

Ontology Alignment Overview

José Manuel Vázquez Naya

University of A Coruña, Spain

Marcos Martínez Romero

University of A Coruña, Spain

Javier Pereira Loureiro

University of A Coruña, Spain

Alejandro Pazos Sierra

University of A Coruña, Spain

INTRODUCTION

At present, ontologies are considered to be an appropriate solution to the problem of heterogeneity in data, since ontological methods make it possible to reach a common understanding of concepts in a particular domain. However, utilizing a single ontology is neither always possible nor recommendable, given that different tasks or different points of view usually require different conceptualizations. This can lead to the usage of different ontologies, although in some cases the different ontologies collectively might contain information that could be overlapping and possibly even contradictory. This, in turn, represents another type of heterogeneity that can result in inefficient processing or misinterpretation of data, information, and knowledge.

To address this problem while at the same time insure an appropriate level of interoperability between heterogeneous systems, it is necessary to find correspondences or mappings that exist between the elements of the (different) ontologies being used. This process is known as ontology alignment.

This article offers an updated overview of ontology alignment, including a detailed explanation of what alignment consists of, and how it can be achieved. First, ontologies are defined using a fusion of different interpretations. This is followed by a definition of the concept of ontology alignment and, using a simple example, some of the most commonly used alignment techniques are illustrated. Subsequently, a case is made for the importance of automating the process of ontology alignment, summarizing some of the main alignment systems currently in use. Finally, in the context of future directions, a discussion is presented of the advantages

associated with integrating ontology alignment into systems that require exchanging information in an automatic fashion.

BACKGROUND

Towards the end of the 20th and beginning of the 21st centuries, the term “ontology” (or ontologies) gained usage in computer science to refer to a research area in the subfield of artificial intelligence primarily concerned with the semantics of concepts and with expressive (or interpretive) processes in computer-based communications. In this context, there are many definitions of ontology, and these definitions have evolved over the years. Gruber offered one of the first definitions of ontology in 1993, as follows (Gruber, 1993):

“An ontology is an explicit specification of a conceptualization”.

Gruber’s definition became the most frequently referenced one in the literature, and became the base or working definition for those working in this area.

At present, ontologies are viewed as a practical way to conceptualize information that is expressed in electronic format, and are being used in many applications including the Semantic Web, e-Commerce, data warehouses, or information integration and retrieval. The basic idea behind these applications is to use ontologies to reach a common level of understanding or comprehension within a particular domain (e.g., a particular industry, medicine, housing, car repair, finances, etc.).

However, certain systems that encompass a large number of components associated with different domains would generally require the use of different ontologies. In such cases, using ontologies would not reduce heterogeneity but rather would recast the heterogeneity problem into a different (and higher) framework wherein the problem becomes one of ontology alignment, thereby allowing a more efficient exchange of information and knowledge derived from different (heterogeneous) data bases, knowledge bases, and the knowledge contained in the ontologies themselves. In this manner, ontology alignment enhances system interoperability.

ONTOLOGY ALIGNMENT

Euzenat et al. defined the problem of ontology alignment in the following manner (Euzenat et al., 2004):

“Given two ontologies which describe each a set of discrete entities (which can be classes, properties, rules, predicates, etc.), find the relationships (e.g. equivalence or subsumption) holding between these entities.”

The key issue in ontology alignment is finding which entity in one ontology corresponds (in terms of meaning) to another entity in one (or many) ontology (or ontologies). Essentially, one might say that ontol-

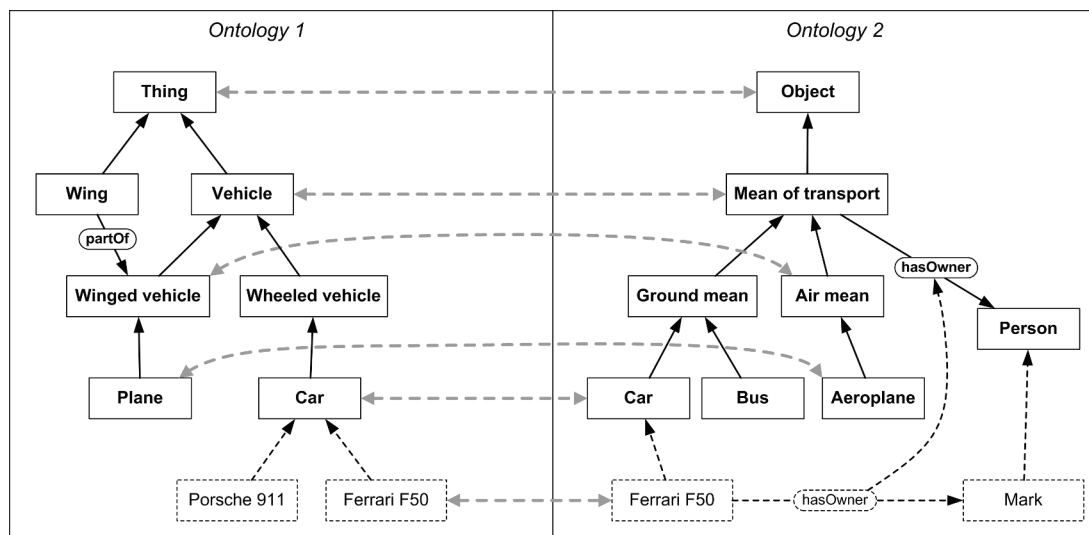
ogy alignment can be reduced to defining a similarity measure between entities in different ontologies and selecting a set of correspondences between entities of different ontologies with the highest similarity measures.

There are different methods to calculate the similarity measures between entities, and collectively these methods are known as **ontology alignment techniques**. Many of these techniques are derived from other fields (for instance, discrete mathematics, automatic learning, data base design, pattern recognition, among others). Consequently, some of these techniques attempt to compare text strings that describe the entities in the ontologies (terminology-based ontology alignment), while others calculate the similarity measures between entities taking into account the structure of their corresponding ontologies (structural ontology alignment). A complete classification of alignment techniques has been developed by Martínez (Martínez, 2007).

Using a simple example, the following discussion illustrates some of the basic ontology alignment techniques that are currently used. In this example, two simple ontologies are examined, as shown in Figure 1.

The ontologies shown in Figure 1 describe various entities in the real world: sets of elements that share certain characteristics or *classes* (e.g., *Wing*, *Car*, *Bus*, etc.), *instances* of classes (*individuals*) and their *relations* (e.g. a specific *Ferrari F50* belongs to a

Figure 1. An example illustrating the alignment between two ontologies



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