

Natural Language Processing Project Phase 1

Data Generation for Personality Detection

Babak Behkamkia (98521099)

Dr. Sauleh Eetemadi

-

Contents

1	Introduction	3
2	Related Datasets	3
3	Collection	3
4	Cleaning	3
5	Generation	4
6	Statistics 6.1 Basic statistics 6.1.1 raw data 6.1.2 clean data 6.1.2 volume 6.2 Number of unique common words 6.1.2 volume	5 5 6 6
	6.3 Number of uncommon unique words 6.4 Most frequent uncommon words 6.5 Reletive Normalized Frequency 6.6 TF_IDF 6.7 Most appeared unique words	7 7 7 8 9
7	Scripts	9

1 Introduction

Nowadays, a huge number of different products are produced worldwide. Thus, people have more options, leading producers to consider increasing their sales. Thus, they tried to determine each individual's personality and advertise the products customized to their personality type. Collecting such data is challenging for researchers because many people don't know their personality type. Thus, researchers tried to extract the personality of people from their tweets. There are other problems because they cannot determine people's personalities just by reading a tweet. Therefore, they are dependent on people again to determine their own personality. With data generation, we hope to overcome these problems. I put my code for this project on my GitHub¹ and the gathered dataset on my Huggingface².

2 Related Datasets

There is a ready dataset for this task on Kaggle (Myers-Briggs Personality Type Dataset³). This dataset consists of more than 8600 samples. Each row of this dataset presents a person's personality type and his/hers most recent 50 posts. I used this dataset in the generation phase.

3 Collection

I used TweeePy and Twitter API to crawl for tweets that contain a personality keyword like "INFJ". I repeated this procedure for all 16 personality types in MBTI and labeled them with the corresponding keyword.

I created a dataframe for each label and saved it in CSV format. These dataframes consist of 5 columns:

- · author name
- author screen_name
- author id
- post
- author personality type

4 Cleaning

In this section, I describe the cleaning methods I have used to make the raw texts of tweets more processable data.

¹https://github.com/Babakbehkamkia/Data-Generation-for-Personality-Detection

 $^{^2 {\}rm https://hugging face.co/datasets/Babak-Behkamkia/Personality}_D et ection$

 $^{^3}$ https://www.kaggle.com/datasets/datasnaek/mbti-type

- Necessary cleaning: Removing special characters, stop words, and Lemmatizing the words are done in this part of cleaning.
- Removing Hashtags
- Removing Links: Links do not have a semantic meaning. Thus, their presents will not improve the model.
- Removing Usernames: A name by itself doesn't contain any information. Therefore, I remove usernames.
- Removing Retweet Sign: All retweets have the string "RT" in their beginning, which is unnecessary.
- Replacing Emojis: In many related works, researchers tend to delete the emojis. However, they contain a lot of information, and even in some cases, they can change the sentence's meaning. In this work, I replaced all emojis with their corresponding text encode by using the demoji⁴ library in Python.

In the end, I saved the cleaned text in a new column named "cleaned_text" in the same dataframe.

5 Generation

As told earlier, collecting and annotating data for personality detection is a challenging task. Thus, I tried to generate new samples for each personality type by using GPT-3 API. I created a prompt that contains the personality, a predefined topic (e.g. depression), and an example post from Myers-Briggs Personality Type Dataset with the given personality. With this prompt, I generated some posts toward the given topic and labeled them with the corresponding personality type. Moreover, in some generated samples, the personality type is mentioned directly. Therefore, I used the [PT] token instead of the personality type.

The generated dataset is saved as a CSV file in the "data/raw" directory. Each file has three columns:

- generated post
- topic of the post
- personality type

⁴https://pypi.org/project/demoji/

6 Statistics

I used nltk library for the following:

 \bullet $\mbox{nltk.tokenize.TweetTokenizer:}$ tokenizing the words

 $\bullet \ \ \, \mathbf{nltk.corpus.stopwords} \colon \ \, \mathbf{removing} \ \, \mathbf{stopwords} \\$

- ${\bf nltk.stem.WordNetLemmatizer}$: Lemmatizing the words

 $\bullet \;\; {\bf nltk.word_tokenize} : \; {\bf breaking \; samples \; to \; words}$

