SentiAspectExtractor Introduction

SentiAspectExtractor is an independent project that can extract aspects from given text and a list of text opinions using several syntax-based rules. In this introduction, we will provide detailed descriptions of its algorithmic workflow and how to use it.

• Algorithmic Workflow

The algorithm requires text and its corresponding list of opinions as input. The aspect extraction algorithm can be divided into five steps: 1) preprocessing text, 2) selecting representative nodes for input opinion expressions, 3) extracting aspect for representative opinion node based on a set of syntactic rules, 4) extending aspect, and 5) trimming aspects. The general flowchart is shown in Figure 1. In the following, we will introduce each step.

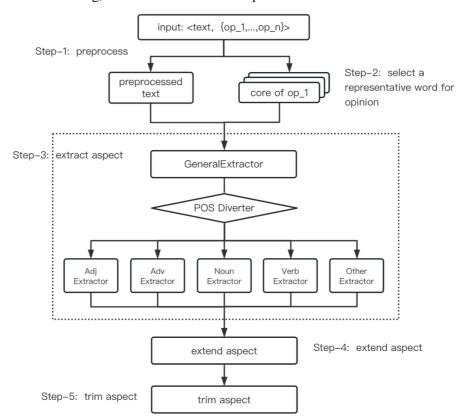


Figure 1 The general flowchart of the aspect extraction algorithm

1) Preprocessing Text

In the algorithm, we will utilize CoreNLP [1] for part-of-speech(POS) tagging, dependency parsing, and constituency parsing of the input text. Some incorrect text writing styles may affect the parsing results of CoreNLP. Therefore, in this step, we will correct these writing styles. Figure 2 provides an example illustrating how different writing styles of text can affect the analysis results

of CoreNLP. On the right-hand side of the figure, there are common or correct writing styles, along with their corresponding dependency parse graphs. On the left-hand side, there are incorrect segmentation styles and their dependency parse graphs. The dependency parse graphs in this figure are extracted from the running examples on the official website of CoreNLP.

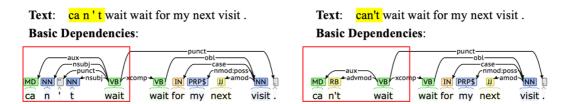


Figure 2 An example illustrating how different writing styles of text can affect the analysis results of CoreNLP.

2) Selecting Representative Nodes For Input Opinion Expressions

Considering that there may be multi-word expressions for opinions, in this step, we will select a representative node for multi-word expressions, and proceed to the next step for aspect extraction. If the last word in a multi-word opinion expression has a POS tag of adjective, verb, or noun, we directly select this last word as the representative node. If the above condition is not met, we use the importance of a node in the dependency graph to select the representative node. Specifically, we calculate the sum of the in-degree and out-degree of each node in the multi-word expression in the dependency graph. The larger the sum, the more connected the node is to other nodes in the graph, and the more important it is. We select the node with the highest sum of in-degree and out-degree as the representative node.

3) Extracting aspects based on a set of syntactic rules

In this step, we will extract aspect for representative opinion node based on a set of syntactic rules. These rules are constructed based on the dependency graph of CoreNLP. A pair of dependency relationships in the graph can be described as " $Head \xrightarrow{reln} Dep$ ". The starting node of the arrow (Head) is a core node (also known as dominant words) in a relationship. The ending node of the arrow (Dep) is a dependent node. " reln " is used to describe the dependency relationship between the two. For example, in the phrase "my next visit", there is a relationship " $visit \xrightarrow{amod} next$ ", which can be described as "next" being a adjectival modifier (amod) for "visit".

The extraction rules can be divided into two parts: general extraction rules, which are applicable to all representative opinion nodes; and customized extraction rules, which will extract specifically based on the different POSs of the representative opinion nodes. Table 1 summarizes

all the rules proposed in this section. Rules are mutually parallel and do not affect each other.

