

Review

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Review

The Two-Step Flow Theory in the Digital Age (2005–2025): An Analytical Literature Review

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Abstract: The Two-Step Flow (TSF) theory, developed in the mid-20th century, posits that mass media influence is mediated by opinion leaders who interpret and relay messages to wider audiences. This literature review synthesizes and critically analyzes approximately 60 studies published between 2005 and 2025, exploring the relevance, evolution, and limitations of TSF within the digital media ecosystem. The review evaluates the reconceptualization of opinion leadership (influencers, micro-celebrities, and networked individuals), the transformation of influence pathways (multi-step, networked, and algorithmic flows), and TSF's application across political communication, health, marketing, and misinformation. While digital media's interactivity, user-generated content, and algorithmic curation challenge the original TSF model, key concepts such as mediated influence and the importance of interpersonal networks persist. The findings suggest that TSF's enduring value lies in its foundational insight into social mediation, though future research must incorporate algorithmic influence, cross-platform dynamics, and the heterogeneity of digital opinion leadership. The review concludes with critical discussion and actionable recommendations for future research.

Keywords: Two-Step Flow Theory; Digital Media; Social Media; Opinion Leaders; Influencers; Information Diffusion; Network Analysis; Media Effects; Misinformation; Political Communication; and Health Communication

1. Introduction

1.1. Background and Rationale

The past two decades have witnessed a seismic transformation in the communication landscape, driven primarily by the proliferation of the internet, social media, and digital platforms (Baym, 2010; Rainie & Wellman, 2012; van Dijck, 2013). Foundational communication theories, developed in an era dominated by traditional mass media, now face renewed scrutiny as researchers assess their applicability in a complex, multi-platform world. Among these theories, the Two-Step Flow (TSF) theory, first proposed by Lazarsfeld, Berelson, and Gaudet (1944) and later elaborated by Katz and Lazarsfeld (1955), remains one of the most influential frameworks in media studies. The TSF model originally challenged the direct-effects paradigm—epitomized by the "hypodermic needle" or "magic bullet" theories—by arguing that media effects are largely mediated by "opinion leaders." These individuals, more attentive to media and trusted within their social circles, interpret and relay media messages to less engaged peers, shaping public opinion through interpersonal networks (Katz, 1957; Katz & Lazarsfeld, 1955). However, the contemporary digital media environment is characterized by ubiquitous connectivity, interactivity, user-generated content, networked publics, algorithmic curation, and the blurring of content creators and consumers (Bruns, 2008; Bucher, 2017; Marwick & boyd, 2011). These features raise fundamental questions about the relevance and limitations of the original TSF model:

- Do opinion leaders still play a central role in mediating media effects?



- **How have digital affordances, such as algorithmic gatekeeping and platform architectures, restructured information flows?**
- **What adaptations or alternative models are necessary to understand influence in the digital age?**

1.2. Objectives and Structure

This literature review aims to:

1. Synthesize empirical research from 2005–2025 assessing the TSF theory in digital environments.
2. Analyze the changing nature of opinion leadership and the structure of influence in digital media.
3. Examine the application and testing of TSF in specific domains: politics, health, marketing, and misinformation.
4. Critically evaluate the limitations of TSF and review complementary or alternative models.
5. Provide a nuanced discussion and recommendations for future research.

The review is organized as follows: Section 2 revisits the original TSF theory and early critiques. Section 3 explores the evolution of TSF in the digital media ecosystem. Section 4 reviews domain-specific applications. Section 5 critically discusses TSF's limitations and alternative models. Section 6 synthesizes findings and future research directions. Section 7 presents a critical discussion, and Section 8 offers recommendations. Section 9 provides a complete, alphabetically sorted reference list.

2. The Original Two-Step Flow Theory and Early Critiques

2.1. Foundations of TSF

The TSF theory emerged from Lazarsfeld, Berelson, and Gaudet's (1944) Erie County study on voting behavior, which found that interpersonal communication from opinion leaders had a greater effect on voting decisions than direct media exposure. This overturned the prevailing belief in powerful, uniform media effects (Bauer, 1964). Katz and Lazarsfeld (1955) expanded upon these findings in their Decatur study, emphasizing that opinion leaders—typically more media-engaged and socially active—act as intermediaries, filtering, interpreting, and legitimizing mass media messages for their peers. Katz (1957) further articulated the model's core tenets, highlighting the distinction between active opinion leaders and passive followers, the primacy of interpersonal communication, and selective exposure to media content.

