

Wiktionary Matcher Results for OAEI 2020

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Abstract. This paper presents the results of the *Wiktionary Matcher* in the *Ontology Alignment Evaluation Initiative* (OAEI) 2020. *Wiktionary Matcher* is an ontology matching tool that exploits *Wiktionary* as external background knowledge source. Wiktionary is a large lexical knowledge resource that is collaboratively built online. Multiple current language versions of Wiktionary are merged and used for monolingual ontology matching by exploiting synonymy relations and for multilingual matching by exploiting the translations given in the resource. This is the second OAEI participation of the matching system. *Wiktionary Matcher* has been improved and is the best performing system on the knowledge graph track this year.³

Keywords: Ontology Matching · Ontology Alignment · External Resources · Background Knowledge · Wiktionary

1 Presentation of the System

1.1 State, Purpose, General Statement

The *Wiktionary Matcher* is an element-level, label-based matcher which uses an online lexical resource, namely *Wiktionary*. The latter is "[a] collaborative project run by the Wikimedia Foundation to produce a free and complete dictionary in every language"⁴. The dictionary is organized similarly to Wikipedia: Everybody can contribute to the project and the content is reviewed in a community process. Compared to WordNet [2], Wiktionary is significantly larger and also available in other languages than English. This matcher uses *DBnary* [13], an RDF version of Wiktionary that is publicly available⁵. The *DBnary* dataset makes use of an extended *LEMON* model [7] to describe the data. For this

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⁴ see <https://web.archive.org/web/20190806080601/https://en.wiktionary.org/wiki/Wiktionary>

⁵ see <http://kaiko.getalp.org/about-dbnary/download/>

matcher, recent DBnary datasets for 8 Wiktionary languages⁶ have been downloaded and merged into one RDF graph. Triples not required for the matching algorithm, such as glosses, were removed in order to increase the performance of the matcher and to lower its memory requirements. As Wiktionary contains translations, this matcher can work on monolingual and multilingual matching tasks.

This is the second OAEI participation of this matching system, *Wiktionary Matcher* initially participated in the OAEI in 2019 [10]. The matcher has been implemented and packaged using the *Matching Evaluation Toolkit (MELT)*⁷, a Java framework for matcher development, tuning, evaluation, and packaging [4,9].

1.2 Specific Techniques Used

This matching system system was initially introduced at the OAEI 2019 [10]. An overview of the matching system is provided in Figure 1. The main techniques used for matching are summarized below.

Monolingual Matching For monolingual ontologies, the matching system first applies multiple string matching techniques. Afterwards, the synonym matcher module links labels to concepts in Wiktionary and checks then whether the concepts are synonymous in the external dataset. This approach is conceptually similar to an upper ontology matching approach. Concerning the usage of a collaboratively built knowledge source, the approach is similar to *WikiMatch* [3] which exploits the Wikipedia search engine. *Wiktionary Matcher* adds a correspondence to the final alignment purely based on the synonymy relation independently of the actual word sense. This is done in order to avoid word sense disambiguation on the ontology side but also on Wiktionary side: Versions for some countries do not annotate synonyms and translations for senses but rather on the level of the lemma. Hence, many synonyms are given independently of the word sense. In such cases, word-sense-disambiguation would have to be performed also on Wiktionary [8]. The linking process is similar to the one presented for the *ALOD2Vec 2018* matching system [12]: In a first step, the full label is looked up in the knowledge source. If the label cannot be found, labels consisting of multiple word tokens are truncated from the right and the process is repeated to check for sub-concepts. This allows to detect long sub-concepts even if the full string cannot be found. Label *conference banquet* of concept *http://ekaw#Conference_Banquet* from the *Conference* track, for example, cannot be linked to the background dataset using the full label. However, by applying right-to-left truncation, the label can be linked to two concepts, namely *conference* and *banquet*, and in the following also be matched to the correct concept *http://edas#ConferenceDinner* which is linked in the same fashion. For multi-linked concepts (such as *conference dinner*), a match is only annotated

⁶ Namely: Dutch, English, French, Italian, German, Portugese, Russian, and Spanish.

