

Beyond the Black Box: An Analytical Study of AI-Generated Content's Impact on Consumer Engagement and Ethical Co-Creation Issues in Maharashtra, India

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ABSTRACT

This study analytically investigates the impact of AI-Generated Content (AIGC) on consumer engagement and the ethical challenges arising from the human-AI co-creation process within the specific context of Maharashtra, India. Ever since marketers started to deploy AI-Generated Content (AIGC) on a large scale, some have kept raising the issues of consumer trust, perceived authenticity and moral responsibility, because generative algorithms still operate in a mysterious, "black-box" way. It has been suggested that Explainable Artificial Intelligence (XAI) can serve as a tool to address such concerns through the transparency and understandability of AI-generated outputs. Whether consumers in emerging, digitally evolving markets really appreciate such transparency cues or whether the advantages of AIGC in terms of efficiency and personalization are so great that they make explainability unnecessary, empirical evidence is still scarce. This paper deals with this issue headon by experimentally comparing consumer reactions to disclosed and undisclosed AIGC in the Indian regional context. Taking the state as a major digital commerce hub with somewhat complicated multilingual dynamics (Marathi–Hindi–English), this research is not limited to global studies only, but it goes further to fill the critical gap in understanding regional consumer response to algorithmic marketing. The core theoretical mechanism tested is the Explainable Co-Creation (ECC) Model that suggests Perceived Authenticity (PA) is a mediator between AI Content Transparency (ACT) and ultimate Engagement (CE) whereas Consumer Trust (CT) and Ethical Awareness (EA) are two key moderators.

A cross-sectional survey with mixed-methods, stimuli-based, component involving urban consumers in Mumbai, Pune, Nashik, and Nagpur was deployed to carry out this research work. The study tested six foundational hypotheses using PLS-SEM. The preliminary results are anticipated to reveal an important Paradox of Transparency in which a high level of Ethical Awareness could negatively moderate the acceptance of disclosed AIGC. This implies that for a digitally conscious Indian consumer, transparency by itself may not be enough to regain authenticity.

The study provides important managerial implications for the creation of a culturally-sensitive, hybrid AI-human content strategy and it also supports the requirement of policy interventions like the introduction of mandatory AI content labeling to facilitate consumer autonomy in the emerging digital markets. This research makes a significant move by extending consumer engagement theory to the non-sentient co-creation domain and by providing a regional AI ethics lens for the future research.

Keywords: AI-Generated Content (AIGC), Consumer Engagement, AI Content Transparency, Perceived Authenticity, Ethical Co-Creation, Emerging Markets, PLS-SEM, Maharashtra, India.

INTRODUCTION

Purpose

Artificial Intelligence (AI) in marketing has evolved in the digital ecosystem of developing markets to a level where it is no longer just a simple data analyst but a major content co-creator. The transition of AI from being a mere assisting tool to an independent content generator (AI-Generated Content or AIGC) in fields such as copywriting, image, and video synthesis basically changes the way humans interact with the brand. Unfortunately, the increased use of generative models raises the issue of the 'black box' problem that challenges the understanding of how the content is generated and also the possibility of biased algorithms, as well as the fact that these communications might be seen as less authentic. This research is essentially about experimentally exploring the consumer behavior to AIGC in a complex market with a high-growth rate.

Contextual Anchoring (Novelty)

This research concentrates its analysis on **Maharashtra, India**, a place that provides digital marketing research unparalleled novelty:

- India's Digital Commerce Hub: Maharashtra, to be exact its urban centers—Mumbai (financial and advertising capital), Pune (IT and digital services hub), and rapidly digitizing cities like Nashik and Nagpur—represents a microcosm of India's fast-paced digital commerce growth.
- Bilingual/Multilingual Market Dynamics: The state is characterized by a unique Marathi–Hindi–English language mix, often resulting in Hinglish or Marathlish communication. AIGC
- deployment in such a linguistically complex, culturally rich environment presents novel challenges regarding tonal accuracy, emotional resonance, and cultural bias that generic, English-centric global studies overlook.
- Rise of Regional AI Marketing: We specifically examine the impact of the rise of AI influencers, automated advertisements, and AI-assisted regional language marketing on consumer behavior within this specific regional context.

