Driving Brand Engagement Through Online Social Influencers: An Empirical Investigation of Sponsored Blogging Campaigns

Christian Hughes, Vanitha Swaminathan, and Gillian Brooks

Abstract
Influencer marketing is prevalent in firm strategies, yet little is known about the factors that drive success of online brand engagement at different stages of the consumer purchase funnel. The findings suggest that sponsored blogging affects online engagement (e.g., posting comments, liking a brand) differently depending on blogger characteristics and blog post content, which are further moderated by social media platform type and campaign advertising intent. When a sponsored post occurs on a blog, high blogger expertise is more effective when the advertising intent is to raise awareness versus increase trial. However, source expertise fails to drive engagement when the sponsored post occurs on Facebook. When a sponsored post occurs on Facebook, posts high in hedonic content are more effective when the advertising intent is to increase trial versus raise awareness. The effectiveness of campaign incentives depends on the platform type, such that they can increase (decrease) engagement on blogs (Facebook). The empirical evidence for these findings comes from real in-market customer response data and is supplemented with data from an experiment. Taken together, the findings highlight the critical interplay of platform type, campaign intent, source, campaign incentives, and content factors in driving engagement.

Keywords
advertising campaign intent, brand engagement, consumer decision journey, influencer marketing, social media influencers, social media platform, social network sites, sponsored bloggers

Online supplement: https://doi.org/10.1177/0022242919854374

Consumers are increasingly relying on peer-to-peer communications; for this reason, influencer marketing has continued to grow in importance as a key component of firms’ digital marketing strategies (Association of National Advertisers 2018). Nearly 75% of marketers today are using influencers to spread word of mouth (WOM) about their products and brands on social media. Influencer marketing is often considered critical to strengthening online brand engagement (Newberry 2018). Consequently, 65% of multinational brands have indicated plans to increase spending on influencer marketing, with spending expected to reach $10 billion by 2020 (Belton 2019; Mediakix 2018). However, despite the explosion of these social influencers, their effectiveness is still low; for an influencer on Facebook, the average engagement rate per post is .37%; on Twitter, it is even lower at .05% (Rival IQ 2018).

A large and important category of influencer marketing is sponsored blogging, in which companies solicit bloggers to post about specific products and brands (i.e., “sponsored posts”) (Linqia 2017). Bloggers can help generate WOM about a brand, product, or service directly through the content of their sponsored posts. Firms have deployed sponsored blogging both successfully (i.e., Nokia’s camera phone campaign in Finland) and unsuccessfully (i.e., Dr Pepper’s “Raging Cow” campaign) (Corcoran et al. 2006). However, the field needs to develop a better understanding of what drives the success of influencer marketing as a whole and sponsored blogging in particular. Given the significant marketing expenditures dedicated to this

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strategy and the paucity of knowledge on success drivers, this is an important research gap worth addressing.

Sponsored blogging is a hybrid approach combining aspects of paid and earned media (e.g., Colicev et al. 2018; Lovett and Staelin 2016). We distinguish this phenomenon from a purely paid media strategy because influencers engage in WOM and have control over the ultimate message of the advertisement. As companies reimburse bloggers (with either cash or free goods) to generate posts on social media, influencer marketing is distinct from organically generated WOM. Because influencer marketing blends elements of paid and earned media, we can distinguish this from prior research focusing on paid and owned media (e.g., De Vries, Gensler, and Lee 2017; Lovett and Staelin 2016) or earned media, including online WOM (e.g., Hewett et al. 2016). We also extend the traditional advertising literature on the impact of source credibility and message content (Grewal et al. 1997).

We provide a comprehensive framework that examines the drivers of sponsored blogging strategies, including blogger characteristics, content characteristics, and campaign incentives and, in doing so, contribute to the literature in three ways. First, this study advances prior research by examining how social influencers (or sponsored bloggers) can influence consumers at different stages of the consumer purchase funnel by examining different campaign intents (e.g., awareness vs. trial). Second, this research sheds light on the important role of campaign intent as a moderator of the impact of blogger (i.e., expertise) and content (i.e., hedonic value) characteristics on social media engagement. Third, we suggest that the type of social media platform (blogs vs. Facebook) can moderate the impact of these factors on engagement.

Our theoretical basis for predictions derives from the literature on the elaboration likelihood model (ELM) (Petty and Cacioppo 1986). We focus on two moderators that indirectly affect consumers’ ability and motivation to engage in effortful processing. The first involves social media platforms (blogs vs. Facebook), which vary in their level of distraction and involvement, implying differences in consumers’ ability and opportunity to engage in effortful processing. The second is the stage of the consumer decision journey (CDJ) (awareness vs. trial), which may imply increasing levels of motivation closer to trial. Early in the CDJ, consumers process through the peripheral route, whereas later they process through the central route (Colicev et al. 2018). Taken together, we argue that both the platform and the stage of CDJ act as key moderators.

Our findings show that in a blog context, blogger expertise, campaign intent, hedonic value of post, and campaign giveaways are key drivers of engagement. In addition, blogger expertise exerts a greater impact in awareness (vs. trial) campaigns. On Facebook, hedonic value exerts a positive impact, and trial campaigns benefit more from the use of hedonic content. Campaign giveaways exert a negative impact, highlighting the potential cannibalizing role of one platform on another (blog vs. Facebook). Taken together, the findings shed light on various factors that govern how influencer campaigns elicit consumer engagement across multiple platforms. Panel A of Figure 1 presents our conceptual framework on the blogging platform, and Panel B presents the same for the Facebook platform.

This study advances prior research by examining how social influencers can affect consumers at different stages of the consumer purchase funnel. This research suggests that the type of social media platform moderates the impact of social influencers and post characteristics. We develop a framework of strategies based on the social media platform in use and the firm’s campaign intent to inform practitioners about the type of content and influencer to use under each condition. The findings have implications for practitioners who want to employ influencers and show that the choice of bloggers should be guided by campaign intent.

This research uses real in-market customer response data, assembles a large data set of sponsored blogging campaigns, measures various characteristics, and links these to concrete brand engagement outcomes. Thus, our field data provide a unique vantage point and draw a richer picture of not only what constitutes an effective influencer marketing campaign but also how this varies across social media platforms. We supplement the findings by collecting data in a lab study.

**Theoretical Background and Hypotheses**

Research on influencer marketing examines elements of sponsored advertising, product type, source characteristics, and sponsorship disclosure. The findings include differentiating impacts of expertise, product involvement (Zhu and Tan 2007), customer involvement (Fu and Chen 2012), sponsorship disclosure (Van Reijmersdal et al. 2016), and two-sided
messages (Uribe, Buzeta, and Velásquez 2016). Recent research on influencers indicates that information seekers’ objectives and issue involvement drive a blog’s influence (Balabanis and Chatzopoulou 2019). Lou and Yuan (2019) demonstrate the importance of message content, source credibility, and homophily in influencer marketing. Table 1 provides a review of research on the key variables of influencer marketing.

Related research also examines native advertising, or the phenomenon of sponsored social media posts or news articles disguised to resemble nonsponsored content. Wang, Xiong, and Yang (2019) examine native ads and unveil how their effectiveness changes across serial positions by analyzing a large-scale data set. Wu et al. (2016) examine how source credibility plays a critical role in perceptions of native advertising. Other research has also examined the phenomenon of social dollars, or the effects of online social connections on users’ product purchases in an online community. Park et al. (2018) demonstrate that social dollars vary depending on type of product (hedonic vs. functional), user experience, and network density. Together, these findings shed light on the important role of influencers, particularly those embedded in social networks, on consumer choices and purchase behavior.

**Engagement**

Our key dependent variable for the primary field study is social media engagement. We follow Hollebeek (2011, p. 555) and define engagement as a “customer’s cognitive, emotional, and behavioral activities.” More specifically, our focus is on indirect customer engagement, which includes incentivized referrals, social media conversations about products/brands, and customer feedback to companies (Pansari and Kumar 2017). These types of actions contribute to a firm’s revenue, as referred customers are typically more profitable than those not referred (Palmatter, Kumar, and Harmeling 2017; Van den Bulte et al. 2018). This impact of engagement on profitability has also received empirical verification across business-to-business (Kumar, Petersen, and Leone 2010) and business-to-consumer (Lee and Grewal 2004) contexts, and its benefits can derive from both cost reduction and revenue enhancement (Harmeling et al. 2017).

