The use of Web analytics for digital marketing performance measurement

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2015
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Date of original submission: November 16, 2013

(Date of first resubmission: August 26, 2014)

Date of second resubmission: November 13, 2014

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Submission declaration

This full-length manuscript is not under consideration for publication elsewhere, and its publication is approved by all of its authors.
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The use of Web analytics for digital marketing performance measurement

Abstract

This study proposes that the benefits gained from marketing performance measurement are determined by how an organization exploits the metrics system under specific circumstances. For this purpose, the authors review performance measurement literature and apply it to the use of Web analytics, which offers companies a metrics system to measure digital marketing performance. By performing an in-depth investigation of the use of Web analytics in industrial companies, the study shows that an organization’s efforts to use marketing metrics systems and the resulting outcomes cannot be understood without considering the reasoning behind the chosen metrics, the processing of metrics data, and the organizational context surrounding the use of the system. Given the continuously growing importance of digital marketing in the industrial sector, this study illustrates how industrial companies characterized by complex selling processes can harness Web analytics to demonstrate how digital marketing activities benefit their businesses.

Keywords: case study, digital marketing, industrial business, performance measurement, Web analytics
1 Introduction

The role of digital marketing\(^1\) (DM) in a firm’s marketing strategy has been expanding in the industrial sector, as evidenced by industrial firms’ increasing investments in DM activities, which currently account for approximately one-quarter (26%) of industrial firms’ total marketing budgets (Gartner, 2013). In addition to cost effectiveness and changes in customer behavior, investments in DM are motivated by its results being more easily measured compared with those of traditional marketing (Hennig-Thurau et al., 2010; Pickton, 2005; Wilson, 2010). As customers are increasingly interacting with companies through digital channels, marketers have realized the need to track these interactions and to measure their performance (Chaffey & Patron, 2012). For this purpose, firms must adopt Web analytics (WA), defined as “the measurement, collection, analysis and reporting of Internet data for the purposes of understanding and optimizing Web usage” (Web Analytics Association, 2008, p. 3). In this study, WA refers to a tool that collects clickstream data regarding the source of website traffic (e.g., e-mail, search engines, display ads, social links), navigation paths, and the behavior of visitors during their website visits and that presents the data in a meaningful format. The WA data are used to understand online customer behavior, to measure online customers’ responses to DM stimuli, and to optimize DM elements and actions that foster customer behavior that benefits the business (Nakatani & Chuang, 2011).

Although it is limited to the digital environment, the use of WA is an important developmental step toward measurable marketing. As the role of the digital world expands through increased

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\(^1\) Digital marketing refers to marketing that uses electronic devices and channels to support marketing objectives. In this study, digital marketing includes marketing via websites, search engines, online advertisements, e-mail and social media channels. Digital marketing is considered to be a synonym for electronic marketing.
digital media consumption and the integration of the online and offline worlds, the proportion of marketing actions covered by WA is growing. Many offline marketing actions already include digital elements that can be tracked by WA. Examples include quick response (QR) codes embedded in print and outdoor media and augmented reality applications used in, e.g., product demonstrations at trade shows. Additionally, firms can design offline campaigns to drive traffic to digital channels and to measure their impact on website customer behavior. However, firms’ ability to harness WA to improve marketing performance remains limited. In a recent survey of 1,000 U.S. marketers, three of four marketers believed that measuring DM performance was important, but less than one-third (29 percent) thought they were doing it well (Adobe, 2013).

WA is used by more than 60% of the top 10 million most popular websites around the globe (Web Technology Surveys, 2014). In addition to the value of the data that WA produces, the high adoption rate is driven by the fact that some WA tools, such as Google Analytics, can be acquired and utilized free of charge. Despite the high adoption rate, academic research on WA remains limited, and much of the research results reveal a discouraging picture of its use. On average, WA is utilized on an ad-hoc basis, the metrics data are not used for strategic purposes, and the benefits of the usage remain unclear (Hong, 2007; Järvinen, Töllinen, Karjaluoto, & Jayawardhena, 2012; Welling & White, 2006). In contrast, a few case studies demonstrate that measuring and optimizing DM performance measurement with WA has improved the efficiency of marketing actions and subsequently increased sales revenue (Phippen, Sheppard, & Furnell, 2004; Wilson, 2010). Hence, the evidence regarding the benefits of exploiting WA for DM performance measurement is contradictory.
In addition, whether performance measurement and the use of measurement data in decision making result in improved firm performance or other business benefits is generally disputed in the literature. For instance, Franco and Bourne (2004) analyzed 99 published papers regarding performance measurement and concluded that more rigorous research methods were associated with a lower likelihood of performance measurement having a positive impact on firm performance. By contrast, various marketing studies have shown that the use of marketing performance measurement data in marketing decisions has positive performance implications (e.g., Kannan, Pope, & Jain, 2009; Lodish, Curtis, Ness, & Simpson, 1988; Mintz & Currim, 2013; Natter, Mild, Wagner, & Taudes, 2008; Silva-Risso, Bucklin, & Morrison, 1999; Zoltners & Sinha, 2005). However, in practice, many marketing managers remain skeptical toward the use of performance measurement data and instead rely on intuition and experience in decision making (Germann, Lilien, & Rangaswamy, 2013; Lilien, 2011). This perspective is also supported by scientific evidence. Heuristics studies demonstrate that less information may in fact result in more accurate and efficient decision making than extensive analysis of past data because heuristic rules can be used to manage uncertainty more efficiently and robustly than rules based on a broader use of information (Gigerenzer & Brighton, 2009; Guercini, 2012; Guercini, La Rocca, Runfola, & Snehota, 2014). Given this contradictory evidence, this study proposes that performance measurement or the use of WA for DM performance measurement does not inherently improve performance. Rather, the benefits gained are determined by how companies exploit the system under specific contextual circumstances.

Against this backdrop, this study has three aims. First, it advances marketing performance measurement theory by elucidating how organizations can design and apply marketing metrics
systems in a way that creates business benefits. Second, although previous findings demonstrate that WA is more beneficial in businesses in which transactions are processed online (Järvinen, Töllinen, Karjaluoto, & Platzer, 2012), this study demonstrates how industrial companies characterized by a long-duration selling process and an emphasis on face-to-face interaction with customers (Webster, Malter, & Ganesan, 2005) can use WA for DM performance measurement. Third, at a time when new analytics tools and technologies are providing marketers with a rapidly increasing volume of digital data regarding online customer behavior (Deighton & Cornfeld, 2009; McAfee & Brynfjolsson, 2012; Russell, 2010), this study examines the limitations of relying on such data and emphasizes the future challenge of achieving a holistic understanding of customers and marketing performance.

