

Personality Prediction System Using Machine Learning

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Abstract: A personality prediction system leverages advanced data analysis techniques to assess and predict an individual's personality traits based on diverse inputs such as text, voice, behaviour, and biometric data. These systems typically utilize psychological models, like the Big Five Personality Traits or Myers-Briggs Type Indicator (MBTI), to derive insights from patterns in the data. Text-based prediction systems use natural language processing (NLP) to analyze written or spoken communication, while voice-based systems analyze vocal attributes such as tone and pitch. Behavioral and biometric data further enhance the accuracy of these systems, with applications spanning areas like human-computer interaction, marketing, recruitment, and mental health. With the aid of artificial intelligence (AI) and machine learning, these systems are increasingly capable of delivering personalized insights, offering the potential for a wide range of applications from customer engagement to psychological well-being. This abstract outline the core mechanisms, applications, and future directions of personality prediction systems, highlighting their potential to transform how individuals are understood and interacted with in both digital and real-world contexts.

INTRODUCTION

A Personality Prediction System is an advanced technological tool designed to assess and predict an individual's personality traits based on various data inputs such as written text, behavioural patterns, voice analysis, and even facial expressions. By leveraging insights from psychology and using techniques from machine learning and data science, these systems aim to analyze and predict personality traits with increasing accuracy. The goal of such systems is to provide deeper insights into human behavior, preferences, and motivations. The core foundation of these systems often lies in established psychological models like the Big Five Personality Traits (OCEAN), Myers-Briggs Type Indicator (MBTI), and other personality frameworks. Using data from social media activity, survey responses, online interactions, or even voice tone and facial expressions, a Personality Prediction System can generate detailed profiles that reflect various aspects of an individual's personality. These systems have a wide range of applications, from improving hiring processes in Human Resources to personalizing marketing efforts and enhancing user experiences in entertainment platforms. However, while they offer valuable insights, their use raises important ethical considerations related to privacy, consent, and potential biases in predictions. In essence, a Personality Prediction System represents the intersection of psychology, data science, and artificial intelligence, providing a unique lens to understand and predict human behavior.

II. EASE OF USE

A. Big Three Personality Traits (OCEAN)

The Big Five personality traits, a widely recognized model known as OCEAN, provides a comprehensive framework for describing human personality across five fundamental dimensions.. These traits include Openness to Experience, which reflects a person's willingness to embrace new ideas, creativity, and curiosity; Conscientiousness, which describes a person's level of organization, responsibility, and dependability; Extraversion, which measures sociability, energy, and the tendency to seek out stimulation and social interaction; Agreeableness, which refers to a person's capacity for compassion, cooperation, and empathy toward others; and Neuroticism, which indicates emotional stability, with higher levels of neuroticism associated with greater emotional reactivity and stress sensitivity. Each of these traits exists on a continuum, meaning individuals can exhibit varying degrees of each trait, leading to unique personality profiles. Together, the Big Five model provides a comprehensive framework for understanding human behavior and predicting outcomes in areas such as work, relationships, and mental health.

B. MBTI (Myers-Briggs Type Indicator)

The Myers-Briggs Type Indicator (MBTI) is a widely used personality assessment tool based on Carl Jung's theory of psychological types. It categorizes individuals into one of 16 distinct personality types by measuring preferences across four pairs of opposing traits: Extraversion (E) vs. Introversion (I), which reflects whether a person is more energized by social interaction or solitary activities; Sensing (S) vs. Intuition (N), indicating whether someone focuses on present realities and facts (Sensing) or future possibilities and abstract concepts (Intuition); Thinking (T) vs. Feeling (F), which assesses whether decisions are made based on logic and objective criteria (Thinking) or personal values and the feelings of others (Feeling); and Judging (J) vs. Perceiving (P), which describes whether a person prefers structure, order, and planning (Judging) or flexibility, spontaneity, and adaptability (Perceiving). The MBTI is often used in personal development, career counseling, and team building, helping individuals gain insight into their own preferences and improve communication and collaboration with others. However, it has faced criticism for its validity and reliability in predicting behavior and outcomes.