Table 1 Extraction Rule Summary

_		raction Rule:				
	redic	otivo Dilloci				
R		Predicative Rules:				
	_1	If $BE \stackrel{cop}{\longleftarrow} govOfBE \xrightarrow{subj} subj$, opinion \longleftarrow subj , $govOfBE$ [Noun];				
	17-1	Extract govOfBE				
		If $BE \xrightarrow{ccomp} ccomp \xrightarrow{subj} subjOfcomp$,				
R	R-2	$BE \xrightarrow{subj} subj$, opinion \leftarrow subj , subj $Ofcomp$ [LegalTag];				
		Extract subj0fcomp				
Subject Rules: $opinion - FindSubject \rightarrow subj0f0P$						
R-3		If subj0f0P [LegalTag]; Extract subj0f0P				
R	-4	If, $subjOfOP[Adj] \xrightarrow{nmod} modOfSubj[LegalTag]$; Extract $modOfSubj$				
R	-5	If, $subjOfOP[Verb] \xrightarrow{obj/obl} objOfSubj[LegalTag]$; Extract $objOfSubj$				
R	-6	If $subjOfOP[Verb] \xrightarrow{obj/obl} NULL$; Extract $subjOfOP$				
	*	If subj0f0P { person }; Skip out				
Subordinate Structure Rules:						
	R-7	If modedByParens[LegalTag] (··· opinion ···); Extract modedByParens				
	R-8					
Г	X-0	$\begin{array}{c} \text{If opinion} \leftarrow \textit{acl}, \textit{acl} \xleftarrow{\textit{acl}} \textit{modedByACL[LegalTag]}; \text{Extract modedByACL} \\ \\ \text{acl} \end{array}$				
١.		If opinion \leftarrow acl, acl \leftarrow modedByACL,				
1	R-9	$modedByACL \xrightarrow{subj} subj[LegalTag], modedByACL \xrightarrow{cop} BE;$				
		Extract subj				
stomized Extraction Rules:						
A	Adjective Extraction Rules:					
		(The modified object of the opinions)				
R	-10	If $adjOP \stackrel{*mod}{\longleftarrow} modedByAdj[Noun]$; Extract $modedByAdj$				
_	11	$If \ adjOP \stackrel{*mod}{\longleftarrow} modedByAdj[Noun] \stackrel{nmod:at/with/in}{\longleftarrow} indirectlyModed[Noun] ;$				
K	R-11	Extract indirectlyModed				
		If $adjOP \stackrel{*mod}{\longleftarrow} modedByAdj[Adj] \stackrel{*mod}{\longleftarrow} indirectlyModed[Noun];$				
R	R-12	Extract indirectlyModed				
		(The object of the opinions)				
		If $adjOP \xleftarrow{advcl} verb\{find, make\} \xrightarrow{obj/obl} objOfAdj[LegalTag];$				
R	-13	Extract obj0fAdj				
		If $opinion - FindSubject \rightarrow subjOfOP\{NULL, person \}$				
Ь	-14					
K	-14	$adjOP \xrightarrow{obj/obl} objOfAdj[LegalTag];$				
		Extract obj0fAdj				
		(The complements of the opinions)				
R	R-15	If $adjOP \xrightarrow{xcomp} compNode[Verb]$;				
		Extract compNode				
	R-16	If $adjOP - FindSubject \rightarrow subjOfOP\{NULL, person \}$				
		$adjOP \xrightarrow{advcl/ccomp} clNode \xrightarrow{subj} subjOfCL[LegalTag];$				
R	-10					
R	-10	Extract subj0fCL				
	-16	Extract $subjOfCL$ If $adjOP - FindSubject \rightarrow subjOfOP\{NULL, person \}$				