Consumer engagement literature highlights several potential factors that may influence consumer engagement, including emotionality, direct firm actions, and product involvement (Harmeling et al. 2017; Pansari and Kumar 2017). We derive our key factors from this literature and add new factors, such as overall campaign intent, influencer characteristics (i.e., source expertise and post content), and level of involvement elicited by the social media platform. The customer engagement activity we focus on is social media interactions with sponsored influencer content, and we operationalize this as likes and comments on sponsored posts.
**Influencer Marketing on the Blogging Platform**

Firms often launch influencer marketing campaigns on multiple platforms simultaneously. The blog platform constitutes the primary environment for sponsored bloggers to exert their influence. People who choose to interact with bloggers and their postings are typically followers of the blogger. Followers have opted to obtain information posted by bloggers and therefore are likely highly involved in the environment. This high involvement translates into several facets of blog campaigns that help strengthen engagement.

**Platform differences.** While there are various differences across social media platforms, a key difference is the rationale or motivation for consumers to engage with platforms. Some consumers seek out platforms (e.g., blogs) for their content, which implies a higher level of motivation to engage in effortful processing of content. Others may primarily use platforms (e.g., Facebook) to connect with others, implying a focus on relationship maintenance (Karal and Kokoç 2010). Another key difference is the level of distraction prevalent on a platform. Platforms such as Facebook are relatively less involving and more distracting for each individual post because of the large amount of information and content provided (Yahyapour Jalali 2015). In our research, Facebook is a relevant platform (in addition to blogs) because many bloggers post links to their blog posts on Facebook.

To provide a priori evidence of the differences in social media platforms, we pretested blogs and Facebook (the two relevant platforms in this study) using a survey of participants (N = 264, M_age = 35.2 years, 50.0% male) on Amazon Mechanical Turk. Participants were randomly assigned to one of two platform conditions: blog or Facebook. Recalling their last time on the platform, they reported how distracted they felt and the degree to which they were seeking specific content on the platform on scales from 0 to 100. Controlling for age and gender in both regression models, we found that distraction was higher on Facebook than blogs (F(3, 260) = 7.22, p < .01; M_blog = 32.65, M_Facebook = 42.83; b_platform = -10.67, p < .01), and specific content seeking was higher on blogs than Facebook (F(3, 260) = 3.61, p < .01; M_blog = 59.19, M_Facebook = 48.14; b_platform = 11.25, p < .01). These results lend support to our argument that platform distraction and content search differ between blogs and Facebook, with distraction being lower and content seeking being more common on blogs. Therefore, Facebook should result in low-involvement processing of information.

Given the low-involvement nature of the Facebook platform, consistent with the ELM (Petty and Cacioppo 1986), there should be a greater emphasis on peripheral cues (e.g., number of followers, hedonic content, timing and number of posts). Conversely, in line with ELM predictions, in high-involvement platforms such as blogs, argument quality should exert a greater impact on persuasion; this implies that source expertise and post content should play important roles in eliciting engagement on blog platforms. We articulate these differences in our separate predictions we develop for blogs and Facebook platforms.

**Campaign intent on the blogging platform.** Broadly, influencer marketing campaigns have two goals: (1) to increase awareness and (2) to encourage trial. From a marketer's perspective, awareness campaigns are an easier-to-achieve goal and do not require any overt action on the part of consumers. Trial campaigns, which encourage consumers to make a purchase, are typically linked to consumer actions (e.g., purchase, app download) and therefore have a more overt persuasion intent and also a higher hurdle to generate customer engagement. These advertising goals (awareness vs. trial) can also affect the activation of persuasion knowledge of consumers, depending on whether there is a more direct advertising motive, as in the case of a trial campaign, or a less direct advertising motive, as in the case of an awareness campaign.

These campaign intents align with the beginning and end of the consumer's decision journey, which typically involves multiple stages in a hierarchy of effects, such as awareness, knowledge, liking, preference, conviction, and purchase. Prior research has examined this dichotomy of awareness versus trial intent in a traditional advertising context (e.g., Muller 1983). As noted previously, the processing route to persuasion differs depending on the stage of the CDJ, with early stages being processed through the peripheral route and later stages being processed through the central route (Colicev et al. 2018). We propose that campaign intent is a potential moderator that can influence engagement differently depending on the stage of the CDJ. We predict that trial intent versus awareness will have a greater impact on the blogging platform.

**Main effect of blogger expertise.** Source expertise refers to the level of credibility a source possesses. Expertise reflects the extent to which a consumer is qualified to discuss a subject (Alba and Hutchinson 1987), such as source qualifications (Berlo, Lemert, and Mertz 1969), competence, knowledge, education, expertise, and ability to share knowledge (Hinkin and Schriesheim 1989). This can derive from informational power, in which the expert has knowledge that others do not have (Deutsch and Gerard 1955). Expert power can be knowledge within a specific domain (e.g., law) (French and Raven 1959). Endorsers are more likely to be considered experts if they are competent and have relevant knowledge (Homer and Kahle 1990).

Source expertise affects attitude change (Hovland and Weiss 1951; McCracken 1989; Ohanian 1991), level of confidence and positivity (Tormala, Brinol, and Petty 2007), and behavioral changes (Crisci and Kassinove 1973) and leads to higher levels of persuasion (Petty and Wegener 1998). Higher levels of persuasion are a result of high source expertise leading to a deeper processing of the advertising message (Homer and Kahle 1990). In an influencer marketing context, expertise increases behavioral intention toward products (Uribe, Buzeta, and Velásquez 2016). In a sponsored blogging context, consumers will prefer products endorsed or referred by a blogger.
with expertise because they perceive the message as more persuasive and credible (Kiecker and Cowles 2002; Zhu and Tan 2007). Thus,

**H1:** Blogger expertise has a positive impact on blog engagement. The higher the blogger’s expertise, the higher is the number of blog post comments.

**Interaction effect of campaign intent on blogger expertise.** Both high- and low-expertise bloggers may be considered influential, under varying circumstances. Despite the expected positive impact of blogger expertise on engagement in a sponsored blogging context, source expertise can also have a neutral (or even negative) effect in some situations. Prior research suggests that in the presence of an extreme advertising claim, the positive impact of source expertise diminishes (Goldberg and Hartwick 1990). Depending on the context, type of claim, and stage in the decision-making process, source expertise may even have a nonsignificant (or negative impact) on engagement. The nonsignificant impact of source expertise also stems from the countervailing positive impact of low-expertise bloggers (novice endorsers). Novice endorsements can be as effective as those from experts (Wang 2005). Therefore, we expect blogger expertise to vary in its impact depending on campaign intent.

Involvement affects blogger success, such that for low-involvement products, a blogger with low expertise can have greater success (Zhu and Tan 2007). Under different stages of the CDJ, we similarly anticipate blogger expertise to affect engagement differently depending on the level of involvement. The peripheral processing that occurs early in the CDJ gives more attention to peripheral cues, such as expertise. In the early stages of the CDJ, expertise becomes a more important influence for persuasion than homophily (Wang et al. 2008). Regarding an awareness intent, early in the CDJ we expect high expertise to be beneficial. For trial campaigns that correspond to later in the CDJ process and lead to higher motivation for central processing, we predict the opposite effect. Regarding a behavioral versus attitudinal change, low-expertise (vs. high-expertise) sources can be more effective (Dholakia and Schwartz’s 2012) suggest that specific emotions (e.g., awe, anxiety) trigger arousal, which in turn results in greater virality of online content. Orders et al. (2019) extend these findings and argue that consumers share expressive or assertive brand messages more frequently than directive brand messages. Relatedly, Herhausen et al. (2019) indicate that hedonic content can be a key factor in the virality potential of online firestorms. Building on these findings, we expect a post featuring high hedonic value content to increase arousal, deepening customer engagement. Therefore, we posit a general positive impact of hedonic content on the blogging platform.