• nltk.sent_tokenize: breaking samples to sentences

6.1 Basic statistics

6.1.1 raw data

	sample num	sentence num	word num	unique word num
ENFP	3142	5758	81077	8107
ENFJ	1205	1700	32802	2842
ESTJ	764	1277	27318	952
ESFJ	494	1486	15653	1345
ESTP	580	995	16953	1523
ESFP	361	856	10176	1554
ENTJ	426	634	10552	2142
ENTP	1419	2442	31964	4725
INFP	5423	10094	145891	11956
INTP	3649	5914	92065	9511
INFJ	1000	1590	25460	3642
INTJ	2048	3387	49572	6986
ISFP	1000	1895	28085	2907
ISTP	1000	1955	28596	2358
ISFJ	1000	1930	29031	2390
ISTJ	1000	1769	28896	2375

6.1.2 clean data

	,		,	. ,
	sample num	sentence num	word num	unique word num
ENFP	3142	3139	34786	4461
ENFJ	1205	1183	14100	1635
ESTJ	764	764	11513	579
ESFJ	494	494	6503	824
ESTP	580	579	7861	923
ESFP	361	361	4266	963
ENTJ	426	426	4543	1309
ENTP	1419	1406	12943	2725
INFP	5423	5419	59540	5890
INTP	3649	3646	38042	4857
INFJ	1000	1000	9901	2165
INTJ	2048	2041	20236	3795
ISFP	1000	1000	11235	1599
ISTP	1000	1000	12310	1341
ISFJ	1000	999	12090	1404
ISTJ	1000	1000	12386	1412

6.2 Number of unique common words

	ENFP	ENFJ	ESTJ	ESFJ	ESTP	ESFP	ENTJ	ENTP	INFP	INTP	INFJ	INTJ	ISFP	ISTP	ISFJ	ISTJ
ENFP		907	365	532	571	616	774	1279	2061	1842	1084	1536	883	723	793	812
ENFJ	907		264	389	397	457	549	752	1030	933	663	843	572	478	537	539
ESTJ	365	264		191	192	209	271	295	396	365	279	341	239	209	234	251
ESFJ	532	389	191		255	273	305	415	565	526	388	489	365	302	350	340
ESTP	571	397	192	255		309	356	485	616	598	418	523	388	347	341	360
ESFP	616	457	209	273	309		403	550	667	617	456	568	423	380	388	387
ENTJ	774	549	271	305	356	403		691	888	840	571	780	478	419	444	472
ENTP	1279	752	295	415	485	550	691		1459	1400	865	1221	717	603	642	644
INFP	2061	1030	396	565	616	667	888	1459		2262	1259	1837	996	846	868	912
INTP	1842	933	365	526	598	617	840	1400	2262		1146	1750	931	778	812	835
INFJ	1084	663	279	388	418	456	571	865	1259	1146		1063	640	555	609	597
INTJ	1536	843	341	489	523	568	780	1221	1837	1750	1063		797	699	759	775
ISFP	883	572	239	365	388	423	478	717	996	931	640	797		513	505	525
ISTP	723	478	209	302	347	380	419	603	846	778	555	699	513		464	473
ISFJ	793	537	234	350	341	388	444	642	868	812	609	759	505	464		505
ISTJ	812	539	251	340	360	387	472	644	912	835	597	775	525	473	505	

6.3 Number of uncommon unique words

	ENFP	ENFJ	ESTJ	ESFJ	ESTP	ESFP	ENTJ	ENTP	INFP	INTP	INFJ	INTJ	ISFP	ISTP	ISFJ	ISTJ
ENFP		4282	4310	4221	4242	4192	4222	4628	6229	5634	4458	5184	4294	4356	4279	4249
ENFJ	4282		1686	1681	1764	1684	1846	2856	5465	4626	2474	3744	2090	2020	1965	1969
ESTJ	4310	1686		1021	1118	1124	1346	2714	5677	4706	2186	3692	1700	1502	1515	1489
ESFJ	4221	1681	1021		1237	1241	1523	2719	5584	4629	2213	3641	1693	1561	1528	1556
ESTP	4242	1764	1118	1237		1268	1520	2678	5581	4584	2252	3672	1746	1570	1645	1615
ESFP	4192	1684	1124	1241	1268		1466	2588	5519	4586	2216	3622	1716	1544	1591	1601
ENTJ	4222	1846	1346	1523	1520	1466		2652	5423	4486	2332	3544	1952	1812	1825	1777
ENTP	4628	2856	2714	2719	2678	2588	2652		5697	4782	3160	4078	2890	2860	2845	2849
INFP	6229	5465	5677	5584	5581	5519	5423	5697		6223	5537	6011	5497	5539	5558	5478
INTP	5634	4626	4706	4629	4584	4586	4486	4782	6223		4730	5152	4594	4642	4637	4599
INFJ	4458	2474	2186	2213	2252	2216	2332	3160	5537	4730		3834	2484	2396	2351	2383
INTJ	5184	3744	3692	3641	3672	3622	3544	4078	6011	5152	3834		3800	3738	3681	3657
ISFP	4294	2090	1700	1693	1746	1716	1952	2890	5497	4594	2484	3800		1914	1993	1961
ISTP	4356	2020	1502	1561	1570	1544	1812	2860	5539	4642	2396	3738	1914		1817	1807
ISFJ	4279	1965	1515	1528	1645	1591	1825	2845	5558	4637	2351	3681	1993	1817		1806
ISTJ	4249	1969	1489	1556	1615	1601	1777	2849	5478	4599	2383	3657	1961	1807	1806	

