French Resources for Extraction and Normalization of Temporal Expressions with HeidelTime

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Abstract
In this paper, we describe the development of French resources for the extraction and normalization of temporal expressions with HeidelTime, an open-source multilingual, cross-domain temporal tagger. HeidelTime extracts temporal expressions from documents and normalizes them according to the TIMEX3 annotation standard. Several types of temporal expressions are extracted: dates, times, durations and temporal sets. French resources have been evaluated in two different ways: on the French TimeBank corpus, a corpus of newspaper articles in French annotated according to the ISO-TimeML standard, and on a user application for automatic building of event timelines. Results on the French TimeBank are quite satisfying as they are comparable to those obtained by HeidelTime in English and Spanish on newswire articles. Concerning the user application, we used two temporal taggers for the preprocessing of the corpus in order to compare their performance and results show that the performances of our application on French documents are better with HeidelTime. The French resources and evaluation scripts are publicly available with HeidelTime.

Keywords: Temporal expressions, normalization, French resources

1. Introduction
The analysis of temporal information is often an essential component in text understanding and is useful in a wide range of information retrieval applications (Alonso et al., 2007; Alonso, 2008; Kanhabua, 2009; Mestl et al., 2009). The task of temporal annotation consists in extracting and normalizing temporal expressions. Normalization is the operation of turning a temporal expression into a formatted, fully specified representation (this includes finding the absolute value of relative dates). The TempEval challenges, for example, focus on the evaluation of temporal information processing using the ISO-TimeML language (Pustejovsky et al., 2010), a specification language for manual annotation of temporal information in texts.

HeidelTime is a multilingual, cross-domain temporal tagger which extracts temporal expressions from documents and normalizes them according to the TIMEX3 annotation standard (Strötgen and Gertz, 2013). It is an open-source tagger which achieved the best results for the extraction and normalization of temporal expressions for English documents in the context of the TempEval-2 and TempEval-3 challenges (Verhagen et al., 2010; Uzzaman et al., 2012). HeidelTime processes documents in English, German, Dutch, Vietnamese, Arabic, Spanish, Italian and we developed French resources that are now publicly available within the official HeidelTime distribution1.

The paper is structured as follows: we first present HeidelTime. Section 3 describes the development of French resources. Finally, we present and discuss the evaluation results in Section 4. This evaluation is twofold: first, HeidelTime with French resources has been evaluated on the French TimeBank corpus and then on a user application which automatically builds event timelines.

2. Presentation of HeidelTime
HeidelTime is a multilingual, cross-domain temporal tagger. It is a rule-based system with a separation between language-dependent resources and generic Java code. Resources consist in extraction rules (regular expression patterns) and lexicons for normalization (for example, weekdays, months, etc). Absolute temporal expressions (July 26th, 2013; 07-26-2013) are extracted and normalized by the extraction rules. Relative temporal expressions (yesterday, July) are extracted by the rules but are left underspecified at this step.

Then, a normalization (Java code) is applied according to the type of documents (news, scientific, etc.) and to the tense of the verb used in the sentence. As a reference time, normalization can use the document creation time (DCT) or the previously mentioned date. HeidelTime requires sentence, token, and part-of-speech information. For English, the TreeTagger is used. Figure 1 shows an example of normalization for an absolute date and a DCT relative expression. Attribute type is the type of the extracted temporal expression and attribute value is its normalization.

DCT: 2009-12-22

Spaniards often choose lottery numbers matching significant dates. One of the most requested ticket numbers <TIMEX3 tid="t1" type="DATE" value="2009">this year</TIMEX3> was 25609, which corresponds to <TIMEX3 tid="t2" type="DATE" value="2009-06-25">June 25, 2009</TIMEX3>, the day pop star Michael Jackson died.

Figure 1: Example of HeidelTime output.

1http://code.google.com/p/heideltime/
3. Development of French Resources

Resources are composed of 3 types of files read by HeidelTime’s resource interpreter and which have to follow HeidelTime’s rule syntax: patterns, normalizations and rules. The pattern files contain words and phrases used to express temporal expressions (months, days, etc.). The normalization files contain normalization information about the patterns (for example, the normalized value of February is 02). Finally, the rule files contain rules for date (100 rules), time (20 rules), duration (25 rules), and set expressions (12 rules). All rules have:

- an extraction part which defines the expressions that have to be matched in a document, using the pattern resources,
- a normalization part which normalizes the extracted expression using the normalization resources.

