

Introduction

In this work, we intent to develop a system that can predict one's personality based on the post or status one uploads on the social media network.

This study is based on the prediction of Big Five Personality, MBTI (Myers-Briggs Type Indicator)

- I-E: Extraversion (E) or Introversion (I)
- N-S: Sensing (S) or INtuition (N)
- T-F: Thinking (T) or Feeling (F)
- J-P: Judging (J) or Perceiving (P)



Why Personality Prediction?

- Recruitment systems
- Personal Counseling Systems
- Computational Advertising,
- Marketing Science
- Enhanced Human-Computer Interaction
- Bank Credit Scoring Systems.

People often update statuses, posts, comments to express their feelings and opinions on social media. These expressions can be used to characterize the individual's behavior, personality, and characteristics of their thought patterns.

NF Valuing

Possible

NT

Visioning

Pulling people with ideas to an optimistic future

Manifesting universal values and valuing people

ENFJ	INFJ	INTJ	ENTJ
Teacher Smooth billing charmers, Mary inspiring & modivatoria. Other clarge. Process leaders & productions. Creat assespectal. Very relationaries creat assespectal. No modivate groups.	Counselor Work is to inspire others to actione great things. Great visioneries of furniar possibilities. Senous academicatins. Other professions or other themselves to a religious order.	Mastermind If they say they are going to do something, they do it. Likely to be cooperate leaders, scientists, theleve everything has room for exprovement. Experior planners and valenance of systems.	Field Marshall Very leadership-oriented. Likely to be top executives, business persone, titig on reducing retificancy, retificativeness. Take charge people. Clart be over- whereing to less outgoing types.
ENFP	INFP	INTP	ENTP
Champion Second only to ESFP's for An. Want their Sled with exchanent and romance. Very enthusiastic and rosative. Other teachers, artists, writers. Great need for diversity and change.	Healer Notes senerals ading accets, Different from ISFPs, they by to tackle long-term problems. Other psychologists or counselons. Want to save the wholes and nantorests.	Architect Despet analysis of problems to be solved. Often physicies, scientists. Most alcoft of types, Ortical thinkers.	Inventor Wart one existing challenge after another. Love to problem solve. Good at analysis, consider themselves full of ingenuity and ideas. Other involved in complicit, systems analysis, design.
ESFP	ISFP	ISTP	ESTP

Personal

and creative. Often teachers, artists, writers. Great need for diversity and change.	Other psychologists or counselors. Want to save the whales and numbers.	Critical tirritors.	Thermoelves full of ingenuity and ideas. Offers involved in corn; sct. systems analysis, design.
ESFP Performer Number one in furn and orthussiams. Alexays invite ESFPs to your party. The most generous of all types. Witers, Invendy, witners people. Escalars at outcomes service.	ISFP Composer Cutelly harmonicus with world. Very obsensing, beneatiers, asched obsend work with people in need. Work to solve problems, of the immediate such as homeless, stopping hunger.	ISTP Operator Flexity to thy smything once. Plunhed with the nuch of this. Beek excellenement. A fove of bools and the utility they offler. Incidend. Lowerd mechanical devices, can take upon a female and anything	ESTP Promoter Exclusivest seelests. Never feel more after than when taking, release. Great relegations on the form end. Excellent previously. A enthropression capabilities if someone dies klaims through
ESFJ Provider Hosts & hostenes, Gracious- ness of this type makes them excellent at entertaining, coordinating, May be feechers,	ISFJ Protector A high sense of duty. Upholders of lamily tradition. Other found in traditional helping professions including nursing, elementary	ISTJ Inspector Does of what should be done. Masters at competing practical details and adding freeling touches. Get-4 done people.	ESTJ Supervisor Administration, workers, pillars of streigh in community, Logial matter, parents, employees. Other promoted to management

SF

Relating Including and building trustworthiness

nurses. Very conscious of

ppeinmoss, should shouldn'ts

Present

Superb administrators. Dutybound & obligates, often military positions. Dependable.

consistent, straightforward.

education, etc.

ST

Logical

Directing
Action from a strategic perspective

Literature Review

 Comparison of machine learning algorithms for content based personality resolution of tweets

Shruti Garg, Ashwani Garg, (https://www.sciencedirect.com/science/article/pii/S2590291121000747)

- Personality Prediction Using Machine Learning
 Shilpa R,Supriya V, Sweta Prasad , Vinaya Varshini R,Uday Shankar SV
 (https://saejournal.com/wp-content/uploads/2021/07/Personality-Prediction-Using-Machine-Learning.pdf)
 - ABOUT MYER-BRIGGS TYPE INDICATOR
 https://rismakov.com/mbti-prediction/category/xgboost

Dataset

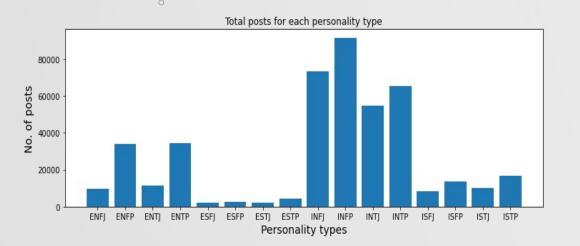
The dataset used in this study is from the Myers–Briggs personality dataset which is openly available on Kaggle (https://www.kaggle.com/datasnaek/mbti-type).

