

The logo for RADemics, featuring the text "RADemics" in white on a blue arrow-shaped background. The arrow points to the right and is part of a larger blue graphic element on the left side of the slide.

RADemics

Sentiment Analysis and Social Media Intelligence Using NLP

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Sentiment Analysis and Social Media Intelligence Using NLP

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Abstract

The exponential growth of social media platforms has generated vast volumes of user-generated textual data that reflect public opinions, emotions, and behavioral patterns across diverse domains. Extracting meaningful insights from this unstructured and rapidly evolving data requires advanced analytical frameworks capable of capturing contextual and affective information at scale. Sentiment analysis, as a core task in Natural Language Processing, plays a pivotal role in transforming social media content into actionable intelligence. This book chapter presents a comprehensive examination of sentiment analysis techniques and their integration within social media intelligence systems. The discussion encompasses traditional lexicon-based and machine learning approaches, deep learning architectures, and transformer-based models, highlighting their comparative strengths and limitations in handling noisy, informal, and context-dependent social media text. Key aspects of data acquisition, preprocessing, sentiment granularity, and performance evaluation are systematically analyzed. The chapter also addresses critical challenges related to multilingual content, emotion detection, bias, and ethical considerations. By synthesizing recent methodological advances and identifying open research challenges, this chapter provides a structured foundation for researchers and practitioners seeking to design robust, scalable, and context-aware sentiment analysis solutions for social media intelligence applications.

Keywords: Sentiment Analysis; Social Media Intelligence; Natural Language Processing; Deep Learning; Transformer Models; Opinion Mining.

Introduction

The unprecedented growth of social media platforms has transformed digital communication into a dominant medium for public expression, information exchange, and social interaction [1]. Millions of users generate vast quantities of textual content daily through posts, comments, reviews, and discussions, reflecting opinions, emotions, and behavioral tendencies across diverse domains [2]. This continuous flow of user-generated data represents a rich source of knowledge for understanding societal trends, consumer preferences, political discourse, and public sentiment [3]. At the same time, the unstructured, informal, and rapidly evolving nature of social media content presents substantial analytical challenges. Linguistic variability, abbreviated expressions, evolving slang, and context-dependent meanings complicate traditional text analysis methods. As a result, effective extraction of meaningful insights from social media data requires advanced computational techniques capable of handling scale, noise, and semantic complexity [4]. The growing reliance on social media data for strategic decision-making in business, governance,

healthcare, and crisis management has further intensified the demand for robust analytical frameworks that can transform raw textual data into actionable intelligence [5].

Sentiment analysis has emerged as a foundational technique for interpreting subjective information embedded within textual data. As a core task in Natural Language Processing, sentiment analysis focuses on identifying emotional polarity, opinions, and affective states expressed in language [6]. Within social media environments, sentiment analysis enables systematic assessment of public attitudes toward events, products, services, and policies [7]. This capability supports the development of social media intelligence systems that provide real-time insights into collective behavior and opinion dynamics. Unlike traditional data sources, social media content reflects spontaneous and unsolicited user expressions, offering a more authentic representation of public perception [8]. Automated sentiment analysis allows large-scale processing of such content, overcoming the limitations of manual analysis [9]. Advances in NLP techniques have significantly improved the ability to capture sentiment patterns in complex textual contexts, making sentiment analysis a critical component of modern data-driven intelligence systems [10].

Early research in sentiment analysis relied primarily on lexicon-based and statistical machine learning approaches [11]. Lexicon-driven methods utilized predefined sentiment dictionaries to assign polarity scores based on word occurrences, offering interpretability and computational efficiency [12]. Statistical learning models introduced supervised classification techniques that leveraged engineered features to improve accuracy. While these approaches established the groundwork for sentiment classification, limitations emerged in handling contextual ambiguity, sarcasm, and evolving language patterns common in social media discourse [13]. The rise of deep learning marked a significant shift by enabling automated feature learning and contextual representation through neural network architectures. Models such as convolutional and recurrent neural networks demonstrated improved performance by capturing semantic and sequential information [14]. More recently, transformer-based architectures have further advanced sentiment analysis through attention mechanisms that model long-range dependencies and contextual relationships, offering state-of-the-art performance across diverse sentiment tasks [15].