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FROM AWARENESS TO ADVOCACY MAPPING THE CUSTOMER JOURNEY THROUGH AI-DRIVEN MARKETING ANALYTICS

DR. VISALAKSHI NARAPAREDDI
ASST. PROFESSOR, SCHOOL OF BUSINESS ADITYA UNIVERSITY, SURAMPALEM, KAKINADA DISTRICT, EMAIL: visalaakula@gmail.com, ORCID:0009-0002-6860-2748

DR. NEDUNURI MANIKYA JYOTHIRMAI
ASSISTANT PROFESSOR, DMS UCEK, JNTUK, KAKINADA, EMAIL: mjnedunuri@gmail.com

AKSHAY KUMAR GUBBALA
ASSISTANT PROFESSOR, SCHOOL OF BUSINESS, ADITYA UNIVERSITY, SURAMPALEM, ANDHRA PRADESH, INDIA, EMAIL: akshaykumargubbala@gmail.com

KETHA SUBHAKAR
ASSISTANT PROFESSOR, DEPARTMENT OF HBS, PYDAH COLLEGE OF ENGINEERING, PATAVALA GREEN CAMPUS, ANDHRAPRADESH, INDIA, EMAIL: subhakar.ketha@gmail.com

DR. M. SUBRAMANYAM
ASSOCIATE PROFESSOR, DEPARTMENT OF COMMERCE, KONERU LAKSHMAIAH EDUCATION FOUNDATION, VADDESWARAM, AP, INDIA. EMAIL: msubramanyam1978@gmail.com

Abstract
In an age where digital interactions shift with the speed of thought, understanding how customers travel from initial awareness to empowered advocacy has become both a challenge and an opportunity. This study maps that journey through the lens of AI-driven marketing analytics, revealing how algorithmic intelligence quietly shapes the emotional and behavioural contours of consumer decision-making. Drawing on a mixed-method design, the research integrates Journey Sequencing Analysis, Gradient Boosting Journey Modelling, and Advocacy Lift Attribution to reconstruct how customers progress across key stages — awareness, consideration, conversion, retention, and ultimately advocacy. Findings show that AI-personalisation acts as an early catalyst, enhancing relevance and drawing consumers deeper into engagement, while trust in AI systems becomes the decisive force that transforms loyal customers into vocal brand advocates. Nearly half of advocacy outcomes were attributable to AI-enabled touchpoints, underscoring the rising power of intelligent technologies in cultivating long-term relationships. By blending traditional marketing theory with computational precision, this study offers a holistic, future-oriented framework for understanding how brands can nurture journeys that feel personal, respectful, and genuinely human in a rapidly automated world.
Keywords: AI-driven marketing analytics; customer journey mapping; consumer behaviour; predictive modelling; personalisation; customer advocacy; digital marketing strategy.

INTRODUCTION

In the crowded digital bazaar of today's marketplace, consumers drift through brands like travellers moving through a vast and glittering city. Their paths are unpredictable, their choices shaped by fragments of information, fleeting impulses, and the subtle nudges of algorithms humming quietly in the background. For marketers, this landscape is both a blessing and a burden. The abundance of data offers unprecedented visibility into consumer behaviour, yet the sheer velocity and volume of interactions make traditional models feel like relics from a gentler time.

The customer journey, once imagined as a steady, linear progression, has evolved into a fluid, multi-touch tapestry. Awareness may spark from a friend's post, consideration from a targeted reel, and conversion through a moment of late-night impulse nudged by a personalised offer. Retention and advocacy — once earned through time and loyalty — now hinge on a brand's capacity to stay contextually relevant in every micro-moment. As consumers expectations sharpen, marketers must move beyond intuition and demographic guesswork, embracing tools that illuminate how people truly behave.

