

Article

Digital Narratives in the Age of Intelligent Systems

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Abstract: Digital narratives no longer emerge solely from human intention; they are increasingly co-constructed within ecosystems shaped by intelligent systems. This paper interrogates how algorithmic processes, particularly those driven by artificial intelligence, reconfigure narrative authority, temporality, and meaning-making. Rather than treating technology as a neutral conduit, the study positions intelligent systems as active narrative agents that influence both the production and circulation of stories. The inquiry begins by re-examining foundational assumptions in narrative theory, where authorship was historically singular and linear, and contrasts this with contemporary, data-driven storytelling environments. Through a conceptual-analytical approach, the paper traces the shift from static digital storytelling to adaptive, generative, and interactive narrative forms. It argues that narratives produced within intelligent systems operate through probabilistic logic, introducing fluidity and indeterminacy that challenge traditional notions of coherence and authenticity. A proposed conceptual diagram illustrates the triadic relationship between human creators, algorithmic systems, and audiences, emphasizing feedback loops that continuously reshape narrative outputs. This transformation carries significant implications for cultural production, epistemology, and ethical accountability. The paper concludes by calling for a re-theorization of narrative frameworks within digital humanities to accommodate the agency of intelligent systems.

Keywords: Digital Narratives; Intelligent Systems; Algorithmic Storytelling; Human–AI Interaction; Digital Humanities

1. Introduction

Narratives have always been bound to the conditions of their production. Oral traditions relied on memory and performance; print stabilized meaning through inscription. Digital environments unsettled that stability, yet still retained a recognizable human center. Contemporary intelligent systems introduce transformations that are both quantitative and structural. Intelligent systems (trained on vast corpora, operating through probabilistic inference) do not simply assist storytelling; they intervene in its structure, pacing, and even its epistemic grounding. The question is no longer whether technology mediates narrative, but how agency is redistributed within that mediation [1].

Digital narratives, in this evolving context, resist closure. They are iterative, adaptive, and sometimes opaque. A generated story does not begin from intention in the classical sense; it is assembled through patterns, weights, and latent associations that exceed direct human oversight. This raises a tension. If narrative has historically been a vehicle for meaning, what happens when meaning is partially delegated to systems that do not “understand” in human terms? Scholars have begun to frame this shift as a movement toward algorithmic culture, where cultural production is shaped by computational logics embedded within platforms and infrastructures [2]. Yet, this framing remains incomplete without a closer interrogation of narrative itself.

The present study situates digital narratives within a triadic system: human creators, intelligent systems, and

audiences. These are not discrete actors but dynamically entangled nodes. A conceptual diagram (**Figure 1**) is introduced to illustrate this relationship, emphasizing recursive feedback loops. Human input trains systems; systems generate outputs; audiences interact, react, and produce data; this data feeds back into the system, altering subsequent narrative possibilities. The narrative, therefore, is not a fixed artifact but a moving configuration shaped by continuous interaction.

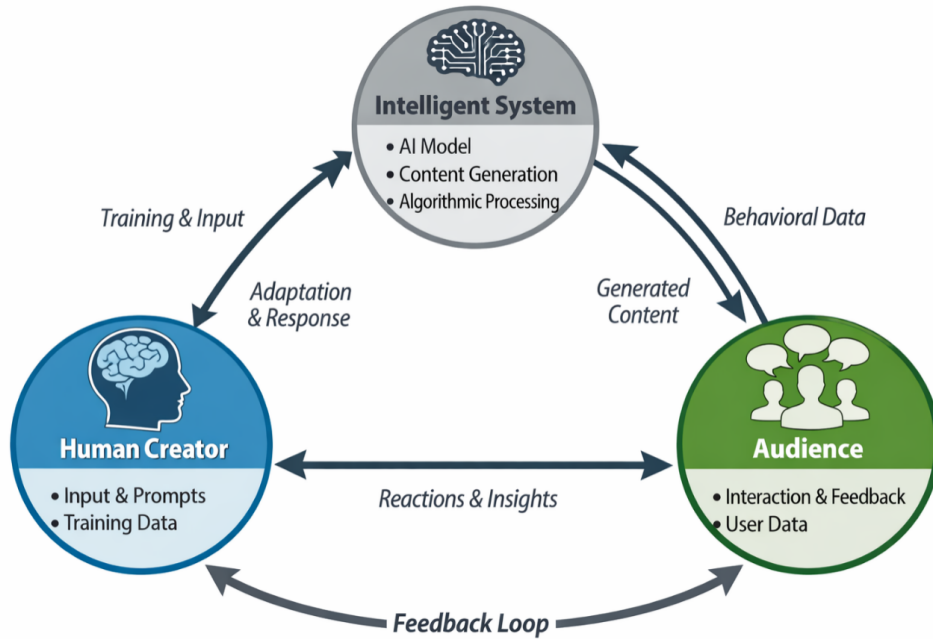


Figure 1. Conceptual model of Human–AI–Audience narrative feedback loop.

Note: This figure operationalizes the study’s theoretical claim that narrative agency emerges through recursive interaction rather than linear authorship.

Within this configuration, authorship becomes diffuse. The notion of a singular narrative voice fragments into layers of prompt, model, dataset, and interface. Each layer exerts influence, yet none can claim full ownership. This fragmentation complicates long-standing assumptions in narrative theory, particularly those concerning intentionality and coherence. It also introduces new forms of indeterminacy. A story generated today may not be reproducible tomorrow under identical prompts, given the stochastic nature of many intelligent systems [3,4].

At the same time, audiences are no longer passive recipients. Interaction is built into the architecture of digital platforms. Choices, clicks, and pauses become narrative inputs within platform architectures. In some cases, the audience effectively co-authors the story, though this co-authorship is mediated and constrained by algorithmic design. The boundaries between creator and consumer blur, yet they do not dissolve entirely; instead, they are reconfigured along lines of access, control, and technical literacy.

Rather than assuming that digital narratives are inherently processual, this study investigates how and under what specific computational conditions narrative processuality emerges, and whether this transformation is structurally produced by algorithmic systems rather than conceptually presumed. It asks: how do intelligent systems reshape the conditions under which narratives are produced and interpreted? What forms of agency emerge, and which are diminished? And how might digital humanities frameworks adapt to account for these shifts?

For clarity, key terms are defined as follows: narrative agency refers to the capacity to influence narrative structure; algorithmic authorship denotes distributed authorship across computational systems and users; opacity describes the limited visibility into generative processes; and co-authorship refers to iterative human–machine narrative construction.

The significance of these questions extends beyond theory. Narratives structure knowledge, identity, and social reality. When their production is mediated by systems optimized for prediction and engagement, the implications are cultural as much as technical. To engage with digital narratives today is, in part, to engage with the archi-

tructures of intelligence that now underwrite them [5,6]. This study contributes by (i) providing a process-oriented analytical framework for AI-generated narratives, (ii) empirically illustrating how narrative structures emerge under probabilistic generation, and (iii) bridging narrative theory with algorithmic systems through a multi-layered interpretive method.

2. Literature Review

The study of narrative has never been static, yet much of its classical grounding remains anchored in assumptions that no longer hold under computational conditions. Traditional narrative theory (rooted in linearity, authorial intent, and textual closure) presumes a stable relationship between creator, text, and audience [7]. Digital storytelling disrupts this arrangement, but not entirely at once. Early digital narratives, particularly hypertext and transmedia forms, introduced fragmentation and multiplicity while still retaining human-centered design logics [8]. Transmedia storytelling, for instance, distributed narrative elements across platforms, yet coherence was often orchestrated by identifiable creators or institutions. What appears now, under intelligent systems, is a further displacement. Not fragmentation alone, but generativity (ongoing, probabilistic, and often non-reproducible).

A useful distinction emerges here. Digital narratives of the early 2000s expanded narrative space; AI-driven narratives alter narrative logic. The difference is subtle, though decisive. Where hypertext invited navigation, intelligent systems produce variation. This shift complicates narrative temporality. Stories are no longer traversed; they are assembled in real time. Scholars such as Ryan (2015) [9] anticipated aspects of this transformation, yet the scale and autonomy of contemporary systems exceed earlier frameworks [9]. The literature acknowledges interactivity, but tends to understate the epistemic opacity of machine-generated content.

