Improved Recognition and Normalisation of Polish Temporal Expressions

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Abstract

In this article we present the result of the recent research in the recognition and normalisation of Polish temporal expressions. The temporal information extracted from the text plays major role in many information extraction systems, like question answering, event recognition or discourse analysis. We proposed a new method for the temporal expressions normalisation, called Cascade of Partial Rules. Here we describe results achieved by updated version of Liner2 machine learning system.

1 Introduction

Temporal expressions tell us when something happens, how long something lasts or how often something occurs. Recognition of temporal expressions is a sequence labelling task and normalisation of temporal expressions is a process of their interpretation. It is often used in many natural language processing tasks, e.g., question answering (Pustejovsky et al., 2005b), text summarisation (Daniel et al., 2003) or event recognition (Andersen et al., 1992; Llorens et al., 2010).

One of the most important thing is the ability to share the temporal information across many languages and systems. A serious problem would be to combine the information from different information extraction systems, e.g., to analyse data from multi-lingual newspapers, where the expected result is the unification of metadata.

A widely used markup language for temporal and event expressions is TimeML (Saurí et al., 2006). At the beginning it was prepared for English, in the context of TERQAS\(^1\) workshop, as a part of the ARDA-funded program AQUAINT\(^2\) in order to improve the performance of question answering systems (Pustejovsky et al., 2005a). One of the most widely used rule-based system HeidelTime\(^3\) (Strötgen and Gertz, 2013; Strötgen et al., 2013) which uses the TIMEX3 annotation standard, currently supports 13 languages: English, German, Dutch, Vietnamese, Arabic, Spanish, Italian, French, Chinese, Russian, Croatian, Estonian and Portuguese.

PLIMEX (Kocoń and Marcińczuk, 2015) is the adaptation of TIMEX3 specification with annotation guidelines presenting how to describe temporal expressions in Polish text documents. It is based on TIDES Instruction Manual for the Annotation of Temporal Expressions (Ferro, 2001), which describes TIMEX2 annotation format. The TIDES manual is also the core of the TIMEX3 annotation format, used in the TimeML specification (Saurí et al., 2006). Both documents present how to use the special Standard Generalized Markup Language tags to annotate temporal expressions, by inserting them directly into the text. We adapted types of temporal expressions from TIMEX3: DATE, TIME, DURATION and SET.

In this work we would like to propose a new method for the temporal expressions normalisation, called Cascade of Partial Rules, which significantly improved the quality of the normalisation system.

2 Related Works

Previous works in this area concerned preparation of a broad description of Polish temporal expressions with annotation guidelines, based in the state-of-the-art solution for English, mainly TimeML specification. The final product, called

\(^{1}\)Time and Event Recognition for Question Answering Systems. An Advanced Research and Development Activity Workshop on Advanced Question Answering Technology

\(^{2}\)http://www.informedia.cs.cmu.edu/aquaint/index.html

\(^{3}\)https://code.google.com/p/heideltime/
PLIMEX (Kocóń et al., 2015), was extended with the solution to capture the local semantics of temporal expressions, by adapting LTIMEX (Mazur, 2012) specification to Polish. Temporal description also supports further event identification and extends event description model, focusing at anchoring events in time, ordering events and reasoning about the persistence of events. PLIMEX was designed to address these issues. All documents in Polish Corpus of Wroclaw University of Technology (KPWr) were annotated using PLIMEX annotation guidelines with adapted types of temporal expressions (timexes) from TIMEX3: Date, Time, Duration and Set. In the following examples of timexes, the extent of the annotation in text (if needed) is marked with square brackets. All English translations of Polish examples are given in parentheses. These examples are presented in (Kocóń and Marcinczuk, 2017), and a broad description of each timex type is presented in (Kocóń et al., 2015).

