Segmentation Granularity in Dependency Representations for Korean

Jungyeul Park Department of Linguistics University of Arizona jungyeul@email.arizona.edu

Abstract

Previous work on Korean language processing has proposed different basic segmentation units. This paper explores different possible dependency representations for Korean using different levels of ent schemes for morphological segmentation of tokens into syntactic words. We provide a new Universal Dependencies (UD)-like corpus based on different levels of segmentation granularity for Korean. The corpus contains 67K words in 5,000 sentences which are split into training, development and evaluation data sets. We report parsing results using the new dependency corpus for Korean and compare them with the previous Korean UD corpus.

1 Dependency Parsing and the Korean Language

Language processing including morphological analysis for Korean has traditionally been based on the *eojeol*, which is a basic segmentation unit delimited by a blank in the sentence. Let us consider the sentence in (1), which contains ten eojoels (the corresponding morphological analysis is found in Figure 1). The number of eojoels is entirely based on the blank space character and the tenth eojeol in (1) also includes the punctuation mark. Almost all natural-language processing systems that have been previously developed for Korean have used the eojeol as a fundamental unit of analysis. As Korean is an agglutinative language, joining content and functional morphemes is very productive and they can be combined exponentially. For example, *yeoghal* ('role') is a content morpheme (a common noun) and -eul, a case marker ('ACC', accusative), is a functional morpheme.¹ They form together a single eojeol *yeoghal-eul* ('role + ACC'). A predicate *gangjo-ha-ass-da* ('focused') also consists of the content morpheme *gangjo-ha* ('focus') and its functional morphemes, *-ass* ('PAST', past tense) and *- da* ('IND', indicative), respectively.

In this paper, we analyze different levels of segmentation granularity in dependency representations for syntactic annotation (§2). We then propose a scheme to build a new Universal Dependencies (UD)-like corpus for Korean based on segmentation granularity (§3). UD has been developed cross-linguistically using a consistent treebank annotation scheme for many languages.² We provide 5,000 sentences based on each of the segmentation granularity possibilities described in this paper. We also present its UD parsing results, compare them with previously proposed UD for Korean (§4), and discuss future perspectives of dependency annotation and parsing for Korean (§5).

2 Segmentation Granularity for Korean

We define the following four different levels of segmentation granularity for Korean. These granularity levels have been independently proposed in previous work on Korean language processing as different basic segmentation units.

2.1 Eojeols

Most language processing systems and corpora developed for Korean have used the eojeol as a fundamental unit of analysis (Figure 2). For example, the Sejong corpus, the most widely-used corpus for Korean, uses the eojeol as the basic unit of analysis as presented in (1). Most morphological analysis systems have been developed based

¹For convenience sake, we add the hyphen-minus (-) at the beginning of functional morphemes, such as *-eul* to distinguish boundaries between content and functional morphemes. The accusative case marker *-eul* or *-leul* vary depending on the preceding character.

²http://universaldependencies.org

(1) 황석영을 비롯해 도서전에 참가한 한국 작가들도 이구동성으로 번역자의 역할을 강조했다.

hwangseogyeong-eul bilos-ha-a doseojeon-e chamga-ha-n jagga-deul-do hangug Hwang Seok-young-ACC including book exhibition-LOC participated Korean other authors-ALSO yeoghal-eul gangjo-ha-ass-da. igudongseong-eulo beonyeogja-ui with one voice translators-GEN role-ACC emphasize-PAST-IND-.

'Hwang Seok-young and other Korean authors who participated in the book exhibition emphasized the role of translators with one voice.'

1	황석영을	황석영/NNP+을/JKO	hwangseogyeong-eul
2	비롯해	비롯/XR+하/XSA+아/EC	bilos-ha-a
3	도서전에	도서/NNG+전/NNB+에/JKB	doseojeon-e
4	참가한	참가/NNG+하/XSV+ㄴ/ETM	chamga-ha-n
5	한국	한국/NNP	hangug
6	작가들도	작가/NNG+들/XSN+도/JX	jagga-deul-do
7	이구동성으로	이구동성/NNG+으로/JKB	igudongseong-eulo
8	번역자의	번역자/NNG+의/JKG	beonyeogja-ui
9	역할을	역할/NNG+을/JKO	yeoghal-eul
10	강조했다.	강조/NNG+하/XSV+았/EP+다/EF+./SF	gangjo-ha-ass-da.