C. Enneagram of Personality

The Enneagram of Personality is a model that describes nine distinct personality types, each with its own core motivations, fears, and behaviours. The system is based on the idea that each person has a dominant personality type that influences how they perceive the world and interact with others. The nine types are divided into three centres: the head centre (types 5, 6, and 7) deals with thinking and fear; the heart center (types 2, 3, and 4) focused on emotions and identity; and the gut centre (types 8, 9, and 1) related to instinct and anger. Each type has its own strengths and challenges, such as Type 1 (The Perfectionist) striving for correctness and order, Type 2 (The Helper) focusing on relationships and caring for others, and Type 5 (The Investigator) valuing knowledge and independence. The Enneagram also explores how each type behaves under stress or growth, highlighting the dynamic nature of personality. This system is widely used for personal growth, self-awareness, and understanding interpersonal dynamics, but it has also faced some criticisms regarding its empirical support and application.

D. HEXACO Personality Model

The HEXACO Personality Model is a six- factor model of personality that expands upon the Big Five by adding a sixth dimension called Honesty- Humility. This model includes the following six traits: Honesty-Humility (the degree to which a person is sincere, modest, and fair versus manipulative, arrogant, and greedy), Emotionality (similar to neuroticism, reflecting emotional stability and the tendency to experience negative emotions), Extraversion (energy, sociability, and enthusiasm), Agreeableness (compassion, cooperation, and a tendency to avoid conflict), Conscientiousness (organization, responsibility, and goal-directed behavior), and Openness to Experience (curiosity, creativity, and a willingness to engage with new ideas and experiences). The HEXACO model provides a more nuanced understanding of personality by incorporating aspects like honesty and humility, which are not fully captured in the Big Five. It is often used in psychology to assess personality traits, predict behavior, and study cross-cultural differences. Applicable in fields such as marketing, recruitment, and mental health.

II. METHODOLOGIES USED IN PERSONALITY PREDICTION

Personality prediction systems employ a variety of methodologies to assess and predict an individual's personality traits. One common approach is through survey-based assessments, where individuals answer a series of questions designed to gauge their behavior, preferences, and tendencies, often based on established models like the Big Five or MBTI. These responses are then used to predict personality traits. Another prevalent method is machine learning, where algorithms are trained on large datasets to identify patterns in behaviors, such as social media activity or written text, which are indicative of personality traits. Techniques like supervised learning are used to classify individuals into personality types based on labeled data, while unsupervised learning can group people into similar personality profiles without prior labeling. Natural Language Processing (NLP) is a critical tool in these systems, analyzing written or spoken language to infer personality traits based on word choice, tone, and sentiment. Additionally, computer vision and facial expression analysis are increasingly utilized to predict personality by interpreting non-verbal cues like facial expressions and body language. These methodologies collectively help in creating more accurate and personalized predictions of personality,

A Survey-based Approaches

Survey-based approaches to personality prediction involve using structured questionnaires or assessments to gather information about an individual's behavior, preferences, and tendencies.

These surveys are designed based on established psychological models, such as the Big Five Personality Traits (OCEAN), MBTI, or the Enneagram, and typically consist of a series of questions that ask respondents to rate how strongly they agree or disagree with various statements about themselves.

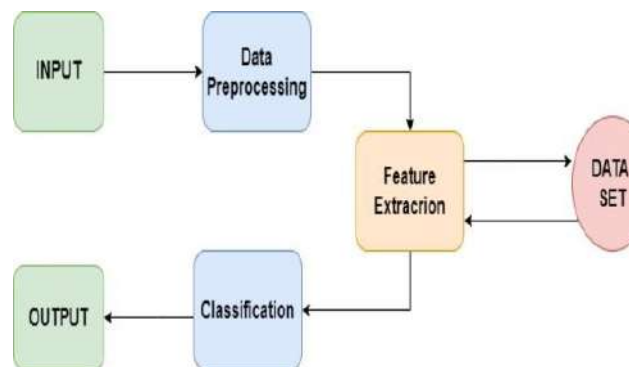


Fig.1 Block Diagram

The answers are then analysed to determine where the individual falls on different personality dimensions. For example, in a Big Five-based survey, questions might assess a person's openness to new experiences, their level of extraversion, their conscientiousness, agreeableness, and emotional stability. Responses are scored and mapped to these five traits, offering insights into the person's overall personality profile. Similarly, MBTI-based surveys categorize individuals into one of 16 personality types based on their answers about their preferences for socializing, decision-making, and planning. Survey-based approaches are widely used because they are relatively easy to administer and understand, and they provide valuable, structured data for personality assessment. However, their effectiveness depends on the accuracy and honesty of the responses, and they can sometimes be limited by the respondent's self-awareness or willingness to disclose personal information. Despite these limitations, they remain a cornerstone for personality prediction in fields such as career counselling, organizational development, and personal growth.

