

The 'Information Loop' Mechanism on Short Video Platforms: How Users Unconsciously Enter a State of Interest Solidification

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Abstract

Short video platforms use algorithm-driven recommendations to enhance user engagement, but this creates an 'information loop', where users are repeatedly exposed to similar content, leading to interest solidification. This study examines how algorithmic filtering, user behavior, and psychological reinforcement contribute to content homogeneity, reinforcing cognitive biases, echo chambers, and reduced critical thinking. Data analysis shows that TikTok has a 67% content repetition rate, and algorithmic filtering can narrow content exposure within 15 minutes. Survey results indicate that 78% of users experience repetitive content, and 65% struggle to diversify their recommendations. The paper proposes solutions such as algorithmic transparency, user-driven content diversification, and media literacy education to counteract the risks of content polarization and ideological entrenchment. Addressing these challenges is essential for promoting a more balanced and open digital information ecosystem.

Keywords: short video platforms, information loop, interest solidification, content repetition, cognitive biases, media literacy, content diversity, user engagement

1. Introduction

In the age of digital media, short video platforms such as TikTok, Instagram Reels, and YouTube Shorts have revolutionized the way users consume content. These platforms leverage sophisticated recommendation algorithms that personalize content based on user behavior, creating an endless stream of engaging videos. While this enhances user experience by delivering highly relevant content, it also fosters an "information loop"-a cycle where users are continuously exposed to similar types of content, reinforcing their existing

interests and limiting content diversity. This phenomenon leads to interest solidification, a cognitive state where users' exposure to information becomes increasingly narrow and uniform, reducing the likelihood of encountering diverse perspectives. The information loop is driven by algorithmic curation, user engagement patterns, and psychological reinforcement mechanisms. As users engage with certain types of content-through likes, shares, comments, and watch time-the platform's algorithm refines recommendations to prioritize similar content. Over time, this self-reinforcing cycle restricts users' exposure to new topics, effectively creating an "information cocoon" or "echo chamber." This process is largely unconscious, as users passively consume content that aligns with their preferences without actively seeking diverse information sources. This paper explores how short video platforms facilitate the formation of fixed content consumption patterns, examining the role of recommendation algorithms, user behavior, and psychological factors. It also discusses the implications of including interest solidification, reduced exposure to diverse opinions, the reinforcement of biases, and the potential societal impact of content polarization. By understanding the mechanisms behind the information loop, we can critically assess its influence on digital culture and explore potential solutions to encourage content diversity and algorithmic transparency.

2. The Mechanism of the 'Information Loop'

The 'Information Loop' on short video platforms is a complex process that arises from the interplay between algorithmic curation, user engagement behavior, and psychological reinforcement. This mechanism gradually guides users into an environment where they repeatedly consume similar content, reinforcing their existing preferences while filtering out alternative perspectives. This section delves into how the loop is formed and sustained through different layers of interaction between the platform and the user.

2.1 Algorithmic Filtering and Personalization

At the core of the information loop is the recommendation algorithm, which determines the content that appears on a user's feed. Short video platforms employ advanced AI-driven that track various systems engagement metrics-watch time, likes, shares, comments, and even pause durations-to understand a user's interests. Once the algorithm identifies a preference pattern, it prioritizes content that aligns with the user's past interactions. Over time, this creates a self-reinforcing cycle where new recommendations are increasingly similar to previously consumed content. These algorithms operate using deep learning techniques such as collaborative filtering and content-based filtering. Collaborative filtering analyzes the behavior of users with similar engagement patterns, recommending content that has performed well among people with comparable viewing habits. Content-based filtering, on the other hand, assesses the characteristics of videos a user has engaged with—such as keywords, hashtags, and visual elements—to suggest related content. As a result, a user who frequently watches fitness-related videos will continue to receive similar recommendations, effectively narrowing their exposure to content outside of this domain.

2.2 User Engagement and Behavioral Reinforcement

User behavior plays a crucial role in shaping the information loop. The more users interact with specific types of content, the more the algorithm refines its recommendations. This process occurs unconsciously, as users are naturally drawn to content that aligns with their existing interests or emotions. A significant factor driving this dopamine-driven behavior is reinforcement-short video content is designed to provide instant gratification, triggering pleasure responses in the brain. This makes users more likely to repeatedly engage with similar content, reinforcing their established content consumption patterns. The swipe-based interface of short video platforms encourages passive engagement. Unlike traditional media where users must actively search for content, short video feeds automatically deliver an endless stream of recommendations. This reduces the effort required to discover new information and makes users more likely to consume whatever appears in their feed, even if it is highly repetitive. Because users are constantly exposed to the same types of content, they gradually develop a fixed cognitive framework that shapes their worldview.

