oriented.

## Ontology, a Definition

The term “Ontology” origins from the early 18th century from the modern Latin *ontologia*, a composition from the Greek words *onto*, which means “being” and *logia* which means “study” (or science, theory). In Philosophy, ontology, is thus, the study of the nature of a being or the existence of things and how these things can be related to each other.

In Artificial Intelligence and Information Systems, the most commonly referred definition is the one presented by Gruber (Gruber, 1993). In particular, an ontology is “*(...) a formal specification of a shared conceptualization of a domain of interest.*” In other words, an ontology represents a formal agreement, where *formal* implies that it has to be machine readable, and *agreement* implies a shared understanding of meaning on the ontological concepts. An Ontology is used when there is the need to share or exchange knowledge within a given domain. Ontologies can be represented as a hierarchically structured set of concepts describing a specific domain of knowledge.

An Ontology is a type of CV that addresses problems like the consistent representation or word ambiguity in information.

### Ontology Purpose

Although ontologies provide structures for concept representation, they face some challenges (Uschold and Gruninger, 1996). Therefore, why use an ontology? Inside an organization people from different domains can have different points of view and different words to communicate. In this sense the benefits of using an ontology relies in the ability to provide a common ground that can lead to a shared understanding for the same concepts. If everyone uses the same words to communicate the same ideas, the understanding of meaning is global across all the peers that access the knowledge in an ontology.

Additionally, the need to communicate remotely and through different types of systems rises each day. Often companies work in different sites or the information is not physically located in the same place where it is accessed (as in a library). Also, for a person can be easy to understand an idea that is being communicated by other people, as they can ask questions between them to clarify possible doubts, on the contrary, IT systems cannot. An IT system by its nature, can only understand bits[[1]](#footnote-1). As a result, when two IT systems need to exchange knowledge, ontologies provide them inter-operability features in order to ease the integration between them (Pouchard et al., 2000). They provide the necessary formalisms to exchange the exact same idea between both. Ontologies provide formal specifications aiming for machine readability, by explicitly defining concepts through terms (eg. words, images, sounds, etc). Therefore, ontologies provide the capability of two systems exchanging knowledge, to understand the exact same ideas. This understanding can be extended to the point of view of human-machine interaction. Indeed, the formalisms used in ontologies are also human readable. This is a requirement in ontologies and allows a human to understand and work with the knowledge from an ontology.

Furthermore, ontologies are useful when there is the need to reuse its contents and features. There is no need to re-invent the wheel (Gangemi and Presutti, 2009). Would not be worth to build an ontology each time anyone would need one. This step is complex and time consuming, so reuse the ontological resources already available is mandatory to motivate the use of the ontologies.

### Ontology Engineering & Components

Specific concepts from a domain are not always easy to understand. Some of them are implicitly understood from other concepts. IT systems do not understand implicit concepts, in contrast they need a conceptualization of explicit ideas in order to be able to understand and work with them easily. Ontology Engineering is a discipline that studies tasks like, Ontology Building (De Nicola et al., 2009; Elsayed et al., 2007) and Ontology Maintenance (Gargouri et al., 2003) which develops approaches for explicit conceptualization of ideas.

One can find ontology tools that can deal with Ontology Engineering tasks as (eg. Protegé (Stanford Center for Biomedical Informatics Research, 2011) or OntoEdit (Sure et al., 2002), however this technology does not have yet sufficient maturation, meaning that the building of an ontology is still a manual, tedious and cumbersome task. Because of this, there is still some reluctance in ontology use. Ontology engineers often face questions and doubts related to ontology development as building time, difficulty, confidence and its maintenance.

In order to build a domain ontology, one first needs to identify the components of an ontology: *Concepts (Ideas), Relations (Meaning), Axioms (Rules)* and *Instances (Individuals).*

#### Concepts (or ideas)

A concept is defined as “An abstract idea; a general notion; an idea formed by mentally combining all its characteristics or particulars” (Oxford University Press, 2006). In other words, a concept is an idea that can be difficult to understand and is constructed in mind. It can be anything, as an object, a place, an image, a task, a reasoning process, etc., it can be whatever can be on a mind.

