

# Relational Graph Attention Network for Aspect based Sentiment Analysis

-- Stargazers

# Aspect Based Sentiment Analysis

- Aspect Based Sentiment Analysis aims to determine the sentiment polarities towards one or more aspects appearing in a single sentence.
- An example of this task is, given a review "*great food but the service was dreadful*", to determine the polarities towards the aspects food and service.
- Since the two aspects express quite opposite sentiments, just assigning a sentence-level sentiment polarity is inappropriate.
- In this regard, ABSA can provide better insights into user reviews compared with sentence-level sentiment analysis.

# Objective of the Work

- obtain the dependency tree of a sentence using an ordinary parser
- reshape the dependency tree to root it at a target aspect and pruning of the tree is performed to retain only edges with direct dependency relations with the aspect
- encode the new dependency trees with relational graph attention network (R-GAT) model

# Related Work

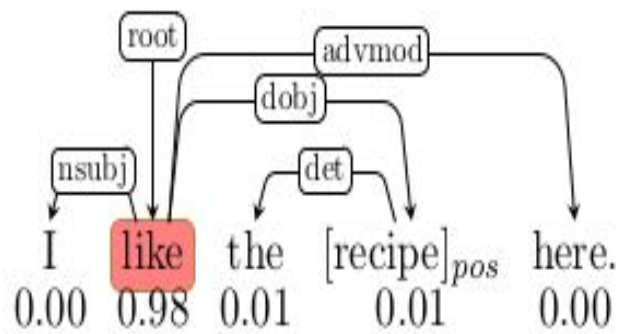
- early attempts rely on handcrafted syntactic rules (Qiu et al., 2011; Liu et al., 2013), though they are subject to the quantity and quality of the rules.
- Dependency-based parse trees are then used to provide more comprehensive syntactic information. For this purpose, a whole dependency tree can be encoded from leaves to root by a recursive neural network (RNN) (Lakkaraju et al., 2014; Dong et al., 2014; Nguyen and Shirai, 2015; Wang et al., 2016a) or the internal node distance can be computed and used for attention weight decay (He et al., 2018a).
- Recently, graph neural networks (GNNs) are explored to learn representations from the dependency trees (Zhang et al., 2019; Sun et al., 2019b; Huang and Carley, 2019).

# Shortcomings of these Approaches

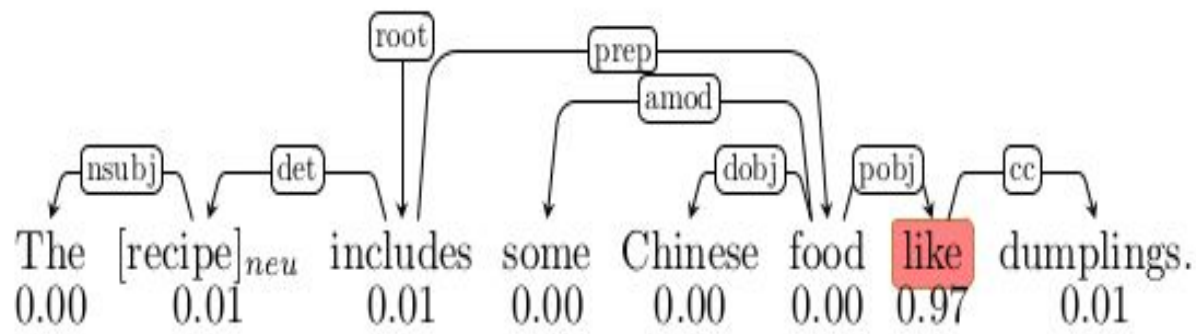
- the dependency relations, which may indicate the connections between aspects and opinion words, are ignored.
- empirically, only a small part of the parse tree is related to this task and it is unnecessary to encode the whole tree.
- the encoding process is tree-dependent, making the batch operation inconvenient during optimization.

