Serge Sharoff · Reinhard Rapp · Pierre Zweigenbaum

Building and Using Comparable Corpora for Multilingual Natural Language Processing



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Building and Using Comparable Corpora for Multilingual Natural Language Processing



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Introduction

1.1 Rationale for Working with Comparable Corpora

The ability of the computers to handle larger amounts of texts and the availability of more texts in electronic form led to the rise of data driven research in computational linguistics. In the case of Machine Translation (MT) and other kinds of multilingual Natural Language Processing (NLP), the first source of large data came from collections of translations, initially in the Statistical MT (SMT) approach from IBM [1], which was based on the Proceedings of the Canadian Parliament in English and French as their data source. This research direction was followed by a proliferation of SMT models, which relied on larger and larger collections of *parallel* data, which consist of exact translations between a pair of languages or several languages at the same time.

However, this early research also demonstrated the limits of using fully parallel corpora, with the most pressing initial concern on the amount of such data [4, 5]. This led to another strand of studies concerning the use of *less* parallel sources of texts, usually under the name of comparable corpora [6]. This book will present an overview of the modern approaches to building and using comparable data in multilingual NLP.

We will start by outlining the basic principles of using comparable resources as well as by comparing them to fully parallel resources (this chapter). This will be followed by specific chapters on building comparable corpora (Chap. 3), aligning their sentences to find a database of suitable translations (Chap. 4), using these corpora to produce dictionaries and termbanks (Chap. 5), to build MT engines (Chap. 6) and to use them in other applications (Chap. 7).



1

1.1.1 Availability of Truly Parallel Data

The first limitation in the use of parallel corpora comes from the process of their production. For truly parallel corpora we need to collect texts that have been carefully translated by highly trained professional translators [8]. Many more people produce monolingual texts in their native languages in comparison to output of a small number of trained translators. Also there is an imbalance in the amount of translations produced for a relatively small number of major languages, primarily for the languages of the United Nations or the EU, while there are thousands of languages with very minor resources. Less resourced languages can still benefit from parallel resources. However, statistically speaking, language can be described as a large number of rare events in the sense that an individual word or expression is relatively infrequent, but the totality of rare events contributes to the mass probability of words and expressions in a text [9]. This creates problems of sparsity even for better-resourced languages. For example, the word *unicyclist* occurs 84 times in 2 billion words of a monolingual English ukWac corpus [10], i.e., about once per 23 million words in ukWac, with no examples of this word in the English parts of large publicly available parallel corpora such as Europarl [15] or the United Nations corpus [16].

Another constraint on corpus data concerns the availability of translation products, as it is easier to obtain translated texts produced by large public bodies, such as the European Parliament or the United Nations than many other kinds of translations. This bias leads to many word choices, which are specific with respect to the genres and topics available in such corpora. For example, there are 75 occurrences of the expression strong voice in Europarl, all of which are used in the sense of political authority, for example, *ensuring that smaller* Member States retain a strong voice in the decision-making procedures. At the same time, out of the 28 occurrences of this expression in the British National Corpus [17] 19 examples refer to the quality of human voices, for example, She had a good, strong voice—an actor's *voice*. When translating *strong voice* into other languages, the political authority metaphor needs to be explicitly unpacked, because the literal translation is not suitable. For example, when translating strong voice into Russian, the political sense is likely to use reshitelno *vystupati* ('to express assertively') or *goryacho osuzhdat* ('to condemn vehemently') as opposed to the straightforward literal translation gromkij golos ('loud voice'). In the end, the mismatch between the domains and genres of parallel corpora and the target applications can lead to errors, which might be corrected by the use of less parallel resources.

1.1.2 Translationese in Parallel Data

Another kind of problems concerning the use of parallel data comes from a particular phenomenon known as translationese, namely a difference between features of translated texts and texts originally produced by native speakers [18]. Translationese is caused by factors inherent in the translation process, such as explicitation [19], i.e., the need to provide more information in the translated text for what remains implicit in the source text. For example, translations tend to use more cohesive markers, such as *therefore, however, nevertheless*, in comparison to original texts by making the logical relations more explicit in translation [18]. Other factors leading to translationese are related to the temporal and cognitive pressures on the translation operations, as the translators need to complete their tasks in a short period of time. This leads such phenomena as (1) normalisation, i.e., the tendendency to re-use stock expressions of the target language even when the source text deviates from the norm, and (2) "shining-through" [20], i.e., the influence of the syntactic and lexical choices stemming from the source texts even when other choices are common in monolingually produced texts in the target languages. In the end it has been shown that translationese effects have statistical significance leading to the ability to build fairly accurate classifiers detecting texts translated by humans [21], even in an unsupervised fashion [22].