To reach our research objectives, we perform an in-depth investigation of a company that has experienced remarkable benefits from the use of WA and compare the company’s WA use with that of two other companies that have not gained notable benefits despite their active use of WA. The differences in the use of WA are examined in three dimensions: the selection of WA metrics, the processing of WA data, and the organizational context of WA use (adapted from Pettigrew, Whipp, & Rosenfield, 1989). A similar approach has been used in the performance measurement literature (Bourne et al., 1999; Bourne, Neely, Platts, & Mills, 2002; Bourne, Kennerley, & Franco-Santos, 2005; Martinez, Pavlov, & Bourne, 2010). However, this study extends Pettigrew et al.’s (1989) model by demonstrating how it can be applied in marketing performance measurement research.
The remainder of the article is organized as follows: We begin by explaining how the dimensions of Pettigrew et al.’s (1989) model are adapted for the purposes of this study. Thereafter, we review and divide the existing findings regarding performance measurement under the adapted dimensions and discuss how the findings are related to evidence derived from the WA research. In the methodology section, we justify the rationale for using a case study approach and describe the data collection and analysis methods that are used in this study. Subsequently, we present the cross-case findings. Finally, we discuss the theoretical contributions and managerial implications of the study, its limitations, and avenues for future research.

### 2 Framework for investigating the use of performance metrics systems

Research on DM performance measurement with WA is scarce and theoretically underdeveloped. Therefore, we consider a broader perspective for the literature review and combine findings from performance measurement and marketing performance measurement literature. We show that the existing findings regarding the use of performance measurement systems are often parallel to available anecdotal evidence regarding the use of WA for DM performance measurement. The literature review is structured according to the three dimensions of Pettigrew et al.’s (1989) framework, which was originally designed to investigate strategic change in organizations. The key idea of the framework is that the content of change, the process of implementing change and the organizational context in which the change occurs are interrelated. Thus, strategic change can only be understood by investigating all three dimensions. Specifically, content (i.e., the what of change) refers to the particular areas of transformation under examination. Process (i.e., the how of change) refers to the frameworks, patterns, actors, and tools that transition the organization from its present to a future state. Context (i.e., the why
of change) refers to the organization’s internal context (i.e., antecedent conditions, resources, capabilities, structure, leadership, dominating frames of thought, culture, and politics) and the external environment (i.e., the economic, business, and political environment and social and economic trends) in which change occurs.

Pettigrew et al.’s (1989) framework was selected as a guide for this study because it provides a sound structure for organizing disparate findings from the performance measurement literature to develop a holistic understanding of the elements that affect the firm’s ability to design and exploit a marketing metrics system. The framework has been adopted in a number of studies on the use of performance measurement systems, all of which have concluded that the content, process, and context of performance measurement affect the outcome of the system (Bourne et al., 1999; 2002; 2005; Martinez et al., 2010). In these studies, the dimensions have been adapted to better harmonize with performance measurement research and the precise research questions. Therefore, the following definitions are formulated by combining and summarizing the core idea of each of Pettigrew et al.’s dimensions in previous performance measurement studies (e.g., Bourne et al., 2005; Martinez et al., 2010):

*Performance measurement content* refers to the actual metrics system that is developed, including what is being measured, what metrics are selected, and how they are structured as a complete metrics system.

*Performance measurement process* refers to the process through which the performance data are refined and managed.
Performance measurement context refers to the internal and external organizational context in which the use of a metrics system occurs.

For the purposes of this study, we use these definitions and extend their use to address marketing metrics and WA metrics systems. However, regarding the performance measurement context, this study exclusively focuses on the internal context, and therefore, the external context is outside the scope of this research.

2.1 Performance measurement content

The exact design of an effective metrics system is likely to depend on the particular organization in question. Thus, there are no clear standards for building a metrics system that would fit the needs of all organizations. However, research indicates that to develop a successful metrics system, organizations should focus on aligning metrics and strategy, as well as the definitions, dimensions, and structure of the metrics (Table 1).

Table 1: Research findings regarding performance measurement content

<table>
<thead>
<tr>
<th>Content issue</th>
<th>Performance measurement</th>
<th>Marketing performance measurement</th>
</tr>
</thead>
<tbody>
<tr>
<td>The alignment of metrics and strategy</td>
<td>Performance measurement should be based on firm strategies and business objectives (Bourne et al., 2003; Eccles, 1991; Kaplan &amp; Norton, 1996; McCunn, 1998; Neely &amp; Bourne, 2000).</td>
<td>The selection of metrics should be based on marketing strategy and objectives (Ambler, 2000; Ambler et al., 2004; Clark, 2001; Norton, 1996; McCunn, 1998; Neely &amp; Bourne, 2000).</td>
</tr>
<tr>
<td>Definitions of metrics</td>
<td>Clearly defined performance metrics help firms avoid common misunderstandings</td>
<td>Determining the marketing contribution to business outcomes requires that the metrics</td>
</tr>
</tbody>
</table>
The dimensions of metrics should be multi-dimensional or “balanced”, including financial and non-financial, internal and external, leading and lagging metrics (Bourne et al., 2003; Eccles, 1991; Kaplan & Norton, 1992; 1996; Keegan et al., 1989; Lingle & Schiemann, 1996; Neely et al., 1996).

To create a thorough understanding of marketing performance, the selected metrics should reflect short- and long-term as well as financial and non-financial results (Ambler & Roberts, 2008; Clark, 1999; O’Sullivan & Abela, 2007; Rust et al., 2004; Seggie et al., 2007).

Marketers need a comprehensive but manageable set of performance metrics, which requires that they understand the interrelationships between metrics and that they are able to focus on the critical ones (Clark, 1999; McGovern et al., 2004; O’Sullivan & Abela, 2007; Pauwels et al., 2009).