## Aspect, Attention and Syntax

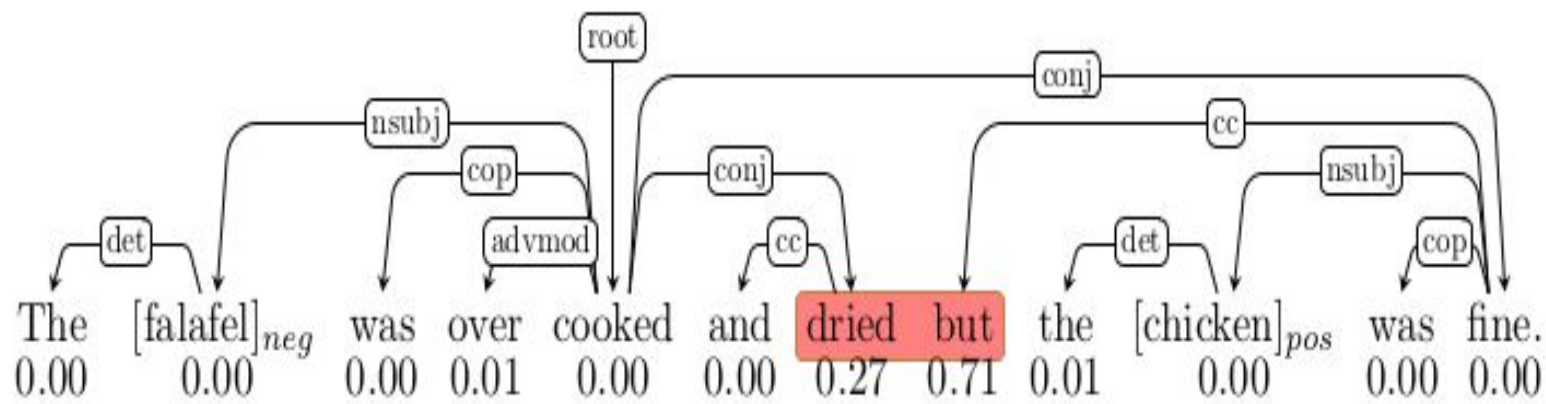
- Intuitively, connecting aspects with their respective opinion words lies at the heart of ABSA.
- Most recent efforts (Wang et al., 2016b; Li et al., 2017; Ma et al., 2017; Fan et al., 2018) resort to assorted attention mechanisms to achieve this goal and have reported appealing results.
- However, due to the complexity of language morphology and syntax, these mechanisms fail occasionally.



(a)



(b)



(c)

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**Algorithm 1** Aspect-Oriented Dependency Tree

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**Input:** aspect  $a = \{w_i^a, w_{i+1}^a, \dots w_k^a\}$ , sentence  $s = \{w_1^s, w_2^s, \dots w_n^s\}$ , dependency tree  $T$ , and dependency relations  $r$ .