	Extract obj0fComp			
	If $adjOP - FindSubject \rightarrow subjOfOP\{NULL, person \}$			
R-18	$adjOP \xrightarrow{xcomp} compNode[Verb] \xrightarrow{ccomp} clNode \xrightarrow{subj} subjOfCL[LegalTag];$			
	Extract subj0fCL			
Verb I	Extraction Rules:			
(The modified object of the opinions)				
R-19	If $verbOP \stackrel{*mod}{\longleftarrow} modedByVerb[Noun]$; Extract $modedByVerb$			
(The object of the opinions)				
	If $verbOP - FindSubject \rightarrow subjOfOP\{NULL, person \},$			
R-20	$verbOP \xrightarrow{obj/obl} objOfVerb[LegalTag];$			
	Extract obj0fVerb			
	(The complements of the opinions)			
	If $verbOP - FindSubject \rightarrow subjOfOP\{NULL, person \},$			
R-21	$verbOP \xrightarrow{ccomp} clNode \xrightarrow{subj} subjOfCL[LegalTag];$			
	Extract subj0fCL			
Noun Extraction Rules:				
(The modified object of the opinions)				
R-22	If $nounOP \stackrel{compound}{\longleftarrow} cpNode[Noun]$; Extract $cpNode$			
R-23	If $nounOP \xrightarrow{compound} cpNode[Noun]$; Extract $cpNode$			
R-24	$\begin{array}{c} \text{If } nounOP \xrightarrow{nmod} modedByNoun[LegalTag]; Extract modedByNoun \end{array}$			
	(The object of the opinions)			
	If $nounOP \stackrel{obj/obl}{\longleftarrow} verb\{find, make\} \stackrel{obj/obl}{\longrightarrow} objOfNoun[LegalTag];$			
R-25	Extract obj0fNoun			
R-26	If $nounOP \xrightarrow{obj/obl} objOfNoun[LegalTag]$; Extract $objOfNoun$			
Adverb Extraction Rules:				
	(The modified object of the opinions)			
R-27	If $advOP \stackrel{*mod/dep}{\longleftarrow} modedByAdv[Noun]$; Extract $modedByAdv$			
R-28	If $advOP \stackrel{*mod}{\longleftarrow} modedByAdv[Verb]$; Extract $modedByAdv$			
	If $advOP \stackrel{*mod}{\leftarrow} modedByAdj[AdjAdv/Verb] \stackrel{*mod}{\leftarrow} indirectlyModed[Noun]$;			
R-29	Extract indirectlyModed			
	(The object of the opinions)			
	If $advOP \xrightarrow{*mod} verb\{find, make\} \xrightarrow{obj/obl} objOfVerb[LegalTag]$;			
R-30	Extract obj0fVerb			
	If $advOP - FindSubject \rightarrow subjOfOP\{NULL, person \}$,			
R-31	$advOP \stackrel{*mod}{\longleftarrow} modedByAdv \stackrel{obj/obl}{\longrightarrow} objOfAdv[LegalTag] ;$			
	Extract obj0fVerb			
Other	Extraction Rules:			
(The modified object of the opinions)				
	(The modified object of the opinions)			
R-32	If other $OP \leftarrow govOfOP[LegalTag]$; Extract $govOfOP$			

General Extraction Rule: it consists of three sub-rules: predicative rules, subject rules, and subordinate structure rules.

Predicative Rules: We will use the BE-verb as a clue to traverse the subject-link verb-predicative structure in the sentence. If the opinion is part of the subject, we take the predicative as the aspect. There are two cases for the predicative: if it is a noun phrase, we take the noun phrase as the aspect; if it is a clause, we take the subject of the clause as the aspect. The dependency relationship between these two cases is also different. When the predicate is a phrase, it is $BE \stackrel{cop}{\longleftarrow} govOfBE \stackrel{subj}{\longrightarrow} subj$, if $opinion \longleftarrow subj$ and govOfBE is a noun, we will select govOfBE as an aspect. Here, we use " $Dep \longleftarrow Head$ " to describe that Dep is directly or indirectly dependent on Head (manifested as a one-way path from Head to the Dep in the graph), or Dep is Head itself. The example can be seen in Figure 3(R-1). When the predicate is a clause, it is $BE \stackrel{subj}{\longrightarrow} subj$, $BE \stackrel{ccomp}{\longrightarrow} ccomp$ $\stackrel{subj}{\longrightarrow} subjOfcomp$, and if $opinion \longleftarrow subj$ and the POS of subjOfcomp belongs to noun(NN*), pronoun(PRP*), number(CD), or determiner (DT) (which are hereinafter referred to as LegalTag), we will select subjOfcomp as an aspect. The example can be seen in Figure 3(R-2).