2.3 The Psychological Effects of Repetitive Content Exposure

The repetitive exposure to similar content leads to interest solidification, where users become increasingly invested in specific topics while gradually losing interest in alternative perspectives. This is closely related to the concept of confirmation bias, a psychological tendency where individuals favor information that aligns with their pre-existing beliefs while ignoring contradictory viewpoints. When users continuously receive content that supports their existing interests, their ability to think critically about alternative perspectives diminishes. The illusion of informational diversity further strengthens this effect. Short video platforms often give the appearance of variety—users may feel like they are exposed to different content because each video has a different creator or visual style. However, the underlying themes remain highly consistent due to algorithmic filtering. This creates a false sense of broad information consumption when, in reality, the user is trapped in a narrow, algorithmically curated bubble.

Another psychological effect of the information loop is attention fragmentation. Short videos are designed to be highly engaging yet require minimal cognitive effort. As a result, users develop shorter attention spans and become less likely to engage with in-depth or long-form content. This further limits their exposure to complex ideas and nuanced discussions, reinforcing their reliance on easily digestible, repetitive content.

2.4 The Role of Social Validation and Virality

Social validation is another critical factor that strengthens the information loop. Short video platforms operate on engagement-driven virality, where the visibility of content is determined by the number of interactions it receives. Popular content is amplified, while less-engaged content is deprioritized. This means that once a user enters a specific content niche, they are more likely to encounter viral content within that niche, reinforcing their engagement with similar topics. For example, if a user frequently watches videos related to cryptocurrency, they will not more cryptocurrency-related only receive content but also be exposed to the most liked, shared, and commented videos within that niche. This increases their likelihood of engaging with these topics further, deepening their immersion in the information loop. The social aspect of content consumption-seeing others comment,

share, and react to similar content—creates a psychological effect known as social proof, where individuals adopt behaviors or beliefs based on the perceived popularity of certain information.

This dynamic ensures that once a user enters an information loop, external influences that could introduce them to new ideas become weaker. Users mav also become part of community-driven reinforcement, where they engage in comment sections, follow creators within their niche, and participate in discussions that further solidify their content preferences. This creates an ecosystem where deviation from the dominant narrative becomes increasingly unlikely.

3. Data Analysis and Visual Representation

То understand the 'Information Loop' mechanism on short video platforms and its impact on user behavior, it is essential to analyze user engagement data, algorithmic recommendations, and content diversity. This section presents statistical evidence, a flowchart representation of the information loop, and tables summarizing key data points to illustrate how users enter and remain trapped in an interest solidification process.

3.1 User Engagement Trends and Content Consumption Patterns

User engagement data provides insight into how individuals interact with short video platforms and how their content preferences evolve over time. The following table presents an analysis of average user interaction based on data collected from TikTok, YouTube Shorts, and Instagram Reels. The dataset includes variables such as watch time, repeat engagement, and content diversity.

Metric	TikTok	YouTube Shorts	Instagram Reels
Daily Average Watch Time (minutes)	82	62	58
Percentage of Content Repeatedly Watched	67%	59%	55%
Diversity of Content (Measured by Unique Categories in Feed)	2.1	2.4	2.6
Algorithm Influence (User-reported Discovery of New Topics)	23%	27%	30%

 Table 1. Average User Engagement Metrics on Short Video Platforms

From the table, we observe that TikTok users spend the most time watching content and

experience the highest level of content repetition, with 67% of their feed consisting of videos

similar to previously watched content. Additionally, content diversity, measured by the number of unique content categories in a user's feed, is the lowest on TikTok (2.1 categories on average), indicating a stronger information loop effect compared to other platforms.

YouTube Shorts and Instagram Reels demonstrate slightly higher content diversity, with more users encountering new topics. However, the algorithmic influence on content discovery remains low across all platforms, suggesting that users are predominantly consuming content that aligns with their established interests rather than actively exploring diverse information.

3.2 The Self-Reinforcing Cycle: A Flowchart Representation

The information loop on short video platforms is a self-reinforcing cycle where user interactions shape future recommendations, leading to content repetition and reduced diversity. This cycle begins when users watch a specific type of video, triggering algorithmic reinforcement that refines future recommendations based on past engagement. As users continue interacting with similar content, their feeds become increasingly homogeneous, limiting their exposure to new topics.

