

# Khwaja Yunus Ali University Journal

Publisher homepage: [www.kyau.edu.bd](http://www.kyau.edu.bd)

**OPEN ACCESS**

ISSN: 2791-3759 (Online), 2521-3121 (Print)

Journal homepage: [www.journal.kyau.edu.bd](http://www.journal.kyau.edu.bd)



## Research Article

### The role of information on social media marketing on purchase decision: Examining the mediating role of brand value

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#### ABSTRACT

*Social media plays a significant role in influencing consumer purchase decisions and has become a key factor in the overall process of buying goods and services. Numerous scholars have conducted separate studies on its impact on consumer behavior and have consistently found social media to be a powerful influencer. In our study, we focused on three specific factors related to social media marketing: informativeness, reliability, and motivation. We collected data using a convenient sampling technique, with 241 individuals participating as samples. To assess their satisfaction level, we employed a Likert scale in our survey. The results of our study revealed that motivation and informativeness had a positive and significant relationship with purchase decisions, whereas reliability did not show the same influence. However, we observed that brand value partially mediated the impact of motivation and informativeness on purchase decisions. Based on the data presented, it is evident that informativeness and motivation are crucial factors in a person's decision to purchase a product from a particular company. Therefore, decision-makers need to implement effective marketing strategies to enhance sales and profitability.*

**Keywords:** Informativeness, Reliability, Motivation, Brand Value, Purchase Decision, Bangladesh

#### 1.0 Background

As the corporate business landscape grows more competitive and dynamic, employees within these enterprises face an ongoing imperative to adapt their business and marketing strategies to maintain a competitive edge. The Internet and prominent social media platforms, including Facebook, Twitter, and Instagram, assume a pivotal role in attracting new customers and ensuring their satisfaction. Consequently, professionals from diverse backgrounds, encompassing practitioners, academics, and business experts, are increasingly directing their attention toward social media (SM) platforms such as Facebook, LinkedIn, Twitter, and Google+. These social media platforms serve as digital communities where users freely exchange ideas, personal messages, photos, and videos, as underscored by Albors *et al.* (2008) in their research. Many businesses have astutely recognized the vast potential of these platforms and are harnessing them for a myriad of purposes. Nonetheless, there exists a research gap concerning how social media can effectively drive product promotion, brand building, and online advertising. Specifically, there is a pronounced deficiency in understanding the impact of social media marketing

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(SMM) on consumer purchasing decisions, particularly in South Asian developing or least-developed countries (LDCs) like Bangladesh, India, Sri Lanka, and Nepal, which collectively house a significant portion of the world's population.

In the realm of business-to-consumer (B2C) interactions, social media has been extensively scrutinized due to its established constructs and concepts pivotal to comprehending customer behavior, as emphasized by Lamberton and Stephen (2016). Furthermore, businesses of all sizes stand to benefit from the cost-efficient or even cost-free advertising opportunities presented by social media platforms, as underscored by Wang *et al.* (2019). The rapid ascendance of social media platforms has profoundly reshaped business models in both the business-to-business (B2B) and business-to-consumer (B2C) sectors, as indicated by Eric *et al.* (2015). The major goal of this research is to evaluate the impact of social media marketing on customer purchase decisions in Bangladesh, taking into account both financial and strategic aspects. Furthermore, the study investigates the unique mediating function of brand equity within this complex connection and assesses the added value delivered by social media marketing (SMM) to customer purchase decisions. Concurrently, this study is positioned to provide useful theoretical developments as well as concrete insights, effectively assisting businesses in the creation of their marketing strategies. By addressing this existing research vacuum, our work aims to open up previously undiscovered avenues for future researchers to further investigate this topic, therefore contributing to the expanding understanding of how social media influences consumer behavior. Finally, the study's results aim to provide businesses with the knowledge they need to improve their social marketing tactics and influence their future commercial success.

## 2.0 Research Question

The preceding literature discussion leads to the formulation of two research inquiries:

- ❖ What is the comprehensive impact of social media marketing (SMM) on customer purchasing behavior within the context of Bangladesh?
- ❖ How does brand value (BV) function as an intermediary factor in the relationship between social media marketing (SMM) and purchase decisions in Bangladesh?

## 3.0 The objectives of the study

Based on existing research, this initiative aims to:

- ✚ Assess the impact of Social Media Marketing (SMM) on the purchase decision-making process.
- ✚ Investigating the mediating effect of brand value on the link between SMM and purchase decisions.
- ✚ Provide practical ideas and guidance to organizations to enhance the effectiveness of social media platforms.

## 4.0 Literature Review

### 4.1 Social Media Marketing (SMM) and its components

Social media" (SM) encompasses a suite of Web 2.0 technologies enabling users to generate and share their online content (Kaplan & Haenlein, 2010). In the marketing context, social media (SM) represents a digital marketplace where buyers and sellers engage and participate through various means (Hennig-Thurau *et al.*, 2013). Consequently, social media marketing (SMM) encompasses activities such as engaging with both existing and prospective consumers via social media platforms to communicate updates on products or services, pricing adjustments, discounts, fostering personal connections, and offering incentives. In the hospitality industry, customers' decisions regarding their next dining experience are significantly influenced by previous patrons' reviews (Verma *et al.*, 2012).

