### SENTIMENT ANALYSIS



By: Vinay Kumar, CSE, IIT Kanpur Guided by: Prof. Amitabha Mukerjee **Sentiment Analysis**: What is it and why are people doing something about it?

But what is sentiment analysis anyway?

#### **Sentiment Analysis**

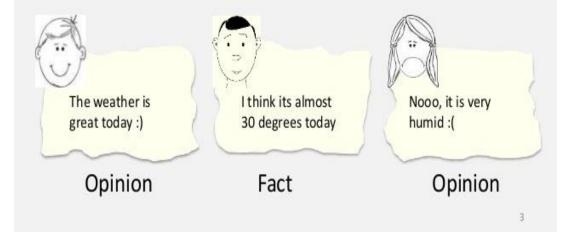
#### Definition:

"It is the field of study that analyzes people's opinions, sentiments, evaluations, appraisals, attitudes, and emotions towards entities such as products, services, organizations, individuals, issues, events, topics and their attributes(from sources like text)."

-Pang, Bo and Lillian Lee

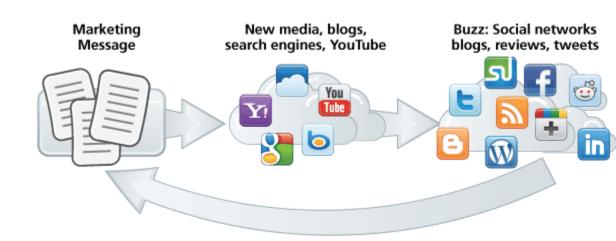
# Sentiment Analysis

"Sentiment analysis is the task of identifying positive and negative opinions, emotions and evaluations in text"



#### Sentiment Analysis (Applications)

- Stock Prediction
- Social Media Monitoring
- TwitterSentiment
- Movie/product reviews





### **Methodology**

Dataset

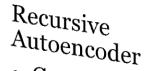


Multinomial regression

 Multiclass classifier

### Sentiment

• (based on all probabilities)



- Creates a Ndimensional
- Phrases and representation Sentences (motor each phrase

#### **Recursive Autoencoder**



### **Multinomial Regression(Softmax):**

$$J(\theta) = -\frac{1}{m} \left[ \sum_{i=1}^{m} \sum_{j=1}^{k} 1 \left\{ y^{(i)} = j \right\} \log \frac{e^{\theta_j^T x^{(i)}}}{\sum_{l=1}^{k} e^{\theta_l^T x^{(i)}}} \right]$$

Cost **Function:**  Notice that this generalizes the logistic regression cost function, which could also have been written:

$$J(\theta) = -\frac{1}{m} \left[ \sum_{i=1}^{m} (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)})) + y^{(i)} \log h_{\theta}(x^{(i)}) \right]$$
$$= -\frac{1}{m} \left[ \sum_{i=1}^{m} \sum_{j=0}^{1} 1 \left\{ y^{(i)} = j \right\} \log p(y^{(i)} = j | x^{(i)}; \theta) \right]$$

$$J(\theta) = -\frac{1}{m} \left[ \sum_{i=1}^{m} \sum_{j=1}^{k} 1\left\{ y^{(i)} = j \right\} \log \frac{e^{\theta_{j}^{T} x^{(i)}}}{\sum_{l=1}^{k} e^{\theta_{l}^{T} x^{(i)}}} \right] + \frac{\lambda}{2} \sum_{i=1}^{k} \sum_{j=0}^{n} \theta_{ij}^{2}$$
: **Weight decay**

Class label probablities:

$$p(y^{(i)} = j | x^{(i)}; \theta) = \frac{e^{(\theta_j - \psi)^T x^{(i)}}}{\sum_{l=1}^k e^{(\theta_l - \psi)^T x^{(i)}}}$$

$$= \frac{e^{\theta_j^T x^{(i)}} e^{-\psi^T x^{(i)}}}{\sum_{l=1}^k e^{\theta_l^T x^{(i)}}}$$

$$= \frac{e^{\theta_j^T x^{(i)}}}{\sum_{l=1}^k e^{\theta_l^T x^{(i)}}}.$$

#### **Dataset**

Consists of 8544 sentences which is converted to 156060 English phrases from movie reviews.

Kaggle. 2014. Sentiment Analysis on Movie Reviews. https://www.kaggle.com/c/sentimentanalysis-on-movie-reviews.

#### **Issues and Resolutions**

- Randomized N-dimensional vector: Using word2vec to map similar words nearby to exploit the similarity of meaning/context.
- Belong to more than one class -> Unnecessary errors : Using
   Softmax regression instead of logistic regression.
- Altering the values of random parameters as N,  $\alpha$ , the parameter that controls the relative weighting between the reconstruction error, error of the unsupervised part, and the cross-entropy error, error of the supervised learning part.

#### Source code for Recursive Auto-encoder:

http://www.socher.org/index.php/Main/Semi-SupervisedRecursiveAutoencodersForPredictingSentimentDis tributions

Thanks to Richard Socher (Phd fellow at Stanford University) and Stanford CoreNLP for source code of Recursive Autoencoder

#### References:

- [1] R. Socher, J. Pennington, E. H. Huang, A. Y. Ng, and C. D. Manning. 2011b. SemiSupervised Recursive Autoencoders for Predicting Sentiment Distributions. In EMNLP Url: http://ai.stanford.edu/ang/papers/emnlp11-
- RecursiveAutoencodersSentimentDistributions.pdf
- [2] Bahareh Ghiyasian and Yun Fei Guo. 2014. Sentiment Analysis Using SemiSupervised Recursive Autoencoders and Support Vector Machines. Stanford.edu
- [3] Google. 2014. word2vec. https://code.google.com/p/word2vec/

## **Questions?**

