



Predicting Social Media Trends with AI: Opportunities for PR Professionals

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Abstract- The modern digital ecosystem is characterized by an unprecedented volume of information and a lightning-fast speed of its dissemination, which renders the forecasting of trends in social networks the cornerstone of effective strategic communications. Within the scope of the study, a critical-systemic evaluation of methodological tools of artificial intelligence (AI) employed for predictive analytics in social media was conducted, followed by their adaptation to the tasks of public relations specialists. The aim of the research is to develop a conceptual framework that integrates various algorithmic AI approaches—from natural language processing (NLP) and computer vision to network analysis—for constructing proactive strategies in corporate reputation management and communication campaigns. The empirical foundation of the study is formed through a systematic review of leading scientific publications dedicated to the described technologies and their applied capabilities in the context of detecting and predicting public sentiment. The scientific novelty of the work lies in the formulation of a unified methodological paradigm that shifts the emphasis of PR activities from passive response to informational challenges to active shaping of media discourse. The conclusions obtained will be useful both to researchers in the fields of communications, data processing, and computational linguistics, and to practitioners — PR directors and heads of digital agencies.

Keywords: artificial intelligence, trend forecasting, social networks, public relations, PR, natural language processing, machine learning, proactive communications, reputation management, predictive

analytics.

Introduction

The modern landscape of mass communications has undergone a radical restructuring due to the monopolization of social networks as a key channel of information exchange [1]. The deep integration of these services into everyday life has transformed them into a complex, unpredictable, and constantly changing environment in which public opinion, corporate and personal reputations are formed. For public relations (PR) specialists this entails the emergence of unprecedented opportunities for prompt interaction with the audience, while at the same time posing fundamental challenges: classical media monitoring tools and manual sentiment analysis prove useless in conditions of information flows measured in petabytes daily. The relevance of the research is determined by the growing gap between the exponential increase in the speed and volume of information processes in social networks and the limitations of modern PR technologies that fail to respond adequately to emerging situations.

In this paradigm artificial intelligence (AI) technologies are regarded as a revolutionary instrument capable of rethinking strategic communications. AI-based predictive analytics systems are not limited to a retrospective analysis of already occurred events but enable with high accuracy the forecasting of emerging information trends, opening possibilities for proactive management of communication processes. At the same time despite significant interest by researchers in applying AI for social media data analysis there is a gap in the academic community: insufficient integration and systematization of these solutions for specific PR tasks. Existing developments mainly focus on the technical aspects of algorithms or on marketing scenarios (such as sales volume forecasting and targeted advertising), whereas issues of reputation management, crisis PR and building long-term relationships with key audiences remain insufficiently explored.

The aim of the research is a comprehensive analysis and systematization of modern AI methodologies for forecasting trends in social media followed by the development of a conceptual model for their practical implementation by public relations specialists.

The scientific novelty of the work lies in the creation of a model that integrates multimodal data analysis — including textual information, visual content and structural analysis of social graphs — for proactive

management of the reputational agenda and strategic planning of communications in PR.

The authorial hypothesis assumes that the integrated application of machine learning methods for sentiment analysis of publications, topic modeling and the study of network interactions will significantly improve the accuracy of forecasts of current trends in social networks which in turn will ensure the evolution of PR practice from reactive measures to systemically substantiated proactive strategies.

Materials and methods

In recent years the pace of digital transformation and digitization of social interactions on the Internet has led to increased interest in tools for forecasting social media trends. Industry reports by Meltwater and We Are Social [1] as well as Talkwalker [11] present a comprehensive picture of global shifts: growth of user activity on platforms, increased saturation of advertising feeds and a shift of focus from textual to visual content. These documents, relying on engagement and reach metrics, predict that by 2027 more than 75 % of analytical content will be generated using generative AI technologies [10], and that the market volume of AI solutions in marketing and sales could reach \$240.6 billion by 2030 [12]. However, these reports are limited to descriptive statistics and SWOT analysis and often do not provide insight into the algorithmic mechanisms of forecasting.

Among academic studies topic modeling remains a popular method for identifying and tracking trends. Musliadi K. H., Zainuddin H., Wabula Y. [4] apply latent Dirichlet allocation (LDA) to cluster a Twitter corpus into topics and analyze the evolution of discussed conversions. Their approach enables discovery of the hidden structure of the discussion but faces limitations of the bag-of-words model and weak incorporation of temporal dynamics.

