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## **A Design science approach to website benchmarking**

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### **Abstract:**

**Purpose** – The literature identifies current approaches to website benchmarking are time-consuming, generally survey-based, and lack agreement on what website components to measure and how these may be measured. A comprehensive and integrated artifact is developed to evaluate and benchmark websites, thereby allowing possible website repositioning to better fit consumer expectations. A theoretical explanation of website benchmarking is provided.

**Design/methodology/approach** – Design Science Research Methodology (DSRM) establishes a website analysis method (WAM) artifact. This WAM artifact solves the DSRM-identified website benchmarking problem using a six activities process (problem identification, solution objective, artifact design/development, artifact demonstration, artifact evaluation, results communication).

**Findings** – This study's rough-cut, mutually-exclusive, causal-summing, set-theory approach to website benchmarking enables scores and comparisons to be established. Functional and domain scores can then be used to modify the website and to encourage consumers into making behaviorally-related transaction decisions.

**Research implications** – The WAM artifact encompasses 28 functions, and 3 domains. This approach allows new emergent components and/or functions to be included as upgrades and redundant components to be removed. The WAM artifact allows benchmarking comparisons or website upgrades to be optimized.

**Practical implications** – The DSRM artifact WAM is software programmable, easy to implement internally or externally and user friendly.

**Originality** – DSRM identifies the benchmarking problem and uses rough cut set-theory with mutual-exclusivity and causal-summing to enable the WAM artifact as an objective solution. This is a new comprehensive approach to website benchmarking and has competitive and website behavioral implications for corporates.

**Keywords:** Website analysis method, benchmark, set-theory, causation, competitive, artefact, behaviour, transaction

**Article Classification: Research Paper**

### **Introduction**

Websites are assets that can deliver high dollar return-on-investment – especially when fully incorporated into a corporate’s business model (Simmons *et al.*, 2008). Although the global e-commerce sales through websites are to exceed \$1.5 trillion in 2014 (eMarketer, 2014), at the national and corporate level, business websites generally do not capture the quality, the innovation and the competitive positioning benchmarks (O’Cass and Ngo, 2012) expected when the corporate is in pursuit of high financial returns.

#### *Website benchmarking approaches*

Currently well over 150 website components have been identified (Cassidy and Hamilton, 2011; Olsina and Rossi, 2002; Stepchenkova *et al.*, 2010), but few studies have attempted to use these large numbers of measures when comparing or benchmarking websites. In addition there is little consistency or consensus on what and how to benchmark. However, some recognize website benchmarking (WB) should deliver a process that is both easy-to-implement and one that delivers timely, cost-effective, and interpretable (internal and external) results (Boisvert and Caron, 2006).

From a corporate perspective although websites do change WB is often assessed irregularly and inconsistently, and may involve specific scoping (Lee and Kozar, 2009) or subjective comparison of convenient components at one particular point-in-time or the occasional comparison against selected competitor websites.

WB employing surveys is complicated because respondents are often restricted by their level of understanding regarding the website’s marketing, aesthetics and design, technical parameters, language terminologies, or its social and interactive components. Also when large amounts of benchmarkable data is to be collected (Krosnick *et al.*, 2002) the respondent often suffers from task-time fatigue (Lee *et al.*, 2011) and their benchmarking decisions show cognitive inconsistencies (Krosnick *et al.*, 2002)

Other corporates choose to derive their WB metrics through computer scanning approaches (Calero *et al.*, 2005; Olsina and Rossi, 2002) and accept comparative ratings where the website is deemed more innovative, or economic, or strategic (Page and Lepkowska-White, 2009; Zhu *et al.*, 2009). As most computer scanning, WB approaches incorporate degrees of human subjectivity in their design or interpretation ratings, these approaches are limited. For example, most are point-in-time developed and scoped by the programming team - who set which website components are to be evaluated and interpreted, and then decide on the intelligence of the software, its comparative mechanisms and its associated metrics.

Although a raft of WB survey software and scanning approaches has emerged each displays limitations in scope, or in the capture of new technologies or features, or in inherent program and reporting subjectivities.

#### *Study motivation*

This study is motivated by an interest in e-business, and information systems research, along with concerns for corporates and their websites when competing globally. Further motivation emanates from the relatively slow and often small advances, deployed in business website comparison approaches, and that to-date WB has lacked both a theoretical approach and an approach to capture relevant analytics. Another motivation is to offer corporates’ a simple, comprehensive, and time-efficient way to interpret their website components and so overcome managerial reluctance to instigate website change(s). A final motivation emanates from the knowledge that when a corporate, through its website, instigates such experimentation, exploration and exploitation approaches, it can release new competitive and

positioning opportunities (Adler *et al.*, 2009). An extensive literature study also identifies (and supports) a need for a different and non-subjective universal approach to business WB.

### *Design science research*

This study adopts a design science research (DSR) approach to design and develop a new WB approach for business.

*Design science research is a research paradigm in which a designer answers questions relevant to human problems via the creation of innovative artifacts - thereby contributing new knowledge to the body of scientific evidence. The designed artifacts are both useful and fundamental in understanding that problem. (Hevner and Chatterjee, 2010).*

Under DSR we apply its seven guidelines to develop our technology-based solution (artifact) to solve this WB issue (Hevner *et al.*, 2004). From the literature, we first explain the gaps that exist, and that no theory shows overall relevance to this WB problem. The artifact can be a construct, or a model, or a method, or an instantiation. In this study we engage the artifact as a website analysis model (termed WAM). WAM is supported by clear, verifiable, and rigorous method, and is applied at the construction and evaluation stages of the artifact. WAM also offers an effective and understandable approach for its relevant audiences (Hevner *et al.*, 2004).

The following sections discuss existing literature approaches to WB and the design science research methodology (DSRM) employed within this study. The application of set-theory; Aristotle's theory of causation; planned behavior, motivation, consumption, and gratification theories are further considered for applicability to WB. The study then describes the deployment and evaluation of the WAM artifact, and concludes with an assessment of implications and other areas befitting future research.

## **Background**

Current literary approaches to business WB include theoretical approaches such as subjective impressions of website consumers (Barnes and Vidgen, 2002), perceived-qualities (Yoo and Donthu, 2001), theory-of-organizational-trust (Fang *et al.*, 2014), and usability. Other WB studies adopt consumer satisfaction and literary assessments (Elling *et al.*, 2007), or prior research across information and service qualities (Webb and Webb, 2004). Loiacono *et al.*, (2007) engage the theory-of-reasoned-action and technology acceptance models for marketing, whilst other WB approaches synthesize the expectation-disconfirmation paradigm and create theories around consumer-satisfaction (McKinney *et al.*, 2002). Still others study how the visual complexity and ordering of components within a website affect the consumer (Deng and Poole, 2010). The above approaches to WB are based around consumer opinion; hence they remain open to subjectivity constraints.

Some WB researchers adopt narrowly-focused, survey-based approaches, and for a specific purpose, generally employ selected survey components (rather than considering a broad-ranging assessment of the website). These study-specific website aspects include: effects of privacy and perceived security on user trust levels (Flavian and Guinaliu, 2006), influences of online trust (Chen and Barnes, 2007), e-service quality (Udo *et al.*, 2010), information flows between websites (Nel *et al.*, 1999), and consumer website satisfaction (Luo *et al.*, 2012). Such WB studies are restrictive and are designed to assess specific components, and so cannot provide comprehensive solutions.

### *Survey-based Website Benchmarking Tools*

Several WB tools are actually survey-based methods. These studies typically target specifically-chosen website components. Such approaches include WebQUAL 4.0 (Barnes

and Vidgen, 2002), WEQ (Elling *et al.*, 2007), SITEQUAL (Yoo and Donthu, 2001), eTailQ (Wolfinger and Gilly, 2003), SiteQual (Webb and Webb, 2004), and PEEIM (Fang *et al.*, 2014). These approaches employ questionnaires and each measures less than fifty-eight website components on a five or seven point Likert type scale.

