

# **Contribution to research new models of knowledge** extraction on **BigData systems**

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### **Motivation**

### Natural Language Processing (NLP) has a wide range of applications such as:



## **Ongoing Work**

Participation in a Sentiment Analysis Competition (SemEval 2015)

We have taken part in the following competition SemEval-2015 Task 10 Subtask B: Sentiment Analysis in Twitter [1]. Goal:

• Given a message from Twitter classify it as positive, negative or neutral.



Scope

intensification

Human performance exceed computers in many complex NLP tasks:



SAEED: #NowPlaying: BEP, Ricky Martin and KT Tunstall! Great songs to get you through your Sunday! Hate the rain!! http://boltonfm.com/listen-live



JACKALS GOAL! Jimmy Martin sneaks a rebound past Killeen to give Elmira a 2-1 lead. Bushee & Bellamy with the assists.1:09 left 2nd period







### **General Approach**

Supervised Strategy with Logistic Regression

Scope adversatives

Scope Scope negation dependencies



Figure 1 : Architecture of the system.

Nonetheless computers are faster and they are able to solve problems at web-scale.

### **Thesis Objectives**

- Objective 1: Research in new unsupervised algorithms for the application in NLP tasks.
- Objective 2: Development of bigdata algorithms to solve NLP problems in the Terabyte-scale.
- Objective 3: Research in new technologies for fast adaptation in different context of text mining models.

Ensemble of classifiers with majority voting strategy

 CRFs for complex feature extraction: negation, comparison, adversative clauses, etc [2, 3]

The steps performed by the system are:

Preprocessing Step: emoticon substitution, multiword hashtag splittage, mentions and URL substitutions, etc

② Data Tagging: polarity dictionaries, verb reversal detection, etc.

OS data extraction

 Over Syntactic Information Extraction: detection of
Over Syntactic Information
Over Syntactic Inform
Ove negation, adversative or polarity reversal scopes using CRFs

**5** Feature extraction and classification of sentences

### Results

Test	<b>F-score</b>
LiveJournal 2014	72.63
SMS 2013	61.97
Twitter 2013	65.29
Twitter 2014	66.87
Twitter 2014 sarcasm	59.11
Twitter 2015	60.62
Twitter 2015 sarcasm	56.45

Table 1 : Performance in progress and input test.

- 16th position out of 40 competitors in both sarcasm and regular 2015 datasets.
- 1st position in 2014 Tweet Sarcasm dataset.

 Generalized degradation between 2014 and 2015 performance results.

#### References

### **Research Plan (Next Year)**



- [1] S. Rosenthal, P. Nakov, S. Kiritchenko, S.M. Mohammad, A. Ritter, and V. Stoyanov. 2015. Semeval-2015 Task 10: Sentiment Analysis in Twitter. In Proceedings of the 9th International Workshop on Semantic Evaluation, SemEval '2015, Denver, Colorado, June
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- [3] E. Lapponi, E. Velldal, L. Øvrelid, and J. Read. 2012b. Uio 2: Sequence-Labeling Negation Using Dependency Features. In Proceedings of the First Joint Conference on Lexical and Computational Semantics-Volume 1.pages 319–327.

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