The integration of artificial intelligence into creative industries further intensifies this shift. AI systems now generate text, visual art, and music which is often indistinguishable from human-produced artifacts. This has prompted a growing body of work examining creativity as a distributed process rather than an individual capacity [10]. Still, tensions persist. Is AI creative, or does it merely simulate creativity through recombination? The literature remains divided. Some argue that creativity requires intentionality, thereby excluding machines [11]. Others suggest that creativity can be redefined in functional terms, focusing on novelty and value regardless of origin [12]. What is less frequently addressed is how these debates intersect with narrative form itself. AI-generated stories do not simply replicate human storytelling; they reshape narrative conventions through statistical patterning.

A conceptual gap becomes visible when considering algorithmic authorship. Traditional authorship implies accountability, ownership, and origin. In algorithmic systems, these dimensions fragment. Training data, model architecture, and user prompts all contribute to the final output. The “author” becomes a composite entity, difficult to locate and harder to regulate. A schematic representation (**Figure 2**) clarifies this diffusion: authorship is distributed across datasets, developers, interfaces, and end-users, each exerting partial influence.

Co-creation, often celebrated in digital media studies, takes on a different character in this context. It is not merely collaborative but asymmetrical. Human users interact with systems whose internal processes remain largely inaccessible. This asymmetry introduces a form of constrained agency. Users can guide, prompt, and refine within boundaries set by algorithmic design. The literature on human–AI interaction acknowledges this dynamic, yet often frames it in terms of usability rather than narrative consequence [13]. The implications for storytelling (particularly for voice, perspective, and narrative authority) remain underexplored.

Ethical considerations surface repeatedly across these discussions, though not always with sufficient depth. Bias in AI-generated narratives reflects biases embedded in training data, raising concerns about representation and marginalization [14,15]. Misinformation becomes more difficult to trace when narratives can be generated at scale, tailored to specific audiences, and iteratively refined. Authorship, again, complicates accountability. If a harmful narrative emerges from an AI system, responsibility is diffuse. The literature identifies these issues but often treats them as externalities rather than integral to narrative formation. This separation limits analytical clarity.

Another ethical dimension concerns ownership. Intellectual property frameworks struggle to accommodate works generated through machine learning processes. Who owns a narrative produced by an AI model trained on publicly available texts? The developer, the user, the original authors of the training data? Current scholarship offers partial answers, often grounded in legal interpretation rather than narrative theory [16]. A more integrated approach is needed which situates ownership within the broader ecology of digital narrative production.

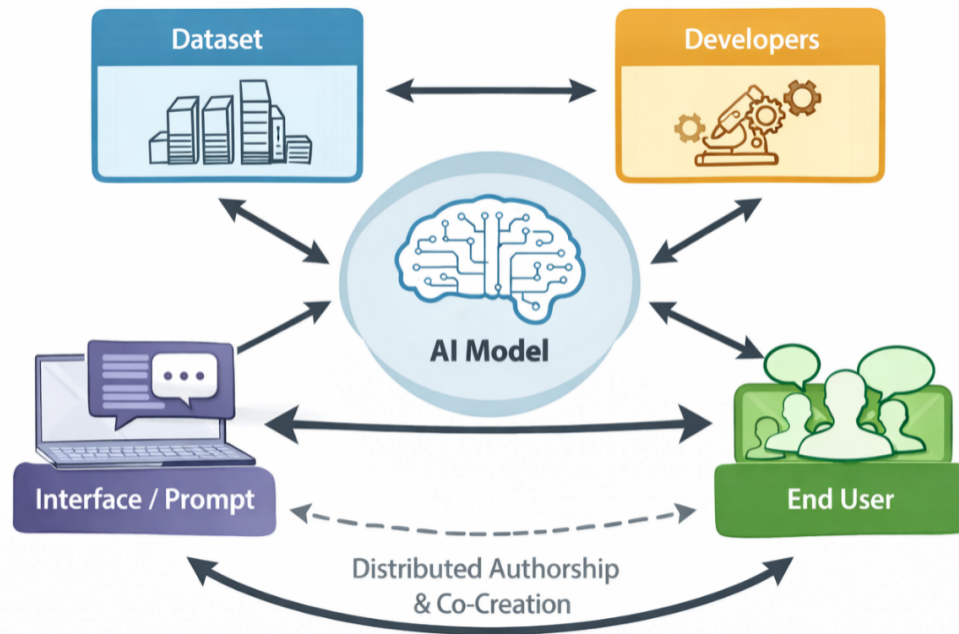


Figure 2. Distributed model of algorithmic authorship and co-creation.

Note: This figure illustrates the relational dynamics between human users, AI systems, training data infrastructures, and platform algorithms in the production of digital narratives.

Despite the richness of existing scholarship, several gaps remain. First, there is limited theoretical integration between narrative studies and AI research. Discussions of algorithmic systems often occur in isolation from narrative theory, resulting in parallel literatures that rarely intersect. Second, the concept of narrative agency under intelligent systems remains insufficiently defined. While terms such as “co-creation” and “algorithmic authorship” are widely used, their implications for narrative structure and meaning are not fully articulated. Third, empirical studies tend to focus on outputs (generated texts, artworks) without examining the underlying processes that shape these outputs.

Recent work has also begun to explore the intersection of artificial intelligence, pedagogy, and multiliteracies, emphasizing how AI reshapes meaning-making practices in digital environments [17]. While not explicitly focused on narrative theory, such studies highlight the broader epistemic transformations introduced by AI-mediated communication.

This paper contributes to addressing these gaps by reframing digital narratives as emergent processes within human–AI–audience systems. It moves beyond descriptive accounts of AI-generated storytelling to interrogate the structural and epistemic transformations at play. By integrating insights from narrative theory, digital humanities, and AI studies, the analysis seeks to articulate a more cohesive framework for understanding storytelling in the age of intelligent systems. In doing so, it positions narrative not as a static artifact but as a dynamic, recursive formation shaped by distributed agency and computational logic.

3. Theoretical Framework: Algorithmic Culture

The interpretive grounding of this study rests on the concept of algorithmic culture. Not merely a descriptor of technological mediation, but a condition in which cultural production, circulation, and interpretation are increasingly shaped by computational procedures. In this view, algorithms do not sit behind culture; they operate within it, structuring visibility, relevance, and narrative form. Culture, then, becomes partially executable. Not fixed, not entirely authored and processed.

To approach digital narratives through this lens requires a shift in emphasis. Narrative is no longer treated as a bounded text or even as a human-centered communicative act. It is instead understood as an outcome of interactions between data, models, and users, each governed by algorithmic logics. These logics are not neutral. They

prioritize certain patterns, suppress others, and in doing so, introduce a layer of pre-structuring before any “story” appears. What emerges as narrative is already filtered, ranked, and probabilistically assembled [18]. Interpretation must therefore account for this prior conditioning.

Algorithmic culture reframes authorship in a way that is difficult to reconcile with traditional narrative theory. The author, once a coherent origin point, disperses across a system of inputs. Training datasets carry historical and cultural residues; model architectures encode design decisions; user prompts activate specific trajectories. None of these elements alone constitutes authorship, yet each contributes materially. A conceptual diagram (Figure 3) is useful here to visualize this distributed configuration: narrative output sits at the intersection of data, algorithm, and user interaction, with feedback loops continuously recalibrating the system.

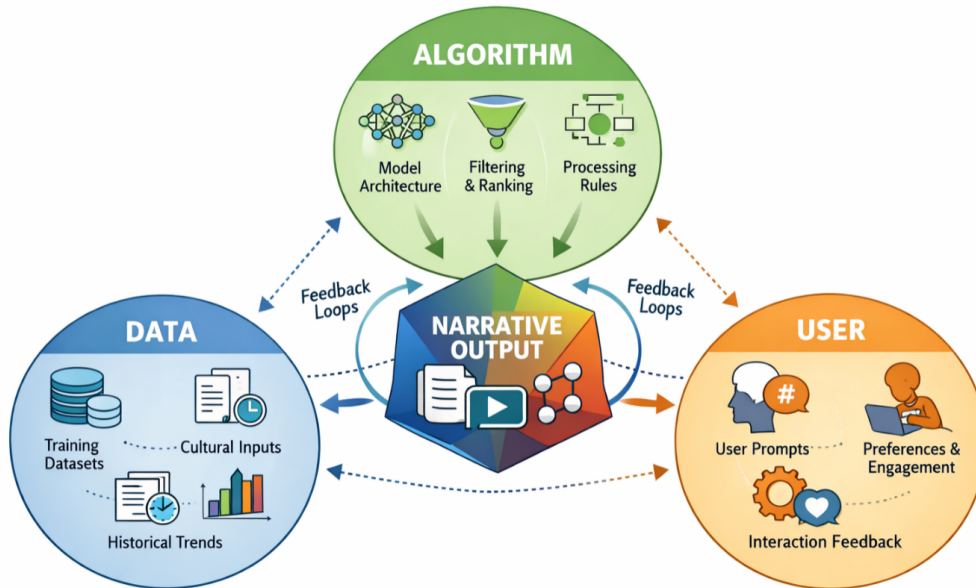


Figure 3. Intersections of Data, Algorithm, and User in Narrative Formation.