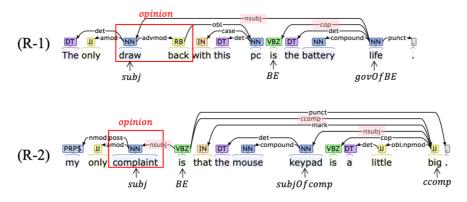


Figure 3 Examples of the predicative rules.

Subject Rules: We will extract the subject (subj0f0P) of the opinion as the aspect. Different treatments will be applied based on the different cases of the subject. If the POS of subj0f0P belongs to LegalTag, the subj0f0P will be extracted as an aspect. (Figure 4(R-3)) If subj0f0P is an adjective (JJ^*) , its modified object mod0fSubj ($subj0f0P \xrightarrow{nmod} mod0fSubj$) will be extracted as an aspect. (Figure 4(R-4)). If subj0f0P is a verb (VB^*) and has an object obj0fSubj ($subj0f0P \xrightarrow{obj/obl} obj0fSubj$) whose POS belongs to LegalTag, the obj0fSubj will be extracted as an aspect. (Figure 4(R-5)). If subj0f0P is a verb and doesn't have objects ($subj0f0P \xrightarrow{obj/obl} NULL$), the subj0f0P will be extracted as an aspect. (Figure 4(R-6)). Noted that if we find subj0f0P is a person during the above processes, we will skip out. Because we cannot

distinguish between "He is good" and "He is very satisfied". The subject extraction algorithm is shown in Figure 5

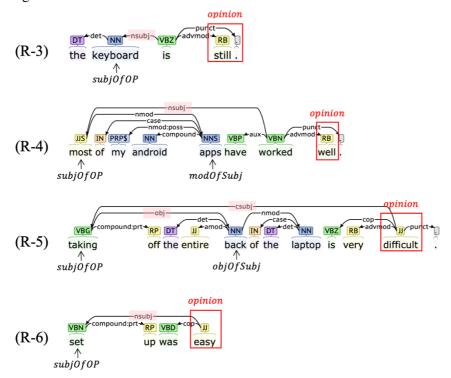


Figure 4 Examples of the subject rules.

```
Algorithm 1 Find Subject
Input: Node in the dependency graph node, the dependency graph graph
Output: The subject of node
    nodeListToRoot = graph.getPathToRoot(node); // get the path from the current node to the root
1:
2:
    for i from 0 to |nodeListToRoot| do:
3:
       nodeCurr = nodeListToRoot[i] ; // the current node
4:
       nodeGov = nodeListToRoot[i+1] \; ; \textit{// the head/parent node of current node}
5:
       reln = graph.getRelation(nodeGov, nodeCurr); // get the relationship between the two
      // If the current node is part of the subject (connected to nodeGov with a subject relationship), then exit the check directly
6:
      if isSubj(reln) then:
7:
         return Null;
8:
9:
        nodeCurr_ChildSet = graph.getChildren(nodeCurr);
10:
       for child in nodeCurr ChildSet do:
11:
12:
         reln = graph.getRelation(nodeCurr, child);
         if \ \ is Subj(reln) \ then:
13:
14:
           return child;
15:
         end if
16:
       end for
17:
     end for
```