**Relations (mean**i**ng)**

Meaning is the concept that is represented by a word, phrase. Is the idea that a person desires to express through the use of words, signs, pictures, etc. A relation in an ontology is a connection between two or more concepts, which represents their proximity in meaning. Relations provide more information about concepts related to its meaning. In other words, they help clarify, and position concepts closer to an explicit clarification.

Table - Examples of relations

|  |  |
| --- | --- |
| RELATIONS | |
| *is-a* | *has* |
| *part-of* | *is-equal-to* |
| *is-about* | *is-similar-to* |

The relations can be manifested by hierarchical or associative relations. Hierarchical are in the form of parent-child connections (or with more levels, like grandparent-grandchild, etc.). These relations, can be found in taxonomies, in which case can be referred as a “is-a” taxonomic relation. In Figure 4 d) above, it can be seen, for instance a relation between PRIMATES (parent level) and HOMINIDAE (child level), or in other words, an HOMINIDAE “*is-a*” PRIMATE. Conversely, associative relations are found in connections in the same level, in the form of siblings, called synonyms. This association represents connections to similar or same meaning in a word or concept. Several other examples of relations can be found in ontologies. A non-exhaustive example list can be found in Table 2 above.

### Axioms (or Rules)

Axioms (or Rules) are formal descriptions of the concepts. They describe additional constraints on the ontology and allow to transform implicit facts into explicit ones. (Maedche and Staab, 2001) Axioms provide descriptions for the characteristics and properties of concepts, and can be seen as the concept definitions. They can include a collections of descriptions, as restrictions, classes, boolean combinations of descriptions and one or more individuals. (W3C, 2004)

### Instances (or Individuals)

Instances (or Individuals) are the units that are used to represent a concept. They can be a word, an image, a number, anything that can be represented and can hold the meaning of a specified concept.

### Ontology Languages

There are several formalisms defined that can provide representation of information in an ontology. Table 2 provides a non-exhaustive list just for demonstration purposes, of several languages used in Ontology Engineering.

Table - Ontology Languages (Lima, 2004)