B Machine learning Models

- Machine learning models have become increasingly important in personality prediction by utilizing large datasets to identify patterns in behaviour, language, and interactions that are indicative of an individual's personality traits. These models go beyond traditional self-report surveys and analyse complex data sources such as social media activity, online interactions, and text data to make predictions about a person's personality [9].
- One of the primary techniques used is supervised learning, where algorithms are trained on labelled data (i.e., data that already contains known personality traits or classifications). For example, a dataset might consist of text written by individuals alongside their known personality profiles based on surveys like the Big Five. The machine learning model learns to associate specific patterns in language or behaviour (e.g., word choice, sentence structure, emotional tone) with particular traits. Once trained, the model can predict the personality of a new individual based on their input data, such as the language used in a social media post or email.
- Another method, unsupervised learning, involves clustering or grouping similar data without predefined labels. For instance, the algorithm might analyze text or behaviour patterns and group individuals into clusters with similar personality characteristics, even without knowing the exact personality traits upfront. This can reveal insights into sub types of personalities or patterns that aren't immediately obvious.
- Natural Language Processing (NLP) is a key tool in machine learning-based personality prediction, analysing text data to assess traits such as openness, neuroticism, or extraversion. For instance, NLP algorithms can evaluate how often a person uses words related to emotion, social interaction, or abstract thinking to infer personality traits.

Additionally, deep learning models, which are more complex neural networks, can be employed to understand highly intricate patterns in large datasets. These models, while more computationally intensive, offer the potential to improve prediction accuracy, particularly in handling multimodal data (such as combining text with images or voice data) to assess personality. Overall, machine learning models in personality prediction provide a more dynamic and scalable approach, allowing for the analysis of vast amounts of data to uncover subtle personality traits that traditional methods may miss. However, these models also face challenges, such as ensuring data privacy, dealing with biases in training datasets, and the complexity of translating subjective human traits into machine-readable patterns.

C Natural Language Processing (NLP)

Natural Language Processing (NLP) plays a crucial role in personality prediction systems by analyzing and interpreting human language to assess individual personality traits. NLP is a branch of artificial intelligence (AI) that enables machines to understand, interpret, and generate human language in a way that is both meaningful and contextually relevant. In the context of personality prediction, NLP algorithms process various forms of written or spoken communication such as social media posts, emails, or even speech transcripts to identify linguistic patterns that are indicative of specific personality traits [10]. For instance, NLP can assess the Big Five personality traits by analysing the words, phrases, and sentence structures an individual uses.

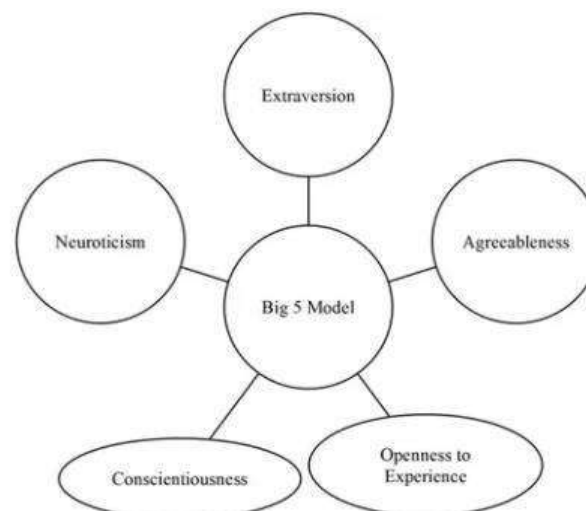
A person who frequently uses words related to social interaction (e.g., "party," "friends," "together") may score higher in extraversion, while someone using more reflective, analytical language (e.g., "think," "consider," "imagine") could be interpreted as having a higher openness to experience. Similarly, NLP can identify emotional tone in language to gauge neuroticism or agreeableness, depending on whether the language is more emotionally reactive or compassionate. Sentiment analysis, a common NLP technique, helps to evaluate the emotional content of text, revealing how a person expresses emotions and their overall mood. By identifying words that convey feelings such as happiness, anger, or sadness, NLP can predict how an individual might score in neuroticism (high neuroticism may correlate with negative emotional language) or agreeableness (a more empathetic and supportive tone may indicate higher agreeableness). Advanced NLP models, including deep learning algorithms, go beyond simple word frequency analysis to consider more complex factors like syntax, context, and semantics, enabling deeper insights into personality traits. These models, such as transformers (e.g., BERT, GPT), are able to understand the nuances of language and can analyse large scale text data to make personality predictions with greater accuracy [10]. NLP powered personality prediction is increasingly used in areas such as recruitment (to assess how a candidate's communication style aligns with team culture), social media analysis (to infer personality traits based on posts or interactions), and customer service (to personalize responses based on inferred personality). However, challenges remain in ensuring that NLP models accurately capture the subtleties of human personality, particularly when language can vary greatly across individuals and cultures.