Figure 1: Sejong corpus-style POS tagging example

on eojeols as input and can yield morphologically analyzed results, in which a single eojeol can contain several morphemes. The dependency parsing systems described in Oh and Cha (2010) and Park et al. (2013) use eojeols as an input token to represent dependency relationships between eojeols. Interestingly, Oh et al. (2011) presented a system of phrase-level syntactic label prediction for eojeols based on morpheme information. Petrov et al. (2012) proposed Universal POS tags for Korean based on the eojeol and Stratos et al. (2016) worked on POS tagging accordingly.

2.2 Separating words and punctuation

As eojeols have been used as a basic analysis unit in Korean corpora, the tokenization task is often ignored for Korean. However, there are corpora which use an English-like tokenization (Figure 3). Words in these corpora are already preprocessed: for example, the Penn Korean treebank (Han et al., 2002), in which punctuation marks are separated from words. Note that among existing corpora for Korean, only the Sejong treebank separates quotation marks from the word. Other Sejong corpora including the morphologically analyzed corpus do not separate the quotation marks. While the Korean Penn treebank separates all punctuation marks, quotation marks are the only symbols that are separated from words in the Sejong treebank. Chung and Gildea (2009) used this granularity of separating words and symbols for a baseline tokenization system for a machine translation system. Park et al. (2014) also used this granularity to develop Korean FrameNet lexicon units.

2.3 Separating case markers

The Sejong corpus has been criticized for the scope of the case marker, in which only a final noun (usually the lexical anchor) in the noun phrase is a modifier of the case marker. For example, Emmanuel Ungaro-ga in the Sejong corpus is annotated as (NP (NP Emmanuel) (NP Ungaroga)), in which only Ungaro is a modifier of -ga ('NOM'). The Korean Penn treebank does not explicitly represent this phenomenon. It just groups a noun phrase together: e.g. (NP Emmanuel Ungaroga). Collins' preprocessing for parsing the Penn treebank adds intermediate NP terminals for the noun phrase (Collins, 1997; Bikel, 2004), and NPs in the Korean Penn treebank will have a similar NP structure in the Sejong corpus (Chung et al., 2010). To fix the problem in the previous treebank annotation scheme, there are other annotation schemes proposed in the corpus and lexicalized parsing grammars for the purpose to correctly express the scope of the case marker (Figure 4).

Park (2006) considered case markers (or postpositions) as independent elements in Tree adjoining grammars (Joshi et al., 1975). Therefore, he defined case markers as an auxiliary tree to be ad-

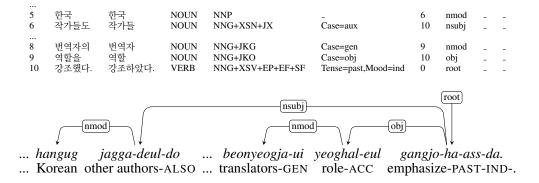


Figure 2: CoNLL-U format for eojeols: the basic elements of dependency relationships are based on eojeols delimited by a blank space character in the sentence. While the actual CoNLL-U data that we provide in this paper contain the results of morphological analysis (such as 73%/NNG+57/XSV+2%/EP+77/EF+./SF for line 10) to conserve the original structure of a combination of the word, for simplicity's sake we do not present them here.

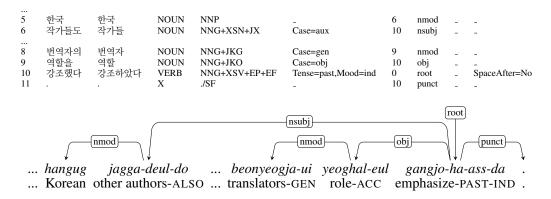


Figure 3: CoNLL-U format for English-like tokenization by separating punctuation marks: it separates the punctuation mark from the word *gangjo-ha-ass-da* ('focused') with punct. Otherwise, it still keeps the original structure of the eojeols.

joined to a noun phrase. For example, the single token *jagga-deul-do* becomes two tokens, *jaggadeul* ('author') and *-do* ('also'). However, verbal endings on the inflected forms of predicates are still in the eojoel and they are represented as initial trees for Korean TAG grammars. The lemma of the predicate and its verbal endings are dealt with as inflected forms instead of separate functional morphemes.

2.4 Separating verbal endings

Government and binding (GB) theory for Korean often proposed a syntactic analysis, in which the entire sentence depends on verbal endings. For example, *gangjo-ha-ass-da* becomes *gangjoha* ('emphasize'), *-ass* ('PAST'), and *-da* ('IND') as described in Figure 5.

