

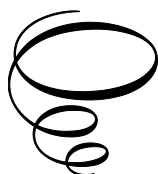
Natural Language
Processing (NLP)
in Understanding
Human Emotion for
Predicting Human
Psychology

Natural Language Processing (NLP) in Understanding Human Emotion for Predicting Human Psychology

Edited by

Jay Kumar Pandey, Mritunjay Rai
and Faizan Ahmad

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TABLE OF CONTENTS

Acknowledgements	viii
Foreword.....	ix
Preface	x
List of Abbreviations	xii
List of Tables	xvii
Chapter 1	1
Bridging Minds and Machines: Philosophical Foundations of Natural Language Processing in Understanding Human Emotion <i>Barbara Gabriella Renzi and Kaddour Chelabi</i>	
Chapter 2	22
Exploring Human Emotion and Psychology through NLP <i>Vivek Veeraiah, Ritu Dahiya, Veera Talukdar, Jay Kumar Pandey and Ankur Gupta</i>	
Chapter 3	42
Enhancing Human Computer Interaction through Emotion Analysis <i>Kanthavel R, Adline Freeda R and Dhaya R</i>	
Chapter 4	67
Advancing Psychological Assessments with Natural Language Processing: A New Era of Precision and Insight <i>Zainab Khan and S Reshma Jamal</i>	
Chapter 5	89
Innovations for Emerging Trends in NLP and Mental Health <i>Bhanu Dwivedi, Abhishek Kumar Saxena and Sunny Kumar</i>	

Chapter 6	120
Role of AI and NLP in Improving Mental Healthcare Through Personalized Therapy	
<i>Megha Agarwal, Nidhi Tiwari and Abhijityaditya Prakash</i>	
Chapter 7	147
The Potential of Natural Language Processing for Enhancing Autism Diagnosis: A Comprehensive Review	
<i>Nisha Banerjee, Neha Majumder</i>	
Chapter 8	170
Understanding Bipolar Disorder: Analyzing Emotional Fluctuations through NLP	
<i>Saikat Santra I, Kushal Roychoudhuri, Bhaskar Pal, Rajdip Goswami and Soumyadeb Dutta</i>	
Chapter 9	199
Recent Trends and Developments in NLP	
<i>Rahul Kumar Ghosh, Sandip Chakraborty, Gourab Dutta and Partha Pratim Dasgupta</i>	
Chapter 10	221
Advancements and Challenges in Sentiment Analysis for Multilingual and Regional Contexts	
<i>Ramesh Ram Naik, Sanah Sayyed, Kalpana B. Khandale, C. Namrata Mahender, Parth R. Gharat and Ms. Ashwini Dattatraya Gopwad</i>	
Chapter 11	258
Advancing Education through Computational Intelligence: Insight from Literature Review	
<i>Ankur Nandi, Santu Karmakar, Tapash Das and Tarini Halder</i>	
Chapter 12	284
Applications of Natural Language Processing in Mental Health	
<i>Ritwik Raj Saxena</i>	
Chapter 13	317
Computational Intelligence (CI) In Techno-Pedagogical Education (TPE) for Enhance Technological Learning	
<i>Shishupal Gorain and Satish Kumar Kalhotra</i>	

Chapter 14	337
Exploring Psycholinguistics Through AI, Ayurveda, and Contemporary Medicine: A Multidisciplinary Framework <i>C. Namrata Mahender, Parth Gharat, Sridhar Raj Sankara Vadivel, Ashwini Gopwad, Rasika Kolhe and Aarti P. Raut</i>	
List of Contributors	361

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Finally, I wish to acknowledge the contributions of the academic and research communities, whose work has been instrumental in shaping the ideas presented in this book. I am also grateful to the editorial team for their diligence and dedication, ensuring the clarity and coherence of this work. To the readers, thank you for your interest and curiosity it is for you that this book has been written.

FOREWORD

The fusion of technology and psychology has given rise to innovative approaches in understanding and addressing the complexities of human emotions and behavior. As we navigate a rapidly evolving digital era, the role of Natural Language Processing (NLP) in decoding human emotion and predicting psychological patterns cannot be overstated. This book, “NLP in Understanding Human Emotion for Predicting Human Psychology,” is a timely and insightful contribution to this emerging interdisciplinary domain.

The unique strength of this book lies in its ability to connect theoretical foundations with practical applications. The author’s exploration of topics such as sentiment analysis, emotion detection, and psycholinguistic profiling demonstrates how NLP tools can be applied to solve real-world challenges, from mental health support to enhancing user experience in digital platforms. By bridging the gap between computational methodologies and human-centered insights, this book lays the groundwork for future research and innovation.

What makes this book particularly compelling is its balanced perspective on the promises and pitfalls of NLP. While highlighting the transformative potential of these technologies, the author does not shy away from addressing the ethical and technical challenges they pose. The discussions on bias in NLP systems, privacy concerns, and the need for responsible implementation underscore the importance of mindful innovation in this field.