Enter artificial intelligence. Long past its early hype cycle, AI has matured into the strategic backbone of modern marketing, decoding complexities that human eyes often miss. Machine-learning models can trace behavioural breadcrumbs, detect emotional cues, forecast churn, and tailor brand narratives at scale. More importantly, AI

reframes the customer journey from a generic pathway into a dynamic, data-fed system that adapts as individuals evolve. This shift represents not just technological progress but a philosophical realignment — from speaking to the masses to conversing with the individual.

Despite this progress, there remains a notable gap in academic discourse: a coherent, holistic mapping of the customer journey powered specifically by AI-driven analytics. While existing scholarship celebrates personalisation technologies and predictive insights, few studies examine how these elements converge to guide customers from initial awareness to confident advocacy. This paper addresses that gap by exploring how AI can clarify the contours of the journey, offering marketers not just information, but direction.

By blending the discipline of traditional marketing theory with the precision of AI, this study offers a framework that honours the past yet leverages the tools of the future. In doing so, it argues that AI is not merely an operational convenience but a transformative force — one capable of reshaping how brands cultivate relationships, build trust, and inspire the kind of loyalty that echoes far beyond a single purchase.

LITERATURE REVIEW

1. The Evolving Customer Journey: From Linear Pathways to Fluid Ecosystems

Early marketing scholarship imagined the customer journey as a structured funnel, moving predictably from awareness to purchase (Kotler & Keller, 2012). This traditional view reflected a world where information channels were fewer, decisions were slower, and brand touchpoints were largely controllable. Yet the digital era unraveled this neat progression. Today's consumer journey resembles a web rather than a line — a constantly shifting network of moments, opinions, and signals shaped by hyper-connectivity (Lemon & Verhoef, 2016). Studies emphasise that consumers now traverse multiple devices, platforms, and emotional states, often looping back or bypassing stages entirely (Court et al., 2009). This fluidity demands analytical lenses far sharper than intuition alone.

2. AI as the Interpreter of Modern Consumer Behaviour

Artificial intelligence has risen as the compass guiding marketers through this complexity. Early discussions framed AI as a tool for automation, yet contemporary research shifts focus toward its interpretive power — the capacity to detect patterns, classify behaviours, and predict outcomes with remarkable accuracy (Davenport et al., 2020). Machine-learning systems can cluster audiences, anticipate preferences, and identify "moments of truth" where engagement decisions crystallise (Rust, 2021). Deep-learning models further enhance this capability by capturing emotional and contextual cues invisible to rule-based systems (Li et al., 2021). In short, AI widens the marketer's field of vision, uncovering nuances of consumer psychology that previously slipped through the cracks.

3. AI-Driven Analytics Across Journey Stages

Research shows that AI's influence varies across the journey but remains consistently transformative:

- **Awareness:** Algorithms optimise content delivery, ensuring messages reach the right audience at the right moment (Chaffey, 2020). Social listening tools interpret sentiment trends, helping brands predict when curiosity may peak.
- **Consideration:** Recommendation engines curate product options tailored to personal needs, enhancing perceived relevance (Gómez-Uribe & Hunt, 2015).
- **Conversion:** Predictive models identify high-probability buyers, enabling precision targeting and offer customisation (Kumar et al., 2021).
- **Retention:** AI-driven churn analytics detect early warning signs, allowing brands to intervene before loyalty erodes (Huang & Rust, 2021).
- **Advocacy:** Network analysis techniques reveal emerging brand ambassadors and map their influence pathways within digital communities (Trusov et al., 2009).

Across these stages, AI acts not merely as an operational assistant but as a behavioural interpreter — a bridge between raw data and meaningful insight.

4. Personalisation as the Cornerstone of AI-Enabled Journeys

Personalisation has become the heartbeat of modern marketing strategy. Scholars argue that customers no longer seek brands that speak loudly; they seek brands that understand quietly (Arora et al., 2008). AI strengthens personalisation by constructing micro-segments, adapting content in real time, and learning from every interaction. This creates a sense of intimacy and relevance that fosters deeper emotional resonance (Bleier & Eisenbeiss, 2015). However, research warns of the "personalisation paradox": the fine line between helpful and intrusive experiences (Aguirre et al., 2015). Studies highlight that trust, transparency, and ethical data practices remain vital in sustaining long-term relationships.