Although questionnaires remain popular for data capture, several psychological and human respondent issues warrant consideration prior to survey implementation. First, when researchers use exchange-theory to improve the WB motivation for survey completion they must account for respondents who weigh-up the reward/benefit they receive when completing a survey against their cognitive-cost (or effort-required) (Albaum *et al.*, 1998; Evangelista *et al.*, 2011). Here, and in self-reported questionnaires, respondents weigh a question's focus and then attempt to anticipate the answer being sought (Cao *et al.*, 2005). Thus, bias can emerge in exchange style approaches.

Second, as survey-length increases, response-time increases. Here, some respondents answer each survey question to the best of their ability, and some may experience survey fatigue (Nathan and Yeow, 2011) when weighing their personal cognitive cost against time to select from near-identical questions (Krosnick, 1991). In contrast, less rigorous respondents may just select 'don't know/neutral', or select randomly across the scale (Krosnick *et al.*, 2002), whilst others may select an answer that appears acceptable, but not necessarily optimal in their view (Krosnick and Alwin, 1987; Klandermans, 1984). Hence, carefully prepared and balanced surveys are imperative, but the number of questions asked of respondents remains numerically-restrictive when compared to the full range of components available for WB.

Third, 'response-order-effects' such as question wording, complexity of rating scale, or order of answers (Krosnick and Alwin, 1987; Krosnick, 1991) can directionally lead responses (Schwarz, 2007).

Fourth, many past studies utilize student respondents (Loiacono *et al.*, 2007), but such convenience sample approaches lacks validity (Wells *et al.*, 2011) and transferability when applied to consumer diversity within global web communities (Lee *et al.*, 2012; Yang *et al.*, 2012; Xu *et al.*, 2013). These studies may also introduce unexpected (or unexplained) biases (Steelman *et al.*, 2014).

Hence, although well-executed, surveys generally account for the majority of psychological and respondent issues (Krosnick *et al.*, 2002); they remain constrained by the above. In addition, most WB studies are lacking in a strong theoretical base (Lee and Kozar, 2009; Yoon and Kim, 2009), with some applying various theories to benchmarking tools - which they then adapt, and apply to websites.

### *Website benchmarking using modified business tools*

Some WB approaches have modified existing business application tools. Kaplan and Norton's (1992) balanced scorecard (BSC) approach has been modified in several tourism and travel WB studies. These studies typically use commercially-available software to measure technical performance of the website (Choi and Morrison, 2005; Lee and Morrison, 2010). However, they show considerable variation in their use of the remaining three BSC dimensions - with Choi and Morrison (2005) using consumer, marketing-effectiveness, and/or business perspectives, Lee and Morrison, (2010) adopting marketing, consumer, and internal-critical-success-factors, and Kline *et al.*, (2004) adding consumer-friendliness, site-attractiveness, and marketing-effectiveness.

Dichotomous (present/absent) WB studies (Kline *et al.*, 2004; Choi and Morrison, 2005; Lee and Morrison, 2010) provide yet another modified BSC approach, however these again show inconsistencies and differing degrees of detail, but they do extend WB studies into classifications that engage between 40 and 150 components.

Technology acceptance model (TAM) variants applied to WB have added online shopping perceptions across 45 Likert scale web components of e-service quality and perceived measures of service value, trust, and usefulness (Lee and Wu, 2011), others used 50 Likert scale web components to consider trust and intention to purchase (Gefen *et al.*, 2003). Whilst in tourism accommodation, TAM is modified using semi-structured interviews across 36 website components (Herrero and San Martin, 2012). Again WB variants are purpose specific and inconsistent in approach.

### *Component-based website benchmarking tools*

From a WB tools perspective, the key-quality-factors approach rates 28 components as present, or absent, or not assessed and graphs comparisons (Cox and Dale, 2002). In pursuit of potential website performance improvement areas, function-benchmarking captures 91 present or absent business components under 18 separate groups and rates a level-of-presence for each group (Boisvert and Caron, 2006). Again WB tool approaches lack consistency.

Content-analysis WB approaches also vary with 53 components across six service quality domains (each containing closely-related components) (Nusair and Kandampully, 2008), or with website-usability-factors tracked using 38 Likert scale consumer-needs components (Nathan and Yeow, 2011).

Expert reviewers of websites often display differing views of website component groupings (Cassidy, 2010), and to improve consistency Stepchenkova *et al.*, (2010) use two experts over 11 weeks to build their 99 component and four domain WebEVAL tool.

The above WB approaches are inconsistent, lack precise point-in-time comparison capabilities, are respondent completion-time restricted, are survey scope-and-size restricted, and at some point each WB approach relies on human involvement and subjective judgment. Hence, automated WB tools are developed.

### *Automated Website Benchmarking Tools*

Automated WB tools such as WebTango, crawl a website, and compute metrics on 157 components (Ivory and Hearst, 2002), but this tool is limited to website design (Ivory and Megraw, 2005). WebQEM is a hyper-document WB tool tracing 150 direct (or indirect) quality aspects (Olsina and Rossi, 2002) but overlap between the WB measures arises, its restrictive scoping leaves out other WB measures.

Hypertext markup language (HTML) validation programs such as Weblint, are best used in combination with other HTML checking tools (Bowers, 1996). For example, EvalIris evaluates the markup of webpages identifying accessibility failures through HTML components (Abascal *et al.*, 2004); however some accessibility issues still require manual checking. HTML tools focus on the technical domain, and so offer narrow WB solutions.

KWARESMI extracts, structures, and organizes web usability guidelines towards an automated evaluation (Beirekdar *et al.*, 2002) but only does so one page at a time. GIST collects information on users; then infers website behavior and attitudes; then identifies and tests new nano-segments; and tracks the gaps across chosen WB segments (Albert *et al.*, 2004). As per previous approaches, these automated WB tools also lack completeness, requiring degrees of human consideration (developer or respondent), and are often costly to implement.

Thus, a significant WB gap remains in the literature, and we now present our proposal for an artifact to solve this business problem. This artifact - termed WAM (our website analysis model) draws together existing web components under three domains (marketing aesthetics and technical) that are associated with a website.

## **Methodology**

*Design Science Research*

This study follows a *Design Science Research Methodology* (DSRM) (Peppers *et al.*, 2007), and applies DSR guidelines (Hevner *et al.*, 2004) to the practical study of WB. Table 1 summarizes the six DSRM activity stages. The above introduction and literature review completes DSRM activity stages one and two, identifies WB approaches, identifies the limitations in the literature, identifies and discusses problems for organizations and then defines objectives for the solution to the identified WB business problem.

This study now discusses activities three, four, and five of the DSRM process. It then considers the implications of this research, suggests what path future research may take, and lastly provides conclusions to the research community at this point-in-time.

