



Features for Modelling Characteristics of Conversations

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Identifying sexual groomers in conversations

What information to use?



What information to use?

- what content in conversations?



What information to use?

- what content in conversations?
- what behaviour within conversations?



What information to use?

- what content in conversations?
- what behaviour within conversations?
- what types of conversations?



Feature types



Feature types

Conversation content

words and short phrases



Feature types

Conversation content

words and short phrases

Conversation behaviour

turn-taking – the flow of interaction



Feature types

Conversation content

words and short phrases

Conversation behaviour

turn-taking – the flow of interaction

Conversation type

number of participants and conversation length



Author profile

Train a maximum entropy classifier with all features from all conversations the author took part in. Some exceptions:

- do not use features with occurrence with only one author.
- remove feature frequency information.



Extracting lexical features

Lexical features



Lexical features

- normalisation and tokenisation
- tokens and token bigrams
- bag of tokens + bigrams for every participant



Lexical features

- normalisation and tokenisation
- tokens and token bigrams
- bag of tokens + bigrams for every participant
- generalize mentions of chat nick-names of other conversation partakers to **OtherName**.
- generalize mentions of nick-name of the “speaker” to **SelfName**.

More than one source to the content of a conversation!

SLEX The things you utter in the conversation.

OLEX The things others utter in the conversation.



More than one source to the content of a conversation!

SLEX The things you utter in the conversation.

OLEX The things others utter in the conversation.

Add the two sets of features to every author profile.

In two ways: Concatenation (SLEX + OLEX) or union (CLEX).



Combination of lexical sources

Features types used	Precision	Recall	F ($\beta=1$)
submitted	0.84	0.89	0.87
SLEX	0.21	0.84	0.33
OLEX	0.27	0.80	0.35
OLEX+SLEX	0.43	0.85	0.57
CLEX	0.48	0.72	0.58
CLEX + CONV features	0.56	0.55	0.55



Extracting conversation features

Conversation type, CTYPE



Conversation type, CTYPE

Monologue

0 – 1 participant

Dialogue

2 participants

Group

> 2 participants



Extracting conversation features

Conversation length, LTYPE



Conversation length, LTYPE

Feature	# of utterances
fail	0 – 1
handshake	2 – 7
prelude	8 – 25
brief	26 – 50
discourse	51 – 100
ldiscourse	101 – 160
vldiscourse	161 –



Adding ctype to ltype

Example

prelude_{mono}, prelude_{dia}, or prelude_{group}

brief_{mono}, brief_{dia}, or brief_{group}

...



Turn-taking behaviour, TTAKE

The flow of utterances within a conversation is modelled by turn-taking trigrams from each participant's perspective.



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Example

	utterance 1	u2	u3	u4	u5	...
participants	A	A	B	A	B	...

Example

A: Self–Self–Other, Self–Other–Self, Other–Self–Other, ...

B: Other–Other–Self, Other–Self–Other, Self–Other–Self, ...



Extracting conversation features

Adding ctype to ttake



Adding ctype to ttake

Example

A: Self–Self–Other_g, Self–Other_g–Self, Other_g–Self–Other_g, ...

B: Other_g–Other_g–Self, Other_g–Self–Other_g, Self–Other_g–Self,

...



Combining it all

Features types used	Precision	Recall	F ($\beta=1$)
S-OLEX+CTYPE+LTYPE+TTAKE	0.84	0.89	0.87
CTYPE+LTYPE+TTAKE	0.68	0.04	0.06
S-OLEX+CTYPE+LTYPE	0.90	0.90	0.90
S-OLEX+LTYPE+TTAKE	0.95	0.93	0.94
S-OLEX+CTYPE+TTAKE	0.97	0.97	0.97



Take home

- it takes two to tango – use that!
- and even simplistic situational features helps!



Normalisation and tokenisation

All lower-case

- ① let all sequences of black characters delimited by white-space constitute a token.
- ② let all initial and final sequences of punctuation characters be a token of its own.
- ③ add extra tokens from (some) tokens with internal black-space:
URL → URL + URL parts.



Identifying flagrant utterances

- ① use the sexual predator classifier.
 - ② classify all predator utterances in training material using only SLEX features.
 - ③ rank the utterances by the probability for the utterance “being a sexual predator”.
 - ④ decide a flagrancy threshold.
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- ① classify all utterances in test material as above.
 - ② pick all utterances above threshold.