

Personality Recognition on Real Persona Chat Corpus

Internship Final Presentation (July-September)

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What's the personality?

The personality traits encompass enduring patterns of human *thoughts*, *feelings*, and *behaviors*, which make each different from one another.



Fig: https://www.verywellmind.com/what-is-personality-2795416

Why personality recognition in coversation?



What are the types of personality?

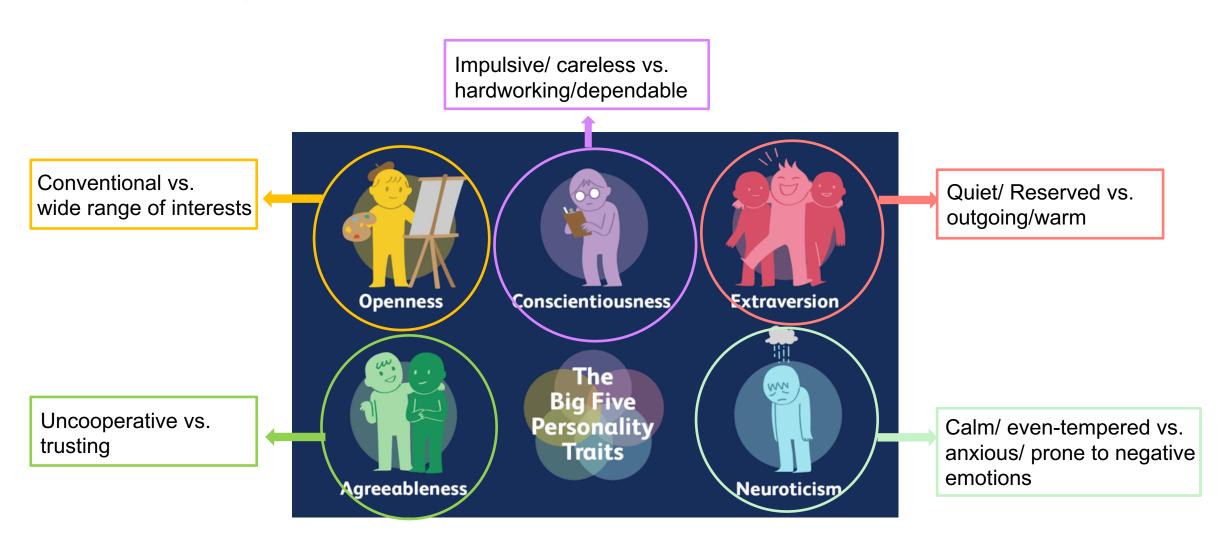


Fig: https://www.verywellmind.com/what-is-personality-2795416

How does personality be expressed in conversation?

Expression of personality is decided by person's inherent personality traits (monologue) and influenced by the interlocutor (dialogue). Neuroticism score:



A: 絵日記が一番つらかった! (the picture diary was the most challenging!)

B: 毎日書かなきゃいけないという。(you had to write it every day.)





Neuroticism: Easily upset/sad, nervous, *pessimistic*, neurotic

A 6.67

B 4.42



A: そうそう。それを溜め込んでしまうと、悲惨でした。(Yeah, that's right. It was *miserable* when I let it pile up.)

> B: 数日前のことですら覚えてないですもんね。(you can't even remember what happened a few days ago)





A: 毎日、朝ごはんに何を食べたとか、そういうことしか書かなかった気がしま す。(I think I only wrote about what I had for breakfast every day.)

> B: 朝ご飯を書いていたんですね、かわいらしい。(You wrote about breakfast, how cute.)





A: いやいや、ちゃんと毎日書くべきでした! (No no, I should have been writing every day properly!)

How does personality be expressed in conversation?

Expression of personality is decided by person's inherent personality traits (monologue) and influenced by the interlocutor (dialogue).



A: 我が家は今晩はアジの南蛮漬けでした! (We had marinated horse mackerel for dinner tonight!)

B: わあ、それ大好きです! 夏って感じしますね。(Wow, I **love** that! It feels like summer.)



Neuroticism score:

A 3.92

B 6.67



A: 今日は小アジを使ったので、丸揚げにしましたよ! (Today, I used small horse mackerel, so I fried them whole!

B: 良いですね、カルシウム満点! (*That's great*, full of calcium!)



Neuroticism: Easily upset/sad, nervous, pessimistic, neurotic



A: 骨まで食べれるのがいいですよねー。(Yes, it's nice to be able to eat the bones too!)