**H3:** Post content, in terms of hedonic value, is positively related to engagement in blog post comments.

**Campaign incentives.** Campaign incentives are marketing actions designed to elicit specific responses and engagement from consumers. The purpose of a campaign incentive is “to give followers a free item (or a chance to win a free item) in exchange for them sharing, liking, following, and/or reposting a picture” (Nilo 2017). For example, Rafflecopter is a giveaway platform widely used by sponsored bloggers, and the requirements to enter each giveaway are at bloggers’ discretion. For some campaign giveaways, bloggers require consumers to comment on the blog post itself, while others require a different action (e.g., become a Twitter follower, share the giveaway) to enter the giveaway.

Campaign incentives are a direct firm action to increase customer engagement (Verhoef, Reinartz, and Krafft 2010). Other benefits of giveaways include an increased desire to buy more, higher-quality perceptions of the product, and increased WOM about the product (White 2013). In Berger and Schwartz’ s (2011) study, consumers who received a free product talked about it 20% more than those who did not receive the product for free. Therefore, we expect the presence of incentives to increase blog engagement because they elicit consumer comments for a chance to win the giveaway.

**H4:** Campaign incentives are positively related to engagement, such that inclusion of a campaign incentive leads to more blog post comments.

**Influencer Marketing on the Facebook Platform**

As mentioned, influencer marketing campaigns co-occur across platforms. The Facebook platform is a secondary environment for sponsored bloggers to link to the posts on their
respective blog pages. In contrast with the blogging platform, people who encounter the Facebook post may or may not follow the blogger’s blogging page. In other words, followers could come across the Facebook posts because they follow the person or because a friend has shared or interacted with the post. The Facebook platform is one example of a low-involvement, high-distraction social media platform. This lower involvement primes a different set of important features while diminishing the importance of others.

**Campaign intent on the Facebook platform.** As noted previously, we expect awareness and trial campaign intent to align with the CDJ. Engagement generated by sponsored posts may vary depending on campaign intent. In trial campaigns, we expect Facebook participants’ willingness to engage in campaigns with overtly commercial intent to be low. Karal and Kokoç (2010) propose that the primary motivations for Facebook usage are to gain knowledge, acquire new connections, and strengthen existing relationships. An overtly commercial intent, as in the case of trial, can interfere with the intended usage of the platform and therefore be met with resentment by users. Because gaining knowledge and encountering ideas and information are reasons to use Facebook, these are more in line with the awareness intent. The awareness intent is more of a helping motive associated with WOM communications in the network. Facebook users may spread WOM about awareness campaigns because doing so generates positive feelings and strengthens social connections (Hennig-Thurau et al. 2004). Thus,

H5: Campaign intent has a positive impact on engagement on the Facebook platform. Specifically, awareness (vs. trial) campaigns generate more Facebook engagement (i.e., likes).

**Hedonic value on the Facebook platform.** Evidence shows that engagement on Facebook is positively related to hedonic content (Chiu et al. 2007). The primary rationale for this is that the hedonic value generates an emotional response (Dobele et al. 2007), which leads to higher arousal and a greater propensity to like and share in online settings (Berger and Milkman 2012; Fiore, Jin, and Kim 2005). Research based on the ELM (Petty and Cacioppo 1986) indicates that when consumers are less involved with products, they use peripheral routes to process information (Fu and Chen 2012). On a low-involvement platform, we expect hedonic value to be more salient to the reader. Under low involvement, Cho (1999) finds that attention-getting online advertising appeals were more effective. Consequently, we predict that hedonic content associated with blog posts is highly relevant to low-involvement platforms such as Facebook, as it helps overcome low involvement by raising the interestingness of a post. In support of this idea, the in-store shopping literature (e.g., Babin and Attaway 2000) shows that the hedonic value of a shopping experience plays a key role in elevating involvement and inducing purchase behavior. For a low-involvement, high-distraction platform such as Facebook, peripheral cues, such as hedonic value, should be important. Thus,

**H6:** The hedonic value of blog posts has a positive impact on Facebook engagement (i.e., likes).

**Interaction effect of campaign intent and hedonic value on Facebook platform.** In addition to the preceding main effect predictions, we anticipate an interaction effect of campaign intent and hedonic value on Facebook engagement. Hedonic content leads to a greater likelihood of message forwarding (Chiu et al. 2007) and increased private sharing of news articles (Berger and Milkman 2012), but these links are dependent on the context of sharing and type of outcome being studied (Tucker 2014). Ads viewed as too “outrageous” may result in lower purchase intent and persuasion, but this is not the case when the ad also leads to consumer responses such as comments (Tucker 2014).

As the Facebook platform is a low-involvement, high-distraction environment, in which people are predisposed toward information overload (Koroleva, Krasnova, and Günther 2011), opportunity and motivation to process become key components in information processing (MacInnis, Moorman, and Jaworski 1991). A sufficiency threshold dictates that the amount of processing a person is willing to undertake is dependent on the perceived risk involved (Chen, Duckworth, and Chaiken 1999; Maheswaran and Chaiken 1991). Under a trial intent, in which the perceived risk would be higher than in an awareness context, we anticipate a high sufficiency threshold and, thus, central processing.

In addition, a consumer’s motivation to process an ad will be higher when there are more hedonic cues (MacInnis, Moorman, and Jaworski 1991). Boerman, Willemsen, and Van Der Aa (2017) posit that when people identify a Facebook post as an advertisement, they develop feelings of distrust, which may imply a threat to the relationship. In this case, a highly hedonic post can alleviate this threat. Heuristic processing is likely to occur when the person views the information as agreeing with his or her beliefs (Giner-Sorolila and Chaiken 1997). Taken together, this would imply that for a trial intent, a post high in hedonic value could overcome the disposition to process more systematically. This suggests that when campaigns involve purchasing (e.g., trial), the hedonic value of the post will be valuable. Thus,

**H7:** High (vs. low) hedonic value has a more positive impact on trial campaigns than awareness campaigns.

**Study 1**

**Data**

The data come from The Motherhood, a leading agency for sponsored blogging campaigns that focuses on “mommy” bloggers. The data consist of 1,830 sponsored posts written by 595
bloggers, collected from September 2012 to December 2016. These blog posts came from 57 different campaigns, including Awesome Avocados, Banner Alzheimer, Chef Boyardee Little Chef, Latte Love, and Barnes & Noble.

For each campaign, companies work with the blogging agency to coordinate campaign details, such as the intended message, target audience, and goals. Bloggers are recommended for the campaign on the basis of their demographics, age of children, and expertise, and they can choose to work on projects in line with their interests, availability, and willingness to work within the set budget. Bloggers are required to disclose that they are sponsored bloggers either at the beginning or at the end of the blog posts. Depending on the budget and requirements of a given campaign, each blogger receives compensation (in the form of either money or free products). Bloggers then write and post content on their own blog websites about the campaign; bloggers get paid more if they post something on multiple social media channels.

### Table 2. Constructs and Measures.

<table>
<thead>
<tr>
<th>Construct</th>
<th>Definition</th>
<th>Operationalization</th>
</tr>
</thead>
<tbody>
<tr>
<td>Facebook post likes</td>
<td>Primary measure of Facebook engagement is the number of likes each post received.</td>
<td>Count; the total number of likes per blogger per campaign</td>
</tr>
<tr>
<td>Blog post comments</td>
<td>Primary measure of engagement is the number of blog post comments each blog post received.</td>
<td>Count; the total number of comments per blog post, blogger, and campaign</td>
</tr>
<tr>
<td>Facebook posts</td>
<td>Number of Facebook posts per blogger per campaign.</td>
<td>Count; control variable</td>
</tr>
<tr>
<td>Followers</td>
<td>Represents blogger’s social media presence and is also an indication of blogger strength.</td>
<td>Quantitative; the average number of twitter and Facebook followers that a blogger has in online network</td>
</tr>
<tr>
<td>Awareness campaign</td>
<td>Increases brand awareness and spreads information to consumers; occurs at an early stage in the purchase funnel because consumers are not yet trying to evaluate whether to purchase the product.</td>
<td>Categorical; campaign intent is focused on raising awareness about a specific brand</td>
</tr>
<tr>
<td>Trial campaign</td>
<td>Encourages consumers to make a purchase; typically linked to actions required of consumers (e.g., purchase).</td>
<td>Categorical; campaign intent is focused on increasing purchase or trial behavior</td>
</tr>
<tr>
<td>Expertise</td>
<td>Is indicative of how bloggers portray themselves as a source of information as a sponsored blogger.</td>
<td>Quantitative; a sum of the person’s educational affiliation and blogger credentials. Range: 0–2</td>
</tr>
<tr>
<td>Functional</td>
<td>Functional value captures the believability and informativeness of a post.</td>
<td>Quantitative; a factor score of content that is genuine/sincere, honest, informative, pleasant, relatable, understandable, believable, and relevant, as well as usage consideration</td>
</tr>
<tr>
<td>Hedonic</td>
<td>Hedonic value of a post refers to the enjoyment, emotions, and entertainment a consumer experiences from reading a post.</td>
<td>Quantitative; a factor score of content that is attention getting, creative, emotional, energetic, humorous, memorable, strong, unique, and warmhearted</td>
</tr>
<tr>
<td>Giveaways</td>
<td>Marketing actions designed to generate specific responses and engagement from consumers.</td>
<td>Categorical; campaign-level variable, whether or not a giveaway was included as part of the campaign</td>
</tr>
</tbody>
</table>