**Table 1.** Design Science Research Methodology – adapted from Peppers *et al.*, 2007

Process Iteration	<b>Activity 1.</b> <i>Problem identification and motivation</i>	<ul style="list-style-type: none"> <li>Approaches in literature: time-consuming, generally survey-based, lack of agreement on components (and how measure), limited components and/ or area measured</li> <li>Organizations need: easy-to-implement, all-encompassing tool that benchmarks internally (and externally) and offers multi-level comparisons – and easy-to-interpret, efficient and cost-effective, and capable of delivering interpretable solutions for website improvement.</li> </ul>
	<b>Activity 2.</b> <i>Define objectives for a solution</i>	<ul style="list-style-type: none"> <li>Extensive literature review and investigation of research developed tools.</li> <li>Develop comprehensive and integrated framework to evaluate websites. One allowing managers to strategize and possibly reposition their website (possibly to better fit user expectations).</li> </ul>
	<b>Activity 3.</b> <i>Artifact Design and Development</i>	<ul style="list-style-type: none"> <li>DSRM, set-theory, causation theory, planned behavior, motivation theory, gratification theory, consumption theory</li> <li>WAM artifact = multi-level, hierarchical, 234 literature-identified, mutually-exclusive components into 28 mutually exclusive functions into 3 mutually exclusive domains. Measure components present=1 or absent=0 = software programmable easy-to-interpret scores at level, cost-effective</li> </ul>
	<b>Activity 4.</b> <i>Demonstration</i>	<ul style="list-style-type: none"> <li>Implement WAM during the design stages of <a href="http://www.therideguide.com.au">www.therideguide.com.au</a></li> <li>A purposed-built publicly-accessible tourism website built in 5 monitored stages - each with more components added to gauge if component additions influence traffic on an active website and if these additions can be deemed website improvements</li> </ul>
	<b>Activity 5.</b> <i>Evaluation</i>	<ul style="list-style-type: none"> <li>Google advanced analytics and server statistics track consumers. Data analyzed and results compared with objectives (est. June 2013)</li> <li>Statistical analysis and interpretation</li> </ul>
	<b>Activity 6.</b> <i>Communication</i>	<ul style="list-style-type: none"> <li>Journal articles and conferences</li> <li>DSR and IS journals, and Benchmarking, an International Journal, and ACIS conferences</li> </ul>

*Design and Development*

*Mutually-exclusive set-theory and causation theory*

The measurable components sourced across the literature each represent a different, and mutually-exclusive, part of the website. Hence, using mutually-exclusive set-theory we can sum the components present on a website. We can then compare this internal WB score, to the WB scores obtained from other (external) websites.

As set-theory houses mutual-exclusivity we collate each website component present into just one mutually-exclusive subset of similar items - termed a *function*. Each function, and its small cluster of similar website components, is grouped with other like functions (each made up of their own mutually exclusive set of similar components). We then group these mutually exclusive functions into one of three mutually exclusive *domains*. The sum of all the components present across the three domains of a website constitutes the total score of all the components present on the website.

When measured against the sum of all the possible components possible for any website, a benchmark score for this website can be generated, and website comparisons at component, functional, and domain level are achievable. Thus, a benchmarking process is theoretically available for websites. The difficulty is - how can this be done, and can the existing website components be meaningfully interpreted?

To understanding the possible website components through mutually exclusive set theory approach we define the artifact WAM as the universal set of all possible website components. We note that within a domain, the full subset of literature-established possible website components making up a function can be different in number when directly compared to another function's full subset of components. Hence, to mathematically compare functional contributions to a website - a functional scaling process is also required.

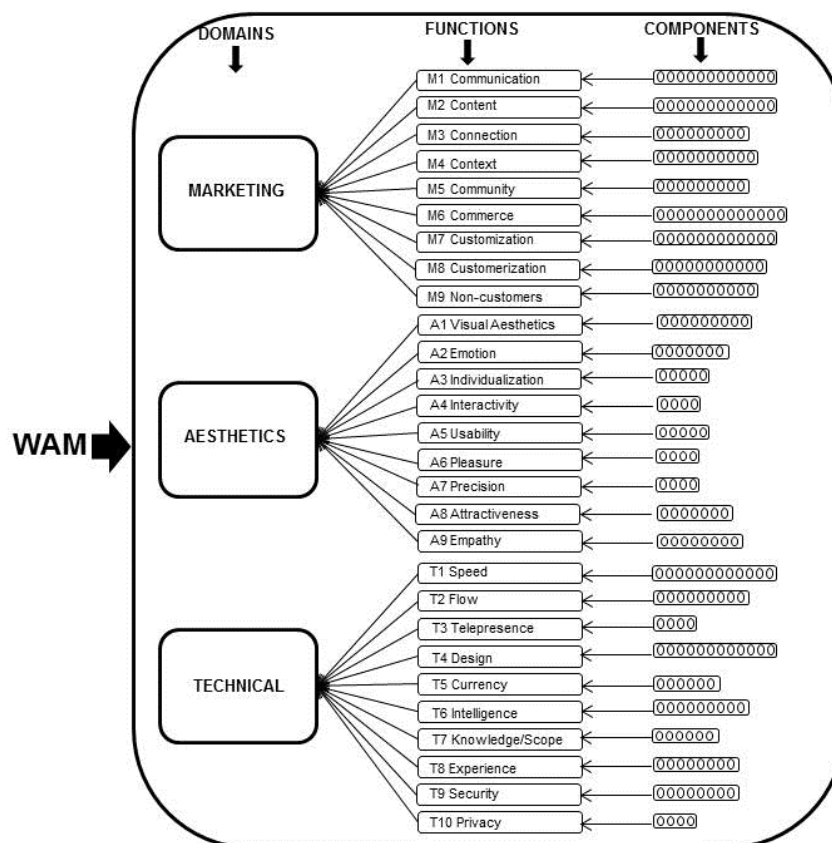


Figure 1. WAM (developed from < remove for refereeing >)

Next we combine mutually exclusive set theory with Aristotle's four step *theory-of-causation* (material-cause, formal-cause, efficient-cause, and final-cause) (Falcon, 2011) and build the following linkages:

1. *Material-cause:* The WAM universal set is defined. WAM captures all possible literature-supported website components.
2. *Formal-cause:* All components are linked and compared through typologically-collated subsets and sets. The presence (or absence) of components within a subset or website set is determinable.
3. *Efficient-cause:* The existing components of a website are collected and compared to the WAM universal set of all possible website components. This allows internal benchmarking at the component, the function, the domain, and at the overall-website level. It also allows external benchmarking against other chosen (or competitive) websites.
4. *Final-cause:* the universal set components not accounted for in a website are presented as both improvement considerations and as possible future inclusion



components. This can be useful when strengthening weaker functions or poorly covered domains within a website.

*Rough-cut mutually-exclusive set-theory and causation theory*

WAM is dynamic and can grow. For example - as the literature advances, and technology advances, new website components continually emerge, and each of these new components potentially adds to a function, then to a domain, and then to the website's sub-set score of the WAM universal set. Alternatively, if the new component represents a breakthrough technology (remove for refereeing), another new function (or even a new domain) may start to emerge for inclusion in the WAM universal set. Likewise, obsolete components (such as obsolete website software-related technical components) are considered for deletion.

Thus, the WAM universal set typifies an infinite number of mutually-exclusive website components (Vilenkin, 1986), and at any point-in-time, functional and domain subset changes may arise. Hence, the currency of the WAM universal set engaged is normally a very close approximation to the actual scope of universal set components available for websites.

Over time, component, functional, and/or domain changes are possible, and these changes affect the WAM universal set. Hence, this study engages a *rough-cut* (Zhu, 2007), *inclusion-exclusion principal of set theory*, and the *causal summing of components into a rough cut, mutually-exclusive, causal-summing, set-theory* approach when benchmarking websites via their components.

The WAM universal set ( $U$ ) contains all available website components at the point-in-time.  $C_p$  is components present on the website. The website's benchmark score ( $W$ ) compares the existing components against the possible components of the universal set as follows:

$$W = \frac{\sum(\text{components present})}{\sum(\text{components possible})} = \frac{\sum(C_p)}{\sum(U)}$$

where:  $c_p \in U$

This study's multi-level benchmarking of websites offers further finely-focused assessments, and more detailed internal or external comparison, and it also delivers an interpretable benchmark score (Boisvert, and Caron, 2006).

To statistically compare components present on the website against the WAM universal set each component present is recorded as '1', and each component missing is scored as '0'. Components in a function are summed into a functional score. Similar functional scores are then summed into a domain score. The functional and the domain scores allow finer internal (or external) website comparisons. A functional or a domain score below its maximum exhibits a degree of 'greyness' (or the extent of missing components) (Deng, 1989; Ho and Wu, 2006) in the score, and these missing components can then be scheduled for inclusion at chosen point-in-time.

