BioNLP Shared Task 2011: Supporting Resources

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Abstract

This paper describes the supporting resources provided for the BioNLP Shared Task 2011. These resources were constructed with the goal to alleviate some of the burden of system development from the participants and allow them to focus on the novel aspects of constructing their event extraction systems. With the availability of these resources we also seek to enable the evaluation of the applicability of specific tools and representations towards improving the performance of event extraction systems. Additionally we supplied evaluation software and services and constructed a visualisation tool, stav, which visualises event extraction results and annotations. These resources helped the participants make sure that their final submissions and research efforts were on track during the development stages and evaluate their progress throughout the duration of the shared task. The visualisation software was also employed to show the differences between the gold annotations and those of the submitted results, allowing the participants to better understand the performance of their system. The resources, evaluation tools and visualisation tool are provided freely for research purposes and can be found at http://sites.google.com/site/bionlpst/

1 Introduction

For the BioNLP'09 Shared Task (Kim et al., 2009), the first in the ongoing series, the organisers provided the participants with automatically generated syntactic analyses for the sentences from the annotated data. For evaluation purposes, tools were made

publicly available as both distributed software and online services. These resources were well received. A majority of the participants made use of one or more of the syntactic analyses, which have remained available after the shared task ended and have been employed in at least two independent efforts studying the contribution of different tools and forms of syntactic representation to the domain of information extraction (Miwa et al., 2010; Buyko and Hahn, 2010). The evaluation software for the BioNLP'09 Shared Task has also been widely adopted in subsequent studies (Miwa et al., 2010; Poon and Vanderwende, 2010; Björne et al., 2010).

The reception and research contribution from providing these resources encouraged us to continue providing similar resources for the BioNLP Shared Task 2011 (Kim et al., 2011a). Along with the parses we also encouraged the participants and external groups to process the data with any NLP (Natural Language Processing) tools of their choice and make the results available to the participants.

We provided continuous verification and evaluation of the participating systems using a suite of inhouse evaluation tools. Lastly, we provided a tool for visualising the annotated data to enable the participants to better grasp the results of their experiments and to help gain a deeper understanding of the underlying concepts and the annotated data. This paper presents these supporting resources.

2 Data

This section introduces the data resources provided by the organisers, participants and external groups for the shared task.

Task	Provider	Tool
CO	University of Utah	Reconcile
CO	University of Zürich	UZCRS
CO	University of Turku	TEES
REL	University of Turku	TEES

Table 1: Supporting task analyses provided, TEES is the Turku Event Extraction System and UZCRS is the University of Zürich Coreference Resolution System

2.1 Supporting task analyses

The shared task included three Supporting Tasks: Coreference (CO) (Nguyen et al., 2011), Entity relations (REL) (Pyysalo et al., 2011b) and Gene renaming (REN) (Jourde et al., 2011). In the shared task schedule, the supporting tasks were carried out before the main tasks (Kim et al., 2011b; Pyysalo et al., 2011a; Ohta et al., 2011; Bossy et al., 2011) in order to allow participants to make use of analyses from the systems participating in the Supporting Tasks for their main task event extraction systems.

Error analysis of BioNLP'09 shared task submissions indicated that coreference was the most frequent feature of events that could not be correctly extracted by any participating system. Further, events involving statements of non-trivial relations between participating entities were a frequent cause of extraction errors. Thus, the CO and REL tasks were explicitly designed to support parts of the main event extraction tasks where it had been suggested that they could improve the system performance.

Table 1 shows the supporting task analyses provided to the participants. For the main tasks, we are currently aware of one group (Emadzadeh et al., 2011) that made use of the REL task analyses in their system. However, while a number of systems involved coreference resolution in some form, we are not aware of any teams using the CO task analyses specifically, perhaps due in part to the tight schedule and the somewhat limited results of the CO task. These data will remain available to allow future research into the benefits of these resources for event extraction.

