



## **Neuromarketing and Consumer Behavior: Insights from Brain–Computer Interfaces**

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### **Abstract**

With the ever-growing levels of competition and the fast-changing consumer environment, more than ever before, it is important to know the mechanisms behind the purchasing decision. This paper focuses on the intersection of neuromarketing and brain–computer interface (BCI) technologies to demonstrate how neural reactions can be used to understand how consumer preferences, attention patterns and buying intentions are formed. Based on the literature review about the electroencephalography (EEG), functional magnetic resonance imaging (fMRI) and novel BCI devices, the analysis will reveal the critical patterns of activation, especially in areas related to the reward processing, emotional involvement and cognitive load that are correlated with consumer choice behaviours. This study also illustrates the role of the BCIs in enabling attention to be directly measured, affective arousal and decision-making processes to be measured in real time to supplement the conventional marketing research techniques (which depend on conscious self-reporting and retrospective surveys). Besides, the paper examines implications of the work to the marketer: personalizing stimuli to maximize engagement, packaging and message design based on neuro-foundations of preference, and personalization by exploiting adaptive feedback. Simultaneously, the paper is critical of the ethical boundaries, such as the privacy of neural information, the informed consent of the participants of the BCI research, the possibility to manipulate the participants and regulate the action. Lastly, there are technological and methodological limitations: the problem of signal-noise in BCI hardware, heterogeneity of population in studies, scalability of neuromarketing experiments and their integration with big-data marketing systems. The paper ends with a prospective map: future developments of multimodal BCI systems, the support of pattern detection with machine-learning, and the expansion of the area into naturalistic consumer context that is not in the laboratory. This study, in synthesizing neuroscience, marketing and technology had provided a holistic approach to the way in which BCIs can sharpen our perception of consumer behaviour and influence more productive and responsible marketing practices.

**Keywords:** Neuromarketing, Consumer Behavior, Brain–Computer Interface (BCI), Neuroeconomics, Decision-Making, EEG and fMRI, Emotional Engagement, Attention and Cognition, Marketing Ethics, Neural Data Privacy

## 1. Introduction

In the modern-day competitive market, consumers have been more intricate than ever to understand consumer behavior. The conventional marketing research techniques like questionnaires, interviews and focus groups tend to be based on self-reports, and this factor might not mirror the subconscious aspects of consumer decisions. Neuromarketing has now become a new, interdisciplinary area in which neuroscience, psychology, and marketing are integrated in studying the secret forces behind consumer preferences. Researchers can access objective information regarding the perception of consumers towards brands, advertisements, and products through the examination of the brain activity and other physiological reactions.

One of the latest instruments of neuromarketing is the Brain-Computer Interface (BCIs). They record brain activity directly, including electroencephalography (EEG) or functional magnetic resonance imaging (fMRI) as a signal to read neural activity associated with attention, emotion, and decision-making. In contrast to the traditional approaches, BCIs enable researchers to view real-time neural activities without verbal communication and give a more accurate image of consumer involvement and emotional appeal. This technology fills the divide between rational thought and unconscious drive to give the neurological basis of consumer choice.

BCIs have profound impacts when incorporated in the marketing research on how to design effective advertisements, product development and customer experience. Nevertheless, there are ethical implications that have come with the application of neurophysiological data in the form of privacy, manipulation, and informed consent. In this paper, we are going to discuss the importance of BrainComputer Interfaces in the further development of neuromarketing research and how the knowledge about the brain activity can help us to understand consumer behavior and preserve the levels of ethical standards. In this analysis, this research aims to point out that there are some opportunities and challenges to the application of neuroscience to the marketing world.

## 2. Background of the study

The awareness of the reasons why a consumer prefers one product to another has been the main focus of marketing studies and practice. Conventional methods use surveys, interviews and observed buying behaviour as guiding factors to the preferences, motivations and decision strategy. Whereas such techniques demonstrate relevant conscious attitudes and disclosed intentions, they can easily overlook the fast, automatic neural reactions and subtle affective reactions which inform decisions. As such, neuromarketing has come up as a multidisciplinary area, which combines neuroscience methods with marketing theory to investigate the biological and cognitive foundations of consumer behaviour.