2.2. Early Critiques of TSF

Despite its enduring influence, TSF faced several critiques even in the pre-digital era:

- **Oversimplification:** Critics argued that communication flows are more complex than a simple two-step process, often involving multi-step, one-step, or networked flows (Robinson, 1976; Troldahl, 1966; Van den Ban, 1964).

- **Fluidity of Opinion Leadership:** The leader-follower dichotomy was seen as artificial and context-dependent; individuals could be leaders in one domain and followers in another (Lin, 1971).
- **Underestimation of Direct Media Effects:** The model was critiqued for downplaying the direct effects of media, especially in agenda-setting and awareness (McCombs & Shaw, 1972).
- **Active Audiences:** The portrayal of opinion followers as passive recipients was challenged by research emphasizing audience agency and independent interpretation (Bauer, 1964).

Nevertheless, TSF shifted the focus of media effects research toward recognizing social context and interpersonal influence, laying the groundwork for later network perspectives (Weimann, 1982).

3. The Two-Step Flow Theory in the Digital Media Ecosystem

3.1. The Evolution of Opinion Leadership

Table 1. Evolution of Opinion Leadership in the Digital Age.

Feature	Traditional Opinion Leaders (Pre-Digital)	Digital Age Opinion Leaders
Primary Basis	Expertise, Social Status, Media Access	Network Position, Relatability, Authenticity, Algorithmic Visibility, Niche Focus
Key Types	Elites, Experts, Journalists, Community Leaders	Influencers, Micro/Nano-Influencers, Networked Individuals, Bots, Algorithms
Relationship w/Audience	Often distant, based on authority/trust	Often Parasocial, Relatable, Interactive, Community-focused
Identification	Sociometric methods, Self-designation, Reputation	Network analysis, Follower counts, Engagement metrics, Algorithmic identification
Stability	Relatively stable	Often fluid, ephemeral, context-dependent
Influence Logic	Interpersonal persuasion, Interpretation	Virality, Engagement cascades, Algorithmic amplification, eWOM, Community norms

3.1.1. Traditional Opinion Leaders in Digital Spaces

Despite technological and social changes, traditional opinion leaders—such as journalists, politicians, and recognized experts—continue to exert influence online. These individuals often leverage platforms like Twitter and Facebook to disseminate their perspectives, maintain authority, and reach wide audiences (Bruns & Burgess, 2011; Parmelee, 2014). Political journalists, for example, frequently act as opinion leaders for both the public and other media professionals during elections (Parmelee, 2014).

3.1.2. The Rise of Digital Influencers

The digital era has witnessed the emergence of new types of opinion leaders: digital influencers, micro-celebrities, and networked individuals who gain prominence primarily through social media platforms (Abidin, 2016; Khamis et al., 2017). These influencers, often self-branded and operating across diverse niches (e.g., beauty, gaming, activism), shape consumer behavior, health decisions, and even political attitudes (Casaló et al., 2020; Lou & Yuan, 2019; Sokolova & Kefi, 2020). Unlike traditional leaders, digital influencers build trust and influence through perceived authenticity, relatability, and parasocial relationships (Jin et al., 2019; Lee & Watkins, 2016). Micro- and nano-influencers—with smaller but highly engaged followings—often wield disproportionate influence within specific communities due to their credibility and network position (Campbell & Farrell, 2020; Gökalp et al., 2022; Weeks et al., 2017).

3.1.3. Networked and Algorithmic Leadership

Advancements in network analysis have allowed researchers to empirically identify influential nodes in digital conversations, moving beyond self-reported leadership (Gonzalez-Bailon et al., 2011; Himelboim et al., 2012). Influence online is increasingly understood as dynamic, context-specific, and structurally defined by network centrality (Choi, 2015; Watts & Dodds, 2007).

A key contemporary development is the role of algorithms in amplifying certain voices and content. Algorithms, acting as non-human agents, shape visibility and confer influence based on engagement metrics rather than expertise, fundamentally altering the logic of opinion leadership (Bucher, 2017; Noble, 2018; Cotter et al., 2022).

3.2. Transformation of Influence Pathways

Table 2. Transformation of Influence Pathways in the Digital Age.