2.1 Types of Temporal Expressions

**Date** is a type of timex which denotes a point on a timeline, i.e., a unit of time greater than or equal to the day. The key question is *when*:

(1) [poniedziałek, 16 marca 1985 roku] (on [Monday, 16th March 1985])

**Time** describes timexes which refer to the time of day. The key question is also *when*. For example Smith wrócił (Smith returned):

(2) [dwadzieścia po dwunastej] (at [twenty past twelve])

**Duration**, in contrast to Date, has two points on a timeline associated with it – a start and an end point. A different name used in literature is period (Saquete et al., 2003). The key question is *how long*. For example Smith był tutaj (Smith stayed there):

(3) [dwa miesiące] (for [two months])

**Set** is relating to more than one instance of a time unit – either a point or a period. The key question is *how often*. Examples – Jan wraca pijany (John comes back drunk):

(4) co dwa dni (every two days)

2.2 Global Semantics

In order to describe timexes we adopted normalisation format from TimeML. The attribute VAL (Value) attached to the temporal expression is important in the normalisation context. It is a textual representation of a timex, which is assigned using guidelines described in the ISO-8601 standard. In order to describe it, the first letters of the English names that specify time are used (e.g. Year, Month, Day, hour, minute, BC – before Christ, AD – Anno Domini, Time, MORning, Half, Quarter, SUmmer, Period). Each timex can be determined with respect to the proper type, using the coding proposed in the ISO-8601 standard, e.g. calendar date (YYYY-MM-DD), week of the year (YYYY-Wxx), hour (hh:mm:ss), date & hour (YYYY-MM-DDTTh:mm:ss), duration (PxW).

Table 1 shows example values of the VAL attribute and the semantic meaning (Kocóń and Marcinczuk, 2017).

| VAL          | Meaning (EN) | Meaning (PL) |
|--------------|--------------|--------------|
| 1992         | year 1992    | rok 1992     |
| 1992-SU      | summer of 1992 | lato 1992 roku |
| BC0346       | year 346 BC  | 346 rok p.n.e. |
| P2Y          | 2 years      | 2 lata       |
| P3W          | 3 weeks      | 3 tygodnie   |
| PT8H2M       | 8 hours and 2 min. | 8 godz. i 2 min. |
| P2DE         | 2 decades    | dwie dekady |

Table 1: Examples of VAL attributes and the semantic meaning of timexes. VAL values can be assigned manually during annotation or by automatic system.

Normalisation in the TimeML standard involves the determination of the global semantics for a timex. There is no indirect form of notation of the local semantics. The introduction of the intermediate stage of determining the global semantics is reasonable from the normalisation point of view. It can be seen in systems that recognise timexes in the English language (e.g. HeidelTime⁴, (Strötgen and Gertz, 2013)), which often use their own intermediate standard of normalisation. For that purpose LTIMEX standard was adapted (Kocóń and Marcinczuk, 2017). It can be used to determine the local semantics of timexes.

⁴https://code.google.com/p/heideltime/
2.3 Local Semantics

LTIMEX standard was designed to be compatible with existing annotation schemes, especially TIMEX2 (and TIMEX3). It is beneficial for both design and evaluation to recognise the semantics of the expression with no context involved, what is called local semantics, representing the partial and underspecified context-free meaning of temporal expressions (Mazur, 2012). The compatibility with the existing schemes has two purposes:

- It is human-readable and requires minimum effort to use for annotators familiar with TIMEX3.

- It provides a relatively easy means of conversion from local semantics to global semantics.

If the temporal expression is explicit, there is no difference between LVAL and VAL representation. The specification is fully described in work (Kocóñ and Marcińczuk, 2017). Table 2 shows the example values of the LVAL attribute and the semantic meaning.

| LVAL      | Meaning (EN) | Meaning (PL) |
|-----------|--------------|--------------|
| xxxx-01-03| January 3rd  | 3 stycznia   |
| xxxx-xx-19| nineteenth   | dziewiątnasty|
| xxxx-SU   | summer       | lato         |
| -0000-00-01| yesterday    | wczoraj      |
| -0000-02  | two months ago| dwa miesiące temu|
| <M06      | last June    | ostatni czerwiec|
| 2D4       | second Thursday| drugi wtorek |
| $1M03     | last February| ostatni luty |

Table 2: Examples of LVAL attributes and the semantic meaning of *timexes*. LVAL values can be assigned manually during annotation or by automatic system.