The WA literature has focused on the extent of WA metrics use by organizations and the types of metrics that organizations have adopted (Hong, 2007; Phippen et al., 2004; Welling & White, 2006). As indicated in the performance measurement and marketing performance measurement literature, aligning WA metrics with a DM strategy and business objectives has been demonstrated to be a viable method to increase the benefits of WA use in certain cases (Phippen et al., 2004; Weischedel & Huizingh, 2006). However, little is known regarding the underlying reasons why organizations select certain WA metrics and ignore others.
Marketing performance measurement suffers from an emphasis on subjective measures, such as brand loyalty and customer satisfaction, which are difficult to link to financial metrics that mainly concern top management (Rust, Ambler, Carpenter, Kumar, & Srivastava, 2004; Stewart, 2009). Seggie, Cavusgil, and Phelan (2007) argue that in addition to dissatisfaction toward subjective marketing measures, the power of the Internet will diminish the importance of subjective measures and increase the importance of objective measures. Indeed, one of the advantages of WA is that it offers a variety of objective, standardized, and quantitative metrics that are relatively easy to communicate to senior management. However, the plethora of metrics complicates WA usage because it is difficult to decide which metrics are the most critical to implement (Phippen et al., 2004; Weischedel & Huizingh, 2006; Welling & White, 2006). Firms should begin WA metrics selection by identifying the key performance indicators\(^2\) (KPIs) and differentiating them from other granular metrics (Chaffey & Patron, 2012). However, little is known regarding how companies resolve the challenge of compiling a comprehensive yet manageable set of WA metrics. Moreover, whether quantitative WA metrics can substitute for subjective marketing measures, which are qualitative in nature, remains unclear.

### 2.2 Performance measurement process

Various studies have investigated how performance measurement systems are implemented and how data are processed at the operational level. As a result, the following key phases of the performance measurement process have been identified: data gathering, data analysis and

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\(^2\) In the DM context, KPIs are defined as metrics that indicate the firm’s overall DM performance in relation to its most important DM goals. The KPIs are supplemented with other more granular metrics that are used to evaluate the effectiveness and efficiency of specific DM activities that support the overall DM performance measured by the KPIs (Chaffey & Patron, 2012).
interpretation, results reporting, taking action, and updating the metrics system (e.g., Bourne et al., 2005; Bourne, Mills, Wilcox, Neely, & Platts, 2000). Table 2 summarizes the research findings related to each of these phases.

Table 2: Research findings regarding the performance measurement process

<table>
<thead>
<tr>
<th>Process phase</th>
<th>Performance measurement</th>
<th>Marketing performance measurement</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Data gathering</strong></td>
<td>A multitude of methods to capture performance data; the challenge is to obtain accurate, standard, and objective data (Eccles, 1991; Lynch &amp; Cross, 1991; Nemetz, 1990).</td>
<td>Marketers have difficulties in gathering reliable and objective data (Stewart, 2009).</td>
</tr>
<tr>
<td><strong>Data analysis</strong></td>
<td>The value of performance data depends on how the information is analyzed and interpreted (Eccles, 1991; Hill et al., 1993; Lynch &amp; Cross, 1991; Neely &amp; Bourne, 2000).</td>
<td>A key challenge in processing marketing performance data is to refine it to actionable insights (McGovern et al., 2004; Pauwels et al., 2009).</td>
</tr>
<tr>
<td><strong>Results reporting</strong></td>
<td>Standardized and regular reporting leads to better performance (Bourne et al., 2005; Hacker &amp; Brotherton, 1998).</td>
<td>Reporting marketing performance to executives positively influences managerial satisfaction, attitudes, and behavior toward marketers (Curren et al., 1992; Pauwels et al., 2009).</td>
</tr>
<tr>
<td><strong>Taking action</strong></td>
<td>Improving performance necessitates that performance data are utilized for taking corrective action toward existing practices (Forza &amp; Salvador, 2000; Lebas, 1995; Lynch &amp; Cross, 1991).</td>
<td>Acting on the basis of marketing performance measurement data results in positive performance implications (Kannan et al., 2009; Lodish et al., 1988; Silva-Risso et al., 1999).</td>
</tr>
<tr>
<td><strong>Updating the metrics</strong></td>
<td>Modifying and updating the metrics</td>
<td>Non-existing</td>
</tr>
</tbody>
</table>
metrics system

system is vital to reflect changes in strategic objectives and targets (Bourne et al., 2000; Johnston et al., 2002; Lingle & Schiemann, 1996; Neely et al., 2000; Wouters & Sportel, 2005).

Gathering reliable data for metrics systems is challenging (Eccles, 1991; Lynch & Cross, 1991; Nemetz, 1990; Stewart, 2009). Thus, an advantage of WA metrics is that the collection of WA data can be standardized and automated (Russell, 2010). For this reason, data gathering is not expected to be a major obstacle to the use of WA data. Instead, in line with the performance measurement literature, WA data are useless without proper analysis and interpretation (Chaffey & Patron, 2012; Court, Gordon, & Perrey, 2012; Phippen et al., 2004). Clearly, the analysis and interpretation phase is a prerequisite for gaining insight and improving current DM practices. Therefore, this phase is presumed to have a major influence on the benefits gained from WA.

Reporting the measured marketing performance outcomes to executives leads to favorable managerial attitudes and behavior toward marketers (Curren, Folkes, & Steckel, 1992; Pauwels et al., 2009). Presumably, reporting DM performance to executives similarly results in positive outcomes. However, how reporting should be organized and how detailed the information that management is willing to receive from DM results remain unclear. In addition, as coordination and clear responsibilities are commonly key factors in successful performance measurement processes (Eccles, 1991; Simons, 1991), one key consideration concerns how responsibilities should be shared and coordination performed regarding WA data.
2.3 Performance measurement context

Research on the internal performance measurement context has identified various factors that influence the use of performance measurement systems. These factors include analytics skills and resources, information technology infrastructure, senior management commitment, leadership, and organizational culture (Table 3).