#### 4.2 Motivations (Mot.)

Advertising incentives, for example, have been recognized as vital strategies for increasing sales while offering valuable advertising (Hossain & Islam, 2019). Discounts, discounts, and freebies, among various types of incentives, impact consumer choices (Martins *et al.*, 2019; Varnali *et al.*, 2012). The willingness of consumers to accept special monetary rewards boosts the success of online social media advertising (Tsang *et al.*, 2004). By utilizing the Ducoffe (1996) model, Kim & Han (2014) introduced incentives as a factor in advertising (Martins *et al.*, 2019). They suggested that increasing incentives for receiving smartphone advertisements can enhance the customer flow experience. Notably, their findings revealed that customers show more interest in tangible benefits and pay greater attention to advertisements that promise monetary gain. Consequently, customers perceive value in ads that incorporate incentives (Martins *et al.*, 2019). Based on the provided information, it can be hypothesized that the presence of motivational factors like incentives in advertisements has the potential to influence customers' decision-making process, and potentially lead to increased purchase intentions.

H<sub>1</sub>: There exists a positive relationship between motivation and purchase decisions.

#### 4.3 Reliability (Rel.)

The suggested study on social media marketing (SMM) takes reliability into account as an independent variable. Ducoffe (1996) underlined the importance of trustworthiness in advertising, including trustworthy, credible, and dependable product information (Wang & Wen, 2017).

Consumer perceptions of information, amusement, and irritation all influence trust in e-commerce, with dependability recognized as the most powerful component in molding attitudes toward advertising (Chowdhury *et al.*, 2010; Gao & Wu, 2010). SMS commercials are frequently disliked by consumers; however, social media ads on sites such as YouTube and Facebook are well received (Chowdhury, 2016; Wang & Wen, 2017).

The dependability of the source has been discovered to be critical to the success of SMS advertising. When reliability was low, SMS advertising was found to be less effective, possibly due to the uncertainty of not having opted-in to receive advertising (Wang & Wen, 2017). Providing an opt-in option was suggested to enhance advertisement effectiveness (Drossos *et al.*, 2007). Youth's attitudes toward SMS advertising were strongly influenced by their perceived credibility (Muzaffar & Kamran, 2011). Users regarded sponsored links as more trustworthy than other advertising characteristics (Lin & Hung, 2009).

Based on the presented information, it can be hypothesized that when advertisements are perceived as credible and trustworthy, they have the potential to positively influence consumers' decision-making process and increase their purchase intentions.

H<sub>2</sub>: There exists a positive relationship between reliability and purchase decisions.

#### 4.4 Informativeness (Info)

The concept of "informativeness" refers to the ability of advertisements to provide customers with relevant information about the advertised products or services. Various factors such as the source, relevance, timing, convenience, and completeness of product information contribute to its usefulness (Ducoffe, 1996; Wang & Wen, 2017). Customers expect advertisements to go beyond product placement and offer substantial information (Gangadharbatla & Daugherty, 2013). Research has shown a positive correlation between consumers' attitudes toward online advertising platforms, including social media (SM), and the level of information provided by digital billboards, suggesting that audience satisfaction is logically linked to informative content (Wang & Wen, 2017; Adetunji *et al.*, 2012). Studies conducted in different cultural contexts have found that Facebook advertising has a

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strong positive influence on consumer attitudes toward Facebook advertising (Alsamydai & Khasawneh, 2013), and consumers' attitudes toward mobile advertising are influenced by the informational content of ads (Blanco *et al.*, 2010). In the case of SMS advertising, its informativeness has been positively correlated with the attitudes of Pakistani youth toward this medium (Muzaffar & Kamran, 2011). Relevant information plays a crucial role in influencing customers' purchasing decisions, and research suggests that highly interactive social media websites increase perceived informativeness, leading to enhanced product and brand likeability and purchase intentions (Ott *et al.*, 2016). However, cultural differences and privacy concerns may impact consumers' perspectives on advertising's ability to inform them, even when their lifestyles are similar (Wang & Wen, 2017). Considering the importance of informativeness and its potential variations across cultures, this proposed study acknowledges informativeness as an independent variable to be examined.

H<sub>3</sub>: There is a positive relationship between informativeness and purchase decisions.

**4.5 Purchase Decision (PD)**

Social media platforms have grown as powerful tools for influencing customer purchase decisions and behavior. Joint decision-making is important because it involves individuals making choices while being affected by others in their social groups, such as family, friends, or coworkers. While conventional media such as advertising, newspapers, and television have historically been used to guide consumer decisions, the emergence of online social networks has changed the scene. Online social networks can influence customer purchasing decisions (Solomon *et al.*, 2010). Consumers have access to a wealth of information and the capacity to share their own product and service experiences in today's digital age. The influence of social media on purchasing decisions can be seen here. People can express their thoughts on products or services they have used in the past.

**5.0 Further extension of the study: Mediating effects of brand value**

To investigate the connection between SMM and purchase decisions, this proposed research makes use of one mediator, as previously stated. In the following sections of this proposal, those variables have been discussed in more detail.