Contextualized text representations and recurrent networks complement topic-based methods, enhancing classification accuracy and semantic interpretation of messages. Alshattnawi S. et al. [2] shift from conventional vector embeddings to transformer-based representations, improving spam detection in social networks by accounting for sentence context and message metadata. Pandey R. et al. [3] integrate a hybrid attention mechanism into an LSTM model to detect irony and sarcasm, which is particularly valuable in PR analytics for monitoring public opinion. Chan J. Y.

L. et al. [5] in a review of sentiment analysis based on sequential transfer learning demonstrate that pre-trained models (BERT-derived and their analogues) outperform classical algorithms in analyzing the sentiment of short messages.

For direct forecasting of content popularity graph-based and hybrid neural network architectures are employed. Jin R., Liu X., Murata T. [6] propose a multilevel temporal graph neural network that models relationships among users, hashtags and messages across different time slices, allowing prediction of engagement peaks by considering both network topology and temporal patterns. Wang J. et al. [7] develop a multimodal hierarchical fusion model combining text, image and post metadata, which yields more accurate popularity predictions through the synergy of heterogeneous feature types.

Finally from the perspective of applied PR practice and ethics Yue C. A. et al. [9] investigate how public relations professionals use AI tools to monitor trends and assess campaign effectiveness, noting the gap between technical capabilities and the human factor—insufficient adaptation of models to PR business objectives. Giovanola B., Granata P. [8] emphasize the necessity of ethical standards in AI training and use, warning of the risks of dishonest manipulation of public opinion through creation of deep fakes and unethical targeting.

Thus the literature reveals a consensus on the high effectiveness of modern neural network and multimodal approaches for trend forecasting, whereas classical topic modeling methods are inferior in flexibility and accuracy. Nevertheless industry reports often provide only a descriptive overview without explaining algorithmic foundations. The contradiction lies in the fact that academic research demonstrates the superiority of complex models but rarely tests them on large-scale, representative data accessible in commercial analytics services. Issues of model interpretability, ethical risks in automated monitoring and the specifics of adapting methods for PR planning and crisis management tasks remain insufficiently addressed.

Results and Discussion

Based on an extensive study of the theoretical propositions and methodological foundations of artificial intelligence a conceptual model has been developed aimed at integrating predictive technologies into the everyday practice of a PR specialist. The

presented model takes the form of a step-by-step framework allowing systematic construction of processes from superficial monitoring of the information field to deep forecasting and proactive management of communication risks. Its structure includes four interconnected blocks: aggregation of multimodal data; multifaceted AI analysis; construction of predictive models and generation of strategic insights; implementation of results into PR strategy and evaluation of effectiveness.

The first block is devoted to the collection and unification of heterogeneous data types underlying any high-precision AI system. In the context of PR activities it is important to accumulate information in real time from diverse sources: official APIs of leading social platforms (X/Twitter, Instagram, Facebook, TikTok, Telegram), as well as news aggregators, blogs, forums and specialized review portals. Equally significant is the processing not only of text messages (posts, comments, articles) but also of visual content (images, videos, memes), accompanying metadata (number of likes, reposts, geotags, timestamps) and social connectivity maps (mutual follows, reposts). Although off-the-shelf solutions such as Brand Analytics or Talkwalker provide a starting platform, the full implementation of analytical scenarios at the level of in-depth research requires the development of specialized parsers and ETL processes [11].

The second stage is a comprehensive AI analysis—the central element of the proposed model responsible for extracting semantic constructs from unstructured data. This stage is subdivided into three equivalent components operating in parallel:

Semantico-tonal analysis (NLP Engine). In addition to classical sentiment analysis (determination of statement polarity: positive, negative, neutral) the module performs recognition of emotional nuances (joy, anger, fear, surprise), identifies named entities (NER) for precise identification of brands, products or personalities, and conducts topic modeling. As a result semantic clusters of messages are automatically formed and dominant and emerging trends are revealed [2, 3].

Visual content analysis (Computer Vision Module). Since images and video sequences often convey emotional and semantic load more powerfully than text this subsystem addresses several key tasks: detection and classification of logos to account for brand mentions without direct textual context identification of objects

and scenes allowing understanding of the environment and situation in which the brand is presented as well as quantitative and qualitative assessment of visual tonality for example determination of the emotional reaction evoked by an image.

Social graph analysis (Network Analysis). This component constructs and examines the network of interactions between accounts (nodes—users, edges—follows, reposts, comments) which enables the identification of key opinion leaders (influencers), measurement of community density and structure, and tracking of content dissemination trajectories. Such an approach makes it possible to accurately determine the sources of trending waves and channels for their most effective propagation [7].