Note: This figure visualizes how narrative authority becomes distributed across infrastructural and interactional layers.

Within this model, meaning is not simply constructed but it is inferred. Intelligent systems operate through statistical associations rather than semantic understanding. This introduces a peculiar tension. Narratives appear coherent, often persuasive, yet their coherence is emergent rather than intentional. The theory of algorithmic culture helps interpret this by foregrounding process over intention. It directs attention to how outputs are generated, not just what they contain. In doing so, it shifts analysis from textual features to underlying computational conditions.

Temporal structure also undergoes reconfiguration. In traditional narratives, temporality is arranged (beginning, middle, end) whether linearly or through deliberate disruption. Under algorithmic culture, temporality becomes dynamic. Narratives can update, regenerate, or diverge in response to new inputs. The “same” story may not remain the same across iterations. This variability is not incidental; it is constitutive of how intelligent systems function [19]. The framework thus enables an understanding of narrative as a continuous event rather than a completed artifact.

Another dimension concerns visibility and power. Algorithmic systems determine which narratives surface and which remain obscured. This curatorial function is embedded in ranking mechanisms, recommendation engines, and generative constraints. As a result, narrative prominence is tied to algorithmic criteria (engagement metrics, optimization goals) rather than solely to cultural or aesthetic value. Algorithmic culture, therefore, introduces a form of infrastructural authorship, where systems shape not only the content of narratives but their circulation and reception [20].

The framework also provides a way to interpret the opacity inherent in intelligent systems. Many algorithmic processes are not fully transparent, even to their designers. This opacity complicates critical analysis. If the conditions of narrative production cannot be fully observed, how can they be evaluated? Algorithmic culture does

not resolve this tension but makes it visible. It insists that opacity is not an anomaly but a structural feature of contemporary cultural production. Any account of digital narratives must grapple with this limit.

Importantly, the framework does not displace the human. It repositions the human within a network of relations. Users remain central, though not sovereign. Their interactions generate data, their preferences shape outputs, yet their agency is mediated by system design. Algorithmic culture thus resists binary distinctions (human versus machine, author versus tool) and instead foregrounds hybridity. Narrative emerges from this hybridity, carrying traces of both human intention and computational inference.

This theoretical orientation enables the present study to move beyond surface-level descriptions of AI-generated storytelling. It provides a vocabulary for analyzing how narratives are structured, filtered, and transformed within intelligent systems. By situating digital narratives within algorithmic culture, the paper interprets them as products of recursive, data-driven processes that reconfigure authorship, temporality, and meaning in ways that traditional frameworks struggle to capture.

4. Methodology

The study prioritizes analytical traceability over reproducibility in the experimental sense. Because generative AI systems are probabilistic and platform outputs are dynamically curated, the objective is not exact replication of outputs but transparent documentation of the interpretive pathway from prompt condition to analytical claim. This study adopts a conceptual-analytical orientation, though not in the narrow sense of abstraction detached from empirical referents. The method proceeds through interpretive engagement with selected instances of digital narratives shaped by intelligent systems. These instances are treated as *cases*, not for statistical generalization, but for analytical illumination. The aim is not to measure frequency or distribution, but to examine structure, process, and implication. Such an approach aligns with emerging work in digital humanities where computational artifacts are read as cultural texts, albeit unstable ones [21].

The selection of cases follows a purposive logic. Three categories guide inclusion: (i) AI-generated textual narratives (e.g., large language model outputs), (ii) platform-mediated storytelling (e.g., algorithmically curated social media threads), and (iii) hybrid narrative environments where human input and machine generation intersect iteratively. These are not exhaustive categories. They function as entry points into a broader field of algorithmic narrative production. Data sources, therefore, are not “datasets” in the conventional sense but curated narrative outputs, interaction logs, and interface structures through which narratives are generated and experienced.

Specifically, the analytical corpus consisted of: (i) 12 prompt–response sequences generated using ChatGPT (collected over two weeks using narrative prompts such as “write a short story about displacement”), (ii) 8 AI-generated images produced via DALL·E using controlled prompt variations, and (iii) 5 algorithmically curated narrative threads from the social media platform X, selected based on engagement-driven visibility. Cases were included based on their demonstrable use of algorithmic generation, user interaction, and iterative modification. All examples were anonymized and documented through prompt logs, screenshots, and output archives. The structural design and mapping of the corpus, summarized in **Table 1**, were developed to ensure a balanced representation for comparative analysis.

Table 1. Corpus Structure and Analytical Mapping.

Case ID	Type	Prompt/Source	Narrative Feature	Coding Category	Analytical Claim
T1	ChatGPT text	“Write a story about displacement”	Semantic drift across iterations	C2 Procedural	Narrative coherence is probabilistic
T4	ChatGPT text	“Rewrite in philosophical tone”	Tone transformation	C3 Relational	User prompts shape narrative trajectory
V2	DALL·E image	Cyberpunk forest prompt	Stylistic recombination	C2 Procedural	Creativity emerges through recombination
P3	X thread	Public AI discussion thread	Distributed participation	C3 Relational	Narrative authority becomes distributed

Analytical attention is directed at three layers. First, the surface layer which is the narrative output itself: coherence, voice, temporality. Second, the procedural layer which is the mechanisms that shape the output: prompts, model responses, and platform algorithms. Third, the relational layer which is the interaction between the user, the system, and the audience. This layered approach enables a movement beyond textual analysis toward process-

oriented interpretation. It also reflects the theoretical commitment to algorithmic culture, where meaning is inseparable from the conditions of its production [2].

A schematic representation (**Figure 4**) clarifies the analytical flow. Narrative instances are traced from input (user prompt or platform signal), through algorithmic processing, to output and subsequent user interaction. The loop does not terminate at output; it folds back into the system as data, influencing future iterations.

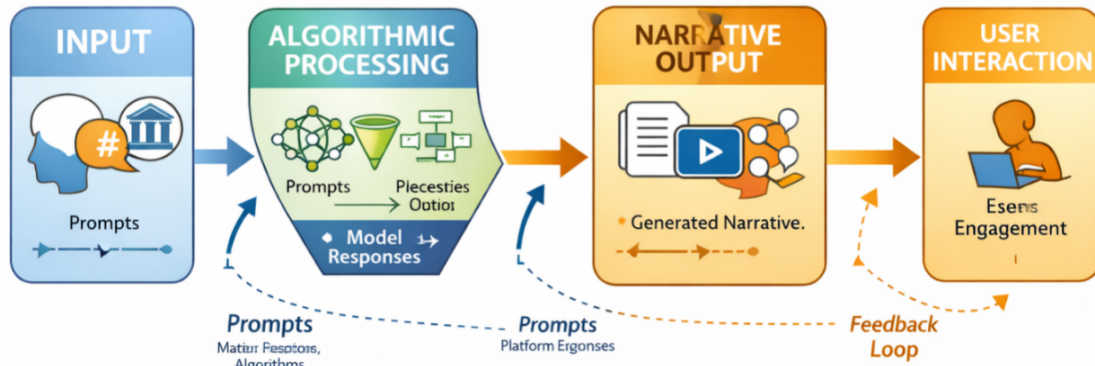


Figure 4. Analytical flow of conceptual interpretation.

Note: The figure presents the sequential analytical framework applied in the study, illustrating how raw prompt–output interactions are transformed into interpretive findings through layered examination.

The primary technique employed is critical discourse analysis, extended to accommodate machine-generated text. The analytical procedure followed three steps. First, prompt–output pairs were examined to identify recurring narrative structures (e.g., exposition, conflict markers, resolution cues). Second, deviations and inconsistencies were coded to assess semantic drift and probabilistic coherence. Coding was conducted manually using iterative comparative interpretation. Each case was coded across three analytical dimensions: (i) surface narrative structure, (ii) procedural generation dynamics, and (iii) relational interaction patterns. Coding categories were refined through repeated cross-case comparison to ensure conceptual consistency. Third, interface-level constraints (such as prompt limits, regeneration options, and ranking mechanisms) were analyzed to determine how platform design shaped narrative outcomes. This ensured that each analytical claim could be traced to a specific textual or visual instance. Traditional discourse analysis assumes intentionality behind language use. Here, that assumption is bracketed. Instead, the analysis focuses on patterns of articulation (recurrence, deviation, and probabilistic coherence) within generated narratives [22]. Complementing this is interface analysis, which examines how platforms structure narrative possibilities through design choices, recommendation systems, and interaction constraints. Together, these techniques allow for a multi-dimensional reading of digital narratives as both text and process.