Each function captures a different aspect of the website, and the number of its literature-established components can vary. Such variations may be exacerbated when new components are added to the website. Hence, to reduce component-masking currently only 13 components are allowed per function, and to ensure sufficient triangulation, the function must possess at least four components.

Hence, the *rough cut, mutually-exclusive, causal-summing, set-theory* approach to the summing of website components, enables current benchmarking scores and comparisons to be established.

The relative contribution of a function ( $F_j$ ) from all its components ( $c_{ij}$ ) is described as the maximum number of components ( $x_j$ ) present in a given function, with '10' norming the components ( $i$ ) into a comparable subset solution. Thus, a function ( $F_j$ ) is benchmarked as follows:

$$F_j = \frac{10}{x_j} \sum_{i=1}^{x_j} c_{ij}$$

A comparable domain level ( $D_{jk}$ ) benchmark score is derived with ( $x_{jk}$ ) as the maximum number of components present in each function ( $j$ ) of the domain( $k$ ), and  $y_k$  is the maximum number of functions in this domain.

$$D_{jk} = \sum_{j=1}^{y_k} \frac{10}{x_{jk}} \sum_{i=1}^{x_{jk}} c_{ijk}$$

A low scoring function contains less than its possible number of components, and where some missing components are deemed important by management, these can be scheduled for point-in-time addition. High scoring functions may be checked by management against industry (or competitor) norms, thus ensuring each existing component remains appropriate. Whichever decision management may make, missing website components within the WAM universal set are highlighted for point-in-time attention. This logic again applies at the domain level, highlighting weaker domains for potential internal WB improvement by management.

To enable external WB, domain scores are collated into an overall website score ( $W$ ) using the following formula (adding  $k$  domains, and  $z$  as the maximum number of domains).

$$W = \sum_{k=1}^z \sum_{j=1}^{y_k} \frac{10}{x_{jk}} \sum_{i=1}^{x_{jk}} c_{ijk}$$

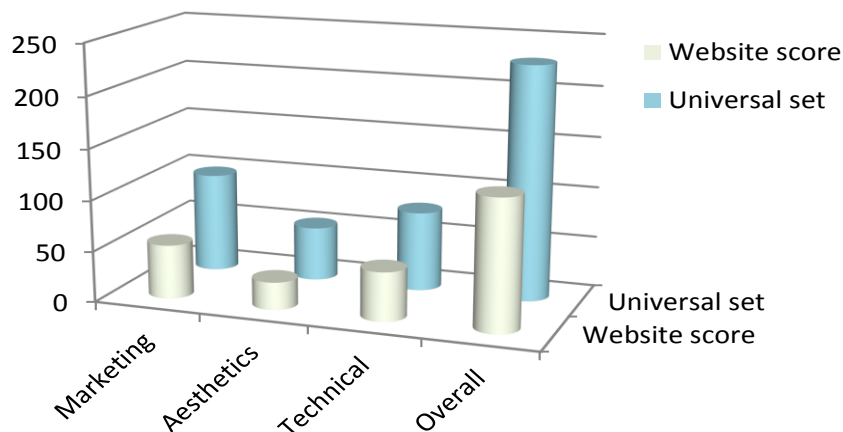
Thus, from a typological perspective, rough-cut set-theory (Zhu, 2007) identifies missing website components, and so highlights grey (or missing component) areas (Deng, 1989), within the function. But although the *rough cut, mutually-exclusive, causal-summing, set-theory* approach for WB delivers benchmarks works it also needs to give effective solutions.

#### *Cause-and-effect considerations*

Benchmarking should contribute towards a superior state of affairs - one consistent with the corporate's purpose to exist, to survive and hopefully to grow (Moriarty, 2011). This is a *cause-and-effect* process. The *cause* satisfies knowledge about something as related to material components, to formal principles, to efficient actions, or to a reason to apply something (and so change a state-of-affairs). The *effect* builds on the antecedents (which themselves cause change), or the opportune/chance alterations, or the efficiencies that are generated (Hamilton, 2010). Therefore, WB also permits management to reflect, and to then change a process, or to then choose to build a different future perception.

We visualize this effect recognition process as Figure 2. In three-dimensional-space, the front pale domain score of all its components within a website can be visually benchmarked

for major weaknesses against the back darker WAM universal set for each websites domain. The same process applies to the overall website versus the WAM universal benchmark set. Such visual benchmarking comparisons (Boisvert and Caron, 2006) also align with our above theory. However, Aristotle's 'final' cause also includes consideration around responsive decision making. Hence, we introduce *cognitive-theory* to understand consumer decisions for visiting a website, and consider how consumer expectations can be met by management.



**Figure 2.** Effects recognition in website benchmarking

### *Cognitive Theory*

Ajzen's (1991) *theory-of-planned-behavior* suggests a consumer intends to do something - such as visiting a website. From *motivational-theory*, some website consumers respond to their intrinsic needs and anticipate personal internal benefits yet to be obtained from a website - such as pleasure, satisfaction, or enjoyment (Lin and Lu, 2011; Luo *et al.*, 2011; Wakefield *et al.*, 2011). Others are motivated to then pursue their extrinsic needs, and to seek useful benefits - such as obtaining information, social interaction, or casually passing time (Luo *et al.*, 2011; van der Heijen, 2011). Thus, consumer motivation to engage a website leads to consumer consumption of some offering on the website, and this is captured as consumption theory. Further, a level of reflective consumer gratification about the website follows consumption. This is captured as user-gratification theory.

At a chosen website, the successes in motivating a consumer's consumption (LaRose and Eastin, 2004) reflects on the website's ability to intrinsically and/or extrinsically meet each consumer's cognitive goals (often gauged against this consumer's acceptable cognitive cost) (Kim *et al.*, 2007). Hence, when the net motivation-to-consumption-to-gratification outcomes pathways are positive (Hamilton and Tee, 2012), then consumers likely display a greater tendency to reuse the website (Lin and Lu, 2011; Wakefield *et al.*, 2011).

Thus, managers can design their website by recognizing their consumers' likely motivations. Past studies suggest that a sequencing of the consumer expectations/motives-to-consumptive-acquisitions-to-gratifying/revisit intentions (Hamilton and Tee, 2012) may increase revisiting. Hence a stronger suite of engaging and consumer-targeted behaviorally-focused website offerings may then draw additional consumers into using (and reusing) the website.

*DSRM Demonstration and Evaluation of the WAM artifact*


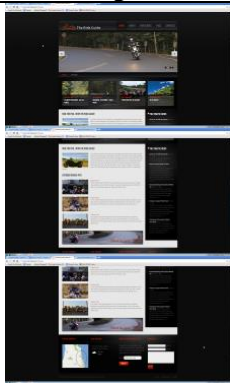

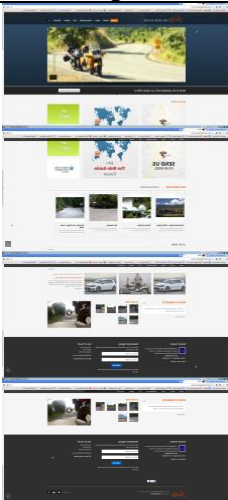

To demonstrate the WAM universal set (or DSRM artifact) as suitable for WB, standards can be developed across chosen industry-sector websites - but commercial IP issues may hinder access to such websites. A second WAM universal demonstration can see a few similar business websites undergoing sequential changes in specific component areas - but obtaining permission to instigate such ongoing changes remains difficult.

This study opts to demonstrate the WAM universal set (or DSRM artifact) through the build of a commercial (and consumer-targeted) website. The study adopts a tourist information website, specifically built for the region’s niche market of motorcycle tourism. This sequential build adds consumer-engaging components stage-by-stage. Each build stage remains commercial, fits the chosen existing marketplace, and has sufficient consumer reach. Also, at the appropriate development stage all normal website tools (SEO, Facebook, Twitter, Flickr, YouTube and Forums) are deployed.

