

Compositional approach applied to analysis of attitudes expressed through written language

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Objectives

We set three-fold focus in our research:

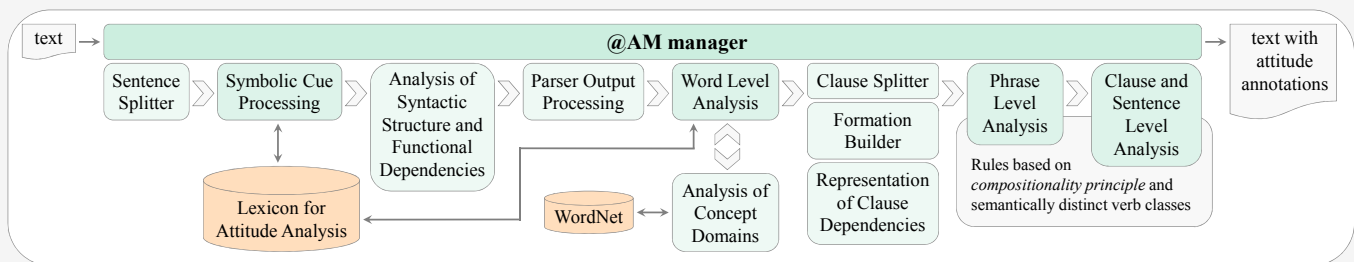
- classification of sentences according to the fine-grained attitude labels
- analysis of the strength of the attitude
- determination of the level of confidence, with which the attitude is expressed.

Approach

We developed **@AM** (ATtitude Analysis Model) system, which automatically recognizes affect, judgment, and appreciation in text.

The algorithm for attitude classification is designed based on the *compositionality principle* and a novel approach relying on the rules elaborated for semantically distinct verb classes.

Architecture and Principles of @AM (ATtitude Analysis Model)



Fine-grained attitude labels:

‘anger’ ‘disgust’ ‘fear’ ‘guilt’ ‘interest’ ‘joy’ ‘sadness’ ‘shame’ ‘surprise’ ‘affect’ ‘POS jud’ ‘NEG jud’ ‘judgment’ ‘POS app’ ‘NEG app’ ‘appreciation’ ‘neutral’

We built the **lexicon for attitude analysis** that includes:

- (1) **affect-conveying symbolic cues** (emoticons and abbreviations)
- (2) **attitude-conveying terms:**

POS	Word	Category: intensity
adjective	<i>honorable</i>	POS jud: 0.3
adjective	<i>unfriendly</i>	NEG aff (sadness): 0.5; NEG jud: 0.5; NEG app: 0.5
adverb	<i>gleefully</i>	POS aff (joy): 0.9
noun	<i>abnormality</i>	NEG app: 0.25
verb	<i>desire</i>	POS aff (interest): 1.0; POS aff (joy): 0.5

- (3) **modifiers:**

- adverbs of degree (*‘significantly’, ‘slightly’*) and adverbs of affirmation (*‘absolutely’, ‘seemingly’*) that have an influence on the strength of attitude of the related words through coefficients (from 0.0 to 2.0)
- negation words (*‘never’, ‘nothing’*), adverbs of doubt (*‘scarcely’, ‘hardly’*) and adverbs of falseness (*‘wrongly’*) that reverse the polarity of related statement
- prepositions (*‘without’, ‘despite’*) and condition operators (*‘if’, ‘even though’*) that neutralize the attitude of related words

- (4) **“functional” words:**

- intensifying adjectives (*‘rising’, ‘rapidly-growing’*), nouns (*‘increase’, ‘up-tick’*), and verbs (*‘to grow’, ‘to rocket’*), which increase the strength of attitude of related words
- reversing adjectives (*‘reduced’*), nouns (*‘termination’, ‘reduction’*), and verbs (*‘to decrease’, ‘to limit’, ‘to diminish’*), which reverse the prior polarity of related words

- (5) **modal operators**, which indicate a degree of person’s belief in the truth of the proposition (*confidence level*):