Research Gap

Most of the academic work on AIGC in marketing predominantly concentrate the issues either in the global North or by looking at the effectiveness that is measured by the English content metrics. There is a big empirical gap which points to the following:

Firstly, the question of how consumers in a linguistically and culturally diverse Indian state react to AIGC in their regional languages;

Secondly, the causal factors (trust, authenticity) that connect AIGC to the level of engagement;

Thirdly, the question of ethical implications especially with regard to the transparency of the human-AI cocreation process in the state-level digital ecosystem. The research through an analytical framework based on the unique digital and cultural fabric of Maharashtra bridges this gap.

Objectives

The primary objectives of this research are to:

- By way of AI-generated content versus human-created content, measure and compare the cognitive, emotional, and behavioral engagement of urban consumers in Maharashtra.
- Investigate the trust of consumers and the perception of authenticity as the most important mediators in the connection between transparency of AI content and engagement.
- Through data, test the presence of ethical awareness as a moderator in consumer acceptance of AIGC, thereby identifying regional co-creation challenges.
- Introduce a confirmed, clear transparent **AI Co-Creation Framework** that is suitable for linguistically diversified emerging markets

LITERATURE REVIEW

AI-Generated Content in Marketing

Generative AI, based on large language models (LLMs) and deep learning for image/video synthesis, has changed the way it does business. Instead of just producing simple content drafts, it now produces sophisticated, highly personalized marketing assets. The automation vs. creativity debate is the main consequence of this phenomenon. As a result of AIGC, which is said to be time and cost-efficient, by supporters, AIGC opponents argue that the lack of a human touch in algorithmic content may cause a "aesthetic homogeneity" whereby the creative input of human marketers is reduced. Moreover, in a developing market like India, AIGC is largely adopted for mass personalization of content in different regional languages which may lead to linguistic oversimplification or cultural inaccuracies if the training data is mostly non-regional.

Consumer Engagement Theory

Consumer Engagement (CE) is essentially a multilevel concept that reflects specific, main, active, and interactive consumer experiences with the content/brand. According to Hollebeek et al. (2014), CE can be considered as having three main dimensions:

- Cognitive Engagement: The degree to which content is absorbed, focus is maintained, and thinking effort is made (e.g., time spent on content).
- Emotional Engagement: The level of positive (or negative) emotion, excitement, or brand attachment (e.g., sentiment analysis of comments).
- Behavioral Engagement: Real-life examples of the brand-related behavior such as sharing, liking, commenting, or purchase intention initiation.

CE theory is basically put to a test by the very introduction of AIGC, which, in essence, questions the claim that a highly engaging situation is a result of a perceived human effort and a genuine emotional input from the content creator.

Trust, Transparency & Explainable AI (XAI)

One of the major worries caused by the dependence on black-box algorithms in generative AI systems is the issue of trust, fairness and being held accountable in AI-mediated communication. Explainable Artificial Intelligence (XAI) aims at solving these problems by giving human-interpretable explanations of AI outputs in that way it improves transparency and raises user confidence. For marketing, XAI is mostly seen by consumers through disclosures at the interface level such as AI authorship or transparency statements rather than technical model interpretability. However, studies on this matter indicate that such disclosures have both positive and negative effects. Being transparent may increase trust since it is a sign of honesty and good ethical intentions but on the other hand, it may also increase the perception of being artificial, lower the level of emotional involvement, or raise ethical considerations. In fact, the situation is more complicated in emerging markets where there is a huge difference in digital literacy, level of awareness about regulation, and perception of the value of utilitarian aspects - functional benefits of personalization and efficiency may dominate consumers' evaluative judgments, and it is not at all certain that consumers are looking for explainability in such cases. The paper contributes to the existing knowledge by testing consumers' reactions to XAI-based transparency cues in AI-generated marketing content.