⁷ see <https://github.com/dwslab/melt>

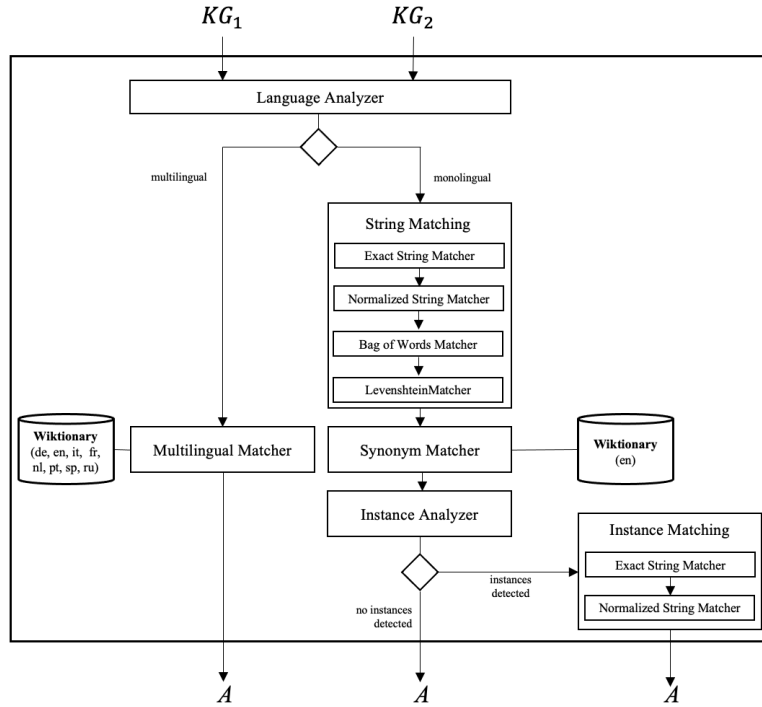


Fig. 1. High-level overview of the *Wiktionary Matcher*. KG_1 and KG_2 represent the input ontologies and optionally instances. The final alignment is referred to as A .

if every linked component of the label is synonymous to a component in the other label. Therefore, *lens* (http://mouse.owl#MA_0000275) is not mapped to *crystalline.lens* (http://human.owl#NCI_C12743) due to a missing synonymous partner for *crystalline* whereas *urinary bladder neck* (http://mouse.owl#MA_0002491) is matched to *bladder neck* (http://human.owl#NCI_C12336) because *urinary bladder* is synonymous to *bladder*.

Multilingual Matching For every matching task, the system first determines the language distributions in the ontologies. If the ontologies appear to be in different languages, the system automatically enables the multilingual matching module: Here, Wiktionary translations are exploited: A match is created, if one label can be translated to the other one according to at least one Wiktionary language version – such as the Spanish label *ciudad* and the French label *ville* (both meaning *city*). This process is depicted in Figure 2: The Spanish label is linked to the entry in the Spanish Wiktionary and from the entry the translation is derived. If there is no Wiktionary version for the languages to be matched or the approach described above yields very few results, it is checked whether the

two labels appear as a translation for the same word. The Chinese label 決定 (juédìng), for instance, is matched to the Arabic label قرار (qrār) because both appear as a translation of the English word *decision* on Wiktionary. This (less precise) approach is particularly important for language pairs for which no Wiktionary dataset is available to the matcher (such as Chinese and Arabic). The process is depicted in Figure 3: The Arabic and Chinese labels cannot be linked to Wiktionary entries but, instead, appear as translation for the same concept.

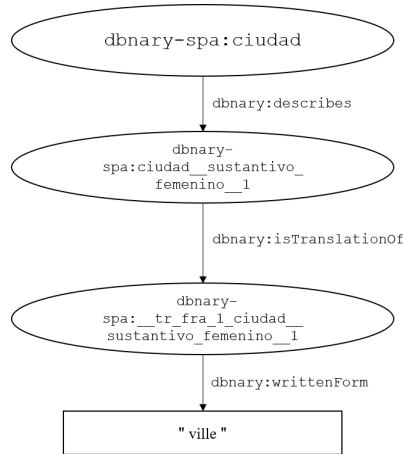


Fig. 2. Translation via the Wiktionary headword (using the DBnary RDF graph). Here: One (of more) French translations for the Spanish word *ciudad* in the Spanish Wiktionary.

Instance Matching The matcher presented in this paper can be also used for combined schema and instance matching tasks. If instances are available in the given datasets, the matcher applies a two step strategy: After aligning the schemas, instances are matched using a string index. As there are typically many instances, Wiktionary is not used for the instance matching task in order to increase the matching runtime performance. Moreover, the coverage of schema level concepts in Wiktionary is much higher than for instance level concepts: For example, there is a sophisticated representation of the concept *movie*⁸, but hardly any individual movies in Wiktionary. For correspondences where the instances belong to classes that were matched before, a higher confidence is assigned. If one instance matches multiple other instances, the correspondence is preferred where both their classes were matched before.

Explainability Unlike many other ontology matchers, this matcher uses the extension capabilities of the alignment format [1] in order to provide a human

⁸ see <https://en.wiktionary.org/wiki/movie>

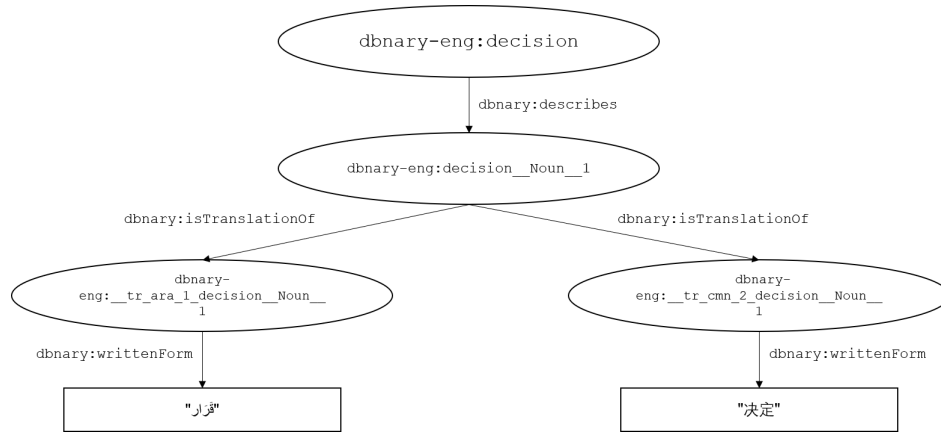


Fig. 3. Translation via the written forms of Wiktionary entries (using the DBnary RDF graph). Here: An Arabic and a Chinese label appear as translation for the same Wiktionary entry (*decision* in the English Wiktionary).