**5.1 Brand Value (BV)**

The concept of "brand community integration" encompasses customers' connections with brands, products, companies, and other customers (McAlexander *et al.*, 2002). When a brand community is well integrated, customer-brand relationships and customer-customer relationships become socially aggregated (Grubb & Grathwohl, 1967; Muniz & O'Guinn, 2001). This integration generates dynamism within the community, leading to added value for both the brand and the customer, ultimately fostering customer loyalty through the perception of association derived from the community (Park & Kim, 2014). Customers' perceptions of a brand are reflected in their brand associations, which are formed based on market information and direct experiences with the brand (Jamal & Goode, 2001). These brand associations play a significant role in influencing buying behavior, as customers' feelings, attitudes, and ideas about the brand strongly impact their purchasing decisions (Steinfeld *et al.*, 2011, Hollenbaugh & Ferris, 2014). Various factors, such as advertising, word-of-mouth, and customer experiences with products and services, can influence customers' purchasing decisions (; Netemeyer *et al.*, 2004). Furthermore, a positive impression of the brand can lead customers to make additional purchases (Porter & Claycomb, 1997). In this context, it is hypothesized that brand value may act as a mediator in the relationship between informational value and purchase decision-making.

Hypothesis 4: A favorable correlation exists between brand value and purchase decisions.

Hypothesis 5: Brand value can serve as a mediating factor in the affirmative correlation between informativeness and purchase decisions.

Hypothesis 6: Brand value can act as a mediating factor in the constructive association between motivation and purchase decisions.

Hypothesis 7: Brand value can function as a mediating element in the positive linkage between reliability and purchase decisions.

## 6.0 Proposed Conceptual Framework

This research study aims to investigate the influence of four distinct factors, namely motivation, reliability, informativeness, and brand value, on purchase decisions in the context of Social Media Marketing (SMM) in Bangladesh. To assess the impact of motivation, reliability, and informativeness on purchase decisions, the study introduces a mediator, specifically 'brand value,' to examine its mediating effect. The relationships between these factors are depicted in the provided Figure, representing the conceptual model of the study.

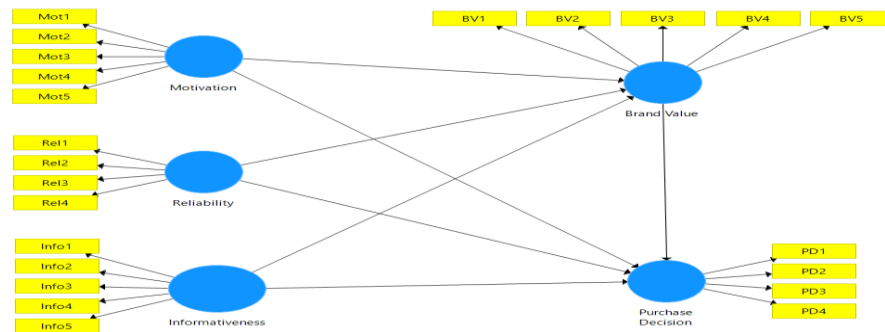


Fig. 1: Conceptual Framework; Source: SmartPLS 3.2.9

## 7.0 Research Methodology

In this study, a blend of primary and secondary data sources was employed. We conducted interviews with a total of 241 individuals residing in diverse regions of the Sirajganj district, ensuring a ratio of ten respondents per questionnaire item. The data collection process was overseen by students from Khwaja Yunus Ali University, who underwent comprehensive training and received close supervision. We utilized both structured and unstructured questionnaires to collect information.

The primary goal of the questionnaire was to evaluate the influence of overall social media marketing (SMM) on purchase decisions. Respondents provided their feedback using a Likert scale ranging from 1 to 5, where 1 signified strong disagreement and 5 indicated strong agreement. To assess the normality of the data, we employed SPSS software. Additionally, we supplemented our primary data with information gleaned from secondary sources, including academic journals.

Convenient sampling was employed to select participants from the pool of potential respondents. The data collected were meticulously prepared and analyzed using Statistical Package for Social Sciences (SPSS) 26.0. For variance-based structural equation modeling (SEM) with the Partial Least Squares (PLS) method, we harnessed SmartPLS-3, a user-friendly graphical interface (GUI) software. This software streamlined the analysis and testing of hypothesized relationships. We also conducted a reliability test based on 20 questionnaire questions using SPSS 26.0, with the acceptability limit determined by the Alpha Coefficient as per Nunnally's (1978) guidelines.

Our data analysis encompassed both descriptive and inferential statistics. To evaluate the validity and reliability of our self-designed questionnaire, we relied on Cronbach's alpha. Furthermore, we investigated the impact of various factors on the dependent variable, the purchase decision, through variance-based structural equation modeling. Notably, we included a mediating factor, brand value, in our analytical framework.

### 7.1 Demographic information

A sum of 241 (n=241) respondents was purposefully chosen from online markets in Bangladesh. Table 1 shows the distribution of sample respondents based on demographic factors included in the questionnaire, such as gender, education level, Age of respondent, and occupation within the sector.