This study employs Google advanced analytics server statistics (and other supporting web-analytics services) to monitor six consecutive build stages of *www.therideguide.com.au*. This study was conducted over 100 weeks between mid-2012 and mid-2014. Each website stage is live, and is tracked for at least twelve weeks, and each stage incorporates additional components - ones designed to advance the website’s overall capabilities, and to grow its consumer base.

Homepage screen captures of each website stage are shown at Table 2 (and all screen capture pages are retained). Once the five website stage developments conclude the website ownership is transferred to a national web-marketing company and is to undergo further regional and national motorcycle ride-tourism development.

**Table 2.** The Ride Guide Homepage at each change

Stage 1	Stage 2	Stage 3	Stage 4	Stage 5
				

Stage one is a static one page ‘brochure’ of limited information. This represents the simplest website presence for *www.therideguide.com.au*. Stage two is a static website of five pages – with basic ride information, photos, static-weather and news, frequently asked questions, an about-us section, and a contact-us section. Stage three adds the basic motivations of communications, static maps, additional ride details, and a YouTube channel.

Stage four commences the build that caters for additional expectations. It includes the interactive consumptive value additions of live map variations (day, street and 3D) and basic social media (twitter, Facebook, forums), along with embedded videos, a photo gallery. The

gratification component of the website includes a ‘send us your rides’ and a forum section. These additions are designed to improve the website’s value and to grow consumer traffic.

Stage five houses a fuller suite of consumptive interactive components, plus a significant rider-selection database of activities, ideas, information, and accommodation - designed to add to the motorcycle rider’s experiences. Again the study maps consumer traffic connections and the cycles of actions they pursue. Hence, we test the WAM artifact using staged component additions across the development cycle of this commercial website. We stepwise expand the motivation, consumption, and user-gratification component areas that likely appeal to motorcycle riders.

Lastly, stage six moves the website beyond this study and into a commercial management and national reach status - where future competitive analysis is limited.

At each stage of [www.therideguide.com.au](http://www.therideguide.com.au)’s development its function, domain and website scores are calculated using the above formulae (from the rough-cut mutually-exclusive set-theory and causation theory section). Table 3 shows the nine Aesthetic Domain scores (A1 to A9) grouped across the five stages of our planned website development (as the number of components possible per function, the number of components present per function, the normed score per function and the domain totals).

Table 4 provides the three website domain scores at each stage of development for [TheRideGuide.com.au](http://TheRideGuide.com.au), along with the corresponding benchmark score and the corresponding percentage of the total score possible.

**Table 3.** Aesthetic Domain calculations

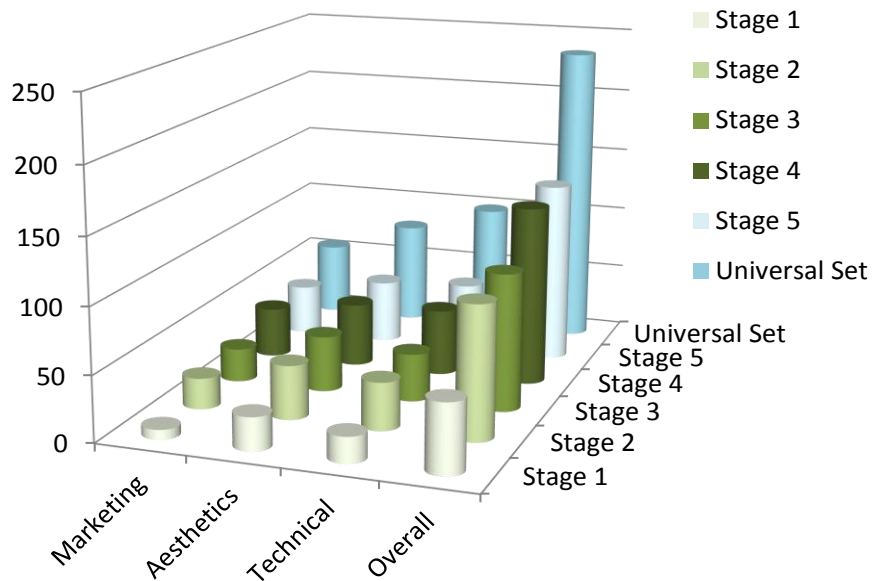
Function	Possible Components	Present Components					Normed out of 10 Component Score				
		Stage 1	Stage 2	Stage 3	Stage 4	Stage 5	Stage 1	Stage 2	Stage 3	Stage 4	Stage 5
A1	9	3	8	8	8	8	3.33	8.89	8.89	8.89	8.89
A2	7	4	5	5	6	6	5.71	7.14	7.14	8.57	8.57
A3	5	0	0	0	0	0	0	0	0	0	0
A4	4	1	2	2	3	3	4	5	5	7.5	7.5
A5	5	3	3	3	4	4	6	6	6	8	8
A6	4	0	0	0	0	0	0	0	0	0	0
A7	4	2	4	4	4	4	5	10	10	10	10
A8	7	1	3	4	4	4	1.43	4.29	5.71	5.71	5.71
A9	8	0	0	0	0	0	0	0	0	0	0
<b>Total</b>	<b>53</b>	<b>14</b>	<b>25</b>	<b>26</b>	<b>29</b>	<b>29</b>	<b>25.47</b>	<b>41.32</b>	<b>42.74</b>	<b>48.67</b>	<b>48.67</b>

**Table 4.** Domain and Stage Calculations

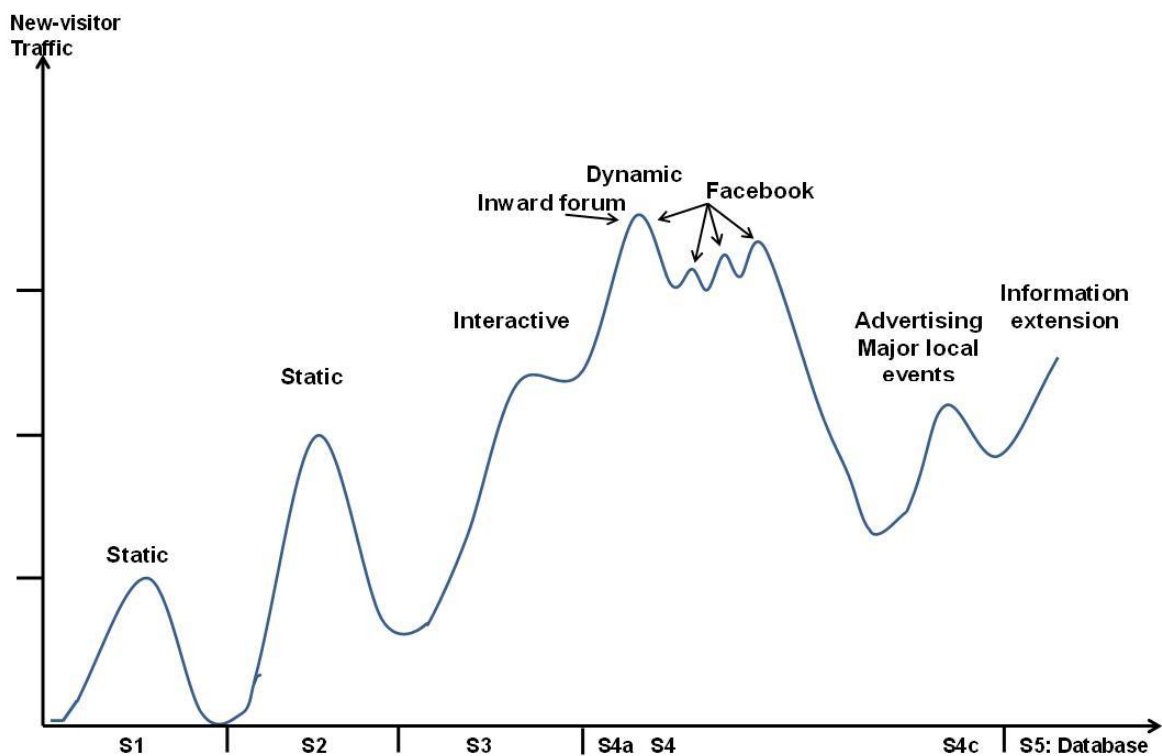
Domain	Stage 1	Stage 2	Stage 3	Stage 4	Stage 5
Aesthetic	25.47	41.32	42.74	48.67	48.67
Technical	19.85	36.59	36.59	50.72	52.66
Marketing	7.57	23.69	25.6	38.33	38.33
Benchmark Score	52.89	101.60	104.93	137.72	138.55
Possible score %	18.89	36.28	37.48	49.19	49.48

