Statistical Affect Detection in Collaborative Chat

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Introduction

- To study the role that affect plays in distributed collaborative teams, we developed methods for detecting expression of affect in chat logs.
- Our data spans 4 years of Nearby Supernova Factory, an international astrophysics collaboration, involving 30 scientists studying supernovae sharing a telescope three nights a week.
- The goal is to label each message with affect expressed. We have labeled 5% of 300k messages manually; automation is necessary.

Each chat message must be classified as one or more of ~30 affect expression labels. The example below shows several anonymized messages, with labels by three human coders.

Time	Speaker	Mes	ssage +	labels	s by 3 d	coder	S		
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Results

The goal of this work is to make coding a large dataset for a set of interpretive affect expression labels more tractable using automation.

We created useful classifiers for the 13 most frequently-occurring labels in our own dataset. The features, shown below, are heavily weighted in SVM classifiers trained independently for each code.

In future work, we will consider graphical models to better incorporate message context during labeling.

Interest (F_1 =0.925)	Amusement (F_1 =0.734)	Agreement (F_1 =0.779)		
???? length	emoticon ";)"	"yes"		
# question marks	emoticon ":)"	"yeah"		
"je" (fr.) (-)	laughter	"yep"		
"sunrise"	emoticon ";-)"	msg. length		
"bert"	"fun"	segment duration		
"est" (fr.) (-)	laughter length	"right"		
"where"	"p"	"yup"		
"wonder"	# people names	"agree"		
"sunset"	"sleep"	"sure"		
"interesting"	"of"	"okay"		
Considering (F_1 =0.761) Confusion (F_1 =0.738)	Acceptance (F_1 =0.773)		
"think"	???? length	ok "ok		
# question marks	# question marks	"okay"		
"maybe"	"understand"	"ah"		
ellipsis length	"confus_"	msg. length		
"or"	"why"	# 1st sg. pronouns		
hmm length	"what"	"oh"		
# hmmm	"nothing"	"yep"		
???? length	"wrong"	# question marks		
"probably"	msg. length	"put"		
"X"	"thought"	segment duration		

 #1: Interest / anger #2: annoyance / confusion #3: interest / frustration 05:58:55 Alice was it the bright blob? #1: interest / anger #2: considering #3: interest 05:59:03 Ben 5876 absorption is much wider than the H alpha in 	
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space	
#1, #2, #3: no affect	
05:59:18 Ben Oh hmmn.	
#1, #2, #3: considering	
05:59:28 Ben Lemme see what [the] coordinates were	
#1, #2, #3: no affect	

However, chat messages have challenging characteristics: they are short; contain jargon; with non-standard spelling, structure, and punctuation. We must be able to detect both "what?" and "WHAAAAAAT"

Our Approach

different We tested combinations feature of extraction methods and machine learning algorithms, comparing precision and recall for each of 13 most frequent codes. retain message context, To despite shortness, we tried segmenting messages by time threshold, and using sliding windows in feature extration.

interest	4351		
amusement	3213		
considering	1763		
agreement	1623		
annoyance	1212		
confusion	1125		
acceptance	975		
apprehension	799		
frustration	541		
supportive	518		
surprise	464		
anticipation	426		
serenity	369		
13 most frequently-occurring codes			

Annoyance (F_1 =0.624)	Apprehension (F_1 =0.638)
# swearing	"bad"
"pascal"	"something"
"" (dash)	"problem"
"all"	"we"
"damn"	"seem"
"again"	"too"
" "	msg. length
"only"	"not"
"me"	# 3rd sg. Pronouns
msg. length	# swearing

Supportive $(F_{1}=0.626)$ Surprise $(F_{1}=0.71)$ Anticipation $(F_{1}=0.748)$

We used lexicons (eg, emoticons); character-counting features; regular expressions (eg, hm+) as well as unigram features.

In short chat messages, even character-level peculiarities can be
meaningful markers for affect expresion.

the telescope is stuck! >:(Punctuation + emoticon \rightarrow frustration.
the telescope is stuuuuuuuuck	Repetition \rightarrow annoyance.
the telescope is stuck??	Multiple question marks \rightarrow confusion

Jup 13C (1 j = 0.71)	Anticipation $(\Gamma_{j}=0.7\pm0)$
# exclamation pts.	"hope"
"wow"	"if"
msg. length	"next"
???? length	"should"
<pre>!!!! length</pre>	"think"
"oh"	"will"
ellipsis length	"try"
# repeated letters	"at"
segment duration	"like"
"right"	'to"
	# exclamation pts. "wow" msg. length ???? length !!!! length "oh" ellipsis length # repeated letters segment duration "right"

Serenity (<i>F</i> ₁ =0.663)	Frustration (F_1 =0.673)		
"good"	# swearing		
emoticon ":)"	# 1st sg. pronouns		
"nice"	msg. length		
"cool"	ellipsis length		
!!!! length	capital. length		
msg. length	chars/second		
"right"	<pre># negation words</pre>		
"too"	"it"		
# 1st pl. pronouns	# repeated letters		
"do" (-)	# interrogative prns		

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