**Table 1:** Demographic information of the respondents

| Name of variables | Category           | Frequency | Percent |
|-------------------|--------------------|-----------|---------|
| Gender            | Male               | 174       | 72.2    |
|                   | Female             | 67        | 27.8    |
| Education         | SSC                | 21        | 8.7     |
|                   | HSC                | 79        | 32.8    |
|                   | Honors             | 92        | 38.2    |
|                   | Master             | 16        | 6.6     |
|                   | Ph.D.              | 33        | 13.7    |
| Occupation        | Businessmen        | 22        | 9.1     |
|                   | Job Holder         | 73        | 30.3    |
|                   | Student            | 98        | 40.7    |
|                   | Others             | 48        | 19.9    |
| Marital Status    | Single             | 198       | 82.2    |
|                   | Married            | 43        | 17.8    |
| Age               | 23-30 years        | 30        | 12.4    |
|                   | 31-38 years        | 123       | 51      |
|                   | 39 years and above | 88        | 36.5    |

Based on the information provided in Table 1, it can be inferred that the study consisted of a predominantly male participant group, with males comprising over 72% of the total respondents. The age distribution of the participants indicates that a majority of 115 individuals (47.7% out of 241) belonged to the age category of 39 years and above. Furthermore, Table 1 reveals that 58.5% of the respondents (141 individuals) had achieved an educational level of Honors or higher. Moreover, more than 40% of the participants were currently enrolled as students. Additionally, the data shows that over 82% of the respondents identified as single.

### 7.2 Univariate Normality Test

As per Hair et al. (2010), data is considered to be normally distributed when skewness falls within the range of -2 to +2 and kurtosis falls within the range of -7 to +7. In our table, both skewness and kurtosis values are within the specified range, indicating that our data follows a normal distribution pattern.

**Table 2: Descriptive Statistics**

|      | N         | Mean      | Std. Deviation | Skewness  |            | Kurtosis  |            |
|------|-----------|-----------|----------------|-----------|------------|-----------|------------|
|      | Statistic | Statistic | Statistic      | Statistic | Std. Error | Statistic | Std. Error |
| Mot1 | 241       | 2.747     | 1.113          | 0.021     | 0.157      | -0.902    | 0.312      |
| Mot2 | 241       | 3.071     | 1.179          | -0.230    | 0.157      | -0.886    | 0.312      |

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|       |     |       |       |        |       |        |       |
|-------|-----|-------|-------|--------|-------|--------|-------|
| Mot3  | 241 | 3.253 | 1.147 | -0.525 | 0.157 | -0.479 | 0.312 |
| Mot4  | 241 | 3.315 | 1.111 | -0.447 | 0.157 | -0.543 | 0.312 |
| Mot5  | 241 | 3.124 | 1.354 | -0.228 | 0.157 | -1.161 | 0.312 |
| PD1   | 241 | 3.610 | 0.860 | -0.187 | 0.157 | -0.013 | 0.312 |
| PD2   | 241 | 3.734 | 0.849 | -0.821 | 0.157 | 1.110  | 0.312 |
| PD3   | 241 | 3.730 | 0.874 | -0.426 | 0.157 | 0.109  | 0.312 |
| PD4   | 241 | 3.631 | 0.983 | -0.582 | 0.157 | 0.013  | 0.312 |
| Rel1  | 241 | 3.515 | 0.975 | -0.422 | 0.157 | -0.164 | 0.312 |
| Rel2  | 241 | 3.535 | 0.908 | -0.358 | 0.157 | -0.428 | 0.312 |
| Rel3  | 241 | 3.456 | 0.912 | -0.235 | 0.157 | -0.255 | 0.312 |
| Rel4  | 241 | 3.361 | 0.948 | -0.216 | 0.157 | -0.340 | 0.312 |
| Info1 | 241 | 3.083 | 1.155 | -0.163 | 0.157 | -0.811 | 0.312 |
| Info2 | 241 | 3.668 | 0.898 | -0.444 | 0.157 | -0.042 | 0.312 |
| Info3 | 241 | 3.643 | 0.925 | -0.601 | 0.157 | 0.170  | 0.312 |
| Info4 | 241 | 3.830 | 0.944 | -0.492 | 0.157 | -0.349 | 0.312 |
| Info5 | 241 | 3.971 | 0.959 | -0.856 | 0.157 | 0.632  | 0.312 |
| BV1   | 241 | 3.913 | 0.902 | -0.962 | 0.157 | 1.288  | 0.312 |
| BV2   | 241 | 3.971 | 0.896 | -0.923 | 0.157 | 0.972  | 0.312 |
| BV3   | 241 | 3.876 | 0.807 | -0.585 | 0.157 | 0.600  | 0.312 |
| BV4   | 241 | 3.743 | 0.983 | -0.525 | 0.157 | -0.038 | 0.312 |
| BV5   | 241 | 3.660 | 0.975 | -0.470 | 0.157 | -0.081 | 0.312 |

**7.3 Multivariate Normality Test**

To identify any potential outliers that may have had a significant impact on our dataset, we performed Cook's distance analysis, as shown in Figure 2. Cook's distance plays a multifaceted role, including the identification of significant data points that have substantial influence. Validity assessment, as well as highlighting areas within the design space where the acquisition of complementary data points would prove advantageous (Cook, 1979; 1977).

