

# Aspect Based Sentiment Analysis

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## 1 Motivation

Sentiment analysis aims to determine the evaluation of an author with respect to a particular topic and detecting the overall contextual polarity of a document. Sentiment analysis is increasingly viewed as a vital task both from an academic and a commercial standpoint. The majority of current approaches, however, attempt to detect the overall polarity of a sentence, paragraph, or text span, irrespective of the entities mentioned (e.g., laptops) and their aspects (e.g., battery, screen).

The project is directed towards aspect-based sentiment analysis, where the goal is to identify the aspects of given target entities and the sentiment expressed for each aspect.

## 2 Problem Statement:

The task is focussed on two domains: restaurants and laptops. For each of these domains, we have a predefined set of entity types(E) and attributes(A). For laptops, we have: 22 entity types and 9 attribute labels, and 6 entity types and 5 attribute labels for restaurants. The E#A pair defines an aspect (category).

The following tasks would be addressed in this project:

### 2.1 Task 1: In domain ABSA

#### Sub-task 1:

Identification of every entity E and attribute A pair E#A towards which an opinion is expressed in the given text. E and A are chosen from predefined inventories of Entity types (e.g. laptop, keyboard, operating system, restaurant, food, drinks) and Attribute labels (e.g. performance, design, price, quality) per domain.

#### Sub-task 2:

Identification of Opinion Target Expression (OTE). An opinion target expression (OTE) is an explicit reference (mention) to the reviewed entity E of the EA pair. The OTE slot takes the value “NULL”, when there is no explicit mention of the opinion entity. Each identified E#A pair of the given text is then assigned a polarity, from a set P = positive, negative, neutral

## 2.2 Task 2: Out of domain ABSA

*(If time permits)*

Domain adaptation considers the setting in which the training and testing data are sampled from different distributions. The implementation will be tested on a previously unseen domain for which no training data is made available.

## 3 Methodology:

### 3.1 Task 1:

**Sub-task 1:** Aspect term detection

Using Conditional Random Fields (CRF) with features extracted using named entity recognition, POS tagging, parsing, and semantic analysis.[2],[3]

**Sub-task 2:** Aspect polarity detection

Using an SVM classifier with features mainly based on n-grams, parse trees, and several out-of-domain, publicly available sentiment lexica (e.g. Wordnet, SentiWordnet).[4]

### 3.2 Task 2:

*(If time permits)*

A deep learning approach can be used to address the problem of domain adaptation for sentiment classifiers. A higher-level feature extraction is learnt in an unsupervised fashion from the text reviews of all the available domains using a Stacked Denoising Autoencoder(SDA). We then use an SVM to adjudicate the polarity of the sentence.[5]

## 4 Dataset:

Two datasets of 550 reviews of laptops and restaurants annotated with opinion tuples are provided by SemEval in XML format. All possible E#A pairs are explicitly mentioned for the laptop and restaurant domain.[1]

## 5 References:

[1].SemEval-2015 Task 12: Aspect Based Sentiment Analysis  
<http://alt.qcri.org/semeval2015/task12/>

[2].Identifying Sources of Opinions with Conditional Random Fields and Extraction Patterns, Yejin Choi and Claire Cardie, Ellen Riloff and Siddharth Patwardhan  
<https://www.cs.cornell.edu/home/cardie/papers/hlt-emnlp05-yejin.pdf>

[3].Dependency Tree-based Sentiment Classification using CRFs with Hidden Variables, Tetsuji Nakagawa,Kentaro Inui and Sadao Kurohashi  
<http://www.aclweb.org/anthology/N10-1120>

[4]. Sentiment analysis using support vector machines with diverse information sources, Tony Mullen and Nigel Collier  
[http://www.aclweb.org/old\\_anthology/W/W04/W04-3253.pdf](http://www.aclweb.org/old_anthology/W/W04/W04-3253.pdf)

[5].Domain Adaptation for Large-Scale Sentiment Classification: A Deep Learning Approach, Xavier Glorot, Antoine Bordes, Yoshua Bengio  
[http://www.icml-2011.org/papers/342\\_icmlpaper.pdf](http://www.icml-2011.org/papers/342_icmlpaper.pdf)