Figurative language annotation

Annotating non literal language on a corpus of tweets

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Annotation task: literal vs figurative

What

Distinguishing between literal tweets vs figurative tweets.

How

- Annotate whether the meaning of a tweet is literal or not (figurative).
- Classify figurative language according to types.
- Train a classifier with the annotated corpus.

Corpus used

The corpus from the Sentiment Analysis of Figurative Language in Twitter task at SemEval 2015 http://alt.gcri.org/semeval2015/task11/)

Pros:

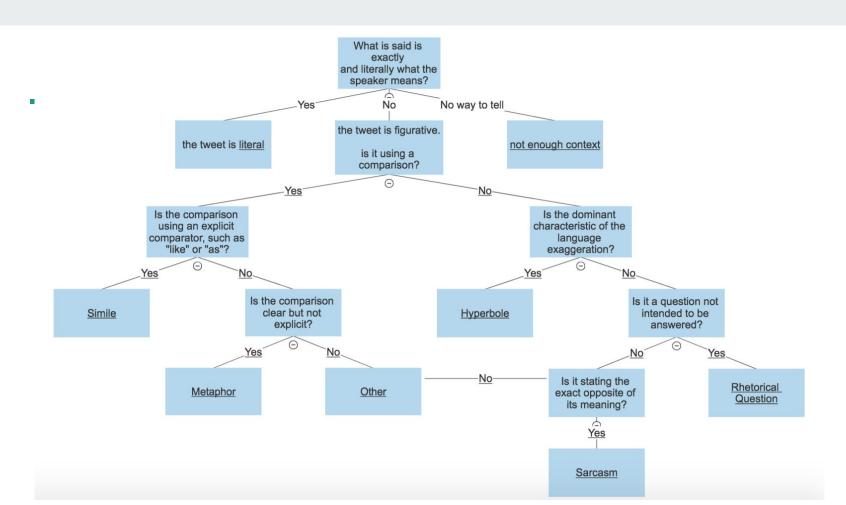
- A corpus of non literal language.
- Originally annotated for sentiment analysis.
- No literal/figurative annotation.

Cons:

 Mostly sarcastic comments (the corpus was scrapped using keywords like #not, #sarcasm or literally)

Guidelines and annotation scheme

- Literal
- Figurative
 - Sarcasm
 - Hyperbole
 - Simile
 - Metaphor
 - Rhetorical question
 - Other
- Not enough context

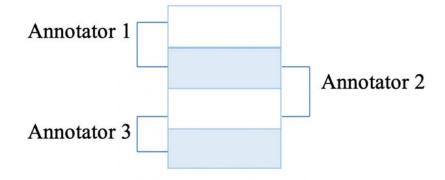


Annotation settings

Tool: MAE

Data preparation:

- 120 tweets for sample annotation
- 480 tweets for full annotation



IAA calculations

| IAA on full annotation | Annotator 1 & 2 | Annotator 2 & 3 |
|---|-----------------|-----------------|
| <pre><cross-tag>Multi-Pi (Fleiss' Kappa) [Figurative, Literal, Not_enough_context]</cross-tag></pre> | 0.6133 | 0.2206 |
| <pre><cross-tag> Multi-Kappa (Huberts' Kappa) [Figurative, Literal, Not_enough_context]</cross-tag></pre> | 0.6184 | 0.2804 |
| <tag-level> Multi-Pi (Fleiss' Kappa) Figurative::-</tag-level> | -0.1496 | -0.3934 |
| <tag-level> Multi-Pi (Fleiss' Kappa) Figurative::type</tag-level> | 0.7874 | 0.7264 |

Gold standard

- overlapped data
 - adjudication on MAE, check the disagreed tweets
- non-overlapped data

Classifier

- Logistic classifier
- BOW features

Shape of model coefficients and intercepts: (3, 1721) (3,) precision recall f1-score support Figurative 0.73 0.86 0.79 57 34 Literal 0.78 0.41 0.54 0.25 Not_enough_context 0.10 0.14 95 micro ava 0.67 0.67 0.67 95 0.54 0.51 0.49 macro avg 95 weighted avg 0.72 0.67 0.67

| | precision | recall | f1-score | support |
|---------------------|-----------|--------|----------|---------|
| Hyperbole | 0.00 | 0.00 | 0.00 | 0 |
| Literal | 0.78 | 0.38 | 0.51 | 37 |
| Metaphor | 0.00 | 0.00 | 0.00 | 0 |
| Not_enough_context | 0.10 | 0.14 | 0.12 | 7 |
| Other | 0.00 | 0.00 | 0.00 | 0 |
| Rhetorical question | 0.00 | 0.00 | 0.00 | 0 |
| Sarcasm | 0.74 | 0.73 | 0.73 | 51 |
| Simile | 0.00 | 0.00 | 0.00 | 0 |
| micro avg | 0.55 | 0.55 | 0.55 | 95 |
| macro avg | 0.20 | 0.16 | 0.17 | 95 |
| weighted avg | 0.71 | 0.55 | 0.60 | 95 |
| | | | | |