# **MBTI Classifier**

Arnav Joshi, Kevin Andrews

### The 16 MBTI Personalities

## What's Your Personality Type?

Use the questions on the outside of the chart to determine the four letters of your Myers-Briggs type. For each pair of letters, choose the side that seems most natural to you, even if you don't agree with every description.

#### 1. Are you outwardly or inwardly focused? If you:

- · Could be described as talkative, outgoing
- environment
- others, think out loud
- · Enjoy being the center of attention

 Could be described as reserved, private

Prefer a slower pace with time for contemplation · Tend to think things

through inside your head Would rather observe than

then you prefer

Introversion

Responsible, sincere. Warm, considerate, analytical, reserved, realistic, systematic pragmatic, thorough Hardworking and levoted caretakers who rustworthy with soun enjoy being helpful to practical judgment

INFJ Idealistic, organized, insightful, dependable compassionate, gentle

Seek harmony and

strategic, logical, reserved, insightful. Driven by their own

novative, independent riginal ideas to achieve

Intellectual, logical,

precise, reserved,

flexible, imaginative.

Original thinkers who

enjoy speculation and

creative problem solving.

impersonal way, using logical reasoning · Value justice, fairness

· Enjoy finding the flaws in an argument Could be described as

· Make decisions in an

reasonable, level-headed then you prefer

Thinking

 Base your decisions on personal values and how your actions affect others

Value harmony, forgiveness

Like to please others and point out the best in people Could be described as warm,

empathetic then you prefer

Feeling

Sensors or iNtuitives

Introverts or Extroverts

Thinkers or Feelers

Judgers or Perceivers

- · Like to be in a fast-naced
- Tend to work out ideas with

then you prefer

Extraversion

be the center of attention

Action-oriented, logical, analytical, spontaneous, reserved, independent, Enjoy adventure, skilled at understanding how mechanical things work.

Outgoing, realistic,

action-oriented, curious

versatile, spontaneous

Pragmatic problem

solvers and skillful



**ESFP** 

Playful, enthusiastic,

friendly, spontaneous,

tactful, flexible, Have

strong common sense,

enjoy helping people ir





Enthusiastic, creative

inspiration, enjoy

starting new projects, see potential in others

## Inventive, enthusiastic

strategic, enterprising inquisitive, versatile. Enjoy new ideas and

· Prefer to have matters Prefer to leave your options

 Think rules and deadlines See rules and deadlines as

> Like to improvise and make things up as you go Are spontaneous, enjoy surprises and new situations

> > then you prefer

Perceiving

#### 2. How do you prefer to take in information? If you:

- · Focus on the reality of how things are
- Pay attention to concrete facts and details
- Prefer ideas that have practical applications
- Like to describe things in a specific, literal way

then you prefer

Sensing

- · Imagine the possibilities of how things could be
- Notice the big picture, see how everything connects Enjoy ideas and concepts
- for their own sake Like to describe things in a figurative, poetic way

then you prefer

Ν Intuition

#### Efficient, outgoing, analytical, systematic ike to run the show an get things done in an

Friendly, outgoing, reliable, conscientious organized, practical. Seel others, enjoy being active and productive



efficient, outgoing, mbitious, independen Effective organizers of people and long-range planners.

#### 4. How do you prefer to live your outer life? If you:

3. How do you prefer to make decisions? If you:

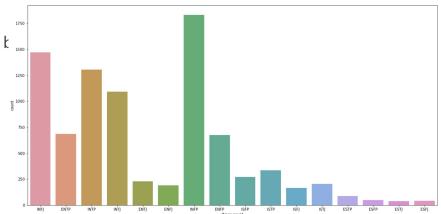
- settled
- should be respected · Prefer to have detailed,
- step-by-step instructions · Make plans, want to know what you're getting into

then you prefer

Judging

### **Dataset**

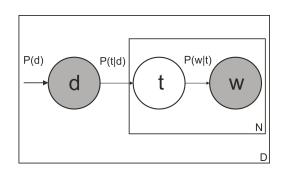
- Text to MBTI Personality dataset found on Kaggle
- 8675 data points
- Multiple instances of text per data point, separated k
   3 pipes '|||'
- Cleaned data to remove links, the pipes, and any other irregular characters



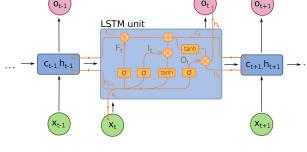
mb.loc[3, 'posts']