As a researcher and practitioner in this area, I believe this book will serve as an essential resource for anyone interested in exploring the intersection of NLP and psychology. Whether you are a student, a seasoned professional, or an academic researcher, the insights offered here will inspire you to think critically and creatively about the ways technology can enhance our understanding of human emotion and behavior.

I congratulate the author for this remarkable work and encourage readers to immerse themselves in its pages. The knowledge and ideas presented in this book have the potential to shape the future of how we interact with and understand one another in an increasingly digital world.

PREFACE

The human mind is a labyrinth of thoughts, emotions, and behaviors, a complex interplay of conscious and subconscious processes that define our interactions, decisions, and perceptions. Understanding human emotion has always been at the forefront of psychology and neuroscience, but the advent of Natural Language Processing (NLP) has opened new frontiers in decoding the nuances of human emotions and predicting psychological patterns. This book, “**NLP in Understanding Human Emotion for Predicting Human Psychology,**” aims to bridge the gap between computational linguistics and psychological analysis, offering a comprehensive exploration of how NLP can be leveraged to unravel the complexities of human emotion.

Over the last decade, NLP has evolved from a niche area of artificial intelligence to a transformative force, impacting diverse domains such as healthcare, education, business, and mental health. The ability of NLP models to process and interpret vast amounts of textual and speech data has revolutionized the way we understand human behavior. By analyzing sentiment, tone, and context in communication, NLP provides unprecedented insights into emotional states and psychological tendencies, enabling applications ranging from mental health diagnostics to personalized human-computer interaction.

This book is designed for researchers, practitioners, and students who are keen to describe into the intersection of NLP and psychology. It covers foundational concepts, advanced techniques, and real-world applications of NLP in understanding and predicting human emotions. Each chapter is structured to provide theoretical insights followed by practical applications, ensuring that readers gain both conceptual knowledge and hands-on expertise.

We explore critical topics such as sentiment analysis, emotion detection, psycholinguistic profiling, and the ethical considerations of deploying NLP in sensitive areas. Case studies and examples illustrate how NLP has been successfully applied to address challenges in mental health, customer experience, and social behavior analysis. The book also highlights the limitations and biases inherent in NLP systems, emphasizing the need for responsible innovation.

I extend my gratitude to the community of researchers and practitioners whose pioneering work has laid the foundation for this field. Their contributions have inspired the discussions and insights presented in this

book. I also thank the readers for their interest in this fascinating domain, as their curiosity and engagement will drive future advancements.

It is my hope that this book serves as a valuable resource, sparking new ideas and fostering collaborations at the nexus of technology and psychology. Together, let us explore the transformative potential of NLP in understanding the human mind and improving the quality of life for individuals and societies alike.

Editors:
Dr. Jay Kumar Pandey
Dr. Mritunjay Rai
Dr. Faizan Ahmad

LIST OF ABBREVIATIONS

ABC	Artificial Bee Colony
ACT	Acceptance and Commitment Therapy
ADHD	Attention Deficit Hyperactivity Disorder
ADI-R	Autism Diagnostic Interview-Revised
ADOS	Autism Diagnostic Observation Schedule
AI	Artificial Intelligence
AMI	Any Mental Illness
ANN	Artificial Neural Networks
APA	American Psychological Association
AR	Augmented Reality
ASD	Autism Spectrum Disorder
ASR	Automatic Speech Recognition
ADHD	Attention-Deficit/Hyperactivity Disorder
BDI	Beck Depression Inventory
BA	Bat Algorithms
BD	Bipolar Disorder
BD-I	Bipolar Disorder Type I
BD-II	Bipolar Disorder Type II
BERT	Bidirectional Encoder Representations from Transformers
BiGRU	Bidirectional Gated Recurrent Unit
BiLSTM	Bidirectional Long Short-Term Memory
BiRNN	Bidirectional Recurrent Neural Network
BLEU	Bilingual Evaluation Understudy
BPD	Bipolar Disorder (context-specific use here; can also be used for Borderline Personality Disorder)
BPD	Borderline Personality Disorder
CAT	Computed Tomography
CBT	Cognitive Behavioral Therapy
CDC	Centers for Disease Control and Prevention
CDSS	Clinical Decision Support System
CHRV	Coronary Heart Rate Variability
CYC	Cyclothymia
CI	Computational Intelligence
CNN	Convolutional Neural Network