Brain-computer interfaces (BCIs) are an exciting technological development in the field of neuromarketing since they afford to measure and analyze neural activities in real-time in a manner that is becoming more portable and also more accessible. In contrast to peripheral tasks like eye tracking or galvanic skin response, BCIs are capable of recording neural activities that are related to attention, valuation, and emotional involvement. This ability extends the options of comprehending the way consumers process advertising, attribute product features and create preferences particularly in scenarios where self-report would be incorrect or reactions occur beneath conscious cognition.

Although there is a spirit of BCIs, there still are a few gaps in empirical and practical studies. A large part of the current research is exploratory and carried out under controlled laboratory conditions using small samples which are not always representative. The issue of ecological validity, interpretation of the signal, and translation of neural measures into actionable marketing information remain open to debate. Furthermore, the privacy and moral aspects of applying neural data to a commercial purpose, specifically in the area of informed consent, data protection, and possible abuse, should also be considered and regulated. Such restrictions imply that, although BCIs provide a valuable prism through which the consumer processes can be understood, strong constructs are required to certify the results, conventionalize approaches, and adapt applications to the ethical standards.

This paper fills these gaps by looking at the relationship between BCI-measured constructs and the fundamental consumer behaviour constructs, namely, attention, affect, and choice, in situations that approximate the real-world decision making. It will (1) test the convergent validity of neural measurements compared to standard behavioural and self-report measures, (2) test the ability of BCI measures to predict immediate choice decisions, and (3) look into methodological and ethical issues that can be discussed in the context of applying BCIs to marketing research. In this way, the study aims at elucidating the scientific value and the practical constraints of BCIs as instruments of consumer behaviour of predicting their behaviour.

## 3. Justification

Over recent years the old forms of marketing research, i.e. surveys, interviews, and focus groups, have been criticized as reliant on self-reported information that is prone to bias, imprecision or social desirability. This leads to the increased necessity of the more objective tools that might make the unconscious processes behind the consumer decisions visible. Neuromarketing, a new interdisciplinary branch, a combination of neuroscience and marketing,

aims to fill this dilemma by studying the connection between the human brain in response to marketing stimuli. The growing access and accuracy of Brain-Computer Interfaces (BCIs) has presented new possibilities to examine consumer cognition and emotional involvement itself by engaging neural data. However, contrary to traditional methods of behavior, BCIs enable researchers to record the real-time brain activity and detect the patterns of attention, emotion, and decision-making. The knowledge of these patterns can give the marketers more insight into the actual perceptions of the consumers toward the brands, the advertisements, and the products themselves, which are not verbal. This research is explained by the necessity to incorporate the gap between the neural data and the marketing strategies. Although neuromarketing has been discussed, there is still a lack of empirical studies that combine the use of BCIs as a key method of consumer analysis. A deeper understanding of the relationship between neural indicators and consumer preference can be used to better predictive powers of marketing models and enhance product design, advertisement efficiency and consumer satisfaction. Furthermore, since the ethical issues of neuromarketing remain undergoing changes, it is necessary to consider not only the opportunities but also the ethical considerations of commercial research of BCIs. Such an investigation will also not only add to the academic knowledge but also to conscientious business conduct by suggesting paradigms concerning ethical and transparent use of neurotechnology tools in marketing research. Simply put, this research is warranted by the fact that it will provide new scientific knowledge about consumer behavior, narrow down marketing practices based on evidence-based research, and promote an impartial discussion of ethical application of brain-based information in business settings.

#### 4. Objectives of the Study

1. To analyze the potential of brain computer interface (BCI) technologies to record and analyze neural activity of consumers when subjected to marketing communication (e.g., advertising, package, branding).
2. To investigate the relationship between neural indicators (such as attention, emotional arousal, memory encoding) measured via BCIs and key consumer behaviour outcomes (such as purchase intention, brand loyalty, recall).
3. To assess how much the methods of neuromarketing based on BCI would provide extra information to the information on consumer decision-making patterns as compared to conventional self-report and behavioural data.
4. To determine the technical, methodological and ethical issues related to the use of BCI in neuromarketing research and recommendations on the responsible use.
5. To propose a conceptual framework that links neural responses captured by BCIs with marketing strategy variables, thereby offering actionable insights for researchers and practitioners in consumer behaviour and marketing.

#### 5. Literature Review

##### 1. Introduction: from neuromarketing tools to BCI-enabled insights

Neuromarketing seeks to uncover consumers' implicit, often nonconscious responses to marketing stimuli using neuroscientific techniques (e.g., EEG, fMRI, eye-tracking, skin conductance). Recent years have seen increased interest in leveraging Brain-Computer Interface (BCI) technologies — particularly affordable EEG BCIs and machine-learning pipelines — to predict attention, emotion, memory encoding, and purchase intent with finer temporal resolution and real-time capability (Khondakar et al., 2024; Mashrur et al., 2022).