Table 3: Research findings regarding the internal performance measurement context

<table>
<thead>
<tr>
<th>Internal context factor</th>
<th>Performance measurement</th>
<th>Marketing performance measurement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Analytics skills and resources</td>
<td>Designing and implementing a performance measurement system requires sufficient skills and human resources (Kennerley &amp; Neely, 2002).</td>
<td>Analytics skills and knowledge of measurement techniques are necessary for the use of marketing performance data (Germann et al., 2013; Lenskold, 2002; O’Sullivan &amp; Abela, 2007; Patterson, 2007).</td>
</tr>
<tr>
<td>Information technology infrastructure</td>
<td>Suitable information technology infrastructure improves the integration and accessibility of performance data (Bourne et al., 2002; Bititci et al., 2002; Eccles, 1991; Marchand &amp; Raymond, 2008; Lingle &amp; Schiemann, 1996; Nudurupati &amp; Bititci, 2005).</td>
<td>Sophisticated information technology infrastructure supports the exploitation of marketing metrics data (Germann et al., 2013).</td>
</tr>
<tr>
<td>Senior management commitment</td>
<td>Management commitment encourages the implementation and active use of a performance measurement system (Bititci et al., 2002; Bourne et al., 2000; Bourne et al., 2002).</td>
<td>Support from top management in terms of attention, budget, and human resources is necessary for the successful deployment of marketing performance data (Germann et al., 2013).</td>
</tr>
</tbody>
</table>
The literature has emphasized the importance of expertise and analytics skills in selecting suitable WA metrics and analyzing WA data for gaining meaningful insight (Chaffey & Patron, 2012; Court et al., 2012), which is also commonly mentioned in marketing performance measurement studies (Germann et al., 2013; Lenskold, 2002; O’Sullivan & Abela, 2007; Patterson, 2007). In addition to expertise and analytics skills, the role of management in the use of WA has been discussed, as senior management can be held responsible for investing in recruitment, training, and suitable information technology infrastructure (Chaffey & Patron, 2012). Furthermore, the use of analytics is more effective when the organizational culture favors data-driven decision making, cooperation, and information sharing, which often requires effective change management practices (Davenport, 2013; McAfee & Brynjolfsson, 2012).
Regarding the information technology infrastructure, anecdotal evidence suggests that one of the primary advantages of WA tools is that they can be synchronized with other enterprise software, such as customer relationship management (CRM) and social analytics software (Digital Marketing Depot, 2014). However, the integration of WA tools with other information technology platforms has not yet been explored in the academic literature.

3 Methodology

For this paper, the case study approach was selected as the research strategy. According to Yin (1981), the case study approach is favored when the study investigates a contemporary phenomenon in its real-life context and when the boundaries between phenomenon and context are not evident. In this study, WA is a contemporary phenomenon because the technology has gained wider attention only during the last decade and because the academic research on WA is still in its infancy. In addition, this study aims to elucidate the underlying reasons for firms’ varying benefits from WA, which can only be achieved through in-depth investigation of selected organizations.

The study was conducted as part of a two-year DM research project that was supported by seven large industrial firms and seven service providers, such as DM agencies. During preliminary discussions with the participating companies, we found that DM performance measurement emerged as a top-priority research theme and that multiple companies had already used WA for this purpose. Studying WA for DM performance measurement specifically in large industrial companies provided a fruitful research setting because the use of WA in industrial companies has not yet gained much interest in the academic literature. Furthermore, generally, marketing
performance measurement is particularly challenging in industrial settings, which are characterized by complex, long-lasting selling processes that render demonstrating the impact of DM on business performance difficult (Webster et al., 2005). In contrast, large industrial companies are more active users of DM measurement than small and medium-sized firms (Järvinen et al., 2012a), which implies that larger companies are more likely to have the resources and knowledge required for the successful use of WA. Therefore, by investigating large industrial companies, we expected not only to identify challenges in the use of WA but also to find insights and solutions for overcoming these challenges.

Three of seven industrial companies that participated in the research project reported actively using WA for DM performance measurement, and all three were willing to participate in the study. These three companies were largely similar in terms of digital marketing activities and channels in use. That is, all of the companies used a company website, campaign websites, search engine marketing (which encompassed both organic and paid search), display advertising, e-mail newsletters and social media (primarily Twitter, LinkedIn, Facebook and YouTube) to achieve digital marketing goals. From this perspective, the opportunities to use WA for DM performance measurement did not differ among the case companies. Two of the firms (which are identified in this research by the aliases Machinery and Paper) stated that they were not satisfied with their current use of WA and had gained only minor benefits thus far. In contrast, the third firm (i.e., Steel) reported that it was highly satisfied with its WA use and that it had experienced substantial improvements in DM performance by using WA. Against this background, we created a comparative study design in which we compared the differences in the use of WA among the companies to discover the reasons for the varying benefits that were experienced.
More specifically, our empirical study design follows the literature review and includes the performance measurement content, process, and context dimensions of WA use. Regarding the context dimension, we limited our study focus to the internal context and concentrated on internal organizational factors related to WA use. Excluding the external context was justified because the case companies were in many ways similar from the external context perspective, sharing the same political, cultural, and social background and operating largely in the same market areas. The details of the three case companies are presented in Table 4.

Table 4: Background information of case companies and interviewees

<table>
<thead>
<tr>
<th>Company code name</th>
<th>Machinery</th>
<th>Paper</th>
<th>Steel</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Ownership</strong></td>
<td>Public limited company</td>
<td>Public limited company</td>
<td>Public limited company</td>
</tr>
<tr>
<td><strong>Main industry</strong></td>
<td>Industrial goods</td>
<td>Paper</td>
<td>Steel</td>
</tr>
<tr>
<td><strong>Annual revenue</strong></td>
<td>USD 5+ billion</td>
<td>USD 10+ billion</td>
<td>USD 3+ billion</td>
</tr>
<tr>
<td><strong>Number of employees</strong></td>
<td>ca. 20.000</td>
<td>ca. 20.000</td>
<td>ca. 10.000</td>
</tr>
<tr>
<td><strong>Headquarters</strong></td>
<td>Finland</td>
<td>Finland</td>
<td>Finland</td>
</tr>
<tr>
<td><strong>Market reach</strong></td>
<td>Global</td>
<td>Global</td>
<td>Primarily Europe</td>
</tr>
<tr>
<td><strong>Interviewees and their positions (names have been changed)</strong></td>
<td>David: team leader of digital communications</td>
<td>Richard: expert in digital communications</td>
<td>Charles: customer data expert in digital marketing</td>
</tr>
<tr>
<td></td>
<td>Lisa: communications expert in branding</td>
<td>Helen: communications manager of Web services</td>
<td>Thomas: campaign manager of digital marketing</td>
</tr>
<tr>
<td></td>
<td>Nancy: team leader of</td>
<td>Sandra: team leader of</td>
<td>Donna: content and SEO</td>
</tr>
</tbody>
</table>
The primary data collection method was interviews. The target group for the interviews was digital marketers who were or had been involved with DM performance measurement and the use of WA. The management in each company performed the actual selection of the key respondents based on the candidate respondent’s role in DM performance measurement tasks. Ultimately, we conducted four to five interviews in each case company (14 in total). After these interviews, the data were determined to be saturated and representative given that fewer than 20 employees in each company were involved with DM performance measurement. The average length of the interviews was 55 minutes, and all the interviews were audio recorded with the permission of the interviewees. The interviews did not include rigidly formulated questions but were open-ended in nature and only guided by six themes: (1) DM strategy and objectives, (2) DM activities and channels, (3) DM performance measurement tools and practices and the role of WA use, (4) WA metrics selection, (5) WA data processing and reporting, and (6) opportunities and challenges in DM performance measurement and the use of WA. In addition to these guiding themes, we kept the interviews as open-ended as possible and allowed the interviewees to freely raise any issues that they thought were relevant to the topic.

