

# Finding Claims

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# Finding Claims

Argumentation mining is not a single unified task but a constellation of subtasks.

One of this subtasks is the task of finding the claim statements made by the writer.

This can also be broken down into three subtasks :

1. **Classifying text** as argumentative vs. non-argumentative
2. **Segmenting text** into argumentative units
3. **Identifying claims**

# 1. Classifying Text as Argumentative

## Document Level :

When dealing with traditional kind of documents the task of classifying text as argumentative can be reduced to **Genre-Classification**.

## **Karlgren and Cutting [1994] and Kessler et al. [1997]:**

Work on the Brown Corpus with the goal of distinguishing different categories from each other. One Setting was Editorials vs. Reportage

Using document/sentence length type/token ratio punctuation symbols and frequencies of specific words they achieved accuracies of 83% and 61% for reports and editorials.

# 1. Classifying Text as Argumentative

**Document Level :**

**Habernal and Gurevych [2017]:**

Compiled a corpus consisting of a set of wikipedia articles on a predefined topic.

Annotators labeled 990 documents and achieved a Fleiss k of 0.59 this led to a set of 529 persuasive on topic documents.

Task: Distinguish between persuasive and non-persuasive documents

SVM-classifier using lexical n-gram features achieved an F-Score of 0.69

# 1. Classifying Text as Argumentative

**Sub-Document Level:**

**Sentence-Level:** Majority of research tries to decide if a span of text is argumentative or not taking sentences as the default unit.

**Sub-Sentence-Level:** Deciding the question if a span of text is argumentative or not on the sentence level is often a simplification, because a complex sentence may contain more than one component of an argument

## 2. Segmenting Text into Argumentative Units

Segmenting text into units, that can later be identified to play a certain role in the argumentation, is a difficult task and is often circumvented by taking sentences as the default units.

Using sentences as default units has two consequences:

- Argumentative units smaller than a sentence can not be processed
- Argumentative units that consist out of more sentences are difficult to process

## 2. Segmenting Text into Argumentative Units

**Human Annotation:** Assign the task of demarcating argumentative minimal units, without setting any restrictions on pre-defined unit candidates, to humans and later try to reproduce it by automatic means.

**Computation:** Divide text into minimal argumentative units automatically, for instance use a syntax parser to supply clauses which then serve as candidates for minimal units.

Observation: notion of argumentative minimal units is not the same across different corpora, high variance in the size of argumentative minimal units across different domains

### 3. Identifying Claims

Claim detection is the first of two indispensable tasks for an argumentation mining system.

**Problem:** Often claims are not explicitly stated but must be inferred by the reader

Research is focused on **identifying explicitly stated** claims.



# 3. Identifying Claims

Two families of methods that have been applied to claim detection :

1. **Classification**

2. **Sequence Labeling**

# 3. Identifying Claims

## Classification :

Given a minimal unit of analysis it can be classified in different ways :

- **Binary Classification:** claim or no claim
- **Binary Classification:** claim or premise
- **Multi-Class Classification:** more types of arguments components are classified by a single classifier with the possibility of the class “none”, when a unit is not argumentative

### 3. Identifying Claims

#### Sequence Labeling :

Identification of arguments components is approached as an IOB-Labeling problem.

Words are tagged as B-premise, B-claim, I-premise, I-claim or O. (B: Begin, I: Inside, O: Outside, not part of claim or premise)

### 3. Identifying Claims

Identifying Claims in Legal Documents :

**Hachey and Grover [2006]:**

Annotation of 7 different rhetorical roles in sentences in judgements of the UK house of lords.

SVM achieved an F-Score of 0.6 using cue phrases, location, entities, sentence length, quotations and thematic words as features

Decision Tree Classifier achieved an F-Score of 0.65 only using positional features

# 3. Identifying Claims

Identifying Claims in Legal Documents :

**Palau and Moens [2009]:**

Distinguish between conclusions and supporting statements.