|  |  |  |
| --- | --- | --- |
| Language | Description | URL |
| DAML+OIL | DAML+OIL is a semantic markup language for Web resources. It builds on earlier W3C standards such as RDF and RDF Schema, and extends these languages with richer modelling primitives. DAML+OIL provides modelling primitives commonly found in frame-based languages. It is important to emphasise that this language was the basis of OWL. | http://www.w3.org/TR/daml+oil-reference |
| EXPRESS / EXPRESS-G | EXPRESS-G is a standard graphical notation for information models. It is a useful companion to the EXPRESS language for displaying entity and type definitions, relationships and cardinality. Used by the ISO DIS 12006-3. | http://www.steptools.com/support/stdev\_docs/devtools/devtools-8.html |
| OIL | OILS stands for Ontology Inference Layer, a language that was developed in the context of the European IST Ontoknowledge project. It is built on top of RDF(S), using as much as possible RDF(S) constructs in order to maintain backward compatibility. | http://www.ontoknowledge.org/oil/ |
| OWL | The OWL Web Ontology Language is designed for use by applications that need to process the content of information instead of just presenting information to humans. OWL facilitates greater machine interpretability of Web content than that supported by XML, RDF, and RDF Schema (RDF-S) by providing additional vocabulary along with a formal semantics. | http://www.w3.org/TR/owl-features/ |
| RDF(S) | Resource Description Framework (RDF) defines a language for describing relationships among Web resources in terms of named properties and values. It is particularly intended for representing metadata about Web resources, such as the title, author, copyright and licensing information about a Web document, or the availability schedule for some shared resource. | http://www.w3.org/TR/rdf-schema/ |
| XML | Extensible Markup Language (XML) is a simple, very flexible text format derived from SGML. Originally designed to meet the challenges of large-scale electronic publishing, XML is also playing an increasingly important role in the exchange of a wide variety of data on the Web and elsewhere. XML has been largely used to represent "semantics" in the Web, here including taxonomies, classification systems, etc.. | http://www.w3.org/XML/ |
| Topic Maps | Topic Maps (ISO/IEC **13250)** define a model for the semantic structuring of knowledge networks and are a solution for organising and accessing large and continuously growing information pools. They provide a ‘bridge’ between the domains of knowledge management and information management. They can also be used to generate navigation for a website, and lots of other metadata tasks. A topic map is a collection of topics (a topic is a resource that acts as a proxy for some subject; the topic map system's representation of that subject), associations, and scopes that may exist in one of two forms: (i) a serialized interchange format (e.g. as a topic map document expressed in XTM syntax); or (ii) Some application-internal form, as constrained by the XTM (XML Topic Maps) Processing Requirements. A topic in a topic Map represents a subject inside the computer. | http://www.topicmap.com/  http://www.topicmaps.org |
| KIF | Knowledge Interchange Format (KIF) is a language designed for use in the interchange of knowledge among disparate computer systems. KIF, a particular logic language, has been proposed as a standard to use to describe things within computer systems, e.g. expert systems, databases, intelligent agents, etc.. Moreover, it was specifically designed to make it useful as an "interlingua". This means a language useful as a mediator in the translation of other languages. KIF has declarative semantics; it is logically comprehensive (i.e. it provides for the expression of arbitrary sentences in the first-order predicate calculus); it provides for the representation of knowledge about the representation of knowledge; it provides for the representation of non-monotonic reasoning rules; and it provides for the definition of objects, functions, and relations. When the computer system needs to communicate with another computer system, it maps its internal data structures into KIF. KIF is a programmer-readable language and thereby facilitates the independent development of knowledge-manipulation programs. | http://logic.stanford.edu/kif/kif.html |

## Ontology Learning

### Problems related to maintain an Ontology

The manual building and maintenance of an ontology is a tedious and cumbersome task. It requires knowledge acquisition, like retrieval and matching of concepts and relations, association of terms to concepts, defining constraints and axioms. The creation of a good taxonomy. It requires classification of the terms based in the taxonomic classes.

Takes time to be able to build a fully functional ontology to be used in the semantic web, for instance.

### Definition of Ontology Learning

Ontology Learning (OL) deals with the creation and maintenance of an ontology, and studies the mechanisms and processes to transform heavy tasks like creation and maintenance of Ontologies, into a semi or complete automatic process. IT is worth noting that relevant literature already presents first results on automatic maintenance of ontologies, but still in a very early stage. Human-based processes are still the current way to create, update and maintain ontology growth (Zhou, 2007).

In fact, manual building of an ontology is an extremely intensive and time consuming process, and because of this, the motivation to automate OL is high. OL provides contributions by offering to the ontology community efficiency and overcoming the bottleneck in content discovery for learning ontologies. (Zhou, 2007)

In literature, commonly OL can be found related to several fields such as *machine learning* (Buitelaar et al., 2005), *knowledge acquisition* (Sánchez, 2010), *natural-language processing* (Liu et al., 2011), *information retrieval* (Zhang et al., 2006), *text mining* (Reinberger and Spyns, 2005) and *artificial intelligence (REF)*, just to name a few.

**Unstructured (non-structured), semi-structured and structured data**

The growth of IT systems increased the quantity of data available. This fact created a new challenge, the diversification of the type of information that can be found. These includes, web pages, documents, images and others, holding more or less structure than others.

One of the ways to categorize OL systems is by the data they use to learn ontologies. One can find the data available as structured, semi-structured and unstructured (or non-structured) (Cimiano et al., 2009; Hazman et al., 2011). Consequently, the OL methods to use in each type of data are different.