The visual effects recognition of this [TheRideGuide.com.au](http://TheRideGuide.com.au) website summarizes the internal website analysis against the WAM universal set. This ([www.therideguide.com.au](http://www.therideguide.com.au)) website enlists fewer components than available in the WAM universal set. After WAM comparison, and focus group feedback, management then selects which missing components are to be added to the website, and at what stage. These focus groups include commercial website developers, website marketing executives, motorcycle riders, and academics. Hence, management-guided changes to the website, whilst introducing subjectivity, do add website

components, and each of these changes yield higher WB scores (Tables 3 and 4, and Figure 3 and 4).



**Figure 3.** Effects recognition (www.therideguide.com.au)



**Figure 4.** Longitudinal tracking: www.therideguide.com.au’s website changes

Figure 4 summarizes the above website change process as a timeline against first-time-consumer/user traffic. Stage one’s single page ‘Brochure’ shows visitors but soon loses its appeal. Stage two’s static five page approach doubles stage ones first-time-consumer/user

traffic, however again quickly loses most of its peak consumer market. Stage three's interactive five page approach wins, and likely retains more consumers. It adds the basic motivations of communications, static maps, additional ride details, and a YouTube channel.

After stages one, two, and three, consumer numbers initially peak each time with more components winning more consumers, but with no ongoing change in the website these numbers decline. However, an increasing, but minimal number of consumers still continue to visit the website as the twelve weeks per stage study continues. Consumer traffic shows this is due to the website's growth in the breadth of its domestic and international market reach over time.

Stage four sees the introduction of broad consumer interactivity. Here inward links from appropriate forums and a Facebook site rapidly build the consumer market to a new peak (more than triple first-time-consumers at stage one). In stage four there is a lot of activity. At 4a traffic growth occurs quickly, however without continued Facebook postings, the traffic again begins to fall. Three subsequent and closely-spaced Facebook postings at 4b generate three small peaks in traffic – with each peak higher than the previous one. This indicates that continual closely-spaced new Facebook postings do continue to draw new consumers to the website. Once again with no additional links, forums or Facebook activity, first-time-consumer numbers again decline rapidly. Thus, social media communication approach can be used to engage with like-minded (or similarly behaving) consumers, and it can be used to drive additional consumer traffic to the website.

The recognizable peak in the data at 4c (prior to stage five update) is produced by visitor 'overflow' from several internationally-advertised major sporting events occurring in this study's region. This indicates new consumer traffic can result from regional activities - especially ones that appear to be complementary, even without building behavioral connections (such as showing these consumers that this website provides great regional images and alternate ways to see some regional attractions).

Lastly stage five adds a useful database of accommodation, eateries, and attractions, but without advertising this is slowly sourced by consumers, and so the growth in new visitors is slower but still significant.

Hence, the addition of website components, specifically designed to target a chosen consumer-group, results in an increase in first-time-consumer traffic over time. This traffic can be increased when more components and more personal interactivity are both added to the website. However when forums, inbound links and social media (Facebook) are each behaviorally-focused at this website's consumer group traffic can rapidly increase.

When a major event aligns to the website's target market the website should cross-promote its complementary offerings to the major event's consumers, and should pursue winning their positive word-of-mouth comments, capturing additional website traffic.

### **Implications of Research**

#### *Theoretical implications*

To be competitive, a corporate's website requires ongoing development, and so must deliver a suitable return-on-investment. However to understand what and how to benchmark first the website must be accurately scored, and then interpreted to competitively draw its consumers into making behavioral decisions regarding the corporate's transaction offerings.

A solution to this website benchmarking problem (Hevner *et al.*, 2004) is developed under a DSRM approach - using set-theory (Boisvert and Caron, 2006) and Aristotle's four-step theory-of-causation (Falcon, 2011) in combination with a causal typological collation of literature described components (that are point-in-time, multi-level, and objectively (present/absent) based). These components are likeness-grouped first into functional subsets, and then into the larger domain subsets of the WAM artifact. Thus, by adding new

component measures to websites, researchers can further develop the WAM artifact (and its subsets of website components). The artifact can also be used to generate component-recognition algorithms, and researchers can investigate these for specific internal or external industry requirements; such as quality, performance, servicing, loyalty and/or competitiveness.

### *Practical implications*

A business benchmarking only against its competitors tends to advance its website sufficiently to cover its comparative weakness (Kim and Mauborgne, 2005). The WAM artifact holds its components within 28 functions and 3 domains of the website. As each component is programmable a complete WB *software solution* can be established against the available universal set of components (or against other chosen external websites). This deployable software approach allows researchers and management to use the WAM components to reposition their website well beyond those of their competition - and into an untapped or 'blue ocean' market space (Kim and Maurborgne, 2005). This shift is applied at activity four of the DSRM.

## **Future Research**

### *Measurement aspects*

With continual literary updates the WAM artifact's components require constant monitoring to remove obsolete components, and to add new components (such as socially-interactive measures develop and new social domain), and also to enhance replacement components. To ensure the WB algorithm remains current researchers must incorporate these alterations into the WAM artifact.

### *Theoretical aspects*

Websites differ in purpose. Hence, not every WAM component is necessary, suitable, or required, for every website. Currently, industry-specific WB is inadequately and inconsistently researched. Thus the WAM artifact may be deployed for industry-specific targeting/niche comparison studies. This is a rich area for future and refined WB industry studies.

### *Management aspects*

Previous WB studies typically engage limited numbers of website components and just concentrate on specific areas (Kim *et al.*, 2009). Other WB approaches do not deliver timely results - forcing management to implement changes post hoc (Pang *et al.*, 2009). Being adaptable to an industry's target requirements, researchers can use the WAM artifact as an efficient and effective monitoring tool, and may even suggest new solutions (as scenarios or options) to managers of the corporate.

## **Conclusion**

As a valuable global reach resource, the website should be a consumer-connecting window, and should help to maintain a corporate's competitiveness. Researchers have spread their studies across well over 150 measures (termed components), but outside this DSRM study, no approach engages a universal set of website components. The WAM artifact, developed through DSRM, is comprehensive and robust (literature-based), programmable (present/absent), adaptable (business/consumer specific), and expandable (new component inclusions). It solves a WB problem, fills a gap in the WB literature, and provides a framework for comprehensive WB assessments linked to relevant information systems and management studies.



This WB problem is solved through an objective and programmable approach using rough cut set-theory and mutual-exclusivity applied at a specific point-in-time. When combined with the theory-of-causation website components are literature linked into functions and domains and WB scores. The WAM artifact's WB scores can be tactically improved when pursued in conjunction with theories encompassing planned-behavior, motivation, consumption and user-gratification, and selectively updated using behaviorally-emphasized new (or remodeled) components, functions and domains.