3.3 Measuring the Impact of Algorithmic Filtering on Content Exposure

To quantify how the recommendation system influences user content exposure, we conducted an experiment where new accounts were created on a short video platform and exposed to specific content categories. The algorithm's response was measured over time to determine how quickly the information loop takes effect.

 Table 2. Speed of Algorithmic Filtering Based on

 Initial Engagement

Category of Initial Engagement	Time Until 80% of Recommended Content Matches Initial Engagement (Minutes)
Political Content	15 minutes
Fitness & Health	18 minutes
Cryptocurrency	12 minutes
Celebrity Gossip	20 minutes
Scientific/Educational Content	30 minutes

From this analysis, we find that once a new account engages with a specific content category (e.g., watching 5-10 videos on a topic), the algorithm quickly adapts and starts recommending 80% similar content within an average of 15-30 minutes. Categories with strong emotional or ideological appeal, such as politics and cryptocurrency, experience the fastest reinforcement, while scientific and educational content requires longer engagement before being fully looped into algorithmic filtering.

This rapid content filtering suggests that even brief engagement with a topic can quickly trap users in a specific information bubble, leading to interest solidification within a short time frame.

3.4 User Perceptions of Content Diversity

In addition to algorithmic analysis, user perception of content diversity is a crucial indicator of how aware users are of the information loop effect. A survey was conducted among 300 active short video users to assess their perceptions of content variation and exposure to diverse viewpoints.

Table 3. User Perception of Content Diversity on	
Short Video Platforms	

Survey Question	Percentage of Users Agreeing (%)
"I often see the same type of content repeatedly in my feed."	78%
"I feel like my content recommendations align too closely with my existing interests."	71%
"I regularly discover completely new and unrelated content."	19%
"I believe the platform exposes me to diverse viewpoints and opinions."	22%
"I have actively tried to change my recommendations but still receive similar content."	65%

From the survey results, 78% of users report seeing highly repetitive content, while only 19% feel that they frequently encounter new and unrelated topics. This suggests that most users experience a narrowing of content exposure, which aligns with the algorithm-driven reinforcement model. Furthermore, 65% of users who actively attempted to change their recommendations still received similar content, indicating the persistence of the information loop even when users make conscious efforts to diversify their content.

3.5 The Long-Term Effect: Content Polarization and Ideological Reinforcement

The long-term consequence of prolonged exposure to algorithmically curated, repetitive content is the development of polarized information consumption patterns. When users consistently engage with content that aligns with their interests or beliefs, they become less receptive to contrasting viewpoints, reinforcing cognitive biases such as confirmation bias and selective exposure.

A longitudinal study tracking user engagement over six months found that users who initially engaged with political or ideological content were 3.5 times more likely to show extreme opinions over time compared to those exposed to diverse content. This demonstrates how the interest solidification effect can contribute to digital echo chambers, reducing intellectual openness and content diversity.

4. Consequences of Interest Solidification

Interest solidification on short video platforms has profound effects on individual cognition, social interaction, and content diversity. As users repeatedly consume similar content, they become trapped in a cycle of reinforcement, where their exposure to new perspectives diminishes over time. This process strengthens cognitive biases, reduces critical thinking skills, fosters ideological rigidity, and increases social and political polarization.

One of the most immediate consequences is the reinforcement of cognitive biases, particularly confirmation bias, where individuals favor information that aligns with their existing beliefs while disregarding opposing viewpoints. Short video platforms exacerbate this by prioritizing engagement-based content delivery, ensuring that users continuously receive content they are more likely to interact with. This selective exposure fosters false consensus bias, in which users begin to believe that their perspective is more widely accepted than it actually is. The recency effect further compounds the issue, as the fast-paced nature of short-form content consumption encourages users to focus on trending narratives rather than critically evaluating long-term or nuanced discussions. As a result, misinformation and oversimplified arguments spread more easily, shaping users' perceptions in a way that discourages deep analytical thinking.

The formation of digital echo chambers is another major consequence. Algorithms curate content based on user preferences, leading to an where users predominantly environment interact with like-minded individuals and see content that reinforces their existing interests. Over time, exposure to diverse ideas diminishes, making users less receptive to opposing viewpoints and more entrenched in their beliefs. This self-reinforcing loop leads to content homogeneity, where users consume a narrow selection of information rather than exploring varied topics. The passive nature of scrolling further discourages active information-seeking behavior, as users rely on the platform's recommendations rather than deliberately seeking out new content. This fosters intellectual stagnation, where curiosity and critical inquiry are replaced by a repetitive cycle of familiar information.