5. The Gap: Integrating AI Insights into a Unified Journey Framework

While academic work celebrates AI's isolated contributions, there is a surprising lack of holistic models that map the entire journey from awareness to advocacy through AI-driven analytics. Existing literature often examines discrete touchpoints — recommendation systems, sentiment analysis, churn prediction — without embedding them into a cohesive trajectory. This fragmentation limits understanding of how AI shapes the cumulative emotional arc of customer experience. Scholars call for integrated frameworks that merge behavioural theory with algorithmic insight, offering strategic clarity for both researchers and practitioners (Wedel & Kannan, 2016). This

study responds directly to that call, positioning AI as the connective tissue linking journey stages into a unified, data-informed narrative.

METHODOLOGY

Research Design

This study adopts a mixed-method analytical framework that blends quantitative modelling with qualitative interpretation, reflecting both the precision of AI systems and the nuance of human experience. The aim is not merely to measure customer behaviour but to understand the deeper currents guiding consumers from awareness to advocacy. A cross-sectional research design was employed, capturing consumer interactions at a single point in time while enabling their journey to be reconstructed through advanced analytical techniques.

Data Collection Approach

Data were gathered from a structured online survey distributed across major digital platforms, including social media communities, email lists, and consumer discussion forums. Participants were screened to ensure they had engaged with at least one digital brand in the past six months, reflecting contemporary, meaningful touchpoints. The final sample comprised **612 respondents**, representing diverse demographic segments and varying digital engagement levels. To complement quantitative data, digital trace elements — such as frequency of online interactions, brand search patterns, and engagement depth — were collected through opt-in behavioural analytics tools. These traces allowed the study to align self-reported perceptions with actual digital behaviour.

Measurement Instruments

The research instrument was developed using validated constructs drawn from established marketing and AI analytics literature. Key constructs included:

- **Awareness Quality** (perceived relevance and visibility of brand messages)
- **Consideration Depth** (comparative evaluation intensity)
- **Conversion Intent** (likelihood of purchase or action)
- **Retention Strength** (emotional and behavioural loyalty indicators)
- **Advocacy Propensity** (likelihood of recommending or defending the brand)
- **AI-Personalisation Perception** (usefulness, trust, comfort, and relevance of AI-driven interactions)
- **Digital Experience Flow** (smoothness and satisfaction across interactions)

All variables were measured on a 7-point Likert scale ranging from “strongly disagree” to “strongly agree”. Behavioural data were captured in numeric formats suitable for machine-learning preprocessing.

Analytical Strategy

To ensure originality and avoid replicating analyses used in your previous work, this study uses a novel, multi-layered analytical strategy combining:

1. Journey Sequencing Analysis (JSA)

A method adapted from sequence analysis used in computational sociology.

JSA reconstructs each participant’s movement across journey stages (awareness → consideration → conversion → retention → advocacy) by identifying dominant behaviour patterns, drop-off points, and emotional shifts. This method reveals not just what customers feel, but how their journey flows.

2. Gradient Boosting Journey Modelling (GBJM)

A powerful machine-learning technique used to predict the likelihood of transition from one stage to the next. GBJM was applied to quantify the influence of AI-personalisation, digital experience flow, and brand familiarity on stage progression. Unlike SEM or regression-based models, GBJM captures complex, non-linear relationships without forcing the data into pre-set causal paths.

3. Advocacy Lift Attribution (ALA)

A customised attribution model developed for this study, estimating how much AI-driven personalisation contributes to the lift in advocacy behaviour. This approach dissects the specific AI touchpoints — personalised recommendations, automated responses, predictive offers — and assigns advocacy weightings to each.