2.2 Syntactic analyses

For syntactic analyses we provided parses for all the task data in various formats from a wide range of parsers (see Table 2). With the exception of the Pro3Gres¹ parser (Schneider et al., 2007), the parsers were set up and run by the task organisers. The emphasis was put on availability for research purposes and variety of parsing models and frameworks to allow evaluation of their applicability for different tasks.

In part following up on the results of Miwa et al. (2010) and Buyko and Hahn (2010) regarding the impact on performance of event extraction systems depending on the dependency parse representation, we aimed to provide several dependency parse formats. Stanford Dependencies (SD) and Collapsed Stanford Dependencies (SDC), as described by de Marneffe et al. (2006), were generated by converting Penn Treebank (PTB)-style (Marcus et al., 1993) output using the Stanford CoreNLP Tools2 into the two dependency formats. We also provided Conference on Computational Natural Language Learning style dependency parses (CoNLL-X) (Buchholz and Marsi, 2006) which were also converted from PTBstyle output, but for this we used the conversion tool³ from Johansson and Nugues (2007). While this conversion tool was not designed with converting the output from statistical parsers in mind (but rather to convert between treebanks), it has previously been applied successfully for this task (Miyao et al., 2008; Miwa et al., 2010).

The text from all documents provided were split into sentences using the Genia Sentence Splitter⁴ (Sætre et al., 2007) and then postprocessed using a set of heuristics to correct frequently occurring errors. The sentences were then tokenised using a tokenisation script created by the organisers intended to replicate the tokenisation of the Genia Tree Bank (GTB) (Tateisi et al., 2005). This tokenised and sentence-split data was then used as input for all parsers.

We used two deep parsers that provide phrase structure analysis enriched with deep sentence struc-

https://files.ifi.uzh.ch/cl/gschneid/parser/

²http://nlp.stanford.edu/software/corenlp.shtml

³http://nlp.cs.lth.se/software/treebank_converter/

⁴http://www-tsujii.is.s.u-tokyo.ac.jp/~y-matsu/geniass/

Name	Format(s)	Model	Availability	BioNLP'09
Berkeley	PTB, SD, SDC, CoNLL-X	News	Binary, Source	No
C&C	CCG, SD	Biomedical	Binary, Source	Yes
Enju	HPSG, PTB, SD, SDC, CoNLL-X	Biomedical	Binary	No
GDep	CoNLL-X	Biomedical	Binary, Source	Yes
McCCJ	PTB, SD, SDC, CoNLL-X	Biomedical	Source	Yes
Pro3Gres	Pro3Gres	Combination	_	No
Stanford	PTB, SD, SDC, CoNLL-X	Combination	Binary, Source	Yes

Table 2: Parsers, the formats for which their output was provided and which type of model that was used. The availability column signifies public availability (without making an explicit request) for research purposes

tures, for example predicate-argument structure for Head-Driven Phrase Structure Grammar (HPSG). First we used the C&C Combinatory Categorial Grammar (CCG) parser⁵ (C&C) by Clark and Curran (2004) using the biomedical model described in Rimell and Clark (2009) which was trained on GTB. Unlike all other parsers for which we supplied SD and SDC dependency parses, the C&C output was converted from its native format using a separate conversion script provided by the C&C authors. Regrettably we were unable to provide CoNLL-X format output for this parser due to the lack of PTB-style output. The other deep parser used was the HPSG parser Enju⁶ by Miyao and Tsujii (2008), also trained on GTB.

We also applied the frequently adopted Stanford Parser⁷ (Klein and Manning, 2003) using a mixed model which includes data from the biomedical domain, and the Charniak Johnson re-ranking parser⁸ (Charniak and Johnson, 2005) using the self-trained biomedical model from McClosky (2009) (McCCJ).

For the BioNLP'09 shared task it was observed that the Bikel parser⁹ (Bikel, 2004), which used a non-biomedical model and can be argued that it uses the somewhat dated Collins' parsing model (Collins, 1996), did not contribute towards event extraction performance as strongly as other parses supplied for the same data. We therefore wanted to supply a parser that can compete with the ones above in a domain which is different from the biomedical domain to see whether conclusions could be drawn as to the

importance of using a biomedical model. For this we used the Berkeley parser¹⁰ (Petrov et al., 2006). Lastly we used a native dependency parser, the GE-NIA Dependency parser (GDep) by Sagae and Tsujii (2007).