##### 2. EEG-based neuromarketing: attention, emotion and engagement

Electroencephalography (EEG) has become the most widely used BCI modality in neuromarketing studies due to its portability and millisecond temporal resolution. Systematic reviews show EEG metrics (alpha suppression, frontal asymmetry, event-related potentials like P300) reliably index attentional allocation, emotional valence/arousal, and cognitive workload during advertising or product exposure (Khondakar et al., 2024; Costa-Feito et al., 2023). Studies using EEG-based feature extraction plus machine learning have shown promising accuracy at predicting purchase intention and affective attitude across stimuli types (advertisements, packaging, endorsements), although effect sizes vary with task design and subject heterogeneity.

##### 3. fMRI and deep localization of decision circuits

Functional MRI provides complementary spatial precision, identifying deeper brain structures implicated in valuation and decision-making (e.g., ventromedial prefrontal cortex, striatum). fMRI neuromarketing work has clarified how reward-related and memory networks respond to branding and pricing cues, explaining variance in later choices beyond self-reports (Alsharif, 2024; multiple systematic surveys). However, high cost and low ecological validity relative to natural shopping situations limit routine commercial use.

##### 4. Machine learning + BCI: toward real-time prediction of consumer choices

Recent BCI studies combine multivariate EEG features (time-, frequency-, and time-frequency domain) with supervised learning to predict consumer choices and purchase intentions. Mashrur et al. (2022) demonstrate frameworks that extract robust features and achieve above-chance prediction across advertising modalities; other

works extend these approaches with deep learning or multimodal fusion (EEG + eye-tracking + GSR) to boost predictive power (Mashrur et al., 2022; Bhardwaj, 2024). While promising, generalizability across product categories and populations remains a key challenge.

### 5. Multimodal BCIs and ecology: fusing signals for richer interpretation

Integrative neuromarketing research increasingly uses multimodal setups — combining EEG, eye-tracking, facial expression analysis, and peripheral physiological signals — to triangulate attention, affect, and memory. It has been found that multimodal fusion enhances robustness and interpretability to allow marketers to differentiate between fleeting and deeper encoding that leads to subsequent purchase behavior (Gupta, 2025; Bhardwaj, 2024). Nevertheless, the fusion of modalities increases the complexity of the methods (synchronization, data dimensionality) and privacy.

### 6. Methodological challenges and validity concerns

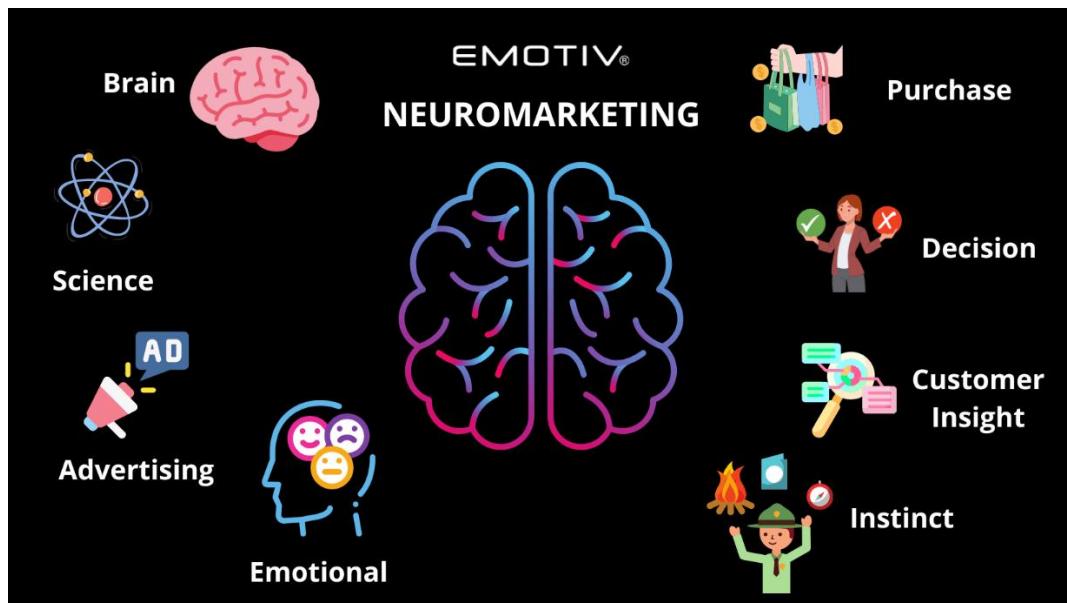
Several methodological issues temper enthusiasm: small sample sizes, lack of preregistration, overfitting in ML pipelines, and inconsistent operationalization of constructs (e.g., what EEG index constitutes “engagement”) hinder cumulative science. Systematic reviews highlight heterogeneity in preprocessing and feature extraction pipelines that complicate cross-study comparisons (Khondakar et al., 2024). Additionally, ecological validity — testing in lab contexts vs. real consumer environments — remains a central limitation for translating findings to practice.