C-NN	Condensed Nearest Neighbors
CRF	Conditional Random Field
DALY	Disability-Adjusted Life Year
DBT	Dialectical Behavior Therapy
DPA	Data Protection Act
DSM IV TR	Diagnostic and Statistical Manual of Mental Disorders, fourth edition, text revision
DSM-V	Diagnostic and Statistical Manual of Mental Disorders, Fifth Edition
DT	Decision Trees
ECG	Electrocardiogram
EDA	Electro dermal Activity
EDA	Exploratory Data Analysis
EEG	Electroencephalogram
ECT	Electroconvulsive Therapy
EHR	Electronic Health Records
EI	Emotional Intelligence
EM	Exact Match
EMA	European Medicines Agency
EMG	Electro muscular
E-NN	Edited Nearest Neighbors
ESG	Environmental, Social, and Governance
EX	User Experience
FACS	Facial Action Coding System
FCNN	Fuzzy Convolution Neural Network
FDA	Food and Drug Administration
FKNN	Fuzzy K-Nearest Neighbor
FPA	Flower Pollination Algorithm
GAN	Generative Adversarial Network
GDPR	General Data Protection Regulation
GLUE	General Language Understanding Evaluation
GNN	Graph Neural Network
GPT	Generative Pre-train Transformers
GRU	Gated Recurrent Unit
GWO	Gray Wolf Optimization
HRQoL	Health-Related Quality of Life

HPA Axis	Hypothalamic-Pituitary-Adrenal Axis
HCI	Human-computer interface
HER	Electronic Health Records
HIPAA	Health Insurance Portability and Accountability Act
HMM	Hidden Markov Model
HRV	Heart Rate Variability
ICD	International Classification of Diseases
ICMLCA	International Conference on Machine Learning and Computer Applications
ICSE	International Conference on Software Engineering
ID3	Iterative Dicotomiser 3
IED	Intermittent Explosive Disorder
IQ	Intelligence Quotient
ISL	Internal State Language
ITS	Intelligent Tutoring Systems
KNN	K--Nearest Neighbors
KSVM	Kernel Support Vector Machines
LASSO	Least Absolute Shrinkage and Selection Operator
LCM	Linguistic Complexity Metric
LIME	Local Interpretable Model-Agnostic Explanations
LLM	Large Language Models
LDA	Latent Dirichlet Allocation
LOPOCV	Leave-One-Person-Out Cross Validation
LSA	Latent Semantic Analysis
LSM	Language Style Matching
LSTM	Long Short-Term Memory Networks
MDE	Major Depressive Episode
MI	Motivational Interviewing
MLM	Masked Language Model
MRI	Magnetic Resonance Imaging
MRR	Mean Reciprocal Rank
ML	Machine Learning
MDD	Major Depressive Disorder
MSE	Mental State Examination
MTL	Multi-Task Learning
NIMH	National Institute of Mental Health

NLG	Natural Language Generation
NLP	Natural Language Processing
NIMH	National Institute of Mental Health
NSDUH	National Survey on Drug Use and Health
NSP	Next Sentence Prediction
PII	Personally Identifiable Information
POS	Part of Speech
PPA	Primary Progressive Aphasia
PTSD	Post-Traumatic Stress Disorder
REBT	Rational Emotive Behavior Therapy
ReLU	Rectified Linear Units
RF	Random Forest
RNN	Recurrent Neural Network
ROUGE	Recall-Oriented Understudy for Gisting Evaluation
SA	Sentiment Analysis
SSRIs	Selective Serotonin Reuptake Inhibitors
SAMHSA	Substance Abuse and Mental Health Services Administration
Seq2seq	Sequence To Sequence
SGD	Stochastic Gradient Descent
SHAP	Shapley Additive Explanations
SMI	Serious Mental Illness
SMOTE	Synthetic Minority Oversampling Technique
STEM	Science Technology Engineering and Management
SOCC	System-on-Chip Conference
SSL	Self-Supervised Learning
SVM	Support Vector Machine
SUD	Substance Use Disorder
Super GLUE	Super General Language Understanding Evaluation
SVM	Support Vector Machine
TPE	Techno-pedagogical Education
T5	Text-to-Text Transfer Transformer
TD	Typical Development
TF-IDF	Term Frequency-Inverse Document Frequency
TTS	Text-to-Speech
QoL	Quality of Life
QA	Question Answering

UCD	User-Concentrated Design
US	United States
VR	Virtual Reality
WEIRD	Western, Educated, Industrialized, Rich, Democratic
XAI	Explainable Artificial Intelligence
XNLI	Cross-lingual Natural Language Inference

LIST OF TABLES

Table 2.1: Literature Survey human emotion and psychology

Table 3.1: Case Study Comparison chart

Table 7.1: Inclusion Exclusion Criteria

Table 7.2: NLP Approaches for Autism Diagnosis

Table 7.3: Factors Contributing to Bias in NLP Models for Autism Diagnosis

Table 8.1 Comparison of Traditional Diagnostic Methods and NLP Approaches

Table 8.2: Potential Applications of NLP in Clinical Settings

10.7.1 Marathi Sentiment Analysis Results

Table10.1: Performance Summary of Sentiment Analysis Models on the Marathi Dataset