B: 羨ましい一。ご近所さんだったら、おすそ分けを奪いに行くのに一! (I'm *envious*! If you were my neighbor, I'd come over for a taste!)



Experiments

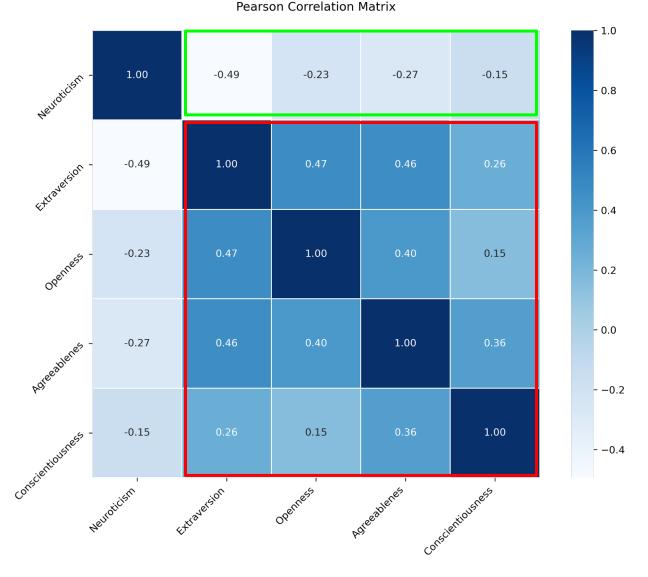
Corpus: RealPersonaChat

No. of dialogues	14,000
Dialogue length	28-49 utts. (Avg. 30.09)
No. of utterances	421,203
Utterance length	1-124 chars. (Avg. 22.92)
No. of participants (personas)	232
Language	Japanese
Personality traits	Neuroticism, Extraversion, Openness, Agreeableness, Conscientiousness (scale of 1-7)

Experiments

Big-5 Correlation





For all the correlation sets, *p-value*<0.05

- Neuroticism and the other four personalities show negative relations.
- Extraversion, Openness, Agreeableness, and Conscientiousness, each two sets show positive relation.

Experiments on Monologue

Threshold of low/high personality traits

Average of the train label, fine-tuning on rinna/Japanese-roberta-base.

	train (0:1)	valid (0:1)	test (0:1)	Acc. (test)	BA (test)
Neuroticism (4.29)	11605/10795	1220/1580	1555/1245	69.07	68.77
Extraversion (4.56)	11625/10775	1274/1526	1655/1145	66.71	67.85
Openness (4.53)	10135/12265	1050/1750	1750/1050	68.36	70.29
Agreeableness (4.87)	10525/11875	1120/ 1680	1545/1255	60.11	61.81
Conscientiousness (4.21)	11410/10990	1520/1280	1525/1275	64.14	64.09
Average				65.68	66.56

Threshold of low/high personality traits

Median of the train label, fine-tuning on rinna/Japanese-roberta-base.

	train (0:1)	valid (0:1)	test (0:1)	Acc. (test)	BA (test)
Neuroticism (4.25)	11195/11205	1115/1685	1545/1255	67.61	68.83
Extraversion (4.5)	10695/11705	1199/1601	1520/1280	64.61	64.50
Openness (4.75)	11110/11290	1105/1695	1810/990	74.57	72.78
Agreeableness (5.0)	11155/11245	1165/1635	1685/1115	67.18	65.59
Conscientiousness (4.17)	10775/11625	1379/1421	1365/1435	68.82	68.72
Average				68.56	68.08

Single vs multi-tasks

- Single-task: separately training five personality tasks.
- ➤ Multi-task: jointly training five personality tasks.

	Single-task		Mult	i-task
	Acc.	BAcc.	Acc.	BAcc.
Neuroticism (4.25)	67.61	68.83	76.11	74.99
Extraversion (4.5)	64.61	64.50	75.68	76.73
Openness (4.75)	74.57	72.78	78.29	80.21
Agreeableness (5.0)	67.18	65.59	72.68	73.16
Conscientiousness (4.17)	68.82	68.72	73.79	73.52
Average	68.56	68.08	75.31	75.72

Results on different pre-trained models

	Ac	C.	ВА	CC.
Models	Lr=1e-5	Lr=2e-5	Lr=1e-5	Lr=2e-5
rinna/roberta-base	-	75.31	-	75.72
microsoft/mdeberta-v3-base	74.25	43.39	74.83	0.5
xlm-roberta-base	75.19	42.89	75.21	0.5
xlm-roberta-large	78.16	-	78.17	-
studio-ousia/luke-japanese-base	80.18	-	79.60	-
studio-ousia/luke-japanese-large	79.98	-	79.76	-
nlp-waseda/roberta-base-japanese	47.44	-	49.98	-
nlp-waseda/roberta-large-japanese	76.37	-	77.01	-

