

Natural Language Processing

CSCI 4152/6509 — Lecture 12

P0 Topics Discussion (4); N-gram Model (start)

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Time and date: 16:05 – 17:25, 21-Oct-2024

Location: Carleton Tupper Building Theatre C

Previous Lecture

- Fully Independent Model (continued)
- Naïve Bayes classification model
 - ▶ Assumption, definition
 - ▶ Graphical representation
 - ▶ Spam detection example
 - ▶ Computational tasks
 - ▶ Number of parameters
 - ▶ pros and cons, additional notes
 - ▶ Bernoulli and Multinomial Naïve Bayes
- P0 Topics Discussion (3)

P0 Topics Discussion (4)

- Discussion of individual projects as proposed in P0 submissions
- Projects discussed: P-14, P-15, P-16, P-17, P-18, P-19, P-20, P-21, P-23, P-24

N-gram Model

- Let us first introduce *language modeling*
- *Language Modeling*: Estimating probability of arbitrary NL sentence: $P(\text{sentence})$
- Example: Speech recognition

$$\begin{aligned}\arg \max_{\text{sentence}} P(\text{sentence}|\text{sound}) &= \arg \max_{\text{sentence}} \frac{P(\text{sentence, sound})}{P(\text{sound})} \\ &= \arg \max_{\text{sentence}} P(\text{sentence, sound}) \\ &= \arg \max_{\text{sentence}} P(\text{sound}|\text{sentence})P(\text{sentence})\end{aligned}$$

- Acoustic model and Language model

Language Modeling

- Task of estimating probability of arbitrary utterance in a language
- Alternative task: Predicting the next token in a sequence: e.g., the next word or words, in a sentence, or next character or characters
- N-gram model: a “natural” model for this task

N-gram Model Assumption

$$P(w_1 w_2 \dots w_n) = P(w_1 | \cdot \cdot) P(w_2 | w_1 \cdot) P(w_3 | w_2 w_1) \dots P(w_n | w_{n-1} w_{n-2})$$

N-gram Model: Notes

- Reading: Chapter 4 of [JM]
- Use of log probabilities
 - ▶ similarly as in the Naïve Bayes model for text
- Graphical representation