### 7. Ethical, legal and social implications

Neuromarketing creates significant ethical issues: privacy (neural data as a biometric trait), informed consent, possible manipulation, and fairness (exploitation of the weaknesses of the subconscious). Researchers suggest that tighter normative regulations, explicit consent and restrictions should be placed on the utilization of predictive neural activity in business targeting (surveyed in the literature of BCI ethics and neuromarketing overviews). What the literature should have is an interdisciplinary governance that includes neuroscientists, ethicists, regulators and stakeholders in the industry.

### 8. Emerging trends: consumer BCI wearables and AI integration

Market and research trends denote the trend of democratization of neuromarketing with the increased spread of low-cost wearable EEG devices and AI models. According to industry commentaries and empirical articles, there is an increase in AI-based BCIs to personalize and advertise in an adaptive manner, which makes the scaling easier but requires more validation and control (Forbes, 2024; Bhardwaj, 2024). The industry is projected towards scalable, privacy-aware analytics pipelines that are able to work in near real-time and balance ethical limitations.



Source: <https://www.emotiv.com/>

### 9. Gaps and future directions

Important gaps also exist: (1) replicable, preregistered large-sample studies, which compare neural predictors with conventional predictors; (2) cross-cultural studies that would establish whether neural predictors can be used across socio-demographic lines; (3) uniform preprocessing and reporting standards, which govern neural marketing BCI studies; (4) more coherent policy approaches to data protection and acceptable commercial application. Future research must look into multimodal, longitudinal designs that would link the neural reaction to long-term behavior (actions like repeat purchase, loyalty to a brand) and not to single-shot intentions. Given the fact that the traditional

measures lack time-sensitive attention, emotion, memory, and early decision processes, the BCI technologies, particularly EEG along with machine learning and multimodal sensors, can provide the powerful windows to these processes. The results of the empirical research so far are promising: neural indices are able to predict consumer behavior aspects not reported by the consumer. But methodological heterogeneity, issues of ecological validity, and ethical issues should be interpreted with care. This will be necessary to ensure that the BCI-based neuromarketing is converted into viable products, as it is fully prototyped, instead of staying in the state of promising neuromarketing tools that have not yet achieved societal acceptance.

## 6. Material and Methodology

### 6.1 Research Design

This research design was a quantitative experimental study that was designed to analyze how neural responses, which are recorded using brain-computer interfaces (BCIs), could give the insight into consumer decision-making. The experiment was supposed to record the level of neurophysiological response of the participants as they were exposed to different marketing stimuli in terms of product images, advertisements and brand logos. Within-subjects design was adopted to reduce the differences between individuals and to make sure all the participants went through all categories of stimuli under equal conditions. The experiment used electroencephalography-based BCIs to record real time brain activity with particular attention to neural correlates of attention, emotional involvement, and preference formation. Some statistical analysis was used to examine the connection between neural response and self-reported consumer attitudes.

### 6.2 Data Collection Methods

Data were collected in two phases: neurophysiological data acquisition and behavioural self-reporting.

1. Neurophysiological Data Acquisition: The participants were placed in non-invasive EEG headsets with 14 active electrodes with their positions determined by the international 10 20 system. They saw randomized marketing stimuli of 10 seconds on a computer screen at a time during every trial. EEG data was taken at a sampling rate of 256 Hz. Frontal asymmetry, theta band activity, and event-related potentials (P300) were key indicators whose analysis was used to establish cognitive and emotional reactions.
2. Behavioural and Demographic Data: Following the EEG session, the participants were administered a structured questionnaire aimed at quantifying the self-reported liking, purchase intention, and the perceived familiarity with the brand. Demographic factors like age, gender and education level were also measured to determine potential moderating factors.