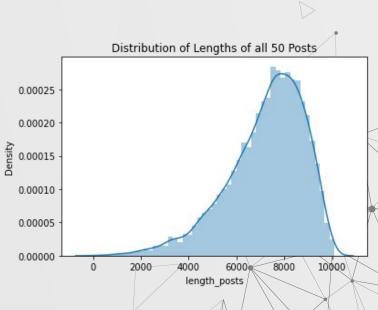
- 8675 unique value
- First column : MBTI
 Personality Type
- Second Column: 50 things they have posted.

	type	posts
8670	ISFP	'https://www.youtube.com/watch?v=t8edHB_h908
8671	ENFP	'Soif this thread already exists someplace
8672	INTP	'So many questions when i do these things. I
8673	INFP	'I am very conflicted right now when it comes
8674	INFP	'It has been too long since I have been on per



Visualizing Dataset





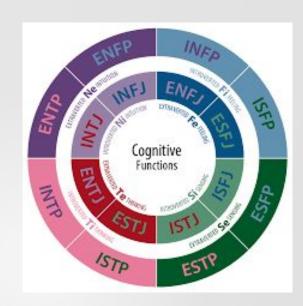


- NLTK (natural language processing toolkit) and XGBoost
- Frequency–Inverse Document Frequency
- Pandas, numpy, re, seaborn, matplotlib and sklearn are other
 Python libraries are used

Pipeline

1. Preprocessing of Dataset

- URL's and Stop words were removed from the dataset.
- b. Splitting the MBTI personality into four letters and binarising it.
- the text was lemmatised, i.e., inflected forms of the words were transformed into their root words



2. Feature Engineering

a. Vectorise with Count and Term Frequency–Inverse Document Frequency (TF–IDF)

Pipeline (Contd.)

3. Splitting the Features:

X: user posts in TF-IDF representation

Y: Personality type in Binarised MBTI form

4. Training Model

a. In total, 70% of the data was used as the training set and 30% of the data was used as the test set. The model was fit onto the training data and the predictions were made for the testing data.



5. Evaluating Model

Random Forest Classifier

IE: Introversion (I) / Extroversion (E) Accuracy: 77.33%

NS: Intuition (N) / Sensing (S) Accuracy: 86.03%

FT: Feeling (F) / Thinking (T) Accuracy: 67.66%

JP: Judging (J) / Perceiving (P) Accuracy: 62.80%

XGBoost Classifier

IE: Introversion (I) / Extroversion (E) Accuracy: 77.65%

NS: Intuition (N) / Sensing (S) Accuracy: 86.06%

FT: Feeling (F) / Thinking (T) Accuracy: 68.77%

JP: Judging (J) / Perceiving (P) Accuracy: 64.83%

KNN Classifier

IE: Introversion (I) / Extroversion (E) Accuracy: 76.67%

NS: Intuition (N) / Sensing (S) Accuracy: 85.82%

FT: Feeling (F) / Thinking (T) Accuracy: 54.70%

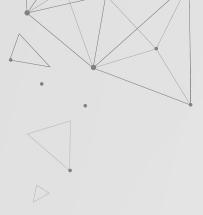
JP: Judging (J) / Perceiving (P) Accuracy: 40.20%



Phase 1- Result and Conclusion

Out of all the models, seen above we see that on an average XG Boost gives relatively good performance, hence we choose it to build our Personality prediction model. This will be beneficial as XGBoost model can even be used to evaluate and report on the performance on a test set for the model during training.

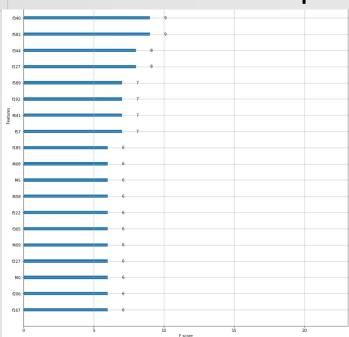
The results show that the methodology has better accuracy and reliability in comparison to other existing methods. Regarding the knowledge contribution in this paper, the presented methodology significantly improved the accuracy of recognising the Intuition (I)—Sensing (S) and Introversion (I)—Extroversion (E) personality categories, as well as slightly better accuracy for the Judging (J)—Perceiving (P) personality category. This can effectively assist NLP practitioners and psychologists in regards to identification of personality types and associated cognitive processes.





WHY XGBOOST???





XGBOOST automatically provide estimates of feature importance from a trained predictive model.

