

Contents lists available at ScienceDirect

Transportation Research Part E

journal homepage: www.elsevier.com/locate/tre



When blockchain meets social-media: Will the result benefit social media analytics for supply chain operations management?



Tsan-Ming Choi^a, Shu Guo^b, Suyuan Luo^{c,*}

- ^a Business Division, Institute of Textiles and Clothing, The Hong Kong Polytechnic University, Hung Hom, Kowloon, Hong Kong
- ^b School of Management, University of Liverpool, Chatham Street, Liverpool, United Kingdom
- ^c Institute of Big Data Intelligent Management and Decision, College of Management, Shenzhen University, Shenzhen, China

ARTICLE INFO

Keywords: Social media analytics Methods Blockchain technology Operations management Case studies

ABSTRACT

Social media analytics is very critical in modern supply chain operations management (SCOM). However, in terms of methods, conducting social media analytics (SMA) for SCOM faces many challenges. Problems such as data accuracy (e.g., fake data), user privacy, data security, etc. are all present. Recently, with the emergence of blockchain technology (BCT), many new social media apps and platforms are developed. Motivated by the fact that (i) SMA is important for SCOM, (ii) the traditional social media (TSM) has insufficiency, and (iii) the "blockchain technology supported social media" (BSM) platforms have emerged, we explore whether and how the BSM would enhance social media analytics for SCOM. To be specific, by reviewing papers published in leading SCOM journals, we identify the applications and limitations of SMA for SCOM. Then, we conduct real case studies to examine the shortcomings of the TSM platforms and reveal features of their BSM counterparts. We investigate how the blockchain technology would potentially improve the use of SMA for SCOM. Finally, a future research agenda is proposed.

1. Introduction

1.1. Motivation and technological background

Social media refers to the mobile or network based application, which supports the creation, exchange and access of usergenerated content among members (Batrinca and Treleaven, 2015; Chen et al., 2019). Popular examples of social media include Facebook, Twitter, Youtube, Linkedin, and Instagram. It is reported that at least 80% of Americans have one social media account. In the whole world, it is estimated that there are over 3 billion users of social media platforms like Facebook, Instagram and Twitter (Mire, 2018). It is also commonly believed that social media is the modern way of connecting people and establishing friendship. The well-establishment of social media platforms and their popularity directly create a large amount of social media data. The proper use of social media data can potentially bring benefits to many real world operations and this directly leads to the emergence of social media analytics (SMA). In the literature, SMA (Zeng et al., 2010) refers to the use of analytical capabilities to analyze social media content so as to achieve a specific goal (Holsapple et al., 2014). It is mainly a passive approach which relies on retrieving the available data, instead of creating data proactively.

It is known that SMA helps business operations along the supply chain in many aspects (Wang et al., 2019a). For example, supporting product quality management in Coca Cola (Choi et al., 2018a; Singh et al., 2018), facilitating demand forecasting in businesses

E-mail addresses: jason.choi@polyu.edu.hk (T.-M. Choi), eunice.guo@liverpool.ac.uk (S. Guo), suyuanluo@126.com (S. Luo).

^{*} Corresponding author.

such as IBM and online fashion retailer Rue La La (Choi et al., 2018a; Cui et al., 2018; Lau et al., 2018), speeding up reaction to emergency events such as disasters (Luna and Pennock, 2018), improving engagement of customers (Holsapple et al., 2014), enhancing marketing programs (Lee et al., 2018; Kumar et al., 2019) including luxury fashion brand Louis Vuitton (Choi et al., 2018a; Chiu et al., 2018), reducing wastes (Mishra and Singh, 2018), facilitating information sharing (Wang et al., 2019a), achieving higher levels of customer services, improving business processes, encouraging innovation and yielding operational efficiency (Lam et al., 2016), etc.

However, SMA faces lots of challenges (Brooker et al., 2016). For example, social media includes dataset that is huge in size which makes it difficult to process. Social media data is also notorious of being incomplete (Stieglitz et al., 2018), unstructured (Cui et al., 2018), inaccurate or even fake, which means conducting SMA is an uneasy task. In some cases, social media data stream is dynamic (Zeng et al., 2010) which further creates challenges. Naturally, social media data is well-qualified as big data with all "V"s being satisfied (Choi et al., 2018a; Kalyan et al., 2018). Here, the "V"s include "Volume, Velocity, Variety, Variability, Veracity, Visualization, and Value". Moreover, different social media platforms have different limitations. Thus, conducting SMA is an uneasy task.

Recently, driven by the bitcoin business model and the importance of supply chain transparency, the blockchain technology arises (Chod et al., 2018; Choi et al., 2019c). The blockchain technology is commonly viewed as a distributed ledger operated in a decentralized manner. The core features of the blockchain technology include its ability to keep permanent data, and achieve high transparency and reliability of data (Choi et al., 2019b,c). The blockchain technology also features important functions such as smart contracting² and cryptocurrency which facilitate and support many innovative measures (Choi and Luo, 2019). Fig. 1.1 depicts the common features of blockchain technology. See Appendix A for more details of Fig. 1.1.