3 Normalisation Improvement

The previous rule-based normalisation system is presented in work (Kocóñ and Marcińczuk, 2017) and contains 224 rules for local normalisation, for 3 classes of temporal expressions: *Date, Time, Duration*. Global normalisation is made as a Java code and contains 16 rules for two classes: *date, time* (for *duration* local and global interpretation is the same). Rules were both created and tuned using *train* data set.

We evaluated rules on *test* set. The evaluation was fully described in the article (UzZaman et al., 2013). We performed the evaluation on a full pipeline for the best model to recognise boundaries of entities and their types (trained on *train+tune* data set). We focus on the end-to-end comparison with other systems presented in the article (UzZaman et al., 2013).

3.1 Metrics

To evaluate temporal expressions we need to check how many entities are correctly identified, if the extents for the entities are correctly identified and how many attributes are correctly identified. We use classical precision, recall and F-measure for the recognition (UzZaman et al., 2013).

3.2 Previous Approach

We prepared module for Liner2 tool (Kocóñ and Marcińczuk, 2017) to perform the normalisation process in order to get the local and global meaning of temporal expressions. We created the normalisation process as a two-step rule-based approach. The first step is to get the local meaning of the temporal expression, and the second (created on the basis of the local interpretation) is to get the global meaning. Rules were created by domain experts. For each class of temporal expression we defined several files with rules, patterns and normalisation dictionaries. The whole number of manually prepared resources is presented in Table 3.

Table 3: Previous normalisation resources for determining local value (LVAL) of temporal expressions, prepared using *train* data set. Table presents number of *Rules, Patterns* and *Normalisations* for each *timex Type*.

| Type  | Rules | Patt. | Norm. |
|-------|-------|-------|-------|
| date  | 110   | 75    | 21    |
| time  | 93    | 52    | 13    |
| duration | 21 | 20    | 11    |
| SUM   | 224   | 147   | 45    |

This approach is very similar to method presented in HeidelTime, except the extraction part, which in Liner2 is performed using Conditional Random Fields. In case of HeidelTime, it is necessary to manually specify the extraction rules and corresponding normalisation rules. Because in our solution we have already recognised chunks, we proposed a rule-based method for partial normalisation of timex chunk constituents.
3.3 Cascade of Partial Rules

We observed that in previous approach the further increase of quality (above 75% of LV AL F1) became very expensive, because only very specific timex examples not covered by rules left and there was no option to cover more than 2-3 of these examples with a single rule. Furthermore we saw that many rules for time are simple extensions of rules for date timexes. Because in Liner2 there is machine learning solution to chunk and classify timexes, we decided to prepare partial rules, which can match specific parts of timexes, which, e.g., always denote year part in each timex containing that part, no matter if its type is date or time.