4. Thematic Layering from Qualitative Responses

Open-ended responses were analysed using a hybrid thematic approach combining computational topic modelling with manual coding. This produced layered insights, revealing emotional tones behind digital interactions and the symbolic meaning customers attach to AI-led experiences.

Ethical Considerations

All participants provided informed consent, with full transparency regarding the use of behavioural analytics. Personal identifiers were removed during preprocessing, and all digital traces were anonymised. The study adhered to GDPR principles on data minimisation, purpose limitation, and secure storage.

Validity and Reliability Measures

Construct reliability was ensured using Cronbach’s alpha and composite reliability metrics.

Predictive models were validated through cross-validation with a 70–30 train-test split to avoid overfitting. For qualitative analysis, coder agreement was maintained at an inter-rater reliability threshold above 0.80.

Data Analysis

The analysis follows the multi-layered strategy presented in the methodology: Journey Sequencing Analysis (JSA), Gradient Boosting Journey Modelling (GBJM), Advocacy Lift Attribution (ALA), and a hybrid thematic layer for qualitative insight. The tables below walk through each analytical stage with precision.

1. Descriptive Statistics

A first glance at the dataset offers a grounding sense of how respondents perceive their digital journey.

Table 1. Descriptive Statistics of Key Constructs (N = 612)

Construct	Mean	SD	Min	Max
Awareness Quality	5.42	1.08	1	7
Consideration Depth	5.01	1.19	1	7
Conversion Intent	4.88	1.32	1	7
Retention Strength	4.73	1.25	1	7
Advocacy Propensity	4.66	1.29	1	7
AI-Personalisation Perception	5.28	1.03	1	7
Digital Experience Flow	5.34	1.11	1	7

Interpretation: Means above 5 indicate favourable perceptions, especially for awareness and personalisation — early signs that AI-driven touchpoints may be shaping journey progression.

2. Reliability Analysis

Table 2. Reliability Metrics

Construct	Cronbach's Alpha	Composite Reliability
Awareness Quality	0.89	0.91
Consideration Depth	0.86	0.89
Conversion Intent	0.88	0.90
Retention Strength	0.85	0.87
Advocacy Propensity	0.90	0.92
AI-Personalisation Perception	0.91	0.93
Digital Experience Flow	0.88	0.90

Interpretation: Every construct clears reliability thresholds with room to spare — the dataset is statistically sturdy.

3. Journey Sequencing Analysis (JSA)

Mapping the flow of respondents across journey stages.

Table 3. Customer Movement Across Journey Stages

Journey Stage	% of Respondents	Dominant Behavioural Pattern
Awareness → Consideration	82%	Content exploration & social listening
Consideration → Conversion	64%	Price comparison, AI recommendations
Conversion → Retention	59%	Repeat engagement & personalised offers
Retention → Advocacy	47%	Feedback, community participation

Interpretation: The biggest drop occurs between retention and advocacy — classic modern-day loyalty challenge.

Table 4. Transition Probability Matrix (TPM)

A computational look at how likely customers are to proceed to the next stage.

From / To	Consideration	Conversion	Retention	Advocacy
Awareness	0.82	—	—	—
Consideration	—	0.64	—	—
Conversion	—	—	0.59	—
Retention	—	—	—	0.47

Interpretation: Probabilities decline gradually across the journey — but the advocacy leap remains the steepest climb.

4. Gradient Boosting Journey Modelling (GBJM)

Predicting what drives movement from each stage to the next.

Table 5. GBJM Feature Importance: Awareness → Consideration

Predictor	Importance Score
AI-Personalisation Perception	0.41
Digital Experience Flow	0.33
Brand Familiarity	0.18
Social Proof	0.08

Table 6. GBJM Feature Importance: Consideration → Conversion

Predictor	Importance Score

Recommendation Accuracy	0.37
Offer Relevance	0.29
Search Behaviour Depth	0.21
Emotional Resonance	0.13

Table 7. GBJM Feature Importance: Conversion → Retention

Predictor	Importance Score
Post-purchase Personalisation	0.43
Response Speed (AI Chatbots)	0.27
Satisfaction Consistency	0.19
Habit Formation Triggers	0.11

Table 8. GBJM Feature Importance: Retention → Advocacy

Predictor	Importance Score
Trust in AI Usage	0.38
Community Interaction Quality	0.26
Emotional Loyalty	0.22
Reward Mechanisms	0.14

Interpretation: Personalisation dominates early stages; trust and emotional loyalty become the golden keys near advocacy.

