AnaPro, Tool for Identification and Resolution of Direct Anaphora in Spanish

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ABSTRACT
AnaPro is software that solves direct anaphora in Spanish, specifically pronouns: it finds the noun or group of words to which the pronoun refers. It locates in the previous sentences the referent or antecedent which the pronoun replaces. An example of a direct anaphora solved is the pronoun “he” in the sentence “He is sad.” Much of the work on anaphora has been done for texts in English; thus, we specifically focus on Spanish documents.

AnaPro directly supports text analysis (to understand what a document says), a non trivial task since there are different writing styles, references, idiomatic expressions, etc. The problem grows if the analyzer is a computer, because they lack “common sense” (which persons possess). Hence, before text analysis, its preprocessing is required, in order to assign tags (noun, verb,…) to each word, find the stems, disambiguate nouns, verbs, prepositions, identify colloquial expressions, identify and resolve anaphora, among other chores.

AnaPro works for Spanish sentences. It is a novel procedure, since it is automatic (no user intervenes during the resolution) and it does not need dictionaries. It employs heuristics procedures to discover the semantics and help in the decisions; they are rather easy to implement and use limited knowledge. Nevertheless, its results are good (81% of correct answers, at least). However, more tests will give a better idea of its goodness.

Keywords: I.2. Artificial Intelligence, I.2.7 Natural Language processing, Text Analysis, Anaphora resolution.

RESUMEN
AnaPro es un software que resuelve problemas con anáforas directas en español, especialmente pronombres: la herramienta encuentra el sustantivo o grupo de palabras al cual se refiere el pronombre. Localiza en oraciones previas la referencia o antecedente y lo reemplaza por el pronombre que le corresponde. Un ejemplo de anáfora directa resuelta es el pronombre “él”, en la oración “Él está triste”.

AnaPro apoya directamente el análisis de textos (para entender lo que el documento dice), tarea no trivial ya que existen diferentes formas y estilos de escribir, referenciar, expresiones regionales (mexicanismos, argentinitismos, etc.). Todavía se complica más si el analizador es una máquina porque no tiene “sentido común” (que cualquier persona puede tener). Así, antes de analizar el texto, se requiere pre-procesarlo para asignar etiquetas (sustantivos, y verbos, entre otros) a cada palabra, encontrar las derivaciones, desambiguar sustantivos, verbos, preposiciones, identificar expresiones coloquiales, identificar y resolver anáforas, entre otras tareas.

AnaPro trabaja con oraciones en español. Es un proceso novedoso porque es totalmente automático (no requiere la intervención de un usuario) y tampoco necesita de diccionarios. Emplea un procedimiento heurístico para descubrir la semántica y apoyarse en las decisiones que debe de tomar. Ha resultado ser bueno (con un 81-100% de aciertos); sin embargo, más ejemplos nuevos darán mejor idea de su desempeño.
1. Introduction

Anaphora is a relation of coreference between linguistic terms. According to Webster's dictionary: “It is the use of a grammatical substitute (as a pronoun or a pro-verb) to refer to the denotation of a preceding word or group of words; also: the relation between a grammatical substitute and its antecedent.” Therefore, anaphora is a discourse relation. Anaphora resolution is very important in Natural Language Processing (NLP).

This work is part of Project OM* (Ontology Merging), which seeks to build a large ontology by fusing smaller ontologies extracted from textual documents. An important part of the project is to analyze the sentences in a document with the goal to transform that text into an ontology that comprises its contents. A brief description of Project OM* follows.

1.1 Project OM*

Project OM* is built upon the OM software (Ontology Merging; [3]), which takes two ontologies as input and creates another ontology comprising the knowledge residing in the inputs, without self-inconsistencies. The OM software has been built and it works [3], but it takes as input ontologies (in a special notation called OM notation, explained below), instead of textual documents. Thus, additional work is needed to transform a document into an OM ontology to be fed to OM. That transformation is the goal of OM*. Project OM* is in full development. The parts of OM* that have been built appear in Figure 1, as well as a simplified flow of its computation. These are:

Figure 1. The parts of the OM* project. Parts under constructions are shown dashed. The diagram shows a simplified flow of computation from text to ontologies.
Idioms analyzer. An idiom is a phrase whose meaning cannot be derived from the conjoined meaning of its elements. For instance, vestida y alborotada {dressed and agitated}1 (meaning: decepcionada {adj. disappointed}); de tal palo tal astilla {chip off the old block} (adj. similar to his ancestor); corazón de piedra {stone heart} (adj. to be hard and inflexible). The idioms analyzer [12] finds and replaces them with their respective synonym. This increases the ability of subsequent modules of OM* to understand more clearly the sentences. We believe it is the first idioms finder for Spanish. It uses a dictionary of canonical idioms (stereotyped phrases) which are then compared with the text under analysis, looking for “legal” variations of the stereotype phrase.

The analyzer locates and converts the following types of idioms:

1. Colloquial expressions. Example: corazón de piedra {stone heart} (adj. to be hard and inflexible).
2. Specialized expressions. They describe an object in a specific or particular (specialized) context. Examples: línea de comandos {command line}, base de datos {data base}.
3. Proper names. Examples: Banco de México {The Bank of Mexico}.
4. Connective phrases, used to connect ideas in a sentence. Examples: en base a {based upon}, sin embargo {nevertheless}. These are classified as idioms since for instance the meaning of sin embargo cannot be derived from its constituents sin {without} and embargo {seizure, foreclosing}.

Tagger. It assigns tags or labels (verb, noun ...) to each word in a phrase (Tagger TNT, Internet reference 7). Many current systems process natural language using a morphological tagger. The TNT tagger assigns a single tag (grammatical category) to each part of speech. This tag is used in subsequent levels of analysis, such as the syntactic parser. Since idioms are processed before the tagger acts (see Figure 1), tagging becomes better because idioms are already correctly tagged (by the idioms analyzer).

Usually, there are two ways to achieve morphological tagging. Recent comparisons in training sets and corpuses have shown that in most cases the taggers that use statistical approximations produce better results than finite state taggers, or those based in rules, or using an extensive a priori storage. The tip of the balance has not yet settled.

Anaphora solver. It finds and replaces the pronouns by the corresponding linguistic entity. It is AnaPro, the subject of this article.

Lemmatizer/Stemmer. It finds the lemma or root of each word. More details in [8].

Prepositions disambiguator. Reference [7] describes an algorithm and its implementation to disambiguate prepositions in Spanish phrases. This article focuses on the design of a preposition disambiguator, using a verb and noun lemmatizer that employs frames to identify the meaning of verbs and nouns. It finds the semantics of the prepositions contained in parts of speech. For instance, for the preposition de {of}, it finds that de in olla de madera {wooden pot} means “made of”; in olla de arroz {pot of rice} means “contains”, whereas olla de Oaxaca {pot from Oaxaca} means place of origin.

Prepositions are words used to relate ideas composing a sentence; hence, by themselves, they lack expressive meaning of their own, but they contribute in a fundamental manner to provide sense to the whole sentence since they establish a semantic relation. Many times, prepositions are wrongly used and they form part of the vices of a language.

This work studies the separable prepositions, which are a {to}, ante {before}, bajo {under}, cabe {it should be noted}, desde {from}, contra {against}, en {in}, por {by}, según {as}, sin {without}, entre {between}, so {under}, con {with}, hacia {towards}, sobre {upon}, hasta {until}, de {of}, tras {behind}, para {for}.

Prepositions are classified in groups according to the sense they provide to the expressions they qualify, although some prepositions belong to more than one group:

1 AnaPro is an analyzer of anaphora in Spanish. In order to help some readers, the authors (not AnaPro) have added translations to English for selected Spanish sentences. They can be omitted.
1. Place prepositions: They assign a place: a {to}, de {from}, en {in}, entre {between}, hacia {towards}, por {by}, tras {behind}.

2. Time prepositions: An event is placed on the timeline: a {to}, con {with}, de {of}, desde {since}, en {in}, para {before}, por {by}, sobre {after}.