- modal verbs (conf(‘may’) = 0.27, conf(‘would’) = 0.8)
- modal adverbs (conf(‘vaguely’) = 0.17, conf(‘arguably’) = 0.63, conf(‘veritably’) = 1.0)

Compositionality Principle

The attitudinal meaning of a sentence is determined by composing the pieces that correspond to lexical units or other linguistic constituent types governed at various grammatical levels by the rules of

polarity reversal: ‘never’ & POS(‘succeed’) => NEG(‘never succeed’)

‘scarcely’ & POS(‘relax’) => NEG(‘scarcely relax’)

‘reduced’ & POS(‘enthusiasm’) => NEG(‘reduced enthusiasm’)

aggregation (fusion): POS(‘beautiful’) & NEG(‘fight’) => POS-neg(‘beautiful fight’)

NEG(‘shamelessly’) & POS(‘celebrate’) => NEG-pos(‘shamelessly celebrate’)

propagation: PROP-POS(‘to admire’) & ‘his behaviour’ => POS(‘his behaviour’)

‘Mr. X’ & TRANS(‘supports’) & NEG(‘crime business’) => NEG(‘Mr. X’)

domination: NEG(‘to deceive’) & POS(‘hopes’) => NEG(‘to deceive hopes’)

NEG(‘It was hard to climb a mountain all night long,’) & POS(‘a magnificent view rewarded the traveler at the morning.’) => POS(whole sentence)

neutralization: ‘despite’ & NEG(‘worries’) => NEUT(‘despite worries’)

intensification: Pos_score(‘extremely happy’) > Pos_score(‘happy’)

Neg_score(‘sad’) < Neg_score(‘sadder’) < Neg_score(‘saddest’).

Consideration of the Semantics of Verbs

Based on the thorough analysis of 270 first-level classes of VerbNet and their members, 73 verb classes (1) were found useful for the task of attitude analysis, and (2) were further classified into **22 classes** differentiated by the role that members play in attitude analysis.

For each of our verb classes, we developed set of rules that are applied to attitude analysis on the phrase/clause-level.

EXAMPLES:

“Object-centered (oriented) emotional state” (subject is ignored):

S & V+(‘admires’) & O-(‘mafia leader’) =>
interior: (verb valence dominance, V_score) => ‘POS aff’;
exterior: (verb valence reversal, max(V_score, O_score)) => ‘NEG jud’.

“Subject-driven change in emotional state (trans.)” (object is ignored):

S-(‘Fatal consequences of GM food intake’) & V-(‘frighten’) & O(‘me’) =>
interior: ‘NEG aff’; exterior: ‘NEG app’.

“Negative judgment”:

S(‘He’) & V-(‘blamed’) & O+(‘innocent person’) => interior: ‘NEG jud’; exterior: ‘NEG jud’.

“Adverse (unfavorable) attitude”:

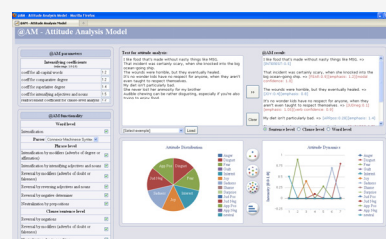
S(‘They’) & [V neg. reinforcement](‘prevented’) & O-(‘the spread of disease’) => ‘POS app’; O_score^.

“Communication indicator/reinforcement of attitude” and

“Termination of activity”:

S(‘Max’) & [V neg. reinforcement, confidence:0.2](‘doubt’) & PP-(‘that’ S+(‘his good fortune’) & [V termination](‘will ever end’)) => ‘POS app’; PP_score^; confidence:0.2.

@AM Interface and Evaluation Results



The evaluation of our method on 1000 sentences, which describe personal experiences (*Experience Project: www.experienceproject.com*), showed promising results: average accuracy on fine-grained level (14 categories) was 62%, on middle level (7 categories) – 71%, and on top level (3 categories) – 88%.