Because durations strongly differ from timexes describing points in a timeline, we prepared 2 sets of rules, the first for dates and times, the second for durations. Each set contains 3 subsets: keys, maps and rules. Examples:

```json
{ "keys": [ "digitM_written": ["zero", "jeden", "dwa", "trzy", "cztery", ...], //zero, one, two, three, four "timeM_written": [ "wieczór", "wieczor", "po północ", //evening, evening, after midnight "o północ", "północ", "rano", ...], //midnight, midnight, morning "tomorrowM_written": ["jutro", "tomorrow"], "następny dzień", "nazajutrz", //next day, morrow "dzień następny", "jutrzejszy"], //day after tomorrow "maps": { "time_written": { "wieczór": "EV", "wieczor": "EV", //evening, evening "o północ": "NI", "północ": "MO", //after midnight morning "północ": "24:00", ...}, //midnight "digit_written": [ "zero": "0", "jeden": "1", //zero one "dwa": "2", "trzy": "3", "cztery": "4", "pięć": "5", ...], "rules": [ {"desc": "[1] jutro", "keys": ["tomorrowM_written"], "groups": [], "match": "(?<date>$digitM_written)+", "map": [], "value": { "year": "+0000", "month": "00", "day": "01"}}, {"desc": "[2] rano", "keys": ["timeM_written"], "groups": ["[hour]", "match": "(?<hour>$timeM_written)"], "map": ["hour": "$hour"], "value": { "hour": "$hour", "separator": "$minute"}], {"desc": "[3] godzina 16.00", // hour "keys": [], "groups": ["hour", "minute"], "match": "godzina (?<hour>[0-9]+)\.(?<minute>[0-9]+)"], "map": [], "value": { "hour": "$hour", "minute": "$minute"}]}
```

Keys are key:value pairs, where value is a list of words. Maps are also key:value pairs, but each value is a key:value pair of word:word. Rules is a list of items. Each item consists of the following elements:

- description – name of the rule,
- keys – a list of keys used in the rule,
- groups – a list of group names used in match regular expression,
- match – regular expression for partial matching, may contain groups in format: (?<group_name>group_regex).

May also contain $key_name elements which are replaced by keys[key_name] joined with regex alternative character, e.g., $digitM_written would be zero|jeden|dwa|trzy|....

- map – a dictionary, where key is a group name used in match, and value is a map key,
- value – a dictionary, where key is a name of the final partial normalisation element and value is the final form of the timex part.

Consider the following example of time:

**Pol.:** jutro rano o godzinie 9:30

**Eng.:** tomorrow morning at 9:30

This example is covered by the whole cascade of the given example rules. Rule [1] is simple: if there is any of words from keys["tomorrowM_written"] (in our example it is jutro) then set the final partial normalisations: year="+0000", month="00", day="01". The order of rules is important, as the final values set by previous rules may be overwritten by next rules.

Rule [2] covers rano part and the match is: (?<hour>$timeM_written). All textual words from temporal expressions are lemmatised and match looks at the chain of lemmas. There is one named group in this match: hour. In this case groups["hour"]="rano". 
There is also a map defined as:
"map": {"hour": "time_written"}.
It means that the group from text has to be transformed like:
groups["hour"] = maps["time_written"]["rano"]
and after the transformation:
groups["hour"] = "MO". Final values to set are:
{"hour": "$hour", "separator": "$hour"}, so it looks like: hour = "MO", separator = "MO".

Rule [3] covers godzina (?<hour>%d%d?) part and the match is: godzina (?<hour>%d%d?) % (?<minute>%d%d?). There are two named groups: hour and minute, but there is no map defined in this rule, so values captured by groups remain unchanged: hour = "9", minute = "30". Note that rule [2] sets the final value of hour = "MO", but rule [3] overwrites this as hour = "9".

After the whole cascade of rules, the final values are:
year = "+0000", month = "00", day = "01", hour = "9", minute = "30", separator = "MO". These values are combined together to obtain the proper LVAL value, which is: +0000-00-01T09:30. Note, that separator is important in this case, because without the information about the exact time of a day rano (Eng. morning), it could be 9AM or 9PM and the final LVAL value would be: +0000-00-00T9:30 (small letter t instead of T, saying that the hour part of the local normalisation value is still ambiguous).