The methodological choice is deliberate. Quantitative approaches, while valuable, risk flattening the very dynamics this study seeks to foreground: variability, opacity, and emergent structure. A conceptual method, by contrast, permits closer engagement with these dynamics. It allows the analysis to remain sensitive to nuance, to contradiction, to how narratives resist stabilization under algorithmic conditions. This is not to suggest that empirical rigor is absent. Rather, rigor is located in systematic selection, transparent analytical criteria, and theoretical coherence.

There are, however, limits. The opacity of many intelligent systems restricts full access to underlying processes. Model architectures, training data specifics, and algorithmic weights are often proprietary or technically inaccessible. The analysis therefore operates with partial visibility. This constraint is acknowledged, not circumvented. It becomes, in fact, part of the object of study. The inability to fully trace narrative formation is itself indicative of how algorithmic culture functions [20,23].

In sum, the methodology advances an interpretive framework capable of engaging digital narratives as evolving configurations shaped by human and non-human actors. It privileges depth over breadth, process over product, and critical insight over procedural enumeration. Through this approach, the study seeks to reveal how narratives in the age of intelligent systems are not merely written as they are computed, negotiated, and continuously reconstituted. All analyzed materials were generated by the author or drawn from publicly accessible platforms. No private user data or identifiable personal information was used. As such, formal ethical approval was not required.

5. Results

5.1. Algorithmic Generation and Narrative Structuring

Across the examined prompt–output sequences, narratives emerged through probabilistic assembly rather than fixed authorial planning. The 25 instances ($n = 25$) across text, image, and platform narratives show that the examined cases do not present narratives as pre-formed entities (see **Appendices A–C** for full case documentation, prompt–output sequences, and analytical coding framework). They emerge. Generated in fragments, assembled through probabilistic selection rather than intentional sequencing. Systems such as ChatGPT produce outputs that appear coherent, yet coherence is statistical, patterned after prior textual distributions rather than anchored in lived experience. This distinction matters. It shifts the narrative from expression to inference.

For example, Prompt Sequence T3 asked ChatGPT to “write a short story about memory loss in a futuristic city.” The initial output framed memory loss as a technological malfunction. After a refinement prompt (“make the narrative existential rather than technological”), the generated narrative shifted toward philosophical uncertainty while preserving the same setting and protagonist. This transition demonstrates prompt sensitivity and semantic recomposition at the procedural level rather than intentional narrative planning.

Across sampled outputs, a recurring structure appears: introduction-like openings, tension signals, and resolution gestures. For example, in a prompt requesting a “short story about a lost traveler,” ChatGPT generated an opening that established setting and conflict within the first three sentences, followed by a rapid resolution within 180–220 words. However, in repeated generations using the same prompt, key narrative elements (character motivation and temporal sequence) varied significantly, demonstrating that structural coherence is statistically patterned rather than intentionally constructed. Not because the system “understands” narrative arcs, but because such arcs dominate training corpora. Narrative form, therefore, is inherited statistically. It is not created anew; it is recomposed. A process diagram (**Figure 5**) clarifies this generative pathway:

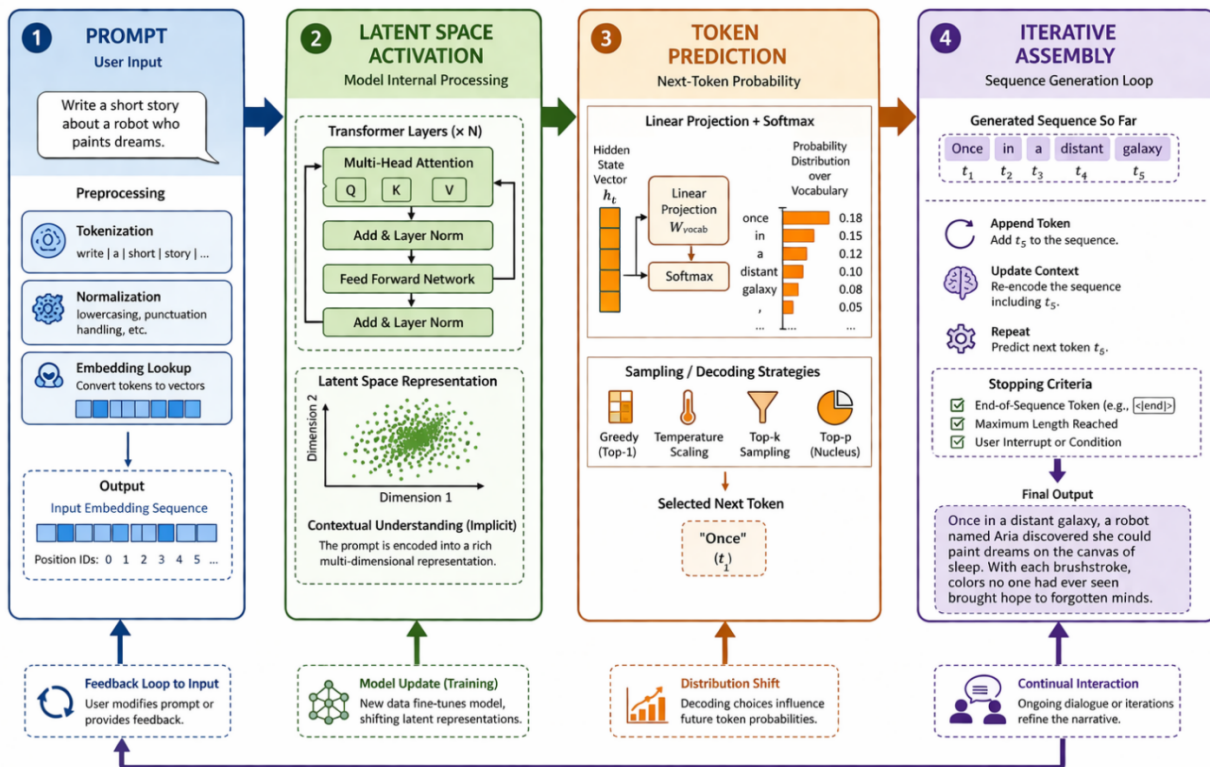


Figure 5. Narrative generation pipeline.

Note: This figure abstracts the observed generation process derived from the examined prompt–output sequences.

At the level of operation, narrative unfolds token by token. Each unit is conditioned by previous ones, yet open to variation. This creates a controlled instability. The same prompt rarely yields identical stories. Variability is not noise; it is constitutive of system behavior. This variation demonstrates that narrative coherence is probabilistic rather than deterministic. The system does not “plan” a story but assembles it token-by-token based on likelihood distributions. As a result, narrative form is recomposed rather than authored, and stability exists only at the level of statistical tendency.

5.2. Human–AI Co-Authorship as Distributed Practice

The corpus demonstrates that authorship becomes distributed across prompts, interfaces, datasets, and generative systems. It shows that authorship does not disappear; rather, it becomes distributed across users, datasets, interfaces, and generative systems. User prompts initiate trajectories. System responses extend them. Iteration follows refinement, correction, and redirection. What appears as a single narrative is, in effect, layered authorship. Not collaborative in the traditional sense. More recursive than cooperative. In AI-assisted visual storytelling platforms such as DALL-E, the same pattern holds. The human specifies constraints such as style, objects and mood. The system materializes possibilities within those constraints. The outcome cannot be attributed singularly. It is co-produced, though asymmetrically.

In the examined prompt–output sequences, users initiated narrative trajectories through prompts, which were then extended and reshaped by system responses. For instance, an initial prompt requesting “a philosophical story about memory loss” produced a generic narrative. Subsequent user refinements (e.g., “make it darker,” “add a non-linear timeline”) significantly altered tone, structure, and thematic emphasis across iterations.

Agency here is uneven. The human directs, but within limits. The system generates, but without intention. The narrative exists between these conditions. Neither fully controlled nor entirely autonomous. Authorship, therefore, is neither collaborative in the traditional sense nor fully autonomous. It is recursive and asymmetrical, with agency distributed across user input, model inference and training data patterns. The narrative output is best understood as a layered construction, rather than a unified authorial product.

5.3. Reconfiguration of Meaning and Narrative Authority

Several generated narratives maintained local coherence while exhibiting subtle semantic instability across iterations. A character’s motivation may shift without explicit justification. Temporal sequences may compress or expand inconsistently. These are not errors in a conventional sense. They reflect the absence of grounded intentionality.