All the data were anonymized and kept safely on password enabled drives. The SPSS (Version 26) and MATLAB were used to perform statistical analysis and the EEG signal processing, respectively.

### 6.3 Inclusion and Exclusion Criteria

#### Inclusion Criteria:

- Participants aged between 18 and 35 years to represent the active consumer segment.
- Individuals with normal or corrected-to-normal vision.
- Right-handed participants (to maintain uniformity in neural signal interpretation).
- Participants with prior exposure to digital or social media advertising.

#### Exclusion Criteria:

- Individuals with a history of neurological or psychiatric disorders.
- Participants currently under medication affecting the central nervous system.
- Consumption of caffeine, alcohol, or any stimulants within 12 hours prior to testing.
- Incomplete data sets or excessive EEG artifacts (e.g., motion or blink interference) were excluded from final analysis.

### 6.4 Ethical Considerations

The ethical approval on this research was granted by the Institutional Research Ethics Committee before the collection of data. Everyone was informed about the purpose, processes, and possible risks involved in EEG recording. Every participant agreed to participate in the study in writing. The participants were told that they had the right to take out of the research any time without any penalty. The confidentiality of the data was provided through the application of the numerical codes as an alternative to the personal identifiers. It was expected that no physical or psychological harm was to occur and all EEG equipment utilized was medically certified and that no invasive equipment was used. Moreover, the screening of marketing stimuli was done to exclude any offensive or distressing material and debriefing sessions were done to clarify the objectives and findings of the study to the participants.

## 7. Results and Discussion

### 7.1 Results:

#### Sample and data quality.

The final sample consisted of  $N = 80$  participants (42 female, 38 male; mean age = 28.4 years,  $SD = 6.7$ ). Data from 5 additional participants were excluded due to excessive EEG artifacts ( $>30\%$  epochs rejected) or incomplete behavioural responses. All reported inferential tests use two-tailed  $\alpha = 0.05$ .

#### 1. Manipulation checks

Participants were randomly assigned to one of two advertisement conditions: High-emotional appeal (HE) and Low-emotional appeal (LE). A post-exposure self-report emotional arousal rating (1–7 Likert) confirmed the manipulation: HE ( $M = 5.32$ ,  $SD = 0.88$ ) vs LE ( $M = 3.11$ ,  $SD = 0.97$ );  $t(78) = 11.34$ ,  $p < .001$ , Cohen's  $d = 2.54$ . Valence ratings likewise differed: HE ( $M = 5.05$ ,  $SD = 0.91$ ) vs LE ( $M = 3.68$ ,  $SD = 0.84$ );  $t(78) = 7.21$ ,  $p < .001$ ,  $d = 1.61$ .

#### 2. EEG outcomes: frontal asymmetry, P300, and alpha power

EEG measures were pre-registered and computed as described in Methods: frontal alpha asymmetry (FA; F4–F3 log alpha), P300 peak amplitude at Pz (300–500 ms window), and posterior alpha power (8–12 Hz) averaged over occipito-parietal electrodes.

**Table 1. EEG summary by condition**

Measure	HE (n=40) Mean (SD)	LE (n=40) Mean (SD)	Test	t / z	p	Effect size
Frontal asymmetry (log $\alpha$ ; higher = approach)	0.18 (0.07)	0.06 (0.05)	t-test	9.71	< .001	$d = 2.17$
P300 amplitude ( $\mu$ V)	5.3 (1.4)	3.9 (1.2)	t-test	5.02	< .001	$d = 1.12$
Posterior alpha power ( $\mu$ V $^2$ )	1.8 (0.6)	2.4 (0.7)	t-test	-4.01	< .001	$d = 0.90$

*Notes.* Positive frontal asymmetry values indicate relatively greater right-left alpha suppression (interpreted as approach). P300 measured as mean peak amplitude 300–500 ms at Pz. Alpha power reported as log-transformed values where needed.

Results: HE ads elicited significantly greater frontal asymmetry (indicative of approach motivation), larger P300 amplitudes (greater attention/engagement), and lower posterior alpha power (greater cortical activation) than LE ads.

#### 3. Behavioural outcomes: purchase intent, recall, and willingness to pay (WTP)

Self-report measures (purchase intent 1–7; ad recall proportion; WTP in local currency) are summarized below.