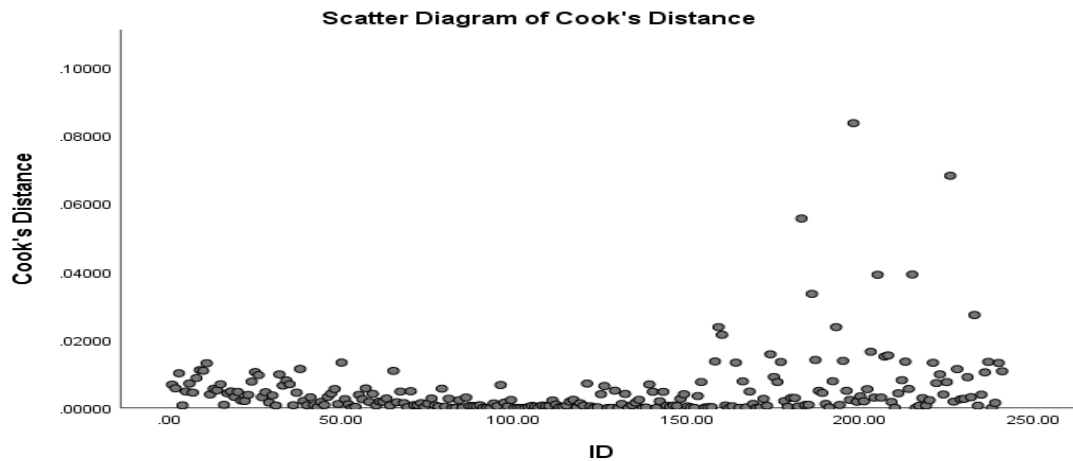


Fig. 2: Multivariate Normality Test

**8.0 Model evaluation**

**8.1 The measurement model**

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Table 3, titled “Validity and Reliability,” provides insights into the convergent validity of our research model. It highlights that all latent variables or model constructs exhibit an average variance extracted (AVE) value greater than 0.50. Specifically, brand value, informativeness, motivation, purchase decision and reliability recorded AVE values of 0.597, 0.532, 0.641, 0.551, and 0.600 respectively. These values unambiguously satisfy the acceptable threshold of 0.50 for AVE across all constructs.

Moreover, each model construct shows composite reliability (CR) values that either meet or exceed the 0.70 benchmarks, registering values of 0.816, 0.772, 0.877, 0.786, and 0.818, respectively. Achieving or exceeding a CR threshold of 0.70 implies strong reliability between constructs.

Table 5 details the convergence validity of the constructs in our model by examining their loading values. According to the guidelines outlined by Sarsted *et al.* (2017), each indicator for the latent variable should exhibit a loading value exceeding 0.70. During this study, four items were excluded from analysis due to low factor loading values ranging from 0.60 to 0.70. Subsequently, we implemented the structural model to identify the main factors influencing purchase decisions about social media marketing in Bangladesh. These results underscore a strong correlation between all construct indicators displayed in Table 3, reinforcing the convergent validity of the research model.

**Table 3:** Exploratory factor analysis with indicator reliability, and model fitting information

| Factor Name       | Indicator                | Factor Loading | SM    | SD    | T Statistics | IR    | CR    | AVE   |
|-------------------|--------------------------|----------------|-------|-------|--------------|-------|-------|-------|
| Brand Value       | BV3 <- Brand Value       | 0.723          | 0.725 | 0.037 | 19.318       | 0.523 | 0.816 | 0.597 |
|                   | BV4 <- Brand Value       | 0.777          | 0.776 | 0.035 | 22.087       | 0.604 |       |       |
|                   | BV5 <- Brand Value       | 0.815          | 0.815 | 0.032 | 25.407       | 0.664 |       |       |
| Informativeness   | Info3 <- Informativeness | 0.759          | 0.758 | 0.044 | 17.229       | 0.576 | 0.772 | 0.532 |
|                   | Info4 <- Informativeness | 0.774          | 0.773 | 0.039 | 19.613       | 0.599 |       |       |
|                   | Info5 <- Informativeness | 0.649          | 0.643 | 0.079 | 8.175        | 0.421 |       |       |
| Motivation        | Mot2 <- Motivation       | 0.792          | 0.789 | 0.033 | 23.874       | 0.627 | 0.877 | 0.641 |
|                   | Mot3 <- Motivation       | 0.826          | 0.825 | 0.025 | 32.822       | 0.682 |       |       |
|                   | Mot4 <- Motivation       | 0.847          | 0.846 | 0.019 | 44.041       | 0.717 |       |       |
|                   | Mot5 <- Motivation       | 0.735          | 0.734 | 0.039 | 18.633       | 0.540 |       |       |
| Purchase Decision | PD1 <- Purchase Decision | 0.749          | 0.742 | 0.058 | 12.863       | 0.561 | 0.786 | 0.551 |
|                   | PD3 <- Purchase Decision | 0.734          | 0.732 | 0.045 | 16.233       | 0.539 |       |       |
|                   | PD4 <- Purchase Decision | 0.744          | 0.744 | 0.041 | 18.128       | 0.554 |       |       |
| Reliability       | Rel2 <- Reliability      | 0.738          | 0.735 | 0.046 | 16.038       | 0.545 | 0.818 | 0.600 |
|                   | Rel3 <- Reliability      | 0.849          | 0.850 | 0.022 | 38.560       | 0.721 |       |       |
|                   | Rel4 <- Reliability      | 0.732          | 0.728 | 0.055 | 13.249       | 0.536 |       |       |

**Table 4: Reliability Indexes and Criteria**



| Reliability Indexes | Criteria     | Reference  |
|---------------------|--------------|--|
| AVE                 | >0.50        | Ringle et al. (2012), Hair et al. (2012), Sarstedt et al. (2017) |
| CR                  | >0.80        | Henseler and Sarstedt (2013)                                     |
| Alpha               | >0.70        | Chin et al. (2008), Henseler and Sarstedt (2013)                 |
| ILV                 | 0.60 to 0.70 | Hair et al. (2012), Ringle et al. (2012), Sarstedt et al. (2017) |

Note. AVE = average variance extracted; CR = composite reliability;  $\alpha$  = Cronbach’s alpha; ILV = Indicator Loading Value.