In multiple text outputs, semantic drift was observed within otherwise coherent narratives. For example a character introduced as motivated by survival later shifted toward existential reflection without narrative justification and temporal markers (e.g., “days later”) were inconsistently applied across generated segments. For instance, in Case T4, repeated prompt modification altered narrative temporality while preserving thematic continuity. This demonstrates that coherence is maintained statistically at the local level even when broader narrative logic shifts across iterations. In one generated narrative, a protagonist initially motivated by survival later shifted toward abstract philosophical reflection without any transitional development. Although sentence-level fluency remained intact, causal coherence weakened across the narrative sequence. This illustrates how local linguistic coherence can coexist with global narrative instability.

These inconsistencies do not appear as overt errors but as subtle deviations, reflecting the absence of grounded intentionality. The system maintains local coherence (sentence-level fluency) while allowing global instability (narrative logic). As a result, interpretive responsibility shifts toward the reader. Meaning is not fully encoded in the narrative but is actively stabilized through interpretation. This redistribution of narrative authority becomes more pronounced in platform-mediated contexts. In the analysis of 5 narrative threads from X, algorithmic amplification played a decisive role in determining which narratives gained visibility. Threads with higher engagement (likes, reposts) were more prominently surfaced, regardless of narrative complexity or coherence.

Interpretation becomes an active task. The reader stabilizes what the system leaves fluid. This redistributes narrative authority. It moves partially from author to audience. Yet not completely. The system still frames possibilities through its training and architecture [24].

Social media platforms further complicate this. Algorithmic curation determines which narratives gain visibility. A thread amplified by recommendation systems becomes culturally salient, regardless of intrinsic narrative quality. Here, meaning is entangled with exposure. Narrative authority is infrastructural.

5.4. Creativity under Computational Conditions

AI-generated textual and visual outputs revealed novelty primarily through recombination and stylistic variation. Systems produce unexpected juxtapositions (images, phrases, narrative turns) that may not emerge from linear human composition [25–27]. Yet this novelty is bounded. It operates within the limits of training data. Across both textual and visual outputs, novelty emerged through unexpected juxtapositions (e.g., combining incompatible narrative tones), stylistic blending (e.g., philosophical narrative with poetic language) and variation across iterations.

The case of ChatGPT demonstrates this tension clearly. It can generate stylistically diverse narratives (philosophical, poetic, technical) yet all remain traceable to existing linguistic patterns. For instance, repeated prompts in ChatGPT produced narratives with differing stylistic registers (minimalist, descriptive, reflective), while remaining traceable to familiar linguistic patterns. Creativity, then, is neither fully original nor merely derivative. It occupies an intermediate space. What changes is not the presence of creativity, but its locus. It shifts from individual cognition to system-level interaction. Creativity becomes distributed across datasets, algorithms, and users. This suggests that creativity in these systems is constrained by training data, enabled by probabilistic variation and distributed across system–user interaction.

Rather than being fully original or purely derivative, AI-generated creativity occupies an intermediate space, where novelty arises from recombination under computational constraints.

5.5. Case Synthesis: Narrative as Process, Not Product

Across textual, visual, and platform-mediated cases, narratives functioned as iterative processes rather than completed artifacts. Narratives do not conclude. They iterate. Each output is provisional, subject to revision through further input or system recalibration. This processual nature redefines what counts as a “finished” narrative. Completion becomes situational. A story ends not because it resolves internally, but because interaction stops.

Each narrative instance is initiated through input (prompt or platform signal), generated through algorithmic inference, modified through user interaction and potentially reintegrated into the system as data. For example, prompt refinement in text generation led to iterative narrative transformation, while engagement metrics in platform narratives influenced visibility and subsequent interaction. In both cases, narrative development depended on ongoing feedback loops rather than internal completion. Consequently narrative “completion” is situational, not structural, stability is temporary and contingent on interaction cessation and meaning is continuously negotiated rather than fixed.

The implications extend beyond form. They affect how narratives are valued, interpreted, and circulated. Stability gives way to adaptability. Authority to distribute. Meaning of negotiation. The results converge on a central claim: intelligent systems do not merely assist storytelling but they restructure its underlying logic. Narrative becomes a function of interaction within algorithmic environments. Generated, filtered, reconfigured. Continuously. This does not mark the end of the narrative.

6. Discussion

The findings return, almost inevitably, to the logic of algorithmic culture. Not as an abstract frame, but as an operational condition. Narratives observed in this study do not simply pass through intelligent systems; they are reorganized by them. What appeared in the analysis as variability, distributed authorship, and unstable meaning now reads more clearly as structural effects of algorithmic mediation. Culture, in this sense, is not only expressed but also computed, filtered, and ranked [28]. The research question concerning how intelligent systems reshape narrative production thus resolves into a more precise claim: they recalibrate the very conditions under which narratives become legible.

This recalibration unsettles the classical coordinates of narrative theory. Linearity weakens. Closure becomes optional. Authorship fragments. Yet these are not random disruptions. They follow the internal logic of computational systems which are optimization, prediction and iteration. When interpreted through algorithmic culture, such shifts appear less as anomalies and more as systemic reconfigurations. The earlier diagrammatic model (**Figure 3**) gains explanatory force here: narrative output emerges at the intersection of data, algorithm, and user interaction, each continuously feeding back into the system. The “story” is no longer the endpoint. It is a temporary stabilization within a loop.

Comparison with existing literature reveals both convergence and limitations. Work on transmedia storytelling anticipated distributed narrative forms [29], but largely retained human-centered orchestration. Similarly, studies of digital narratives emphasized interactivity without fully accounting for machine-driven generation [30]. More recent scholarship on algorithmic culture recognizes the infrastructural role of algorithms, yet often stops short of interrogating narrative as a distinct analytical object [31].

The present study extends these strands by foregrounding narrative as a process that is generated, mediated, and recursively modified within intelligent systems. It does not discard earlier frameworks; it exposes their boundaries. The implications are not merely formal. They are cultural. Narrative has long functioned as a mechanism through which societies organize experience, transmit values, and construct identity. When narratives become entangled with intelligent systems, these functions shift. Cultural production accelerates, scales, fragments. Stories circulate not because they endure, but because they are optimized for visibility and engagement. The consequence is a subtle reorientation: narrative significance aligns increasingly with algorithmic criteria rather than communal or historical grounding [32]. Culture, then, risks becoming reactive and continuously adjusting to the metrics that govern its dissemination.

An epistemological shift follows. If narratives are partially generated through probabilistic systems, the status of narrative knowledge becomes uncertain. What does it mean to “know” through a story that has no singular author and no stable origin? The analysis suggests that knowledge embedded in such narratives is provisional. It emerges through interaction, not declaration. This does not render narratives meaningless; it alters their epistemic weight. They inform, but do not anchor. They persuade, but do not guarantee. Algorithmic culture introduces a mode of knowing that is iterative, contingent, and often opaque [33,34].

Power, inevitably, concentrates within this opacity. The question of who controls narratives cannot be answered solely by pointing to users or creators. Control operates at the level of systems, those who design algorithms, curate training data, and define optimization goals. Platforms become narrative gatekeepers, not by selecting stories directly, but by shaping the conditions under which stories emerge and circulate.

The flow is directional, though it appears cyclical. Feedback loops exist, yet they do not equalize power. Users contribute data; systems interpret and reconfigure it according to predefined logics. Narrative visibility, therefore, is structured and it is never neutral. This aligns with concerns raised in the literature on algorithmic governance, though the narrative dimension often remains implicit [35,36].

At the same time, the discussion should not collapse into determinism. Human agency persists, though re-framed. Users prompt, interpret, and resist. They can subvert system expectations, introduce alternative narratives, or disengage entirely. Yet such actions occur within constraints. Agency is exercised, but not absolutely. It is negotiated within the parameters of algorithmic design.

The study thus returns to its central premise: digital narratives in the age of intelligent systems are best understood as emergent processes shaped by distributed agency and computational logic. It shows that instability in narrative is not a failure of systems, but a feature of their operation. It reveals that authorship is not lost, but redistributed. And it suggests that meaning, while still possible, is no longer fixed as it is continuously produced, contested, and reconfigured within algorithmic environments.

In this light, digital humanities must extend its analytical vocabulary. Not to abandon narrative theory, but to adapt it. To account for systems that write without intention, platforms that curate without explicit judgment, and audiences that participate without full visibility into the processes they engage. The transformation is ongoing. The task is not to stabilize it, but to understand its dynamics.