D Computer Vision and Facial Expression Recognition

Computer vision and facial expression recognition are emerging methodologies used in personality prediction systems to assess individuals' personality traits based on their visual and non-verbal cues. These technologies analyze facial expressions, body language, and other visual indicators to infer psychological characteristics, often complementing traditional methods like surveys or text-based analysis [13]. Computer vision refers to the ability of machines to interpret and process visual information, allowing them to see and analyse images or videos in a way similar to human perception. In personality prediction, computer vision is often employed to study the facial expressions, gestures, and movements of individuals. By using algorithms trained on large datasets of human faces, these systems can detect subtle changes in facial muscles, eye movement, posture, and other body language indicators, which are then mapped to specific personality traits. Facial expression recognition is a key component of computer vision in personality prediction. It focuses on analyzing facial expressions to gauge emotional states, which can be linked to certain personality traits. For example, extroverted individuals may exhibit more expressive, frequent smiles and stronger engagement in facial interactions, while introverted individuals might show more subdued or less frequent facial expressions. Similarly, **neuroticism** may be associated with more frequent displays of negative emotions like anxiety, sadness, or worry, which can be detected through facial analysis [13]. In practice, facial expression recognition works by detecting key facial features such as the mouth, eyes, and eyebrows, using machine learning algorithms to identify specific emotions. For instance, a smile can indicate happiness or contentment, while a furrowed brow may suggest confusion or anger. More complex systems can analyze these facial cuts over time and in various contexts to identify patterns in an individual's emotional responses, which can be used to predict personality traits like **agreeableness** (calm and pleasant expressions) or **neuroticism** (frequent stress-related expressions). These methods are especially valuable in settings like recruitment, customer service, and mental health assessments, where non- verbal cues can offer insights into a person's emotional stability, openness, or interpersonal dynamics. However, challenges remain in accurately interpreting facial expressions across different cultural contexts, as facial cues can vary greatly between individuals and cultures. Additionally, ethical concerns surrounding privacy and the potential misuse of facial recognition technology in profiling individuals also need to be considered as these systems become more prevalent.

III. APPLICATIONS OF PERSONALITY SYSTEMS

Consumers' preferences, motivations, and purchasing behaviors [11]. By leveraging tools such as survey-based assessments, Personality prediction systems have a wide range of applications across various industries [10].



Fig,2 Personality Traits

In human resources, these systems help in selecting candidates who fit organizational cultures, guiding employee development, and optimizing team dynamics. In marketing, personality insights are used for targeted advertising, personalized product recommendations, and crafting brand messages that resonate with specific audiences. In healthcare, these systems assist in mental health assessments, tailored therapies, and personalized wellness programs. In education, they support personalized learning experiences, career counseling, and student engagement strategies. Social media platforms leverage personality prediction for content personalization and user profiling, enhancing engagement. In customer support, businesses can adapt their communication to better suit customer personalities, improving satisfaction and service outcomes.

A Marketing and Consumer Behaviour

In marketing and consumer behaviour, personality prediction systems are increasingly being used to gain deeper insights into machine learning, NLP, and behavioural data analysis, marketers can create more personalized and targeted campaigns, improve customer engagement, and increase overall brand loyalty [11]. Consumer segmentation is one of the key areas where personality prediction is applied. Traditional market segmentation often relies on demographic factors like age, gender, or income, but personality prediction offers a more nuanced approach by categorizing consumers based on traits such as openness, extraversion, conscientiousness, agreeableness, and neuroticism. For instance, individuals with high openness may be more receptive to innovative, cutting-edge products or creative advertising, while those with high neuroticism might be more responsive to messaging that emphasizes security, reliability, and reducing stress [11]. Extraverted consumers could be more likely to engage with social media campaigns, events, and interactive promotions, whereas introverted consumers might prefer more private, thoughtful interactions with brands.