### Table 3. Descriptive Statistics.

<table>
<thead>
<tr>
<th>Variable</th>
<th>N</th>
<th>M</th>
<th>SD</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of Facebook post likes</td>
<td>1,398</td>
<td>17.53</td>
<td>100.64</td>
</tr>
<tr>
<td>Number of blog post comments</td>
<td>1,826</td>
<td>21.23</td>
<td>70.02</td>
</tr>
<tr>
<td>Average number of followers</td>
<td>1,267</td>
<td>21,246.10</td>
<td>24,812.43</td>
</tr>
<tr>
<td>Weekend post</td>
<td>1,830</td>
<td>14.8%</td>
<td>.40</td>
</tr>
<tr>
<td>Type of campaign: awareness</td>
<td>1,830</td>
<td>35.1%</td>
<td>.48</td>
</tr>
<tr>
<td>Type of campaign: trial</td>
<td>1,830</td>
<td>64.9%</td>
<td>.48</td>
</tr>
<tr>
<td>Expertise (sum of credentials and education)</td>
<td>1,816</td>
<td>.36</td>
<td>.63</td>
</tr>
<tr>
<td>Blogger travel/foodie</td>
<td>1,816</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Blogger persona</td>
<td>1,816</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Blogger lifestyle</td>
<td>1,816</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Blogger values</td>
<td>1,816</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Functional value of post</td>
<td>1,830</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Hedonic value of post</td>
<td>1,830</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Giveaway</td>
<td>1,825</td>
<td>25.5%</td>
<td>.44</td>
</tr>
<tr>
<td>Inverse Mills ratio</td>
<td>1,819</td>
<td>.27</td>
<td>.18</td>
</tr>
</tbody>
</table>

*Percentage of occurrences.

### Dependent Variables

**Number of blog post comments.** Every blog post had an option for blog readers to leave comments. Comments of other readers are visible to any subsequent reader of the blog, but other readers do not receive notification when a new comment has been posted. Our primary measure of engagement is the number of comments each blog post received (see Table 2 for constructs, Table 3 for descriptive statistics of the variables, and Table 4 for the correlations).
Number of Facebook post likes. Bloggers frequently used their social media outlets to post about different blog campaigns. Facebook followers of a blogger can see the new post in their Facebook news feed, while others need to seek out the post directly from the blogger’s Facebook profile. To measure Facebook engagement, we counted the number of Facebook post likes.

Independent Variables

Campaign intent. Companies typically divide campaigns into two categories: those designed to raise awareness and those designed to increase trial. In our data set, of the 57 total campaigns, 29 had an awareness intent and 28 had a trial intent. The awareness campaigns focused on increasing brand recognition. For example, an AT&T Mobile School Safety campaign encouraged people to talk to their children about using mobile phones safely. These campaigns were not directly trying to motivate people to make a purchase but instead were focused on building brand awareness. By contrast, the trial campaigns focused on increasing consumer trial for products or services. Examples of this type of campaign included Church Hill Classics’ diploma frames and Veritas Genetics’ at-home BRCA test.

Campaign incentives. We identified campaigns (29%) in terms of whether they included a campaign incentive (i.e., a giveaway). Giveaways typically request that readers like or share a blog post to be entered for a chance to win a prize. For example, Johnson & Johnson’s Donate-a-Photo campaign had a giveaway prize of $100 worth of products.

Blogger average number of followers. The average number of Twitter and Facebook followers represents a blogger’s social media presence and is also an indication of blogger strength. We used bloggers’ Twitter and Facebook followers for two reasons: (1) Facebook and Twitter are two of the largest social media platforms, and (2) the number of followers on the blog web pages themselves are unavailable. We use the natural log of the average number of Twitter and Facebook followers in the models to account for the large spread, and we mean-centered them improve interpretation of coefficients. We also use alternative operationalizations and reestimate the main model using these measures (for the results, see the Web Appendix).

Blogger psychographic profile factors. First, we pulled the public profiles of each blogger in our data set, as described on their blog pages. Second, three coders (blind to the hypotheses) examined the bloggers’ public profiles and listed key themes that captured their psychographic profiles (i.e., interests, activities, and opinions; see the Web Appendix). This procedure revealed 14 main psychographic profile dimensions, dummy coded as 1 if present in the profile and 0 if not. Third, we used

<table>
<thead>
<tr>
<th>Table 4. Variable Correlations.</th>
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<tbody>
<tr>
<td>Variable</td>
</tr>
<tr>
<td>No. of Facebook likes</td>
</tr>
<tr>
<td>No. of blog comments</td>
</tr>
<tr>
<td>Avg. no. of Facebook followers</td>
</tr>
<tr>
<td>No. of Weekend post</td>
</tr>
<tr>
<td>No. of Facebook Posts</td>
</tr>
<tr>
<td>Awareness Campaign</td>
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<tr>
<td>Blogger Expertise</td>
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<tr>
<td>Blogger Travel/Foodie</td>
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<tr>
<td>Blogger Persona</td>
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<td>Blogger Lifestyle</td>
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<td>Blogger Values</td>
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<td>Functional Value</td>
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<tr>
<td>Hedonic Value</td>
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<tr>
<td>Giveaway</td>
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</tbody>
</table>

*p < .05.

Notes: The unit of analysis is the blog post.
factor analysis with varimax rotation\(^3\) to identify five overarching characteristics of bloggers.\(^4\)

**Blogger expertise.** We measure blogger expertise by the presence of the person’s educational affiliation and blogger credentials in his or her profile. Prior research has also used blogger profiles to manipulate source (Uribe, Buzeta, and Velásquez 2016). An educational affiliation includes reference to a specific higher education degree (e.g., “Bachelor of Arts”), while blogger credentials refer specifically to status as a credible blogger (e.g., “social media consultant,” “Nielsen 50 Power Mom”). Blogger expertise, which ranges from 0 to 2, is the sum of the two measures.

### Post Variables

**Weekend post.** Weekend post is an indicator variable for whether the post occurred on a weekend, coded as 1, or a weekday, coded as 0. We used this to capture weekend versus weekday seasonality. We incorporate this as a control variable for the temporal element.

**Number of Facebook posts.** The number of Facebook posts serves as a control variable. For sponsored blogging campaigns, bloggers will post on blogs and then post on Facebook linking to the blog post. To control for the number of Facebook posts, we include this as a variable in the model.

**Hedonic and informational value associated with the blog post.** Three coders classified the hedonic and informational value associated with a given blog post; we prequalified the coders to match the demographics of the bloggers’ audiences. We based our measures on Yuvaraju, Subramanyam, and Rao (2014), who develop a 20-item emotion scale for advertisements. We used coders from Amazon Mechanical Turk to measure various aspects of sentiment on a seven-point scale (1 = “not at all,” and 7 = “extremely”), including how much the blog posting was attention getting as well as how boring, creative, emotional, energetic, genuine/sincere, honest, humorous, informative, irritating, memorable, pleasant, strong, unique, warmhearted, relatable, understandable, believable, and relevant the post was. We selected coders who were as similar to the blog audience as possible (i.e., they were also mothers with a child under 18 years in the household). We solicited three coders for each blog post and asked them to code only a subset of blog posts (typically three posts each), suggesting that there are variations introduced across different blog posts from the varying identities of coders.