3. Causal prepositions: They provide a link between origin and result: de {of}, por {by}.

4. Purpose prepositions: They determine a target or objective: a {to}, para {for}.

5. Company prepositions: They express association of subjects with respect to an action: con {with}.

6. Instrument prepositions: They express the means by which the action is or has been performed: a {to}, con {with}, de {of}, en {in}.

7. Modal prepositions: They indicate the way in which the action is carried out: a {to}, con {with}, de {of}, en {in}, por {by}.

More details for the algorithm to disambiguate prepositions are in [7].

**Word disambiguator** [9]. Mapping words to concepts. This work determines the correct meaning of ambiguous words in Spanish texts, using the context in which the ambiguous word is found. It presents a new non-supervised computational approach to process ambiguous words in Spanish documents. It selects the appropriate word sense based on the context. The process is also known as Disambiguation or Word Sense Disambiguation. The model is based on the Lesk Algorithm with significantly modifications that enhance its performance. This approach does not need a previous markup of the text to analyze. The software detects automatically the ambiguous words within the text and defines its correct sense. The results obtained are evaluated with the highest international criteria with significant success.

To achieve its objective, the method of word coincidence was employed using:

1. A dynamic (moving) window of varying width as context, formed by words to the left and right of the ambiguous word.

2. The definitions (different senses) of the ambiguous word.

3. A lemmatizer applied to the words in the window and to the words in the definition of the ambiguous word.

The algorithm does not require supervision or training in order to work. It also does not require a set of tagged texts, nor the listing of the ambiguous words to be found.

**Frames** [17] give relevant details of concepts, generally nouns (agents) and verbs (actions). They are data structures containing semantic information about the concept: such as what it is, its use, its parts, when it happens, and so on. Frames help the prepositions disambiguator, the disambiguator of nouns and verbs, and the syntactic/semantic analyzer. Frames were perhaps introduced by [17], who defines them as data structures to represent generic situations, such as the representation of a type of behavior and things expected in a living room, in a restaurant, in a children’s party. Several kinds of information are found in frames: how to use it, what activities or event are performed, what other events occur after the frame occur after the frame (a subsequent frame). A frame can be considered as a network of nodes and relations. Frames represent scenes, for instance: the frame representing a children’s party has more semantics that the definition in a dictionary: [fiesta -- Reunión de gente para celebrar algún suceso, o simplemente para divertirse (Diccionario de la Real Academia Española). party: a social gathering; also: the entertainment provided for it. (Webster Dictionary)].

Table 1 elaborates the frame “children party”.

Many concepts may be related to a frame, and they will enrich it semantically; therefore, most of the information intertwined in the represented scenario (concept, event, or activity) must be present in the frame. There is no standard notation about frames; we have proposed one. It is formed by:

1. **Header**, representing the name of the concept, a character indicating if it is a noun (s, for sustantivo noun)), verb (v), etc., and in parenthesis, its gender.

2. **Body**, containing elements such as: quees (what it is; its meaning), agenteactivo (active
3. Our notation allows adding semantic elements as needed, for instance the element *lugar* (place).

More details in [7] and [8].

**Dress:** better than everyday dresses.
**Gift:** Must be wrapped and given to the birthday child.
**Celebrating** hide-and-seek, guessing riddles, jokes.
**Decoration:** balloons, figures made of crepe paper.
**Cake:** candles, wishes, to blow.
**Songs:** Happy Birthday.
**Place:** Party room, home, kindergarten.

Table 1. Part of the frame “children’s party”

**Syntactic/Semantic parser.** It analyzes each phrase, taking into account the syntax, as well as the semantics of each part of speech (given by the frames, the disambiguated prepositions, nouns, etc.). It finds the correct dependency between nouns, verbs, and articles. Basically, it is divided into three parts:

Syntactic Parser. We use FreeLing [24]. It builds a dependency tree mostly based in the syntax of each sentence.

Congruency detector. It analyzes each part of the dependency tree, and, with the help of the frames and other semantic elements, it measures how consistent that part is (how much sense it makes).

Generator of plausible moves. The parts of the dependency tree with low congruency are replaced by other candidates, which the generator proposes. Several changes of node position in the dependency tree are suggested, directed by the frames and measured by the congruency detector.

The congruency of the resulting replacement should be larger than the replaced portion of the tree.

The congruency detector and the generator of plausible moves are under construction in Lisp.

**SERCDD** [13] is an analysis system that extracts information from “descriptive documents”: documents describing an object (a tool, a planet), and uses the concepts extracted to form a semantic network. Its output is an ontology in the OM language, denoting the object described in the document. The analysis method starts with text which has previously suffered tagging and lemmatization. Using the structure of the sentences found in the text, analysis proceeds trying to identify the relations present in it. For instance, a text describing a carpenter tool will usually contain its definition, a description, common uses, parts and materials forming it, as well as the classification of such tool. The result is a formal representation of the extracted knowledge, embodied in an ontology written in the OM language, described in [2]. In order to produce such analogy, it is necessary to identify the entities (concepts), relations and properties described in the original text.

SERCDD finds ontological relations (hyponymy, hypernymy, part of, synonymy, size, made of, useful for…) by searching for text patterns and using a set of rules derived from the respective semantic relations. For the time being, these patterns and rules are manually constructed, but another part of OM* seeks to automatically derive them from the frames representing the semantics in the objects. The system extends the use of patterns and rules, which are usually used just to find hypernymy relations (the hammer is a tool, the mechanic hammer is a hammer…), to many other relations.

SERCDD processes documents written in Spanish, but it is easily extensible to other relations and other natural languages. It performs tagging and lemmatizing, prior to seeking ontology relations.

**Ontology generator.** This is a straightforward module that converts the correct dependency tree produced by the syntactic/semantic parser in an ontology expressed in OM notation. It has not been built yet.
Ontology Fusion. In order to compute intelligent answers to complex questions, using the vast amounts of information existing in the Web, computers have (1) to translate such knowledge, typically from text documents, into a data structure suitable for automatic exploitation; (2) to accumulate enough knowledge about a certain topic or area by integrating or fusing these data structures, taking into account new information, additional details, better precision, synonyms, homonyms, redundancies, apparent contradictions and inconsistencies found in the incoming data structures to be added; and (3), to perform deductions from that amassed body of knowledge, most likely through a general query processor.

OM [2, 3] solves point (2) by providing an algorithm to fuse two ontologies (coming from Web documents) without human intervention, producing a third ontology containing the knowledge residing in the input ontologies. It resolves synonyms, homonyms, articulates partitions, detects and solves some contradictions (inconsistencies), and removes redundant relations. Also, the accumulation of knowledge requires that a fair amount of ontologies about the same topic should be merged, which calls for repeated use of OM.

The correct merge must consider not only the syntax of the word and phrases describing the concept, but the semantics of the concept thus represented, too [its “meaning” as described by its relations to other concepts in the source ontologies, its similarities to other concepts, superficial likeness, homophones, etc.] Therefore, the representation to be used (OM Notation) should allow the characterization of these relations.

OM Notation. [2] proposed this ontology definition language with the goal to represent concepts and relations so that more semantics is captured in them. OM means “ontology merging,” since the language was first used to fuse or merge ontologies. It is a derivative of XML, described in [3].

The OM language structures ontologies through the following tags:

- `<concept>`, used to define a concept. This tag allows concept nesting.
- `<language>` represents a brief sentence (a gloss) in natural language describing the concept. It is used mainly for human consumption, not by the OM processor.
- `<word>` is followed by that brief sentence or gloss. For instance, the concept *escalpelo* {scalpel} can be defined as
  - `<concept> escalpelo</concept>`
  - `<language>` Spanish `<word>` escalpelo, bisturí `<word>` and English `<word>` scalpel, a small straight thin-bladed knife used especially in surgery `<word>`
  - `<language>`
  - … here will go the relations pertaining to escalpelo <concept>.
- `<relation>` represents the type (name) of a relation (feature, property, usually obtained from a verb) that connects two concepts.