The Kaist treebank (Choi et al., 1994), the first treebank created for Korean adapted this type of analysis (Figure 6). While the Kaist treebank separates case markers and verbal endings with their lexical morphemes, punctuation marks are not separated and they are still a part of the preceding token. Therefore, strictly speaking, this granularity level is not exactly same as in the Kaist treebank.

2.5 Discussion

The different levels of segmentation granularity described in this section have been proposed mainly because of different syntactic analysis in several previously proposed Korean treebank

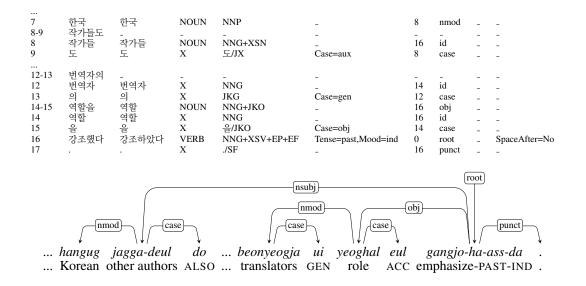


Figure 4: CoNLL-U format by separating case markers, which requires a dependency relationship between the noun phrase and the case marker (case), for example *yeoghal* ('role') and *eul* ('ACC').

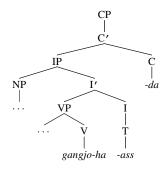


Figure 5: X-bar schema for *gangjo-ha*, *-ass*, and *-da* in Korean

datasets: Kaist (Choi et al., 1994), Sejong³, and Penn (Han et al., 2002) treebanks. Even for the segmentation granularity which we deal with, syntactic theory is implicitly presented in the corpus for Korean words. Granularity described in §2.1 and §2.3 is based on the Sejong treebank. Granularity described in §2.2 and §2.4 is based on the Korean Penn treebank and the Kaist treebank, respectively.

Many applications for Korean language processing are based on another level of segmentation granularity, in which all morphemes are separated: phrase-structure parsing (Choi et al., 2012; Park et al., 2016) and statistical machine translation (SMT) (Park et al., 2016), etc. Such morphemebased analysis for the word can be generated by a morphological analysis system, and most POS tagging systems such as Hong (2009) and Park et al. (2011) can produce all morpheme-based analysis. For example, *jagga-deul-do* ('authors-ALSO') is separated into *jagga* ('author'), *deul* ('PLUR'), and *do* ('ALSO'). However, we do not deal with this granularity to represent dependencies. It shows rather how words are formed, and it should include the fine-grained relationships between morphemes. This type of representation of words does not conform with the current dependency schemes for other languages and especially, neither with UD best practices.

3 UD for Korean

Since Universal Dependencies (UD) has been released (Nivre et al., 2016), several studies have been published, both theoretical (Schuster and Manning, 2016) and practical (Zeman et al., 2017). As for other morphologically rich languages, specific Universal Dependencies for Japanese were introduced relatively recently to meet the requirement of UD's cross-linguistically consistent treebank annotation (Tanaka et al., 2016). In the current UD, other morphologically rich languages such as Kazakh (Tyers and Washington, 2015) and Turkish (Sulubacak et al., 2016) are also available. In this section, we describe how to build UDs for Korean based on the different levels of segmentation granularity.

³https://www.sejong.or.kr

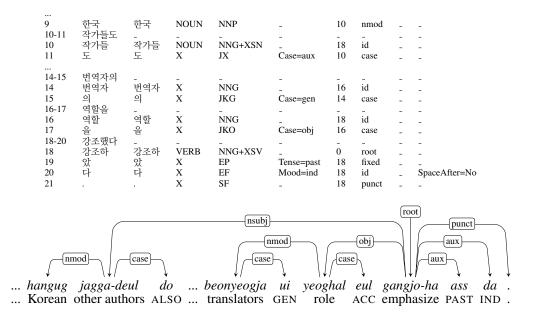


Figure 6: CoNLL-U format that separates verbal endings, which requires a dependency relationship between the verbal head and the verbal ending (aux), for example a verb *gangjoha* ('focus'), and two verbal endings *ass* ('PAST') and *da* ('IND') for tense and mood.

