

# ATEX: A RULE-BASED SENTIMENT ANALYSIS SYSTEM PROCESSING TEXTS IN VARIOUS TOPICS

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# Sentiment analysis

- Started in 1990's in works by J. Wiebe
- Subjectivity, sentiment analysis, polarity classification
- Russia: special in being commerce- and business-oriented
  - Fine-grained tuning capabilities
  - Easy access to rules and output analysis
  - Thus rule-based algorithms appear to fit better

# The ATEX sentiment analysis algorithm

- Implemented in EPAM Systems
  - A project in Kazakhstan
  - For Russian news texts
- A rule-based algorithm
- Linguistically rich
  - Processes deep linguistic information
- Participated in the ROMIP 2012 sentiment analysis tracks
  - Results

# ATEX: morphological analysis

- A morphological dictionary
  - generated initially with data from [Zaliznyak]
  - contains all word-forms and respective morphological information
- Analysis of unknown words and typos
  - Word-ending: defines normal form and morphology
  - Word-prefix: is deleted and the rest of the word is looked up in the dictionary
  - Fuzzy search in dictionary: Levenstein distance

# ATEX: syntactic analysis

- Formal grammar:
  - Homonymy resolution
  - Phrases
  - Bigger phrases, up to complex sentences
- Algorithm applying dependencies to the resulting phrase structure
- Some additional rules applying negation
- Finally: a dependency tree representing a sentence

# ATEX: sentiment analysis

- Sentiment: a polarity value for one or more words
  - +1
  - -1
  - 0
  - null
- No sentiment degree: all polarity values are equally “sentimental”

# Keyword sentiment

- Keywords
  - *хороший, плохой, неприятный, трус, успех, провал, угроза, позитив, оперативно, своевременно, слишком*

# Sentiment rules

- Phrase
  - *пойти навстречу, на лапу, душа компании, по фазе, так себе, промыть мозг, поставить крест, с ума, из ума, ниже плинтуса*
- Inverted sentiment
  - *нет, без, отсутствие, удаление, лишение, отрицание, устранение, отсутствовать, удалять, лишать, отрицать, устранять*
  - Cases of special words
    - Нет проблем
    - Не удастся
- Dependency-based rules
  - Degree, loss, problems, lack, flexibility



# ATEX results: examples

Sample id	Text	Sentence sentiment
1049	"На данный момент не вижу <u>перспективы(-1)</u> никаких военных действий за исключением мер по защите дипломатических представителей, а также справедливого наказания ответственных за эту ужасную <u>акцию(-1)</u> ", - сказал Терци.	-1
1068	"Льоренте все еще принадлежит Атлетике и, похоже, готов играть. Впрочем, мы все равно <u>потеряли(-1)</u> одного <u>отличного(0)</u> <u>футболиста(0)</u> и <u>хорошего(0)</u> <u>человека(0)</u> ", - сказал Бьелса, намекая на уход Хави Мартинеса в мюнхенскую "Баварию".	-1
1108	"В период после нашей предыдущей встречи мировая экономика по-прежнему испытывала немалые трудности и продолжает подвергаться рискам падения; финансовые рынки <u>остаются(-1)</u> нестабильными, тогда как высокий <u>уровень(-1)</u> дефицита госсектора и государственной задолженности в некоторых развитых экономиках в значительной мере сдерживает процесс восстановления экономики", - отмечается в документе.	-1

# Task description: ROMIP sentiment analysis tracks

- 3-class polarity classification of direct and indirect speech in news texts: positive, negative and neutral (=no polarity)
- 2-class polarity classification of product reviews: positive and negative
- 3-class polarity classification of product reviews: positive, negative and containing significant positive and negative polarity at the same time

# Task description: tuning the ATEX system

- No training applied
- Tuning:
  - A corpus of new texts from Russian and Kazakh web resources in Russian: 3000 sentences manually annotated
- Sentence polarity value:
  - Sign of words' mean value
  - 0 = neutral = null
- 2 modes: with/no sentence boundaries for resulting document polarity

# Task description: ROMIP data

Topics	Number of samples	Positive	Negative	Neutral/ containing + and -	Percent of the most frequent class, %
News	4573	1448	1234	1890	41
Movies	408	330	78	-	80
		266	63	79	65
Books	129	112	17	-	87
		100	9	20	78
Cameras	411	397	14	-	97
		351	7	53	85

# ATEX results: evaluation

- Evaluation: F-measure, Accuracy, Precision, Recall
- Unbalanced data
- -> F-measure appears adequate
- Example:
  - 90% documents negative,
  - A goldstandard approach assigning the most frequent class achieves 90% accuracy
  - Although totally insensitive to the other class
  - While F-measure accounts for Recall and Precision for all classes

# ATEX results: news samples

Number	System ID	Object	Classes	Precision Macro	Recall Macro	F_Measure Macro	Accuracy	
1	xxx-4	news	3	0.626	0.616	0.621	0.616	
2	ATEX	news	3	<b>0.606</b>	<b>0.579</b>	<b>0.592</b>	<b>0.571</b>	with no sentences
3	ATEX	news	3	<b>0.606</b>	<b>0.576</b>	<b>0.590</b>	<b>0.569</b>	with sentences
4	xxx-5	news	3	0.579	0.568	0.574	0.575	

# ATEX results: blog reviews

- Movies:
  - 2-class: 1<sup>st</sup>, 0.707 F-measure. Rel. lower accuracy – 0.806
  - 3-class: 2<sup>nd</sup>, 0.503 F-measure. Relatively lower accuracy – 0.596
- Cameras:
  - 2-class: 5<sup>th</sup>,
  - 3-class: 5<sup>th</sup>,
- Books:
  - 2-class: 3<sup>rd</sup>,
  - 3-class: 6<sup>th</sup>.

# Conclusions

- Relatively high results in native news sample
  - Despite difference in sources and time in tuning/testing data
- Considerable results in all other topics
  - Highest performance in movie reviews
  - More modest in books, cameras.
- -> the results reflect the degree of technical information relevant to sentiment in the topic:
  - Movies and news appear the most common and the closest to each other
  - Books and especially cameras include more special terms in their sentiment expressions
- The rule-based algorithm with no training performs with high replicability
  - High F-measure
  - High results in various topics



# Future work

- Statistical account of rules applied:
  - Retrieve most significant rules
  - Retrieve important differences between topics
  - A step towards automatic lexical rule filling
- Model a neutral class and a class containing both polarity values
  - ML appears to draw fine lines between plarity, neutral and containing both

# Thank you!

- Questions?