From the viewpoint of the translation direction, there is always a source text which needs to be translated into other languages, so only one text in a parallel corpus is primary, while other texts exhibits features of translationese. At the same time, when parallel corpora are used for MT applications, this directionality of translation is usually ignored, so that the Slovenian \rightarrow English MT built from the Europarl corpus uses slightly unnatural Slovenian texts exhibiting features of translationese as its source texts. These aspects call for the greater use of texts originally produced in the respective languages.

1.2 Levels of Comparability

These constraints on the availability and the use of parallel data led to numerous studies utilising less parallel resources. This research direction generally goes under the name of 'comparable corpora'. However, it is important to note that there is no clear dividing line between fully parallel and comparable corpora, as multilingual resources vary with respect to the degree of linking between documents in the two languages. Consider some examples of documents along the cline of comparability:

translations identifiable originally produced source texts and their translations.

truly parallel	true high quality translations, such as the proceedings of the
	European Parliament.
modified parallel	translations with some modifications to cater for the target audience.
	For example, language-specific descriptions of the Search dialogue
	box in the OpenOffice manual are translations from English with
	necessary modifications, such as searching for New York is
	replaced with searching for Berlin in the German version.
adaptations	translators exhibit freedom in rendering the source text, as it is the
	case with many of the fan-produced subtitles as in the
	OpenSubtitles corpus [23].

en	The Adena culture was a Pre-Columbian Native American culture that existed from 800 BC to 1 AD, in a time known as the Early Woodland period
de	Adena-Kultur ist die Bezeichnung für eine im mittleren Ohio-Tal ansässige prähistorische Indianerkultur. Sie lässt sich für die Zeit von etwa 1000 v. Chr. bis 200 n. Chr. nachweisen
fr	La Civilisation Adena était une culture pré-colombienne amérindienne ayant existé de l'an 1 000 à l'an 200 avant JC., durant l'ère connue sous le nom de Période sylvicole
ja	アデナ文化(Adena)は、アメリカ合国オハイオ州を中心 に1000B.C.から元前後にえた文化。アデナ文化は、初期(ないし 前期)ウッドランド期(Early Woodlannd Period)の文化として 位置付けられ、アデナ文化の出によって、後のホプウェル文化を はじめとするウッドランド文化の先となるウッドランド式土器や 丘墓、トウモロコシ耕のすべてがでそろった。
ru	Культура Адена - доколумбовая индейская археологическая культура, существовавшая в период 1000-200 г. до н. э., в период, известный как ранний Будлендский период.

 Table 1.1
 Adena culture example

strongly comparable closely related texts produced in several languages.

Wikipedia entries entries on exactly the same topic are linked across languages via iWiki links, see Table 1.1, with individual entries varying in the amount of information.

news items	very specific events are covered by numerous news agencies in
	various languages. Often the same agency reports the same story
	in various languages without fully relying on their translation,
	see the BBC News in English and Spanish.

weakly comparable similar texts which cannot be directly linked to each other across languages, while still being in the same domain and genre, for example:

- texts in the same narrow subject domain and genre, but describing different events, e.g. parliament debates on health care from the German Bundestag, the British House of Commons and the French parliament;
- texts within the same broader domain and genre, but varying in subdomains and specific genres, e.g. parallel queries in the renewable energy domain mostly returning wind energy research articles in English vs solar panel producers in Russian [24].
- **unrelated** collections of unrelated texts, which are nevertheless collected using comparable methods from comparable sources. For instance, this concerns the use of random snapshots of the Web for Chinese, English, German and Russian [25] or the use of filtered Common Crawl data [26] with the assumption that different cultures use the Web for broadly similar purposes.