3.1 Universal POS

Using the eojeol and morpheme level mapping tables to Universal POS tags for the Sejong tagset proposed in Petrov et al. (2012) and Park et al. (2016), we can convert the single tags (morphemes) and the sequences of tags (eojeols and tokens) in the Sejong corpus into Universal POS tags. We also use additional mapping rules by using the approach to find Universal POS tags described in Oh et al. (2011) in which they predict phrase tags for the eojeol. In addition, the Sejong tags (morphemes) and the sequence of tags (tokens and eojeols) represented as immediate nonterminal nodes in the eventual parse tree can be used as a language-specific part-of-speech tag in the CoNLL-U format. Figure 7 shows example mapping rules for each segmentation granularity level. Tagsets in the Sejong corpus are mapped to the Universal POS tag sets either individually (NNP \rightarrow PROPN) or by a sequence of the POS tags (NNP+JKS \rightarrow PROPN). Figure 8 represents the 1-to-1 mapping from the POS tags in the Sejong corpus to Universal POS tags described in Park et al. (2016). These 1-to-1 mapping rules are used throughout segmentation granularity schemes described $\S2.1$ to $\S2.4$ if the eojeol is composed only by a single morpheme.

3.2 Universal features

Park (2006) detailed an approach to extract features from the Sejong treebank. Syntactic tags and morphological analysis allow us to extract syntactic features automatically and to develop universal features. For example, NP-SBJ syntactic tag is changed into NP and a syntactic feature Case=Nom is added. Syntactic tags which end with -sbj (subject), -obj (object) and -CMP (attribute), we extract Case features which describe argument structures in the sentence. Alongside Case features, we also extract Mood and Tense from the morphological analyses in the Sejong treebank. Since however morphological analyses for verbal and adjectival endings in the Sejong treebank are simply divided into ep (nonfinal endings), ef (final endings) and ec (conjunctive endings), Mode and Tense features can not be extracted directly. Park (2006) analyzed 7 non-final endings and 77 final endings used in the Sejong treebank to extract automatically Mood and Tense features. In general, ef carries Mood inflections, and ep carries Tense inflections. Conjunctive endings are not concerned with Mood and Tense features and we only extract ec features with their string value. We also add Hor for the honorific feature, which we can extract from lexical information of non-final endings

		참가한 <i>chamga-ha-n</i> ('participated')	작가들도 <i>jagga-deul-do</i> ('authors')	강조했다. <i>gangjo-ha-ass-da.</i> ('emphasize-PAST-IND')
eojeol	§2.1 (S) (U)	chamga-ha- n/NNG+XSV+ETM chamga-ha-n/VERB	jagga-deul- do/NNG+XSN+JX jagga-deul-do/NOUN	gangjo-ha-ass- da./NNG+XSV+EP+EF+SF gangjo-ha-ass-da./VERB
separating	§2.2 (S)	chamga-ha- n/NNG+XSV+ETM	jagga-deul- do/NNG+XSN+JX	gangjo-ha-ass- da/VV+EP+EF /SF
symbols	(U)	chamga-ha-n/VERB	jagga-deul-do/NOUN	gangjo-ha-ass-da/VERB ./PUNCT
separating	§2.3 (S)	chamga-ha- n/NNG+XSV+ETM	jagga-deul/NNG+XSN -do/JX	gangjo-ha-ass- da/VV+EP+EF ./SF
case markes	(U)	chamga-ha-n/VERB	jagga-deul/NOUN -do/ADP	gangjo-ha-ass-da/VERB ./PUNCT
separating verbal endings	§2.4 (S) (U)	chamga-ha/NNG+XSV -n/ETM chamga-ha/VERB -n/PRT	jagga-deul/NNG+XSN -do/JX jagga-deul/NOUN -do/ADP	gangjo-ha/VV -ass/EP -da/EF JSF gangjo-ha/VERB -ass/PRT -da/PRT JPUNCT

Figure 7: Example of the tag sequences at each granularity level. We show the examples for the converting mapping table between Sejong and Universal POS tag sets described in §3.1: *e.g.* NNG+XSV+ETM is converted into VERB in §2.1. (S) and (U) are for the Sejong and Universal POS tag sets.

such as -si.