Figure 5 The subject extraction algorithm

Subordinate Structure Rules : If the opinion is in a subordinate structure, the object modified by the subordinate structure will be considered as the aspect. Here, we consider two types of

subordinate structures, namely parentheses clauses and adnominal clauses. For parentheses clauses, if we detect a structure like "modedByParens ($\cdots opinion \cdots$)", and the POS of modedByParens belongs to LegalTag, we extract modedByParens as the aspect (Figure 6(R-7)). For adnominal clauses (acl), if the opinion is part of an adnominal clause ($opinion \leftarrow acl$), and there is a LegalTag modedByACL modified by acl ($modedByACL \xrightarrow{acl} acl$), we extract modedByACL as the aspect (Figure 6(R-8)). Combining with the subject rules, if the node being modified by the acl is also a predicate in a subject-copula-predicate relationship ($modedByACL \xrightarrow{subj} subj$, $modedByACL \xrightarrow{cop} cop$), and the POS of subj belongs to LegalTag, we will extract subj as the aspect (Figure 6(R-9)).

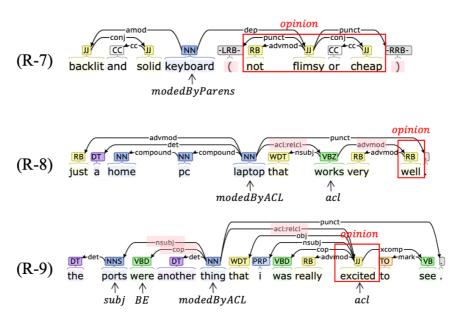


Figure 6 Examples of the subordinate structure rules.

Customized Extraction Rules: Nodes with different POS in the dependency graph are positioned in different dependency relationships. Therefore, we have custom rules for four important POSs (i.e. adjectives, verbs, nouns, adverbs) to search aspects. These rules are roughly divided into the extraction rules for the modified object of the opinions, the extraction rules for the object of the opinions, and the extraction rules for the complements of the opinions.

Adjective Extraction Rules: In order to extract the modified objects of adjective opinions, we have set the following rules: 1) $adjOP \stackrel{*mod}{\longleftarrow} modedByAdj[Noun]$, if the modified object (modedByAdj) of the adjective opinion is a noun, then it will be extracted as an aspect (Figure 7(R-10)); 2) $adjOP \stackrel{*mod}{\longleftarrow} modedByAdj[Noun] \stackrel{nmod:at/with/in}{\longleftarrow} indirectlyModed[Noun]$, If the POS of

modedByAdj is noun, it has a noun dependent (indirectlyModed[Noun]), the relationship between them is a compound noun modifier (nmod), and the preposition between the two is "at", "with", or "in", then indirectlyModed will be extracted as an aspect (Figure 7(R-11)); 3) adjOP $\stackrel{*mod}{\longleftarrow} modedByAdj[Adj] \stackrel{*mod}{\longleftarrow} indirectlyModed[Noun]$, If the POS of modedByAdj is adjective, it modifies a noun ($modedByAdj[Adj] \stackrel{*mod}{\longleftarrow} indirectlyModed[Noun]$), then indirectlyModed will be extracted as an aspect (Figure 7(R-12)).

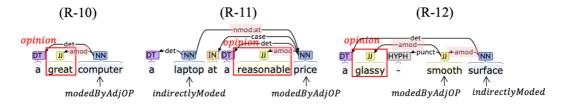


Figure 7 Examples of extracting the modified objects of adjective opinions

In order to extract the **objects of adjective opinions**, we have set the following rules: 1) $adjOP \stackrel{advel}{\longleftarrow} verb\{find, make\} \xrightarrow{obj/obl} objOfAdj[LegalTag]$, in the structure like 'make/find sth. adj.', we will extract 'sth.' as the aspect (Figure 8(R-13)) ; 2) $adjOP - FindSubject \rightarrow subjOfOP\{NULL, |person|\}, adjOP \xrightarrow{obj/obl} objOfAdj[LegalTag]$, if the opinion does not have a subject or the subject is a person, we will extract its object whose POS belongs to LegalTag as the aspect (Figure 8(R-14)).