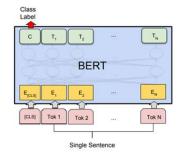
''Dear INTP I enjoyed our conversation the other day Esoteric gabbing about the nature of the universe and the idea that every rule and social code being ar bitrary constructs createdDear ENTJ sub Long time no see Sincerely AlphaNone of them All other types hurt in deep existential ways that I want no part offro bably a sliding scale that depends on individual preferences like everything in humanityDraco Malfoy also I'd say he's either or I'm either or though in whi ch stacking to me is a somewhat arbitrary distinction to make as I believe that the core indicates primary motivation and has a hand in every action Therefo re al'm not particularly introverted or extraverted personally That said I would say I'm somewhat unphased by either social interactions or being alone What I'd say I crave more so than anything isDear Type INFP Your absolute admiration of me is refreshing You're a great girlfriend and I wish we both didn't have such busy schedules so we could be around one another more ofte...'

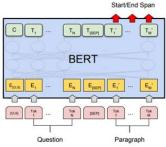
## **Different Approaches**



- pLSA
- RNN (LSTM)
- BERT







### pLSA

#### Recap:

- Method for semantic analysis in text
- Finds topic distributions in text
- optimized using E-M steps
- $\Theta$  (N x K): topic probabilities per document
- φ (K X V): word probabilities per topic
- Z (N X V X K): probability of word w in document i being generated by topic k

16 PERSONALITIE	S REPORT			
р	recision	recall	f1-score	support
0	0.17	0.06	0.09	1470
1	0.08	0.06	0.07	685
2	0.13	0.06	0.08	1304
3	0.12	0.05	0.07	1091
4	0.02	0.06	0.03	231
5	0.02	0.05	0.03	190
6	0.24	0.08	0.12	1832
7	0.12	0.11	0.11	675
8	0.03	0.07	0.04	271
9	0.04	0.07	0.05	337
10	0.01	0.04	0.02	166
11	0.02	0.05	0.03	205
12	0.01	0.08	0.02	89
13	0.00	0.04	0.01	48
14	0.00	0.03	0.00	39
15	0.00	0.05	0.01	42
accuracy			0.06	8675
macro avg	0.06	0.06	0.05	8675

0.06

0.08

8675

0.14

weighted avg

- First tried pLSA with all 16 personality types

				pred	ision	recall	f1-score	support
				0	0.76	0.49	0.59	6676
				0				
				1	0.22	0.48	0.30	1999
		LSA	ac	curacy			0.49	8675
	P		mac	ro avg	0.49	0.48	0.45	
	-		weight		0.63	0.49	0.53	8675
			_	-				
								', 'things', 'love', 'don', 'really', 'people', 'just', 'like', 'know']
N V. S REPOR	т					[ t	ime, good	od', 'way', 'people', 'say', 'like', 'don', 'thing', 'just', 'think']
	precision	recall	f1-score	support				
0		0.50	0.63	7478			1	11/hal libinal bal basalal Basal bisal libinal
1	0.13	0.48	0.21	1197				'like', 'things', 've', 'people', 'know', 'just', 'thing'] , 'say', 'people', 'really', 'don', 'time', 'like', 'think']
					[ Tee.	L , KIIOW	, just,	, say , people , really , don , time , like , think ]
accuracy			0.50	8675				
macro avg		0.49	0.42	8675				
weighted avg	0.76	0.50	0.58	8675				
F V. T REPORT								
	precision	recall f1	-score su	pport				
0	0.56	0.56	0.56	4694				
1	0.48	0.48	0.48	3981				
				['tir	ne', 't	hings',	'know',	'really', 'good', 'don', 'way', 'like', 'just', 'think']
accuracy			0.52	8675 ['tvr			-	don', 'sure', 'time', 'know', 'make', 've', 'people']
macro avg	0.52	0.52	0.52	8675	, ,			, , , , , , , , , , , , , , , , , , , ,
weighted avg	0.52	0.52	0.52	8675				
J V. P REPORT								
0 11 11210111	precision	recall	f1-score	support				
	F			FF				
0	0.39	0.51	0.45	3434				
1	0.60	0.48	0.53	5241				
_	internal state of	con costs	2000 C-2012	50.750-4608700E	[	'time',	'thing',	'just', 'really', 'like', 've', 'know', 'feel', 'think', 'people']
accuracy			0.49	8675	]	'love',	'really',	, 'think', 'know', 'time', 'way', 'just', 'like', 'don', 'good']
	0 50	0 50	0.40	0675				

