Assignment 7 + Text Classification Basics

(SNLP Tutorial 7)

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15th, 17th June

Assignment 7

- Exercise 1: Count Tree
- Exercise 2: Kneser-Ney Smoothing
- Bonus: Smoothing Techniques

Text Classification

Fill in the classes:

- f : Text $\rightarrow C$ (classes/categories)

Issues with this?

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- Sense Disambiguation: Word+sentence $\rightarrow$

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- $\{NN, VERB, PART., \ldots\}^{|S|}$
- Sense Disambiguation: Word+sentence $\rightarrow$
- Senses of Word

Issues with this?

Classification vs. Clustering

	Classification	Clustering
Method	???	???
Classes	???	???
# Classes	???	???

Classification vs. Clustering

	Classification	Clustering
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	Classification	Clustering
Method	Supervised	Unsupervised
Classes	Given	Unknown
# Classes	Given	(Mostly) unknown

Binary vs. Multi-Class Classification

Multi-Class

- $f : D \rightarrow \{\text{politics}, \text{NLP}, \text{healthcare}, \text{sport}, \ldots\}$

How to turn this into a binary classification?

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How to turn this into a binary classification?

Binary

- $f_1 : D \rightarrow \{\text{politics}, \text{not politics}\}$
- $f_2 : D \rightarrow \{\text{NLP}, \text{not NLP}\}$
- $f_3 : D \rightarrow \{\text{healthcare}, \text{not healthcare}\}$
- $\ldots$

Binary vs. Multi-Class Classification

Multi-Class

- $f : D \rightarrow \{\text{politics}, \text{NLP}, \text{healthcare}, \text{sport}, \ldots\}$

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How to turn multiple multi-class into a single multi-class?

Flat vs. Hiearchical

Flat Classification

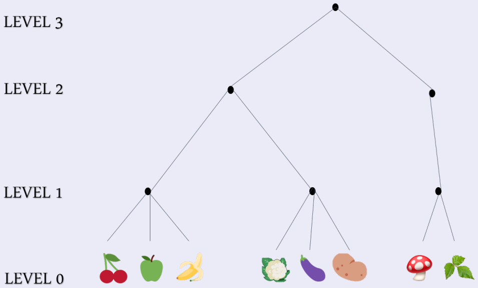
$f_1 : D \rightarrow$ 

Flat vs. Hiearchical

Flat Classification

$f_1 : D \rightarrow$ 

Hierarchical Classification



Single-Category vs Multi-Category

- $f : D \to 2^C$

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- Topic detection: Document $\rightarrow \{\text{politics}, \text{NLP}, \text{healthcare}, \text{sport}, \ldots\}$

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- Sentiment analysis: Document $\rightarrow \{\text{positive}, \text{negative}, \text{interested}, \ldots\}$

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- Topic detection: Document $\rightarrow$ $\{(\text{politics, news}), (\text{NLP, Machine Learning}), (\text{healthcare, nutrition}), (\text{sport, biography}), \ldots\}$

Single-Category vs Multi-Category

- $f : D \to 2^C$
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Feature Extraction

- Move from text to more processable domain
- How? (at least three “approaches”)

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Binary/indicator features

$$f_b(doc) = \begin{cases} 1 & \text{Contains string } \texttt{"Super free \$\$\$ discount"} \\ 0 & \text{Otherwise} \end{cases}$$

Integer features

$f_i(doc) =$ Number of occurences of `"buy"`

Real-valued features

$$f_r(doc) = \frac{\text{Number of occurences of } \texttt{"buy"}}{|doc|}$$

Feature Selection

TODO

Document Frequency

DF

$$df(term) = \frac{|\{doc|term \in doc, doc \in D\}|}{|D|}$$

- Remove rare items ($df \leq \frac{2}{|D|}$)
  Won't occur in new documents anyway
- Remove frequent items ($df = 1$)
  Usually stop words
  No information

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- Remove frequent items ($df = 1$)
  Usually stop words
  No information
- Sometimes not a good idea (interaction with other terms, etc.)
- Stopword distribution gives information in author identification

Information Gain

- Information gained (reduction in entropy) by knowing whether a term is present

$$
\begin{aligned}
G(C,t) =& H(C) - H(C|t) \\
=& -\sum_i p(c_i) \log p(c_i) \\
& + \ p(t) \sum_i p(c_i, t) \log p(c_i, t) \\
& + \ p(\overline{t}) \sum_i p(c_i, \overline{t}) \log p(c_i, \overline{t})
\end{aligned}
$$

Pointwise Mutual Information

- Difference between observed distribution and independent

$$\text{pmi}(c_i, t) = \log \frac{p(c_i, t)}{p(c_i) \cdot p(t)}$$

- TODO (relation to mutual information)

Chi Square χ^2

$$\chi^2(c_1, c_2) = \sum_{tt,tf,ft,ff}(O - E)^2$$

- TODO example
- TODO table

Term Strength

- Two documents: d_1, d_2
- Term t
- $p(t \in d_2 | t \in d1)$
- *What is the probability that the term t will be in* d_2 *given that it is in* d_1*?*
- If two documents related $\rightarrow$ high probability
- If two documents not related $\rightarrow$ low probability
- "Constant" with stop words

Resources

1. UdS SNLP Class, WSD: https://teaching.lsv.uni-saarland.de/snlp/