

Semantic Features for Dialogue Act Recognition

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Table of Content

- 1 Introduction
- 2 Approaches
- 3 Results
- 4 Conclusions & Perspectives

Definition of the Dialogue Acts (DAs)

- introduced by Austin in 1962 [Aus62]
- developed by Hary Bunt in [Bun94]
 - DA = meaning of an utterance in the context of a dialogue
- this work
 - DA = function of an utterance, or its part, in the dialogue

Example:

- question → requesting of some information
- answer → providing this information

Applications

- dialogue systems
- machine translation
- automatic speech recognition
- topic tracking
- talking head animation
- etc.

Objectives

- to propose semantic features and integrate them into a dialogue act recognition task to improve the recognition score in Czech
- three different feature computation approaches proposed, evaluated and compared:
 - Latent Dirichlet Allocation (LDA)
 - Hyperspace Analogue to Language (HAL)
 - Correlated Occurrence Analogue to Lexical Semantics (COALS)

Related Work

Features

- lexical [Jet *al.*97] (and syntactic [KC14])
- prosodic [SB98]
- dialogue history [Set *al.*00]
- **semantic** [KUX10] (few work × our focus)

Models

supervised machine learning

- Bayesian Networks [KRN02]
- Discriminative Dynamic BN [JB05]
- Maximum Entropy [ALS05]
- Conditional Random Fields [QIR11]
- Neural Networks [LLL⁺03]
- ...

Latent Dirichlet Allocation (LDA)

- unsupervised topic model \rightarrow a topic to each word in the sentence
- semantically close words \approx similar topics (e.g. synonyms)
- standard LDA model \rightarrow a sentence topic for each word
- used together with word labels for DA recognition

Semantic Spaces

- words \approx high dimensional semantic vectors
- semantically close words \rightarrow similar vectors
- opportunity to use a clustering method to create word clusters
- two semantic space models:
 - Hyperspace Analogue to Language (HAL [LB96])
 - Correlated Occurrence Analogue to Lexical Semantic (COALS [RGP04])
- additive composition of *word-level* vectors (by HAL or COALS methods) \rightarrow *sentence-level* semantic vectors
- = additional semantic information for DA recognition

Note

- never used for dialogue act recognition before

Some Assumptions about the Models

LDA

- full sentences as a context
- → long word dependencies
- → information about the topic of the conversation

Expectation

- *HAL and COALS will give better results than LDA*

HAL and COALS

- (relatively) short context window
- → local dependencies between words
- → syntactic structure information
- → important for DA recognition

Dialogue Act Recognition

- W .. sequence of n words w_i in the sentence
- F .. sequence of semantic features f_i ($i \in [1; n]$)
- C .. dialogue act class

Two classifiers used:

Naive Bayes [Ris01]

- sometimes also referred as an *unigram*
- modelling of $P(W|C)$ (*first baseline*)

Maximum Entropy (ME) [BPP96]

- modelling of $P(C|W)$ in lexical case (*second baseline*)
- $P(C|W, F)$ in semantic case

DA Corpus

- evaluation of our approach on both types of transcriptions: manual and automatic
- automatic transcription with the jLASER [PE07] recognizer
 - training on about nine hours (6234 sent.)
 - testing on about three hours (2173 sent.)

DA corpus			
DA	No.	Example	English translation
S	566	Chtěl bych jet do Písku.	I would like to go to Písek.
O	125	Najdi další vlak do Plzně!	Look at for the next train to Plzeň!
Q[y/n]	282	Řekl byste nám další spojení?	Do you say next connection?
Q	1200	Jak se dostanu do Šumperka?	How can I go to Šumperk?
Sent.	2173		

Tools & Parameters I.

LDA

- MALLET [McC02] tool-kit
- Dirichlet distributions parameters initially set to (see [GS04])
 - $\alpha = 50/K$ ($K = \text{topic number}$)
 - $\beta = 0.1$

HAL and COALS semantic space models

- S-Space package [JS10] for implementation
 - four-words context window in both directions
 - matrix composed of 1,000 columns
 - dimensionality reduction not used

Tools & Parameters II.

- LDA and both semantic spaces trained on the training part of the Railways corpus (i.e. 6234 sentences)
- Brainy [Kon14] tool-kit for implementation of Maximum Entropy classifier
- 10-fold cross-validation used (10% of the corpus for testing)
- confidence interval of $\pm 1\%$
- ASR Accuracy (ACC)
 - Sentence ACC = 39.78%
 - Word ACC = 83.36%

Results with Manual Word Transcription

Approach/ Classifier	Accuracy in [%]				
	S	O	Q[y/n]	Q	Glob.
1. Lexical information (baselines)					
NB	93.5	77.6	96.5	89.9	91.0
ME	90.3	88.0	97.2	96.5	94.6
2. Semantic information					
LDA + ME	93.3	87.2	96.5	98.5	96.4
HAL + ME	95.1	96.0	97.9	97.9	97.2
COALS + ME	96.1	97.6	99.3	99.2	98.4

Table: Dialogue acts recognition accuracy for different approaches/classifiers and their combination with manual word transcription

Results with Automatic Speech Recognition

Approach/ Classifier	Accuracy in [%]				
	S	O	Q[y/n]	Q	Glob.
1. Lexical information (baselines)					
NB	93.1	68.8	94.7	86.3	88.2
ME	87.5	77.6	89.7	95.2	91.6
2. Semantic information					
LDA + ME	88.3	80.8	89.0	96.3	92.5
HAL + ME	92.2	86.4	93.6	96.9	94.8
COALS + ME	95.9	96.8	97.5	99.0	98.0

Table: Dialogue acts recognition accuracy for different approaches/classifiers and their combination with word transcription by ASR

Conclusions & Perspectives

- three approaches to create semantic features proposed, implemented and evaluated on the Czech Railways corpus
- semantic features **important** for dialogue act recognition
- semantic spaces, HAL & COALS, significantly outperform the LDA model
- explanation: **semantic spaces** \approx modelling of local dependencies between words \times **LDA** \approx global word dependency

Perspectives

- adaptation and evaluation of the proposed methods on larger corpora and in other languages with more dialogue acts
- evaluation of the other classifiers

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