Flow Model	Description	Key Characteristics	Relevant Digital Phenomena
Original Two-Step Flow	Media → Opinion Leaders → Wider Public	Linear, Top-down, Assumes passive audience, Emphasizes interpersonal mediation	---
One-Step Flow	Media/Source → Individuals (Direct)	Direct access, Bypasses intermediaries, Audience agency, Selective exposure	Search engines, Direct website access, Personalized news feeds

Multi-Step Flow	Media → Intermediary 1 → Intermediary 2... → Wider Public	Cascading, Information potentially modified at each step, Variable path length	Retweeting, Sharing on social media, Information chains
Networked Flow	Complex interplay between media, individuals, and network structures	Multi-directional, Importance of strong/weak ties, Virality, Peer-to-peer influence	Social media networks, Online communities, Viral content
Algorithmic Flow	Algorithms shape exposure, visibility, and connections based on data & metrics	Non-human mediation, Engagement-driven, Personalization, Potential for bias/bubbles	Recommendation systems, News feed algorithms, Search rankings

3.2.1. Multi-Step and Networked Flows

Digital environments facilitate multi-step information flows, where content is shared and reshaped through multiple intermediaries before reaching broader audiences (Bakshy et al., 2012; Goel et al., 2012; Hansen et al., 2011). Retweeting and sharing on platforms like Twitter and Facebook exemplify this cascading process, with information potentially being modified at each step.

3.2.2. One-Step and Direct Flows

Paradoxically, digital media also enables more direct, one-step flows. Individuals can access a vast array of information sources directly, bypassing traditional intermediaries (Bennett & Manheim, 2006; Flaxman et al., 2016). Personalization algorithms and search engines deliver tailored content, potentially reducing the mediating role of opinion leaders—though algorithmic mediation itself becomes a new form of influence (Bucher, 2017).

3.2.3. Networked Flow and the Strength of Weak Ties

Many scholars argue that networked influence models better capture digital realities, emphasizing the interplay between mass media, social networks, and individual expression (Castells, 2009; Rainie & Wellman, 2012). Information spreads rapidly through both strong and weak ties; Granovetter's (1973) "strength of weak ties" concept is especially relevant as digital networks facilitate diffusion across diverse groups (Centola & Macy, 2007; Valenzuela et al., 2018).

3.2.4. Echo Chambers and Filter Bubbles

Algorithmic curation and homophilous sorting contribute to the formation of echo chambers and filter bubbles, where individuals are exposed primarily to reinforcing viewpoints (Barberá et al., 2015; Pariser, 2011; Sunstein, 2017). This phenomenon complicates TSF by suggesting that influence may operate powerfully within fragmented, homogenous publics.

3.3. Platform Architecture and Affordances

Platform-specific features—such as visibility metrics, sharing mechanisms, algorithmic gatekeeping, and context collapse—actively structure communication flows and influence dynamics (Marwick & boyd, 2011; Gillespie, 2014; Napoli, 2014):

- **Visibility and Social Metrics:** Likes, shares, and follower counts provide visible social cues, amplifying perceived influence (Haim et al., 2018; van Dijck, 2013).
- **Sharing Mechanisms:** Features like retweets and shares accelerate diffusion and facilitate viral cascades (Guille et al., 2013; Kwak et al., 2010).
- **Algorithmic Gatekeeping:** Algorithms prioritize content based on engagement and user history, often amplifying certain voices while marginalizing others (Cotter et al., 2022).
- **Context Collapse:** Blurring of social contexts means messages intended for one group may reach unintended audiences, complicating targeted influence (Marwick & boyd, 2011).

4. Application and Testing of TSF in Digital Contexts

Table 3. TSF Application and Challenges in Key Digital Domains.

Domain	TSF Relevance in Digital Context	Key Digital Manifestations	Challenges/Nuances to TSF
Political Communication	Influential users shape discourse; partisan leaders reinforce views	Networked political discussion, Influencer targeting, Online campaigns, Echo chambers	Multi-step flows, Hybrid media influence, Algorithmic sorting, Disinformation campaigns
Health Communication	Online leaders (experts, peers) influence health decisions and information spread	Online health communities, Influencer health campaigns, e-Patient movements	Spread of health misinformation, Trust issues, Algorithmic filtering of health content