Hyperparameter Tuning

The parameters to consider tuning are:

- The number and size of trees (n_estimators and max_depth).
- The learning rate and number of trees (learning_rate and n_estimators).
- The row subsampling rates (subsample)

Our Tuning FORMULATION!!

```
IE: Introversion (I) / Extroversion (E) ...
* Best: -0.508298 using {'learning rate': 0.2, 'n estimators': 200}
* -0.508298 (0.012022) with: {'learning rate': 0.2, 'n estimators': 200}
* -0.511190 (0.014109) with: {'learning rate': 0.2, 'n estimators': 300}
* -0.515801 (0.013401) with: {'learning rate': 0.3, 'n estimators': 200}
* -0.523666 (0.014271) with: {'learning rate': 0.3, 'n estimators': 300}
NS: Intuition (N) / Sensing (S) ...
* Best: -0.395608 using {'learning rate': 0.2, 'n estimators': 200}
* -0.395608 (0.002532) with: {'learning rate': 0.2, 'n estimators': 200}
* -0.407233 (0.004062) with: {'learning rate': 0.2, 'n estimators': 300}
* -0.418423 (0.003287) with: {'learning rate': 0.3, 'n estimators': 200}
* -0.436628 (0.004779) with: {'learning rate': 0.3, 'n estimators': 300}
FT: Feeling (F) / Thinking (T) ...
* Best: -0.559850 using {'learning rate': 0.2, 'n estimators': 200}
* -0.559850 (0.019320) with: {'learning rate': 0.2, 'n estimators': 200}
* -0.566251 (0.021509) with: {'learning rate': 0.2, 'n estimators': 300}
* -0.575489 (0.024986) with: {'learning rate': 0.3, 'n_estimators': 200}
* -0.588334 (0.027208) with: {'learning rate': 0.3, 'n estimators': 300}
JP: Judging (J) / Perceiving (P) ...
* Best: -0.650526 using {'learning rate': 0.2, 'n estimators': 200}
* -0.650526 (0.014655) with: {'learning rate': 0.2, 'n estimators': 200}
* -0.662368 (0.015849) with: {'learning rate': 0.2, 'n estimators': 300}
* -0.667228 (0.012558) with: {'learning rate': 0.3, 'n estimators': 200}
* -0.685591 (0.018371) with: {'learning rate': 0.3, 'n estimators': 300}
```

- → learning rate as 0.2
- → number of estimators as 200
- → maximum depth of tree as 2
- → number of threads as 8



Random post



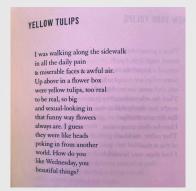
Sshayni 🦟 @Sshayni_ · Apr 18

Hey guys! • My name is Sshayni. I am 22 years old, I am a student and I am engaged in Digital art. I believe that we have incredible visual power in our hands. Art is something that can not only change a particular person, art can change the whole world.

The result is: INFP

TEST CASES Let'S PREDICT!!

A Poem



The result is: INFP

A Tweet



The result is: INFJ



Failures!!

Bill Gates tweet's

The result is: INFP

Reality:



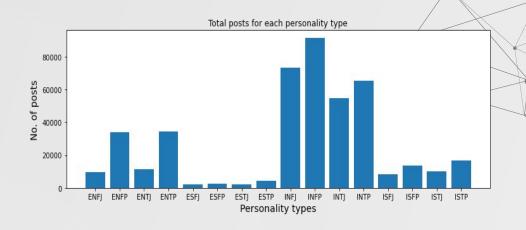
Taylor Swift's tweet's

The result is: INFP

Reality:



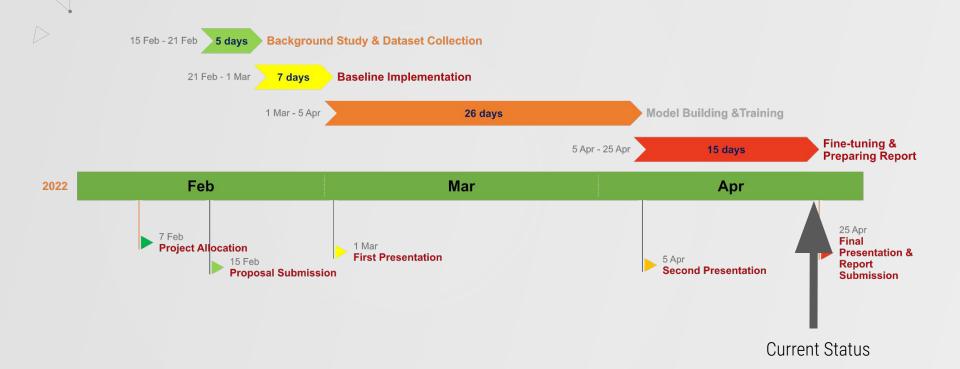




- Data Imbalance
- Myers-Briggs provides inconsistent, inaccurate results

Implementing with more data, natural language processing methods, or newer text features will help us predict it accurately

Timeline and Progress





https://colab.research.google.com/drive/1kwwCyo_-HA9hIXDbXzT7xYVa-Qjg2L1y?usp=sharing



- 1. https://www.personality-database.com/profile?pid=1&sort=top
- 2. https://www.sciencedirect.com/science/article/pii/S2590291121000747
- 3. https://web.stanford.edu/class/archive/cs/cs224n/cs224n.1184/reports/6839354.pdf
- 4. https://www.kaggle.com/datasets/datasnaek/mbti-type

Group Members

- 1. Darshan Patil 21120006
- 2. Shoaib Alam 21250011
- 3. Iram Nawab 21250024
- 4. G. V. Sai 21210036