6.4 Most frequent uncommon words

According to the number of labels (16), I can only show part of the table in this section. There is a total of 120 (choosing 2 out of 16) possible combinations for the rows of this table.

	word 0	word 1	word 2	word 3	word 4	word 5	word 6	word 7	word 8	word 9
ENFP/ENFJ	sinb	vote	virgo	yangyang	literal	wayv	faceewe	sauropod		
ENFJ/ENFP	gold	unwavering	desire	empathetic	headbandage	skzsvt	gel	eya	ford	teddy
ENFP/ESTJ	enfp	enfj	hi	junhoe	donghyuk	intp	ikon	passport	jinhwan	yunhyeong
ESTJ/ENFP	exol	estj	lucky	bgm	gt_lt	mama	exo	baekhyuns	050623	kyoong
ENFP/ESFJ	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	w	1
ESFJ/ENFP	demiboy	gepard	izzie	gold	serving	nurturing	jia		jianna	kye
ENFP/ESTP	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	idol	game
ESTP/ENFP	jung	ahyeon	april	2007	position	rapper	kaden	bg	lyon	army
ENFP/ESFP	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	idol	sinb
ESFP/ENFP	sean	herb	04	kep	1er	lsfm	nmixx	artistic	rinn	hsrgenshin
ENFP/ENTJ	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	sinb	helloo
ENTJ/ENFP	entjartist	brain	brawn	table	savvy	jihoons	doyoung	entjs	style	opportunist
ENFP/ENTP	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	idol	sinb
ENTP/ENFP	youngk	xh	uid	position	charge	heitthey	rara	8540672	93	endless
ENFP/INFP	junhoe	donghyuk	ikon	passport	jinhwan	yunhyeong	chanwoo	sinb	preseason	baekho
INFP/ENFP	weather	sserafim		reii	emotion	butterfly		230510	743pm	tag
ENFP/INTP	sinb	sam	yangyang	wayv	faceewe	sauropod	preseason	option	baekho	yeonjun
INTP/ENFP	lizzie	stanlist	lina	int	moacaratfearnot	sserafim	gaby	baemon	blackpink	twic
ENFP/INFJ	junhoe	isfp	donghyuk	ikon	passport	jinhwan	yunhyeong	bobby	chanwoo	sinb
INFJ/ENFP	jungwoo	doyoung	jaehyun	reset	kofi	bffs	mae	loona	taeyang	bust

6.5 Reletive Normalized Frequency

Due to the same problems as the above section, I only show the first 20 rows of this table.

	word 0	word 1	word 2	word 3	word 4	word 5	word 6	word 7	word 8	word 9
ENFP/ENFJ	sam	another	ten	idol	v	yet	3rd	game	ask	telling
ENFP/ESTJ	sheher	interactive	cry	game	amp	e	according	virgo	another	v
ENFP/ESFJ	enfp	according	game	there	cat	hello	intp	option	yet	havent
ENFP/ESTP	enfp	isfp	according	w	people	sinb	there	many	enfj	first
ENFP/ESFP	according	enfp	enfj	another	many	time	whats	tear	intp	qrt
ENFP/ENTJ	enfp	according	idol	game	minor	vote	isfp	many	active	v
ENFP/ENTP	helloo	vote	according	enfp	game	telling	isfp	cat	virgo	others
ENFP/INFP	game	ceo	ht	sam	rabbit	standard			enfp	headphone
ENFP/INTP	ten			3rd	ceo	ht	helloo	acc	mbtis	haii
ENFP/INFJ	enfp	according	virgo	idol	pronoun	enfj	intp	havent	3rd	game
ENFP/INTJ	bobby	enfp	chanwoo	active	option	3rd	ht	voting	member	lst
ENFP/ISFP	helloo	05	please	idol	pronoun	enfp	option	yet	yeonjun	dream
ENFP/ISTP	enfp	enfj	vote	please	according	name	w	many	today	said
ENFP/ISFJ	enfp	w	enfj	according	intj	ten	05	yet	qrt	full
ENFP/ISTJ	wayv	please	idol	option	yet	flag	1	mainly	there	acc
ENFJ/ESTJ	sheher	guy	according	interactive	born	leader	boy	let	help	cry
ENFJ/ESFJ	according	guy	enfj	scorpio	intp	enfp	lee	hello	isfp	born
ENFJ/ESTP	enfj	isfp	according	leader	guy	enfp	help	heart	born	scorpio
ENFJ/ESFP	enfj	according	intp	leader	boy	guy		15	gel	enfp
ENFJ/ENTJ	according	minor	enfj	isfp	sheher	name	others	enfp	help	3