Experiments on big-5 regression

Jointly train five tasks: Normalize the label to (0,1) last_hidden_size (768)-->512 (relu)-->256 (relu)-->128 (relu)-->1

	rinna/ roberta-base				
	MSE	Pearsonr (p-value)	Spearmanr (p-value)		
Neuroticism (4.25)	.0034	.3056 (.2498)	.3383 (.1999)		
Extraversion (4.5)	.0298	.1335 (.6222)	.2197 (.4137)		
Openness (4.75)	.0152	.1966 (.4656)	.3051 (.2506)		
Agreeableness (5.0)	.0065	.1244 (.6461)	.2197 (.4137)		
Conscientiousness (4.17)	.0048	.4867 (.0559)	.5125 (.0424)		

Experiments on big-5 regression

Jointly train five tasks: Normalize the label to (0,1) last_hidden_size (768)-->512 (relu)-->256 (relu)-->128 (relu)-->1

	studio-ousia/luke-japanese-base				
	MSE	Pearsonr (p-value)	Spearmanr (p-value)		
Neuroticism (4.25)	.0060	0539 (.8428)	.0208 (.9389)		
Extraversion (4.5)	.0260	.1434 (.5962)	.1708 (.5270)		
Openness (4.75)	.0216	.3789 (.1478)	.4027 (.1220)		
Agreeableness (5.0)	.0079	.3571 (.1745)	.4698 (.0663)		
Conscientiousness (4.17)	.0047	.6814 (.0037)	.8024 (.0002)		

Experiments on Dialogue

Proposed Methods

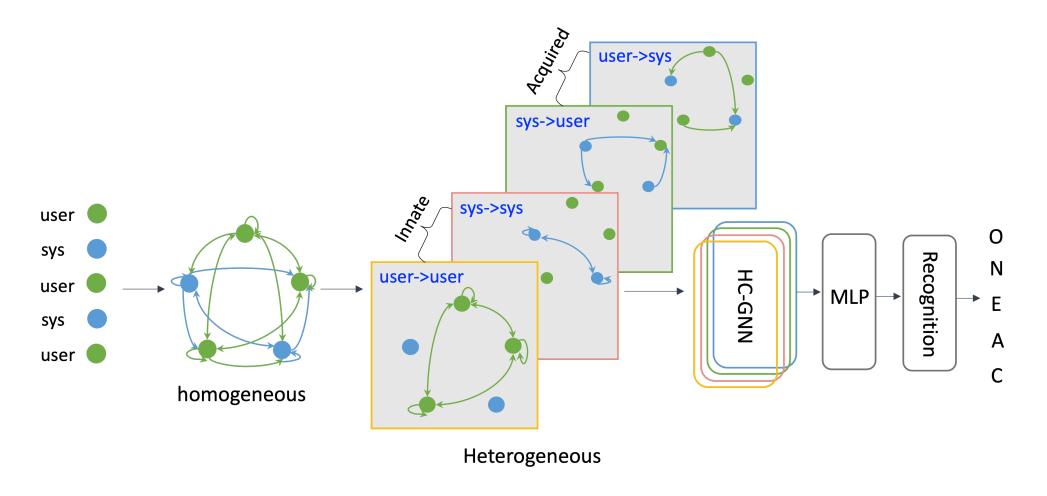


Fig. Personality Recognition by Heterogeneous Conversation Modeling Using Graph Representation. HC-GNN includes GATv2 and GCN layers.

Results on dialogue

	Acc.	BAcc.
Temporal (luke-japanese-base)	66.91	66.49
Spatial (luke-japanese-base)	57.67	57.11
Spatial (HC-GNN) (1 relation)	67.34	66.18
Spatial (HC-GNN) (2 relations)	67.66	66.88
Spatial (HC-GNN) (4 relations)	73.63	72.83