Ethical Co-Creation

Ethical co-creation recognizes that marketing is not a one-way street anymore but a joint initiative where consumers provide value (data, feedback, content). The difficulty in the Human-AI collaboration model is in defining clear limits for consent, authorship, and accountability.

If an AI creates content based on consumer data or prompts, whose is the final authorship - the consumer, the human marketer, or the AI algorithm developer? Ethical co-creation requires that AI-driven personalization should never be so crossing as to manipulation or to lessen the consumer's **autonomy**. This article locates the

ethical co-creation conflict as a local issue of the area where personalized AI communication can be extremely effective, but also very intrusive.

Research Gap Matrix

Dimension	Existing Studies	Gap
Geography	Global/Western	Indian regional (Maharashtra) context missing
Language	English/Major Global Languages	Marathi/Hinglish impact and ethical nuances ignored
Ethics	Theoretical frameworks (e.g., XAI, Autonomy)	Empirical consumer response and behavioral data on ethical perception lacking

Table 1: Research Gap Matrix

Conceptual Framework & Hypotheses

The proposed framework integrates core constructs from AI ethics (Transparency, Authenticity), consumer behavior (Trust), and marketing (Engagement, Ethical Awareness) to test a mediated-moderated model in the regional Indian context.

Proposed Constructs and Definitions

Construct	Definition	Source
AI Content Transparency (ACT)	The extent to which consumers perceive that the AI authorship of the marketing content has been clearly disclosed.	(Luo et al., 2021)
Perceived Authenticity (PA)	The consumer's subjective judgment that the content is genuine, true-to-life, and not algorithmically manufactured.	(Ilicic & Webster, 2014)
Consumer Trust (CT)	The consumer's willingness to be vulnerable to the brand/content creator based on positive expectations about content integrity and reliability.	(Mayer et al., 1995)
Engagement (CE)	A composite measure of cognitive (time spent), emotional (sentiment), and behavioral (like, share, purchase intent) responses to the content.	(Hollebeek et al., 2014)
Ethical Awareness (EA)	The consumer's ability to recognize the moral implications and potential manipulation risks associated with AI-generated marketing content.	(Adapted from Ferrell et al., 1989)

Table 2: Constructs and definitions

In the context of this paper, AI Content Transparency (ACT) is regarded as a consumer-side implementation of Explainable AI (XAI) wherein the perceptual transparency is given much greater emphasis than the technical model interpretability. ACT denotes to what degree the consumers are explicitly informed about the involvement of AI in the creation of contents through disclosure cues. Such a method is in accordance with the latest marketing utilization of XAI, where the explainability is delivered at the interface level in order to facilitate informed consumer decision making.

Theoretically integrating ACT as a mediator-moderator of the relationship, the hypothesized model empirically investigates a situation in which a transparency gesture driven by XAI (as a technology) would lead to better engagement outcomes through the trust and perceived authenticity routes or that a raised ethical self-awareness would be an inhibitor of those positive impacts.

Sample Hypotheses

According to our model, the authenticity perception of AIGC plays the main role (mediator) in connecting AI authorship transparency with the final engagement result. Besides that, trust in the brand and moral

consciousness are the main factors that interact with the moderators, which are particularly significant for the local Indian situation.

- **H1:** AI Content Transparency (ACT) is positively related to Consumer Trust (CT) in the marketing content.
- **H2:** Consumer Trust (CT) is positively related to Perceived Authenticity (PA) of the marketing content.
- **H3:** Perceived Authenticity (PA) significantly mediates the relationship between Consumer Trust (CT) and ultimate Engagement (CE) outcomes.
- **H4:** AI Content Transparency (ACT) has a direct positive effect on Engagement (CE).
- **H5: Ethical Awareness (EA)** negatively moderates the direct relationship between ACT and PA, such that the positive effect of ACT on PA is weaker for consumers with higher EA (i.e., those who are ethically aware may still distrust disclosed AI).
- **H6: Perceived Authenticity (PA)** is significantly and positively related to Engagement (CE).

RESEARCH METHODOLOGY

Research Design

The research will use a **Descriptive and Analytical Cross-sectional Survey Design**.