Structured data is data that is already organized like in databases schemas or in some different type of CVs, like a dictionary or an ontology. As the data is already structured, the main goal in OL from structured data is to find which pieces of structural information are valuable and can provide relevant knowledge. For instance, one can identify concepts and their relations based in a database schema. (Kashyap, 1999)

Semi-structured data is related to text and data that can be found in HTML pages, XML files, etc. This data already includes some structuring, as a schema and also some free text. It takes advantage of learning methods from structured data, although also needs methods applied on unstructured data to process free text.

Finally, unstructured data relates to text or data in its raw form, without any kind of organization nor processing. This kind of data is related to natural language texts and other kinds of data found in e-books, word, pdf documents, web pages, etc. The methods used to retrieve this kind of data does not rely in any kind of structured information, therefore they are supported by statistical or natural language process approaches. (Hazman et al., 2011)

### State of the art of OL

The growth of Semantic Web increased the interest to develop methods that could ease the creation and maintenance of semantic resources as ontologies. The automatic learning of ontologies is yet an utopic task, however several researches provided approaches for semi-automatic methodologies for OL.

One example of an approach for OL commonly referred in academic documents is Text2Onto (Cimiano and Völker, 2005). Text2Onto is a tool for ontology learning from unstructured textual sources aimed for the extraction of ontologies from text documents. In particular, this tool targets the components of an ontology (concepts, taxonomical and non-taxonomical relations, and other properties), to whom are applied different algorithms. For instance, Text2Onto relies in machine learning techniques, to learn concepts.

OntoLearn (Velardi et al., 2005) is an OL system that provides a methodology also for ontology extraction from free text sources. Similar to Text2Onto, it likewise targets several steps in the OL cycle, such as term extraction, natural language definitions extraction, expert parsing of knowledge found and ontology mapping.

OntoEdit (Sure et al., 2002) is a tool aimed for OL from the Semantic Web that proposes a method composed by modules that serve different steps in the ontology engineering cycle. The main steps considered in the methodology of OntoEdit relies in requirements specifications, refinement and evaluation of resources extracted from web documents. In particular, this process includes extracting, pruning, refining, applying, importing and reusing data from web documents.

In brief, OL discipline is taking a lot of attention from the academic community.

## Ontologies in Building and Construction Sector – E-Cognos project

Development of semantic sources

The following lines will present a perspective related to the development initiatives of semantic sources European institutions and companies.

Special attention is given to the European Project in Knowledge systems aimed for building and construction sector (B&C), specifically E-Cognos project.

### Historical perspective

Europe, in the last decade developed efforts, focused in the development of controlled vocabularies aimed for B&C sector. Some initiatives include

### Creation of an ontology in B&C – E-Cognos approach

### Other practical cases in B&C sector for Ontology use

3

# Pattern extraction from non-structured sources of information

Chapter 3 – Pattern Extraction from unstructured information sources

## Data mining / Knowledge Discovery.

### Definition - What is DM/KD?

### Techniques used today?

## Pattern Discovery

### FP-Growth – definition

### Algorithms to discover patterns Weaknesses/Strengths between them

[ECLAT, APRIORI, FP-GROWTH]

### Association Rules (Definition, Rules)

* + Algorithms to discover patterns [ECLAT, APRIORI, FP-GROWTH]
  + Weaknesses/Strengths between them
  + Why FP-Growth?
* Application domain. (Practical cases where association rules are used)
* measures

4

# Concept Model

Chapter 4 – Concept Model

- Explain conceptual model/solution

- Describe an application example

From unstructured information to knowledge representation and ontology structure

- Dimensions included in the model???

- Enrichment process / Ontology learning process

FP-Growth how to build and FP-Tree

Association rule evaluation

- DER / MVC / UML Diagrams

5

# Model Design and Development

Chapter 5 – Model Design and Development (Proof of concept)

- Method proposal to address the question.

