Bureau for Rapid Annotation Tool: Collaboration can do More over Variety-oriented Annotations

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ABSTRACT

A high-quality manually annotated corpus is crucial for many text mining and information extraction tasks. Several workbenches have been developed in the literature to facilitate collaborative annotation. However, given the growing volumes of un-annotated documents, these variety-oriented annotation workbenches have many shortcoming in terms of teamwork, quality control and time effort. For this purpose, we develop a novel workbench such that collaboration can do more over variety-oriented annotation. Our workbench is named as Bureau for Rapid Annotation Tool (Brat for short). Main functionalities include enhanced semantic constraint system, Vim-like shortcut keys, annotation filter and graph-visualizing annotation browser. Until now, over 500,000 mentions have been annotated with our Brat workbench.

1 INTRODUCTION

A high-quality manually annotated corpus is very crucial for many text mining and information extraction tasks [1, 3, 4, 6, 7, 14, 15]. Several workbenches have been developed in the literature to facilitate collaborative annotation [8, 11, 13, 16]. However, given the growing volumes of un-annotated documents, these variety-oriented workbenches still have many shortcomings in terms of teamwork, quality control and time effort. Let's take the sentence "Depending on the model, a Tesla costs somewhere between 1 and 3.33 BTC" [2] as an example. A practical issue we face is whether or not to assign "Price" type to the mention "between 1 and 3.33 BTC". This actually depends on a consensus acknowledging *Bitcoin* as actual money [5, 17].

Reaching this consensus is extremely time-consuming and heavily rely on two types of annotation collaborations in Table 1: *grounded collaboration* and *trusted collaboration*. By grounded collaboration, we mean that the resulting annotators are restricted with sounded pre-arrangements. For example, U-Compare only supports named entity annotations in the UIMA-type system [10], which can avoid many conflicts. An alternative [13] takes the form of semantic constraints. In more details, a certain relationship should take parameters with specific entity types. As for trusted collaboration, YEDDA [16] recognized common gestures from BRAT [13] and embedded many functionalities including teamwork, multi-annotator analysis and pairwise annotators comparison. Then, on the basis of various annotations, the inter-project agreement can be calculated. Another strategy of trusted collaboration, user-independent workspace, was utilized in TeamTat [8].

This paper combines these two types of annotation collaborations to structure various mentions annotated by each annotator

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Table 1: Two types of annotation collaborations used in previous workbenches.

Grounded	Trusted
UIMA-type system[10]	Teamwork [8, 16]
Annotation semantic constraint [13]	Personal workspace TeamTat [16]
	Multi-annotator analysis [8]
	Pairwise annotators comparison [8]

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2 FUNCTIONALITIES

2.1 Enhanced Semantic Constraint System

It is well known that not all parameters are valid to a specific relationship. To limit invalid annotated results for an annotation project, its manager can customize the schema at any time. Once the schema is modified, all involved annotated mentions will be adjusted correspondingly. A readable name is usually assigned to each type of entity and relation. In addition, a list of rules are also attached to expression the constraint conditions between parameters in each type of relation. In this way, the understanding on entities and relations from the manager can be delivered to all annotators.

2.2 Vim-like Shortcut Key

According to our observation, the conventional annotating operations (marking, selecting and confirming [13]) is time-expensive to choose a proper candidate from more than 5 entity types or relationships. To speed up the annotation procedure, our workbench embeds many Vim-like shortcut keys [12]. In this time, one can annotate smoothly an entity by the following steps (cf. Figure 2): 1) to move cursor and select a span of text with Figure 1, 2) to acknowledge one command from recommended candidates with TAB and ENTER, 3) to type leading characters and confirm entity types. Similar operations can be followed for relation mention annotation. It is worth noting that the key feature of this functionality is code auto-completion. This is based on enhanced semantic constraint system and polymorphic type inference [9].

2.3 Configurable Annotation Filter

It's unknown in advance how many mentions should be annotated for a single document, especially a very long document. To correct wrong annotations in time and reduce the conflicts among multiple documents, a feasible solution is to only display the mentions

Figure 1: Vim-like Shortcut Keys Mapping
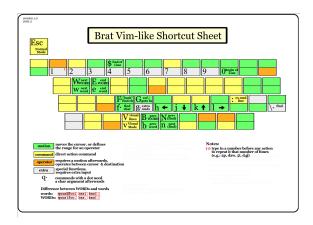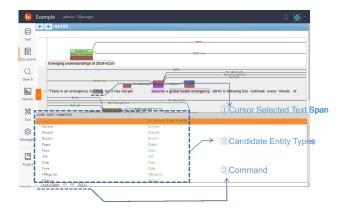

Figure 2: The novel annotation procedure powered by Vim-like Shortcut Keys

Figure 3: The visualization of the entity "Disk" and its relevant information in the TFH-2020 corpus [4] via our browser

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with interested types in current workspace. Thereupon, we provide a configurable annotation filter by toggling or un-toggling entity types and relationships.

2.4 Graph-visualizing Browser

In real-world scenario, it is not trivial to reach an agreement when multiple annotators are involved, and an entity or relation is mentioned simultaneously in multiple documents. To inspect the underling disagreements, our workbench can load and index all texts, mentions and their types, and visualize them in a graph browser, as illustrated in Figure 3.

3 CONCLUSION

Many projects utilized our Brat workbench to annotate interested entities and/or relations, and inspect potential conflicts over variety-oriented annotations. Nowadays, over 500,000 mentions have been annotated with our Brat workbench. In the near future, the Vim-like shortcut keys will be strengthen further, and machine learning methods will be incorporated to accelerate conflict inspection.
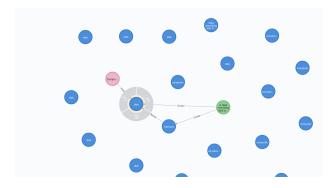
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