readable explanation of why a correspondence was added to the final alignment. Such explanations can help to interpret and to trust a matching system’s decision. Similarly, explanations also allow to comprehend why a correspondence was falsely added to the final alignment: The explanation for the false positive match (<http://confOf#Contribution>, <http://iasted#Tax>), for instance, is given as follows: “The first concept was mapped to dictionary entry [contribution] and the second concept was mapped to dictionary entry [tax]. According to Wiktionary, those two concepts are synonymous.” Here, it can be seen that the matcher was successful in linking the labels to but failed due to the missing word sense disambiguation. In order to explain a correspondence, the `description` property⁹ of the *Dublin Core Metadata Initiative* is used.

1.3 Extensions to the Matching System for the 2020 Campaign

For the 2020 campaign, the matching system has been improved. The instance matching module has been extended to better exploit the string indices. As a consequence, the matcher is the best performing system in the knowledge graph track [6] this year. Furthermore, *Wiktionary Matcher* now gives more detailed explanations in terms of why a correspondence has been added to the alignment. Lastly, the background knowledge has been updated: The system uses Wiktionary dumps as of late July 2020. The 2020 system uses the latest version of MELT [5]. The implementation is now also publicly available on GitHub.¹⁰

⁹ see <http://purl.org/dc/terms/description>

¹⁰ see <https://github.com/janathan/WiktionaryMatcher>

2 Results

2.1 Anatomy Track

On the anatomy track, recall and F_1 could be improved compared to the 2019 version of the matcher. Due to further improvements of the implementation, the matching system’s runtime performance could be significantly increased and the system is able to align the two ontologies in less than 100 seconds.¹¹ The system performs above the median of all 2020 systems with an F_1 score of 0.842 (precision = 0.956, recall = 0.753).

2.2 Conference Track

The matching system achieves almost the same results as in 2019 on the conference track with a slightly improved precision. With an F_1 score of 0.65 on rar2-M1, the system performs slightly above the median in terms of F_1 .

2.3 Multifarm Track

Wiktionary Matcher is one of the few systems capable of matching multilingual ontologies. This year, *Wiktionary Matcher* is the system with the highest precision on the aggregated results (precision = 0.8 on different ontologies). In terms of f-measure, the system scores at the exact median. Compared to the 2019 campaign, the results improved slightly. This effect is caused by the updated DBnary dataset used this year – the system improved itself due to a growing knowledge source (the multilingual matching implementation has not been changed compared to 2019).

2.4 LargeBio Track

Although the system has not been optimized for the LargeBio track, the matcher could complete all matching tasks within the given time. The system performs surprisingly competitive despite not using any other background knowledge source than Wiktionary. With the exception of task “FMA/NCI Whole”, the matching system performed significantly better than the 2019 version in terms of F_1 . A small contributor to better results is also the new Wiktionary version which carries more synonyms in 2020 than in 2019.

2.5 Knowledge Graph Track

Due to an improved instance matching module, the overall instance matching performance in terms of F_1 could be increased from 0.79 to 0.87. With an overall

¹¹ In the 2020 campaign, only 4 out of 11 systems were able to align the ontologies in less than 100 seconds.

f-measure of 0.87, *Wiktionary Matcher* is the best matching system on this track.¹²

3 General Comments

It is important to note that the matching system currently exploits only a small share of semantic relations available on Wiktionary. The system is restricted by the available relations extracted by the DBnary project. The additional exploitation of the relations *alternative forms* or *derived terms*, for instance, would likely improve the system. However, those are not yet extracted and are consequently not used for the matching task as of today.¹³

4 Conclusion

In this paper, we presented the *Wiktionary Matcher*, a matcher utilizing a collaboratively built lexical resource, as well as the results of the system in the 2020 OAEI campaign. Overall, the results of the matching system could be significantly improved compared to its last OAEI participation. Given Wiktionary’s continuous growth, it can be expected that the matching results will improve over time – for example when additional synonyms and translations are added. Small improvements due to new synonyms and translations could already be observed within a one year time frame for example on the Multifarm or the LargeBio track. In addition, improvements to the DBnary dataset, such as the addition of alternative word forms, may also improve the overall matcher performance in the future.

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¹² *ALOD2Vec Matcher 2020* [11] achieves the same F_1 score – however, as the performance of the latter matcher on classes and properties is slightly worse, *Wiktionary Matcher* comes in first.

¹³ We contacted the developers and will include the additional relations in our matching system as soon as those are available.

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