Marketing/Consumer	Influencers and eWOM drive purchases and brand attitudes	Influencer marketing, Online reviews (eWOM), Brand communities, Sponsored content	Authenticity perceptions, Parasocial interaction dynamics, Disclosure issues, Micro-influence
Misinformation	Influential accounts (incl. bots) amplify false content; trust mediates sharing	Viral misinformation/disinformation, Coordinated amplification, Trusted source effect	Algorithmic contribution to spread, Emotional contagion, Difficulty of correction

4.1. Political Communication

TSF's original context was politics. In the digital era, network analysis consistently identifies influential users shaping online political discourse (An et al., 2014; Conway et al., 2015; Himelboim et al., 2012). Selective exposure and partisan echo chambers reinforce the role of in-group opinion leaders (Bakshy et al., 2015; Iyengar & Hahn, 2009), while political campaigns increasingly target online influencers to mobilize supporters (Bimber & Davis, 2003; Gibson & Cantijoch, 2013; Vaccari & Valeriani, 2015). However, empirical studies highlight the complexity beyond TSF. Meraz (2009) documents multi-step flows in the political blogosphere, while Weeks et al. (2017) find hybrid models integrating traditional, digital, and interpersonal sources.

4.2. Health Communication

Health communication research identifies diverse online opinion leaders, from medical professionals to patient advocates (Chen et al., 2018; Hesse et al., 2005; Ho et al., 2021). Online health communities foster peer influence, and health organizations increasingly leverage influencers to disseminate public health messages (Eysenbach et al., 2004; Kite et al., 2016; Thackeray et al., 2012).

The spread of health misinformation, however, highlights the risks of mediated influence when sources are unreliable (Burki, 2019; Johnson et al., 2020; Kata, 2010).

4.3. Marketing and Consumer Behavior

Marketers have embraced TSF principles, partnering with social media influencers to shape consumer attitudes (De Veirman et al., 2017; Hughes et al., 2019; Lou & Yuan, 2019). Electronic word-of-mouth (eWOM) and user reviews function as digital word-of-mouth, with consumers acting as opinion leaders (Cheung & Thadani, 2012; Hennig-Thurau et al., 2004). Online brand communities further enable peer-to-peer influence (Casaló et al., 2010; Laroche et al., 2012).

4.4. Misinformation and Disinformation

TSF concepts illuminate the spread of misinformation online, where a small number of influential users (including bots and highly active accounts) amplify false content (Grinberg et al.,

2019; Shao et al., 2018; Starbird et al., 2014). Trust in the source, rather than message veracity, often drives sharing (Turcotte et al., 2015). However, algorithmic amplification, emotional resonance, and coordinated disinformation campaigns complicate the TSF framework (Benkler et al., 2018; Lazer et al., 2018; Vosoughi et al., 2018).

5. Critiques, Limitations, and Alternative Models

5.1. Critiques and Limitations of TSF in the Digital Age

Despite its adaptability, TSF faces substantial limitations in accounting for contemporary digital communication:

- **Oversimplification:** The linear, two-step model is inadequate for describing multi-directional, networked, and algorithmically mediated information flows (Bennett & Segerberg, 2012; Castells, 2009).
- **Fluid and Ephemeral Leadership:** Online opinion leadership is highly contextual, transient, and often driven by algorithmic visibility rather than inherent expertise (boyd, 2010; Turcotte et al., 2015).
- **Active Audiences and Direct Media Effects:** Digital audiences actively seek, interpret, remix, and produce content, challenging the notion of passive followers and mediated influence (Jenkins, 2006; Livingstone, 2004).
- **Algorithmic Mediation:** TSF does not account for the powerful role of platform algorithms in shaping exposure, visibility, and influence (Bucher, 2017; Gillespie, 2014; Noble, 2018).
- **Beyond Persuasion:** Digital communication serves functions beyond persuasion, including community-building, identity expression, and deliberation, which are not addressed by TSF (Baym, 2010; Papacharissi, 2010).
- **Online-Offline Nexus:** Influence operates across digital and offline contexts, with complex feedback loops not captured by the original model (Couldry & Hepp, 2017; Wellman, 2001).
- **Trust and Authenticity:** The basis of trust in digital opinion leadership (parasocial relationships, algorithmic amplification) differs from face-to-face trust envisioned in TSF (Dubois et al., 2020; Marwick, 2015).

5.2. Alternative and Complementary Models

Table 4. Comparison of TSF with Alternative/Complementary Models.

