# Assignment11 - Natural Language

Given: July 17 Due: July 22

## Problem 11.1 (Ambiguity)

1. Explain the concept of ambiguity of natural languages.

*Solution: Ambiguity* is the *phenomenon* that in *natural languages* a single *utterance* can have multiple *meanings*.

2. Give two examples of different kinds of ambiguity and explain the readings.

*Solution:* Here are some examples

- "bank" can be a financial institution or a geographical feature.
- In "I saw her duck" the word "duck" can be a verb or a noun.
- "Time flies like an arrow" could be about the preferences of special insects ("time flies") or about the fact that time passes quickly – e.g. in an exam.
- In "Peter saw the man with binoculars", it could be Peter who is using binoculars, or it could be that Peter saw "the man" who had "binoculars".

#### Problem 11.2 (Language Identification)

You are given an English, a German, a Spanish, and a French text corpus of considerable size, and you want to build a language identification algorithm *A* for the EU administration. Concretely *A* takes a string as input and classifies it into one of the four languages  $\ell^* \in \{English, German, Spanish, French\}$ . The prior probability distribution for the strings being English/German/Spanish/French, is  $\langle 0.4, 0.2, 0.15, 0.15 \rangle$ .

How would you proceed to build algorithm *A*? Specify the general steps and give/derive the formula for computing  $\ell$  given a string  $\mathbf{c}_{1:N}$ .

### Solution:

Build a trigram *trigram language model* P(c<sub>i</sub> | c<sub>i-2:i-1</sub>, ℓ) for each candidate language ℓ by counting trigrams in an ℓ-corpus.

2. Apply Bayes' rule and the Markov property to get the most likely language:

$$\ell^* = \underset{\ell}{\operatorname{argmax}} (P(\ell \mid \mathbf{c}_{1:N}))$$
  
= 
$$\underset{\ell}{\operatorname{argmax}} (P(\ell) \cdot P(\mathbf{c}_{1:N} \mid \ell))$$
  
= 
$$\underset{\ell}{\operatorname{argmax}} (P(\ell) \cdot (\prod_{i=1}^{N} P(\mathbf{c}_i \mid \mathbf{c}_{i-2:i-1}, \ell)))$$

## Problem 11.3 (Language Models)

1. How can we obtain a *trigram* model for a *natural language*? Explain the *probability distribution* involved.

Solution: We need a *corpus* of words over *L*. Then we count how often each *trigram* occurs in it and use that to estimate the probability distribution P(T = t) of trigrams *t*.

2. Explain informally how we can use *trigram* models to identify the language of a document *D*.

Solution: We build a *trigram* model for each candidate language. Then we use each model to compute the probability of D occurring in that language. We choose the language with the highest probability.

3. Explain briefly what *named entity recognition* is.

*Solution:* The task of finding, in a text, names of things and deciding what class they belong to.

#### **Problem 11.4 (Information Retrieval)**

Let *D* be the set containing the following three texts:

- *d*<sub>1</sub>: Decision theory investigates decision problems: how an agent deals with choosing among actions.
- *d*<sub>2</sub>: Reinforcement learning is a type of unsupervised learning where an agent learns how to behave in an environment.

•  $d_3$ : *Information retrieval* deals with representing information objects. Let q be the query "agent action". 1. Give the list of words occurring in any of these texts and the word frequency tf(t, d), i.e., the number of occurrences of t in d divided by the length of d (measured in words), for each text d. Normalize all words so that inflection (plural, -ing, etc.) is ignored.

Solution: The order of the list does not matter as long as it is fixed. We use decision theory investigate problem how a agent deal with choose among action reinforcement learn be type of unsupervised where to behave in environment *information retrieval* represent object The word frequency vectors for the three texts are

2. For every word *t*, give the inverse document frequency idf(t, D).

3. For every word *t* and every document, give *tfidf*(*t*, *d*, *D*). Do the same for the query *q* "agent action".

Solution: tfidf(t, d, D) is obtained by multiplying tf(t, d) with 0.48 (if N(t) = 1) or 0.18 (if N(t) = 2) or 0 (if N(t) = 3), e.g.,

 $tfidf(\_, d_1, D)$ :  $\langle 0.96, 0.48, 0.48, 0.48, 0.48, 0.18, 0.18, 0.18, 0.18, 0.48, 0.48, 0.48, 0, ..., 0 \rangle / 13$ 

For the query, all values are 0 except for  $tfidf(agent, q, D) = 1/2 \cdot 0.18$  and  $tfidf(action, q, D) = 1/2 \cdot 0.48$ .

4. Compute the cosine similarity for q and each  $d_i$ .

Solution: Let  $A_i = tfidf(\_, d_i, D)$  and  $B = tfidf(\_, q, D)$ . We have  $A_1 \cdot B =$ 

 $(1 \cdot 1 \cdot 0.18^2 + 1 \cdot 1 \cdot 0.48^2)/(13 \cdot 2)$  and  $|A_1| = \sqrt{0.96^2 + 4 \cdot 0.18^2 + 7 \cdot 0.48^2}/13$ and  $|B| = \sqrt{0.18^2 + 0.48^2}/2 = 0.25$ . Then, we obtain  $\cos \theta_1 = A_1 \cdot B/(|A_1| \cdot |B|)$ .  $\cos \theta_2$  and  $\cos \theta_3$  are obtained accordingly.

5. How is the cosine similarity used to answer the query?

*Solution:* We return the document with the highest cosine similarity or return the list of documents ordered decreasingly by cosine similarity.