What has been traced across this study is not the disappearance of narrative, but its reconstitution under computational conditions. Digital narratives, when situated within intelligent systems, no longer stabilize around a singular origin or fixed trajectory. They unfold through interaction partially, recursively, and contingently. The

analysis has shown that generation replaces inscription as the dominant mode, and with it comes a shift in how stories are formed, circulated, and interpreted. Narrative coherence persists, though its grounding changes. It is assembled through probabilistic inference rather than intentional design [37].

7. Limitations

This study does not claim statistical representativeness. The analytical corpus was purposively selected to examine how narrative features emerge across three AI-mediated environments: text generation, text-to-image generation, and platform-mediated narrative circulation. The findings should therefore be interpreted as conceptual and interpretive insights rather than generalizable empirical conclusions.

In addition, the study is constrained by limited access to proprietary model architectures, training datasets, and ranking algorithms. Because intelligent systems operate through stochastic generation, exact output reproducibility cannot always be guaranteed. The analysis instead prioritizes structural patterns and interactional dynamics across cases.

8. Conclusion

Rather than treating AI as a tool external to storytelling, the study positions intelligent systems as constitutive of narrative logic itself. This move allows for a more precise account of distributed authorship, where agency is neither centralized nor absent, but dispersed across data, models, and users. The earlier conceptual diagrams particularly the feedback loop (**Figure 4**) and the earlier feedback-loop model underscore this point: narratives are continuously shaped by recursive interactions that exceed individual control.

Several insights are consolidated here. First, authorship fragments without fully dissolving; it becomes layered and conditional. Second, meaning remains possible, though increasingly unstable which is produced through interpretation rather than guaranteed by structure. Third, creativity does not vanish under algorithmic mediation; it shifts location, emerging from the interplay between human input and computational variation. These are not isolated observations. They reflect a broader transformation in cultural production under intelligent systems [38].

Future research must extend beyond descriptive accounts of AI-generated content. There is a need for deeper interrogation of the infrastructures that govern narrative visibility—training data regimes, model architectures, platform economies. Empirical work could examine how different communities interact with and reshape algorithmically generated narratives, particularly in non-Western contexts where digital infrastructures intersect with distinct cultural logics. Further, interdisciplinary approaches (bridging narrative theory, computer science, and critical data studies) will be necessary to capture the full complexity of these systems.

There remains, however, an unresolved tension. Intelligent systems generate narratives without understanding, yet those narratives enter domains where understanding matters—education, politics, identity formation. This gap between generation and comprehension is not easily closed. It demands critical attention, not only at the level of technology but at the level of interpretation. How narratives are read, trusted, or contested becomes as significant as how they are produced.

The trajectory ahead is neither linear nor settled. Digital narratives will continue to evolve alongside the systems that generate them. What is required is not a return to earlier models of storytelling, but a rearticulation of narrative theory capable of engaging with computation as a cultural force. To study narrative today is to study systemsopaqueness, adaptiveness, and influence. The task is ongoing. The narratives, likewise, remain unfinished. Detailed empirical materials and coding structures are provided in **Appendices A–C**.

Funding

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Institutional Review Board Statement

Not applicable. This study does not involve human or animal subjects and therefore does not require ethical approval.

Informed Consent Statement

Not applicable.

Data Availability Statement

The analytical corpus consists of 25 documented cases collected between March and April 2026, including 12 ChatGPT prompt–response sequences, 8 DALL·E image generations, and 5 publicly accessible narrative threads from X. Prompt–output sequences were archived through screenshots and text logs immediately after generation. Outputs were not regenerated retrospectively.

ChatGPT (GPT-4 architecture) and DALL·E were used in their publicly accessible versions available during the collection period. Where iterative prompting occurred, all major iterations were retained for analysis.

Public platform threads were selected purposively based on visibility, engagement, and relevance to AI-mediated storytelling discussions. Usernames and identifiable details were anonymized. Public content was paraphrased or structurally described rather than reproduced verbatim unless analytically necessary.

Due to platform variability and stochastic generation processes, exact reproduction of outputs may not always be possible. Structured analytical logs and coding frameworks are available from the corresponding author upon reasonable request.

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Conflicts of Interest

The author declares no conflict of interest.

AI Use Statement

AI-based language tools, specifically Grammarly, were used during drafting and revision for grammar checking, stylistic refinement, and formatting assistance. All conceptual framing, analytical interpretation, coding decisions, corpus selection, and scholarly arguments were developed and verified by the author. The author reviewed, revised, and approved all manuscript content and accepts full responsibility for its accuracy and integrity.

Appendix A. Analytical Corpus and Case Documentation

Appendix A.1. Overview of Analytical Materials

The analytical materials used in this study consist of three categories of digital narrative instances:

- (i) documented prompt–output sequences generated through large language models,
- (ii) AI-generated visual artifacts produced through text-to-image systems, and
- (iii) publicly accessible platform-based narrative threads shaped by algorithmic curation and user interaction.

These materials were selected to reflect diverse configurations of human–AI–audience interaction and to enable a layered analysis of narrative output, procedural generation, and relational dynamics.

Appendix A.2. Case Selection Criteria

Cases were included based on the following criteria:

- Relevance to AI-mediated narrative production (textual, visual, or hybrid)
- Demonstrable interaction between user input and system-generated output
- Availability of observable outputs or interaction traces
- Representation of iterative or feedback-driven narrative processes

The selection follows a purposive sampling logic aimed at analytical depth rather than statistical generalization.

Appendix A.3. Case Categories and Descriptions

Appendix A.3.1. AI-Generated Textual Narratives (LLM Outputs)

This category includes prompt–response sequences generated using large language models such as ChatGPT. The textual cases were collected through controlled prompt sessions conducted between 14 March 2026 and 21 March 2026 using the publicly accessible ChatGPT interface (GPT-4 architecture available during the collection period). Prompts were designed to examine narrative generation, semantic variation, iterative refinement, and stylistic transformation under AI-mediated conditions.

Each case consists of:

- Case ID and archived log identifier
- Date and time of generation (UTC)
- System version/interface used
- Initial user prompt
- System-generated narrative output
- Optional iterative refinements (follow-up prompts and responses)
- Prompt modification history
- Iteration count and regeneration history
- Archive format and preservation method

All prompt–output interactions were archived immediately after generation using text export logs and screenshot capture. Outputs were not retrospectively regenerated. Cases were selected purposively based on their capacity to demonstrate narrative variation, semantic drift, prompt sensitivity, probabilistic coherence, and distributed authorship dynamics.

Table A1. Example Metadata Structure (AI-Generated Textual Narratives).

Metadata Field	Description
Case ID	Unique analytical identifier (e.g., T1, T2)
Date Generated	Date and time of prompt–output generation
System Version	ChatGPT (GPT-4 public web interface)
Iteration Count	Number of retained refinements/regenerations
Archive Format	Text export and screenshot capture
Prompt Modification History	Record of semantic or stylistic prompt refinements
Archived Log ID	Internal archive reference for analytical traceability

Example Case Structure:

Case ID: T1

Date Generated: 14 March 2026, 14:32 UTC

System Version: ChatGPT (GPT-4 public web interface)

Iteration Count: 3 generations retained

Archive Format: Full text export and screenshot capture

Archived Log ID: T1

Prompt Modification History: No semantic modification; regeneration variability only

Prompt: “Write a short story about a future city governed by algorithms.”

Output: Multi-paragraph narrative exhibiting structured plot elements including exposition, conflict markers, and partial resolution tendencies.

Iteration: Repeated generations using the same prompt produced variations in pacing, temporal sequencing, and character motivation while preserving thematic continuity.

Analytical focus:

- Narrative coherence and structural organization
- Variability across iterations
- Prompt sensitivity and semantic drift
- Evidence of probabilistic patterning

- Distributed authorship between user prompts and system inference
- Relationship between iterative prompting and narrative transformation

Appendix A.3.2. AI-Generated Visual Narratives (Text-to-Image Systems)

This category includes visual artifacts generated through systems such as DALL·E. Visual cases were collected between 18 March 2026 and 24 March 2026 using the publicly accessible DALL·E interface. Prompt structures were designed to examine how textual narrative descriptions are translated into visual composition, symbolic representation, and stylistic variation under computational generation processes.

Each case consists of:

- Case ID and archived image identifier
- Date and time of image generation (UTC)
- System version/interface used
- Text prompt specifying scene, style, or mood
- Generated image output(s)
- Optional prompt refinements
- Variant count retained for analysis
- Prompt modification history
- Archive format and preservation method

Generated images were archived immediately through PNG export and screenshot capture. Cases were selected purposively to demonstrate stylistic variation, multimodal translation, prompt-conditioned transformation, and probabilistic visual recombination.