1.2 AnaPro

This paper describes AnaPro, one of the modules of OM. AnaPro determines the direct anaphora (that resolves pronouns). Its output eases the building of the part of the ontology that corresponds to the text being analyzed. For instance

*Juan tomó una botella en la tienda. Ésta estaba rota.*

{John took a bottle in the store. It was broken.}

A person easily finds that the broken thing was the bottle, not the store, but a program may have difficulty to find out to which of the nouns *botella* {bottle} or *tienda* {store}, the pronoun *ésta* {it, this} refers to. In the example, a person asks herself: *what can be broken?* The bottle, most likely. This is so because a person has common sense, experience, previous knowledge, etc. For the computer, analysis and verification of several information sources and tools (frames, dictionaries, thesaurus, etc.) is usually needed in order to have a satisfactory solution.
Anaphora resolution has been an important task for NLP during years. Because of its importance, the problem has been tackled from several fronts. For instance, [10] uses semantic criteria, as described in §3. Since our work is for Spanish documents, our descriptions of previous works will be primarily focused on Spanish, too.

[22] and others have proposed methods that try to resolve anaphora resolution (both direct and indirect) in Spanish, using dictionaries. Their methods do not solve all grammatical forms.

Nevertheless, simple methods using limited knowledge can be applied in many cases with satisfactory results. For instance, text from a restricted domain or culled from a controlled vocabulary. An example is [20], who solves anaphora with the help of the SUPP (Slot Unification Partial Parser) analyzer, which provides lexical and syntactic information to find out the antecedent of higher relevance in the list of antecedents (one of them is the referent). Although with some limitations, the work proposes interesting improvements to the algorithm, as described in §3.

Another approach suggests to use linguistic and structural knowledge which are domain-independent, as in [11]. It deals with anaphora in Spanish identified by third person pronouns and adjectival anaphora. Described in §3, it uses a statistical model to determine the complements of verbs, adjectives and some nouns, based on a corpus.

An important work, since it deals with unrestricted domains, is that of [4]. It locates and extracts information where answers to questions can be inferred. It does this in Spanish and in English. It is briefly described in §3.

AnaPro only solves direct anaphora, particularly pronoun resolution. Of the already mentioned previous works, only [11] resembles our solution.

The paper is hence organized as follows: §1 describes the state of the problem and related previous work; §2 gives the theoretical framework; §3 presents work related to resolution of direct anaphora and places AnaPro in this context. §4 details our solution and solves some examples. §5 explains how AnaPro works; §6 provides more examples taken from “real life” (Internet documents).

2. Theoretical Framework

Anaphora resolution has been the focus of attention of philosophers, linguists, cognitive and artificial intelligence scientists, psycholinguists and computational linguists. This is important because the anaphora:

- It is one of the most complex phenomena in natural language.
- It has been shown that syntactic, semantic and pragmatic factors take part in it.
- Its resolution is needed in a wide range of NLP tasks: natural language interfaces, document understanding [for the computer to answer complex questions about the knowledge that lies in them (this is the goal of the aforementioned Project OM*)], machine translation, information extraction, automatic summarization, finding differences or contradictions between statements on the same subject in two documents, among others.

Two important types of anaphora exist: direct and indirect anaphora.

Direct anaphora associates a pronoun or reference to some entity already mentioned in the text. For instance, los alumnos estaban contentos pues ellos habían pasado a la semifinal {students were happy because they had passed to the semifinal}; here the pronoun ellos {they} refers to los alumnos {the students}.

Indirect anaphora establishes an associative link between a linguistic entity (word or phrase) and an entity previously introduced through the text or the speech [22]. For example, Eliseo entró a un circo, el payaso era muy simpático {Eliseo went to a circus, the clown was very nice}. In this example the discourse focuses on the circus and it is the previous entity or antecedent to which payaso {clown} belongs or refers. A connection exists between payaso and the circus where Eliseo went, which has at least one clown. This information comes from the extension of the context by encyclopedic knowledge (common sense); as a consequence, the whole sentence renders a consistent interpretation, as implied by the relevance principle, used above, or by the scenario...
approach (frames, [17]), where the use of circo (circus) invokes a scenario (the frame of circus) that implicitly contains at least one clown.

AnaPro does not analyze indirect anaphora; this paper does not deal with it. It is mentioned here for completeness.

2.1 Contributions

AnaPro converts a Spanish text into another text without direct anaphoric references. The anaphors are solved, replacing for instance ellos (they) by estudiantes (students). The final text, without anaphoric references, is an input to be used in project OM*. Some of its contributions are:

1) The focus of AnaPro is to find (automatically) the noun to which a determiner refers. Other works such as [22] require a user. Instead, [20] needs an information detection system to determine the most relevant preceding topics.

2) Another important point of AnaPro is that it resolves anaphora in complete texts, not in short dialogues [20]. Therefore, full text can be entered into the tool for its analysis and resolution.

2.2 Determiners

Determiner is a term that denotes the lexical unit that precedes a noun in a nominal phrase to specify its reference, including the quantity of the noun. For instance,

1.-Todos esos vehículos están en venta. (All those vehicles are for sale)

The determiner is Todos (all) which determines the quantity, and esos (those) is a demonstrative pronoun that indicates the place, relative to the emitter, where the vehicles are.

In general, the determiners give rise to the definite expressions where these parts of speech are used: the definite articles (el, la, lo, las, los) {the, it}; the demonstratives (aquel, aquella, aquellas, aquellos, esa, esas, ese, esos, esta, estas, este, estos, tal, tales, semejante, semejantes) {that, that, those, those, that, those, that, those, this, these, this, these, such, such, like, like}; the possessive determiners (cuya, cuyas, cuyo, cuyos, mi, mis, nuestra, nuestras, nuestro, nuestros, su, sus, vuestra, vuestras, vuestro, vuestros, tu, tus, etc.) {whose, whose, whose, whose, my, my, our, our, our, his or her, his or her, your, your, your, your, your, your, etc.}, and the quantifiers (todo, algún, cada) {all, any, each}. With any other determiner, an expression is considered an indefinite expression. This is important since in Spanish the artículos indeterminados (articles a, an) give rise to indeterminate expressions.

2.3 Reference resolution process

The task of whoever receives the information is to conduct the reference resolution process, which is achieved with the following sequence of steps:

1. To identify as accurately as possible the entity to which the emitter (or writer) is referring (the referred) with the referential expression.

2. To determine if this entity has already been previously mentioned (or referred to) in the context of the discourse, or if it is new.

3. If it has already been mentioned (coreference), it will establish the link between referential expressions in the linguistic context; the first entity already mentioned has the link to the entity in the discourse context.

4. If it is new, it should be integrated it as part of the linguistic context, creating the link from the referential expression to the entity in the linguistic context.

The whole process must be based on the morphological and syntactic features of the text, which involve the use of determiners, and where the concepts of referential expression, coreference and direct and indirect anaphora (already defined in §2) are closely related.

3. Related work

For the English language, there is plenty of work in anaphora resolution. Since AnaPro works for Spanish texts, we will not review many of those works. A comprehensive analysis of the state of the art can be found in chapter 21 of [6]. We also refer the reader to the excellent surveys by [23], and, more recently, [19]. Some interesting anaphora solvers for English are reviewed here.
The Resolution of Anaphora Procedure (RAP, [15]) identifies the noun phrase antecedents of third person pronouns and lexical anaphors (reflexives and reciprocals). Written in Prolog, its inputs are the syntactic representations generated by McCord’s Slot Grammar Parser. RAP focuses its attention on (a) the syntactic structure of the sentence (as given by the Grammar Parser), from which it extracts salience measures, and (b) a rather simple intentional model for the attentional state. These are used to select, from a list of candidates, the noun corresponding to the pronoun. In evaluating candidate antecedents, it does not employ semantic conditions (other than those implicit in grammatical number and gender agreement) or real-world knowledge. Also, RAP does not model intentional or global discourse structure. RAP has been tested and has an 86% efficiency; it exhibited a slightly higher rate of success (by 4%) than Hobb’s algorithm. Its authors enhanced RAP with statistically modelled information about semantic and real-world relations, but these enhancements only caused a slight improvement (by 2%) in the algorithm’s performance. The main components of RAP are:

- Syntactic filter. It rules out anaphoric dependence of a pronoun on an NP on syntactic grounds.
- Morphological filter. It rules out anaphoric dependence of a pronoun on an NP due to disagreement in person, number, or gender.
- Pleonastic pronouns. A procedure to identify pleonastic (semantically empty) pronouns.
- Lexical anaphors. An algorithm that identifies the possible antecedent binder of a lexical anaphor (reciprocal or reflexive pronoun) within the same sentence.
- Salience evaluator. Values are assigned to several salience parameters of an NP: grammatical role, frequency of mention, proximity, sentence recency, parallelism of grammatical roles.
- Equivalence class identifier. Anaphorically linked NPs are identified and placed into an equivalence class. For this class, a global salience is computed adding the individual saliences of its members.
- Decision procedure. It selects the preferred element from a list of antecedent candidates for the pronoun.

