QA Session (SNLP tutorial)

Vilém Zouhar

21st July 2021

• How can we define it?

- How can we define it?
- Suppose that we generate a random sequence of digits S in the decimal system, where each digit in S is drawn from a uniform distribution. What is the perplexity of S in the following cases:

- How can we define it?
- Suppose that we generate a random sequence of digits S in the decimal system, where each digit in S is drawn from a uniform distribution. What is the perplexity of S in the following cases:

- How can we define it?
- Suppose that we generate a random sequence of digits S in the decimal system, where each digit in S is drawn from a uniform distribution. What is the perplexity of S in the following cases:
- S is a sequence of length 5

- How can we define it?
- Suppose that we generate a random sequence of digits S in the decimal system, where each digit in S is drawn from a uniform distribution. What is the perplexity of S in the following cases:
- S is a sequence of length 5

- How can we define it?
- Suppose that we generate a random sequence of digits S in the decimal system, where each digit in S is drawn from a uniform distribution. What is the perplexity of S in the following cases:
- S is a sequence of length 5
- S is a sequence of length 25

# Word Sense Disambiguation

• Formal definition of the problem

### Word Sense Disambiguation

- Formal definition of the problem
- Naive Bayes for WSD

### Word Sense Disambiguation

- Formal definition of the problem
- Naive Bayes for WSD
- Flip-Flop for clustering (???)

#### Conditional Random Fields

First and second-order HMM

#### Conditional Random Fields

- First and second-order HMM
- Suppose you have a sequence of length M and tagset of size T, what would be the complexity of the normalization factor Z(x) in this case?

# Naive Bayes

Naive Bayes pseudocode

### Naive Bayes

- Naive Bayes pseudocode
- Suppose you have a Naive Bayes classifier for topic categorization that is defined over a vocabulary of size V and a category label set of size C. How many parameters does this model have?

### Naive Bayes

- Naive Bayes pseudocode
- Suppose you have a Naive Bayes classifier for topic categorization that is defined over a vocabulary of size V and a category label set of size C. How many parameters does this model have?
- Exercise 12 from 2020 exam

### Naive Bayes - Code

```
for each class c \in C # Calculate P(c) terms
   N_{doc} = number of documents in D
   N_c = number of documents from D in class c
  logprior[c] \leftarrow log \frac{N_c}{N_{doc}}
   V \leftarrow vocabulary of D
   bigdoc[c] \leftarrow \mathbf{append}(d) for d \in D with class c
   for each word w in V # Calculate P(w|c) terms
     count(w,c) \leftarrow \# of occurrences of w in bigdoc[c]
     loglikelihood[w,c] \leftarrow log \frac{count(w,c) + 1}{\sum_{w' \text{ in } V} (count(w',c) + 1)}
return logprior, loglikelihood, V
```

MANCHESTER

The University of Mancheder

# Naïve Bayes

- · Algorithm: Continuous-valued Features
  - Numberless values for a feature
  - Conditional probability often modeled with the normal distribution

$$\hat{P}(X_j | C = c_i) = \frac{1}{\sqrt{2\pi}\sigma_{ji}} \exp\left(-\frac{(X_j - \mu_{ji})^2}{2\sigma_{ji}^2}\right)$$

 $\mu_{ji}$ : mean (avearage) of feature values  $X_j$  of examples for which  $C = c_i$  $\sigma_{ji}$ : standard deviation of feature values  $X_j$  of examples for which  $C = c_i$ 

- Learning Phase: for  $\mathbf{X} = (X_1, \dots, X_n)$ ,  $C = c_1, \dots, c_L$ Output:  $n \times n$  Output:  $P(C = c_i)$   $i = 1, \dots, L$
- Test Phase: Given an unknown instance  $\mathbf{X}' = (a'_1, \dots, a'_n)$ 
  - Instead of looking-up tables, calculate conditional probabilities with all the normal distributions achieved in the learning phrase
  - Apply the MAP rule to make a decision

# Significance Testing

• How to use Chi-Square?

### Compression

- Kraft's inequality and trees
- Optimal code length:  $-log_D p(w_i)$

#### Resources

- https://miro.medium.com/max/1400/1\*neaBooRXSloZAz6A2XfHJA.png
- https://image.slidesharecdn.com/naive-bayes-150514165844-lva1-app6891/95/naive-bayes-15-638.jpg?cb=1431622795