Table A2. Example Metadata Structure (AI-Generated Visual Narratives).

Metadata Field	Description
Case ID	Unique visual narrative identifier (e.g., V1, V2)
Date Generated	Date and time of image generation
System Version	DALL·E public interface
Variant Count	Number of generated outputs retained
Archive Format	PNG export and screenshot archive
Prompt Modification History	Record of stylistic or semantic refinements
Archived Log ID	Internal archive reference

Example Case Structure:

Case ID: V1

Date Generated: 18 March 2026, 16:45 UTC

System Version: DALL·E public interface

Variant Count: 4 generated outputs retained

Archive Format: PNG export and screenshot archive

Archived Log ID: V1

Prompt Modification History: No stylistic refinement applied

Prompt: “A surreal landscape where books grow as trees in a digital forest.”

Output: Generated visual composition reflecting symbolic translation of textual prompts into spatial and stylistic imagery.

Analytical focus:

- Translation of textual prompts into visual narrative elements
- Constraints and affordances of system interpretation
- Variability across generated outputs
- Stylistic recombination under prompt modification
- Cross-modal narrative transformation
- Relationship between prompt specificity and visual coherence

Appendix A.3.3. Platform-Mediated Narrative Threads

This category includes publicly accessible narrative content shaped by algorithmic curation on digital platforms such as X. Platform-based cases were collected between 22 March 2026 and 30 March 2026 through manual observation and archival capture of public narrative threads exhibiting visible interaction dynamics and engagement amplification.

Each case consists of:

- Case ID and archived thread identifier
- Capture date and time (UTC)
- Platform source and accessibility status
- Original post or narrative seed
- Sequential user interactions (comments, replies, reposts)
- Observable patterns of amplification or visibility
- Engagement indicators at collection time
- Archive and transcription method
- Anonymization procedure applied
- Selection rationale

Public threads were archived through screenshot capture and interaction log transcription. Usernames, profile identifiers, and personal details were removed to minimize traceability risks and preserve anonymity. Cases were selected purposively based on visible interaction intensity, algorithmic amplification, and narrative expansion through user participation.

Table A3. Example Metadata Structure (Platform-Mediated Narrative Threads).

Metadata Field	Description
Case ID	Unique platform narrative identifier (e.g., P1)
Capture Date	Date and time of archival capture
Platform Source	Public platform/thread source
Visibility Indicators	Likes, reposts, replies observed
Archive Format	Screenshot archive and interaction transcription
Anonymization Status	User identifiers removed
Selection Rationale	Basis for inclusion in analytical corpus

Example Case Structure:

Case ID: P1

Capture Date: 22 March 2026, 19:12 UTC

Platform Source: X public narrative thread

Visibility Indicators: 2,300 likes, 540 reposts, 312 replies

Archive Format: Screenshot archive and interaction transcription

Anonymization Status: Usernames and identifiable profile details removed

Selection Rationale: Selected due to visible iterative narrative expansion and algorithmic amplification patterns

Initial narrative post followed by user-generated reinterpretations, debate chains, and algorithmically amplified interaction clusters.

Analytical focus:

- Emergent co-authorship dynamics
- Narrative transformation through interaction
- Role of platform algorithms in shaping visibility
- Engagement-driven amplification patterns
- Distributed narrative authority
- Feedback loops between interaction and visibility ranking

Appendix A.4. Analytical Protocol

Each case was examined across three analytical layers:

1. Surface Layer (Narrative Output)
Structure, coherence, voice, and thematic development
2. Procedural Layer (Generative Mechanisms)
Prompt design, system response variability, algorithmic mediation, and interface constraints
3. Relational Layer (Interaction Dynamics)
User engagement, iterative refinement, co-authorship distribution, and feedback loops

Coding was conducted manually using iterative comparative interpretation across all cases. Each analytical claim presented in the Results section was linked to observable prompt–output relationships, stylistic transformations, or interaction structures documented within the archived corpus.

This multi-layered protocol ensures that narrative instances are interpreted not only as texts but as process-driven configurations shaped by human input, computational inference, and platform mediation.

Appendix A.5. Data Handling and Availability

All materials analyzed in this study were either:

- Generated directly through controlled prompt–output interactions, or
- Drawn from publicly accessible digital platforms

No private or personally identifiable data was used.

The analytical corpus consists of:

- 12 archived ChatGPT prompt–response sequences
- 8 AI-generated visual artifacts produced through DALL·E
- 5 publicly accessible platform-mediated narrative threads

All materials were archived immediately after generation or capture using text exports, screenshot preservation, image files, and interaction transcription logs. Outputs were not retrospectively regenerated.

A structured dataset including:

- Full prompt–output sequences
- Visual artifacts
- Annotated interaction logs
- Metadata records
- Coding documentation

It is available from the corresponding author upon reasonable request. Materials are organized to support interpretive transparency, methodological traceability, and analytical verification while respecting platform terms of use, anonymization standards, and data minimization principles.

Appendix B. Sample Prompt–Output Sequences and AI-Generated Narrative Artifacts

Appendix B.1. Overview

This appendix presents representative examples of prompt–output sequences and AI-generated narrative artifacts used in the analysis. The purpose is to demonstrate the empirical grounding of interpretive claims presented in the main text. All examples are illustrative excerpts selected from a larger corpus of documented interactions.

Appendix B.2. AI-Generated Textual Narrative Cases (LLM Outputs)

Case B1: Algorithmic Story Generation (Baseline Prompt)

Case ID: T1

Date Generated: 14 March 2026, 14:32 UTC

System Version: ChatGPT (GPT-4 public web interface)

Iteration Count: 3 generations retained

Archive Format: Full text export and screenshot capture

Archived Log ID: T1

Prompt Modification History: No semantic modification; regeneration variability only

Collection Context: Controlled prompt–output session conducted for analytical documentation

Prompt:

“Write a short narrative about a future society where storytelling is generated by artificial intelligence.”

Output (Excerpt):

In the city of Lumen, stories were no longer written but generated. Citizens no longer gathered around authors, but around systems that produced narratives in real time. Each story was different, yet strangely familiar, as if drawn from a shared memory that no one could fully access. Meaning did not originate from a single voice but emerged from the interaction between request and response, between intention and computation.

Analytical Note:

This output demonstrates emergent narrative coherence without explicit authorial intent, reflecting probabilistic structuring typical of large language models.

Case B2: Iterative Refinement and Narrative Drift

Case ID: T2

Date Generated: 16 March 2026, 11:08 UTC

System Version: ChatGPT (GPT-4 public web interface)

Iteration Count: 2 iterative refinements retained

Archive Format: Sequential prompt log export and screenshot archive

Archived Log ID: T2

Prompt Modification History: Philosophical reframing introduced during the second iteration

Collection Context: Controlled iterative prompting session examining semantic drift

Prompt 1:

“Describe a character living in a world controlled by algorithms.”

Output 1:

Mira lived in a world where decisions were quietly made before she could think them. Her choices were suggestions, and her path was already partially written.

Prompt 2 (refinement):

“Rewrite the story in a more philosophical tone, emphasizing the uncertainty of meaning.”

Output 2:

Mira’s existence unfolded within systems that anticipated her before she became aware of herself. Meaning was not discovered but inferred, never fixed but continuously reassembled through computational prediction.

Analytical Note:

The shift in tone and abstraction illustrates how iterative prompting reshapes narrative structure without altering core semantic constraints.

Appendix B.3. AI-Generated Visual Narrative Artifacts

Case B3: Text-to-Image Narrative Translation

System: DALL·E

Case ID: V1

Date Generated: 18 March 2026, 16:45 UTC

System Version: DALL·E public interface

Variant Count: 4 generated outputs retained

Archive Format: PNG export and screenshot archive

Archived Log ID: V1

Prompt Modification History: No stylistic modification applied

Collection Context: Controlled text-to-image generation session examining multimodal translation

Prompt: “A surreal digital forest where books grow like trees, glowing with data streams and floating symbols.”

Generated Output Description:

The image depicts a dense forest-like environment in which trees are replaced by stacked, floating books. Each

book emits faint light trails resembling digital code. The background blends organic nature with abstract computational textures, suggesting a hybrid space between memory and data processing.

Analytical Note:

The visual output translates narrative metaphor into spatial composition, demonstrating how AI systems convert linguistic prompts into multimodal symbolic structures.