Table10.2: The results for the Gujarati dataset

Table 14.1: Ayurveda's view of Psycholinguistics with Modern Psychology

CHAPTER 1

BRIDGING MINDS AND MACHINES: PHILOSOPHICAL FOUNDATIONS OF NATURAL LANGUAGE PROCESSING IN UNDERSTANDING HUMAN EMOTION

BARBARA GABRIELLA RENZI
AND KADDOUR CHELABI

Abstract

The advancement of Natural Language Processing (NLP) has led to a significant transformation in how machines interpret and engage with human language. This progress, while predominantly technical, has also brought to the forefront complex questions about the nature of language, thought, and emotion. NLP's goal to enable machines to understand, analyse, and respond to human communication has revealed deeper philosophical challenges that go beyond mere algorithmic processing. As machines attempt to decode emotions expressed through language, they must grapple with centuries-old philosophical debates about consciousness, the mind-body connection, and the contextual meaning of emotions.

This paper seeks to bridge the gap between the technical aspects of NLP and the philosophical foundations that inform our understanding of language and emotion. By examining the works of influential philosophers such as Aristotle, Descartes, and Heidegger, the paper will explore how their views can inform and challenge the current approaches of NLP systems in interpreting emotions. Aristotle's perspective on emotions as rational guides to behavior aligns with the structured way NLP models analyse language. Descartes' dualism, however, raises questions about whether machines, devoid of a physical body, can truly understand emotions. Heidegger's existential view on the contextual nature of

emotions further complicates this, suggesting that emotions are deeply intertwined with our broader experience of the world. Through this exploration, the paper aims to highlight the potential and limitations of NLP, arguing that its engagement with emotion is not merely a technical pursuit but a continuation of philosophical inquiries into the nature of human consciousness and experience.

Keywords: Emotion recognition, Contextual analysis, Embodiment, Sentiment analysis, Dualism.

1.1 Introduction

The field of Natural Language Processing (NLP) represents a fascinating convergence of technology and humanistic inquiry, particularly when it comes to understanding the nuances of human language, thought, and emotion (Manning et al. 2008). At its core, NLP seeks to enable machines to process and interpret human language, allowing for applications ranging from chatbots and sentiment analysis to advanced language translation (Jurafsky and Martin 2024). However, the quest to make machines understand and respond to human emotions reveals deeper, more complex challenges that extend beyond mere technical problem-solving. These challenges are rooted in long-standing philosophical questions about the nature of language, emotion, and consciousness (Carruthers 1996, 1-8).

This chapter explores how philosophical frameworks can provide valuable insights into the capabilities and limitations of NLP when it comes to interpreting human emotions. By drawing on the works of influential philosophers such as Aristotle, Descartes, and Heidegger, the analysis highlights different perspectives on the relationship between emotion and reason, the mind-body connection, and the contextual nature of human experience (Solomon 1993, 1-26). Each of these perspectives offers a unique lens through which to assess the potential of machines to truly understand or replicate human emotions. Aristotle's philosophy, which sees emotions as rational and interconnected with reason, aligns with the way NLP models attempt to decode emotional content from language (Nussbaum 2001, 1-16). Conversely, Descartes' dualistic approach, emphasizing the separation of mind and body, raises questions about the feasibility of machine-based emotional comprehension. Heidegger's existential framework, which views emotions as deeply intertwined with the broader context of human experience, challenges the ability of machines to fully grasp the subtleties of human feelings (Dreyfus 1990, 14-15).

This introduction sets the stage for a detailed exploration of these philosophical perspectives and their relevance to the future of NLP in understanding and replicating human emotions.

1.2 Aristotle’s Perspective on Emotions and Rationality

Aristotle’s theory of emotions, as articulated in his seminal work *Rhetoric*, presents emotions as integral to human reason and rationality, rather than as irrational forces to be suppressed or ignored (Aristotle trans. 2004). According to Aristotle, emotions are appropriate and reasoned responses to specific situations, guiding human behavior and decision-making. For instance, feelings such as anger, fear, or joy are not random outbursts but are deeply connected to how individuals perceive and respond to their circumstances (Fortenbaugh 2002, 9-22). This interconnection between emotion and reason allows for emotions to be predictable, understandable, and, to some extent, manageable (Solomon 1993, 3-28). Aristotle’s view suggests that emotions can be systematically analysed and understood, as they follow patterns that can be deciphered through careful observation of language and context.

This perspective offers a compelling framework for understanding how NLP models can interpret emotional content from text. In many ways, NLP systems echo Aristotle’s approach by treating emotions as data that can be analysed and categorized (Calvo et al. 2015). By using algorithms that parse language, NLP can detect patterns that signify particular emotional states. These systems can analyse word choices, sentence structures, and contextual cues to infer emotions, much like how Aristotle believed emotions could be rationally understood based on their causes and expressions (Jurafsky and Martin 2024). For example, an NLP model might recognize phrases like “I am thrilled” or “This is disappointing” and classify them as expressions of happiness or sadness, respectively.