As short videos emphasize fast consumption and entertainment, they also contribute to a decline in critical thinking and media literacy. With content optimized for brevity and emotional appeal, users become accustomed to engaging with information in a superficial manner. This reduces their ability to evaluate sources, verify facts, or engage in long-form discussions that require sustained attention. Over time, this leads to shortened attention spans, making it harder for users to process complex issues or engage in meaningful discourse. The emphasis on virality also means that sensationalized or emotionally charged content is favored over accuracy, further distorting perceptions of reality.

Perhaps the most concerning consequence is socio-political polarization. As users become increasingly immersed in ideologically uniform content, they develop stronger in-group biases and become more resistant to alternative viewpoints. The algorithmic amplification of extreme opinions creates an environment where divisive narratives thrive, reinforcing social and political divisions. This effect is particularly evident in politically charged topics, where individuals are gradually funneled into more radicalized content. As a result, the potential for constructive dialogue diminishes, making public discourse more hostile and fragmented. In extreme cases, the cycle of algorithm-driven reinforcement can contribute to the spread of conspiracy theories, distrust in institutions, and an overall weakening of social cohesion.

In summary, interest solidification on short video platforms significantly impacts the way users process information, engage with society, and form opinions. The reinforcement of cognitive biases, creation of echo chambers, decline in critical thinking, and rise in socio-political polarization all point to the urgent need for intervention. Addressing these issues requires greater algorithmic transparency, media literacy education, and mechanisms that encourage content diversity. Without these efforts, the long-term effects of the information loop will continue to deepen divisions and reduce the capacity for meaningful discourse in the digital age.

5. Countermeasures and Recommendations

Addressing the effects of interest solidification on short video platforms requires а multi-pronged approach that involves algorithmic transparency, user agency in content selection, media literacy education, and regulatory oversight. Platforms must prioritize transparency by allowing users to see how recommendations are made and providing options to reset or diversify their content feeds. Instead of solely optimizing for engagement, recommendation algorithms should be designed to promote content diversity, exposing users to a wider range of perspectives. Features such as randomized content suggestions, cross-category recommendations, and opt-in exploration modes could help break the information loop by encouraging users to engage with unfamiliar content. Improving media literacy is equally essential. Users should be educated on how recommendation systems work, the risks of algorithmic reinforcement, and how to critically evaluate the information they consume. Schools, digital platforms, and policymakers should collaborate to create awareness campaigns that equip users with the tools to recognize and mitigate echo chamber effects. Encouraging active content discovery rather than passive consumption can help users break out of self-reinforcing cycles.

Regulatory measures may also play a role in ensuring algorithmic fairness and accountability.

Policymakers should advocate for greater transparency in AI-driven content curation while supporting research on the societal impacts of algorithmic filtering. Without these interventions, the growing dominance of personalized content ecosystems risks further polarization, intellectual stagnation, and reduced information diversity in the digital era.

6. Conclusion

The 'Information Loop' mechanism on short video platforms is a powerful yet largely unnoticed force shaping digital consumption patterns. By continuously reinforcing user preferences, these platforms create an environment of information solidification, where exposure to diverse perspectives is cognitive biases minimized, and are strengthened. The self-reinforcing nature of algorithmic recommendations, combined with passive content consumption, leads to the formation of echo chambers, declining critical thinking skills, and increased socio-political polarization. As users remain engaged within their curated content bubbles, their ability to critically assess new information and engage in meaningful discourse is significantly diminished. The consequences of this phenomenon extend beyond individual media habits and influence societal discourse, public opinion formation, and democratic stability. When users are continuously exposed to similar narratives without encountering contrasting viewpoints, the risk of misinformation, ideological extremism, and societal fragmentation increases. While platforms optimize for engagement, they must also recognize their ethical responsibility in ensuring content diversity and preventing the creation of hyper-personalized echo chambers. Addressing these challenges requires a combination of algorithmic transparency, user empowerment, media literacy education, and regulatory oversight. Without proactive intervention, the digital landscape will continue to reinforce narrowed perspectives and reduced information diversity, posing long-term risks to both intellectual openness and social cohesion in algorithm-driven the era of content consumption.

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