To complement the data from interviews (and increase the validity of the study), two workshop sessions were organized to allow for informal group discussions on the key respondents’ opinions and experiences regarding the research topic. The participants in the workshops
included the same individuals who participated in the interviews and representatives with an interest in the topic from other companies that participated in the DM research project. However, in the analysis of workshop sessions, we included only the comments made by the interviewees. The topics of the workshop sessions were open in nature but were designed based on the interviews, as we pursued issues that multiple interviewees’ had considered particularly challenging or important in the interviews. For example, many interviewees noted the difficulty of deciding what WA metrics to select and what metrics to ignore, so we tried to elaborate this issue by asking the participants to share their thoughts and rationale for making metrics selection decisions. The data gathering in the workshops was conducted such that one researcher participated by raising topics to discuss while two other researchers observed and took notes of comments and reactions by the participants of the case companies. Finally, e-mail exchanges were used in the data collection where an interview request was declined.

The analysis of the case data followed a three-step thematization process comprising condensing the data, displaying the data, and drawing and verifying conclusions (Miles, Huberman, & Saldaña, 2013, pp.12-14). First, the audio-recorded interviews were transcribed and combined with notes from the workshop discussions and e-mail exchanges, after which the data were reviewed several times. Second, content analysis was performed by descriptive coding to create relevant categories (such as metrics selection, data gathering, analytics skills) followed by second-cycle coding in which the descriptive codes were grouped according to the content, process, and context of WA usage (Miles et al., pp. 74, 86-93). After a carefully conducted process of coding and recoding the data, we reviewed the cases individually and compiled case descriptions based on this review process. Thereafter, a comparative analysis was conducted to
examine the differences and similarities across the individual case findings. Finally, to verify the study results, our interpretations were presented in final meetings with each firm where key respondents were invited to comment on the study’s findings and conclusions. Against this background, the results obtained by the researchers were reliable in terms of managerial relevance.

4 Findings
The case companies substantially differed in terms of their satisfaction toward and benefits gained from the use of WA. The participants from Machinery and Paper considered WA to make their DM more measurable but noted that the greatest benefit was being able to track how many people visit their website and how much traffic different marketing actions attract to the website. In comparison, Steel’s participants reported multiple major benefits from using WA: First, Steel’s marketers are now able to measure financial outcomes of DM and demonstrate their contribution to top management. As a result, the digital marketers’ influence in the company, as well as their budget, has increased. Second, the marketers are much more aware of the relative effectiveness of various DM channels and actions to attract visitors to their website. Third, Steel’s marketers have a better understanding of the type of content that attracts potential customers to interact with the company and which customer actions at the website indicate customer interest in the company’s offerings. Overall, Steel’s marketers can employ this information for planning new DM actions and modifying existing actions to improve
Steel’s DM performance had been improving month by month as measured by the website traffic, sales leads\(^3\), revenue, and profits generated.

*Lately, digital marketing has been systematically invested in. You can clearly see direct monetary and human resource investments as well as the top management commitment to this thing. With the help of analytics, the management and sales teams have undoubtedly noticed that our digital services, website, and all our activities have a powerful impact, and the change has been radical in the last few years. The budget is still bigger for offline marketing, but the digital marketing budget has multiplied. Last year, I think we more or less tripled our budget.* (Joseph, director of digital marketing, Steel)

### 4.1 Web analytics measurement content

In the design of the content of a WA metrics system, two distinctive approaches were identified: Steel built its metrics system by considering its top priority marketing goals and by including WA metrics that would indicate how their DM activities support the achievement of these goals. By contrast, Machinery and Paper primarily included the metrics that were easily available and that provided information that is considered to be meaningful for DM performance.

Steel’s study participants continuously emphasized that all of their efforts in digital channels are primarily aimed at increasing sales. Acknowledging the complex purchase process in the metal industry, the company has designed a WA metrics system that measures the purchase process at

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\(^3\) A sales lead is a website visitor who displays interest in the company’s products or services and leaves his or her contact information. The definition applies to all three studied companies.
three levels: traffic generation to the website, user behavior and interaction on the website, and revenue and profits gained through online sales leads (Table 5). By measuring these different levels, Steel hopes to examine the different stages of the customer’s path to purchase and to improve its understanding of how DM influences the customer’s buying behavior. In addition to setting specific metrics for the different stages of the customer’s path to purchase, the metrics are classified in KPIs that indicate the overall performance at each level of the purchase process and other metrics that are linked with KPIs and provide more specific information on how to define overall performance.

Table 5: Steel’s Web analytics metrics for the different stages of customers’ path to purchase

<table>
<thead>
<tr>
<th>Traffic generation to website</th>
<th>Website behavior</th>
<th>Revenue &amp; profits</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Key performance indicators</strong></td>
<td><strong>Key performance indicators</strong></td>
<td><strong>Key performance indicators</strong></td>
</tr>
<tr>
<td>• number of all website visits</td>
<td>• number of sales leads</td>
<td>• sales revenue through sales leads</td>
</tr>
<tr>
<td>• website visit growth (%)</td>
<td>• sales leads growth (%)</td>
<td>• profits through sales leads</td>
</tr>
<tr>
<td>• conversion rate (^{a})</td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Examples of other metrics</strong></td>
<td><strong>Examples of other metrics</strong></td>
<td><strong>Examples of other metrics</strong></td>
</tr>
<tr>
<td>number of website visits and website visit growth (%) driven per traffic source:</td>
<td>(number of)</td>
<td></td>
</tr>
<tr>
<td>• campaign website</td>
<td>• visits in product information pages</td>
<td>• number of sales leads that lead to a transaction</td>
</tr>
<tr>
<td>• organic search</td>
<td>• product information sheet downloads</td>
<td>• percentage of sales leads that lead to a transaction</td>
</tr>
<tr>
<td>• paid search</td>
<td>• product comparison tool uses</td>
<td>• average costs incurred per sales lead</td>
</tr>
<tr>
<td>• display advertisements</td>
<td>• product video views</td>
<td>• number of transactions per traffic source</td>
</tr>
<tr>
<td>• e-mail</td>
<td>• visits in contact request form</td>
<td>• sales revenue per traffic source</td>
</tr>
<tr>
<td>• social media (Twitter, LinkedIn, Facebook, YouTube)</td>
<td>• sales leads per traffic source</td>
<td>• profits per traffic source</td>
</tr>
</tbody>
</table>