Table 4 further highlights that all T-statistic values surpassed 2.33 at a 1% level of significance, underscoring the exceptional significance of the outer model loadings. Consequently, based on this data, our structural equation modeling (SEM) model can be deemed valid. In path modeling, a global goodness-of-fit (GoF) measure, calculated as the geometric mean of average commonality and average R2, particularly for endogenous variables, can be employed as recommended by Chin (2010).

$$\text{Formula for calculating GoF} = \sqrt{(\text{AVE} \times R^2)}$$

In this study, the GoF value was computed to be 0.49 (with R2 = 0.316 and average AVE = 0.76). As evident, this GoF value exceeds the highest cutoff value of 0.36, signifying that the proposed model in this research possesses superior explanatory capacity when compared to the suggested GoF thresholds of 0.1, 0.25, and 0.36 (Akbar *et al.*, 2011).

### 8.2 Discriminant Validity

To evaluate the discriminant validity of the model's constructs using the Fornell-Larker criterion (1981), the researchers conducted a comparison between the square root of the Average Variance Extracted (AVE) for each latent variable and their correlations with all other latent variables. According to this criterion, the square root of the AVE coefficients must exceed the correlations. This condition was rigorously verified by scrutinizing both the correlation matrix and the diagonal elements, by the guidelines outlined by Hair *et al.* in their 2012 work.

Table 5 displays the outcomes of the discriminant validity analysis, conclusively revealing that the square root of the AVE values surpasses the correlations with any other constructs or latent variables within the model. This significant finding serves as strong confirmation that the model's constructs indeed demonstrate robust discriminant validity, as delineated in Table 5.

**Table 5: Discriminant Validity**

|                   | Brand Value | Informativeness | Motivation | Purchase Decision | Reliability |
|-------------------|-------------|-----------------|------------|-------------------|-------------|
| Brand Value       | 0.773       |                 |            |                   |             |
| Informativeness   | 0.455       | 0.73            |            |                   |             |
| Motivation        | 0.386       | 0.139           | 0.801      |                   |             |
| Purchase Decision | 0.441       | 0.467           | 0.403      | 0.742             |             |
| Reliability       | 0.504       | 0.302           | 0.381      | 0.354             | 0.775       |

### 8.3 HTMT Test

Table. 6: HTMT Test

|                 | Brand Value | Informativeness | Motivation | Purchase Decision | Reliability |
|-----------------|-------------|-----------------|------------|-------------------|-------------|
| Brand Value     |             |                 |            |                   |             |
| Informativeness | 0.725       |                 |            |                   |             |

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|                   |       |       |       |       |  |
|-------------------|-------|-------|-------|-------|--|
| Motivation        | 0.523 | 0.22  |       |       |  |
| Purchase Decision | 0.696 | 0.772 | 0.568 |       |  |
| Reliability       | 0.745 | 0.475 | 0.517 | 0.549 |  |

**8.4 Common Method Bias Test**

Variance Inflation Factors (VIFs) encompass a numerical range between 1 and 10 or beyond, signifying the degree to which the variance of each coefficient is inflated. The VIF serves as a valuable metric for gauging the extent of inflated variance associated with each coefficient. When interpreting VIF values, a score of 1 signifies the absence of correlation, while values falling between 1 and 5 suggest a moderate level of correlation. VIFs exceeding 5 indicate a high level of correlation, as elucidated by Hair et al. in 1998.

To assess the presence of multicollinearity within the variables, we computed the VIFs and identified that the highest value observed was 1.63, as indicated in Table 5. This value remains well within the recommended threshold set by Hair et al. in 1998, indicating the absence of multicollinearity issues among the factors. Multicollinearity concerns typically arise when VIFs exceed 3.3, suggesting potential pathological collinearity and the presence of common method bias in a model. However, in our model, as displayed in Table 5, all VIF values are either equal to or lower than 3.3, providing assurance that there is no evidence of common method bias, in line with Kock's insights from 2015.

**Table 7: Variance Inflation Factors**

|                   | Brand Value | Informativeness | Motivation | Purchase Decision | Reliability |
|-------------------|-------------|-----------------|------------|-------------------|-------------|
| Brand Value       |             |                 |            | 1.654             |             |
| Informativeness   | 1.101       |                 |            | 1.279             |             |
| Motivation        | 1.171       |                 |            | 1.25              |             |
| Purchase Decision |             |                 |            |                   |             |
| Reliability       | 1.264       |                 |            | 1.435             |             |

**8.5 Structural Model Assessment**

Our assessment of the structural model encompassed a comprehensive examination of path coefficients and the explained variances (R<sup>2</sup> values). This analysis involved a thorough scrutiny of all relationships within the proposed model, encompassing both direct and indirect connections, each examined separately. To derive the coefficients and t-statistics, we employed the bootstrapping technique, conducting 5,000 re-samples. The structural model effectively illustrates the intricate interplay between the dependent and independent constructs, shedding light on their complex relationships.