Modern NLP models, especially those leveraging machine learning, apply a form of rationalization similar to Aristotle’s when analysing text for emotional content (Poria et al. 2017). They do this by being trained on large datasets where language is labelled with emotional states. Through this training, the algorithms learn to identify subtle indicators of emotion, such as positive or negative sentiment, intensity, and context. For instance, advanced sentiment analysis tools use techniques like word embedding’s where words are represented as vectors in a high-dimensional space. These vectors capture semantic relationships between words, allowing the system to infer emotions even when they are not explicitly stated (Mikolov et al.

2013). A sentence like “The news left me feeling numb” might be identified as conveying sadness or despair, even though the word “sad” is not used.

Moreover, NLP models often rely on contextual understanding to interpret emotions accurately (Cambria et al. 2017, 71-92). For example, a phrase like “I’m fine” might typically be seen as neutral or positive, but if accompanied by sarcastic or negative words in the surrounding text, it could be inferred as conveying frustration or sadness. In this way, NLP systems attempt to replicate the rational analysis of emotional content that Aristotle described, seeking to decode the underlying sentiments through language patterns and context (Pang and Lee 2008). By rationalizing emotions in text, these systems demonstrate the practical application of Aristotle’s notion that emotions are not arbitrary but can be understood within a framework of reason and causality.

Through this lens, Aristotle’s philosophy provides a foundational rationale for the development of NLP models focused on emotion (Nussbaum 2001, 139-173). By approaching emotions as rational and systematic, it becomes possible to teach machines how to recognize and interpret them, creating a bridge between human experience and computational understanding. However, as will be explored further, this rationalization has its limits, especially when considering the deeper, embodied aspects of emotions that extend beyond mere linguistic expression (Dreyfus 1990, 56–91).

1.3 Descartes’ Dualism and the Challenges for Machine Understanding

René Descartes, a central figure in modern philosophy, is well known for his theory of dualism, which posits a fundamental distinction between the mind (the domain of thought and reason) and the body (the domain of physical processes and sensations) (Descartes 1641/2017, 57-72). In his *Meditations on First Philosophy*, Descartes argued that the mind, or soul, is a non-material entity that governs rational thought, while the body operates within the physical world, responding to sensory stimuli. This separation implies that mental experiences, including emotions, are deeply intertwined with bodily states, even though they are processed by the rational mind (Cottingham 1996, 122-136).

In the context of NLP, Descartes’ dualism presents significant challenges. Machines, as non-embodied entities, lack the physiological basis that humans have for experiencing emotions. Human emotions are not purely cognitive but are often accompanied by physical sensations and bodily reactions like a racing heart during anxiety or a warm sensation of

joy (Damasio 1994 3-11). According to Cartesian dualism, this physical aspect of emotion is inseparable from the experience itself. Therefore, a machine that lacks a body cannot fully replicate or understand the complete nature of emotional experiences, as it misses the physiological dimension that contributes to human feelings (Shanahan 2016, 191-214).

This raises critical questions about the limitations of machine understanding when it comes to emotions. While NLP systems can analyse and interpret textual data to infer emotional states, they do so by identifying patterns and correlations rather than truly experiencing emotions (Poria et al. 2017). For instance, an NLP model might recognize that phrases like “I feel stressed” or “I can’t stop crying” are indicative of anxiety or sadness, but this recognition is based on statistical learning rather than a direct experience of those states (Jurafsky and Martin 2024). The model processes emotional content by analyzing words, syntax, and context, but it does not “feel” anxiety or sadness because it has no physical or sensory interface to mirror human emotional experiences (Calvo et al. 2015, 1-8).

Descartes’ philosophy underscores this limitation, suggesting that true understanding of emotions requires an embodied experience, where the mind and body interact (Cottingham 1996, 141-210). Machines, devoid of this interaction, can only simulate understanding based on patterns rather than genuinely comprehending the full emotional spectrum. For example, while an advanced NLP system can detect sarcasm or subtle emotional undertones by learning from large datasets, it is still merely mapping statistical probabilities rather than intuitively grasping the emotional complexities (Cambria et al. 2017, 172-188). In a sense, it is mimicking the outward signs of understanding without accessing the internal, felt experience that humans have when they process emotions.

This philosophical stance raises important questions about the depth of machine comprehension. If emotions are indeed deeply connected to the physical responses of the body, as Descartes would argue, then machines that lack a body are fundamentally limited in their ability to understand emotions in the same way humans do (Dreyfus 1990, 92-129). This has ethical and practical implications, especially in applications where machines are expected to interact empathetically with humans, such as in mental health chatbots or customer service agents (Crawford and Calo 2016). Can a machine that merely processes language data ever truly engage with the complexities of human emotions, or is it perpetually confined to surface-level approximations?