**Output:** aspect-oriented dependency tree  $\hat{T}$ .

- 1: Construct the root  $R$  for  $\hat{T}$ ;
  - 2: **for**  $i$  to  $k$  **do**
  - 3:     **for**  $j = 1$  to  $n$  **do**
  - 4:         **if**  $w_j^s \xrightarrow{r_{ji}} w_i^a$  **then**
  - 5:              $w_j^s \xrightarrow{r_{ji}} R$
  - 6:         **else if**  $w_j^s \xleftarrow{r_{ij}} w_i^a$  **then**
  - 7:              $w_j^s \xleftarrow{r_{ij}} R$
  - 8:         **else**
  - 9:              $n = distance(i, j)$
  - 10:              $w_j^s \xrightarrow{n:con} R$
  - 11:         **end if**
  - 12:     **end for**
  - 13: **end for**
  - 14: **return**  $\hat{T}$
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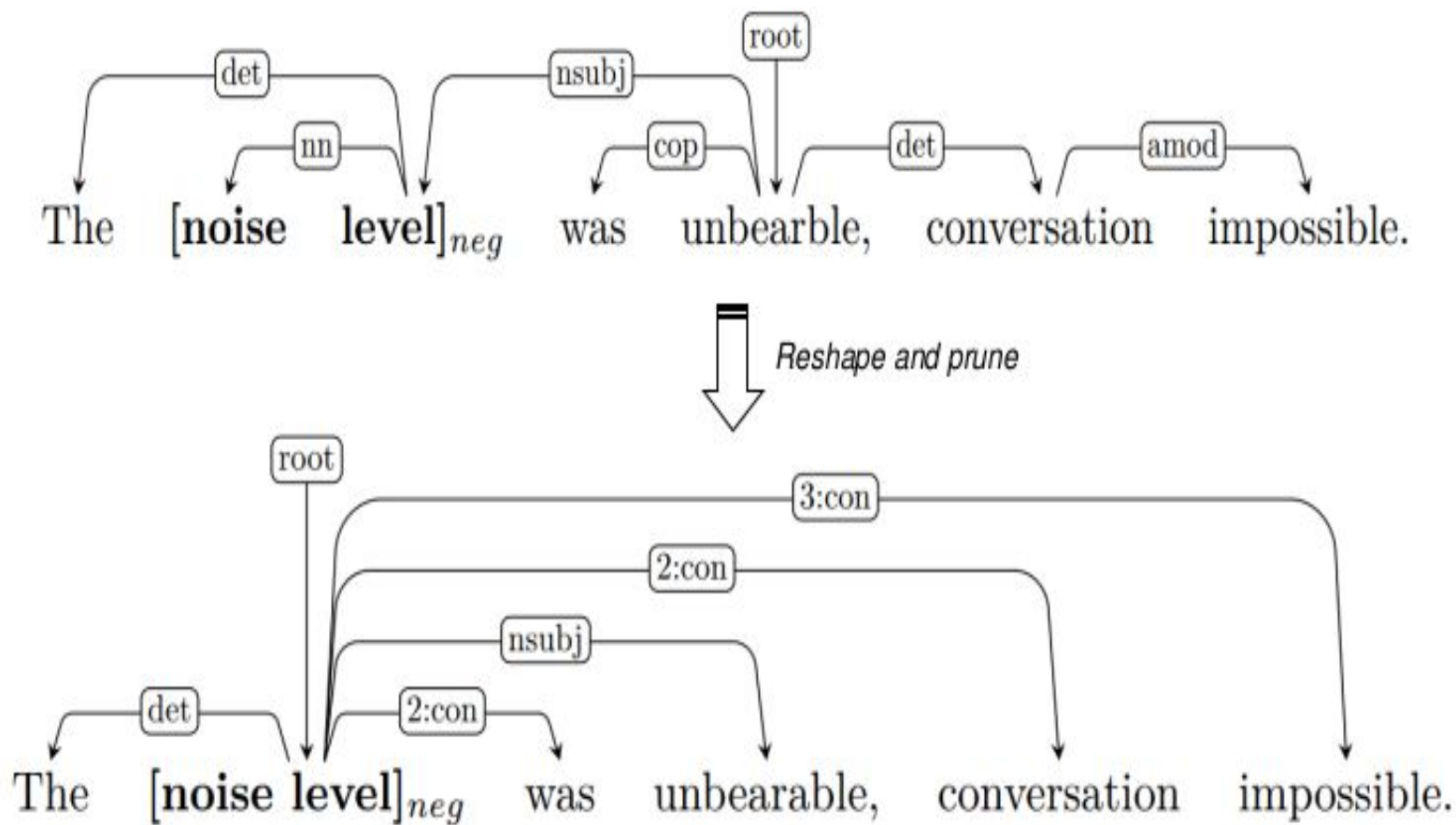


Figure 2: Construction of an aspect-oriented dependency tree (bottom) from an ordinary dependency tree (top).

**Thank You!**