B Human Resources

In Human Resources (HR), personality prediction systems are increasingly being used to enhance recruitment, team dynamics, employee development, and overall organizational effectiveness [14]. By leveraging tools such as survey-based assessments, machine learning, NLP, and computer vision, HR professionals can gain deeper insights into the personality traits of candidates and employees, which can be crucial in making informed decisions [14,15]. In recruitment and hiring, personality prediction tools can help HR departments assess whether a candidate's personality aligns with the role they are applying for, the organizational culture, and the team dynamics. For instance, tools based on the Big Five personality traits can help evaluate whether a candidate is conscientious (a valuable trait for roles requiring attention to detail), extraverted (for roles that require collaboration and social interaction), or emotionally stable (important for high-stress environments). Personality assessments can complement traditional interviews and resumes by providing an objective analysis of traits that could indicate future job performance, job satisfaction, and the likelihood of a successful long-term fit.

C Mental Health and Therapy

Mental health and therapy are areas where personality prediction models, including those using computer vision, NLP, and survey-based approaches, are increasingly being applied to enhance assessments and improve treatment outcomes [16]. Personality traits are closely linked to mental health, as they can influence how individual scope with stress, interact with others, and perceive the world. By predicting personality traits, mental health professionals can gain a more comprehensive understanding of a patient's emotional and psychological state, tailor therapeutic interventions, and monitor progress over time [16]. In mental health assessments, personality prediction tools can help identify underlying traits that may contribute to mental health conditions. For instance, individuals with high neuroticism may be more prone to anxiety, depression, or mood disorders, while those with lower levels of agreeableness might struggle with interpersonal relationships or conflict. By integrating personality prediction into mental health evaluations, clinicians can better understand the root causes of a patient's symptoms and design more personalized treatment plans.

IV. CONCLUSION

In conclusion, personality prediction systems have become valuable tools across various fields, including mental health, human resources, marketing, and consumer behaviour, by offering deeper insights into individual traits and behaviours. Whether through survey-based approaches, machine learning, NLP, or computer vision, these systems allow for a more nuanced understanding of personality, which can enhance decision making, improve personalization, and foster better outcomes. In mental health, personality prediction tools support tailored therapy, emotional regulation, and progress monitoring, contributing to more personalized care. In human resources, these systems assist in hiring, team-building, and employee development by providing data-driven insights into candidates' and employees' traits.

REFERENCES

1. Alam Sher Khan, Hussain Ahmad, Muhammad Zubair Asghar, Furqan Khan Saddozai, Areeba Arif, Hassan Ali Khalid, "Personality Classification from Online Text using Machine Learning Approach", International Journal of Advanced Computer Science and Applications, Volume:11, No. 3, 2022
2. P. Srinivasa Rao, PradeepBheemavarapu, P. S. Latha Kalyampudi and T. V. Madhusudhana Rao, "An Efficient Method for Coronavirus Detection Through X-rays using deep Neural Network", Journal of Current Medical Imaging,[online Available] Vol.18, No. 6, with ISSN: 1875-6603, 2022
3. Shilpa R., Supriya V., Sweta Prasad, Vinaya Varshini R., Uday Shankar SV, "Personality Prediction Using Machine Learning", Science and Engineering, Journal, Volume:25, 2021
4. P. MaheshKumar,P. Srinivasa Rao, "Frequent Pattern Retrieval on Data Streams by using Sliding Window", EAI Endorsed Transactions on Energy web, Volume:5,issue:35, 2021.