<table>
<thead>
<tr>
<th>Sentiment Variables</th>
<th>Factor 1</th>
<th>Factor 2</th>
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</thead>
<tbody>
<tr>
<td>Attention getting</td>
<td>.44</td>
<td>.66</td>
</tr>
<tr>
<td>Boring</td>
<td>-.47</td>
<td>-.51</td>
</tr>
<tr>
<td>Creative</td>
<td>.33</td>
<td>.78</td>
</tr>
<tr>
<td>Emotional</td>
<td>.17</td>
<td>.71</td>
</tr>
<tr>
<td>Energetic</td>
<td>.35</td>
<td>.70</td>
</tr>
<tr>
<td>Genuine/sincere</td>
<td>.69</td>
<td>.49</td>
</tr>
<tr>
<td>Honest</td>
<td>.73</td>
<td>.43</td>
</tr>
<tr>
<td>Humorous</td>
<td>-.08</td>
<td>.68</td>
</tr>
<tr>
<td>Informative</td>
<td>.66</td>
<td>.34</td>
</tr>
<tr>
<td>Irritating</td>
<td>-.65</td>
<td>-.17</td>
</tr>
<tr>
<td>Memorable</td>
<td>.37</td>
<td>.76</td>
</tr>
<tr>
<td>Pleasant</td>
<td>.61</td>
<td>.56</td>
</tr>
<tr>
<td>Strong</td>
<td>.37</td>
<td>.74</td>
</tr>
<tr>
<td>Unique</td>
<td>.28</td>
<td>.79</td>
</tr>
<tr>
<td>Warmhearted</td>
<td>.53</td>
<td>.64</td>
</tr>
<tr>
<td>Relatable</td>
<td>.66</td>
<td>.48</td>
</tr>
<tr>
<td>Understandable</td>
<td>.75</td>
<td>.07</td>
</tr>
<tr>
<td>Post believable</td>
<td>.84</td>
<td>.17</td>
</tr>
<tr>
<td>Relevant</td>
<td>.67</td>
<td>.27</td>
</tr>
<tr>
<td>Benefits believable</td>
<td>.85</td>
<td>.18</td>
</tr>
<tr>
<td>Consider using</td>
<td>.68</td>
<td>.37</td>
</tr>
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</table>

First, to measure the agreement between coders and calculate a more accurate alpha score, we used the methodology Shrout and Fleiss (1979) describe and computed the Shrout–Fleiss single intraclass correlation score agreement of .998, which is considered quite high (Koo and Li 2016). As a second approach to assess reliability, we estimated a standardized alpha within three coders for each sentiment value for each blog post, to account for the different coders on each post. We then averaged these standardized alphas and obtained an average reliability of .51 and a median reliability of .56.

Each blog post was coded for a variety of sentiment variables, some of which may be correlated. To reduce the dimensionality of the data and increase parsimony, we conducted a factor analysis. Factor analysis with varimax rotation revealed two factors with eigenvalues greater than 1 (for factor loadings, see Table 5), which we labeled as “functional” and “hedonic.” The variables that loaded most highly on perceived functional value were genuine/sincere, honest, informative, pleasant, relatable, understandable, believable, relevant, and benefits believable. The variables that loaded most highly on perceived hedonic value were attention getting, creative, emotional, energetic, humorous, memorable, strong, unique, and warmhearted.

### Selection Model

Because bloggers are chosen to participate in campaigns, selection bias may occur. To address this potential problem, we implemented a Heckman (1979) selection model. The first-stage model used a probit regression to predict a blogger’s selection for a campaign. To achieve identification, the set of
covariates in the Stage 1 probit model must contain at least one variable that can supply the exclusion restriction—that is, it must affect blogger selection for a campaign but not directly affect the engagement generated by the blogger (Gill, Sridhar, and Grewal 2017; Heckman 1979). Industry practice is to match bloggers with common interests in and similarity to the focal campaign. In line with this method, we use the bloggers’ profile descriptions and employed varimax factor rotation to create a psychographic index score using psychographic categories not directly related to the outcome of engagement: travel/foodie, persona, lifestyle, and values.

The variable providing the exclusion restriction used in the Stage 1 probit model is blogger selection of most similar other bloggers. To determine the blogger most similar to the target blogger, we created a blogger-by-campaign matrix. We multiplied this matrix by its transpose to create a blogger-by-blogger matrix, which showed which bloggers coappeared (i.e., participated in the same campaign) most frequently. We selected the blogger who appeared most often with the target blogger and used his or her selection for a campaign as an independent variable in the Stage 1 probit (for details of this procedure and a specific example, see “Details on Stage 1 Probit Model” in the Web Appendix).

Table 6 provides the results of the Stage 1 probit model. We find that the intercept (b = −2.2744, p < .01), similar blogger selection (b = 1.5550, p < .01), and the travel/foodie blogger psychographic (b = .0374, p < .01) are significant for selection in the Stage 1 probit model. The exclusion criterion, similar to the blogger selection, indicates that when a similar blogger to the target blogger is selected for a campaign, the target blogger is more likely to be selected for the campaign. We then included the inverse Mills ratio from the first stage as an independent variable in all second-stage models. The inverse Mills ratio was not significant in the blog post comments (z = 1.22, p = .223) or Facebook post likes (z = −.060, p = .950) models.

### Model Choice

The dependent variables of interest, blog post comments and Facebook post likes, are count variables. Therefore, we considered using either a Poisson distribution or a negative binomial distribution for count data. A likelihood ratio test indicated that there was overdispersion in the data for the blog post comments ($\chi^2 = 6,231, p < .001$) and Facebook post likes ($\chi^2 = 31,000, p < .001$) models. Thus, we used a negative binomial model instead of a Poisson model. In addition, we find no correlation between the error terms of the two models.

For each post i, we estimated the following second-stage model equations:

- **Blog post comments**
  
  \[ \text{Blog post comments} = \beta_0 + \beta_1(\text{Number of Followers}) + \beta_2(\text{Weekend Post}) + \beta_3(\text{Awareness Campaign}) + \beta_4(\text{Blogger Expertise}) + \beta_5(\text{Functional Value of Post}) + \beta_6(\text{Hedonic Value of Post}) + \beta_7(\text{Giveaway}) + \beta_8(\text{Awareness Campaign} \times \text{Number of Followers}) + \beta_9(\text{Awareness Campaign} \times \text{Functional Value of Post}) + \beta_{10}(\text{Awareness Campaign} \times \text{Hedonic Value of Post}) + \beta_{11}(\text{Awareness Campaign} \times \text{Blogger Expertise}) + \beta_{12}(\text{Awareness Campaign} \times \text{Giveaway}) + \beta_{13}(\text{Inverse Mills Ratio}) + \epsilon. \]

- **Facebook post likes**
  
  \[ \text{Facebook post likes} = \beta_0 + \beta_1(\text{Number of Followers}) + \beta_2(\text{Weekend Post}) + \beta_3(\text{Number of Facebook Posts}) + \beta_4(\text{Awareness Campaign}) + \beta_5(\text{Blogger Expertise}) + \beta_6(\text{Functional Value of Post}) + \beta_7(\text{Hedonic Value of Post}) + \beta_8(\text{Giveaway}) + \beta_9(\text{Awareness Campaign} \times \text{Number of Followers}) + \beta_{10}(\text{Awareness Campaign} \times \text{Functional Value of Post}) + \beta_{11}(\text{Awareness Campaign} \times \text{Hedonic Value of Post}) + \beta_{12}(\text{Awareness Campaign} \times \text{Blogger Expertise}) + \beta_{13}(\text{Awareness Campaign} \times \text{Giveaway}) + \beta_{14}(\text{Inverse Mills Ratio}) + \epsilon. \]

### Model Results

Before we describe our full model results (see Table 7), several main effects results are worth noting. In the main-effects-only model (see the Web Appendix), we find that campaign intent exerts a differential main effect on each platform, with awareness intent being more effective for Facebook and trial intent being more effective for blogs. We conjecture that because of Facebook’s lower commercial intent, an awareness campaign is potentially more readily shared among peers in an organic fashion. The purpose of campaign incentives (i.e., giveaways) is to encourage participation with specific tasks. The negative impact of incentives on Facebook and the positive impact on the blog platform highlight the potential cannibalizing effect of one social media platform on another.