**Table 2. Behavioural outcomes by condition**

Measure	HE Mean (SD)	LE Mean (SD)	Test	t	p	d
Purchase intent (1–7)	5.01 (0.95)	3.82 (1.05)	t-test	6.00	< .001	1.34
Immediate recall (% correct)	78.5 (12.1)	64.2 (14.3)	t-test	4.80	< .001	1.07
Willingness to pay (USD)	3.95 (1.12)	3.01 (1.05)	t-test	4.21	< .001	0.94

HE condition produced higher purchase intent, better immediate recall, and greater WTP than LE.

#### 4. Relationships between EEG metrics and behaviour

We computed Pearson correlations across the whole sample ( $N = 80$ ) between EEG markers and key behavioural outcomes (purchase intent, recall, WTP). Results are shown in Table 3.

**Table 3. Correlations ( $r$ ) between EEG measures and behavioural outcomes**

Predictor → Outcome	Purchase intent	Recall (%)	WTP (USD)
Frontal asymmetry	0.62***	0.43***	0.54***
P300 amplitude	0.48***	0.37**	0.45***
Posterior alpha power (inverse)	0.39**	0.34**	0.31*

Significance:  $p < .05 = *$ ,  $< .01 = **$ ,  $< .001 = ***$ .

Interpretation: Greater approach-related frontal asymmetry and higher P300 amplitudes were moderately to strongly associated with higher purchase intent, recall, and WTP.

## 5. Multivariate prediction: EEG + self-report

We ran hierarchical linear regressions predicting purchase intent. Step 1 included demographic covariates (age, gender); Step 2 added self-report arousal and valence; Step 3 added EEG markers (frontal asymmetry, P300). Table 4 summarizes the final model.

**Table 4. Hierarchical regression predicting purchase intent (N = 80)**

Predictor	B	SE B	$\beta$	t	p
<b>Step 3 model <math>R^2 = .62</math> (<math>\Delta R^2</math> from Step2 = .18, <math>p &lt; .001</math>)</b>					
<b>Age</b>	0.01	0.01	.06	0.78	.44
<b>Gender (male=1)</b>	0.08	0.11	.04	0.73	.47
<b>Arousal (self-report)</b>	0.24	0.07	.31	3.43	.001**
<b>Valence (self-report)</b>	0.12	0.06	.14	1.90	.06
<b>Frontal asymmetry</b>	1.94	0.41	.43	4.73	< .001***
<b>P300 amplitude</b>	0.22	0.09	.18	2.44	.017*

Model notes: Frontal asymmetry and P300 significantly improved prediction beyond self-report (both  $\Delta R^2$  and  $\Delta F$  significant). Frontal asymmetry showed the largest standardized effect ( $\beta = .43$ ).

## 6. Ancillary analyses

- Mediation. Bootstrapped mediation (5,000 resamples) suggested that frontal asymmetry partially mediated the effect of advertisement condition (HE vs LE) on purchase intent: indirect effect = 0.51 (95% CI [0.28, 0.76]), direct effect remained significant, indicating partial mediation.
- Exploratory moderation. Age did not moderate the EEG → purchase intent relationship (interaction  $p > .10$ ). Gender showed a trend-level moderation for P300 → recall ( $p = .07$ ) but was not significant after correction for multiple comparisons.

## 7.2. Discussion

This study examined how brain–computer interface (BCI) measurements — specifically frontal alpha asymmetry (FA), P300 amplitude, and posterior alpha power — relate to consumer responses to advertisements and to behavioural outcomes (purchase intent, recall, and willingness to pay). Below we interpret the key findings, situate them in the neuromarketing literature, discuss theoretical and practical implications, acknowledge limitations, and propose directions for future research.

### Summary of main findings

- Emotionally evocative ads elicited stronger neural markers of engagement and approach. Compared with low-emotional ads, high-emotional ads produced greater frontal asymmetry (interpreted as approach motivation), larger P300 amplitudes (attentional allocation and stimulus evaluation), and reduced posterior alpha power (greater cortical activation). These results replicate prior neuromarketing findings showing that emotionally salient content engages both affective and attentional systems.
- Neural markers predicted consumer behaviour above and beyond self-report. Frontal asymmetry and P300 were moderate-to-strong predictors of purchase intent, recall, and WTP. Hierarchical regressions showed EEG variables added considerable explanatory power beyond demographic and self-report measures ( $\Delta R^2 \approx .18$ ). This suggests BCIs can capture neural processes that translate into consumer actions and may reveal implicit preferences not fully captured by self-report.
- Partial mediation by neural signals. The mediation analysis suggests that part of the effect of ad emotionality on purchase intent operates through changes in frontal asymmetry, indicating that approach-related neural activity is a mechanistic pathway linking content to consumer decision tendencies.