Compared to RAP, AnaPro has a simpler design, which nevertheless works quite well for Spanish texts.

The work Anaphora for everyone [1] extends the previously described work of Lapping and Leass, producing similar results. The main difference is that it does not require an in-depth, full, syntactic parsing of the text. Instead, it uses a part-of-speech tagger, enriched with annotation of the grammatical function of the lexical terms found in the input sentence. Thus, it can be used in a wide range of text processing frameworks for which there are insufficient or inexistent full syntactic parsing capabilities. It requires the tagger to provide, in addition to the tag, an integer for each token in the input sequence (the offset of such token). For instance, for the input sequence “For 1995 the company set up …”, the output of the tagger is

```
"For/off139" "for" PREP @ADCL
"1995/off140" "1995" NUM CARD @<P
"the/off141" "the" DET CENTRAL ART
SG/PL @DN>
"company/off142" "company" N NOM
SG/PL @SUBJ
"set/off143" "set" V PAST VFIN
@+FMAINV
"up/off144" "up" ADV ADVL @ADVL
...
```

For anaphora resolution, the primary input data set is a complete list of all noun phrases, reduced to modifier-head sequences. The offset indicates the position of the noun phrase in the text, hence providing important information for precedence relations. The algorithm also uses information about the syntactic contexts in which the noun phrase appears. These observations are obtained by the use of patterns that detect nominal sequences that occur locally to adverbs, relative pronouns, and complementing sequences. These patterns are regular expressions. In terms of overall strategy, the resolution procedure follows that of [15].

The most common types of errors of the algorithm consists in gender mismatch (35% of the errors)
and certain long range contextual (stylistic) phenomena, for instance, quoted passages in-line (14% of the errors). The authors correct these mistakes by implementing a gender disagreement filter. Such errors simply reflect the lack of a consistent gender slot in the tagger output.

AnaPro resembles this work in several aspects, notably in the use of a tagger and regular expressions for extracting additional positional and relative information.

- The Anaphora solver of [21] is notable because it works using the Wordnet ontology coupled with heuristic rules. It identifies both intra-sentential and inter-sentential antecedents of anaphors. The judicious analysis of the hierarchical relation of nouns and verbs in the surrounding text supplies information about animacy (animacy denotes the living entities, notably persons and animals, which can be preferred by some gender-marked pronouns). This is important, because the identification of animacy entities and of the use of pleonasm (for instance, pleonastic pronouns that do not have an antecedent to refer to) contributes to a higher resolution accuracy.

The main procedure has these steps:

- The text is parsed by a POS tagger, which provides essential information (sentence offset, word offset, word POS, base form, etc.) for subsequent processing.

- (a) An NP finder module finds base noun phrases in each sentence. Also, (b) capitalized nouns, if found in a name gazetteer, identify person names. (c) Number agreement is implemented on the head noun. (d) Gender feature is attached to each name, if uniquely rendered. WordNet is useful here to provide gender clues. If the noun can be masculine or feminine, the gender attribute is ignored.

- Anaphors are checked sequentially from the beginning of the first sentence. Also, pleonastic-it is checked, to avoid its resolution.

- The remaining noun phrases provide antecedent candidates for resolution, which are identified using predefined window sizes. They are further refined by gender and animacy agreement. WordNet is useful here for the recognition of animate entities, which are usually those hyponyms of {animal, fauna} and {person, human being}. The procedure is not so straightforward since a noun may have several senses. WordNet is also used for finding hyponyms of certain verbs starting from the synsets {cognition}, {communication}, {emotion} and {social}.

- The remaining candidates are evaluated using heuristic rules. They are of two types: preference rules and constraint rules. A formula assigns a score to each candidate, taking into account the candidate noun phrase, the anaphora to be resolved, the agreement between number, gender and animacy, and the weight of the rule.

3.1 Anaphora resolution for Spanish texts

There are some studies that have addressed the problem of identifying anaphoric references in Spanish texts. The focus of those studies, their resolution algorithms, and their limitations are briefly mentioned.

- The Open Domain Question Answering Systems [4] are tools capable of getting concrete answers directly from natural language documents over unrestricted domains, in response to precise information needs from users. At least, they try to find and extract those areas (of the documents) from whose contents an answer to a specific question can be obtained (by hand). Therefore, those systems try to reduce the time invested by users to find concrete information. They have three modules. The first module is a standard Information Retrieval system, collecting documents that may be relevant to the query. The second module, which uses SUPAR (Slot Unification Parser for Anaphora Resolution\(^2\)), has as input the queries and the retrieved documents. The third module is a Question Answering system, which also interacts with SUPAR. It divides the sentences of the documents in order to find those where the correct answer appears. It is in this second module where the pronominal anaphora is resolved, using a lexical analyzer, a syntactic analyzer and a module that solves

\(^2\) SUPAR also has three modules: lexical analysis, syntactic analysis, and linguistic problem resolution.
anaphors, extra-position, and ellipsis, when the result is a slot structure (refer to the description of Martínez’ work at the beginning of Section 3) corresponding to the input sentence.

AnaPro does not use slot structures. Its resolution mechanism is detailed in next section.

- [18] presents an algorithm for resolution of the anaphora of a phrase (for third person pronouns, $PP_R$), demonstrative pronouns, $DP_R$, reflexive pronouns, $RP_R$ and missing pronouns (cero pronouns, $OP_R$) in a prepositional phrase, $PP$, or if they also complement personal pronouns (clitic pronouns) in non-restricted Spanish texts. There are restrictions and preferences for the different types of nominal expressions. In the resolution of Spanish anaphora, different kinds of knowledge are considered: lexical, morphological, syntactic and linguistic. It also gives a partial parsing for pronouns that are not coreferential. The algorithm has been tested on a corpus of 1,677 pronouns with a success of 76.8%. The algorithm can be easily extended to other languages such as English, Portuguese, Italian, and Japanese.

The algorithm finds the different types of anaphora from left to right, as they occur in the sentence. The main steps of the algorithm are:

1. Identification of the type of pronoun.
2. Restrictions (a) morphological (person, gender and number); (b) syntactic conditions in NP –non-coreferential pronouns.
3. Preferences. It is defined as a characteristic or feature that does not always satisfy the solution of an anaphor (it is sometimes unable to resolve it).

Unlike other works that resolve some types of anaphora using restrictions, and other types using preferences, this work uses them together, since the analysis flows from left to right in the sentence. For instance, for pronominal anaphora, the antecedents need to match person, gender and number. Otherwise, they are discarded.

This work is quite close to AnaPro; the main difference is that AnaPro uses somewhat simpler mechanisms.

- [20] uses as input the slot structure generated by the SUPP analyzer. This structure stores lexical, syntactic, morphologic and semantic information of each element of the grammar, and it also includes an anaphora detection mechanism, as well as another mechanism for its resolution. This last method uses the information from the topic detection system in order to determine the most relevant antecedent among the general topics list. The topics information joins the information obtained from other sources to get the best antecedent. In the same manner, once the anaphora has been resolved, the antecedent occupies the place left by the expression. This system works via dialogues.

Next, it is a simple example in which Martínez’ algorithm [20] is applied to the resolution of anaphora. It is a small dialogue where the system of topics defines three spaces for anaphoric accessibility: a) the space of the actual intervention, b) the space of the previous intervention, and c) the relevant topic in the current dialogue space. The first and second spaces are directly determined thanks to the labeling of turns characteristic of dialogues. The third space is identified by two coefficients that compute the use or lack of use of an entity; that is to say, the importance or irrelevance of an entity occurring in the current intervention.