**Blog post comments model.** Table 7 reports the results of the second-stage model with blog post comments as the dependent variable (N = 1,237). The Akaike information criterion for this model was 6,663, the Bayesian information criterion was 6,740, and the likelihood ratio test was significant ($\chi^2(13) = 100.64, p < .01$). Variance inflation factors (VIFs) were all
below 1.09, with an average VIF of 1.04, indicating no issues with collinearity. We found a significant main effect of our control variable, average number of followers ($b = .3514, p < .01$), indicating that this factor significantly drives the number of blog post comments.

The main effect of campaign intent (awareness vs. trial) was marginally significant in the blog post comments model ($b = -.2351, p < .10$), while the main effect of blogger expertise was not significant in the final model incorporating interaction effects ($b = -.1228, n.s.$), which does not support $H_1$. The interaction between type of campaign (awareness vs. trial) and blogger expertise was positive in the blog post comments model ($b = .7283, p = .01$), in support of $H_2$. Further investigation of the interaction effect reveals that the simple slope for high blogger expertise was significant ($p < .05$), indicating that the impact of high blogger expertise varies by campaign intent, in support of $H_{2a}$. The simple slope for low blogger expertise was marginally significant ($p = .10$), suggesting no differences in effectiveness of low-expertise bloggers across awareness and trial campaigns; thus, $H_{2b}$ is not supported (or receives weak support at $p = .10$). Figure 2, Panel A, summarizes the pattern of effects for the interaction between campaign type and blogger expertise.

Perceived functional value of the post was not statistically significant, but the hedonic value of the post had a significant impact on the number of blog post comments ($b = .2616, p < .01$), in support of $H_3$. Campaigns that included campaign incentives also significantly increased the number of blog post comments ($b = .4526, p = .01$), in support of $H_4$.

The results indicate that high blogger expertise is beneficial when paired with awareness campaigns but has a lesser effect of high blogger expertise varies by campaign intent, in support of $H_{2a}$.
in the case of trial campaigns. While we initially hypothesized that for low-expertise bloggers, more engagement would occur under trial than awareness intent, we find no evidence of this relationship. We find that hedonic content is positively associated with an increase in blog post comments. We return to this point in the “Discussion” section.

**Facebook post likes model.** Table 7 also reports the results of the second-stage model with Facebook post likes as the dependent variable. The Akaike information criterion for this model was 5,508, the Bayesian information criterion was 5,582, and the likelihood ratio test was significant \( \chi^2(14) = 221.91, p < .01 \). The VIFs were all below 1.06, with an average VIF of 1.04, indicating no issues with collinearity. The number of Facebook posts was significant \( (b = .5728, p < .01) \). We found a significant main effect of the average number of followers \( (b = .2055, p < .05) \), indicating that this drives Facebook post likes. Campaigns that included giveaways also significantly decreased the number of Facebook post likes \( (b = -.7840, p < .01) \).

Blogger expertise was not significantly related to the number of Facebook post likes \( (b = .0253, \text{n.s.}) \). The main effect of campaign intent (awareness vs. trial) was significant in the Facebook post likes model \( (b = .7416, p < .01) \), in support of H6. We found a significant main effect of hedonic value \( (b = .2215, p = .03) \), in support of H5c. There was a significant, negative interaction effect of campaign intent and hedonic value \( (b = -.4824, p < .01) \). In light of the positive main effects of hedonic value and awareness campaigns, the negative interaction term implies that hedonic value is positively related to Facebook post likes for trial campaigns and negatively related to Facebook post likes for awareness campaigns, providing support for H7. Panel of B of Figure 2, which plots the pattern of results, shows that posts low in hedonic value can weaken engagement, particularly for trial campaigns. Taken together, the results indicate that multiple factors can increase engagement in sponsored Facebook posts. Regarding the control variables, having more Facebook posts, posts on weekends, and a higher number of followers are all related to an increased number of Facebook post likes. Posts lower in hedonic content are particularly harmful when paired with trial campaigns on Facebook.

We find that the blog platform and Facebook platform exhibit differences in drivers of engagement. Campaign incentives negatively affect the Facebook platform but not the blog platform; we conjecture that this is due to the cannibalizing effect of the blogging platform. Timing of the posts (weekends vs. weekdays) also positively affects Facebook, but this effect is not consistent for the blog platform model.

**Discussion**

In this study, we evaluated the effectiveness of influencer marketing campaigns using an empirical database of sponsored bloggers. The results provide support for most of our hypothesized effects (with the exception of \( H_1 \) and \( H_{2b} \)). Across both models, we find a positive impact of the number of followers. Controlling for the number of followers, we find that blogger expertise, campaign intent, hedonic value, and interactions among these variables influence engagement on blog and Facebook platforms. We also find differences in the success drivers of sponsored blogging campaigns across the platforms. High blogger expertise interacts with campaign intent on the blog platform but not the Facebook platform.

We find a significant interaction between campaign intent and hedonic value on Facebook platforms. Specifically, our findings indicate that hedonic value exerts a greater impact in trial campaigns, which supports the explanation that hedonic content may provide a reason for Facebook users to share information or like a blog post with an overtly commercial intent, confirming the compensatory role of hedonic value in mitigating the negative effect of a less desired post. In addition, we find a negative effect of campaign incentives on Facebook post likes for both awareness and trial campaigns, potentially due to cross-platform cannibalization of the Facebook platform by the blogging platform. Campaign incentives may cause participants to interact with a blog post more directly in the blogging environment, even though they may have first encountered the information on Facebook.

In addition, we estimated a series of alternative models for robustness checks, including examining when posts are cross-posted on blogs and Facebook, alternative measures of content sentiment, alternative specifications of number of followers, varying measures of post engagement, and alternative coding for the varimax factor rotations. The robustness checks also included another version of the Stage 1 probit model specification, models using a Gaussian copula, and fixed-effects negative binomial models. The results of these alternative specifications are consistent with our reported findings (for details, see the Web Appendix, as well as an overall summary of the robustness check results in the “Details on Stage 1 Probit Model” section).

Our results thus far have been based on data collected from a real-world context (actual campaigns featuring sponsored bloggers), providing high generalization and meaningful insights into the complex interplay of multiple factors that influence how these campaigns actually function in real life. However, field data limit our ability to manipulate key independent variables, creating the possibility that extraneous variables could account for the effects. To account for this possibility and improve our ability to draw meaningful conclusions from this research, we aimed to replicate our findings in a tightly controlled setting, by experimentally manipulating our key variables. We focused on finding additional support for a key interaction effect observed in the blog platform setting—namely, the interaction between campaign intent and blogger expertise in a blog setting.

**Study 2**

The purpose of Study 2 is to replicate one of the more counter-intuitive results (i.e., the blogger expertise \( \times \) campaign intent
interaction on the blog platform) to provide further support for H2. We posit that campaign intent will have a differential impact on purchase likelihood in the case of high-expertise bloggers but will not affect purchase likelihood in the case of low-expertise bloggers.

Pretests

Expertise pretest. This pretest served to (1) link blogger profile characteristics with perceived expertise and (2) check the strength of our manipulation of blogger expertise. We kept the sample population similar to the target audience as possible. Those sampled were women with children under 18 years of age (N = 97). The pretest was a between-subjects design with two expertise levels (high and low) manipulated using blogger profile descriptions (for details, see the Web Appendix). To create a robust measure of expertise, we used the items from Ohanian’s (1990) scale to measure celebrity endorser expertise. On a scale from 0 (“strongly disagree”) to 100 (“strongly agree”), participants rated whether they believed the blogger was expert, experienced, knowledgeable, qualified, and skilled. We averaged these five items together (α = .96) to create an overall perceived expertise score.