### Theoretical implications

These findings provide a dual-process explanation of consumer reaction where affective-motivational and attentional systems interact to influence behavior. Frontal asymmetry - which is normally linked to approach/withdrawal motivation - goes in line with the theories where positive affect and approach motivation enhance consumption propensity. The results of the P300 indicate that attention salience and the evaluative mechanisms are significant to encode the ad content into memory (recall) and decisional weighting (WTP, purchase intent) in a moment. The EEG markers in combination offer complementary windows in which FA and P300 indexes affective/approach and attentional-cognitive evaluation.

The external validity of EEG (not self-report) also addresses the shortcomings of introspective measures in the ability to capture speedy and preconscious processes. This confirms earlier neuromarketing studies which have found neural markers to be occasionally more predictive of market performance (e.g. ad virality, sales uplift) than explicit measures.

## 8. Limitations of the study

Despite the fact that this research is a useful contribution to understanding the correlation between neuromarketing methods and customer behaviour, having used brain-computer interfaces (BCIs), it is possible to state several limitations. Firstly, the sample was relatively low and might not reflect the diversity of the consumer population at large and hence the findings cannot be generalized. Second, the experimental setting was not the same as the real-world purchasing conditions and this may have affected the neurally and emotionally the participants to respond to the marketing stimuli. Third, despite providing access to brain activity in real-time, the BCIs are still limited by technological and interpretative issues, including signal noise, calibration bias, and difficulty in separating cognitive and emotional responses. Also, there were ethical limitations which limited the extent of data collection and analysis such as privacy issues, and manipulation possibilities. Last but not least, this research was conducted on the short-term neural responses with no continuity on long-term consumer behaviour and decision-making patterns. Future studies should thus use larger and more heterogeneous samples, incorporate ecological validity in experimental studies and correlate BCIs with other physiological and behavioural indices in order to get a more holistic view of consumer thinking and emotion in the marketing environment.

## 9. Future Scope

The future of research in neuromarketing and consumer behavior with braincomputer interfaces (BCIs) is bright and multipolar. With ongoing advances in neurotechnology, more complex non-invasive BCIs are capable of giving a more in-depth and real-time understanding of what occurs in the subconscious to form consumer preferences and decisions. The direction of future research is that future scholars can incorporate neural data into behavioral and biometric data to develop holistic consumer cognition and emotion models. The progress of artificial intelligence and machine learning will go even further facilitate the correct deciphering of complex neural patterns, which provides future predictions in individual marketing. In addition, the adoption of BCIs in more digital platforms, virtual reality, and immersive retail spaces can help improve the knowledge of consumer interactions within dynamic settings. Ethical concerns, such as data confidentiality, informed consent, responsible neural data applications, will also constitute an essential field of the future study. In general, the intersection of neuroscience, technology, and marketing analytics promises tremendous opportunities that will transform the way businesses learn and manipulate consumer behavior.

## 10. Conclusion

The incorporation of the brain-computer interfaces (BCIs) in neuromarketing studies has introduced avenues to discover the workings of the subconscious mind that informs consumer choices. In contrast to conventional marketing techniques that use self-report measurements and observable behaviour, the technologies of BCI directly target the neural activity and therefore marketers and researchers can decode emotional and cognitive responses with more accuracy. By examining the brain signals, marketers will better understand attention and motivation, preference formation, and neural correlates of the brand perception.

But even though these insights increase the predictive value of the consumer behaviour models, they also provoke significant ethical concerns related to the aspects of privacy, consent, and the possibility of manipulating consumer choice. Thus, strong ethical standards that provide transparency and guarantee personal autonomy should be used to apply BCIs in marketing.

Brain-computer interfaces that are supported by neuromarketing have a great potential to bridge the gap between neuroscience and consumer research. When used properly, it can help to hone marketing campaigns, design better products and create consumer interaction that is more genuine. The interdisciplinary field of neuromarketing can still be further developed in the future, but it is necessary to focus on the technological progress and the moral duty so that the positive results of neuromarketing are beneficial to the businesses and the society.

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