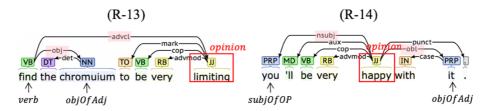


Figure 8 Examples of extracting the objects of adjective opinions

In order to extract the **complements of adjective opinions**, we have set the following rules: 1) $adjOP \xrightarrow{xcomp} compNode[Verb]$, if the open clausal completion node (xcompNode) of adjOP is a verb, we will extract xcompNode as the aspect (Figure 9(R-15)); 2) $adjOP - FindSubject \rightarrow subjOfOP\{NULL, |person|\}, adjOP \xrightarrow{advcl/ccomp} clNode \xrightarrow{subj} subjOfCL[LegalTag]$, if adjOP has no subject or has a person subject, we will extract the subject of its complement clause as aspect (Figure 9(R-16)); 3) $adjOP - FindSubject \rightarrow subjOfOP\{NULL, |person|\}, adjOP \xrightarrow{xcomp} compNode[Verb] \xrightarrow{obj/obl} objOfComp[LegalTag]$, if adjOP has no subject or has a person subject, we will extract the object of its complement verb as aspect (Figure 9(R-17)); 4)

 $adjOP - FindSubject \rightarrow subjOfOP\{NULL, |person|\}, adjOP \xrightarrow{xcomp} compNode[Verb] \xrightarrow{ccomp} clNode \xrightarrow{subj} subjOfCL[LegalTag]$, if adjOP has no subject or has a person subject, we will extract the subject of its complement verb clause as aspect (Figure 9(R-18)).

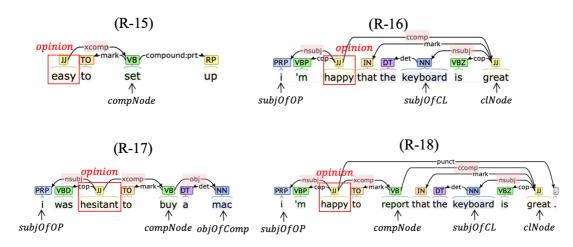


Figure 9 Examples of extracting the complements of adjective opinions

Verb Extraction Rules: In order to extract the **modified objects of verb opinions**, we have: $verbOP \stackrel{*mod}{\longleftarrow} modedByVerb[Noun]$ (Figure 10(R-19)). In order to extract the **objects of verb opinions**, we have: $verbOP - FindSubject \rightarrow subjOfOP\{NULL, | person|\}$, $verbOP \stackrel{obj/obl}{\longrightarrow} objOfVerb[LegalTag]$ (Figure 10(R-20)). In order to extract the **complements of verb opinions**, we have: $verbOP - FindSubject \rightarrow subjOfOP\{NULL, | person|\}$, $verbOP \stackrel{ccomp}{\longrightarrow} clNode \stackrel{subj}{\longrightarrow} subjOfCL[LegalTag]$ (Figure 10(R-21)).

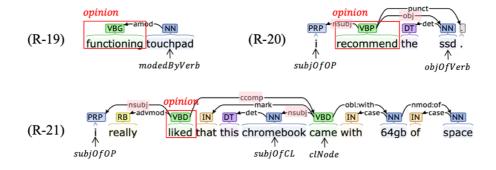


Figure 10 Examples of verb extraction rules

Noun Extraction Rules: In order to extract the modified objects of noun opinions, we have: 1) $nounOP \stackrel{compound}{\longleftarrow} cpNode[Noun]$ (Figure 11(R-22)); 2) $nounOP \stackrel{compound}{\longrightarrow} cpNode[Noun]$ (Figure 11(R-23)); 3) $nounOP \stackrel{nmod}{\longrightarrow} modedByNoun[LegalTag]$ (Figure 11(R-24)). In order to extract the objects of noun opinions, we have: 1) $nounOP \stackrel{obj/obl}{\longleftarrow} verb\{find, make\} \stackrel{obj/obl}{\longrightarrow} objOfNoun[LegalTag]$ (Figure 11(R-25)); 2)

 $nounOP \xrightarrow{obj/obl} objOfNoun[LegalTag]$ (Figure 11(R-26)).