3.3 Universal dependency representations

We use basic dependencies (core, non-core, noun dependents) for eojeols for segmentation granularity in §2.1. We add punct between word and punctuation marks (§2.2), and case between noun phrase and case markers (§2.3). We also employ fixed for verbal endings (§2.4). Initial dependency labels are based on phrase information in the Sejong treebank such as np-sub, np-obj, etc. We create conversion rules to conform to Universal Dependency relations.

nsubj (nominal subject) and csubj (clausal subject) can be assigned in which np-sbj occurs and nouns ended with either jks (nominative marker) or jx (topic marker). We distinguish nsubj and csubj as follows:

- if a subject noun is a derivational noun from the verb or the adjective, which are usually ended with etn+jks or etn+jx (where etn is a derivational morpheme for the noun), then csubj.
- otherwise, nsubj.
- (2) a. *unggaro-ga* ... *naseo-eoss-da* Ungaro-jx ... become-PAST-IND 'Ungaro became ...'

 b. unggaro-ga naseo-gi-ga ... sib-eoss-da Ungaro-NOM become-etn-jx ... easy-PAST-IND
 'Ungaro's becoming ... was easy'

While the previous UD for Korean uses nsubj:pass for the passive construction in Korean, we do not use it for the following two reasons: First, passive and causative verbs are often in the same form if they use passive or causative derivational morphemes such as *-i*, *-hi*, etc. and they are very ambiguous. Second, intransitive verbs are also allowed in the passive construction unlike in English.

obj (direct object) can be assigned in which np-obj occurs and nouns ended with jko (accusative marker). There are several cases where nouns can be ended with jx (topic marker). There are also some cases where nouns can be ended with jx (topic marker) for obj. iobj (second core dependent) can be assigned when np-alt (NP adjunct) occurs and nouns ended with jkb (auxiliary marker) such as *-ege*, *-e*, *-gge* (dative markers).

(3) ... sagoa-leul unggaro-ege ju-eoss-da
... apple-jko Ungaro-jkb give-PAST-IND
'... gave an apple to Ungaro'

ccomp (clausal complement) can be assigned when vp-cmp or vnp-cmp occurs. ccomp normally ends with ec and we identify 71 verbal

Sejong POS (S)	description	Universal POS (U)
NNG, NNB, NR, XR	Noun related	NOUN
NNP	Proper noun	PROPN
NP	Pronoun	PRON
MAG,	Adverb	ADV
MAJ	Conjunctive adverb	CCONJ
MM	Determiner	DET
VV, VX, VCN, VCP	Verb related	VERB
VA	Adjective	ADJ
EP, EF, EC, ETN, ETM	Verbal endings	PRT
JKS, JKC, JKG, JKO, JKB, JKV, JKQ, JX, JC	Postpositions (case markers)	ADP
XPN, XSN, XSA, XSV	Suffixes	PRT
SF, SP, SE, SO, SS	Punctuation marks	PUNCT
SW	Special characters	Х
SH, SL	Foreign characters	Х
SN	Number	NUM
NA, NF, NV	Unknown words	Х

Figure 8: POS tags in the Sejong corpus and their 1-to-1 mapping to Universal POS tags.

ending (among 410) in the Sejong treebank for ccomp. Otherwise, if vp or vnp occurs, and a phrase ends with ec, we consider it as a non-core dependent clause and assign advcl (adverbial clause modifier).

(4) ... unggaro-ga ... naseo-eoss-dago malha-eoss-da
 ... Ungaro-NOM ... become-PAST-ec tell-PAST-IND
 '... told that Ungaro became ...'

acl (adnominal clause) and amod (adjectival modifier) for Korean, in which vp-mod occurs, are defined as follows:

- if a verb ends with etm (verbal/adjectival ending for the relative clause) and it modifies a noun, we assign acl.
- if a adj ends with etm and it modifies a noun, we assign amod.

UD for Korean annotates acl:rel instead of acl to specify a relative clause for the verb ended with etm. ajt (adjunct) or nmod (nominal dependents) can be assigned where np-ajt or np occurs, respectively. det:poss is assigned for noun ended with jkg (genitive marker). Other UD relations such as advmod, det, etc can be assigned as a 1-to-1 mapping table by using Sejong POS labels as described in Table 1.