Notes: \(^{a}\) Conversion rate: the percentage of visitors who take a desired action such as purchasing products, leaving a contact request, subscribing to newsletters, and downloading brochures.

The second most important goal for Steel’s DM activities is to enhance customer relationships by providing customer service through digital channels. All of the respondents mentioned providing
superior service for existing customers as a vital goal for their DM activities. However, the respondents noted that because of its qualitative nature, measuring service quality with WA is difficult. Therefore, they must rely on customer surveys and informal feedback.

The third goal established for Steel’s DM is to improve brand awareness and image. However, the importance of this goal slightly differed between the participants. The major arguments against improving brand awareness and image were that brand is less important in the metal industry relative to other industries and that reliably measuring brand awareness and image is impossible. One participant mentioned that they focus on increasing sales because of difficulties of measuring brand awareness and image.

*I admit that it is a little bit shortsighted to measure digital marketing performance by comparing costs to produce a sales lead resulting in sales with monetary value. Investing in brand building might yield even better results in the long run, but then again, lead generation metrics make it easy to justify the costs of a campaign and show its direct monetary value.* (Thomas, campaign manager of digital marketing, Steel)

In comparison with Steel, Machinery and Paper encountered more difficulties in designing a holistic WA metrics system. The participants from both companies easily recalled various WA metrics that they use (Table 6) but had more difficulty justifying the selection of those particular metrics. Machinery and Paper have adopted numerous similar metrics to those used by Steel. However, unlike Steel, they have no clear structure that defines the practical significance of the selected metrics and their relative importance. The reason for this difficulty is partly the result of
Paper’s and Machinery’s inability to specify and prioritize their DM goals, which range from increasing sales and improving brand image to strengthening service processes, creating meaningful content and fostering customer dialogue. In the absence of clearly defined DM goals, it is unfeasible for Paper and Machinery to design a metrics system that would reflect their ultimate DM performance.

Table 6: Selected Web analytics metrics of Machinery and Paper

<table>
<thead>
<tr>
<th>Machinery</th>
<th>Paper</th>
</tr>
</thead>
<tbody>
<tr>
<td>• traffic volume to website</td>
<td>• number of website visits</td>
</tr>
<tr>
<td>• traffic volume to website from search engines/paid online advertisements/e-mail/social media</td>
<td>• number of unique visitors</td>
</tr>
<tr>
<td>• unique website visitors</td>
<td>• sources of website traffic</td>
</tr>
<tr>
<td>• pages views per visit</td>
<td>• average time spent on website</td>
</tr>
<tr>
<td>• time spent on website</td>
<td>• click-through rate $^a$ from paid online advertisements</td>
</tr>
<tr>
<td>• top pages on website (pages with the most views)</td>
<td>• product demonstration views (videos on website)</td>
</tr>
<tr>
<td>• sales leads</td>
<td>• page views on where-to-buy section</td>
</tr>
<tr>
<td></td>
<td>• sales leads</td>
</tr>
</tbody>
</table>

Notes: $^a$ Click-through rate: the ratio of clicks to impressions of an online advertisement (e.g., display and search engine advertisements).

Furthermore, although Machinery and Paper measure the generation of sales leads, they do not follow how many of these leads result in transactions. Thus, they are unable to link their DM activities with financial outcomes.

_We have been planning and working with [digital] measurement issues, but it is always challenging to decide which metrics to include and which not to include. As in all marketing themes, it is particularly challenging to find financial metrics that show our return on investment. We would really need a comprehensive metrics set that we could use_
to demonstrate the value of our work in monetary terms. For now, we haven’t been able to develop a proper system. (Nancy, team leader of branding, Machinery)

4.2 Web analytics measurement process

Steel had outlined a clear process and clear responsibilities for their use of WA data (Figure 1). The WA data are automatically collected with Google analytics and a specific online survey (E-space) that randomly targets website visitors with a short survey regarding their website experience. Steel uses these survey data to identify the types of visitors who visit their website and to evaluate how well the website serves different customer and other stakeholder needs. Google analytics forms the core of Steel’s WA measurement, as it enables them to measure the effectiveness of specific digital marketing activities and to connect these activities with the generation of sales leads. The generated leads are stored in the CRM system (sales force automation), which allows the company to track whether the leads resulted in sales. With the CRM system, Steel is also able to follow the generated leads over time and to determine how they react to various marketing inputs.

All customer-related data from digital surveys to campaigns goes directly into our CRM system under a specific customer profile. Leads from a certain campaign are automatically directed to the correct salespersons, and we can follow the yield of such a campaign in real time. (Charles, customer data expert in digital marketing, Steel)
Although data gathering is automated, it is supervised by Steel’s customer data expert and campaign manager whose main responsibilities are to analyze and interpret the data and to draw insights from the data. The DM director coordinates the data analysis, makes detailed inquiries regarding the interpretations, and offers weekly feedback to the team based on the results. Finally, the director summarizes the KPIs and key insights and presents them in a monthly meeting with senior management. Based on the results, management offers feedback to the DM director, who disseminates the information to his team.
Such a systematic measurement and reporting process does not exist at Paper or Machinery. On the contrary, the absence of a systematic approach in the deployment of WA was widely discussed in the interviews and workshops. The respondents from both firms explained that in the course of implementation, they had failed to develop a process to systematically analyze and report the WA data. The responsibilities were considered unclear, and the reporting had been *ad hoc*. One respondent noted:

_Honestly, I think that only successful campaigns are reported, because we lack systematic reporting._ (Helen, communications manager of Web services, Paper)

The measurement tools of Paper and Machinery were partly the same as those of Steel. Both companies used Google analytics and online survey applications similar to E-space. However, without a systematic measurement, analysis and reporting process in place, Paper and Machinery were unable to effectively use these tools. Another key difference in comparison with Steel’s use of WA was that Machinery and Paper were not able to integrate WA and CRM data. Consequently, the WA data of Paper and Machinery resided in a separate database, which made tracking leads over time and obtaining customer-level insights impossible.