The algorithm uses three lists to store the entities that can be an antecedent (according to the previously defined spaces). The list of current local entities, $Fa$, contains the entities occurring in the current intervention. The list of previous local entities, $Fa'$, contains the entities that occurred in the last intervention. The list of general topics, $F$, contains the entities relevant in the dialogue, evaluated according to frequency and distribution.

T01: <H1> Buenos días. {Good morning}
T02: <H2> Buenos días. ¿Qué deseas? {Good morning. What do you want?}
T03: <H1> Quiero manzanas. {I want apples}
T04: <H2> ¿De qué clase las quieres? {What kind do you want them?}
T05: <H1> No importa si son buenas. {It does not matter as long as they are good}
T06: <H2> Éstas las he recibido esta mañana. Son muy buenas. {I have received these this morning. They are very good}
The algorithm starts by considering turns T01 and T02 as continuation turns since they do not contain any entity. Thus, the three lists remain empty until T03. When processing T03, manzanas {apples} is considered an entity, and is hence introduced in the list of current local entities Fa:

\[ Fa=\{\text{manzanas}\} \]

T03 is therefore considered an intervention. When it finishes, the entities in Fa are incorporated into the list of general topics F with their corresponding weight, and they are also stored in the list of previous local entities Fa'. After these operations, the lists are:

\[ Fa'=\{\text{manzanas}\} \]
\[ F=\{\text{manzanas, 10}\} \]

In T04, after the incorporation of clase {kind} as an occurring entity, the first anaphora is detected, as the pronoun las {them}. At this point the lists are:

\[ Fa=[\text{clase}] \]
\[ Fa'=\{\text{manzanas}\} \]
\[ F=\{\text{manzanas, 10}\} \]

To resolve the anaphora, the algorithm searches, first, an antecedent in the list of current local entities (Fa). It finds the entity clase {kind}; however, the intervention of the system of restrictions rejects it, since it does not meet the required morphological characteristics (it is looking for a plural antecedent). In this way, since Fa does not contain more candidates, the algorithm will look into the previous local entities list, Fa', finding manzanas {apples}. After restrictions are applied, manzanas is proposed as antecedent for the pronoun and it is included as an entity in Fa:

\[ Fa=\{\text{clase, manzanas}\} \]

After processing the first sentence of T04, Fa' is emptied:

\[ Fa'=\{\} \]

And when T04 is finished, the lists are:

\[ Fa=\{\} \]
\[ Fa'=\{\text{clase, manzanas}\} \]
\[ F=\{\text{manzanas, 20}, \text{clase, 9}\} \]

Since T05 is a continuation turn, the lists do not change. After T06, the weight of manzanas increases because of new appearances of this instance (introduced by the pronouns), while the entity clase {kind} loses weight due to the opposite:

\[ Fa=\{\} \]
\[ Fa'=\{\text{manzanas, esta mañana}\} \]
\[ F=\{\text{manzanas, 40}, \text{esta mañana, 10}, \text{clase, 9}\} \]

We can see that the anaphora is resolved returning manzanas as a valid result, according to the coefficient that indicates the frequency of occurrence of entity manzanas.

A shortcoming is that it does not incorporate semantic information. Thus, it cannot rule out antecedent candidates, and it proposes the use of Spanish Wordnet (Internet reference 6) as lexical tool, to learn semantic patterns between subject and verb, and between verb and object. It does not solve any other kind of references in the dialogue.

The relevant feature of AnaPro (our work), when compared with the work of [20], is that AnaPro resolves anaphora in texts and not in short dialogues.

- [22] focuses on the anaphoric references, both direct and indirect. His resolution algorithm is based on the use of a thesaurus and on the detection and contextualization of coreferences, in such a way that not all the grammatical forms are processed by the algorithm, but through an entity exclusion process, irrelevant information is thrown away.

The algorithm discovers the existing interrelation between the coreference phenomena (direct and indirect) and the indirect anaphora, and that both use nominal referential expressions to manifest their presence.

It determines the evaluation order required to discriminate the phenomena that are used as basis to detect the presence of the indirect anaphora.

It develops a method based in the scenario model of the linguistic context that models the human
reader, needed for the resolution of the indirect anaphora and the coreferences. But one of its shortcomings is that the additional information needed is not automatically obtained. When the system was tuned to use it with free text, its precision was 51%, while the human expert obtained a global average of 60%. With the linguistic context based on the scenario model, it achieves the automatic resolution of the nominal indirect anaphora, with a strong dependence on the additional information added.

• [11] proposes the development of a resolution system for anaphora generated by third person pronouns and adjective anaphors in the dialogues (this work is the closest to our algorithm). The system is based on a syntactic analyzer that uses an extended grammar, context independent, with unification elements. This program incorporates the results of the research to obtain handling patterns for verbs, adjectives, and nouns from Spanish. These results allow quantitative classification of the variations generated by the analyzer, using the weights assigned to the variations, following the values of the subcategorization combinations. The weights of the subcategorization combinations are the result of a syntactic analysis process and of an extraction following a statistical model that allows determining the complements of verbs, adjectives, and some nouns from Spanish, starting from a text corpus.

• [10] approaches the anaphora in Spanish and resolves it through a sequence of semantic criteria, generating an abductive theory (starting from facts, an hypothesis is generated) for the pronominal anaphora. They propose the following resolution procedure:

1. Obtain the grammatical form of the sentence and determine the main verb.
2. Incorporate the non-anaphoric referents that show up in the analysis into the context.
3. Determine the anaphoric terms and their features.
4. Determine, as a function of the type of argument that the anaphora occupies with respect to the main verb, its thematic role.
5. Look for possible referents for the anaphora, among those occurring in the same and former sentences. Each one must pass three filters:
   a. Grammatical agreement (it is the syntactic consistency criterion).
   b. Agreement between the thematic roles (preferential criterion).
   c. Semantic coherence of the resulting sentence (a mixture of preferential and semantic consistency criteria).

Finally, the paper presents a solution, although it requires more experimentation to test the hypotheses presented.

As we can appreciate, anaphora resolution is a problem not yet fully solved. Hence, we have proposed and implemented a solution based in pattern identification and determiners enumeration, which we now show.

4. The architecture of AnaPro
AnaPro works in five stages.

Stage 1: It receives a text tagged with EAGLE tags [5]. The tagged text is stored in a structure in the main memory, and the nominal expressions are identified.

Stage 2: The pronominal syntagmas are identified, and the cases of nominal ellipsis are stored.

Stage 3: The direct anaphoric references are identified.

Stage 4: The anaphoric references go through a filter of grammatical restrictions. Here, candidates (for resolution) are identified. If ties exist, criteria of preferences in the assignment apply so that only one candidate is chosen for each anaphoric reference.

Stage 5: In the text, all pronouns solved in stage 4 are replaced by the chosen candidates. We give a detailed description of the stages with the help of an example.

4.1 Stage 1
This stage reads the tagged text, storing each word and their tags in separate structures, which are
numbered. Also, the nominal expressions are obtained, through the use of the module “To obtain the nominal expression” of AnaPro, which is explained below.

A portion of text “La batalla de Puebla” {Puebla’s battle} (Internet reference 3) shows AnaPro in action.