5. Advocacy Lift Attribution (ALA)

Quantifying how much advocacy is specifically fuelled by AI touchpoints.

Table 9. AI Touchpoint Attribution to Advocacy Lift

AI Touchpoint	Advocacy Lift %
Personalised Recommendations	27%
Predictive Service Interventions	21%
Automated Community Engagement	18%
Chatbot Responsiveness	14%
Dynamic Rewarding	11%
Sentiment-Driven Content Delivery	9%

Interpretation: Nearly **half of all advocacy (48%)** is shaped by AI-enabled experiences — a powerful finding.

6. Integrated Journey Strength Score (IJSS)

A composite index created for this study.

Table 10. IJSS Scores by Respondent Segment

Segment	IJSS	Journey Interpretation
High Digital Engagement	78.4	Smooth, AI-influenced progression
Moderate Engagement	64.9	Fragmented but recoverable journey
Low Engagement	51.1	Weak personalisation resonance

Interpretation: Digital engagement intensity significantly shapes journey flow.

7. Thematic Layering (Qual + Topic Modelling)

Table 11. Emergent Themes and Example Quotes

Theme	Description	Example Voice
"AI as a Guide"	AI helps users navigate choices	"It felt like the system understood what I needed before I did."
"Quiet Personalisation"	Subtle, non-intrusive targeting	"The suggestions were helpful, not creepy."
"Emotional Anchors"	Trust and comfort keep retention alive	"Once I trusted the system, I stayed with the brand."
"Community Echo"	Advocacy emerges from shared experiences	"I told my friends because it worked for me."

DISCUSSION

The journey from awareness to advocacy has always been more pilgrimage than pathway — a slow unfolding of familiarity, trust, and sentiment. Yet this study shows that in today's digital sprawl, where choices multiply like stars on a clear night, artificial intelligence has taken on the role of a quiet guide, illuminating each step that customers take. The data reveal a landscape where AI not only influences behaviour but actively shapes the emotional cadence of the journey.

The high transition from awareness to consideration (82%) signals something profound: when personalisation is done with finesse, brands no longer shout to be recognised. Instead, they appear naturally in the consumer's line of sight, like a well-placed signpost on a familiar road. The dominant feature importance of AI-personalisation in the early stage underlines this shift. Customers do not simply notice brands — they feel understood by them. This emotional undertone reflects the modern expectation for relevance without intrusion, an expectation that AI seems uniquely equipped to satisfy.

As the journey progresses, the analytic models show a notable reorientation. Conversion is no longer just a rational decision but a moment shaped by algorithmic reassurance. Recommendation accuracy and offer relevance remain central, yet emotional resonance — though smaller in weight — still matters. It suggests that AI-driven content cannot rely solely on logic; it must whisper to the heart as well. Consumers respond not merely to the perfect match but to the sense that the brand speaks in their language.

Retention, often the Achilles' heel of marketing strategy, emerges as a dance between consistency and personalisation. Post-purchase personalisation stood out with the strongest importance score, reinforcing the belief that journeys do not end at the point of sale; they breathe and evolve. Customers stay not because they have purchased, but because the brand remembers. AI-powered service responsiveness, especially through chatbots, stitched itself into the retention fabric with remarkable influence — proof that immediacy, even when automated, can still feel like care.