4 Experiments and Results

We collected sentences from news articles in one of Korean News websites published during 2016.⁴ We select the length of sentences in which there

Sejong POS	UD relations
mag	advmod
maj	CC
mm	det
sn	nummod

Table 1: Miscellaneous conversion between Sejong POS labels and UD representations

are words (eojeols) between 10 and 20 and the sentence should end with the final verbal ending such as -da (IND) or -gga (INT) and the punctuation mark such as *period* or *question mark* (sf).⁵ We perform initial automatic preprocessing tasks using existing tools for Korean such as POS tagging (Hong, 2009), assigning Universal POS labels (Petrov et al., 2012; Park et al., 2016), and MaltParser-based dependency parsing (Park et al., 2016). We manually correct the initial preprocessing tasks especially focused on dependency relation as described in §3.3.6 First, we build a corpus as described in $\S2.1$, then convert it into other levels of segmentation granularity as described in from $\S2.2$ to $\S2.4$. As a result, we provide a new UD for Korean which contain 5,000 sentences. We split them into 3K-1K-1K sentences for training, development, and evaluation data sets. Table 2 shows the brief statistics of the new UD for Korean. The number of words indicates the number of eojeols as described in $\S2.1$. We train and evaluate four different dependency segmen-

⁴http://hani.co.kr

⁵Similar criteria for selecting sentences are used for the Kaist and the Penn Korean treebank.

⁶Manual verification was done by two linguists in a month.

	sentences	words
training	3,000	40,648
dev	1,000	13,492
eval	1,000	13,623
total	5,000	67,763

Table 2: Statistics of the new UD for Korean. A number based on granularity $\S2.1$)

tation schemes based on segmentation granularity for Korean. Table 3 shows results produced by UDPipe (Straka et al., 2016). Upper granularity (towards granularity described in $\S2.4$) generally gives better results than lower granularity (towards $\S2.1$) because lexical items with functional morphemes in lower granularity can yield data sparseness. Bengoetxea and Gojenola (2010) presents a system that also changes segmentation granularity. They converted back the result of parsing to the original granularity to decide whether the new representation is effective for parsing. Additionally, the usual attachment score metrics used to evaluate dependency parsers are biased as described in Nivre and Fang (2017) for the cross-lingual setting. This bias can be equally applied to different segmentation granularity for Korean. We leave the evaluation as future work.

The current Universal Dependencies treebank for Korean used for the *CoNLL 2017 UD Shared Task* (Zeman et al., 2017)⁷ uses the same segmentation granularity as described in §2.2. We obtain 59.64% (UAS) and 51.05% (LAS) using the current version of UD for Korean (Nivre et al., 2017). While the current UD for Korean has a more sentences in the training data (4400 sentences vs. 3000), its results are comparable with the results by our corpus of §2.2 where we obtain 65.72% (UAS) and 48.44% (LAS).

5 Conclusion

The different levels of segmentation granularity described in this paper are mainly due to different representations of syntactic structure in the various Korean treebank datasets. They have used different word segmentation depending on their linguistic and computational requirements. While a certain segmentation granularity may be well suited for some linguistic phenomena or applications, it

	eojeol §2.1	symbol §2.2	case marker §2.3	verbal ending §2.4
UPOS	93.04	94.13	97.12	98.31
XPOS	82.59	85.22	90.63	95.19
UAS	62.08	65.72	76.19	79.59
LAS	40.51	48.44	71.29	78.07

Table 3: POS tagging and parsing results using UDPipe trained with four different UDs for Korean.

does not mean that this granularity is a better representation than the other in general. We need to find the most adequate segmentation granularity to adapt to our requirements for Korean language processing. The UDs corpus for Korean based on different levels of segmentation granularity will be publicly available.

Acknowledgement

We thank Francis Morton Tyers, Loïc Dugast, and the anonymous reviewers for their helpful comments and suggestions.

References

- [Bengoetxea and Gojenola2010] Kepa Bengoetxea and Koldo Gojenola. 2010. Application of Different Techniques to Dependency Parsing of Basque. In Proceedings of the NAACL HLT 2010 First Workshop on Statistical Parsing of Morphologically-Rich Languages, pages 31–39, Los Angeles, CA, USA. Association for Computational Linguistics.
- [Bikel2004] Daniel M. Bikel. 2004. Intricacies of Collins' Parsing Model. *Computational Linguistics*, 30(4):479–511.
- [Choi et al.1994] Key-Sun Choi, Young S Han, Young G Han, and Oh W Kwon. 1994. KAIST Tree Bank Project for Korean: Present and Future Development. In *Proceedings of the International Workshop on Sharable Natural Language Resources*, pages 7–14.
- [Choi et al.2012] DongHyun Choi, Jungyeul Park, and Key-Sun Choi. 2012. Korean Treebank Transformation for Parser Training. In Proceedings of the ACL 2012 Joint Workshop on Statistical Parsing and Semantic Processing of Morphologically Rich Languages, pages 78–88, Jeju, Republic of Korea. Association for Computational Linguistics.
- [Chung and Gildea2009] Tagyoung Chung and Daniel Gildea. 2009. Unsupervised Tokenization for Machine Translation. In *Proceedings of the 2009 Conference on Empirical Methods in Natural Language*

⁷http://hdl.handle.net/11234/1-1983

Processing, pages 718–726, Singapore. Association for Computational Linguistics.