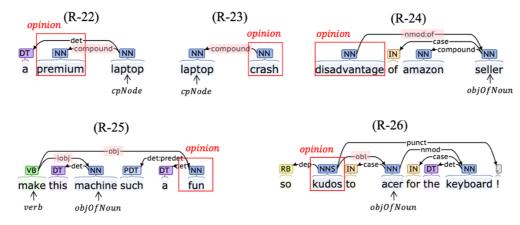


Figure 11 Examples of noun extraction rules

Adverb Extraction Rules: In order to extract the modified objects of noun opinions, we have: 1)

 $advOP \stackrel{*mod/dep}{\longleftarrow} modedByAdv[Noun]$ (Figure 12(R-27)); 2) $advOP \stackrel{*mod}{\longleftarrow} modedByAdv[Verb]$ (Figure 12(R-28)); 3) $advOP \stackrel{*mod}{\longleftarrow} modedByAdj[AdjAdv/Verb] \stackrel{*mod}{\longleftarrow} indirectlyModed[Noun]$ (Figure 12(R-29)). In order to extract the **objects of noun opinions**, we have: 1) $advOP \stackrel{*mod}{\longleftarrow} verb\{find, make\} \stackrel{obj/obl}{\longrightarrow} objOfVerb[LegalTag]$ (Figure 12(R-30)); 2) $advOP - FindSubject \rightarrow subjOfOP\{NULL, |person|\},$ $advOP \stackrel{*mod}{\longleftarrow} modedByAdv \stackrel{obj/obl}{\longrightarrow} objOfAdv[LegalTag]$ (Figure 12(R-31)).

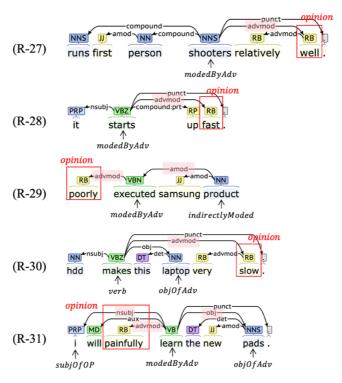


Figure 12 Examples of adverb extraction rules

Other Extraction Rules: For opinions that do not belong to the above POSs, we will check

 $otherOP \leftarrow govOfOP[LegalTag]$ (Figure 13(R-32)).

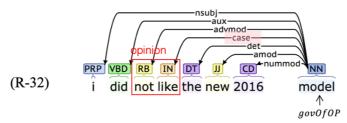


Figure 13 Examples of other extraction rules

4) Extending aspect from the core node

According to step three, we can only find one node, but in real-world data, there are often multi-word aspects, such as 'backlit keyboard'. To address this issue, we have set up several aspect extension rules (AER). **AER-1**: If the core word is a verb, we will expand based on the "compound:prt" relation (Figure 14(AER-1)). **AER-2**: If the core word is a pronoun, pronoun possessive, or any determiner in 'this',' that ',' that ', and' these ', we will use CoreNLP for coreference resolution (Figure 14(AER-2)). For the vast majority of aspects with noun POS, we have set three granularity extension rules. **AER-3.1**: Only use 'compound' relationship to extend (Figure 14(AER-3.1)). **AER-3.2**: On the basis of AER-3.1, further extend through the " *mnod* " relationship (Figure 14(AER-3.2)). **AER-3.3**: Extension using constituency parsing. Starting from the leaf node of the core node in the constituency tree, the range of the aspect is determined based on the last noun constituents (Figure 14(AER-3.3)). The granularity of AER-3.1 to AER-3.3 extension is gradually relaxed. AER-3.1 used by default in the algorithm.