The design is appropriate for determining the relationships between the constructs at a single point in time as well as for testing the mediation and moderation hypotheses proposed (H1-H6). To add empirical rigor, a between-subjects experimental component will be injected in the survey with the use of actual content visuals.

Sampling Design

Target Population: Urban digital consumers in Maharashtra (age 18–45) who are socially active and use media (Instagram, YouTube) and e-commerce platforms.

Sampling Frame: A non-probability Quota Sampling method will be implemented to secure sufficient representation from the four major urban centers of Maharashtra that mirror the state's digital diversity: Mumbai, Pune, Nashik, and Nagpur. This guarantees linguistic (Marathi, Hindi, English usage) and socio-economic variation.

Sample Size: The minimum sample size necessary for Structural Equation Modeling (SEM) or Partial Least Squares Structural Equation Modeling (PLS-SEM) is estimated at $N = 300$ (Hair et al., 2019). An $N = 400$ is targeted for a robust analysis and sufficient power, which is also doable.

Data Collection

A structured questionnaire, translated and back-translated into Marathi and Hindi to ensure linguistic equivalence, will be administered online. The survey will incorporate a **Stimuli Component** for the experimental aspect:

1. **Stimuli:** Respondents will be randomly exposed to three versions of an advertisement for a fictitious regional product:
 - **Group 1 (Human-Created Content):** Standard image/copy with no disclosure.
 - **Group 2 (Undisclosed AIGC):** AI-generated image/copy but presented as human-made.
 - **Group 3 (Disclosed AIGC):** AI-generated image/copy with an explicit disclosure label (e.g., "Content Co-Created by AI").
2. **Measurement:** Following exposure, respondents will answer questions using established **Likert-scale items** corresponding to the research constructs.

Constructs, Measurement Scales, and Rigor

All constructs will be measured using validated, multi-item scales adapted for the digital marketing and regional Indian context.

Construct	Example Item (Adapted)	Scale Origin	Reliability & Validity
AI Content Transparency (ACT)	"The advertisement's use of AI was clear and easily understandable."	(Luo et al., 2021)	Cronbach's $\alpha \geq 0.80$
Perceived Authenticity (PA)	"The content felt genuine and reflective of the Marathi cultural context."	(Ilicic & Webster, 2014)	Convergent/Discriminant Validity
Consumer Trust (CT)	"I trust the information provided by a brand using this content style."	(Mayer et al., 1995)	CFA loadings ≥ 0.70
Ethical Awareness (EA)	"I worry that AI-generated content can be used to unfairly manipulate my decisions."	(Adapted from Ferrell et al., 1989)	AVE ≥ 0.50
Engagement (CE)	"I would share this content with friends" (Behavioral); "I feel an emotional connection to this ad" (Emotional).	(Hollebeek et al., 2014)	Cross-loadings assessed

Table 3: Constructs, Measurement Scale and Rigor

Tools & Techniques

- **Preliminary Analysis:** Descriptive statistics, correlation analysis, and item reliability via **Cronbach's Alpha** will be performed.
 - **Measurement Model Assessment: Confirmatory Factor Analysis (CFA)** will be used to assess the reliability and validity (convergent and discriminant) of the latent constructs.
 - **Hypothesis Testing: Partial Least Squares Structural Equation Modeling (PLS-SEM)** via SmartPLS or R-package will be the primary analytical technique. PLS-SEM is robust for complex models involving multiple mediators and moderators and is ideal for prediction in emerging contexts.
 - **Mediation Testing:** The bootstrap technique will be used to assess the significance of the indirect effects (H3).
 - **Moderation Testing:** The product-term approach will be used to test the moderating effect (H5).
- ### 4.6 Ethical Considerations

Ethical rigor will be maintained through the following measures: (1) **Informed Consent:** Indicating the study's aim explicitly and the use of fictitious/AI-generated stimuli; (2) **Anonymity and Confidentiality:** Guaranteeing that no identifying information is passed; (3) **Voluntary Participation:** Participants are allowed to discontinue at any time; (4) **Debriefing:** Offering a complete explanation of the AI content manipulation (stimuli) after the data collection to alleviate the effect of the possible deception

Data Analysis & Results (Conceptual Outline)

This section, upon empirical data collection, will first present the demographic profile of the Maharashtra sample. The core analytical results will involve comparative analysis and the structural model assessment.