- [Chung et al.2010] Tagyoung Chung, Matt Post, and Daniel Gildea. 2010. Factors Affecting the Accuracy of Korean Parsing. In Proceedings of the NAACL HLT 2010 First Workshop on Statistical Parsing of Morphologically-Rich Languages, pages 49–57, Los Angeles, CA, USA. Association for Computational Linguistics.
- [Collins1997] Michael Collins. 1997. Three Generative, Lexicalised Models for Statistical Parsing. In Proceedings of the 35th Annual Meeting of the Association for Computational Linguistics, pages 16–23, Madrid, Spain. Association for Computational Linguistics.
- [Han et al.2002] Chung-Hye Han, Na-Rae Han, Eon-Suk Ko, Heejong Yi, and Martha Palmer. 2002. Penn Korean Treebank: Development and Evaluation. In Proceedings of the 16th Pacific Asia Conference on Language, Information and Computation.
- [Hong2009] Jeen-Pyo Hong. 2009. Korean Part-Of-Speech Tagger using Eojeol Patterns. Master's thesis. Changwon National University.
- [Joshi et al.1975] Aravind K. Joshi, Leon S. Levy, and Masako Takahashi. 1975. Tree Adjunct Grammars. *Journal of Computer and System Sciences*, 10(1):136–163.
- [Nivre and Fang2017] Joakim Nivre and Chiao-Ting Fang. 2017. Universal Dependency Evaluation. In Proceedings of the NoDaLiDa 2017 Workshop on Universal Dependencies (UDW 2017), pages 86– 95, Gothenburg, Sweden. Association for Computational Linguistics.
- [Nivre et al.2016] Joakim Nivre, Marie-Catherine de Marneffe, Filip Ginter, Yoav Goldberg, Jan Hajic, Christopher D. Manning, Ryan McDonald, Slav Petrov, Sampo Pyysalo, Natalia Silveira, Reut Tsarfaty, and Daniel Zeman. 2016. Universal Dependencies v1: A Multilingual Treebank Collection. In Luis von Ahn, editor, *Proceedings of the Tenth International Conference on Language Resources and Evaluation (LREC 2016)*, Portorož, Slovenia. European Language Resources Association (ELRA).
- [Nivre et al.2017] Joakim Nivre, Željko Agić, Lars Ahrenberg, et al. 2017. Universal dependencies 2.0 – CoNLL 2017 shared task development and test data. LINDAT/CLARIN digital library at the Institute of Formal and Applied Linguistics, Charles University.
- [Oh and Cha2010] Jin-Young Oh and Jeong-Won Cha. 2010. High Speed Korean Dependency Analysis Using Cascaded Chunking. *Korean Simulation Journal*, 19(1):103–111.