Modules in HC-GNN

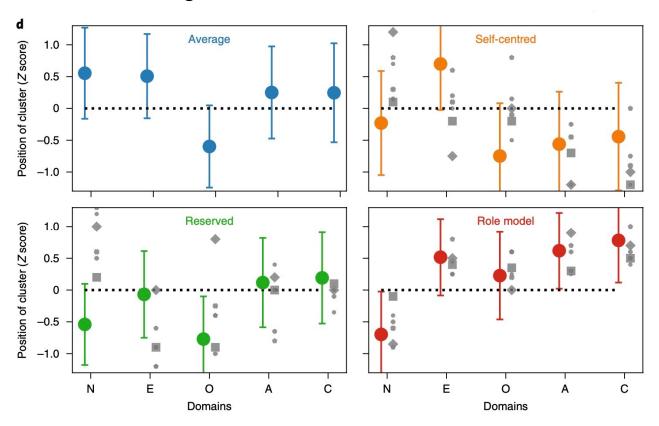
	Acc.	BAcc.
Spatial (GCN) [Kipf, T. N., & Welling, M. (2016)]	62.30	61.38
Spatial (GATv2, head=2) [Brody et al. 2021]	66.75	66.61
Spatial (HC-GNN) (Ours)	73.63	72.83

Results on monologue+dialogue

	Acc.	BAcc.
Monologue (temporal)	80.18	79.60
Dialogue (spatial, HC-GNN)	73.63	72.83
Monologue (temporal)+ Dialogue (spatial, HC-GNN)	78.13	77.65

Character = a combination of big-five personality traits. How to define?

Martin et al. (2018) clustered person's personality into "Role model", "Reserverd", "Self-centered", "Average" based on the score on big five traits.



Character = a combination of big-five personality traits. How to define?

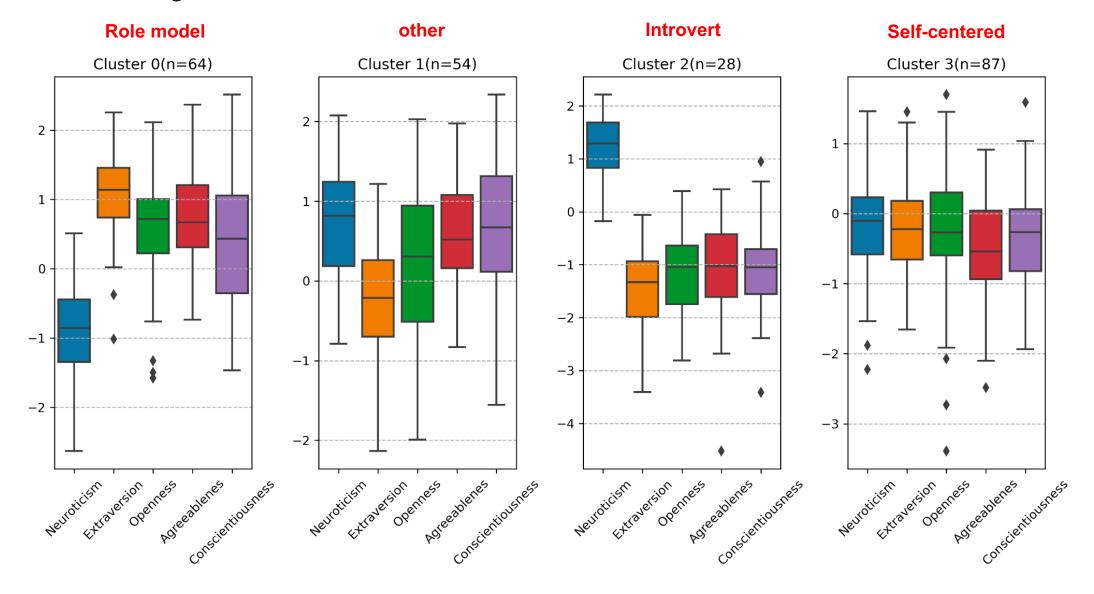
> Yamamoto et al. (2023) clustered person's personality into "Role model", "Reserverd", "Self-centered", "Introvert" based on the score on big five traits.