In our research, we discovered that four independent factors – motivation, reliability, informativeness, and brand value had a positive association with purchase decisions among customers of the online market in Bangladesh. Additionally, we identified a mediator construct, brand value, that also showed a significant relationship with purchase decisions.

Using SmartPLS for variance-based structural equation modeling (SEM), we found that all hypotheses were supported, except for H<sub>2</sub>. Moreover, the corresponding relationships exhibited t-values greater than 1.96 at a significance level of 5%, indicating their statistical significance.

After employing structural equation modeling with the SmartPLS software, we observed that motivation had a significant positive relationship with purchase decision ( $\beta = 0.271$ ,  $t = 5.225$ ,  $p < 0.01$ ), and informativeness also had a significant positive relationship with purchase decision ( $\beta = 0.342$ ,  $t = 6.082$ ,  $p < 0.01$ ). Therefore, our hypotheses

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H<sub>1</sub> and H<sub>3</sub> were supported. However, reliability did not have a significant relationship with Purchase Decision ( $\beta=0.075$ ,  $t=0.989$ ,  $p>0.05$ ). Consequently, our hypothesis H<sub>2</sub> was not supported.

**Table 8: Regression weight**

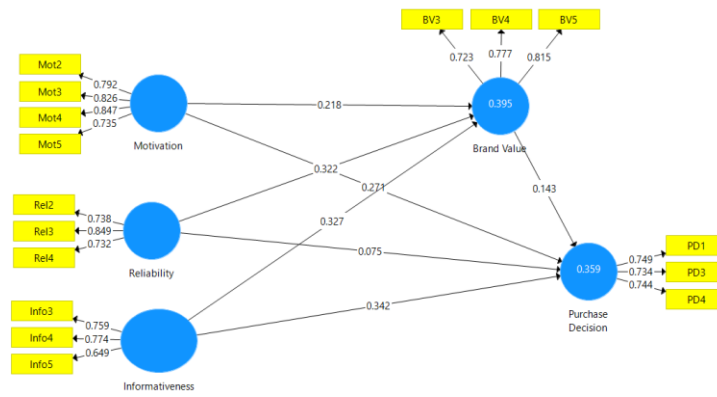
| Hypotheses     | Association                          | Beta     | SM    | SD    | LL     | UL                | T Statistics | P Values | Comment       |
|----------------|--------------------------------------|----------|-------|-------|--------|-------------------|--------------|----------|---------------|
| H <sub>1</sub> | Motivation -> Purchase Decision      | 0.271    | 0.276 | 0.052 | 0.170  | 0.374             | 5.225        | 0.000    | Supported     |
| H <sub>2</sub> | Reliability -> Purchase Decision     | 0.075    | 0.076 | 0.076 | -0.089 | 0.209             | 0.989        | 0.323    | Not Supported |
| H <sub>3</sub> | Informativeness -> Purchase Decision | 0.342    | 0.350 | 0.056 | 0.226  | 0.442             | 6.082        | 0.000    | Supported     |
| H <sub>4</sub> | Brand Value -> Purchase Decision     | 0.143    | 0.141 | 0.068 | 0.009  | 0.266             | 2.095        | 0.037    | Supported     |
|                | Motivation -> Brand Value            | 0.218    | 0.222 | 0.057 | 0.096  | 0.313             | 3.826        | 0.000    | Supported     |
|                | Reliability -> Brand Value           | 0.322    | 0.318 | 0.067 | 0.198  | 0.467             | 4.801        | 0.000    | Supported     |
|                | Informativeness -> Brand Value       | 0.327    | 0.329 | 0.058 | 0.202  | 0.430             | 5.649        | 0.000    | Supported     |
|                |                                      | R Square |       |       |        | R Square Adjusted |              |          |               |
|                | Brand Value                          | 0.395    |       |       |        | 0.388             |              |          |               |
|                | Purchase Decision                    | 0.359    |       |       |        | 0.348             |              |          |               |
|                | SRMR                                 | 0.079    |       |       |        |                   |              |          |               |

Note:  $SRMR \leq 0.08$  (Hu & Bentler, 1998);  $R^2 > 0.30$  (Hair et al., 2011); Collinearity Statistic (VIF)  $< 10$  (Hair et al., 1995).

Cohen (1998) outlined a categorization scheme for R<sup>2</sup> values about endogenous latent variables, designating thresholds of 0.26 (significant), 0.13 (reasonable), and 0.02 (insignificant). As evidenced in Table 6, the coefficient of determination (R<sup>2</sup>) for the dependent variable, Job Happiness, stands at 0.316. This signifies that the four independent variables collectively account for 35.9% of the variance in purchase decisions within the Bangladeshi social media marketing sector, aligning closely with Cohen's (1998) suggested benchmarks.

Furthermore, an assessment of model fit indicators reveals that the Standardized Root Mean Squared Residual (SRMR) is 0.043, falling comfortably below the recommended threshold for a favorable fit to the data (Hu & Bentler, 1999,  $SRMR \leq 0.08$ ). The fit indices presented in Table 5 affirm that the model exhibits a satisfactory fit with the data.