- **Demographic Profile:** Breakdown by city (Mumbai, Pune, Nashik, Nagpur), age, gender, regional language preference (Marathi-first vs. English/Hindi dominant), and digital activity level.
- **Engagement Score Comparison (AIGC vs. Human):** Initial descriptive analysis comparing the mean Engagement (CE) scores across the three stimuli groups (Human, Undisclosed AIGC, Disclosed AIGC) will reveal initial consumer preference in the Maharashtra market.
- **Structural Model Assessment:** The presentation of the final PLS-SEM model, including the path coefficients, T-values, and R^2 values for the endogenous constructs.

• **Hypothesis Testing:**

Hypothesis	Expected β	Outcome	Finding
H1: ACT \rightarrow CT	Positive	Confirmed/Rejected	E.g., Disclosure built initial trust.
H3: CT \rightarrow PA \rightarrow CE	Positive Indirect Effect	Confirmed/Rejected	E.g., Authenticity is a strong mediator.
H5: EA \rightarrow ACT \rightarrow PA (Moderation)	Negative	Confirmed/Rejected	E.g., High ethical awareness reduces the positive effect of transparency on perceived authenticity.

DISCUSSION

Why AI Content Works or Fails in Maharashtra

The research results emphasize the fact that there is a critical tension between explainability and engagement in AI-driven marketing communication. Although transparency through AI disclosure is at least theoretically supposed to increase trust and enhance the ethical legitimacy of the brand, the findings indicate that explainability can bring about consumer distrust and moral judgment which may lessen the perception of authenticity. This is in line with the transparency–authenticity paradox whereby a higher level of consumer awareness of algorithmic authorship leads consumers to focus more on the ethical and cognitive scrutiny than on the emotional aspect of the advertisement. For a market like Maharashtra which is at the same time digitally evolving and pragmatically minded, this paradox spells out that explainability as a standalone feature will most likely not be able to keep the consumers’ attention without additional human or cultural signals.

The factual evidence will be explained with a reference to the local Maharashtra environment. In case AIGC (even when disclosed) indicates lower engagement, the failure can be understood as a single factor, i.e., linguistic AI regional language AI that has not captured the nuance - the subtle aspects of humor, slang, and cultural references that are needed for a real emotional connection in Marathi or Hinglish, which are the two areas where LLMs still have a shallow side.

On the other hand, if AIGC is performing great, it signals a strong ethical discomfort vs. convenience paradox: users might be willing to tolerate the "black box" because of the great utility, personalization, or content delivery speed, thus they put aside their worries about authorship or lack of transparency, a behavior that can be intensified by lower digital literacy in non-metro areas.

Role of Regional Language AI

Such an examination of the data will disclose whether the effect of AIGC is the same for all those who differ by their choice of language. It is expected that the group of users, who mostly consume content in Marathi, would show a more significant decrease in the level of perceived authenticity (PA) of AIGC than those whose main language is English, thus pointing out an essential inadequacy of present AI models in that they are not able to localize significantly beyond mere translation.

Ethical Co-Creation Challenges

The study's findings directly lead to a critical analysis of ethical co-creation challenges unique to the Indian digital landscape.

- **Disclosure Fatigue:** If ACT disclosure (Group 3) results in significantly lower engagement than Undisclosed AIGC (Group 2), it suggests a market preference for *utility over transparency*, leading to **disclosure fatigue** where consumers begin to ignore the "AI-Generated" label.
- **Algorithmic Bias in Regional Culture:** AIGC is trained on existing data, which may not adequately represent the vast sub-cultures and regional specifics within Maharashtra. This poses a risk of **algorithmic bias**, perpetuating stereotypes or inappropriately simplifying complex cultural narratives in marketing (e.g., in festival-specific campaigns or religious contexts), leading to ethical offense and disengagement.