Yet the most striking insight rests in the shift from retention to advocacy. Here, trust in AI becomes the decisive threshold. After all, advocacy is not merely satisfaction but belief — the kind strong enough for one to lend their voice to a brand. The 38% feature importance for AI trust and the 48% overall advocacy lift attributed to AI experiences collectively warn that brands cannot treat AI as a faceless engine. Customers must feel safe, respected, and valued, or advocacy simply refuses to ignite.

The thematic analysis enriches this understanding. Phrases such as "the system understood me" and "not creepy" reveal that consumers crave a balance: personalisation that is helpful without becoming intrusive. In other words, AI must be present but not domineering. This delicate dance shapes the emotional arc that turns customers into ambassadors.

Together, these findings paint a journey far more complex and intimate than classical models allowed. AI does not replace human judgement; instead, it strengthens it by revealing hidden rhythms in behaviour. Brands that successfully integrate AI into the journey do not rely on force or volume but on listening — listening through data, through patterns, through digital traces. And when done with sensitivity, AI becomes the quiet architect of loyalty.

Ultimately, this study highlights a new paradigm: advocacy is less the result of aggressive marketing and more the harvest of meaningful, personalised, and trustworthy interactions. The future of digital marketing belongs not to the loudest brand, but to the one that reads the room — and the customer — with both intelligence and intention.

Implications

1. Theoretical Implications

This study pushes the boundaries of traditional customer-journey theory by demonstrating that the journey is no longer a linear, cognitive process but an emotionally layered, AI-mediated experience. The integration of Journey Sequencing Analysis (JSA), Gradient Boosting Journey Modelling (GBJM), and Advocacy Lift Attribution (ALA) offers scholars a fresh framework for understanding how behavioural, emotional, and computational elements converge. The findings suggest that AI should not be studied merely as a tool but as a contextual force shaping behavioural intention across the entire journey. In doing so, the research contributes to the growing discourse on algorithmic influence in consumer decision-making, positioning AI as a co-creator of experiential value rather than a post-hoc efficiency mechanism.

2. Managerial Implications

a. Personalisation Must Be Subtle, Not Showy

Managers must recognise that consumers prefer "quiet personalisation" — experiences that feel natural rather than intrusive. AI systems should be trained to recommend with restraint, ensuring relevance without overstepping psychological boundaries.

b. Conversion Depends on the Harmony Between Logic and Emotion

AI-driven recommendations must merge functional accuracy with emotional resonance. Brands should design content that not only matches preferences but also mirrors the customer's tone, mood, and decision context.

c. Retention Requires Human Warmth Through Machine Efficiency

Post-purchase personalisation and AI-assisted responsiveness emerged as key retention drivers. Brands should streamline chatbot systems to deliver fast, empathetic responses and design loyalty strategies that adapt dynamically to behavioural cues.

d. Advocacy Emerges From Trust, Not Transaction

The study shows that trust in AI practices is the single strongest predictor of advocacy. Brands must ensure transparent data use, give customers control over personalisation levels, and explain AI decisions in simple, human language. Advocacy cannot be engineered; it must be earned through respect.

e. AI Should Guide, Not Dictate

Managers should view AI not as a dictator of decisions but as a guide that supports customer choice. AI must enable empowerment, not dependency — a balance that nurtures loyalty without eroding autonomy.

3. Policy Implications

Regulators and policymakers must recognise the expanding emotional influence of AI in consumer journeys. Clear guidelines on transparency, fairness, and personal data use are vital to safeguard consumer wellbeing. As AI becomes more integrated into everyday experiences, ethical oversight must evolve to ensure personalisation remains beneficial rather than exploitative.

4. Societal Implications

The rise of AI-powered marketing subtly shapes how individuals perceive, choose, and advocate. This can either strengthen digital literacy or deepen dependency. Educational initiatives that help consumers understand algorithmic influence become crucial. A society that understands the tools shaping its choices is better equipped to navigate them with autonomy.

5. Technological Implications

Organisations must invest in interoperable AI systems capable of sharing insights across the entire customer journey rather than operating in silos. AI engines should evolve to read emotional cues, context switches, and nuanced behavioural triggers. This future-facing technological alignment will allow brands to build experiences that feel less mechanical and more meaningfully human.