- What were the technologies used for the solution.

Technologies used,

- Implementation description.

(Present the server / front end solution)

- Include use cases (Relations discovered, new concepts discovered, etc.)

(Discover a relation between two concepts, update a relation between two concepts, and discover new concepts)

- Front end

Brief explanation of the functionality of the front end. Explain in a form of manual??

6

# Assessment

Chapter 6 – Assessment

* Present list of relations discovered and discuss them
* Present new concepts discovered

7

# Conclusion and Future directions

Through the following lines will be presented an overview of the work developed in this thesis. The objectives were defined in Chapter 1 that intended to guide the path of the study. For these objectives it will be described which ones were achieved and which ones were not, describing also the problems and difficulties found during the development and research, and also, how were these difficulties solved. Similarly, this will also bring to attention some possible future research topics, where achievements addressed by this work can constitute a solid basis.

## Work overview

As presented through this dissertation, it was described the importance of concept representation in contrast to word representation, in the Semantic Web area. It was highlighted the advantage of the use of mechanisms (e.g.. Controlled Vocabularies), as these provides means for semantic representation, which allows more than just simple word representation. These mechanisms provide the possibility to make semantic search in contrast to the currently used search technology, providing results more adapted to what users want.

Similarly, frequent pattern discovery in texts may enhance the recognition of semantic relations between words. As a result this recognition can help discover the meaning associated to a word. The Data Mining techniques adopted to achieve this task were FP-Growth to discover frequent patterns and Association Rules to provide more than the just recognition of relations between the words (refer to chapter XX.XX). Based in the AR algorithm, this work demonstrated that it is possible to measure the strength of a relation.

Pattern recognition by itself can be an indicator of relations between words, however this can be enhanced through the use of a domain ontology. In this sense, this work proposed Frequent Itemset Mapping, a process to match frequent items discovered in a document corpus and keywords associated to concepts from a domain ontology related to B&C (refer to chapter 4).

As explained in Chapter 2.3, Ontology Learning is the area related to the automatic or semi-automatic (meaning without human supervision) maintenance of an ontology. Through newly discovered knowledge sources it is possible to learn a domain ontology, in the sense that one can use this new knowledge that could otherwise be unknown, be difficult to discover or be unavailable to improve concept relations inside the ontology. The method proposed in this dissertation, based in the AR algorithm, provides metrics in the form of numeric values to evaluate the strength of semantic relations between concepts. Through these value is possible to know if a concept A is more related to a concept B than to a concept C, therefore learn or maintain the concepts from a domain ontology, as the one used in this work related to B&C.

## 7.2 Research Contributions

The development of this work proposed four expected outcomes in Chapter 1 as follows:

* To develop a method to describe how to extract concepts and recognize relations between them from a data document corpus, and to find new knowledge sources in order to update a domain ontology.

The proposed method relies on applying Data Mining techniques to discover knowledge in documents that could be useful to update a domain ontology. Knowledge, meant the discovery of new concepts, relations or the improvement of the relations between the concepts already in a domain ontology (e.g. the ontology used in this work adopted from the B&C domain). The initial resources were a set of documents from the ICONDA[[2]](#footnote-2) database and a domain ontology adopted from SEKS framework developed under a MSc. Dissertation (Figueiras, 2012) both related to B&C domain. The documents were initially processed in the Rapidminer software tool. Rapidminer proved to be a satisfactory tool, as it also allowed to apply the algorithms FP-Growth to discover frequent patterns and Association Rules to discover the relations. The process created for the matching between the frequent items discovered in the documents and the keywords associated to the concepts from the domain ontology was the Frequent Itemset Matching (refer to chapter XX.XX). This process allowed to search through the ontology in order to verify if the frequent items discovered in the documents were associated to any concept inside the ontology, or if it originated new knowledge.