At least one team (Choudhury et al., 2011) performed experiments on some of the provided lexical analyses and among the 14 submissions for the EPI and ID tasks, 13 submissions utilised tools for which resources were provided by the organisers of the shared task. We intend to follow up on whether or not the majority of the teams ran the tools themselves or used the provided analyses.

2.3 Other analyses

The call for analyses was open to all interested parties and all forms of analysis. In addition to the Supporting Task analyses (CO and REL) and syntactic analyses provided by various groups, the University of Antwerp CLiPS center (Morante et al., 2010) responded to the call providing negation/speculation analyses in the BioScope corpus format (Szarvas et al., 2008).

Although this resource was not utilised by the participants for the main task, possibly due to a lack of time, it is our hope that by keeping the data available it can lead to further development of the participating systems and analysis of BioScope and BioNLP ST-style hedging annotations.

3 Tools

This section presents the tools produced by the organisers for the purpose of the shared task.

⁵http://svn.ask.it.usyd.edu.au/trac/candc/

⁶http://www-tsujii.is.s.u-tokyo.ac.jp/enju/

⁷http://nlp.stanford.edu/software/lex-parser.shtml

⁸ftp://ftp.cs.brown.edu/pub/nlparser/

⁹http://www.cis.upenn.edu/~dbikel/software.html

¹⁰http://code.google.com/p/berkeleyparser/

Figure 1: Text output from the BioNLP'09 Shared Event Viewer with line numbering and newline markings

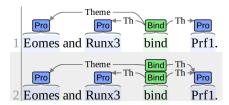


Figure 2: An illustration of collective (sentence 1) and distributive reading (sentence 2). "Theme" is abbreviated as "Th" and "Protein" as "Pro" when there is a lack of space

3.1 Visualisation

The annotation data in the format specified by the shared task is not intended to be human-readable – yet researchers need to be able to visualise the data in order to understand the results of their experiments. However, there is a scarcity of tools that can be used for this purpose. There are three available for event annotations in the BioNLP ST format that we are aware of.

One is the BioNLP'09 Shared Task Event Viewer¹¹, a simple text-based annotation viewer: it aggregates data from the annotations, and outputs it in a format (Figure 1) that is meant to be further processed by a utility such as grep.

Another is What's Wrong with My NLP¹², which visualises relation annotations (see Figure 3a) – but is unable to display some of the information contained in the Shared Task data. Notably, the distributive and collective readings of an event are not distinguished (Figure 2). It also displays all annotations on a single line, which makes reading and analysing longer sentences, let alone whole documents, somewhat difficult.

The last one is U-Compare¹³ (Kano et al., 2009),

which is a comprehensive suite of tools designed for managing NLP workflows, integrating many available services. However, the annotation visualisation component, illustrated in Figure 3b, is not optimised for displaying complex event structures. Each annotation is marked by underlining its text segment using a different colour per annotation type, and a role in an event is represented by a similarly coloured arc between the related underlined text segments. The implementation leaves some things to be desired: there is no detailed information added in the display unless the user explicitly requests it, and then it is displayed in a separate panel, away from the text it annotates. The text spacing makes no allowance for the annotations, with opaque lines crossing over it, with the effect of making both the annotations and the text hard to read if the annotations are above a certain degree of complexity.

As a result of the difficulties of these existing tools, in order to extract a piece of annotated text and rework it into a graph that could be embedded into a publication, users usually read off the annotations, then create a graph from scratch using vector drawing or image editing software.