Fig. 3: Path Model



### 8.6 Mediating Effect

Regarding the mediator, the brand value was found to partially mediate the relationship between informativeness and purchase decision ( $\beta= 0.047$ ,  $t\text{-value}= 2.04$ ,  $p<0.05$ ). This is because both the direct effect (Informativeness  $\rightarrow$  Purchase Decision) ( $\beta= 0.342$ ,  $t\text{-value}= 6.082$ ,  $p<0.01$ ) and the indirect effects (Informativeness  $\rightarrow$  Brand Value) ( $\beta= 0.327$ ,  $t\text{-value}= 5.649$ ,  $p<0.01$ ) and (Brand Value  $\rightarrow$  Purchase Decision) ( $\beta= 0.143$ ,  $t\text{-value}= 2.095$ ,  $p<0.05$ ) were significantly related to purchase decision. Therefore, our hypothesis H5 was supported.

Similarly, brand value was found to partially mediate the relationship between motivation and purchase decision ( $\beta= 0.031$ ,  $t\text{-value}= 2.067$ ,  $p<0.05$ ). This is because both the direct effect (Motivation  $\rightarrow$  Purchase Decision) ( $\beta= 0.271$ ,  $t\text{-value}= 5.225$ ,  $p<0.01$ ) and the indirect effects (Motivation  $\rightarrow$  Brand Value) ( $\beta= 0.218$ ,  $t\text{-value}= 3.826$ ,  $p<0.01$ ) and (Brand Value  $\rightarrow$  Purchase Decision) ( $\beta= 0.143$ ,  $t\text{-value}= 2.095$ ,  $p<0.05$ ) were significantly related to purchase decision. Therefore, our hypothesis H6 was supported.

However, when examining the relationship between reliability and purchase decision with the mediating effect of brand value, it was found that brand value does not mediate between reliability and purchase decision ( $\beta= 0.046$ ,  $t\text{-value}= 1.699$ ,  $p>0.05$ ). Thus, our hypothesis H7 was not supported.

**Table 9: Mediating effect analysis**

|                | Hypotheses  | Beta  | SM    | SD    | LL    | UL    | T Statistics | P Values | Comment       | Mediation          |
|----------------|---|-------|-------|-------|-------|-------|--------------|----------|---------------|--------------------|
| H <sub>5</sub> | Informativeness -> Brand Value -> Purchase Decision | 0.047 | 0.046 | 0.023 | 0.005 | 0.095 | 2.040        | 0.042    | Supported     | Partially Mediated |
| H <sub>6</sub> | Motivation -> Brand Value -> Purchase Decision      | 0.031 | 0.031 | 0.015 | 0.003 | 0.072 | 2.067        | 0.048    | Supported     | Partially Mediated |
| H <sub>7</sub> | Reliability -> Brand Value -> Purchase Decision     | 0.046 | 0.046 | 0.027 | 0.004 | 0.113 | 1.699        | 0.090    | Not Supported | Not Mediated       |

### 9.0 Discussion of Results

The findings of this study closely support prior research, adding further credence to established knowledge in the field of social media marketing (SMM). Notably, we identified a significant positive relationship between motivation and purchase decision, seamlessly aligning with the findings of previous investigations conducted by Hossain and Islam (2019), Varnali *et al.* (2012), Kim & Han (2014), and Martins *et al.* (2019).

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Similarly, our study revealed a significant positive relationship between informativeness and purchase decision, which echoes studies conducted by Ducoffe (1996), Wang & Wen (2017), Gangadharbatla & Daugherty (2013), Adetunji et al. (2012), Alsamydai & Khasawneh (2013), Muzaffar & Kamran (2011), and Ott *et al.* (2016).

In the context of mediation, we found that brand value acts as a partial mediator in the relationship between informativeness and purchase decision. Additionally, brand value also acts as a partial mediator of the relationship between motivation and purchase decision. These findings are consistent with previous studies conducted by McAlexander et al. (2002), Grubb & Grathwohl (1967), Muniz & O'Guinn (2001), Park & Kim (2014), Jamal & Goode (2001), Steinfield *et al.* (2011), Park *et al.*, (2009), Hollenbaugh & Ferris (2014), and Netemeyer *et al.* (2004).

**9.1 Conclusions and Recommendations**

In conclusion, this study's findings align closely with prior research, validating existing knowledge in the field of Social Media Marketing (SMM). The significant positive relationship between motivation and purchase decision confirms the impact of motivational factors on consumers' buying behavior. Similarly, the observed significant positive relationship between informativeness and purchase decision emphasizes the importance of providing informative content to influence consumer purchasing decisions.

Furthermore, the study reveals that brand value acts as a partial mediator, influencing the relationship between informativeness and purchase decision as well as motivation and purchase decision. This highlights the role of brand perception and value in shaping consumers' decision-making process.

Based on these findings, it is recommended that businesses and marketers focus on fostering consumer motivation and delivering informative content through social media channels. Additionally, they should invest in building strong brand value and perception, as these factors can enhance the effectiveness of marketing strategies and ultimately influence purchase decisions.

By understanding and leveraging the mediating role of brand value, marketers can craft targeted and impactful campaigns that resonate with consumers. Future research should explore additional factors and variables that may influence the relationship between social media marketing, motivation, informativeness, brand value, and purchase decisions to further enrich our understanding of consumer behavior in the digital landscape.

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**Citation:** Mustafi MAA, Anonthe F, Hossain MJ *et al.*, (2023). The role of information on social media marketing on purchase decision: Examining the mediating role of brand value. *Khwaja Yunus Ali Uni.J*, 6(1):20-34  
<https://doi.org/10.61921/kyauj.v06i01.003>