6.6 TF_IDF

I used **TfidfVectorizer** from **sklearn** in order to compute TF_IDF.

	ENFP	ENFJ	ESTJ	ESFJ	ESTP	ESFP	ENTJ	ENTP	INFP	INTP	INFJ	INTJ	ISFP	ISTP	ISFJ	ISTJ
0	day	isfp	puppy	skx	dancer	04	new	93	philippine	jinhwan	new	ten	isf	amp	edp	admins
1	sam	nj	spade	nurturing	position	If	put	rara	stans	junhoe	moot	new	cains	hi	fact	according
2	sinb	filo	auce	hehim	mbti	creativity	become	rice	avt	yunhyoong	looking	infj	genshinpjsk	blackpink	fun	isíp
3	interactive	ribbon	number	finding	south	newjeans	like	heitthey	weather	restarting	mbti	personalityloudly	day	ber	know	enfj
-4	bobby	hi	6000	sheher	rapper	looking	add	found	helio	enfp	bffs	literal	jelli	sheher	day	moot
5	sheher	looking	page	main	moot	itzy	face	add	put	lina	amp	orv	active	fandom	moot	intp
6	face	sheher	gt lt	future	11th	artistic	character	character	amp	88	reset	face	jeongin	hot	dive	looking
7	chanwoo	mbti	always	new	april	mbti	amp	friend	fundom	hi	kofi	hi	15	atenjay	new	chant
8	yunhyoong	bobby	backhyuns	looking	looking	1er	jihoons	hi	day	interactive	esfp	wayv	trixx	welcomed	preferably	bobby
9	jinhwan	moot	kyoong	interactive	bg	acspa	bringing	uid	add	stan	fun	put	Itit	castals	sheher	im
10	passport	chanwoo	050623	14	plece	staye	intj	new	diameter	lissele	fact	sauropod	sky	loseraf	looking	chanwoo
11	ikon	ikon	mbti	auco	im	kep	brain	looking	hi	new	know	facerwe	hi	mbti	extrovert	ikon
12	donghyuk	junhoe	today	im	train	hethey	palette	xh	sheher	day	extrovert	yangyang	new	yori	thai	yunkyeong
13	junhoe	jinhwan	thank	izzie	korea	nmixx	table	young	interactive	stanlist	iefj	ndd	mbti	beverage	unit	pasoport
14	moot	donghyuk	mama	suit	jung	moot	brawn	moot	new	sheher	im	moot	sheher	interactive	апапра	donghyuk
15	looking	passport	gt _i t	demilooy	lyon	lsfm	жичу	mbti	mbti	mbti	unit	character	interactive	DAW.	doyoung	junhoe
16	new	yunhyeong	bgm	gepard	kaden	im	entjartist	guon	moot	moot	doyoung	looking	moot	moot	jachyun	jinhwan
17	mbti	enfp	estj	heart	2007	scan	mbti	youngk	looking	looking	jackyun	mbti	looking	looking	im	enfp
18	im	im	Incky	moot	ahyeon	herb	im	im	im	im	infj	im	im	im	jungwoo	mbti
19	enfp	enfj	exal	esfj	estp	esfp	entj	entp	infp	intp	jungwoo	intj	isfp	istp	idj	istj

6.7 Most appeared unique words

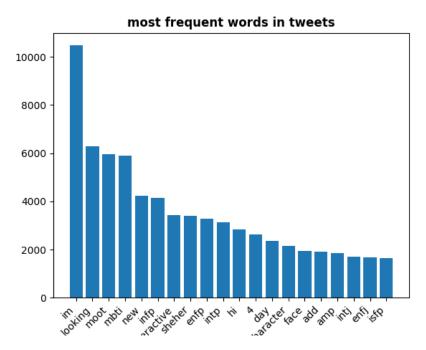


Figure 1: most Frequent Words

7 Scripts

For this work, bash and batch scripts are provided to run each part or the entire project. The necessary instructions are inside those files.