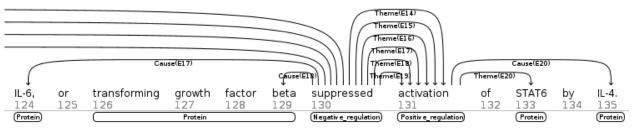
To address these issues, we created a visualisation tool named *stav* (*stav* Text Annotation Visualizer), that can read the data formatted according to the Shared Task specification and aims to present it to the user in a form that can be grasped at a glance. Events and entities are annotated immediately above the text, and the roles within an event by labelled arcs between them (Figure 3c). In a very complex graph, users can highlight the object or association of interest to follow it even more easily. Special features of annotations, such as negation or speculation, are shown by unique visual cues, and more in-depth, technical information that is usually not required can be requested by floating the mouse cursor over the annotation (as seen in Figure 5).

We took care to minimise arc crossovers, and to

¹¹http://www-tsujii.is.s.u-tokyo.ac.jp/GENIA/SharedTask/downloads.shtml

¹²http://code.google.com/p/whatswrong/

¹³http://u-compare.org/bionlp2009.html



(a) Visualisation using What's Wrong with My NLP

we examined the ability of type I and type II IFNs to regulate activation of STAT6 by IL-4 in primary human monocytes. Pretreatment of monocytes with IFN-beta or IFN-gCause, but not I L-1, IL-2, macrophage colory-activating factor, granulocyte/macrophage colony-stimulating factor, IL-6, or transforming growth factor acceptable to suppressed actGause on of STAT6 by IL-4. The Theme Negation ham type in and Cause type in the suppressed actGause on of STAT6 and was not evident unless the celThemes re preincubated with IFN for at least 1

(b) Visualisation using U-Compare

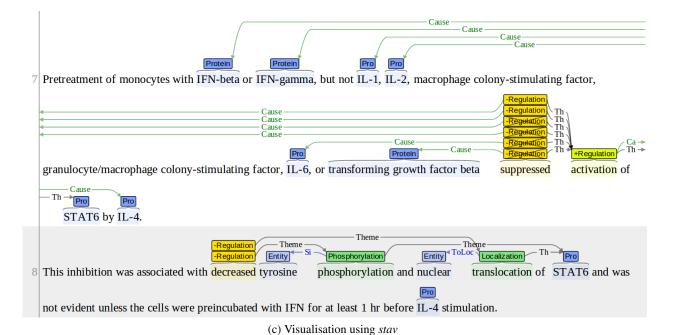


Figure 3: Different visualisations of complex textual annotations of Dickensheets et al. (1999)

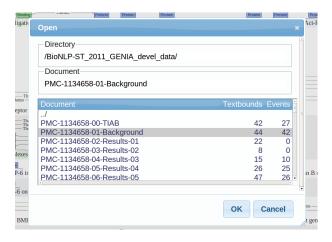


Figure 4: A screenshot of the stav file-browser

keep them away from the text itself, in order to maintain text readability. The text is spaced to accommodate the annotations between the rows. While this does end up using more screen real-estate, it keeps the text legible, and annotations adjacent to the text. The text is broken up into lines, and each sentence is also forced into a new line, and given a numerical identifier. The effect of this is that the text is laid out vertically, like an article would be, but with large spacing to accomodate the annotations. The arcs are similarly continued on successive lines, and can easily be traced – even in case of them spanning multiple lines, by the use of mouseover highlighting. To preserve the distributionality information of the annotation, any event annotations are duplicated for each event, as demonstrated in the example in Figure 2.

stav is not limited to the Shared Task datasets with appropriate configuration settings, it could also visualise other kinds of relational annotations such as: frame structures (Fillmore, 1976) and dependency parses (de Marneffe et al., 2006).

To achieve our objectives above, we use the Dynamic Scalable Vector Graphics (SVG) functionality (i.e. SVG manipulated by JavaScript) provided by most modern browsers to render the WYSIWYG (What You See Is What You Get) representation of the annotated document. An added benefit from this technique is that the installation process, if any, is very simple: although not all browsers are currently supported, the two that we specifically tested

against are Safari¹⁴ and Google Chrome¹⁵; the former comes preinstalled with the Mac OS X operating system, while the latter can be installed even by relatively non-technical users. The design is kept modular using a dispatcher pattern, in order to allow the inclusion of the visualiser tool into other JavaScript-based projects. The client-server architecture also allows centralisation of data, so that every user can inspect an uploaded dataset without the hassle of downloading and importing into a desktop application, simply by opening an URL which can uniquely identify a document, or even a single annotation. A screenshot of the *stav* file browser can be seen in Figure 4.

