

Analyzing Differences in Subjective Annotations by Participants and Third-party Annotators in Multimodal Dialogue Corpus

Kazunori Komatani, Ryu Takeda (SANKEN, Osaka University), Shogo Okada (JAIST)



Background and objectives

User's subjective impression is important

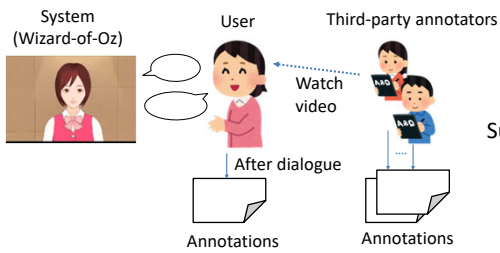
- The system should respond adaptively to it

Subjective impressions are inherently ambiguous

- not always agree among annotators (and with users themselves)

Analyses in multimodal dialogue corpus Hazumi

- Users themselves vs. third-party annotators
- Use of dispersion of third-party annotation results



Multimodal dialogue corpus Hazumi



Publicly available

- Movies: a written oath is required <https://www.nii.ac.jp/dsc/idr/rdata/Hazumi/>
- Annotations, feature files, etc <https://github.com/ouklab/Hazumi2010>, etc.

Version numbers: the year and month when the data collection started

Version	Recorded env.	No. of participants (dialogues)	No. of exchanges	Subjective annotations			
				Self-sentiment	Third-party sentiment	18 rapport items	Personality traits
Hazumi1712	In-person	29	2,422		○		
Hazumi1902		30	2,514	○	○	○	
Hazumi1911		30	2,859	○	○	○	
Hazumi2010	Online	33	2,798	○	○	○	○
Hazumi2012		63	5,334	○	○	○	○
Hazumi2105		29	2,235	○	○	○	○
Total		214	18,162				

Subjective Annotations

- Sentiment:** how much the user enjoyed the dialogue in the exchange (7-point scales)
 - 18 rapport items:** 18 questionnaire items about the dialogue (8-point scales)
 - Personality traits:** the user's Big Five traits via TIPI-J inventory (10 items on 7-point scales)
- Each was annotated by users themselves (self) and five third-party annotators

Analyses (I): Relationship between self- and third-party annotation results

① Personality traits

Correlation: users themselves vs. averages by third-party annotators

	E	A	C	N	O
Hazumi1911	0.53	0.08	0.43	0.25	0.29
Hazumi2010	0.58	-0.44	0.17	0.10	0.34
Hazumi2012	0.39	0.19	0.11	0.14	0.19
Hazumi2105	0.57	0.37	0.06	0.21	0.17
Total	0.49	0.06	0.16	0.15	0.21

$p = 8.4 \times 10^{-11}$ $p = 0.041$ $p = 0.0077$

- Extraversion (E): statistically significant correlations
- Consistent with a result in psychology [Bokenau+ 2009]
- Dialogue itself is an extroverted act
- A and N were not sufficiently expressed by users

E: Extraversion, A: Agreeableness, C: Conscientiousness, N: Neuroticism, O: Openness

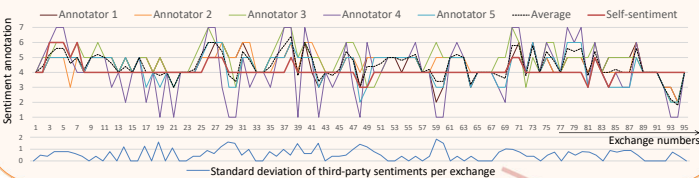
Bold: the correlation is statistically significant $p < 0.05$

③ Sentiment

Correlation between self-sentiments and averages of third-party sentiments

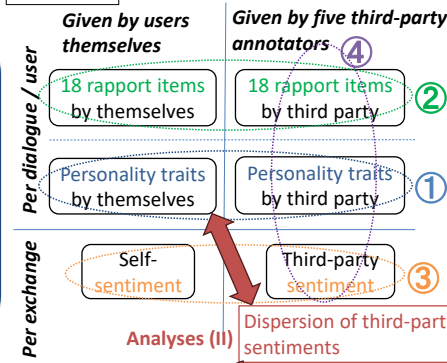
	Macro average	(max, min)
Hazumi1902	0.45	(0.69, 0.11)
Hazumi1911	0.41	(0.79, 0.01)

- Self-sentiments are not always expressed and perceived by the annotators
- Large individual differences among users



Standard deviation of third-party sentiments per exchange

Overview



Analyses (II) Dispersion of third-party sentiments
Negatively correlated
Machine learning performance of sentiment estimation

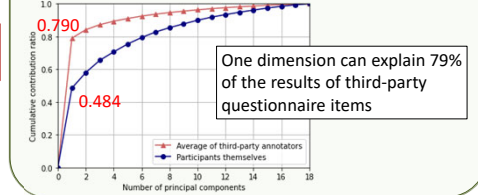
② Rapport 18 items

Correlation between self- and third-party annotations

5* Unsatisfying	0.38
9 Engrossing	0.35
2* Boring	0.32
17 Worthwhile	0.29
8* Awkward	0.27
16* Dull	0.25
10* Unfocused	0.23
6* Uncomfortably paced	0.23
1 Well-coordinated	0.22
12* Intense	0.21
11 Involving	0.21
14 Active	0.20
4 Harmonious	0.20
7* Cold	0.19
18* Slow	0.17
13 Friendly	0.13
15 Positive	0.09
3 Cooperative	0.07
Average of 18 items	0.34

* denotes inverted items

$p < 0.05$ Correlated significantly ($p = 1.23 \times 10^{-5}$)



One dimension can explain 79% of the results of third-party questionnaire items

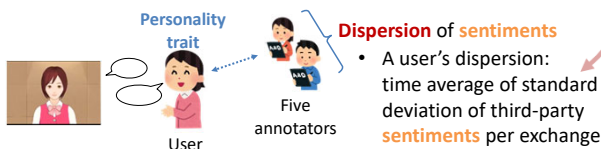
④ Relation among results by third-party annotators

	Personality traits					Average sentiments
	E	A	C	N	O	
Average 18 rapport items	0.53	0.68	0.21	-0.22	0.52	0.55
Average sentiments	0.21	0.30	0.12	0.00	0.36	

- Correlation among average sentiments, average 18 rapport items, and E, A, and O
- The third-party annotation results had some correlations, but the self-annotation results seem more complex because not all factors are expressed during dialogues.

Analyses (II): Use of dispersion of third-party sentiments

For what users can the estimation results be reliable?

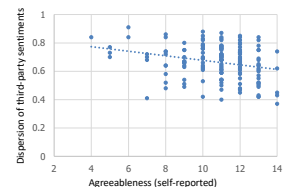


- Annotation results agree → higher machine learning performance
- Emotion recognition for children speech [D. Seppi+ 2008]
- Our preliminary results: regression errors and the dispersions correlated

Correlation between the dispersions and personality traits

	E	A	C	N	O
Hazumi1911	0.24	-0.44	0.16	0.12	0.27
Hazumi2010	0.38	-0.38	-0.15	0.11	0.04
Hazumi2012	-0.13	-0.20	0.00	0.08	-0.05
Hazumi2105	-0.20	-0.13	0.29	-0.04	0.03
Total	-0.05	-0.26	-0.05	0.03	-0.04

$p = 9.1 \times 10^{-4}$



- The **dispersions** were negatively correlated with Agreeableness (self-reported)
- More agreeable users had smaller **dispersion** of third-party **sentiments**
- Such users may express their sentiments in a way that others can perceive → Sentiment estimation results for agreeable users would be reliable