For example, the following rule is used to extract and normalize the temporal expressions jeudi 4 octobre 2012 or le lundi 23 sept. 2013:

```plaintext
RULENAME="date_r1",
EXTRACTION= "([Ll]e )g\reWeekdayg2
%reMonthLongg1|%reMonthShortg1)%2
%reYearDigitg3
%normMonth(group(1)) g4
%normDay(group(3)) g5",
NORM_VALUE="group(7)-%normMonth(group(4))-%normDay(group(3))"
```

where:
- group(7) is the string extracted with the pattern %reYearDigit (i.e. 2012 or 2013),
- %normMonth(group(4)) is the normalization of octobre extracted with the pattern %reMonthLong (i.e. 10) or sept. extracted with the pattern %reMonthShort (i.e. 09),
- %normDay(group(3)) is the normalization of 4 (i.e. 04) or 23 (i.e. 23) extracted with the pattern %reDayNumber.

The normalized values are then 2012-10-04 and 2013-09-23.

To develop the French pattern and normalization resources, we translated the English and Spanish resources and adapted them to the French patterns. To develop the French rules for dates, times, durations and sets, we used a corpus composed of 350 news articles from Agence France Presse (AFP). Note that during the process of rule development, an important point that has to be taken into account is that French is an inflected language (nouns, adjectives, determiners, verbs).

HeidelTime uses the tense of the sentence verb to determine if a temporal expression refers to a past or future date w.r.t. the DCT. This information is given by the French TreeTagger which is used for preprocessing the French documents (sentence, token and part-of-speech annotation). For example, the following rule is used to extract and normalize the temporal expressions mars in Il est parti en mars (He left in March) or in Il reviendra en mars (He will come back in March):

```plaintext
RULENAME="date_r7",
EXTRACTION= "(%reMonthLongg1|%reMonthShortg1)%2
%normMonth(group(1))",
NORM_VALUE="UNDEF-year-%normMonth(group(1))"
```

where:
- %normMonth(group(1)) is the normalization of mars (i.e. 03) extracted with the pattern %reMonthLong,
- UNDEF-year is an undefined year that is then calculated with the DCT and the tense of the verb. The DCT of both documents is 2009-09. In Il est parti en mars (He left in March), a past tense verb is identified so the normalized value is 2009-03-XX whereas in Il reviendra en mars (He will come back in March), a future tense verb is identified so the normalized value is 2010-03-XX.

Figure 2 shows some examples of extraction and normalization for each temporal type.

![Example of HeidelTime output on French documents](https://gforge.inria.fr/projects/fr-timebank/)

### 4. Evaluation

The French resources have been evaluated in two different ways: on the French TimeBank corpus and on a user application for automatic building of event timelines. We present in this section the results obtained for both evaluations.

#### 4.1. Evaluation Results on the French TimeBank

We first evaluated HeidelTime with the French resources on the French TimeBank corpus\(^2\) (Bittar et al., 2011).

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\(^2\)https://gforge.inria.fr/projects/fr-timebank/

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composed of 108 newspaper articles in French annotated according to the ISO-TimeML standard. In this corpus, there are 425 temporal expressions in texts, among which 227 dates, 130 time expressions, 52 duration expressions and 16 temporal sets.

To evaluate the extraction performance, we used the measures used in TempEval challenges: precision, recall and F1-score for strict and relaxed matching. We also computed the F1 measures on the two most important TIMEX attributes value and type. The value F1 score captures the performance of the system to extract a temporal expression with a correct normalization. The type F1 is the performance of the system to extract a temporal expression with a correct type (i.e. date, time, duration, set). Evaluation scripts that we developed are also publicly available.

4.1.1. Results
Table 1 presents the results on the French TimeBank:

|                      | Precision | Recall | F1  |
|----------------------|-----------|--------|-----|
| Strict match         | 0.86      | 0.84   | 0.85|
| Relaxed match        | 0.92      | 0.89   | 0.91|
| Value F1             |           | 0.74   |     |
| Type F1              |           | 0.83   |     |

Table 1: HeidelTime’s results with French resources on French TimeBank.

These results are quite satisfying as they are comparable to those obtained by HeidelTime in English and Spanish on newswire articles (Strötgen et al., 2013). Indeed, on the TempEval 3 English corpus, F-score for strict matching is 0.81, attribute value F1 is 0.77 and attribute type F1 is 0.82. On the TempEval 3 Spanish corpus, F-score for strict matching is 0.85, attribute value F1 is 0.85 and attribute type F1 is 0.87 (if values are similar, results are hardly comparable, since the corpora are different).