### 4.3 Web analytics measurement context

*Analytics skills and resources*

Marketers’ analytics skills in using WA were found to be inadequate at both Paper and Machinery. Although the study participants rarely mentioned skills as a major obstacle, their lack of skills was apparent from their inability to understand the opportunities offered by WA and to tailor its usage to performance measurement. The selection of metrics for company needs was commonly mentioned as a challenge, and the failure to understand marketing strategy often
undermined the selection process. The interviewees’ responses showed that when the marketing strategy and key objectives are unclear or abstract, transferring them to actionable WA metrics becomes difficult. In addition to the lack of skills, another related problem at Paper and Machinery was the insufficient human resources dedicated to the use of WA. The respondents complained that daily routines occupied so much of their time that they had little time to determine how they could make the most of WA. The same pitfalls in skills and resources were not found at Steel. The study participants from Steel had a clear understanding of their DM strategy, and the WA data were used to measure DM performance in relation to the strategic objectives and to optimize DM activities to continually improve performance.

**Information technology infrastructure**

Information technology infrastructure was not identified as a major issue for the use of WA at any of the case companies. The study participants from Steel reported that synchronizing WA data with CRM software was straightforward and that this practice enabled them to tailor their DM actions for specific customers based on their website behavior. Although Paper and Machinery had not linked WA data with CRM, none of the respondents from these companies mentioned this issue as a challenge. The participants were either unaware of this opportunity or did not consider it to be a key issue. Integrating WA data with social analytics software was not a topical issue for the studied companies because none of them had significant experience in using such software. Currently, the studied companies were satisfied with the information that their WA tool provided regarding social media performance, that is, how much traffic social media attracted to the website and the subsequent outcomes.

**Senior management commitment**
The level of senior management commitment to DM and the use of WA substantially varied between Steel and the other companies. At Steel, senior management commitment is evidenced by considerable investment in recruiting an expanding team of specialists to operate the firm’s DM and the allocation of substantial monetary resources to DM activities. In addition, the senior management demonstrated substantial interest in the results generated by establishing clear reporting criteria that the DM team must satisfy monthly and provided constant feedback based on the results.

Although only a few participants from Machinery and Paper mentioned limited budgets and other resources as an obstacle, senior management’s low commitment to WA was evidenced by the lack of attention to DM performance measurement. In fact, the respondents from both companies reported that senior management had not made any requests to report the performance of DM activities. Moreover, the digital marketers at Paper and Machinery felt that making DM measurable was solely in their own interests. To obtain senior management’s perspective on this issue, we approached the head of marketing and communications at each firm with an interview request. However, both of our requests were politely declined because of the leaders’ (self-reported) limited understanding of DM and its performance measurement.

**Leadership**

A lack of leadership was found to be one of the pitfalls in the use of WA at Machinery and Paper. Multiple employees were responsible for measuring DM performance to differing degrees. However, no one was certain who was in charge of the process. In comparison, Steel’s DM director was clearly responsible for coordinating the DM performance measurement process. The director had assigned clear responsibilities for each member of his team and was actively
participating in the analysis and interpretation of the WA data. In addition, he reported the results to senior management and disseminated the feedback from management to his team.

The study data also indicated that Steel’s DM team members trusted the director’s expertise. For example, when the team members were unsure about the answer to a question, they typically responded that Joseph (the director) would know the answer to the question. It was also mentioned that Joseph has been the company’s chief DM advocate and that he initiated the digitization of Steel’s marketing strategy. In fact, the interview and the workshop discussions clearly indicated that Joseph genuinely believed in the power of DM and that he was eager to demonstrate the results of his team’s efforts. The interviews with other team members indicated that the director had managed to inspire other team members with the same enthusiasm, as was shown by their general belief that DM was finally enabling them to demonstrate the contribution of their daily work.

**Organizational culture**

Two organizational culture issues identified in the case data clearly differentiated Steel’s use of WA from that of Machinery and Paper. First, Steel’s approach to DM decision making was largely data driven in the sense that it was exploiting WA data to evaluate and learn which DM activities performed best, and this information was used to optimize subsequent DM activities. In comparison, Machinery and Paper relied primarily on their intuition and perception of what activities bring value to customers. The WA data were used as a supplemental source for DM decision making in situations where they wanted to get support for their decisions e.g., when they wished to know whether video clips on the website were receiving any attention.
The second organizational culture condition that emerged from the data was related to cooperation and information sharing between the digital marketers. Cooperation was weakest at Machinery, whose DM team was split into branding and digital communications sections. The two sections of the team lacked awareness of one another’s activities and a coherent view of DM performance measurement as a result of limited information sharing. Although Paper’s DM team is spread across multiple geographic locations, the team members regularly communicate and meet to ensure that everyone is aware of one another’s tasks and activities. However, Steel’s situation, in which the whole team works in the same building, was regarded as the best arrangement to foster cooperation because it enabled the team members to continuously interact and to plan DM activities and measurement practices together.

5 Discussion

The study findings provide a number of theoretical contributions. First, although an ample body of literature discusses the inability of marketers to demonstrate the contribution of marketing activities to business outcomes (Li, 2011; O’Sullivan & Abela, 2007; Rust et al., 2004; Stewart, 2009; Wiersema, 2013), this study shows that the ability to demonstrate marketing performance depends on the organization’s content, process and context of the marketing metrics system in use. Specifically, the findings emphasize the importance of the following: 1) designing a manageable metrics system that demonstrates the progress toward marketing objectives, 2) establishing a process that fosters the effective use of metrics data within the organization, and 3) ensuring that the organizational context supports the use of the metrics system. By investigating the use of marketing metrics systems multidimensionally, the study extends Pettigrew et al.’s
(1989) framework by demonstrating how this framework can be applied to marketing performance measurement research.