This work tries to develop a method for Ontology Learning where it is possible to turn a domain ontology more up to date. Even with a small sample, this process provided some good results (refer to chapter XX.XX), as it discovered new concepts, and also provided some interesting relations between the concepts. However, OL relies on automatic methods, this work did not intended to provide a full automatic method to learn an ontology. Alternatively, it was intended to develop a semi-automatic method that relies in human interaction to complete the OL task with the knowledge discovered through all the results.

* To develop a proof of concept, a software platform, based in the previous method in order to reflect the application of the studied techniques.

In order to execute all the steps from the method proposed it was developed a software tool, DOKS (refer to chapter XX.XX). DOKS is a client-server application developed using Java technology to implement all the processes and components in this tool. To interact with the ontology, it was used Jena API. The communication to the database was made by JavaBeans technology. The ontology was developed in OWL. Rapidminer provided an API to access its results, and they were exported through a script represented in Groovy. To hold the results for later access, it was created an XML message. Both DBs for the ontology and for the AR results were saved in a MySQL RDBMS.

* Present results of the semi-automatic OL process. Results composed by patterns discovered in the documents, their relations and the new concepts discovered. They should be presented in an understandable way to the user.

To present the results from DOKS, a FrontEnd was implemented in web technology. Here the set of technologies used were: (i) Html5+CSS3 as a base to support the layout; (ii) The communication with the server was made through HTTP requests based on PHP technology to send the results; (iii) To present the results in the web page, the technology chosen was PHP + XPath. The results were presented in a first page, in which the user could choose two concepts, based on the Frequent Itemset Mapping, and the values of the metrics from each association rule presented by the FrontEnd. This way, a relation between two concepts could be chosen for later processing. It is worth mentioning, the creation of a colour scheme for the Frequent Itemset Mapping process, in order to help the user choose the concept from the domain ontology that best matches the frequent item.

* Finally, publication of scientific documents about the work, to be assessed by the academic community.

The following scientific documents were published after assessment by the academic community during the development of this work:

* Luis Paiva, Ruben Costa, Paulo Figueiras, Celson Lima, “Discovering Semantic Relations from Unstructured Data for Ontology Enrichment - Association rules based approach”, 8ª Conferência Ibérica de Sistemas e Tecnologias de Informação: CISTI'2013, pp 579-584, 2013
* Ruben Costa, Paulo Figueiras, Luis Paiva, Ricardo Jardim-Gonçalves, Celson Lima, “Capturing Knowledge Representations Using Semantic Relationships An Ontology-based Approach”, Sixth International Conference on Advances in Semantic Processing: SEMAPRO 2012, pp 75-81, 2012
* Paulo Figueiras, Ruben Costa, Luis Paiva, Ricardo Jardim-Gonçalves, Celson Lima, “Information Retrieval in Collaborative Engineering Projects-A Vector Space Model Approach”, International Conference on Knowledge Engineering and Ontology Development: KEOD2012, pp 233-238, 2012

## 7.3 Future Directions

As this work relates to some areas from Semantic Web and Ontology Engineering, some possible directions can be identified for further work and improvement. Two paths are proposed, one related to the improvement of the presented method, the second related to its applicability and reuse.

Sometimes, the knowledge that results from the method proposed herein can be huge, and if the process is not fully automated it can be an exhaustive task to analyse these results. This suggests further research related to DOKS ability to deal with the size growth of data used in the Ontology Learning process, can be identified in three areas: (i) speed to process large sets of data as it can be really slow. Research can be taken in methods to, for instance, take advantage of multi-core processor technology in order to use parallelization techniques to improve the speed of the matching process; (ii) way to present results for evaluation by an expert, although this work provided a colour scheme to represent the strength of the matching process (refer to chapter XX.XX). This means to improve the way in which the results are shown, by using more graphics (e.g. graphs to represent relations). This will provide a better efficiency of the method itself and allow for faster reasoning of the results; (iii) method to process large/huge and complex sets of data, also known as Big Data. Big Data is the nouvelle sub domain of Data Mining that studies solutions to the problem of big and complex sets of data.