In summary, Descartes’ dualistic view highlights a critical limitation of NLP systems: while they can analyse and respond to emotional content in text, their lack of physical embodiment means they cannot fully under-

stand or replicate the human experience of emotions (Descartes 1641/2017, 28-41). The separation of mind and body, as posited by Descartes, suggests that true emotional understanding involves more than just cognitive processing. It is also about how the mind interprets the signals from the body (Hatfield 2018, 1-6). This perspective calls into question the depth of machine comprehension and underscores the inherent challenges of attempting to bridge the gap between computational systems and the embodied human experience of emotion.

1.4 Ludwig Wittgenstein: Language as a Form of Life: The Contextual Limits of NLP

Ludwig Wittgenstein's philosophy offers a profound critique of how language is understood, challenging the notion that meaning can be fully captured through static definitions or purely linguistic structures. In his later work, particularly in *Philosophical Investigations* (1953), Wittgenstein introduces the concept of language as a "form of life," emphasizing that the meaning of words is not an intrinsic property but is derived from their use within specific social practices and contexts. According to this view, language is fundamentally tied to the activities, behaviours, and interactions of people within a community. This means that understanding language is not merely about deciphering words but about grasping the underlying practices and forms of life that give those words their meaning (Wittgenstein 1953, Remarks 1-43).

Wittgenstein's idea that "meaning is use" suggests that understanding language, and by extension emotions expressed through language, requires an appreciation of the broader social and cultural contexts in which language is embedded. For instance, the word "love" might have different connotations depending on whether it is used in a romantic, familial, or platonic context, and these nuances cannot be fully understood without participating in, or at least being familiar with, the social practices that shape these forms of expression. Therefore, meaning is not static; it is dynamic and evolves as language is used in various contexts. This poses a fundamental challenge for NLP systems, which primarily rely on analysing textual patterns and correlations without engaging in the social practices that inform the meaning of those patterns (Bender Koller 2020).

Wittgenstein's approach raises important questions about the contextual limits of NLP. While modern NLP models can be trained on vast datasets and learn to associate certain phrases with specific emotional states, they do not participate in the "form of life" that human beings do. For example, an NLP system may be able to recognize that the phrase "I am fine" can some-

times indicate well-being and, in other contexts, may be an expression of sarcasm or discomfort. However, it arrives at this conclusion through statistical patterns and probabilities rather than through an intuitive understanding of social cues, body language, tone, and shared experiences those humans rely on to interpret such phrases accurately (Cambria et al. 2017, 47–70). This mechanistic approach to language processing can lead to misinterpretations, particularly when dealing with subtle or complex emotional expressions that are deeply rooted in cultural practices or require a shared background of experiences to understand fully (Hovy Spruit 2016).

Furthermore, Wittgenstein’s perspective suggests that much of what we convey through language goes beyond the literal content of our words. Emotions, attitudes, intentions, and even social roles are often communicated implicitly, through the way language is used in interaction rather than through explicit statements. Consider, for example, the way humour or irony operates: these forms of expression depend heavily on the shared assumptions and context between speaker and listener. An NLP system might struggle to detect irony because it lacks the social background knowledge to recognize when a statement is meant to be taken literally or humorously. For Wittgenstein, this inability to grasp the subtleties of language use underscores a fundamental limitation of trying to capture meaning purely through text analysis, as NLP systems attempt to do (Wittgenstein 1953, Remarks 23–43).

The concept of language as a form of life also implies that understanding language involves more than processing words; it requires engaging in the practices and contexts that give words their meaning. For human beings, communication is not just a matter of exchanging information but also a way of building relationships, establishing social norms, and navigating complex social dynamics (Kripke 1984, 7-54). When two people communicate, they draw on a wealth of shared experiences, cultural norms, and situational awareness that inform how they interpret each other’s words. Machines, however, do not share this rich background of experiences. While they can be programmed to recognize patterns in language, they do not truly understand the social functions of communication because they are not participants in the form of life that humans share (Bender and Koller 2020). This lack of social embeddedness means that even the most sophisticated NLP systems are limited in their ability to understand the deeper, contextual meanings of language.

For example, consider a situation where a friend says, “I can’t believe you did that!” Depending on the context—whether it was said after a surprising achievement, a playful prank, or a serious mistake—the meaning of this phrase can vary greatly. Humans can easily interpret this be-

cause they are attuned to the tone, the situation, and their relationship with the speaker. An NLP system might recognize the words and even detect a potential positive or negative sentiment based on the dataset it was trained on, but it would miss the nuances that make the interaction meaningful. This limitation becomes even more pronounced when trying to understand emotions that are deeply tied to cultural contexts, such as expressions of honour, shame, or love that might have different connotations across societies (Poria et al. 2017).

Wittgenstein's philosophy suggests that there is a performative aspect to language use: we do not just describe emotions; we express them, enact them, and sometimes even create them through our interactions. This performative quality is something that NLP systems cannot replicate because they do not engage in the act of communication as participants. They can simulate responses based on learned data but lack the intentionality and expressiveness that characterize human emotional communication (Searle 1980). As a result, while machines can analyse patterns in language and even predict emotional content with a degree of accuracy, they do not truly understand what it means to feel or to convey an emotion. They lack shared human experiences that inform our understanding of why someone might feel a certain way or how best to respond to that emotion (Hovy and Spruit 2016).