I V. E REPORT

0.50

0.52

macro avg weighted avg

0.50

0.49

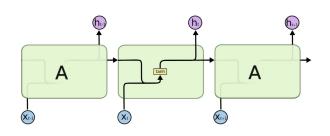
0.49

0.50

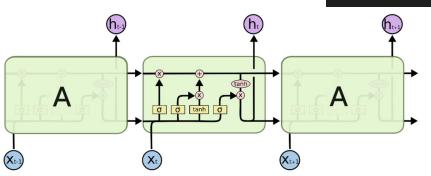
8675

8675

## **LSTM**



A regular RNN



epoch	train_loss	valid_loss	accuracy	time
0	1.477793	2.277507	0.317579	07:02
1	1.468069	2.078241	0.424207	07:03
2	1.265659	1.962566	0.436311	07:03
3	1.044665	1.961544	0.445533	07:03
4	0.878513	1.902411	0.458213	07:03

 Used fastai library, had built in methods for using an LSTM and predicting

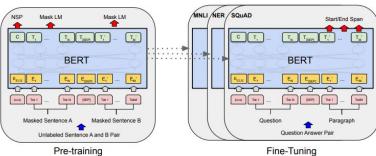
personality\_pred = learn\_data.predict("I am doing machine learning all day inside at home")

LSTM Architecture

#### **BERT**

#### Recap:

- Newer model (2018): Bidirectional Encoder Representations from Transformers
- Jointly conditions on left and right context for all layers
- "Masked Language Model" mask some tokens and predict original ID by context
- "Next Sentence Prediction" not as relevant, most inputs were single sentences
- Bert base 12 encoder layers
- Pre-trained; fine tuned by final additional output layer without substantial architectural change
- 85/15 Training/Validation Split



#### **BERT**

```
Epoch 6
Epoch 1
                                                         Training loss: 0.4170884359764327
Training loss: 2.1211794058826636
                                                          Test loss: 2.4628626195930994
Test loss: 1.8534048400773593
                                                         F1 Score test (Weighted): 0.4852421048274305
F1 Score test (Weighted): 0.3792028764849477
                                                         Epoch 7: 0%
                                                                               0/922 [00:00<?, ?it/s]
Epoch 2: 0%
                      | 0/922 [00:00<?, ?it/s]
                                                         Epoch 7
Epoch 2
                                                         Training loss: 0.2737534848411028
Training loss: 1.6583584517305172
                                                         Test loss: 2.745829509811167
Test loss: 1.6190826106656548
                                                         F1 Score test (Weighted): 0.4855002776617087
F1 Score test (Weighted): 0.4927139860794328
                                                         Epoch 8: 0% | 0/922 [00:00<?, ?it/s]
Epoch 3: 0% 0/922 [00:00<?, ?it/s]
                                                         Epoch 8
Epoch 3
                                                         Training loss: 0.1800585792387569
Training loss: 1.3012166120871547
                                                          Test loss: 3.0763083005244014
Test loss: 1.6957366400701137
                                                         F1 Score test (Weighted): 0.4873826868478445
F1 Score test (Weighted): 0.5049519647334698
                                                         Epoch 9: 0% | 0/922 [00:00<?, ?it/s]
Epoch 4: 0%
                       | 0/922 [00:00<?, ?it/s]
                                                         Epoch 9
Epoch 4
                                                         Training loss: 0.12046974006213991
Training loss: 0.9403691974401862
                                                          Test loss: 3.1599525183987764
Test loss: 1.718616721088901
                                                         F1 Score test (Weighted): 0.49749752156010946
F1 Score test (Weighted): 0.5043785361836136
                                                                                 | 0/922 [00:00<?, ?it/s]
                                                         Epoch 10: 0%
Epoch 5: 0% 0/922 [00:00<?, ?it/s]
                                                         Epoch 10
Epoch 5
                                                         Training loss: 0.08876481469031959
Training loss: 0.630955840388228
                                                         Test loss: 3.2372621572090803
Test loss: 2.0997796651044505
                                                         F1 Score test (Weighted): 0.4918067876294191
F1 Score test (Weighted): 0.5028874022377327
```

### **BERT Predictions**

```
predict("I love staying indoors!")
predict("I love going outside with my friends!")
predict("I love partying!!!")
```



### Caveats and Roadblocks

- Biased dataset
  - Overfitting and bad training
  - BERT struggles with this currently
- Slow training for some models
- Bugs caused by training on relatively new models difficult to debug
- Training takes forever for new models (GPU)

### **Ethics & Considerations**

- Must follow Value Sensitive Design
  - Ex. discrimination in the workplace
    - Direct Stakeholders: Interviewees
    - Indirect Stakeholders: Consumers/Public
  - F1 score vs accuracy
- The current design of all our model is dangerous
- Must only be for recreational use
- Clear biases in datasets that can be disadvantageous
  - Likely measurement bias
  - Pulled from "PersonalityCafe forum" likely biased source

## Thank you!