3.2 Evaluation Tools

The tasks of BioNLP-ST 2011 exhibit very high complexity, including multiple non-trivial subproblems that are partially, but not entirely, independent of each other. With such tasks, the evaluation of participating systems itself becomes a major challenge. Clearly defined evaluation criteria and their precise implementation is critical not only for the comparison of submissions, but also to help participants follow the status of their development and to identify the specific strengths and weaknesses of their approach.

A further challenge arising from the complexity of the tasks is the need to process the relatively intricate format in which annotations are represented, which in turn carries a risk of errors in submissions. To reduce the risk of submissions being rejected or the evaluation showing poor results due to formatting errors, tools for checking the validity of the file format and annotation semantics are indispensable.

For these reasons, we placed emphasis in the organisation of the BioNLP-ST'11 on making tools for format checking, validation and evaluation available to the participants already during the early stages of system development. The tools were made available in two ways: as downloads, and as online services. With downloaded tools, participants can perform format checking and evaluation at any time without online access, allowing more efficient optimisation processes. Each task in BioNLP-ST also

¹⁴http://www.apple.com/safari

¹⁵ http://www.google.com/chrome

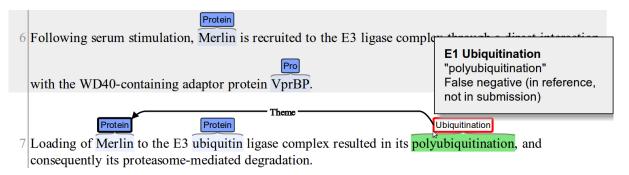


Figure 5: An example of a false negative illustrated by the evaluation tools in co-ordination with stav

maintained an online evaluation tool for the development set during the development period. The online evaluation is intended to provide an identical interface and criteria for submitted data as the final online submission system, allowing participants to be better prepared for the final submission. With online evaluation, the organisers could also monitor submissions to ensure that there were no problems in, for example, the evaluation software implementations.

The system logs of online evaluation systems show that the majority of the participants submitted at least one package with formatting errors, confirming the importance of tools for format checking. Further, most of the participants made use of the online development set evaluation at least once before their final submission.

To enhance the evaluation tools we drew upon the *stav* visualiser to provide a view of the submitted results. This was done by comparing the submitted results and the gold data to produce a visualisation where errors are highlighted, as illustrated in Figure 5. This experimental feature was available for the EPI and ID tasks and we believe that by doing so it enables participants to better understand the performance of their system and work on remedies for current shortcomings.

4 Discussion and Conclusions

Among the teams participating in the EPI and ID tasks, a great majority utilised tools for which resources were made available by the organisers. We hope that the continued availability of the parses will encourage further investigation into the applicability of these and similar tools and representations.

As for the analysis of the supporting analyses provided by external groups and the participants, we are so far aware of only limited use of these resources among the participants, but the resources will remain available and we are looking forward to see future work using them.

To enable reproducibility of our resources, we provide a publicly accessible repository containing the automated procedure and our processing scripts used to produce the released data. This repository also contains detailed instructions on the options and versions used for each parser and, if the software license permits it, includes the source code or binary that was used to produce the processed data. For the cases where the license restricts redistribution, instructions and links are provided on how to obtain the same version that was used. We propose that using a multitude of parses and formats can benefit not just the task of event extraction but other NLP tasks as well.

We have also made our evaluation tools and visualisation tool *stav* available along with instructions on how to run it and use it in coordination with the shared task resources. The responses from the participants in relation to the visualisation tool were very positive, and we see this as encouragement to advance the application of visualisation as a way to better reach a wider understanding and unification of the concept of events for biomedical event extraction.

All of the resources described in this paper are available at http://sites.google.com/site/bionlpst/.

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