Table 2 Character and personality classes

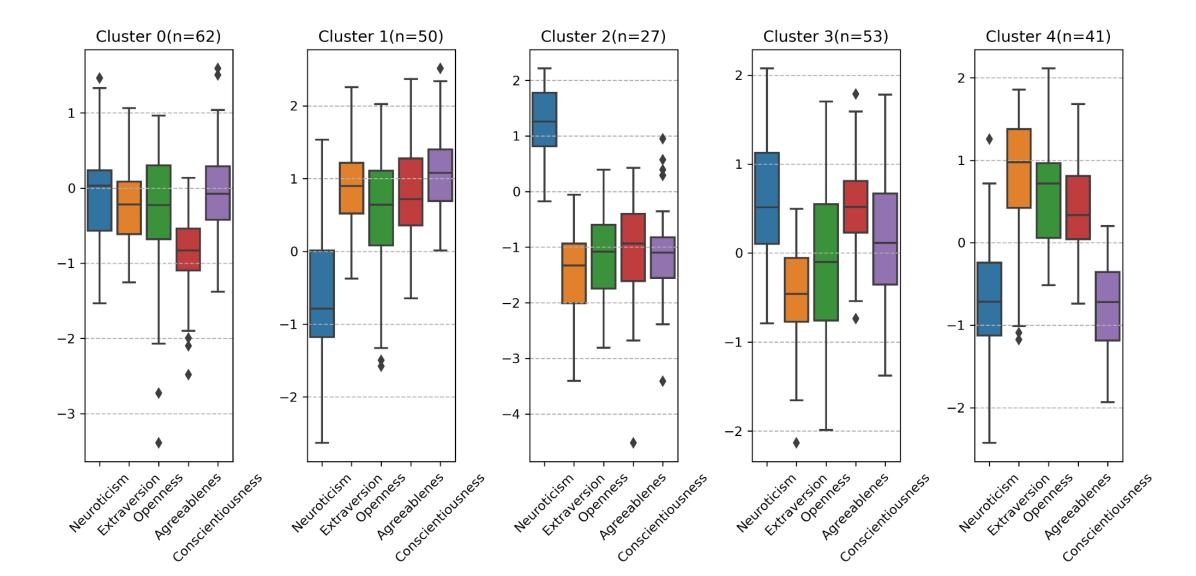
Class	User	System		Big	g Five tra	aits	
name	personality	Character	Em	Ex	Op	Ag	Co
Role model	0	0	Low	High	High	High	High
Reserved	0	0	Low	Low	Low	High	High
Self-centered	0		High	High	High	Low	Low
Introvert	0	0	High	Low	Low	Low	Low
Neutral (Baseline)		0	Middle	Middle	Middle	Middle	Middle

Source: Yamamoto et al. (2023)

- > Data preprocessing: Normalized the big5 scores using the mean and standard deviation of each score.
- Clustering: K-means++, 4 classes.



- > Data preprocessing: Normalized the big5 scores using the mean and standard deviation of each score.
- Clustering: K-means++, 5 classes.



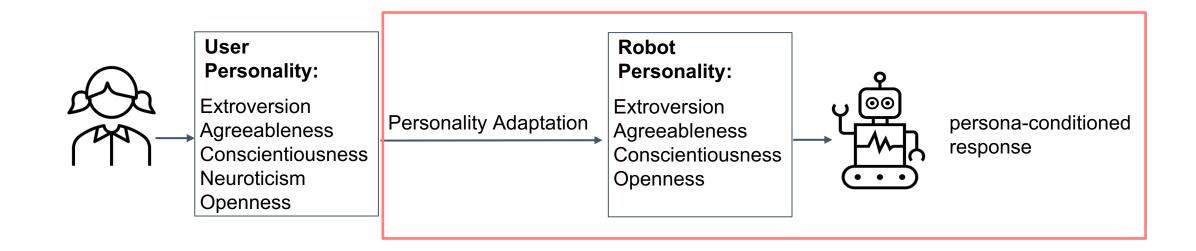
Methods		4-classes		5-classes	
		Acc.	BAcc.	Acc.	BAcc.
	Temporal	65.11	62.23	60.50	58.36
rinna/roberta-base	Spatial	58.05	56.18	56.92	55.34
	Temporal+Spatial	-	-	-	-
studio-ousia/luke-japanese-base	Temporal	63.18	62.87	62.36	59.45
	Spatial	56.53	57.28	51.44	48.07

Conclusions

- 1. Big Five personality traits exhibit correlations with each other, which makes them effective for coarse-grained low/high classification tasks. Although this may not hold true for fine-grained regression task.
- 2. For personality recognition, *monologues carry more weight than dialogues*. This means that the context provided by an interlocutor has a limited influence on a person's personality.
- 3. In the context of personality recognition during dialogues, our *proposed HC-GNN* (spatial-based) method outperforms the temporal-based approach.
- 4. Classifying a user's character remains a challenging task, and there is a need to explore precise methods for clustering the Big Five personality traits into distinct character classes.

Future Work

- 1. Better method to fuse spatial and temporal features.
- 2. Testing HC-GNN on monologue.
- 3. Personality adaptable system, specific to user's personality and needs (empathy, sympathy etc.)





Thanks for your attention!

Q&A