The third stage is characterized by predictive modeling and generation of strategic insights combining the analytical power of AI with practical decision-making

tasks. At this stage based on aggregated and cleaned data a specially constructed feature space is formed: in addition to raw time series of mentions the model includes additional indicators—reach intensity, discussion growth rate, geographic differentiation and audience engagement metrics.

For forecasting the evolution of topics recurrent neural network architectures are applied primarily LSTM and GRU [6] capable of accounting for long-term dependencies and memory of contextual factors. The model is trained on historical sequences where target variables are the rate of change of mentions and calculated life cycle stages (initiation, exponential growth, saturation, decline). During training regularization techniques (dropout, L2 normalization) and optimization algorithms (Adam, RMSProp) are used to enhance robustness to noise and prevent overfitting (Fig.1.)

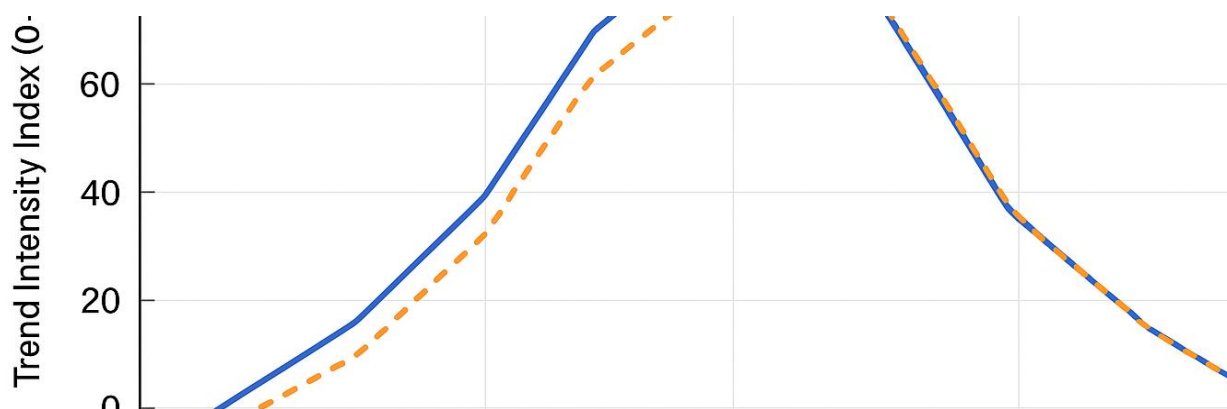


Fig. 1. Model for forecasting the life cycle of a trend (compiled by the author based on [1, 4; 10-12]).

This forecast enables a PR specialist to make a well-founded decision on whether to engage in an actively developing trend or whether it is already in decline, in which case any invested efforts would prove ineffective. At the same time the system calculates the viral potential of the generated content: it analyzes the emotional component of messages, the level of trust in the source and the parameters of network interaction.

The resulting metrics are consolidated on a specialized analytical panel (dashboard), visually structuring the key findings and recommendations. Among the insights presented on it may be early signals of impending reputational risks, opportunities to launch positive PR campaigns and a list of the opinion leaders most suitable for collaboration. The conceptual architecture of such a framework is illustrated in figure 2.

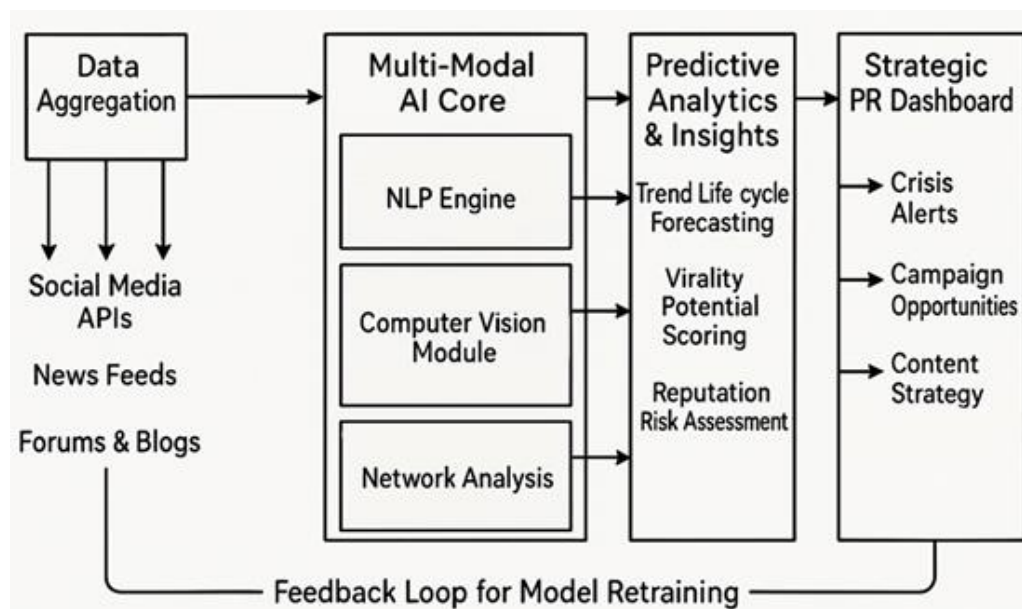


Fig. 2. Conceptual model of AI integration for predictive PR (compiled by the author based on [3, 5, 6])