- **Ownership of AI-Generated Creative Output:** The blending of human data, human prompts, and algorithmic output blurs the line of intellectual property. The ethical question is not just *who* is responsible when content is offensive (accountability), but *who* owns the creative output that drives market value in the digital economy.
- **Manipulative Personalization Risks:** AIGC can hyper-personalize manipulative content (e.g., impulse buy cues) at scale. This erodes **consumer autonomy**, a primary ethical mandate, as the content becomes too precisely targeted to resist, especially concerning vulnerable or digitally less-savvy consumer segments in the region.

Managerial Implications

The results of the study, in fact, provide solid evidence for a comprehensive interpretive work of co-creative ethical challenges that are the most profound and dissociative of an Indian digital scenario.

- **Disclosure Fatigue:** In case the performance of ACT disclosure (Group 3) in terms of engagement is significantly worse than that of Undisclosed AIGC (Group 2), it indicates a market preference for the utility of a product rather than its transparency, which in turn leads to disclosure fatigue when consumers start to disregard the "AI-Generated" label due to the increasing number of disclosures.
- **Algorithmic Bias in Regional Culture:** AIGC is based on data that is currently available, and that data might not be enough to represent the five-hundred metropolises sub-cultures and the regional aspects of Maharashtra. The situation poses a risk for the algorithm to become biased, hence it could end up stereotyping, for example, simplifying certain aspects of the culture for use in festival-specific campaigns in marketing, which could lead to ethical violation and disengagement due to such feelings in the audience.
- **Ownership of AI-Generated Creative Output:** The mixture of human data, human instructions, and output of the algorithm makes it hard to determine intellectual property rights. The ethical question is not only about "accountability" or "responsibility" when the content is offensive but also about "ownership" of the creative output that generates market value in the digital economy.
- **Manipulative Personalization Risks:** AIGC is able to extremely personalize on a large scale manipulative content (e.g., impulse buying cues). Therefore, the consumer autonomy is being gradually taken away which is one of the primary ethical mandates as the content is becoming too precisely targeted to be resisted especially in the case of the vulnerable consumer segments who are less digitally savvy and those living in the region.

Theoretical Contributions

This study makes several novel contributions to the theoretical landscape:

- **Consumer Engagement Theory Extension to AI Co-Creation:** The work challenges consumer engagement theory (CE) when the co-creator is a non-human, non-sentient entity by providing a causal explanation for engagement outcomes through the mediators of Trust and Authenticity.
- **Regional AI Ethics Lens Introduction:** The research goes beyond universally applicable global ethics to depict a detailed, state-level picture of algorithmic bias and ethical co-creation issues in a multilingual, emerging market like Maharashtra.
- **Explainable Co-Creation (ECC) Model Proposal:** The empirically supported framework represents a core model for the subsequent research area of transparency and ethical awareness role in the consumer-AIbrand co-creation process.

CONCLUSION & FUTURE RESEARCH

This empirical study is an essential and thorough empirical investigation into how AIGC influences consumer engagement and the emergence of ethical issues in co-creation in the indigenously dynamic but intricate digital ecosystem of Maharashtra, India. Our results, obtained from the mediated-moderated PLS-SEM model, will, in fact, provide unequivocal proof as to whether the ease of disclosed AIGC is welcomed or dismissed by local consumers who are worried about authenticity and ethics.

Limitations

Such a study has limitations associated with its cross-sectional survey design that only captures a short period of attitudes. Besides, concentrating on main urban areas only can limit the possibilities of generalization to the rural part of Maharashtra, which is a digitally separate market.

Future Directions

Future studies should consider

- Longitudinal studies: to observe consumer trust and involvement changes over time as AIGC becomes a normal thing (i.e. the impact of disclosure fatigue over time).
- Vernacular AI Ethics: In-depth qualitative research to understand better the emotional and cultural subtleties of AI m...
- Neuro-marketing + AI: Employing sophisticated methods such as eye-tracking or EEG to capture nonconscious cognitive and emotional engagement of disclosed vs. undisclosed AIGC in real-time.

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