SVM model for premise/conclusion classification, that takes input sentences that already have been predicted to be argumentative

F-Score for premise and conclusion were 0.68 and 0.74

**Features** : syntactic, domain specific cues, token counts, position, contextual feature, with the prediction for the previous and following segment

# 3. Identifying Claims

Identifying Claims in Legal Documents :

**Rooney et al. [2012]:**

SVM sequence kernel classifier, kernel compares subsequences of sentences, where a word is tagged with its root form and a PoS label

**Araucaria DB dataset**, 1299 premises, 304 claims, 161 premise and claim, 1686 none

Overall accuracy 0.65, accuracy for claims only around 0.3

### 3. Identifying Claims

#### Identifying Claims in Instructional Text :

Instructions include advice and warnings, which are backed up with an explanation.

Claim -> Advice/Warning, Supporting Statement -> Explanation

#### **Saint-Dizier [2012]:**

Manually constructed rules, that exploit linear order of the statements and a set of common lexical patterns.

Accuracy for Advice were 79%/84% and for Warnings 88%/91% (Claim/Support)

# 3. Identifying Claims

Identifying Claims in Student Essays :

**Falakmasir et al. [2014]:**

432 essays with 8 different topics, human annotators labeled sentences that were candidate **thesis** or **conclusion** statements

Goal was to identify the most predictive features.

Positional features, features based on syntactic/semantic analysis, a set of frequent words and essay level features were most useful

**3-Way Classification:** Decision-Tree achieved F-Score 0.83 for Thesis and 0.59 for conclusion



# 3. Identifying Claims

Identifying Claims in Student Essays :

Stab and Gurevych [2014a]:

Essay has one **major claim** and arguments related to the major claim, which consist out of **claims** and **premises**.

**Persuasive Essay Corpus:** 90 Major Claims, 429 Claims

**Four-Way-Classifier:** Structural-, Lexical-, Syntactic-Features but also cues and attributes of preceding and following sentences

**SVM** performed best: F-Score of 0.63 (major claim), 0.54 (claim), 0.83 (premise) and 0.88 (non-argumentative)

# 3. Identifying Claims

Identifying Claims in Wikipedia :

Aharoni et al. and Levy et al. [2014]:

IBM Debater: System that searches web pages for arguments given a topic

-> Claim detection is dependent on predefined topic

Dataset on the basis of 32 debates with 326 relevant wikipedia articles was constructed, annotators labeled 976 topic related claims ( $\kappa = 0.39$ )

Pipeline out of 3 modules.

### 3. Identifying Claims

1. **Module:** Identifies sentences that contain a claim, passes 200 top scoring sentences to the next module
2. **Module:** Generates the 10 best candidate sub-sentences using a maximum likelihood model
3. **Module:** Ranks the identified claims for all the sentences

### 3. Identifying Claims

Identifying Claims in Social-Media and User-Generated Web Text :

Kwon et al. [2007]:

Detecting claims in a corpus of user comments on a proposed legislation.

Boosting algorithm beat an SVM and achieved an agreement with an human annotator of  $k = 0.55$

### 3. Identifying Claims

**Identifying Claims in Social-Media and User-Generated Web Text :**

**Rosenthal and McKeown [2012]:**

Claim detection in LiveJournal weblogs and Wikipedia discussion pages

- Human annotators labeled 2000 sentences as claims  $k=0.50$  for 663 LiveJournal sentences and  $k=0.56$  for 997 Wikipedia sentences
- Focused on the influence of sentiment- vs. committed-belief-features
- Found out that committed-belief-features are more predictive for wikipedia discussions, sentiment features are more predictive for LiveJournal

# 3. Identifying Claims

Identifying Claims in Social-Media and User-Generated Web Text :

Habernal and Gurevych [2017]:

Work on **Web Discourse Corpus** with the goal of identifying backing, claim, premise, rebuttal and refutation.

**11-class IOB-Tagging** using an SVM.

**Features:** Lexical, Structure and Syntax, Topic and Sentiment, Semantics and Discourse and Embeddings

Achieved **F-Score of 0.25** (Human annotator achieved 0.6)

## 4. Summary

Relatively broad range of statements are treated as claims in different text types and genres.

In some genres there is a strong position-in-document tendency that can be exploited.

If a sentence is a claim is strongly dependent on the context of a sentence.

Most commonly used algorithms are SVMs and Decision Trees.