As its second major theoretical contribution, this study is the first to demonstrate how industrial companies characterized by complex and lengthy selling processes can harness WA to improve their DM performance measurement practices. Whereas previous case studies have demonstrated the benefits of WA for measuring marketing performance in e-commerce businesses (Phippen et al., 2004; Wilson, 2010), this study demonstrates that the benefits gained from WA are not limited to those business sectors in which transactions can be processed online. Compared with traditional measurement methods, such as customer surveys and interviews, which are subjective and vulnerable to response bias, the advantage of WA is its ability to gather objective data on genuine online customer behavior and subsequent business outcomes. Although the actual purchase decision in the industrial sector is often made through personal selling, industrial marketers can use WA to measure, for example, which DM activities attract potential customers to interact with the company, how many sales leads are generated and how many of these leads result in transactions. Consequently, industrial companies that use WA are in a better position to demonstrate the influence of marketing actions on business benefits. Moreover, in agreement with the previous marketing performance measurement literature (O’Sullivan, Abela, & Hutchinson, 2009; Pauwels et al., 2009), the findings demonstrate that reporting the objective and standardized metrics provided by WA to top management increases the influence of industrial marketers in an organization.
As its third theoretical contribution, this study illustrates the need for multiple methods in measuring overall marketing performance. The study findings support Seggie et al.’s (2007) assessment that the Internet will diminish the importance of subjective marketing measures and increase the importance of objective measures. However, there is no evidence that objective measures provided by WA would obviate the need for subjective measures. In addition to the fact that WA is largely limited to the digital environment, the study revealed two major weaknesses related to the use of WA data. First, the data provided by WA are backward-looking. That is, they present customer behavior and DM results in retrospect but are less helpful for evaluating the future intentions of customers. Second, WA data are exclusively quantitative and cannot be used to measure the fulfillment of qualitative objectives, such as enhancing brand image and increasing customer satisfaction or positive word-of-mouth that may be ultimately more important for a company with respect to, e.g., maintaining customer relationships. This deficiency is a significant disadvantage, particularly in industrial marketing, in which business relationships are especially important and in which businesses therefore require relationship-specific information to interact with customers (La Rocca & Snehota, 2011). In conclusion, relying solely on WA data may result in suboptimal or harmful marketing decisions. Thus, companies should only use WA data as a component of performance evaluation. This suggestion is supported by numerous studies that demonstrate that selected marketing metrics should reflect short- and long-term as well as financial and non-financial results (Ambler & Roberts, 2008; Clark, 1999; O’Sullivan & Abela, 2007; Rust et al., 2004; Seggie et al., 2007).

5.1 Managerial implications

The study has three managerial implications for using WA. First, instead of adopting a variety of WA metrics, managers should primarily focus on designing a manageable metrics system that is
linked to the firm’s top priority marketing objectives. A vital part of this process is to identify and select firm-specific KPIs with respect to the major marketing objectives and to differentiate them from other secondary metrics. By prioritizing the WA metrics, marketers can focus on the most important marketing objectives and avoid information overload. Once the KPIs are selected, the relevance of other metrics should be evaluated based on the information that they provide in relation to the KPIs. That is, the other metrics should be used to obtain more detailed information on why the overall performance measured by the KPIs is below or above the target. Generally, we recommend that managers create a WA metrics system that illustrates the interrelationships among the metrics. Steel’s metrics system that outlines the different stages of customers’ path to purchase is an innovative way to construct a metrics system but is not the only way to do so.

Second, to achieve optimal outcomes, managers must plan a systematic process for managing WA metrics data. Because data gathering can be automated with WA and therefore can be relatively effortless, the largest obstacle in the WA measurement process is to analyze and interpret the data to gain meaningful insight and inform marketing decisions. Managers can advance the analysis and interpretation of the data by clarifying clear responsibilities for WA users and by appropriately coordinating the process. In addition, managers should ensure that DM outcomes are reported to top management, as observed in Steel’s case, where reporting KPIs convinced top management of the contribution of DM to business performance and the feedback from the top management to the digital marketers encouraged and motivated them to continuously develop their activities.
Third, managers should ensure that the organizational context supports the use of WA. They must ascertain that WA users have sufficient time and expertise to use the system and acquire new talent if necessary. Moreover, we recommend that managers play an active role in coordinating WA use and providing feedback regarding DM outcomes. When management has only limited interest in DM outcomes, digital marketers are not motivated to develop a proper metrics system and apply it. Finally, to foster active WA use, the WA users should have a suitable leader. The leader should be able to manage a variety of tasks in the WA use process, including sharing responsibilities with team members, coordinating and participating in the execution of tasks, and creating a culture that fosters cooperation, information sharing, and data-based decision making.

5.2 Limitations and future research

The results of this study must be interpreted in light of its limitations. Notably, only three industrial companies were investigated. Thus, the generalizability of the study results may be diminished. By investigating companies from other sectors, we may have encountered other circumstances that are relevant to WA use and that were not revealed by our case data.

The study focused on the use of WA metrics systems, and it is unclear whether the findings are applicable to other marketing metrics systems. Although our findings were largely consistent with previous results reported in the marketing performance measurement literature, additional research is required to confirm the applicability of our framework to other settings. It is also noteworthy that the study did not examine the external context of organizations with respect to DM performance measurement. Future research should investigate how the external context of organizations influences the use of WA and other marketing metrics systems. In addition, the
study discusses various benefits regarding the use of WA for DM performance measurement. However, given the qualitative nature of our study, we cannot demonstrate a causal relationship between the benefits of WA use and firm performance.

Finally, although WA is likely the most prominent new technology for DM performance measurement, many other analytics innovations exist, and more analytics innovations are emerging at an increasing pace. For example, the social media revolution has resulted in new social analytics innovations that are used to measure and understand the social behavior of customers in the digital environment and its influence on the marketing performance of firms (Hoffman & Fodor, 2010; Peters, Chen, Kaplan, Ogniben, & Pauwels, 2013). Although this study supports the perception that business marketers currently do not actively measure their social media performance (Michaelidou, Siamagka, & Christodoulides, 2011) and that they thus have limited interest in using social analytics, such attitudes will likely change with the increasing sophistication of analytics tools. Additionally, with the increasing use of analytics tools, the amount of data that can be used to evaluate marketing performance is rapidly increasing. Thus, a future challenge is to unify these “big data” in different forms, databases, and platforms to provide a complete understanding of how customers behave in today’s world and to assess marketing performance in this changing environment.
References:


