POLIMI GRADUATE MANAGEMENT

INTRODUCTION TO NATURAL LANGUAGE PROCESSING

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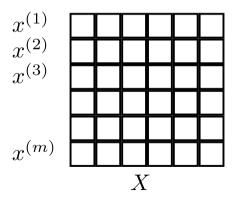


CONTENTS

- Introduction
- Data Preparation
- Sentiment Analysis
- ▶ Topic Modeling

NATURAL LANGUAGE PROCESSING

From structured data to unstructured data



- $x^{(1)}$: Walking across the sitting room, I turn the television off.
- $x^{(2)}$: @UnitedAirlines second flight in a month #cancelled thanks a lot #worstairlineever

SOME IMPORTANT QUESTIONS

- We would like to exploit "all" that we know on structured data
- How can we represent text in a structured way?
- Which are the limits/problems of a representation?



THE ELEMENTARY UNIT

The elementary unit of a digital image is the pixel, what about text?

→ words? letters? tokens?

- ► The same words can have different meanings
- The same meaning can be expressed by different words



CONTEXT AND SEQUENCES

- The meaning of a word depends on its context
- Context is not always defined by proximity

This summer I went to Italy. I had a lovely time there, engaging in both recreational and cultural **<blank#1>**, meeting old friends and making new ones [...] but most of all I had my food, **<blank#2>**.

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This summer I went to Italy. I had a lovely time there, engaging in both recreational and cultural **<blank#1>**, meeting old friends and making new ones [...] but most of all I had my food, **<blank#2>**.

Order plays an important role

Only he can play that instrument **He only** can play that instrument



TEXT CLEANING

- Convert to lower case
- Remove punctuation
- Remove numerical values
- Typos
- Remove special characters ([?@)
- Remove stop words (the, it, etc)
- Remove special description words ([chorus], [fade], [applause])
- ► Tokenize text (O'Neill \rightarrow [o] [neill], [o'neill]?; aren't \rightarrow [arent], [are][nt]?)
- Create bi-grams or tri-grams ([United Kingdom] vs [United][Kingdom])
- Normalization:
 - Stemming (car, cars, car's, cars' → car;)
 - Lemmatization (am, are, is \rightarrow be)

POLIMI SRADUATE MANAGEMENT

TEXT REPRESENTATION

- 1. Corpus: a collection of text
- 2. Document-Term Matrix: word counts in matrix format
- 3. TF-IDF: Term Frequency Inverse Document Frequency

$$\mathsf{TF}\mathsf{-IDF} = f_{t,d} \times \mathsf{idf}(t,D)$$

where

- $f_{t,d}$: the number of times that term t occurs in document d.
- $\mathrm{idf}(t,D)=\log\frac{|D|}{|\{d\in D: t\in d\}|+1}$: log of the inverse of the number of documents containing term t.



SENTIMENT ANALYSIS

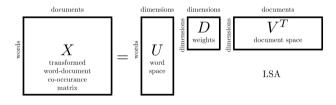
TextBlob Module: Linguistic labeled the sentiment of words.

```
https://github.com/sloria/TextBlob/blob/
eb08c120d364e908646731d60b4e4c6c1712ff63/textblob/en/
en-sentiment.xml
```

- Sentiment Labels: Each word is labeled in terms of
 - Polarity: negative(-1) or positive(+1)
 - Subjectivity: subjective(0) or fact(+1)
- Sentiment of words can vary based on where it is in a sentence.
 - Negation multiplies the polarity by -0.5



TOPIC MODELING - LSA

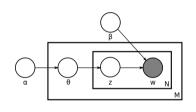


Latent Semantic Analysis (LSA)

Singular Value Decomposition (SVD) of the Document-Term Matrix



TOPIC MODELING - LDA

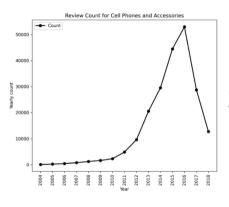


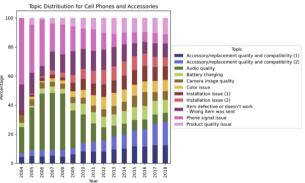
Latent Dirichlet Allocation (LDA)

- **Documents (**M**) are probabilistic distribution over topics**: let say that a document is p_i % of topic i.
- Topics are probabilistic distribution over words: given a topic chosen according to the distribution of the document, we generate a word according to the topic distribution
- Random initialisation: assign each word to a random topic
- Update each word by considering
 - proportion of words in the document of topic
 - proportion of topics in all documents for the word
- Repeat until stopping condition



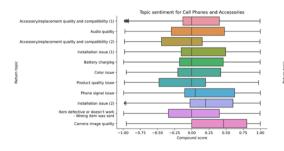
METADATA: TOPICS IN TIME

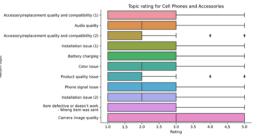






TOPIC + SENTIMENT AND METADATA







THANK YOU