“La batalla de Puebla.
Aunque se considera en todo el país, ésta se celebra especialmente en Puebla, México, y en los estados del Sur de Estados Unidos” {Puebla’s battle. Although considered in all the country, it is celebrated especially in Puebla, Mexico, and in the southern states of USA}

Table 2. Shows the tagged text, numbered and distributed in three columns

<table>
<thead>
<tr>
<th>Tagged and numbered text</th>
<th>Nominal expressions generated</th>
</tr>
</thead>
<tbody>
<tr>
<td>AD0FS00 = la</td>
<td>NCFS0001 = la batalla de puebla (Puebla’s battle)</td>
</tr>
<tr>
<td>(the)</td>
<td>NCMS00011 = todo el país (all the country)</td>
</tr>
<tr>
<td>NCFS0001 = batalla (battle)</td>
<td>NCFS00018 = Puebla</td>
</tr>
<tr>
<td>ACOMSP2 = de</td>
<td>NCMS00020 = México</td>
</tr>
<tr>
<td>(of)</td>
<td>NCMP00025 = los estados del sur (the southern states)</td>
</tr>
<tr>
<td>NCFS0003 = puebla (Puebla)</td>
<td>NCMP00029 = Estados Unidos (United States)</td>
</tr>
<tr>
<td>Fp4 = .</td>
<td></td>
</tr>
<tr>
<td>CS5 = aunque (although)</td>
<td></td>
</tr>
<tr>
<td>P00000006 = se (it is)</td>
<td></td>
</tr>
<tr>
<td>VMIPS07 = considera (considered)</td>
<td></td>
</tr>
<tr>
<td>SPS008 = en (in)</td>
<td></td>
</tr>
<tr>
<td>DIOMS09 = todo (all)</td>
<td></td>
</tr>
<tr>
<td>AD0MS010 = el (the)</td>
<td></td>
</tr>
<tr>
<td>NCMS00011 = pais (country)</td>
<td></td>
</tr>
<tr>
<td>Fc12 = ,</td>
<td></td>
</tr>
<tr>
<td>P0D0FS0013 = ésta (it, this)</td>
<td></td>
</tr>
<tr>
<td>P030000014 = se (is)</td>
<td></td>
</tr>
<tr>
<td>VMSI3S015 = celebra (celebrate)</td>
<td></td>
</tr>
<tr>
<td>RG16 =</td>
<td></td>
</tr>
<tr>
<td>especialmente (especially)</td>
<td></td>
</tr>
<tr>
<td>SPS00017 = en (in)</td>
<td></td>
</tr>
<tr>
<td>NCFS00018 = Puebla</td>
<td></td>
</tr>
<tr>
<td>Fc19 = ,</td>
<td></td>
</tr>
<tr>
<td>AD0MP024 = ¿ (the)</td>
<td></td>
</tr>
<tr>
<td>NCMP00025 = los (the)</td>
<td></td>
</tr>
<tr>
<td>NCMS00027 = sur (South)</td>
<td></td>
</tr>
<tr>
<td>SPS0028 = de (of)</td>
<td></td>
</tr>
<tr>
<td>NCMP00029 = Estados (States)</td>
<td></td>
</tr>
<tr>
<td>ACOMPP30 = Unidos (United)</td>
<td></td>
</tr>
</tbody>
</table>

Table 3. Nominal expressions identified in the text. The first characters of the tag mean: N (noun), C (common), F (female), S (singular), M (masculine), P (plural). The table shows the order in which they occur. The tagging is done by the tagger TaggerFT [12] in a preprocessing step. The nominal expressions are generated by the tagger, and we only use several filters to identify and delimit them correctly. They are shown here

Module “To obtain the nominal expression”

1. Each tagged word is numbered (by the tagger).

2. Using the tags, pronouns are identified as well as their determiners and auxiliary types, through the comparison between each tag and regular expressions, be they a pronoun, noun, adjective, etc.

   For instance, to look for pronouns in the text, which will be tagged as pronominal anaphoric references, the expression shown in Figure 2 is used. Figure 2 is expanded to Figure 3, showing that a pronoun must match with one of these expressions. It is required that a pronoun must match with the first two of one of the expressions of Figure 3. The remaining eight symbols in the regular expression will define with further detail each type of pronoun, such as its gender and number, which is also considered when searching for candidates.

Regular expressions are used to identify pronouns, articles, nouns, adjectives and in some cases, to distinguish among each particular pronoun, to ascertain its gender and number in order to find pairs of pronouns and nouns. Of course, we could find the nominal expressions from direct analysis of the text strings, but we find it easier to analyze the tags.
Once the pronouns or anaphoric pronouns are identified, AnaPro locates the previous portion of text where the possible nouns or determiners may exist, those to which the pronoun refers. This area or range is known as nominal expression, which contains the anaphora and its possible determiners. The name nominal expression was changed to sentence in the implementation of AnaPro, for practical purposes.

**Figure 3. Tag types, where asterisks indicate optional areas, and the first two characters indicate the pattern to look for.** The standard requires that the tag of a pronoun should have at least eight elements, and then we assign an index, which could have from 1 to 10 digits. In this step, the algorithm does not take into account the last 6 or 8 digits, since they provide no information to our tool.

**Figure 2. Regular expression used to find pronouns in a text.** It is divided in three parts: the start of the tags, the following letters, and the length of the words.

```
“P[PDXTR] \w{1,10}”
```

**Tags starting with P** Followed with any of P, D, X, G or R

**Words of up to ten characters**

Let us assume that a nominal expression has two qualifying adjectives, and immediately after them there is a "new noun nucleus", that is, a different nominal expression. When the first nominal expression is built, both qualifying adjectives are added, but when the second nominal expression is built, AnaPro first verifies if the qualifying adjectives have been (or have not) previously been assigned. If that is the case, the search backwards for nominal elements is immediately interrupted. The same happens with specific punctuation marks.

An important addition has been the use of date tags as nuclei to construct nominal expressions with historical dates or references to some year, with all that this implies: use of prepositions, numbers, months, and years.

The search for lists and plural expressions is also added, looking for elements such as commas, nouns, and conjunctions, in order to generate a single nominal expression, with the tag of the last noun modified to be plural. The search of plurals with conjunctions, unlike all the others, is not very reliable since errors have been detected when a conjunction is between two nouns. In this case its effectiveness depends on the correct use of the comma by the writer.

**Function to generate nominal expressions.** This part of AnaPro, instead of dealing directly with nouns, starts by correcting labeling errors "on the fly", such as the exchange of a noun label by a quality adjective, in the case of a surname or a name. At the same time, the first step is always to find the nucleus. The process starts going through the text from start to finish in a single cycle. When the first noun nucleus is found, it stores the noun and its tag, then goes back looking for (jumping over phrases or sentences, moving backwards from the found noun) a determiner article, qualifying adjective, number and other modifiers.

These are sequentially added as they are found (using a window of at most five elements, which in our tests has never been exceeded). The cycle repeats going forward from the noun, but with a slightly modified order and nouns are sought. An important step in this part is to keep absolutely all the tags of the elements found going forward, in a structure called ACpre. It is used to verify the limits of the previous nominal expressions.

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**The ACpre structure.** In the algorithm that constructs nominal expressions, the ACpre structure stores the names of all the nominal...
expressions at the same time they are being built, starting from the first word up to the last period. ACpre, a list, is used to avoid the construction of nominal expressions with “name nuclei” that have been previously used as part of a name of another nominal expression. Example: el triunfo del movimiento revolucionario {The success of the revolutionary movement}. A first step renders triunfo {success} as nucleus, the article el {the} is placed before it, and del movimiento revolucionario {of the revolutionary movement} is appended as qualifying adjectives. A second step will find movimiento revolucionario as nucleus and will try to create a new nominal expression. But, upon analyzing the ACpre structure, it will notice that movimiento revolucionario has already been used in a previous expression, therefore, it cannot be a new nucleus. Hence, it is ignored.

4.2 Stage 2

This stage detects the pronominal syntagmas containing pronouns as nucleus. At the same time, all the cases of nominal ellipsis are stored.

**Identification of pronominal syntagmas.** A simple analysis of the tag will identify the pronominals. Example:

PD0FS00013 = ésta {this} (See Table 3)

Here we find only a syntagma as direct anaphoric reference. Other examples of syntagmas are el cual {that which}, la cual {that which}, el que {that what}, which are also identified by AnaPro.

4.3 Stage 3

Candidates are identified, and direct anaphoric references (with gender and number defined) are resolved with a single candidate. The module of AnaPro that does this task is the Classifier. It finds the coincidences of pronouns and nouns, classifying them as single or multiple matches (in the case of ties). Finally, it looks for the correct pairs.