- [Oh et al.2011] Jin-Young Oh, Yo-Sub Han, Jungyeul Park, and Jeong-Won Cha. 2011. Predicting Phrase-Level Tags Using Entropy Inspired Discriminative Models. In *International Conference on Information Science and Applications (ICISA) 2011*, pages 1–5.
- [Park et al.2011] Jungyeul Park, Jeong-Won Cha, and Seok Woo Jang. 2011. Korean POS Tagging using Noisy Channel Model with Syllable Lattice Based OOV Words Resolution. *Information - an international interdisciplinary journal*, 14(8):2835–2843.
- [Park et al.2013] Jungyeul Park, Daisuke Kawahara, Sadao Kurohashi, and Key-Sun Choi. 2013. Towards Fully Lexicalized Dependency Parsing for Korean. In Proceedings of The 13th International Conference on Parsing Technologies (IWPT 2013), Nara, Japan.
- [Park et al.2014] Jungyeul Park, Sejin Nam, Youngsik Kim, Younggyun Hahm, Dosam Hwang, and Key-Sun Choi. 2014. Frame-Semantic Web : a Case Study for Korean. In Proceedings of ISWC 2014 : International Semantic Web Conference 2014 (Posters and Demonstrations Track), pages 257– 260.
- [Park et al.2016] Jungyeul Park, Jeen-Pyo Hong, and Jeong-Won Cha. 2016. Korean Language Resources for Everyone. In Proceedings of the 30th Pacific Asia Conference on Language, Information and Computation (PACLIC 30), pages 49–58, Seoul, Korea.
- [Park2006] Jungyeul Park. 2006. Extraction automatique d'une grammaire d'arbres adjoints à partir d'un corpus arboré pour le coréen. Ph.D. thesis, Université Paris 7 - Denis Diderot.
- [Petrov et al.2012] Slav Petrov, Dipanjan Das, and Ryan McDonald. 2012. A Universal Part-of-Speech Tagset. In *Proceedings of the Eighth International Conference on Language Resources and Evaluation (LREC-2012)*, pages 2089–2096, Istanbul, Turkey. European Language Resources Association (ELRA).
- [Schuster and Manning2016] Sebastian Schuster and Christopher D. Manning. 2016. Enhanced English Universal Dependencies: An Improved Representation for Natural Language Understanding Tasks. In Proceedings of the Tenth International Conference on Language Resources and Evaluation (LREC 2016), Paris, France, 5. European Language Resources Association (ELRA).
- [Straka et al.2016] Milan Straka, Jan Hajic, and Jana Straková. 2016. UDPipe: Trainable Pipeline for Processing CoNLL-U Files Performing Tokenization, Morphological Analysis, POS Tagging and Parsing. In Proceedings of the Tenth International Conference on Language Resources and Evaluation (LREC 2016), Paris, France, 5. European Language Resources Association (ELRA).

- [Stratos et al.2016] Karl Stratos, Michael Collins, and Daniel Hsu. 2016. Unsupervised Part-Of-Speech Tagging with Anchor Hidden Markov Models. *Transactions of the Association for Computational Linguistics*, 4:245–257.
- [Sulubacak et al.2016] Umut Sulubacak, Memduh Gökırmak, Francis M. Tyers, Çağrı Çöltekin, Joakim Nivre, and Gülşen Eryiğit. 2016. Universal dependencies for Turkish. In *Proceedings of COLING 2016*.
- [Tanaka et al.2016] Takaaki Tanaka, Yusuke Miyao, Masayuki Asahara, Sumire Uematsu, Hiroshi Kanayama, Shinsuke Mori, and Yuji Matsumoto. 2016. Universal Dependencies for Japanese. In Proceedings of the Tenth International Conference on Language Resources and Evaluation (LREC 2016), Paris, France, 5. European Language Resources Association (ELRA).
- [Tyers and Washington2015] Francis Morton Tyers and Jonathan North Washington. 2015. Towards a Free/Open-source Universal-dependency Treebank for Kazakh. In *Proceedings of the 3rd International Conference on Turkic Languages Processing (Turk-Lang 2015)*, pages 276–289.
- [Zeman et al.2017] Daniel Zeman, Martin Popel, Milan Straka, Jan Hajič, Joakim Nivre, Filip Ginter, Juhani Luotolahti, Sampo Pyysalo, Slav Petrov, Martin Potthast, Francis Tyers, Elena Badmaeva, Memduh Gökırmak, Anna Nedoluzhko, Silvie Cinková, Jan Hajič jr., Jaroslava Hlaváčová, Václava Kettnerová, Zdeňka Urešová, Jenna Kanerva, Stina Ojala, Anna Missilä, Christopher Manning, Sebastian Schuster, Siva Reddy, Dima Taji, Nizar Habash, Herman Leung, Marie-Catherine de Marneffe, Manuela Sanguinetti, Maria Simi, Hiroshi Kanayama, Valeria de Paiva, Kira Droganova, Hěctor Martínez Alonso, Hans Uszkoreit, Vivien Macketanz, Aljoscha Burchardt, Kim Harris, Katrin Marheinecke, Georg Rehm, Tolga Kayadelen, Mohammed Attia, Ali Elkahky, Zhuoran Yu, Emily Pitler, Saran Lertpradit, Michael Mandl, Jesse Kirchner, Hector Fernandez Alcalde, Jana Strnadova, Esha Banerjee, Ruli Manurung, Antonio Stella, Atsuko Shimada, Sookyoung Kwak, Gustavo Mendonça, Tatiana Lando, Rattima Nitisaroj, and Josie Li. 2017. CoNLL 2017 Shared Task: Multilingual Parsing from Raw Text to Universal Dependencies. In Proceedings of the CoNLL 2017 Shared Task: Multilingual Parsing from Raw Text to Universal Dependencies. Association for Computational Linguistics.