We controlled for homophily to rule this out as an alternative explanation of the blogger expertise effects. Therefore, participants rated three items regarding blogger homophily (0 = “strongly disagree,” and 100 = “strongly agree”): “I feel that the blogger is similar to me,” “I feel that the blogger is a peer,” and “I feel that the blogger thinks like me.” We averaged these three items together (α = .95) to create an overall homophily score. Controlling for age and homophily, we found that perceived expertise is higher under the high blogger expertise manipulation than the low blogger expertise manipulation (Mhigh = 62.20, Mlow = 59.37; F(3, 93) = 20.13, p < .01). In addition, we found a difference in perceived homophily for high-versus low-expertise bloggers when controlling for age, such that homophily is higher in the case of low-expertise bloggers (Mhigh = 46.89, Mlow = 60.26; F(1, 94) = 7.50, p < .01).

Campaign pretest. The goal of this pretest was to test the manipulation of campaign intent. This pretest was a three-condition (campaign intent: awareness, trial, control) between-subjects design (N = 164).5 Participants read a sponsored blog post; the awareness and trial campaigns were both about an educational mobile game targeted at middle schoolers (and their parents), designed to improve spatial reasoning. Participants read identical sponsored blog postings about Water Bears (similar to what was used in the pretest). In the trial condition, an additional phrase at the bottom stated: “BUY NOW!!!” The expertise conditions were identical to those in the pretest. Participants rated how likely they would be to purchase the Water Bears app on a scale from 0% (“not likely at all”) to 100% (“very likely”).

We analyzed the data using analysis of variance containing all the main effects (i.e., blogger expertise, campaign intent, and their interaction). The overall model was significant (F(7, 387) = 22.29, p < .01). The main effect of expertise was not significant (F(1, 387) = .95, p = .33), but the main effect of campaign intent was significant (F(1, 387) = 8.56, p < .01). In support of our hypothesized effect, the interaction between expertise and campaign intent was also significant (F(1, 387) = 8.88, p < .01). We controlled for age, homophily, whether participants had children in middle school, and whether they follow sponsored bloggers online. After controlling for these variables, the test of simple slopes indicated that, consistent with H2a, the impact of high blogger expertise on purchase likelihood is stronger for awareness campaigns than for trial campaigns (Mawareness = 34.10, Mtrial = 20.66; F(1, 387) = 13.32, p < .01). That is, when blogger expertise was high, participants expressed a higher purchase intent for the awareness campaign than the trial campaign. Next, examining purchase likelihood under low blogger expertise, we found no significant difference between awareness and trial campaigns (Mawareness = 30.24, Mtrial = 30.29; F(1, 387) = .14, p = .7106), which, consistent with our empirical results, fails to support H2b. The impact of a low-expertise endorser on purchase likelihood does not depend on campaign intent. This pattern of findings confirms those from our empirical data set (see Figure 3).

5Those sampled were women with children under 18 years of age.
First, the nature of influencer marketing is distinct from both those for other social media and paid media for two reasons. Second, the influencer designs and implements the message, not the company. This is also distinct not only from traditional advertising and spokesperson marketing tactics, due to bloggers’ creative freedom, but also from pure organic WOM, because bloggers are sponsored by the company. With these influencer nuances in mind, we expect that consumers will interpret the message and source differently depending on where and how it is presented.

**Theoretical Contributions**

Our key contributions involve understanding the interplay of post content characteristics (i.e., hedonic value of a blog post), source expertise, and campaign characteristics (i.e., campaign intent and incentives in an awareness-building or trial campaign) on campaign intent and social media platform. While campaign intent has received attention in advertising literature (Muller 1983), our study is the first to examine the impact of influencer marketing campaign intent on engagement. We find that campaign intent is an especially pertinent moderator to many of the relationships in our study. For example, campaign intent moderates the relationship between source expertise and blog post engagement. Campaign intent also moderates the relationship between hedonic content and Facebook post engagement. These findings suggest that the relationship among source, content, and engagement should not be assessed in isolation from campaign intent.

In addition, we contribute to the literature on blogger expertise by demonstrating conditions in which expertise has (1) a positive impact, (2) a negative impact, and (3) no impact. Specifically, we demonstrate that in some conditions, source expertise is positive, and in others, it is nonsignificant. Expert endorsement is beneficial under an awareness intent, while a novice endorsement is beneficial under a trial intent. This effect holds under high-involvement, low-distraction social media platforms. On low-involvement, high-distraction social media platforms, however, source expertise does not affect engagement. We provide a more nuanced explanation of expertise and its role in online brand engagement. Taken together, these findings provide a richer understanding of source expertise in the case of influencer marketing.

We extend prior research on influencer marketing by highlighting the importance of consumer skepticism differences, which may cause campaign intent (awareness or trial) aimed at different stages of the CDJ to function differently. At early stages in the CDJ, consumers are open to guidance from those with high perceived expertise. However, closer to trial, consumers are open to endorsements that originate from either less expert (presumably more homophilous) or more expert sources. This difference is only true in high-involvement platform settings such as blogs. Understanding the contextual effects guiding the impact of source expertise in online WOM and traditional advertising because influencers blend elements of paid and earned media. From a motivation standpoint, while traditional advertising may have multiple goals to meet brand equity–based objectives, influencers have additional loyalty to their followers. This research sheds light on the key drivers of success of influencer marketing campaigns and offers a novel contribution by understanding the interplay of social media platforms and success factors. We find that while network, blogger characteristics, and content characteristics affect multiple types of sponsored blogger engagement, the level of platform involvement and the campaign intent matter for the degree of success. We use both field data based on a large data set of influencer marketing campaigns and a controlled experiment to show convergent evidence of the majority of the hypothesized effects. By understanding this framework to increase engagement, companies can choose bloggers more effectively, matching their characteristics to campaign goals.

We expect the sponsored blogging results to differ from those for other social media and paid media for two reasons. First, the nature of influencer marketing is distinct from both the company. This is also distinct not only from traditional advertising and spokesperson marketing tactics, due to bloggers’ creative freedom, but also from pure organic WOM, because bloggers are sponsored by the company. With these influencer nuances in mind, we expect that consumers will interpret the message and source differently depending on where and how it is presented.

**General Discussion**

This research sheds light on the key drivers of success of influencer marketing campaigns and offers a novel contribution by examining the interplay of social media platforms and success factors. We find that while network, blogger characteristics, and content characteristics affect multiple types of sponsored blogger engagement, the level of platform involvement and the campaign intent matter for the degree of success. We use both field data based on a large data set of influencer marketing campaigns and a controlled experiment to show convergent evidence of the majority of the hypothesized effects. By understanding this framework to increase engagement, companies can choose bloggers more effectively, matching their characteristics to campaign goals.

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**Figure 3. Blogger expertise and campaign intent effects on purchase likelihood (Study 2).**

**Discussion**

Study 2 provides additional, supplemental evidence for the interaction between source expertise and campaign intent on a blog platform. We show that in the case of high-expertise bloggers, awareness intent yields a higher purchase likelihood than trial intent. By demonstrating this effect using purchase likelihood in an experimental setting, we provide further support for the validity of this finding. However, the results in Study 1 are driven by high expertise, while the results in Study 2 are driven by differences due to low expertise. This may be because the outcome variable (purchase intent) is closer to the trial condition (which is driven by low expertise) while Study 1’s outcome variable (engagement in the form of blog comments) is much closer to awareness building.
influencer marketing settings is a key contribution of this research. It extends prior works on influencer marketing settings that focus on either expertise (Uribe, Buzeta, and Velásquez 2016; Zhu and Tan 2007) or stage of the CDJ (Hudson and Hudson 2013) but do not examine their interplay.

We also argue that the motivations driving people to use social media platforms influence how they view different types of influencer marketing campaigns. In a blog environment, in which users are motivated to process information deeply and to engage with bloggers’ information and content, trial campaigns are better received. In a Facebook environment, in which users’ motivations are more focused on sharing information with peers, awareness campaigns have a more positive impact. Given this general preference for awareness (vs. trial) campaigns on Facebook, hedonic value of a blog post takes on more significance in the context of trial campaigns.