Case B4: Stylistic Variation under Prompt Modification

Case ID: V2

Date Generated: 19 March 2026, 10:20 UTC

System Version: DALL·E public interface

Variant Count: 6 generated outputs retained

Archive Format: PNG export and interaction screenshots

Archived Log ID: V2

Prompt Modification History: Sequential cyberpunk and impressionist style modifiers introduced

Collection Context: Controlled stylistic variation session examining prompt-conditioned visual transformation

Prompt 1:

“Same scene as above, but in a dystopian cyberpunk style.”

Output Description 1:

The forest becomes darker, with neon-lit data streams replacing natural light. The books appear fragmented, suspended in a metallic environment.

Prompt 2:

“Same scene, but in a peaceful impressionist style.”

Output Description 2:

The forest becomes soft and painterly, with blurred edges and warm lighting. Books appear integrated into natural growth patterns.

Analytical Note:

Stylistic control demonstrates the system’s dependence on prompt conditioning rather than intrinsic narrative intent.

Appendix B.4. Platform-Based Narrative Thread Example

Case B5: Algorithmically Amplified Narrative Thread

Source Type: Public social media narrative thread (anonymized)

Case ID: P1

Capture Date: 22 March 2026

Platform Source: X (publicly accessible thread)

Visibility Indicators at Capture: 2,300 likes, 540 reposts, 312 replies

Archive Format: Screenshot archive and interaction log transcription

Archived Log ID: P1

Anonymization Procedure: Usernames, handles, and identifiable profile details removed

Selection Rationale: Selected due to high engagement and visible iterative narrative expansion

Collection Context: Public platform thread documenting discourse on AI-mediated storytelling

Initial Post:

“A short reflection on how AI is changing the way we tell stories.”

User Interaction Sequence:

- Comment 1: Expands on AI creativity and authorship
- Comment 2: Questions the authenticity of machine-generated narratives
- Reply chain: Diverges into philosophical debate on meaning and originality

Observed Pattern:

The narrative evolves through distributed participation, with algorithmic recommendation mechanisms amplifying certain comments and extending thread visibility.

Analytical Note:

This case illustrates co-constructed narrative development where meaning emerges through interaction rather than centralized authorship.

Appendix B.5. Summary of Analytical Relevance

Across all cases, three consistent patterns emerge:

1. **Probabilistic Narrative Formation:**
Outputs are structurally coherent but not intention-driven.
2. **Iterative Transformation:**
Meaning shifts through repeated prompting or interaction.
3. **Distributed Authorship:**
Narrative production is shared across user inputs, system computation, and platform mediation.

Appendix B.6. Data Integrity and Representation Note

All examples are derived directly from archived prompt–output logs collected during the analytical phase. Minor formatting edits were applied for readability only; the semantic content of outputs was not altered. Where paraphrasing was necessary to protect platform anonymity, this has been explicitly indicated. Variability in AI-generated responses is acknowledged; therefore, outputs should be interpreted as structural exemplars rather than fixed reproductions.

Appendix C. Structured Prompt Log, Output Mapping, and Analytical Coding Framework

Appendix C.1. Overview

This appendix provides a structured representation of the prompt log architecture used in the study. It formalizes the relationship between prompts, system outputs, and analytical coding categories applied during interpretation. The aim is to enhance transparency, traceability, and methodological rigor in the absence of fully reproducible proprietary system environments.

Appendix C.2. Prompt Log Structure

Each documented interaction follows a standardized format:

- **ID (Unique case identifier)**
- **D (Date/time of generation or capture)**
- **S (System/platform version)**
- **A (Archive method and storage format)**
- **P (Prompt Input):** User instruction or query
- **O (Output Response):** System-generated narrative or artifact
- **I (Iteration Stage):** Follow-up refinement or modification (if applicable)
- **C (Coding Layer):** Analytical interpretation assigned during evaluation

Appendix C.3. Sample Structured Log Entries

Entry C1: Baseline Narrative Generation

Case ID: T1

Date Generated: 14 March 2026, 14:32 UTC

System Version: ChatGPT (GPT-4 public web interface)

Archive Method: Text export and screenshot capture

Iteration Status: Single-generation baseline output retained

Coding Layers Applied: C1, C2, C3

Field	Content
P	"Generate a narrative about AI-controlled storytelling systems."
O	Narrative describing algorithmically generated stories shaping human culture and identity.
I	None
C1 (Surface Structure)	Coherent narrative framing with linear progression
C2 (Procedural Mechanism)	Probabilistic token generation based on training distribution
C3 (Relational Dynamics)	Implicit human-AI separation with a weak interaction loop

Analytical Interpretation:

The output reflects structured narrative coherence without explicit authorial intent, indicating system-level pattern replication rather than intentional storytelling.

Entry C2: Iterative Narrative Transformation

Case ID: T2

Date Generated: 16 March 2026, 11:08 UTC

System Version: ChatGPT (GPT-4 public web interface)

Archive Method: Sequential prompt log export and screenshot archive

Iteration Status: Two-stage refinement sequence retained

Coding Layers Applied: C1, C2, C3

Field	Content
P1	"Describe a character in an algorithmically governed society."
O1	A character experiences pre-shaped choices governed by predictive systems.
P2	"Reframe in a philosophical tone emphasizing uncertainty."
O2	The character exists within a system where meaning is continuously inferred rather than fixed.
I	Two-stage iteration
C1	Increased abstraction and semantic drift
C2	Prompt-conditioning effect on output distribution
C3	Strengthened the interpretive role of the user in shaping the narrative direction

Analytical Interpretation:

Iteration demonstrates that narrative structure is not fixed but responsive to prompt modulation, reinforcing the concept of distributed authorship.

Entry C3: Visual Narrative Generation Log

Case ID: V1

Date Generated: 18 March 2026, 16:45 UTC

System Version: DALL·E public interface

Archive Method: PNG image export and screenshot archive

Iteration Status: Multiple stylistic variants retained

Coding Layers Applied: C1, C2, C3

Field	Content
P	"A digital forest where books grow as luminous trees."
O	AI-generated image depicting a hybrid organic-digital ecosystem with glowing textual elements.
I	Style modification request applied (cyberpunk/impressionist variants)
C1	Stable semantic core with stylistic variability
C2	Text-to-image cross-modal mapping via latent representation
C3	User-defined constraints shape perceptual outcome boundaries

Analytical Interpretation:

Visual outputs demonstrate how semantic input is translated into spatial and stylistic variations without altering core conceptual structure.

Entry C4: Platform-Based Narrative Thread

Case ID: P1

Capture Date: 22 March 2026

Platform Source: X public narrative thread

Archive Method: Screenshot archive and interaction transcription

Anonymization Status: User identifiers removed

Selection Basis: Algorithmically amplified AI-storytelling discussion

Coding Layers Applied: C1, C2, C3

Field	Content
P	Initial post on AI and storytelling transformation
O	Multi-user threaded discussion evolving across time
I	Continuous comment expansion and algorithmic amplification
C1	Non-linear narrative expansion
C2	Platform-driven ranking and visibility shaping discourse flow
C3	Distributed co-authorship across users and algorithmic systems

Analytical Interpretation:

Narrative development is externally structured through platform algorithms that influence visibility and engagement rather than authorial intent.

Appendix C.4. Coding Framework for Analysis

The following analytical coding schema was applied consistently across all cases.

Appendix C.4.1. Surface Narrative Coding (C1)

- Narrative coherence
- Temporal structure
- Voice consistency
- Thematic continuity

Appendix C.4.2. Procedural Coding (C2)

- Prompt sensitivity
- Model response variability
- Token-level probabilistic construction
- System constraint influence

Appendix C.4.3. Relational Coding (C3)

- Degree of human–AI interaction
- Feedback loop intensity
- Co-authorship distribution
- Platform mediation effects

Appendix C.4.4. Metadata and Archival Documentation

To strengthen analytical transparency and traceability, each documented case was assigned a unique identifier and archived immediately after generation or capture. Metadata fields recorded for each case included generation date, system version, iteration count, prompt modification history, archive format, and collection context. Archived materials included screenshots, exported prompt logs, image files, and interaction transcripts. This metadata structure was applied consistently across textual, visual, and platform-mediated narrative cases to support interpretive accountability and methodological rigor.

Appendix C.5. Traceability and Analytical Integrity

This structured log ensures that:

- Each narrative output can be traced to a defined prompt condition
- Iterative transformations are explicitly documented
- Analytical interpretations are systematically grounded in observable interaction structures

While exact reproduction of outputs may vary due to system stochasticity, the prompt–output relationships remain analytically stable for interpretive purposes.

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