Searching for patterns in a document, was proved by this work that it is not an easy task, although it is possible. The relation between words in a document can lead to the discovery of a central concept or idea that could represent its context or domain, for instance a document including the words “*bridges*” and “*buildings*”. The central concept from this document could be identified as “*Civil Engineering*”. However, how can one discover the central idea in a document? Is it even possible? Can this discovery be done? How to find the central idea? Syntactic Context[[3]](#footnote-3) or Latent Semantic Analysis[[4]](#footnote-4) are areas that tries to address this questions, and can be a promising future direction.

It is worth mentioning that the intention of this research was not to develop a fully functional model to deal with data mining. However, the author thinks that it could be a good contribution to the following areas:

* Information Systems: Search engines like Google, Bing or Yahoo, just to name a few, could use semantic search capabilities to improve its results instead of just statistical ones. For instance, if one would like to search for a car, the search engine could provide the pages where the word “*car”* appears, and also the pages where the word *“automobile”* appears based on the relation between these two words. This means that one could search for a concept, instead of having to know every word that represents it. Additionally, it could also provide suggestions, for instance, related to their brands, turning the search results more close of what the user really searched for.
* Cybersecurity: This is an area of great interest in the present, based on several world events related to terrorism. The method proposed in this work, could help in this area, for instance, if one could use a search engine to search in the web for a person A that could be known as related to Al-Qaeda. After using the method proposed, one could also discover a person B that appears frequently in some pages related to person A, although not directly related to Al-Qaeda. This could be a proof of the relation between both people, and the discovery of the relation of person B to Al-Qaeda.
* Cybersecurity and human rights: MEMEX is a project from DARPA with an initial goal of using search technology to help fight human trafficking, as they identified this as a serious problem to solve. The secondary goal of this project was identified as to improve the search mechanisms and tools that are used today. Semantic search could help in the sense that it could discover pages with terms related to human traffic, for example *human trade* or *modern slavery*, which could represent the same idea.
* Team Sports: GloballCoach is a software tool idealized by former Liverpool and Chelsea coach, Rafael Benitez, that targets Football Teams and their coaches, providing them data analysis capabilities. Amongst others, this software aims to recognize patterns in game data to show to team players and improve their tactical and technical behaviour. This kind of software systems are focused in the individual behaviours of each player. In this sense, the method proposed in this work could be a great aid in the recognition of relations between the players, augmenting the analysis from an individual up to team perspective analysis capabilities. For instance, the coach could analyse which is the best relation in their left side. Meaning that if he wants to select player A for a game, he could know if the relation between players A and B provides more goals than with A and C. Other example could be, to whom a player A provides more assistances[[5]](#footnote-5), meaning that the relation between player A and B provides more goals to the team than a relation between player A and C.

Summing up, the semantic search is here to stay and is spreading along all research areas. In this sense, controlled vocabularies are useful tools to enhance semantic search capabilities in information systems. Ontologies themselves are great mechanisms to provide search capabilities to users, experts or not, in theirs daily search quests. Consequently, knowing how to provide or obtain the best results will throw companies or entities one technological step ahead of their competitors.

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1. Bits (also referred as binary digits) are the basic units in a digital system. They commonly can have values of 0 or 1 in which are used to represent data. [↑](#footnote-ref-1)
2. ICONDA is a large database of technical documents related to B&C domain. [↑](#footnote-ref-2)
3. Syntatic Context relates to the order of the words in a sentence, and states that through language rules, one can infer the context of a sentence. [↑](#footnote-ref-3)
4. Latent Semantic Analysis is the area that analyse the relations between documents, trying to find correspondence between its terms and concepts in order to infer its context. [↑](#footnote-ref-4)
5. Assistance is the word used in a football game and represents the moment when a player passes the ball to a teammate, and this teammate scores a goal. [↑](#footnote-ref-5)