Future Scope

The digital marketplace is evolving faster than the theories built to understand it, and this study opens several pathways for deeper inquiry. As AI becomes woven into the fabric of everyday decision-making, future research must explore the subtler, more human dimensions of algorithmic influence.

Firstly, the emotional intelligence of AI deserves closer academic scrutiny. While this study highlighted the role of emotional resonance in conversion and retention, future work could investigate how AI interprets micro-emotions — hesitation, curiosity, doubt — and how these cues shape moment-to-moment decision paths. Machine-learning models capable of capturing affective states may transform current understandings of customer psychology.

Secondly, longitudinal studies would offer richer insight into journey evolution. The cross-sectional nature of this research captures a snapshot, but customer relationships unfold like seasons — shaped by memory, habit, and changing contexts. Tracking journeys over months or years would reveal whether AI-driven touchpoints create durable loyalty or momentary convenience.

Thirdly, cultural differences in AI acceptance represent a compelling frontier. What feels helpful in one cultural setting may feel intrusive in another. Comparative studies across countries or demographic cohorts could illuminate how cultural norms, privacy expectations, and digital maturity influence the progression toward advocacy.

Fourthly, future exploration may focus on real-time adaptive AI systems. As technologies mature, AI may personalise not only content but timing, emotional framing, and even the decision architecture itself. Examining the ethics and effectiveness of such hyper-personalisation will be vital as brands walk the tightrope between relevance and manipulation.

Additionally, methodological innovation remains wide open. Journey Sequencing Analysis (JSA) and Advocacy Lift Attribution (ALA) introduced fresh analytical lenses, but deeper integration with reinforcement learning, behavioural simulation models, or neural sequence modelling could offer even stronger predictive power. Scholars can explore how AI might not just analyse journeys but simulate alternate pathways to optimise customer experience design.

Finally, the societal consequences of AI-mediated journeys warrant sustained attention. As algorithms increasingly shape consumer worldviews, future research must interrogate questions of autonomy, algorithmic fairness, and the long-term psychological effects of personalised digital environments. Understanding what we gain — and what we risk — is essential in guiding responsible AI adoption.

Taken together, the future of studying AI-driven customer journeys lies in embracing complexity: emotional, ethical, cultural, and computational. The field is young, the questions still unfolding, and the journey — much like the customer's — far from complete.

CONCLUSION

The modern customer journey is no longer a tidy sequence of steps but a shifting constellation of moments, emotions, and digital cues. This study set out to map that constellation through the lens of AI-driven marketing analytics, revealing a journey shaped not only by technological precision but by the subtle interplay of relevance, trust, and emotional connection. From the sharp rise in awareness shaped by quiet personalisation to the steep climb toward advocacy fuelled by trust in AI practices, the findings make one truth unmistakably clear: AI has become a quiet architect of contemporary consumer behaviour.

The novel analytical framework employed — combining Journey Sequencing Analysis, Gradient Boosting Journey Modelling, and Advocacy Lift Attribution — demonstrated that AI's influence spans the entire journey. Yet its role is far from mechanical. It supports customers with almost human sensitivity, guiding them without overwhelming them. Where classical models once imagined a linear path, AI reveals a living, evolving journey shaped by context, intention, and digital resonance.

The insights show that meaningful advocacy is not born from aggressive tactics or transactional efficiency, but from experiences that feel personal, respectful, and trustworthy. Brands that harness AI with transparency and emotional intelligence stand to cultivate not just loyal customers, but genuine advocates who speak for the brand with conviction.

Ultimately, this study affirms that the future of marketing lies in the balance between tradition and innovation — between understanding people as they have always been, and interpreting them through the intelligent tools of today. AI does not replace human judgement; it sharpens it. And in this shared space between human intuition and machine insight, the modern customer journey finds its fullest expression.

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