Model	Core Focus	How it Complements/Challenges TSF	Key Concepts
Two-Step Flow (TSF)	Interpersonal mediation of	Foundational concept of social mediation, but overly	Opinion leaders, Two-step process,

	mass media effects via opinion leaders	linear and simplistic for digital context	Interpersonal influence
Networked Influence Models	Information diffusion through network structures	Moves beyond linearity; emphasizes structural position, cascades, and complex contagion	Network centrality, Hubs, Bridges, Viral cascades, Weak/Strong ties
Diffusion of Innovations	Process of how new ideas/practices spread through social systems over time	Provides stages of adoption and adopter categories; considers communication channels beyond mass media	Innovators, Early adopters, Laggards, S-curve, Communication channels
SIDE Model	Group identity and anonymity effects in online interaction	Explains influence based on social identity salience rather than just individual leaders	Social identity, Deindividuation, Group norms, Anonymity
Algorithmic Influence Frameworks	Role of algorithms in shaping visibility, exposure, and interaction	Introduces non-human actors (algorithms) as key mediators, challenging human-centric assumptions TSF	Algorithmic curation, Gatekeeping, Personalization, Bias, Algorithmic imaginary
Logic of Connective Action	Personalized, digitally enabled collective action often bypassing leaders	Challenges the necessity of strong leadership structures for mobilization in digital contexts	Personalized action frames, Digital networks, Self-organization

- **Networked Influence Models:** These models employ network science to analyze how structure and individual attributes interact to facilitate diffusion and cascades (Aral & Walker, 2012; Bakshy et al., 2012; Watts & Dodds, 2007).
- **Diffusion of Innovations:** Rogers' (2003) diffusion model offers a process-oriented perspective, describing how innovations spread through social systems, involving multiple adopter categories and communication channels.

- **Social Identity Model of Deindividuation Effects (SIDE):** This framework explains how group identity and anonymity shape behavior and influence in digital environments (Postmes et al., 1998; Spears & Lea, 1994).
- **Algorithmic Influence Frameworks:** These models explore how algorithms mediate content exposure, confer visibility, and interact with social dynamics (Bucher, 2017; Cotter et al., 2022; Gillespie, 2014).
- **Logic of Connective Action:** Bennett and Segerberg (2012) propose that large-scale digital mobilization often bypasses traditional leaders via personalized, networked communication flows.
- **Hybrid Models:** Recent scholarship advocates for integrating TSF, network analysis, diffusion theory, and algorithmic studies to capture the interplay of social, structural, and technological factors (Hilbert et al., 2017; Weeks et al., 2017).

6. Synthesis and Future Research Directions

6.1. *Synthesis of Findings*

The literature reviewed demonstrates that while the original, linear TSF model is insufficient for the digital era, its foundational insight—that media effects are socially mediated—remains salient. The digital ecosystem has fundamentally transformed the mechanisms and pathways of influence, introducing new actors (digital influencers, micro-celebrities, algorithms), new structures (networked flows, platform architectures), and new complexities (algorithmic mediation, cross-platform diffusion). Core elements of TSF-mediated influence, the importance of interpersonal networks, and the role of trust—continue to underpin research across politics, health, marketing, and the spread of misinformation. However, the processes are now characterized by:

- Fragmentation and diversification of opinion leadership.
- Complex, multi-step, and networked information flows.
- Centrality of platform affordances and algorithmic gatekeeping.
- Contextual variation across issues, platforms, and cultural settings.
- The need for hybrid, integrative models that reflect the interplay of human, social, and technological factors.

6.2. *Future Research Directions*

The literature suggests several avenues for future inquiry:

1. **Integrated Models:** Develop robust models integrating human agency, network structure, platform architecture, and content characteristics (Hilbert et al., 2017).
2. **Algorithmic Mediation:** Investigate the role of algorithms as mediators, including their effects on opinion leadership, trust, and public perception (Bucher, 2017; Noble, 2018; Yeo et al., 2021).