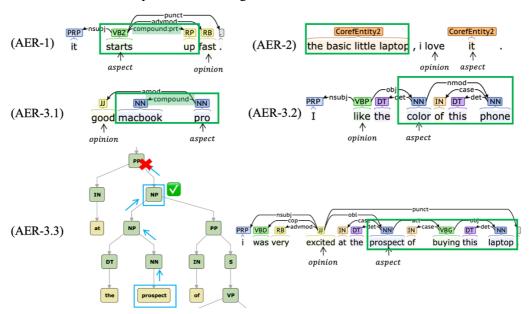


Figure 14 Examples of aspect extension rules. The "aspect" represents the aspect that was initially found through the rules in step three. The green box indicates the extended aspect.

5) Trimming aspect

In this step, we will trim the aspects generated during the above steps. We have several trimming rules:

TR-1: When extending, nodes with POS such as "PRP", "DT", and "IN" will not be included in the list representing multi-word aspects, even if they are included in the scope according to the extension rules. However, in general, the scope of the aspect is determined by the first and last nodes that will not be trimmed. Therefore, if the above POS appear between these two nodes, they may still exist in the final output aspect. Taking Figure 14 AER3.3 as an example, the initial range found is "the prospect of buying this laptop". Based on the POS, "the", "of", and "this" will be trimmed, but the first and last nodes that are not trimmed are "prospect" and "laptop". Therefore, the final aspect is "prospect of buying this laptop". This trimming rules are mainly for AER-3.3 to prevent the scope it extends from being too large.

TR-2 Words in English that have lower meanings such as "thing" and "feeling" are often referred to as "generic" or "vague" words. We believe that such words cannot be regarded as explicit aspects, so before expanding the core, we check whether the core node is a vague word. If it is a word connect with vague and does not anv non-vague words $\xrightarrow{nmod/obj/obl/compound/appos} non-vagueNode$), we won't extend it, and the output "[-1, (vaaueNode -1]" indicates that we have found an implicit aspect.

TR-3: Some words, such as 'quality' and 'time', while not vague words, we believe that if the extended aspect only contains one of them, it still cannot be considered an explicit aspect, so we will also output '[-1,-1]' in this case. In the algorithm, we use implicitAspectWordList.txt to store these words.

TR-4: Several rules in Step 3 are designed to extract verb aspects, but we are not interested in all actions. We have set up lists of verb aspects, and only the verb aspects that exist in these lists will be kept in the final output result.

If no aspect is left after the above trimming, the output will also be [-1,-1].

• Instruction For Use

This algorithm is published in the form of a jar package. Below, we will introduce some parameter descriptions when using jar packages:

Parameter	Parameter Description
-help	Output parameter description.
-inputFile	Set the input file address.
-outputFile	Set the output file path. If not set, the analysis result will be output in the same directory as the input file.
-dict	Set the address of the dictionary to be used for analysis.
-explain	Set whether to output the reason for extracting the aspect when outputting the result.
-coreExtendType	Set the expansion method for noun aspects. 0 corresponds to AER-3.1, 1 corresponds to AER-3.2, 2 corresponds to AER-3.3. The default parameter is 0.
-outputFormat	Set the output format of the result 0 represents the text-based output format, such as '[4,5]:[0,1], [5,6];' 1 represents the JSON output format, such as '{"opinion": [1,2], "aspect": [3,4]}' The default parameter is 1.
-implicitOpinionDealType	Set the handling method for illegal or implicit opinions. 0: Output 'cannot_deal' 1: Output [-1, -1] as an aspect 2: Do not output this opinion The default parameter is 1.
-isTextPreprocessing	Set whether to perform text preprocessing. The default parameter is true.

The words in the vagueWordList.txt and implicitAspectWordList.txt files will not be output as explicit aspects, but in some fields, there may be situations where the words in the two files are used as explicit aspects. To evaluate this risk, the "/dataProcess/ListRiskChecker.java" in the jar package can be used to calculate the number of times the words in the file appear in the marked explicit aspects. (This requires providing a certain amount of labeled data.)

• Reference

[1] https://corenlp.com