5. Hernandez, R.; Knight, I.S. Predicting Myers- Bridge Type Indicator with text classification. (Accessed on 9 September 2022).
6. T.V. Madhusudhana Rao, P. Srinivasa Rao, P.S. Latha Kalyampudi, "Iridology based Vital Organs Malfunctioning identification using Machine learning Techniques", International Journal of Advanced Science and Technology, Volume: 29, No. 5, PP: 5544–5554, 2022.
7. Sanjit Kumar, Shrivatson R., Rishi Priyan, Padmavathy, "Personality Prediction using TwitterData" International Research Journal of Engineering and Technology, Volume: 07 Issue: 07 | July 2022.
8. Vidya sagar Appaji setti, P Srinivasa Rao, "ANovel Scheme For Red Eye Removal With ImageMatching", Journal of Advanced Research in Dynamical & Control Systems, Vol. 10, 13-Special Issue, 2023.
9. S.Vidyasagar Appaji, P. V. Lakshmi, P. Srinivasa Rao, "Maximizing Joint Probability in Visual Question Answering Models", International Journal of Advanced Science and Technology Vol. 29, No. 3, pp. 3914 – 3923,
10. P. V. Lakshmi, P. Srinivasa Rao, "Maximizing Joint Probability in Visual Question Answering Models", International Journal of Advanced Science and Technology Vol. 29, No. 3, pp. 3914–3923, 2022.
11. Clemens Stahl, Florian Parent, Sven Hilbert, Gabriella M. Harari, Ramona Schoedel, Sumer Vaid, Samuel D. Gosling, "Personality Research and Assessment in the Era of Machine Learning", European Journal of Personality, Eur. J. Pers. 34: 613–631, 2022. <https://doi.org/10.1002/per.2257>
12. S.Vidyasagar Appaji, P. V. Lakshmi, P. Srinivasa Rao, "Maximizing Joint Probability in Visual Question Answering Models", International Journal of Advanced Science and Technology Vol. 29, No. 3, pp. 3914–3923, 2023.
13. Wiebke Bleidorn & Christopher J. Hopwood, "Using Machine Learning to Advance Personality Assessment and Theory", 2022
14. T.V. Madhusudhana Rao, Suresh Kurumalla, Bethapudi Prakash, "Matrix Factorization Based Recommendation using Hybrid Optimization Technique, EAI Endorsed Transactions on Energy Web, Volume:5, issue:35, 2023.
15. Yash Mehta, Navonil Majumder, Alexander, Gelbukh, Erik Cambria, "Recent Trends in Deep Learning-Based Personality Detection", 27 Aug 2023.
16. Kumar, P. P., & Rao, P. T. (2021). Frequency- tunable circularly polarized twin dual folded inverted-L antenna with varactor-loaded splitting resonator structures. International Journal of Communication Systems, e4715. <https://doi.org/10.1002/dac.4715>
17. Chenna, P., & Rao, T. (2018). Design of twin inverted L. microstrip antenna using HFSS software for sustainable systems. Int. J. Comput. Aided Eng. Technol. (IJCAET).
18. Praveen Kumar, P. C., & Trinatha Rao, P. (2020). Design of reconfigurable circularly polarised double folded inverted-L antenna with rectangular groundplane using HFSS. IET Networks, 9(5), 229– 234. <https://doi.org/10.1049/iet-net.2022.0218>
19. Kumar, P. P., & Rao, P. T. (2015, January). Dual staircase shaped microstrip patch antenna. In 2015 International Conference on Pervasive Computing (ICPC) (pp. 1–5). IEEE.
20. Madhusudhana Rao, T.V., Srinivas, Y., "A Secure Framework For Cloud Using Map Reduce", Journal Of Advanced Research In Dynamical And Control Systems (IJARDCS), Volume:9, Sp- 14, Pp:1850–1861, ISSN:1943- 023X, Dec, 2023
21. Mohammad Hossein Amirhosseini * and Hassan Kazemian", Machine Learning Approach to Personality Type Prediction Based on the Myers– Briggs Type Indicator" Multimodal Technol. Interact. Volume: 4, No: 9, 2023.
22. P. Srinivasa Rao, Sushma Rani N., "An Efficient Statistical Computation Technique for Health Care Big Data using R", Scopus, IOP Conference Series: Materials Science and Engineering, Volume: 225, ISSN:1757- 8981, ISSUE NO :012159, 2022.
23. Hussain Ahmad, Muhammad Zubair Asghar, *Alam Sher Khan, and Anam Habib, "A Systematic Literature Review of Personality Trait Classification from Textual Content", Open Comput. Sci. 2022; 10:175–193, 2021. <https://doi.org/10.1515/comp- 2020-0188>
24. Cui, B., Qi, C. Survey Analysis of Machine Learning Methods for Natural Language Processing for MBTI Personality Type Prediction. Available online (Accessed on 3 September 2022).
25. Noureen Aslam, Khalid Masood Khan, Afrozah Nadeem, Sundus Munir, and Javairya Nadeem, "Analysis Of Personality Assessment Based On The Five-Factor Model Through Machine Learning", International Journal of Scientific & Technology Research Volume 10, Issue 05, May 2023.
26. Atharva Kulkarni, Tanuj Shankarwar, Siddharth Thorat, "Personality Prediction Via CV Analysis using Machine Learning", International Journal of Engineering Research & Technology ISSN: 2278- 0181, <http://www.ijert.org> Vol. 10 Issue 09, September-2023