In conclusion, Wittgenstein's concept of language as a form of life highlights the contextual limitations of NLP systems in understanding emotions. While these systems can identify and respond to linguistic patterns, they do so without the social context and shared experiences that are essential to genuine understanding. This suggests that no matter how advanced NLP becomes, there will always be aspects of human emotional communication that elude machine comprehension. To bridge this gap, developers might need to consider ways to integrate more contextual and cultural awareness into AI systems, though this poses its own set of challenges. Ultimately, Wittgenstein's insights remind us that language is not just a tool for transmitting information but a complex, socially embedded practice that cannot be fully reduced to algorithms and data patterns (Wittgenstein 1953, Bender and Koller 2020).

1.5 Heidegger's Existential Contextualization of Emotions: Understanding Moods Beyond Language

In his seminal work *Being and Time*, Martin Heidegger introduced a unique perspective on emotions, framing them as "moods" (*Stimmungen*) that fundamentally shape our experience of the world (Heidegger

1927/1962). Unlike traditional views that treat emotions as mere internal responses to external stimuli, Heidegger argued that moods are existentially significant; they reveal how we are attuned to the world, influencing our perception, thoughts, and actions (Dreyfus 1990). Moods, for Heidegger, are not just reactions but modes of being that disclose our situatedness and engagement with the world around us (Ratcliffe 2008, 56–91). This existential contextualization emphasizes that emotions are not isolated events but are embedded in a broader framework of our existence (Heidegger 1927/1962, 438–488).

Heidegger's view poses a significant challenge to modern NLP systems, which aim to analyse and interpret emotions based on textual data. While machines have become adept at recognizing and categorizing basic emotions like happiness, sadness, or anger, they struggle to grasp the more nuanced and contextual aspects of moods (Cambria et al. 2017, 71–92). This is because, in Heidegger's sense, moods cannot be fully captured through discrete, definable categories; they are not merely cognitive states but are deeply intertwined with our being-in-the-world. For example, the mood of anxiety in Heidegger's philosophy is not just a feeling of unease but a fundamental awareness of the contingency and openness of existence, something that cannot be easily parsed into data points (Heidegger 1927/1962, 274–311).

This brings us to the limits of language-based emotional analysis by machines. NLP systems rely on language patterns, word frequencies, and syntactic structures to infer emotions, but such approaches often miss the existential depth that Heidegger attributes to moods (Poria et al. 2017). Machines can identify keywords associated with sadness or joy, but they cannot grasp the underlying mood that informs how those emotions are experienced within a particular context. For instance, the same phrase might carry different emotional connotations depending on the speaker's mood, historical context, or existential state, making it difficult for algorithms to accurately interpret the richness of human emotions (Crawford and Calo 2016).

Moreover, language itself can be an imperfect vessel for conveying the full depth of moods. Heidegger believed that moods often reveal themselves indirectly, through actions, silence, or subtle shifts in attention, rather than explicit statements (Ratcliffe 2008, 92–128). This non-verbal dimension of moods poses an additional challenge for NLP systems, which are primarily designed to process and analyse textual information (Cambria et al. 2017). Therefore, while machines can become increasingly proficient at detecting explicit emotional expressions, they remain limited

in their ability to understand the existential layers of moods as described by Heidegger (Dreyfus 1990, 179 - 255).

In conclusion, Heidegger's existential contextualization of emotions as moods reveals the profound gap between human emotional experience and machine interpretation. Moods, as fundamental ways of being, resist reduction to simple, analysable units, and their complexity highlights the limitations of current NLP approaches (Mulhall 2013, 105-137). As we continue to develop more sophisticated systems for emotional analysis, we must recognize these philosophical insights, acknowledging that true understanding of emotions may require more than just parsing language it may require a more holistic grasp of the human condition that machines are not yet equipped to achieve (Ratcliffe 2008, 61–104).

1.7 Gilles Deleuze and Félix Guattari: Emotions as Flows - Moving Beyond Individualized Analysis

Gilles Deleuze and Félix Guattari's philosophical framework offers a radically different approach to understanding emotions, one that moves away from viewing them as fixed, isolated states and instead conceptualizes them as dynamic flows that traverse both individual and collective bodies. In their works, particularly *A Thousand Plateaus* (1987), Deleuze and Guattari challenge traditional notions of identity, subjectivity, and emotion, proposing a model that sees emotions as processes that are constantly in motion, interacting with and influencing various entities and environments. This perspective can greatly enrich the discourse around NLP by suggesting new ways of designing systems that can capture the fluid, networked nature of emotional expression.