Table 2 presents the detailed results for each temporal type. Here, a matching is correct if there is a strict or relaxed matching and if the type attribute is correct. A value attribute is considered as correct only if the matching is correct.

|                      | Correct Match | Correct Match & Correct Value |
|----------------------|---------------|-----------------------------|
|                      | #  | % w.r.t ① | #  | % w.r.t ① | % w.r.t ② |
| Total (227)          | 212| 93.4 % | 187| 82.4 % | 88.2 % |
| DATE (227)           | 84 | 64.6 % | 62 | 47.7 % | 73.8 % |
| TIME (130)           | 40 | 76.9 % | 40 | 76.9 % | 100 % |
| DURATION (52)        | 8  | 50 %   | 6  | 37.5 % | 75 %   |

Table 2: Detailed results on French TimeBank.

4.1.2. Error Analysis
As we can see from Table 2, most extraction errors are on TIME and SET expressions. For time expressions, we noticed in the French TimeBank that adverbs maintenant, aujourd’hui and dèsormais (now, today, henceforth) are inconsistently annotated either as a TIME or a DATE expression and that their value is either PRESENT-REF or a normalized value. In our French resources, we considered that these adverbs are DATE expressions. We found 22 occurrences of mismatches due to this problem. Another error cause is when a date is associated with a time expression (for example, the interview will be on <TIMEX3 type=”DATE” value=”2012-06-05”>June, 5th</TIMEX3> at <TIMEX3 type=”DATE” value=”2012-06-05T17:00”>5 pm</TIMEX3>).

Concerning duration expressions, some have not been extracted mainly because we did not develop rules for very specific cases (8 occurrences): for example, time expressions expressed in minutes or seconds, durations like half-century, quinquennium or greco-roman period, etc. But we can note that when a duration expression is correctly extracted, it is always normalized correctly.

4.2. Evaluation on a User Application
We have also evaluated HeidelTime with the French resources on a user application that automatically builds event timelines from a search query.

4.2.1. User Application: Event Timelines
We developed an approach for detecting salient (important) dates in texts in order to automatically build event timelines from a search query (Kessler et al., 2012b). In order to extract salient dates that warrant inclusion in an event timeline, a newswire article corpus is first pre-processed and temporal expressions are normalized. Then, the corpus is indexed by the Lucene search engine. Given a query, a number of documents are retrieved by Lucene. Dates are extracted from documents and ranked in order to show the most important ones to the user together with the sentences that contain them.

4.2.2. Document Collection
So far, our system used the linguistic analyzer XIP (Ait-Mokhtar et al., 2002) which performs a deep syntactic analysis, named entity recognition and extraction and normalization of temporal expressions for English and French. We wanted to evaluate the performance of our system with another free temporal tagger. Thus we used two temporal taggers for the preprocessing of the corpus, HeidelTime and XIP, in order to compare their performance for the user application.

We used a corpus of newswire texts provided by the AFP French news agency. The French AFP corpus is composed of 1 million texts that span the 2004-2011 period (499 documents/day in average and 390 millions words). Each document is an XML file containing a title, a date of creation

3http://code.google.com/p/heideltime/
wiki/ReproduceEvaluationResults

4http://lucene.apache.org/
(DCT), set of keywords, and textual content split into paragraphs. Note that absolute dates are quite infrequent in this corpus (about 7%).

4.2.3. Results

Processing runs were evaluated on 94 manually-written chronologies according to Mean Average Precision (MAP), which is a widely accepted metric for ranked lists. These chronologies (textual event timelines) are a specific type of articles written by AFP journalists in order to contextualize current events. These chronologies consist in a list of dates (typically between 10 and 20) associated with a text describing the related event(s). With the corpus processed by XIP, MAP is 0.60 (Kessler et al., 2012a) whereas it is 0.64 with the corpus processed by HeidelTime. This result shows that the performances of our application on French documents are better with HeidelTime than with XIP which achieved good results in the TempEval campaign (Verhagen et al., 2007; Hagège and Tannier, 2008).

The main cause for incorrect value normalization of under-specified expressions in AFP corpus is wrong tense identification: for normalization, HeidelTime considers the tense of the verb close to the temporal expression but this verb may not be the main verb of the sentence and be in a different tense. In the following example, the temporal expression mercredi (Wednesday) is not normalized correctly because the closest verb which is used for normalization (devra (will have to)), is in future tense whereas the main verb of the sentence in present tense should be considered for normalisation:

François Hollande assure que le prochain président de la République devra être l’inverse de Nicolas Sarkozy”, dans un entretien à Libération mercredi.

(François Hollande declares that the next president will have to "be the opposite of Nicolas Sarkozy," in an interview with Libération on Wednesday)

5. Conclusion

In this paper, we presented the French resources we developed for HeidelTime, an open-source multilingual, cross-domain temporal tagger that achieved the best results in the TempEval-3 challenge. Our French resources allow for the extraction and normalization of French temporal expressions with HeidelTime and are now publicly available. HeidelTime with French resources achieved good results in both evaluations that we performed. In future work, we intend to perform evaluations on other types of documents than news articles.

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