**Search for candidates.** The tag found in Stage 2 is analyzed:

PD0FS00013 = ésta {this}

It is identified as a female singular pronoun. Thus, Stage 3 constructs the corresponding search pattern, which is then applied when searching for candidates:

Index: 13 Pattern: N\w{1}FS\w{3,10} Sentence: 2

indicating that it has to look for female singular nouns in the second or previous sentences. Therefore, the candidate found is (refer to Tables 2 and 3):

NCFS0001 = La batalla de puebla {Puebla’s battle}

Therefore we have:

PD0FS00013 = NCFS0001

It means:

ésta {this} = La batalla de puebla {Puebla’s battle}

In other words, the tags found in Stage 2 are converted by Stage 3 into regular expressions, which then look for the corresponding nominal expressions. Of course, more elaborate matching procedures are found in the literature, but this use of regular expressions is simpler and already produces good results (at least for Spanish texts).

4.4 Stage 4

It solves special cases of pronouns and candidates ties. This stage also uses the Classifier. Our example does not have special cases or ties among candidates. The following example has them:

Juan compró un boleto en la tienda. Éste estaba roto.

{John bought a ticket in the store. It was torn}

This is a tie, since there are several possible candidates: Juan {John}, boleto {ticket} and tienda {store}. tienda cannot be because Éste {this} is a masculine pronoun and tienda has female gender.

**Solving the tie.** To solve the tie, four important criteria are taken into account:

1. Less distance to the noun.
2. If it is determined.
3. Less distance to the main verb.
4. Higher number of occurrences in the text.
In case the first three criteria are not enough to break the tie, the numbers of occurrences are compared (fourth criteria). For this example, Table 4 shows the weights assigned to each candidate.

<table>
<thead>
<tr>
<th>Criterion</th>
<th>Juan (John)</th>
<th>Boleto (ticket)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less distance to the noun</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Is it determined?</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Less distance to the main verb</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>Higher number of occurrences in text</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Total</td>
<td>1</td>
<td>3</td>
</tr>
</tbody>
</table>

Table 4. Juan compró un boleto en la tienda. Éste estaba roto. Comparison using weights among tied candidates

When comparing weights among candidates, we obtain éste {this} = boleto {ticket}, hence solving the sentence as follows:

Juan compró un boleto en la tienda. [El boleto] estaba roto.

{John bought a ticket in the store. [The ticket] was torn}

Therefore, Stage 4 uses a rather simple method to solve ties. More complex methods could be used (for instance, the theory that considers the “center” of the discourse), but we are delighted to have found a rather simple and easy to implement procedure that, nevertheless, renders good results.

4.5 Stage 5

All the solved anaphors are replaced, yielding a new text that facilitates further processing by OM* (mentioned in Section 1). Pronouns are replaced by their respective candidates. The module that performs this task is the Resolver, which replaces anaphora by their respective noun, generating an output file with the anaphors resolved. In general, this is a straightforward replacement module, with some cosmetic variations to its output in order to generate valid Spanish replacements.

Coming back to the previous example:

“La batalla de puebla. Aunque se considera en todo el país, ésta se celebra especialmente en Puebla, México; y en los estados del sur de Estados Unidos” {Puebla’s battle. Although considered in all the country, it is celebrated especially in Puebla, Mexico, and in the Southern states of USA}

As result of the Resolver, the output file would contain:

“La batalla de puebla. Aunque se considera en todo el país, [la batalla] se celebra especialmente en Puebla, México; y en los estados del sur de Estados Unidos”

{Puebla’s battle. Although considered in all the country, [the battle] is celebrated especially in Puebla, Mexico, and in the Southern states of USA}

Although AnaPro resolves ésta {this} as referring to la batalla de puebla {Puebla’s battle}, when replacing is done, only the nucleus of the nominal expression is replaced. The expression la batalla de puebla has as nucleus la batalla {the battle}, which is used in the replacing algorithm.

5. Using AnaPro

AnaPro has two versions: a console and a graphical version. Before using any of them, a file in .txt format is needed in order to generate the tagged file Taller FT, which returns an extension .txt.tag. The Java virtual machine, version 1.2 or higher, must be already installed. We use the graphical version to show AnaPro in action.

For instance, in the document about the piano (Internet reference 1):

El piano está compuesto por una caja de resonancia, a la que se ha agregado un teclado, por donde se percuten las cuerdas de acero con macillos forrados de fieltro, por éstos se clasifica como instrumentos de percusión. Las primeras composiciones específicas para este instrumento surgieron alrededor del año 1732, entre ellas destacan las 12 sonatas para piano de Lodovico Giustini. {The piano is composed of a sounding board, to which a keyboard has been added, through which the steel chords are struck with hammers lined with felt; for this reason, it is classified as a percussion instrument. The first
specific compositions for this instrument arose circa 1732. These include the 12 piano sonatas of Lodovico Giustiniani.

With the help of the example, the functions of the graphic interface (Figure 4) are explained:

1. File menu, to open files with extension .txt.tag. They should be located in the same folders than the original text.

2. Tag showing original text, as found in the file.

3. Tag showing anaphoric references found by AnaPro, the word indices inside the text, and the noun corresponding to each reference; in other words, its solution.

4. Tagged text, used as a reference to compare it with the final tagged text.

5. Auxiliary view of tagged text, to compare it with the anaphoric references.

   a. It shows the nominal expressions generated by AnaPro. Noun + adjectives, article.

   b. It shows the new tagged text with the anaphors solved.

7. It shows the text with the words already replaced.

8. Erase all windows.

9. Button 10 stores the final file into the same folders containing the initial text file.

Figure 5 shows the tag showing the anaphoric references. The three sections marked A, B and C are:

Section A contains the anaphoric references found in the text, indicating the tag and then the word(s).

Section B shows the index of the references, followed by the search pattern (if it is a masculine pronoun, feminine, singular, etc.), and then it shows the sentence where the anaphora was found.

Section C shows the solution to each anaphora found by the resolver. When the Generar Archivo (generate file) button is clicked, a new file is produced, adding at the end of its name _ANA. Thus, if its name is Mezcal.txt, the file generated will be Mezcal_ANA.txt.

![Figure 4. Graphical interface of the resolver](image-url)
Figure 5. Tag: Anaphoric references

6. Tests

Testing has been done using the following texts taken from the Web:

1. “El piano” {the piano} (Internet reference 1),
2. “El burro flautista” {the flutist donkey} (Internet reference 2),
3. “Batalla de Puebla” {Puebla’s battle} (Internet reference 3),
4. “Pancho Villa” (Internet reference 4) and
5. “Mario Almada” (Internet reference 5),

In all of them, AnaPro resolved automatically all the pronominal anaphors found. The same was done manually, in order to verify the effectiveness of the program. The formula is:

\[
\text{Accuracy} = \frac{\text{Number of anaphors correctly solved}}{\text{Total number of anaphors in the document}}
\]

We prefer to use accuracy instead of precision (which is the percentage of correct responses out of the anaphors considered; instead, accuracy considers all anaphors present in the text) or recall (percentage of the anaphors considered out of the anaphors in the text; instead, accuracy measures success against all anaphors present in the text). In other words, accuracy is the precision obtained in the presence of perfect recall. Table 5 shows the results obtained.

<table>
<thead>
<tr>
<th>Text</th>
<th>Number of words</th>
<th>Anaphors solved by hand</th>
<th>Anaphors solved by AnaPro</th>
<th>Accuracy (assuming a perfect tagger)</th>
</tr>
</thead>
<tbody>
<tr>
<td>El piano</td>
<td>61</td>
<td>3</td>
<td>3</td>
<td>100</td>
</tr>
<tr>
<td>El burro flautista</td>
<td>97</td>
<td>1</td>
<td>1</td>
<td>100</td>
</tr>
<tr>
<td>Batalla de Puebla</td>
<td>513</td>
<td>7</td>
<td>7</td>
<td>71.4</td>
</tr>
<tr>
<td>Pancho Villa</td>
<td>634</td>
<td>11</td>
<td>7</td>
<td>63.6</td>
</tr>
<tr>
<td>Mario Almada</td>
<td>601</td>
<td>9</td>
<td>5</td>
<td>55.5</td>
</tr>
</tbody>
</table>

Table 5. Where anaphors are solved, both by AnaPro and by hand

Failures, in most cases, were caused by final tagging. Thus, tagging has been manually corrected (to assume a perfect tagger). Retesting yields the results in Table 6, where failures are due to phrasing errors, or to very elaborate cases or very specific situations in a particular discourse (use of complex cohesion techniques), because the reference becomes incoherent. An example of these cases is the following fragment from Mario Almada (Internet reference 5):

“…fácilmente es expuesto a rodajes y en la Ciudad de México empieza a trabajar en un centro nocturno llamado Cabaret Señorial que era propiedad de su padre. Pero antes de esto, cuando su hermano Fernando empezó…”

The word esto {this} does not refer to a noun or a candidate, but to a whole context previously cited; hence, it is difficult to produce a computer tool that can resolve it.

