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Towards the extraction of partial instances of N-Ary relations in textual data

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Abstract :

This paper presents a generic approach in order to extract experimental information from scientific documents in specialized domains. We are here describing the first phase of our work: the use of an *Ontological and Terminological Resource (OTR)* to research *partial instances* of *N-Ary relations*. The *OTR* drives the extraction with its domain vocabulary and guides the creation of a *Representation* of the results. Our contributions in this process is the identification of terminological variations and acronymic forms recognition to increase the coverage of the *OTR* vocabulary. The recognition and classification of text *Segments* (ie. document sections, figures...) is also used to contextualize the extraction of the *partial instances* of *N-Ary relations*.

Mots-clés : N-Ary relations, Ontological and Terminological Resource, Information Extraction in Specialized Domain, Terms Variation, Text Segments

1 Context

The ARTEXT4LOD project (n-ARy relaTions EXTraction for Linked Open Data — AR-TEXT4LOD) aims is to develop a method enabling the extraction of experimental data from a corpus of scientific documents in a specialized domain relying on an *Ontological and Terminological Resource (OTR)*. Our generic method is an extension of (Berrahou *et al.*, 2017).



Figure 1: Structure and arguments of an N-Ary relation (food packaging O^2 permeability)

The *OTR* we use is a structured representation of a specific domain, its different concepts and their relations. It defines the information we seek to extract as *N*-*Ary relations*, with a specific structure and *arguments*, and drives the different steps of the process. This resource also comes with a vocabulary associated with each of its concepts, useful to identify the terms of interest in the documents and their corresponding *arguments*.

One of the biggest hurdles we face is that the *arguments* of *N*-Ary relations are scattered throughout the document. Considering for example the *N*-Ary relation in Figure 1, we may encounter the *argument* concerning the *packaging* name in the *Introduction* section, its *thickness* in *Materials and Methods* and the actual O^2 permeability value in Results and Discussion.

This involves two necessary phases: (I) the extraction of *partial instances* of *N*-Ary relations and (II) the reconstitution of *complete instances*.

This paper focuses on the first phase (I), its general process and the contributions we are offering.

General Process 2

Experimental information are characterized by the presence of unit of measure. The extraction of *partial instances* of *N*-Ary relations (I) then start with the identification of Measure Units as defined in the OTR. We consider them as *pivot terms*, defining a context favorable to the presence of *arguments* in textual *Windows* formed by ± 1 sentences around *pivot terms*. This choice of *pivot term* offers the best agreement between quantity and quality of the results and is supported by previous research (Berrahou et al., 2017). To identify and extract the arguments we use features, associations of OTR vocabulary and concepts, to produce Representations (in example 1 the term "polyethylene" corresponds to the feature $\langle packaging \rangle$).

However, no OTR can be absolutely comprehensive and we set up two resources to create an expansion of its vocabulary. Identifying the variations of the terms contained in the OTR vocabulary is a simple way to allow the recognition of more *arguments* in the *Rep*resentation. For that, we pre-processed the corpus and the OTR vocabulary with FASTR (Bourigault & Jacquemin, 1999), a tool used to extract term variations. Scientific publications use acronyms to represent important terms and being able to identify them is essential. We adapt an acronym recognition/disambiguation method (Okazaki & Ananiadou, 2006) by driving it with the OTR. Term variations and acronyms are afterwards added to the *features* used to build the *Representations* and should increase its coverage.

Example 1

Window [...]OP of the polyethylenimine film at 50% RH were $0.60 * 10^{-18} m^3.mm^2.Pa[...]$

Representation $< quantity_OPacro > < packaging_{termVar} > < numval > < measure$ $Unit_RH > < quantity_RHacro > < numval > < measureUnit_OP >$

In addition, scientific publications are structured documents where the information is contextualized in text Segments (ie. sections, figures ...) (Shah et al., 2003). With a recognition and classification of these Segments (Hofmann et al., 2009) we can use terms frequency measures to associate a *confidence score*, based on *arguments* and *Segment* associations, to Representations. Thresholds will later be applied on *confidence scores* to filter the partial instances before phase (II).

3 Conclusion

This first phase highlights two contributions in the process of *N*-Ary relations extraction: the extension of an OTR concepts vocabulary with terms variation and acronyms recognition, and the consideration of the context in which the *arguments* are expressed. Our expectations are that these tasks will respectively improve recall and precision of the results: by an increasing number of arguments identified and the selection of the best partial instances before the reconstitution of complete *N*-Ary relations in phase (II).

References

- BERRAHOU S. L., BUCHE P., DIBIE J. & ROCHE M. (2017). Xart: Discovery of correlated argu-
- ments of n-ary relations in text. *Expert Systems with Applications*, **73**, 115–124. BOURIGAULT D. & JACQUEMIN C. (1999). Term extraction-i-term clustering: An integrated platform for computer-aided terminology. In Ninth Conference of the European Chapter of the Association for Computational Linguistics.
- HOFMANN K., TSAGKIAS M., MEIJ E. & DE RIJKE M. (2009). The impact of document structure on keyphrase extraction. In Proceedings of the 18th ACM conference on Information and knowledge management, p. 1725-1728: ACM.
- OKAZAKI N. & ANANIADOU S. (2006). A term recognition approach to acronym recognition. In Proceedings of the COLING/ACL on Main conference poster sessions, p. 643–650: Association for Computational Linguistics.
- SHAH P. K., PEREZ-IRATXETA C., BORK P. & ANDRADE M. A. (2003). Information extraction from full text scientific articles: where are the keywords? BMC bioinformatics, 4(1), 20.