At the final, fourth stage the direct integration of the obtained analytical conclusions into the structure of the overall PR strategy and the simultaneous development of a system for evaluating its effectiveness takes place. This means that the conclusions obtained through monitoring and analysis of media and social signals must become an integral part of the daily workflows of PR department specialists. Thus, upon detection of an early signal of the emergence of a negative narrative regarding the company the PR manager is able to promptly develop and disseminate counterarguments

through verified brand ambassadors thereby neutralizing a potential crisis at its inception. Moreover, when developing a new communication campaign the intelligent system suggests the most relevant thematic emphases for the target audience and the optimal presentation formats whether infographic expert interview or interactive content. Such a proactive and adaptive approach demonstrates significantly higher engagement and reach metrics compared to classical methodologies as confirmed by the data in Table 1.

Table 1. Comparative analysis of approaches to PR monitoring and analytics (compiled by the author based on [9, 12]).

Parameter	Traditional approach (manual monitoring)	AI-predictive approach
Reaction speed	Reactive (from several hours to days)	Proactive/preventive (from minutes to hours)
Data coverage	Selective, limited by physical resources	Full-scale, analysis of millions of content units
Depth of analysis	Surface (keywords, sentiment)	Deep (topics, emotions, sarcasm, visual context)
Focus	What has already happened?	What will happen next and why?
Strategic value	Reporting on fact	Agenda setting, crisis planning

Implementation of such solutions is associated with a number of significant challenges. Firstly, the costs of development and subsequent maintenance of advanced AI systems remain quite high, which often limits their widespread adoption. Secondly, there is a risk of

algorithmic bias: systems trained on historical datasets are capable of reproducing and even amplifying entrenched stereotypes and social prejudices [8]. In addition, a fundamental set of issues arises concerning the ethical aspects of technology use and ensuring the

confidentiality of user data.

Nevertheless, the transition to a predictive model in the field of PR yields tangible benefits. The expansion of the range of cloud AI services and MLaaS (Machine Learning as a Service) platforms gradually lowers the entry barrier for organizations of various scales. The presented model serves not merely as an abstract scheme but as a step-by-step guide for the evolution of the PR function. It enables a shift from empirical decision-making to system management based on data-driven analytics, as well as a move from a reactive strategy oriented toward past events to proactive construction of future communication triggers. To achieve this, public relations specialists must master new domains of knowledge, including big data processing methods and machine learning algorithms. However, the benefits offered by such a transition—enhanced forecast accuracy, increased campaign efficiency, and competitive advantage—fully compensate for the costs of implementing modern tools and training personnel.

Conclusion

The conducted study enabled the examination of the application of artificial intelligence to address one of the key tasks of contemporary PR — forecasting the dynamics and trends in social media. It was revealed that the use of individual techniques, whether sentiment analysis of messages or topic modeling, although yielding certain benefits, does not provide a comprehensive effect and does not allow a transition from reactive event monitoring to proactive communication management.

The achievement of this work was the proposal of a conceptual framework that integrates multimodal analysis (textual messages, visual content and social graph structure) with predictive analytics. The developed four-stage structure — data collection and aggregation, processing with AI tools, forecast construction and subsequent integration of results into PR strategy — serves as a practical guide for specialists seeking to transform passive monitoring into active modeling of the information agenda.

The scientific novelty lies in the fact that the proposed framework bridges the existing gap between theoretical advances in machine learning and the real needs of the public relations industry. The emphasis on the analysis of diverse data types and the capability to forecast topic development enable not only the early detection of potential reputational risks but also the identification of

white spots in the media landscape for launching proprietary viral initiatives. This fundamentally transforms the role of the PR specialist: they cease to be a chronicler of events and become an architect of public discourse.

Despite challenges such as high implementation costs, ethical considerations and the need for workforce upskilling, the demonstrated strategic advantages — improved forecast accuracy, accelerated response to shifts in public opinion and the opportunity for anticipatory influence — define predictive AI approaches as a key direction in the evolution of PR communications in the coming years.

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