

### Sentiment

 Computational identification, catagorization, quantification, extraction, and study of attitudes, emotional states, and/or subjective information



Figure 4-2. A review that praises some aspects and criticizes few

### Different Tasks

#### Simplest task:

• Is the attitude of this text positive or negative?

#### More complex:

• Rank the attitude of this text from 1 to 5

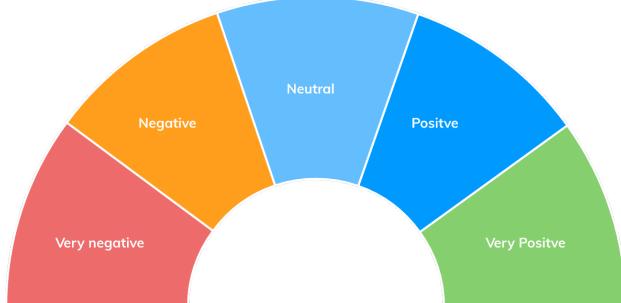
#### Advanced:

• Detect the target, source, or complex attitude types

## Polarity and Intensity

- Polarity refers to the overall sentiment conveyed by a particular text, phrase or word
- Intensity defines how strong or weak the sentiment/emotion is with respect to the context

 Difference between "this restaurant was good" and "this restaurant was the best"



## Subjectivity

- Texts can be objective or subjective
- Subjective sentence expresses some personal feelings, views, beliefs, opinions, allegations, desires, beliefs, suspicions, and speculations whereas Objective sentences are factual

#### **Objective**

### Subjective

- 1.The sky is blue.
- 2. The temperature outside is 20 degrees Celsius.
- 3. The Earth orbits around the Sun.

- 1. The movie I watched last night was incredibly boring.
- 2. I love listening to music while driving.
- 3. The beach is the most relaxing place in the world.

### How are these calculated?

- AFINN Lexicon
- SentiWordNet
- Pattern
- VADER (Valence Aware Dictionary and Sentiment Reasoner)
- Multi-Perspective Question Answering

### Lexicon-based solutions

- "Lexicons" or lists of positive and negative words are created
- Can use a series of heuristics to encode how contextual elements increment, decrement, or negate the sentiment of text

#### Advantages

- Sorting Data at Scale
- Real-Time Analysis
- Consistent criteria

### ML and DL solutions

• Sentiment Analysis is just another type of classification problem

- Common models used for sentiment analysis:
  - Naïve Bayes
  - SVM
  - Random Forest
  - LSTMs
  - Transformers

### Example dataset features

- Counts of number of positive and negative words
- POS tags
  - adjectives and adverbs may be more likely to indicate sentiment than nouns or verbs
- Word occurrence may matter more than word frequency
  - The occurrence of the word "fantastic" may tell a lot
  - If it occurs 5 times, it may not tell much more
- Boolean Multinomial Naïve Bayes
  - Clip all the word counts in each document at 1

### Example using TextBlob

- Sentiment calculation from the pattern library
- Measure polarity and subjectivity
- It can be used in place for NLTK and spaCy library

```
doc=TextBlob("We are having fun here")
doc.polarity # -1 to +1
doc.subjectivity # 0 to +1
doc.sentiment # returns both
```

### Example using TextBlob

- Ignores one-letter words
- Ignore words it doesn't know anything about
- Finds words and phrases it can assign polarity and subjectivity to, and it averages them all together for longer text

## Example using Vader

- A lexical dictionary contains regular words, as well as phrases like "not good", "really cool", or "the best", emoticons like happy faces:) and sad faces:(, and abbreviations that convey strong feelings such as LOL or OMG
- Lexical features were rated for the polarity and intensity on a scale from "-4: Extremely Negative" to "+4 Extremely Positive" by 10 independent human raters
- The average of these scores is used as the sentiment indicator for each lexical feature in the dictionary

## Example using Vader

• The compound score is the sum of positive, negative & neutral scores which is then normalized between -1(most extreme negative) and +1 (most extreme positive).

$$\frac{x}{\sqrt{x^2 + \alpha}}$$

x is the sum of the sentiment scores of the constituent words alpha is a normalization parameter that we set to 15

 Heuristics for capitalization, punctuation, sentences containing "but", degree modifiers, negation

## Example using Vader

```
import nltk
nltk.download('vader_lexicon')
from nltk.sentiment.vader import SentimentIntensityAnalyzer
sid = SentimentIntensityAnalyzer()
scores = sid.polarity_scores(text)
print(scores)
# {'neg': 0.0, 'neu': 0.0, 'pos': 0.0, 'compound': 0.0}
```

## Challenges

- Negation, inverted word ordering
- Metaphors
- Sarcasm
- Negative term used in a positive way
- Context
- Comparison
- Subtlety

# Aspect-level Sentiment Analysis

- If you identify the sentiment of a piece of text it may be beneficial to know what is being written about
- Supervised or unsupervised

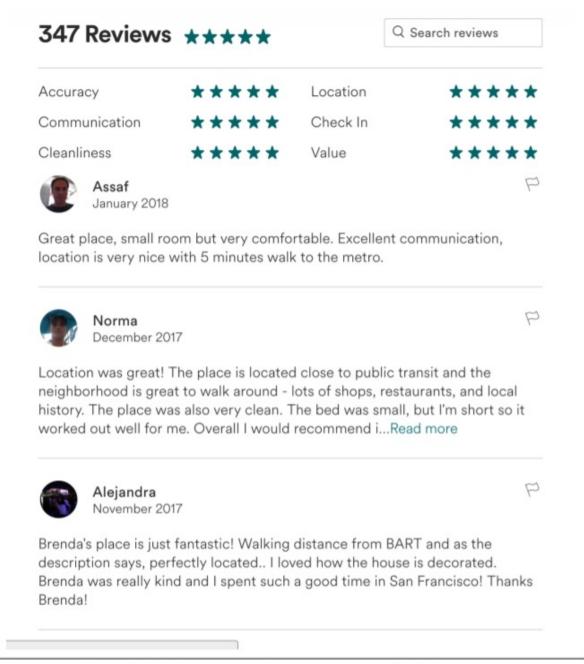


Figure 9-15. Aspect-level ratings on reviews given in a travel website

### Next time

Information Extraction