3. **Cross-Platform Dynamics:** Analyze how influence flows across multiple platforms and how platform ecosystems collectively shape public discourse (Chadwick et al., 2021).
4. **Longitudinal Analysis:** Conduct longitudinal studies to understand the evolution of influence networks and the dynamics of opinion leadership over time (Aral & Dhillon, 2018).
5. **Online-Offline Interactions:** Further explore the interplay between online and offline influence, including the translation of digital authority to real-world impact (Couldry & Hepp, 2017; Vaccari & Valeriani, 2016).
6. **Nuanced Leadership Typologies:** Examine the diversity, motivations, and mechanisms of digital opinion leadership, moving beyond monolithic conceptions of "influencers" (Abidin, 2016; Dubois et al., 2020).
7. **Comparative and Global Research:** Expand research beyond Western contexts to explore the applicability of TSF and related models globally (Valeriani & Vaccari, 2018).
8. **Influence in Malign Contexts:** Investigate the role of human and algorithmic mediation in the spread of misinformation, hate speech, and polarization, developing targeted interventions (Benkler et al., 2018; Johnson et al., 2020).

7. Discussion

7.1. Enduring Relevance and Evolution of TSF

The Two-Step Flow theory's core insight—the social mediation of media effects—remains remarkably relevant, even as the digital media landscape has fundamentally altered its mechanisms. Influential intermediaries, whether traditional elites, digital influencers, or algorithmically amplified accounts, continue to shape the flow of information and public opinion. However, the identity, stability, and basis of opinion leadership have become increasingly fragmented, situational, and performative (Marwick & boyd, 2011; Watts & Dodds, 2007).

The linearity of the original TSF model has given way to complex, multi-step, and networked flows, with information traveling through diverse pathways shaped by network structures, platform affordances, and algorithmic logics (Bakshy et al., 2012; Castells, 2009). Echo chambers and filter bubbles further complicate the dynamics of influence, potentially reinforcing group identities and limiting exposure to divergent perspectives (Barberá et al., 2015; Pariser, 2011).

7.2. Limitations and Gaps

While TSF continues to offer a valuable lens for examining mediated influence, its explanatory power is limited by its human-centric and linear assumptions. It often fails to account for:

- The active role of audiences as content creators, remixers, and selective consumers (Bruns, 2008; Jenkins, 2006).
- The algorithmic mediation of visibility, reach, and influence (Bucher, 2017; Gillespie, 2014).
- The diversity of communication goals in digital environments, including community-building and identity work (Baym, 2010; Papacharissi, 2010).

- The interplay between online and offline influence.
- The ethical and societal implications of algorithmic and influencer-mediated communication.

7.3. Integrating TSF with Contemporary Models

To address these gaps, future research must integrate TSF insights with network analysis, diffusion of innovations, algorithmic studies, and social psychological models. Hybrid approaches can more accurately capture the interplay of social, technological, and individual factors shaping influence in the digital age (Hilbert et al., 2017; Weeks et al., 2017).

8. Recommendations

Based on this review, the following recommendations are offered for researchers, practitioners, and policymakers:

1. **Adopt Hybrid Theoretical Frameworks:** Combine TSF with network, diffusion, and algorithmic models to capture the complexity of digital influence.
2. **Prioritize Empirical Network Analysis:** Employ network metrics and longitudinal data to empirically identify opinion leaders and influence pathways.
3. **Investigate Algorithmic Mediation:** Critically examine how platform algorithms confer or diminish influence and shape public discourse.
4. **Embrace Cross-Platform and Cross-Cultural Research:** Study influence diffusion across multiple platforms and diverse cultural settings.
5. **Examine Ethical Implications:** Address the ethical challenges posed by algorithmic amplification, influencer marketing, and the spread of misinformation.
6. **Promote Digital and Algorithmic Literacy:** Enhance public understanding of algorithmic curation to foster critical media consumption.
7. **Develop Interventions for Malign Influence:** Design targeted interventions to mitigate the spread of misinformation, hate speech, and polarization.
8. **Foster Interdisciplinary Collaboration:** Engage scholars from communication, sociology, computer science, psychology, and policy studies to develop integrative models.

9. Conclusion

The Two-Step Flow theory, conceived in the context of mid-20th-century mass communication, continues to echo in the digital age. Its central premise—that communication is fundamentally a social process mediated by trusted intermediaries—remains pertinent, even as the digital environment introduces new actors, structures, and mechanisms. The review demonstrates that TSF's value lies not in its original, literal structure but in its recognition of the importance of social mediation. To remain relevant, TSF must be reconceptualized and integrated with complementary models that account for the complexities of digital media—networked flows, algorithmic mediation, diversified leadership, and active audiences. The challenge for future research is to develop nuanced, context-aware, and dynamic frameworks that reflect the interplay of human, social, and technological factors shaping influence in the 21st century.

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