## References

- Abascal, J., Arrue, M., Fajardo, I., Garay, N. and Tomás, J. (2004), "The use of guidelines to automatically verify web accessibility", *Universal Access in the Information Society*, Vol. 3 No. 1, pp. 71-79.
- Adler, P.S., Benner, M., Brunner, D.J., MacDuffie, J.P., Osono, E., Staats, B.R., Takeuchi, H., Tushman, M.L. and Winter S.G. (2009), "Perspectives on the productivity dilemma", *Journal of Operations Management*, Vol. 27 No. 2, pp. 99–113.
- Albaum, G., Evangelista, F. and Medina, N. (1998), "Role of response behavior theory in survey research: a cross-national study", *Journal of Business Research*, Vol. 42 No. 2, pp. 115-125.
- Albert, T. C., Goes, P. B. and Gupta, A. (2004), "GIST: a model for design and management of content and interactivity of customer-centric web sites", *MIS Quarterly*, Vol. 28 No. 2, pp. 161-182.
- Barnes, S.J. and Vidgen, R.T. (2002) "An integrative approach to the assessment of e-commerce quality", *Journal of Electronic Commerce Research*, Vol. 3 No. 3, pp. 114-127.
- Beirekdar, A., Vanderdonckt, J. and Noirhomme-Fraiture, M. (2002), "KWARESMI—Knowledge-based Web Automated Evaluation with REconfigurable guidelineS optiMization", in the 9th International Workshop on Design, Specification, and Verification of Interactive Systems DSV-IS2002, 2002, Springer-Verlag London, UK.
- Boisvert, H. and Caron, M. (2006), "Benchmarking web site functions", *Benchmarking: An International Journal*, Vol. 13 No. 1/2, pp. 174-189.
- Bowers, N. (1996), "Weblint: quality assurance for the world wide web", *Computer Networks and ISDN Systems*, Vol. 28 No. 7 to 11, pp. 1283-1290.
- Calero, C., Ruiz, J. and Piattini, M. (2005), "Classifying web metrics using the web quality model", *Online Information Review*, Vol. 29 No. 3, pp. 227-248.
- Cao, M., Zhang, Q. and Seydel, J. (2005), "B2C e-commerce web site quality: an empirical examination", *Industrial Management and Data Systems*, Vol. 105 No. 5, pp. 645-661.
- Cassidy, L. (2010), "Website Benchmarking: Evaluating Scaled and Dichotomous Approaches", Honours Thesis, James Cook University, Cairns, Qld. 1-89.
- Cassidy, L., and Hamilton, J. (2011) "Website benchmarking: theoretical and measurement aspects", in *Electronic Business: Borderless E-Business for the Next Decade 2011 proceedings of the 11th International Conference in Bangkok, Thailand, 2011*, pp. 413-416.
- Chen, Y.-H. and Barnes, S. (2007), "Initial trust and online buyer behavior", *Industrial Management and Data Systems*, Vol. 107 No. 1, pp. 21-36.
- Choi, S. and Morrison, A.M. (2005), "Website effectiveness for bricks and mortar travel retailers", *International Journal of Tourism and Hospitality Research*, Vol. 16 No. 1, pp. 63-78.

- Cox, J. and Dale, B.G. (2002), "Key quality factors in web site design and use: an examination", *International Journal of Quality and Reliability Management*, Vol. 19 No. 7, pp. 862-888.
- Deng, J. (1989), "Introduction to grey system theory", *The Journal of Grey System*, Vol. 1 No. 1, pp. 1-24.
- Deng, L. and Poole, M.S. (2010), "Affect in web interfaces: a study of the impacts of web page visual complexity and order", *MIS Quarterly*, Vol. 34 No. 4, pp. 711-730.
- Elling, S., Lentz, L. and De Jong, M. (2007), "Website evaluation questionnaire: development of a research-based tool for evaluating information websites", in *Electronic Government 2007 proceedings of 6th International Conference in Berlin, Heidelberg*, pp. 293-304.
- eMarketer. (2014), "Global B2C e-commerce sales to hit \$1.5 trillion this year driven by growth in emerging markets", available at: <http://www.emarketer.com/Article/Global-B2C-Ecommerce-Sales-Hit-15-Trillion-This-Year-Driven-by-Growth-Emerging-Markets/1010575> (accessed 4 June 2014).
- Evangelista, F., Poon, P. and Albaum, G. (2012), "Using response behavior theory to solicit survey participation in consumer research: an empirical study", *Journal of Marketing Management*, Vol. 28 No.9/10, pp. 1174-1189.
- Falcon, A. (2011), "Aristotle on Causality", in *The Stanford Encyclopedia of Philosophy*, E. N. Zalta (ed.) available at: <http://plato.stanford.edu/archives/fall2011/entries/aristotle-causality> (accessed 2 June 2013).
- Fang, Y., Qureshi, I., Sun, H., McCole, P., Ramsey, E. and Lim, K.H. (2014), "Trust, satisfaction, and online repurchase intention: the moderating role of perceived effectiveness of e-commerce institutional mechanisms", *MIS Quarterly*, Vol. 38 No. 2.
- Flavian, C., Guinaliu, M. and Gurrea, R. (2006), "The role played by perceived usability, satisfaction and consumer trust on website loyalty", *Information and Management*, Vol. 43 No. 1, pp. 1-14.
- Gefen, D., Karahanna, E. and Straub, D.W. (2003), "Trust and TAM in online shopping: an integrated model", *MIS Quarterly*, Vol. 27 No. 1, pp. 51-90.
- Hamilton, J. (2010), "Rating the website: an objective approach", *International Journal of Electronic Business*, Vol. 9 No. 4, pp. 1-24.
- Hamilton, J.R., and Tee, S. (2012), "Growing the marketspace: driving the social media networks to websites", in *12th International Conference on Business 2012 proceedings of the international conference 2012*, Honolulu, Hawaii, pp.. 628-632.
- Herrero, A. and San Martin, H. (2012), "Developing and Testing a Global Model to Explain the Adoption of Websites by Users in Rural Tourism Accommodations", *International Journal of Hospitality Management*, Vol. 31 No. 4, pp. 1178-1186.
- Hevner, A. and Chatterjee, S. (2010), *Design Research in Information Systems: Theory and Practice*, Springer: Dordrecht.
- Hevner, A.R., March, S.T., Park, J. and Ram, S. (2004), "Design science in information systems research", *MIS Quarterly*, Vol. 28 No. 1, pp. 75-105.
- Ho, C.-T. and Wu, Y.-S. (2006), "Benchmarking performance indicators for banks", *Benchmarking: An International Journal*, Vol. 3 No. 1/2, pp. 147-159.
- Ivory, M.Y. and Hearst, M.A. (2001), "The state of the art in automating usability evaluation of user interfaces", *ACM Computing Surveys*, Vol. 33 No. 4, pp. 470-516.
- Ivory, M.Y. and Megraw, R. (2005), "Evolution of web site design patterns", *ACM Transactions on Information systems*, Vol. 23 No. 4, pp. 463-497.
- Kaplan, R. and Norton, D. (1992), "The balanced scorecard - measures that drive performance", *Harvard Business Review*, Vol. 70 No. 1, pp. 71-79.