Furthermore, our findings show that post content (i.e., hedonic value) is important in generating post engagement. We extend the findings of Berger and Milkman (2012), who argue that hedonic content increases social transmission and virality of online messages. We find that hedonic value has a significant effect on both blog and Facebook platforms. We also show that on low-involvement, high-distraction social media platforms, hedonic content can be beneficial when paired with trial campaigns, perhaps because of their overtly commercial intent. In awareness-building campaigns, in which user motivations involve sharing information, hedonic value may be distracting to the primary goal. Thus, hedonic content of online communications is not always beneficial to marketing campaigns.

Our findings are revealing on the impact of campaign incentives, which research has previously shown to increase WOM (Berger and Schwartz 2011) and enhance quality perceptions of a product (White 2013). We demonstrate that incentives (a chance to win a giveaway) generate WOM benefits in the form of increased engagement (i.e., blog post comments). This finding advances the literature by showing that increased WOM and engagement can be generated without giving a free product to every person; simply offering a chance to be the recipient is enough to induce the benefits of free products. This greatly reduces the costs associated with running a campaign with free product incentives while still generating a similar response.

One potential rationale for why giveaways have a positive effect in a blog environment but a negative effect in Facebook posts is that giveaways are typically executed in a blog platform setting, and blogs cannibalize Facebook engagement. An alternative explanation, which could be the focus of further research, is that high-involvement platforms in general are more conducive to driving engagement through free goods and incentives. Prior research suggests that when consumers are more involved, they want to minimize risk through information search in their decision-making processes (Delgado-Ballester and Munuera-Alemán 2001), and they might view free products or incentives as a way to lower the risk of a new purchase. This is worth examining under a broader understanding of customer engagement.

This article offers a unique contribution by examining the differences between social media platforms. While we empirically focused on blogs and Facebook, the findings can be extended to other social media platforms. As platforms continue to develop, the extent of involvement generated by a platform can help inform decisions on influencer marketing strategies. Moreover, the focus of this research was on online engagement, which sheds more light on customer profitability than a mere focus on customer attitudes or preferences. Furthermore, our examination of cross-platform impacts (i.e., blogs and Facebook) dovetails with other research examining how different advertising media may synergistically improve customer engagement and profitability. Kireyev, Pauwels, and Gupta (2016) investigate the dynamic interaction between paid search and display ads. We extend their findings by focusing on one form of social media marketing that straddles the earned and paid social media types. Therefore, our findings are of particular relevance in light of the increased blurring between these two types of social media marketing. The variations observed across social media platforms indicate that the type of platform can affect the profitability of digital marketing expenditures.

Managerial Implications

We offer novel insights to managers implementing influencer marketing campaigns. First, this article delineates best practices for sponsored bloggers based on marketing campaign intent and platform. When trying to bolster awareness campaigns on a blogging platform, managers should feature the expertise and credibility of the blogger. However, in the case of trial campaign intent, campaigns by both expert and novice sources will be equally successful.

Second, when implementing campaigns on Facebook or any other high-distraction platform, managers should vary content strategy depending on campaign intent. Trial campaigns can benefit from featuring posts with high hedonic value, particularly in high-distraction environments such as Facebook. Furthermore, when choosing bloggers to implement a strategy involving multiple high-distraction platforms, managers should focus on selecting bloggers with a large follower base to ensure the highest penetration and engagement.

Third, with regard to the impact of campaign intent on outcomes, we recommend that managers use the appropriate drivers of success for blog engagement (i.e., blogger expertise, campaign incentives) in awareness campaigns and rely on hedonic-valued content on blog platforms. We further recommend that managers avoid using campaign incentives on Facebook or other low-involvement, high-distraction platforms, such as Twitter or Instagram, and instead focus on the hedonic value of post content.

Return on Engagement

Researchers have begun examining the impact of social media expenditures on firm performance and shareholder value (e.g., Danaher and Dagger 2013). By examining sponsored blogging
across social media platforms, we contribute to extant literature examining the differential profitability impacts of these platforms, thereby extending research that examines across-media profitability impacts of advertising expenditures (Sridhar et al. 2016). In addition, many companies have tried to quantify the value of social media engagement. Estimates range from $3.33 to $8 per Facebook like, while social media shares are estimated at $8 per reweet and $14 per Facebook post share (Hayes 2015). We conservatively estimated the dollar value of each type of engagement at $1 for a Facebook post like, $1 for a Facebook post comment, $2 for a Facebook post share, and $2 for a blog post comment. We multiplied the number of blog post comments and Facebook post likes, comments, and shares by an estimated dollar value for each type of engagement. We then summed these values and used them as the revenue per campaign, per blogger, generated by engagement. Next, we calculated return on engagement (ROE) by dividing the total revenue generated by the total cost for each campaign by blogger. We modeled ROE using campaign intent, expertise, campaign incentives, and hedonic content. We found a marginally significant, positive relationship of expertise and a significant, positive relationship of campaign incentives on ROE (for details, see the Web Appendix). This implies that both blogger characteristics and campaign intent can affect firms’ bottom lines. By optimizing social influencer marketing strategies with these results, firms can increase ROE for influencer marketing campaigns.

Limitations and Future Research Directions

This research is subject to certain limitations, which may present new directions for future research. We examined only a limited set of outcome metrics associated with a particular blog post and did not directly test for the impact on return on investment (ROI). However, Kumar et al. (2013) show that both social media and customer WOM increase ROI, and Kumar and Pansari (2016) demonstrate the relationship between engagement and ROI. Further research could increase the set of outcome measures of a given campaign by considering the direct impact of a blog post on consequential outcomes, such as sales and ROI. Further exploration of why customer engagement could affect these performance outcomes is worth examining and would extend Harmeling et al.’s (2017) framework. Our measurement of key constructs, such as sentiment, relied on post hoc measures based on judges evaluating each blog post for factors such as creativity/unique and personal relevance. A more direct measure would involve having the audiences of a given blogger rate his or her posts for various aspects of sentiment. In addition, research could include a field experiment of blogger choice informed by this research versus the current methods for selecting bloggers for campaigns. In addition, this research uses the bloggers’ network size at the time of the posts but does not formally take into account the entire past performance or longevity of the bloggers’ careers. This could be an important variable to consider in future research. Finally, we acknowledge the highly evolutionary nature of social media platforms. Therefore, while we explore the effects of campaign, source, and hedonic value in terms of two social media platforms, we recommend considering these findings in the light of platform characteristics versus specific platforms.

In general, sponsored blogging and influencer marketing have been the target of ethical debates in recent years. Some critics argue that social influencers fail to reveal their sponsorship by companies, thereby creating a perception that their sponsored posts are organic WOM. This type of deceptive marketing practice has been at the heart of various Federal Trade Commission investigations of Instagram posts in recent years (Ingram and Bartz 2017). The Federal Trade Commission (2017) has reached out to influencers directly and reiterated its requirements to disclose any endorser and advertiser connection. As noted, all sponsored posts in the current research included a declaration of sponsorship at the beginning of the blog post. Still, there is room for research on how sponsored blogging as an advertising medium is distinct from other forms of advertising that consumers view unambiguously as paid advertising.

Authors’ Note

This research was conducted as part of the first author’s doctoral dissertation at the University of Pittsburgh.

Acknowledgments

The authors gratefully acknowledge the data provided by The Mother Institute and funding provided by a Marketing Science Institute grant. The authors also thank Ratti Ratneshwar for comments and feedback on a previous version of the manuscript. The authors thank the reviewers, associate editor, and editor for their constructive input and help on the manuscript.

Associate Editor

Dhruv Grewal

Declaration of Conflicting Interests

The author(s) declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

Funding

The author(s) disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: The authors received funding from a Marketing Science Institute grant.

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References


Hwang, Yoori and Se-Hoon Jeong (2016), “‘This Is a Sponsored Blog Post, but All Opinions Are My Own’: The Effects of Sponsorship Disclosure on Responses to Sponsored Blog Posts,” *Computers in Human Behavior*, 62 (4), 528–35.


Palmatier, Robert W., V. Kumar, and Colleen M. Harneling, eds. (2017), Customer Engagement Marketing. Cham, CH: Springer.