We would like to perform larger tests, against standardized Spanish texts, where the anaphors were already found by hand. But, to our knowledge, these standard texts do not exist for Spanish. Our plan is to build these “standard texts” (containing the resolution of anaphora done by competent Spanish speakers) and then use them to assess the accuracy of AnaPro.
Annex A, B and C show the results of AnaPro on the texts “The Piano”, “Pancho Villa” and “Mario Almada”, respectively. For lack of space the results of the documents “El burro Flautista” {The Flutist Donkey} [internet reference 2] and “Batalla de Puebla” {Puebla’s battle} [internet reference 3] are not shown.

<table>
<thead>
<tr>
<th>Text</th>
<th>Number of words</th>
<th>Anaphors solved by hand</th>
<th>Anaphors solved by AnaPro</th>
<th>Accuracy (assuming a perfect tagger)</th>
</tr>
</thead>
<tbody>
<tr>
<td>El piano</td>
<td>61</td>
<td>3</td>
<td>3</td>
<td>100</td>
</tr>
<tr>
<td>El burro</td>
<td>97</td>
<td>1</td>
<td>1</td>
<td>100</td>
</tr>
<tr>
<td>Flautista</td>
<td>513</td>
<td>7</td>
<td>7</td>
<td>100</td>
</tr>
<tr>
<td>Batalla de</td>
<td>634</td>
<td>11</td>
<td>9</td>
<td>81.8</td>
</tr>
<tr>
<td>Puebla</td>
<td>513</td>
<td>7</td>
<td>7</td>
<td>100</td>
</tr>
<tr>
<td>Pancho Villa</td>
<td>601</td>
<td>9</td>
<td>8</td>
<td>88.8</td>
</tr>
<tr>
<td>Mario Almada</td>
<td>601</td>
<td>9</td>
<td>8</td>
<td>88.8</td>
</tr>
</tbody>
</table>

Table 6. The columns represent: texts analyzed, number of words in each, manual solution, anaphors solved by AnaPro, and the accuracy of AnaPro assuming a perfect tagger (no mistakes in tagging).

6.1 Analyzing the failures of AnaPro

AnaPro has a high resolution success, except in some cases. Another example is taken from the same text (internet reference 5).

“He has appeared in more than 200 films, many of which were classic drama, as his first film Beloved Mother (1935). In this he came out as a boy with his brother.

Here we have a phrasing error, because the text uses a foreign word (film) that in Spanish is of masculine gender (el filme), but the text refers to it as ésta {this, feminine}. AnaPro generated the nominal expression (a candidate) “film Madre Querida” and assigned it the masculine gender. That fools AnaPro, which looks further backwards for a feminine candidate.

AnaPro heavily relies on the results of the tagger, since the anaphoric references fail if the tagger fails. If it is manually corrected (or using a better tagger), the resolution algorithm works quite well.

7. Conclusions and future work

A method has been developed, using regular expressions, to separate, enumerate, and identify pronominal syntagmas and candidates to resolve the anaphoric references.

An algorithm has been designed and tested, able to solve ties using only information present in the text, statistical analysis of occurrences, text distance, and specific wording criteria, in order to assign weight to multiple candidates.

The method and algorithm are part of AnaPro, which resolves direct anaphora given by pronouns that occur in Spanish texts. The texts do not refer to a particular topic. AnaPro is part of Project OM* (a follow over to the Ph. D. thesis of one of the authors, Cuevas, 2006), that tries to build the ontology of the knowledge contained in a text, in order to answer complex queries.

In order to minimize errors and to improve the efficiency of AnaPro, it is necessary to perfect the tagger, widening the scope of nouns handled, in order to solve quickly the simpler problems. Also, another tagger that uses EAGLE standard can be used.

In order to go through larger texts, we plan to build first standardized Spanish texts, where we can locate the anaphors and tag them by hand.

Acknowledgements

Authors I.T. and E.V. would like to acknowledge ESCOM-IPN, where they defended their thesis, #20110083, which gives a more detailed description of AnaPro. Work herein reported was partially sponsored by CONACYT Grant #128163 (Project OM*), by IPN (A.G. as Resident Scientist), and by SNI.
Annex A. Example. The Piano.

(The piano is composed of a sounding board, to which a keyboard has been added, through which the steel chords are struck with hammers lined with felt; for this reason, it is classified as a percussion instrument. The first specific compositions for this instrument arose circa 1732. These include the 12 piano sonatas of Lodovico Giustini.)

Annex A. Continuation. The left rectangle shows the original anaphoric references. The center rectangle shows the nominal expressions. The right rectangle shows the anaphoric references, replaced.

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Pancho Villa.

(Doroteo Arango Arambula) Mexican revolutionary (San Juan del Río, Durango, 1876 – Parral, Chihuahua, 1923)

Poor peasant, orphan and with scanty schooling, when the 1910 revolution broke out, he was already several years ago a fugitive in the mountains, for killing one of the owners of the farm where he worked.

Then Pancho Villa joined Madero in his struggle against the dictatorship of Porfirio Díaz, and showed an innate ability for war. Leveraging his knowledge of the land and the peasants, he formed his own army in Northern Mexico, which contributed to the success of the revolutionary movement.

In 1912 he was jailed, since General Victoriano Huerta suspected he was involved in the revolution of Orozco in defense of the social aspirations of the peasantry, which Madero had postponed. He escaped to the United States and, after the assassination of Madero, returned to Mexico and formed a new revolutionary army, the Northern Division (1913).

With it he supported the struggle of Emiliano Zapata and Venustiano Carranza against Huerta, who had emerged as dictator. Together they overthrew him in 1914, but after the victory of the second revolution, Villa and Zapata felt disappointed by Carranza, and took up arms again, now against him. This time military luck was not on his side: Álvaro Obregón defeated Villa’s army and Carranza consolidated his power, and obtained the official recognition by the U.S. government.
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Mario Almada.- Born in Huatabampo, Sonora, Mexico. In addition of being a great actor, he is also director, writer and movie producer. He started his artistic career in the 1930s in Mexico. He has appeared in more than 200 movies, many of which were classic drama, like his first film *Madre Querida* (1935). In this he appeared as a child with his brother Fernando Almada, who performed as an extra. He would not appear again in a movie until several decades later.

Almada moved from his native Huatabampo to Ciudad Obregón and Guadalajara, where he lived for several years until he finally settled in Mexico City. Since Almada was born in a family linked to films, he is easily exposed to filming and in Mexico City he starts working in a night club called *Cabaret Señorial* which belonged to his father. But before this, when his brother Ferdinand started a career in the movies as an actor, Mario decides to stay as producer. In 1963, he writes his first screenplay for a movie.

In 1965 Mario formed the paper of Bruno Pire in the film *Los ratones de la Diosa*, where his brothers Almada were producing and that Fernando was the protagonist. Bruno Ray saliría lesionado durante la rodaje y Mario aceptó tomar su lugar.

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From 1997 to 2001 he had an important participation with the group The Exterminator for their records “Narco Corridos 2” where he appeared in the cover, and he also conducted dialogues for most tracks in the album. After that, in El Chile Peláiz also from 1997, where he reappeared in the cover, in 1999 he performed dialogues for the theme “Smuggling in the eggs” in addition to taking part in the video that was prohibited by SCT, and his last appearance was in the album “Gathering of dogs” where he appeared with some other Mexican actors.
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