There are some other elements, which are not presented in examples, e.g., it is possible to specify the limit parameter to apply rule only to a defined subset of temporal types, for example this rule runs only on Time expressions:

```
{ "desc": "3 .", "keys": [ ], "groups": ["minute"], "match": "(?<minute>%d+)\s\("|\')\$", "map": {}, "limit": ["t3_time"], "value": { "year": "+0000", "month": "00", "day": "00", "hour": "+00", "minute": "$minute" } }
```

The number of items prepared in this approach is presented in Table 4.

| Type            | Rules | Keys | Maps |
|-----------------|-------|------|------|
| date & time     | 122   | 24   | 13   |
| duration        | 45    | 4    | 3    |
| SUM             | 167   | 28   | 16   |

### 4 Experiments and Results

Results in Table 5 show, that our system achieves similar quality as the best systems in case of extraction of temporal expressions. It is not easy to compare the quality for different languages, but the result shows, that the complexity of that task for Polish is more close to English in case of normalisation. A strict F-measure of Liner2 tool for the extraction task is very high (88.76%) and outperforms all other systems. In case of global normalisation there is probably still a room for improvement, mainly from the perspective of validating the training data. Still guidelines are not precise enough in case of determining the global value and it requires further work to prepare the exact procedure of determining VAL value for temporal expressions. Obtained results for VAL F1 are very close to those achieved by best system for English (HeidelTime) and the proposed method (Liner2-new) is almost 11 p.p. better than the previous one (Liner2-old).

### 5 Conclusions

The comparison of the recognition results (see: Section 3.1) for previous (Liner2-old) and current approach (Liner2-new) is presented in Table 5. We analysed the statistical significance of differences, using paired-differences Student’s t-test with a significance level $\alpha = 0.05$ (Dietterich, 1998). The improvement of the recognition quality is statistically significant.

The proposed Cascade of Partial Rules method, applied for determining the LVAL attribute for temporal expressions, outperformed the previous solution by almost 11 p.p. with using only 167 rules instead of 224. We believe, that the further improvement is still possible at the level of de-
| System         | Extr. | Lang. | Rel.F1 | Rel.P | Rel.R | Str.F1 | VAL F1 | LVAL F1 |
|---------------|-------|-------|--------|-------|-------|--------|--------|---------|
| HeidelTime    | RB    | SP    | 90.10  | 96.00 | 84.90 | 85.30  | 87.50  |
| TIPSem        | DD    | SP    | 87.40  | 93.70 | 81.90 | 82.60  | 82.00  |
| HeidelTime    | RB    | EN    | 90.30  | 93.08 | 87.68 | 81.34  | 77.61  |
| NavyTime      | RB    | EN    | 90.32  | 89.36 | 91.30 | 79.57  | 70.97  |
| ManTime       | DD    | EN    | 89.66  | 95.12 | 75.36 | 78.69  | 65.57  |
| SUTime        | DD    | EN    | 84.90  | 97.20 | 75.36 | 81.63  | 65.31  |
| ATT           | DD    | EN    | 85.25  | 98.11 | 75.36 | 78.69  | 65.57  |
| TIPSem        | DD    | EN    | 89.36  | 93.75 | 86.96 | 82.71  | 64.66  |
| JU-CSE        | DD    | EN    | 86.38  | 93.28 | 80.43 | 75.49  | 63.81  |
| KUL           | H     | EN    | 83.67  | 92.92 | 76.09 | 69.32  | 62.95  |
| FSS-TimEx     | RB    | SP    | 65.20  | 86.60 | 52.30 | 49.04  | 58.24  |
| FSS-TimEx     | RB    | EN    | 85.06  | 90.24 | 80.43 | 49.04  | 58.24  |
| Liner2-old    | DD    | PL    | 88.50  | 90.25 | 86.81 | 85.06  | 66.71  | 75.14  |
| Liner2-new    | DD    | PL    | 92.83  | 94.56 | 91.15 | 88.76  | 77.23  | 89.23  |

Table 5: Evaluation of the end-to-end Systems for recognition and normalisation of temporal expressions – a comparison of old and new version of Liner2 tool with other systems presented during SemEval 2013 (only the best results regarding VAL F1 measure for each tool). Extraction methods of temporal entities in these systems are: RB – rule-based, DD – data-driven, H – hybrid. Languages: SPanish, ENglish, PL – Polish. We used Strict and Relaxed variants of evaluation measures presented in Section 3.1, also with measures for VAL and LVAL.

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