Exploring predicate-arguments structures in texts to relate biological entities

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Abstract : Verbal and nominal predicate structures present interesting linguistic properties. In biological texts they express relations between biological entities. We describe PredXtract, a platform which extracts predicate argument structures, and present the results.

Mots-clés : Predicate structure, Nominalizations of verbs, Dependency grammar, Link Parser.

1 Introduction

This paper focuses on the parsing of verbal and nominal predicate structures, expressed in a great variety of ways (Meyers et al., 2004a). Defining a uniform representation for these structures is decisive to converge on a VerbNet or FrameNet representation (Wattarujeekrit et al., 2004; Miyao et al., 2006) and to acquire semantic relations. In this perspective, we have developed a robust platform, PredXtract, based on the Link Parser (Sleator and Temperley, 1991). This platform is a general tool, which extracts verbal and nominal predicate argument structures (PAS) in english texts. More specifically, it allows to exhibit relations between biological entities.

2 Classification of the predicates

Verbs and their nominalizations are the most productive predicates and have the same argument relations, where arguments play precise conceptual roles: subjects and complements, which are core arguments, and adjuncts. We present here seven important classes of predicates, defined from their core arguments. (i) Verbs accepting a direct object are grouped together in Class 1 and 2; in the corresponding predicate noun phrases (PNPs), the preposition *of* marks the direct object. (ii) In Class 3 to 5, verbs do not accept a direct object, and in the PNP the preposition of marks the subject. (iii) Class 6 and 7 concern symmetric predicates with interchangeable arguments.

Class 1: $N_0 V N_1 = N^{pred}$ of $N_1 by N_0$. Example: *IFN-gamma activates protein kinase C delta / activation of protein kinase C delta by IFN-gamma*. **Class 2:** $N_0 V N_1$ *Prep* $N_2 = N^{pred}$ of N_1 *Prep* N_2 *by* N_0 . Example: N_0

Class 2: $N_0 \ V \ N_1 \ Prep \ N_2 = N^{pred}$ of $N_1 \ Prep \ N_2$ by N_0 . Example: N_0 attributes a protein fragment to a sequence / attribution of a protein fragment to a sequence by N_0 .

Class 3: $N_0 V = N^{pred}$ of N_0 . Example: the femoral head necroses / necrosis of the femoral head.

Class 4: N_0 V Prep $N_1 = N^{pred}$ of N_0 Prep N_1 . Example: tryptophans fluctuates in gramicidin / fluctuation of tryptophans in gramicidin.

Class 5: N_0 V Prep N_1 Prep $N_2 = N^{pred}$ of N_0 Prep N_1 Prep N_2 . Example: temperature decreases from 200 K to 70 K / decrease of temperature from 200 K to 70 K.

Class 6: $N_a V$ with $N_b = N^{pred}$ of N_a with $N_b = N^{pred}$ of/between N_a and N_b . Examples: genes interact with proteins; interaction of genes with proteins / interaction of/between genes and proteins.

Class 7: $N_0 V N_a Prep N_b = N^{pred}$ of N_a with/to N_b by $N_0 = N^{pred}$ of/between N_a and N_b by N_0 . Examples: N_0 connects a new sequence with/to a cluster; connection of a new sequence with/to a cluster / connection of/between a new sequence and a cluster.

3 PredXtract, an extracting platform

PredXtract uses the Link Parser (LP) and its native Link Grammar (LG), a variant of dependency grammars (Sleator and Temperley, 1991). In LG, generic links attach verbs (MVp link) or nouns (Mp link) to any preposition which introduces an NP. In order to mark, in the parses, the precise role of each argument of the predicates, we have defined specific argument links, which are searched during the extraction process. PredXtract is constructed with the following components.

An extended lexicon. Following Szolovits (2003), we have added in the grammar all the words of the "Specialist Lexicon" (SL), which includes UMLS terms (http://www.nlm.nih.gov/research/umls). We have then added a lexicon of genes and proteins extracted from corpus.

A special PNP grammar. Several teams in biomedecine use the LP but without modifying its grammar (Ding et al., 2003; Hakenberg et al., 2009). According to our classification of the nominalizations, we have added to the native LP a grammar module of PNPs with 89 subclasses, where each subclass corresponds to a syntactic pattern with core arguments (including clauses with *that*) and adjuncts. Each nominalization belongs to one or more subclass.

The frame below gives an example of a short sentence with two nominalizations. The first one (*response*), concerns a subclass (noted **ni2**) of Class 4, where the preposition *to*, inherited from the verb, introduces the complement. The MSI

+0s+MCIT0+									
+Sp+		+D*u+MSI+-Jp-+			+Jp+MCDTWI-+Jp-+				
	- T		1		- I		- I	1	- I
we	examined.v	the	response.ni2	of	cells.n	to	treatment.ndt7	with	drugs.n

link marks the subject introduced by the preposition of, while the MCITO link marks the complement introduced by to. The second shows the prepositional use of *treatment*, here not saturated, corresponding to a subclass (noted ndt7) of Class 2, where the preposition with (link MCDTWI), inherited from the verb, introduces the complement.