Deleuze and Guattari's idea of "assemblages" is central to this understanding. An assemblage refers to a network of interconnected elements that come together to form a temporary, yet functional, entity. Emotions, in this sense, are not properties confined within a single person but are flows that can be distributed across multiple bodies, including humans, machines, and environments (Deleuze and Guattari 1987, 351-423). For example, the joy felt at a concert or the shared outrage on social media platforms are instances where emotions are not just personal experiences but collective phenomena. Such a perspective challenges the reductionist view that emotions can be neatly categorized into discrete states like happiness, sadness, or anger. Instead, emotions are seen as ever-evolving, multi-layered processes that emerge from interactions within complex systems.

In the context of NLP, this understanding pushes for a shift from traditional models that attempt to classify emotions into fixed categories

towards more sophisticated systems that can recognize and analyse the fluidity and multiplicity of emotions as they manifest across different contexts. Current NLP techniques often rely on sentiment analysis, which categorizes text as positive, negative, or neutral based on the presence of specific keywords or patterns (Poria et al. 2017). However, this approach can miss the nuances and overlapping qualities of real-world emotional expressions, where emotions are not always neatly separable. For instance, a tweet expressing sarcasm might contain both elements of humour and frustration, which a standard sentiment analysis model might fail to capture accurately.

By adopting Deleuze and Guattari's perspective, NLP systems could be designed to account for the interconnectedness and flow of emotions across digital and social platforms. Rather than merely focusing on individual expressions, these systems could analyse how emotions propagate, evolve, and transform within networks. This could be particularly useful in understanding phenomena like viral trends, where emotions are not just responses but active forces that spread and mutate across communities. For example, an incident that sparks a wave of outrage online may not remain static; the initial anger could transform into calls for action, solidarity, or even humour, as different groups engage with and reinterpret the original event. Understanding this dynamic process requires models that can track the evolution of emotions over time and across different platforms, rather than treating them as isolated incidents (Massumi 2002).

Deleuze and Guattari also introduce the concept of "rhizomes," which refers to non-hierarchical, decentralized networks that can expand in multiple directions without a clear beginning or end. Emotions, viewed as rhizomatic, are not linear progressions but can branch out, intersect, and recombine in unpredictable ways (Deleuze and Guattari 1987 3-25). This challenges NLP systems to move beyond linear models of language processing and consider the complexity of emotional networks that may not follow predictable patterns. For example, during a global event like a pandemic, fear, hope, and solidarity can spread simultaneously through various channels, interacting and influencing each other in ways that are not easily traceable or reducible to simple cause-and-effect relationships. An NLP system inspired by the rhizomatic approach would need to identify these non-linear connections and patterns to provide a more holistic understanding of emotional dynamics.

The concept of emotions as flows also aligns with the way emotions manifest in digital environments. Social media platforms, for instance, serve as conduits where emotions are shared, amplified, and modified, creating a kind of emotional ecosystem. Emotions can ripple through digital

spaces, gaining momentum or dissipating depending on how they are engaged with by users. An individual post may not simply reflect the author's emotions but can resonate with a collective mood, leading to a cascade of responses that amplify or transform the initial sentiment. Deleuze and Guattari's philosophy, with its emphasis on the flow and interaction of forces, provides a framework for understanding these complex, emergent phenomena. It suggests that to truly grasp the nature of emotional expression in the digital age, NLP systems should be designed to track how emotions move and change as they pass through networks of individuals, platforms, and contexts (Deleuze and Guattari 1987, 111-148; Massumi 2002).

This perspective could inspire the development of more nuanced and adaptable NLP models that do not just recognize emotions but also understand how they interact and evolve within broader networks. For instance, instead of merely categorizing tweets as "angry" or "happy," a more advanced system could map the trajectory of emotions, showing how a tweet that starts with sarcasm might shift towards earnestness as it gets shared and commented on by others. Such a model would be better equipped to handle the complexities of modern communication, where emotional expressions are often ambiguous, layered, and subject to change depending on the context and audience.

In conclusion, Deleuze and Guattari's concept of emotions as dynamic flows offers a novel way of thinking about emotional analysis in NLP. Their view challenges the reductionist, static approach to emotions, advocating instead for models that can capture the fluidity, multiplicity, and interconnectedness of emotional expressions. By treating emotions as emergent properties of complex systems rather than fixed states, NLP systems could become more adept at understanding the subtleties of human communication. This shift would not only improve the accuracy of emotional analysis but also open up new possibilities for exploring how emotions spread and evolve within digital and social networks, ultimately leading to a more comprehensive understanding of the interplay between language, emotion, and technology (Dhabliya et.al 2024, 52-74).

1.8 The Philosophical Debate on Machine Understanding

The ability of machines to process and interpret emotional content raises profound ethical and existential questions. As NLP systems become more sophisticated in detecting and responding to emotions expressed through language, society must grapple with the implications of machines that can seemingly understand human feelings (Cambria et al. 2017, 47-70). This