- Kim, H.-W., Chan, H.C. and Gupta, S. (2007), "Value-based adoption of mobile internet: an empirical investigation", *Decision Support Systems*, Vol. 43 No. 1, pp. 111-126.
- Kim, J., Kim, M. and Kandampully, J. (2009), "Buying environment characteristics in the context of e-service", *European Journal of Marketing*, Vol. 43 No. 9/10, pp. 1188-1204.
- Kim, C. and Mauborgne, K. (2005), *Blue Ocean Strategy*, Harvard Business School Press: Boston, MA.
- Kline, S.F., Morrison, A.M. and St. John, A. (2004), "Exploring bed and breakfast websites: a balanced scorecard approach", *Journal of Travel and Tourism Marketing*, Vol. 17 No. 2-3, pp. 253-267.
- Krosnick, J.A. (1991), "Response strategies for coping with the cognitive demands of attitude measures in surveys", *Applied Cognitive Psychology*, Vol. 5 No. 3, pp. 213-236.
- Krosnick, J.A. and Alwin, D.F. (1987), "An evaluation of a cognitive theory of response-order effects in survey measurement", *Public Opinion Quarterly*, Vol. 51 No. 2, pp. 201-219.
- Krosnick, J.A., Holbrook, A.L., Berent, M.K., Carson, R.T., Hanemann, W.M., Kopp, R.J., Mitchell, R.C., Presser, S., Rudd, P.A., Smith, V.K., Moody, W.R., Green, M.C. and Conaway, M. (2002), "The impact of "no opinion" response options on data quality: non-attitude reduction or an invitation to satisfice?", *Public Opinion Quarterly*, Vol. 66 No. 3, pp. 371-403.
- LaRose, R. and Eastin, M.S. (2004), "A social cognitive theory of internet uses and gratifications: toward a new model of media attendance", *Journal of Broadcasting and Electronic Media*, Vol. 48 No. 3, pp. 358-377.
- Lee, Y., Chen, A.N.K. and Ilie, V. (2012), "Can online wait be managed? The effect of filler interfaces and presentation modes on perceived waiting time online", *MIS Quarterly*, Vol. 36 No. 2.
- Lee, Y. and Kozar, K.A. (2009), "Designing usable online stores: a landscape preference perspective", *Information and Management*, Vol. 46 No. 1, pp. 31-41.
- Lee, J. and Morrison, A.M. (2010), "A comparative study of web site performance", *Journal of Hospitality and Tourism Technology*, Vol. 1 No. 1, pp. 50-67.
- Lee, S.-M., Ungson, G.R. and Russo, M.V. (2011), "What determines an engaging website? An empirical study of website characteristics and operational performance", *Journal of High Technology Management Research*, Vol. 22 No. 1, pp. 67-79.
- Lin, K.-Y., and Lu, H.-P. (2011), "Why people use social networking sites: an empirical study integrating network externalities and motivation theory", *Computers in Human Behavior*, Vol. 27 No.3, pp. 1152-1161.
- Loiacono, E.T., Watson, R.T. and Goodhue, D.L. (2007), "WebQual: an instrument for consumer evaluation of web sites", *International Journal of Electronic Commerce*, Vol. 11 No. 3, pp. 51-87.
- Luo, J., Ba, S. and Zhang, H. (2012), "The effectiveness of online shopping characteristics and well-designed websites on satisfaction", *MIS Quarterly*, Vol. 36 No. 4, pp. 1131-1144.
- Luo, M.M., Chea, S. and Chen, J.-S. (2011), "Web-based information service adoption: a comparison of the motivational model and the uses and gratifications theory", *Decision Support Systems*, Vol. 51 No. 1, pp. 21-30.
- McKinney, V., Yoon, K. and Zahedi, F.M. (2002), "The measurement of web-customer satisfaction: an expectation and disconfirmation approach", *Information Systems Research*, Vol. 13 No. 3, pp. 296-315.
- Moriarty, J. (2011), "A theory of benchmarking", *Benchmarking: An International Journal*, Vol. 18 No. 4, pp. 588-612.

- Nathan, R.J. and Yeow, P.H.P. (2011), "Crucial web usability factors of 36 industries for students: a large-scale empirical study", *Electronic Commerce Research*, Vol. 11 No. 2, pp. 151-180.
- Nel, D., Niekerk, R. v., Berthon, J.-P. and Davies, T. (1999), "Going with the flow: web sites and customer involvement", *Internet Research: Electronic Networking Applications and Policy*, Vol. 9 No. 2, pp. 109-116.
- Nusair, K.K. and Kandampully, J. (2008), "The antecedents of customer satisfaction with online travel services: a conceptual model", *European business Review*, Vol. 20 No. 1, pp. 4-19.
- O'Cass, A. and Ngo, L.V. (2012), "The antecedents of customer satisfaction with online travel services: a conceptual model", *Industrial Marketing Management*, Vol. 41 No. 1, pp. 125-135.
- Olsina, L. and Rossi, G. (2002), "Measuring web application quality with WebQEM", *Multimedia IEEE*, Vol. 9 No. 4, pp. 20-29.
- Page, C. and Lepkowska-White, E. (2002), "Web equity: a framework for building consumer value in online companies", *Journal of Consumer Marketing*, Vol. 19 No. 3, pp. 231-248.
- Pang, M.-S., Suh, W., Kim, J. and Lee, H. (2009), "A benchmarking-based requirement analysis methodology for improving websites", *International Journal of Electronic Commerce*, Vol. 13 No. 3, pp. 119-162.
- Peppers, K., Tuunanen, T., Rothenberger, M.A. and Chatterjee, S. (2007), "A design science research methodology for information systems research", *Journal of Management Information Systems*, Vol. 24 No. 3, pp. 45-77.
- Schwarz, N. (2007), "Cognitive aspects of survey methodology", *Applied Cognitive Psychology*, Vol. 21 No. 2, pp. 277-287.
- Simmons, G., Armstrong, G.A. and Durkin, M.G. (2008), "A conceptualization of the determinants of small business website adoption: setting the research agenda", *International Small Business Journal*, Vol. 26 No. 3, pp. 351-389.
- Steelman, Z.R., Hammer, B.I. and Limayem, M. (2014), "Data collection in the digital age: innovative alternatives to student samples", *MIS Quarterly*, Vol. 38 No. 2.
- Udo, G.J., Bagchi, K.K. and Kirs, P.J. (2010), "An assessment of customers' e-service quality perception, satisfaction and intention", *International Journal of Information Management*, Vol. 30 No. 6, pp. 481-492.
- Vilenkin, N.Y. (1986), *Stories about Sets*, Academic Press: New York, NY.
- Wakefield, R.L., Wakefield, K.L., Baker, J. and Wang, L.C. (2011), "how website socialness leads to website use", *European Journal of Information Systems*, Vol. 20 No. 1, pp. 118-132.
- Webb, H.W. and Webb, L.A. (2004), "SiteQual: an integrated measure of website quality", *The Journal of Enterprise Information Management*, Vol. 17 No. 6, pp. 430-440.
- Wells, J.D., Valacich, J.S. and Hess, T.J. (2011), "What signal are you sending? How website quality influences perceptions of product quality and purchase intentions", *MIS Quarterly*, Vol. 35 No. 2, pp. 373-396.
- Wolfinger, M. and Gilly, M.C. (2003), "eTailQ: dimensionalizing, measuring and predicting eTail quality", *Journal of Retailing*, Vol. 79 No. 3, pp. 183-198.
- Xu, J., Benbasat, I. and Cenfetelli, R.T. (2013), "Integrating service quality with system and information quality: an empirical test in the e-service context", *MIS Quarterly*, Vol. 37 No. 3, pp. 777-794.
- Yang, T., Linder, J. and Bolchini, D. (2012), "DEEP: design-oriented evaluation of perceived usability", *International Journal of Human-Computer Interaction*, Vol. 28 No. 5, pp. 308-346.

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- Yoo, B. and Donthu, N. (2001), "Developing a scale to measure the perceived quality of an internet shopping site (SITEQUAL)", *Quarterly Journal of Electronic Commerce*, Vol. 2 No. 1, pp. 31-47.
- Yoon, C. and Kim, S. (2009), "Developing the causal model of online store success", *Journal of Organizational Computing and Electronic Commerce*, Vol. 19 No. 4, pp. 265-284.
- Zhu, W. (2007), "Topological approaches to covering rough sets", *Information Sciences*, Vol. 177 No. 6, pp. 1499-1508.
- Zhu, Y., Basil, D.Z. and Hunter, M.G. (2009), "The extended website stage model: a study of Canadian winery websites", *Canadian Journal of Administrative Sciences*, Vol. 26 No. 4, pp. 286-300.