A verb-noun alignment module. Rather than writing a grammar for verbs, which would have been very complex, we have defined a module that aligns verb arguments to nominalization arguments during a post-processing step. This module performs several tasks: (i) distinguish complements from adjuncts of verbs, by using the data of SL, and substitute the generic MVp link with a specific argument link when appropriate; (ii) identify each "verbal sequence", compound with a verb and a set of possible auxiliaries, negation, and modal verbs; (iii) identify arguments in passive or active voice, and interchangeable arguments.

A recognition module of predicate structures. For each parse of a sentence, all the predicates and their arguments are identified (argument links point on the heads of core arguments). Then the surface structure of each argument is reconstructed via the links, by using linguistic criteria. The reconstructed arguments can be NPs (most cases), clauses or adverbs.

A filtering module of parses. For each sentence, the parses (often several thousands) are re-ordered by attributing to each parse a score defined from several criteria. Among the main criteria: (i) a higher score is given to parses whose number of argument links is maximum in the case of multiple prepositional attachments to verbs or nouns; (ii) a specific score is calculated in the case of PNPs containing several nominalizations, to favour prepositional arguments attached to the head of the PNP.

We present below an example: from the sentence Hyperoxic exposure induced an S-phase arrest associated with acute inhibition of Cdk2 activity and DNA synthesis, 9168 parses were found and PredXtract outputs :

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Nominalization 1: exposure
Nominalization 2: arrest
subject or object: S-phase
Nominalization 3: inhibition
direct object: Cdk2 activity
direct object: DNA synthesis
Nominalization 4: synthesis
subject or object: DNA
Verb 1: induced (verbal sequence: induced ; active)
subject: hyperoxic exposure
direct object: an S-phase arrest associated with acute inhibition of [...] synthesis
Verb 2: associated (verbal sequence: associated ; passive)
direct object A: an S-phase arrest
direct object B: acute inhibition of Cdk2 activity and DNA synthesis
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This example shows a short sentence with six predicate structures. We can notice that (i) *exposure* has no argument, (ii) *inhibition* has two coordinated objects, (iii) the role of the argument of *arrest* and *synthesis* is underspecified (subject or object), and (iv) the verb *associated* has two interchangeable arguments (object A and object B).

4 Results and discussion

PredXtract has been evaluated with a corpus of 335 Medline abstracts given by biology researchers. From the 3,500 sentences of this corpus, we have selected 700 random sentences; 300 of them have been used to finalize our system and the evaluation has been done on the 400 others. Nominalizations represent 42.3% of all predicates. Because of the possibility of wrong segmentation of arguments, we have calculated two values for recall, precision and F-measure, with: [Case 1] only the true and complete arguments, [Case 2] the true and complete arguments and the true but incomplete arguments. In Case 1, the F-measure score is 78% for nominalizations (precision: 79%; recall: 77%) and 77% for verbs (precision: 78%; recall: 77%). In Case 2, it is 85% for nominalizations (precision: 79%; recall: 77%) and 88% for verbs (precision and recall: 88%). Thus, we observe a very small difference between values for nominalizations and verbs.

Much research has been published on predicate argument structures but it is difficult to compare research because objectives are often different. In biomedicine, research focuses on PAS dedicated to gene /protein interaction, where two genes or proteins are in a subject and a complement position in a proteomic relation. For example, McDonald et al. (2004) obtain a precision rate of 89% and a recall rate of 61% with a complete parsing; Leroy et al. (2003) use a shallow-parsing with finite state automata and obtain 90% of precision; Huang et al. (2004) have a precision rate of 80.5% and a recall rate of 80% with a pattern-matching processing. Few studies process nominalizations. Leroy et al. (2002) use templates built around a set of prepositions to capture relations with genes, proteins, gene locations, diseases, etc., with a precision of 70%. A specific work on PP attachments on nominalizations (Schuman and Bergler, 2006) in proteomic texts achieves good results (precision: 82%) with linguistic heuristics using information of "Specialist Lexicon" nominalizations, but the system does not produce information on the PP roles (subject, object or adjunct). Concerning nominalizations in other texts than biology, the first version of NOMLEX is used in information extraction (Mevers et al., 1998). The NOMBANK project (Mevers et al., 2004b) annotates automatically, semi-automatically and manually, in corpus (the Wall Street Journal Corpus of the Penn Treebank), predicate nouns (verbal, adjectival and other) with their argument relations and improves the lexical base of predicate nouns (NOMLEX-PLUS).

As PredXtract is based on very large lexicons, it is possible to say that PredXtract is a platform which extensively recognizes PAS, independently from the predicate type. To refine PredXtract outputs, the next step will need the annotation of arguments with UMLS or other ressources term types.

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