As a result, the blockchain technology is known as a disruptive force, which will potentially revolutionalize many business operations. For example, the blockchain technology can be used to help trace the sources and show the details of food (as what Walmart and IBM are exploring). It is already employed for product authentication as what Everledger is doing for diamonds (Choi, 2019). The blockchain technology also helps with ethical sourcing in fashion apparel.³

In addition to the above functions, the blockchain technology is also changing social media and SMA. First, the presence of many cryptocurrency rewarding schemes can entice more users to be engaged with the social media. Second, it gives higher control power of data to users as well as the content's distribution. Third, the available data will be more authentic, reliable and secure. Fourth, the identity of creators can be traced. Thus, the issue of fake data can be addressed. Fig. 1.2 depicts how blockchain technology can be used in a social media platform. Table 1.1 shows some features of the blockchain technology, which are related to social media and SMA. Table 1.2 further summarizes some current blockchain technology supported social media platforms that aim to overcome some shortcomings of the existing traditional social media platform counterparts.

Fig. 1.2 shows how the blockchain technology is used in a social media platform. When User A posts a new thread in the social media platform, a new record is created in his ledger and announced to every node in the blockchain. Other users will verify User A's identity and the authenticity of the thread by the hash function. If the information is true, the HyperLedger will save this new record by generating a new block. What is more, User A can get cryptocurrency to compensate for his contribution from the platform. However, if other users find out that User A is a "fake" user or he is telling lies, this new record will be declined and everyone knows User A tells lies. The credit of User A will be affected (see Fig. 2.1.).

1.2. Research questions and paper's structure

Motivated by the fact that (i) SMA is important for supply chain operations management (SCOM), (ii) SMA faces many challenges, and (iii) the emergence of blockchain technology supported social media, we attempt to address the following open research questions:

- 1. For the social media platforms which have been used in operations analyses in the most authoritative SCOM literature, what are the major functional areas involved?
- 2. With respect to the commonly seen social media platforms employed for social media analytics, are there any problems?
- 3. What problems can blockchain technology solve for social media analytics and how? Which topical areas related to SMA for SCOM will be interesting for future research?

This paper aims to address the above three main research questions and derive the respective important insights.

To the best of our knowledge, this paper is the first study which comprehensively explores the use of blockchain technology to enhance SMA for SCOM. It includes original research (e.g., examining the real world blockchain technology supported social media (BSM) platforms). It also systematically examines the current state-of-the-arts SCOM literature on SMA and provides the latest details of blockchain technology supported social media platforms. The findings are valuable to both the practitioners and academicians. It contributes to the SCOM literature, especially in the SCOM-technology interface (Kumar et al., 2018).

This paper is organized as follows. Section 2 presents the review methodology. Section 3.1 shows the summary results of the systematic review. Section 3.2 reports some SCOM studies related to the blockchain technology. Section 4 presents the traditional

¹ https://impact.com/marketing-intelligence/7-vs-big-data/ (accessed 14 December 2019).

² The smart contracting function refers to the ability of automating the contracting agreements in a speedy and transparent manner, following the pre-specified steps.

³ See DHL Trend Research (2018).

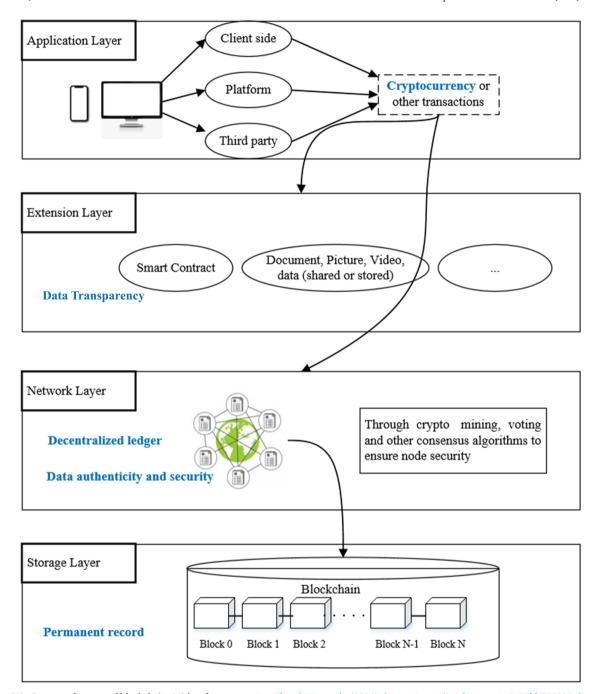


Fig. 1.1. Common features of blockchain. With references to: Ismail and Maerwala (2019). https://www.jianshu.com/p/5e72bb776280. https://blog.csdn.net/qq_35624642/article/details/78138077 (accessed: 26 November 2019).

social media platforms, and how SCOM can be enhanced by SMA conducted on them. TSM examples and features as well as problems of TSM platforms are also discussed. Section 5 examines the blockchain technology supported social media (BSM) platforms. Some real world BSM platforms are introduced and how specifically the blockchain technology can help is to support SMA also discussed. Section 6 presents the future research agenda and Section 7 concludes this paper with comments on research limitations. To enhance readability, a list of abbreviations used in this paper (excluding the ones for journal names) is shown in Table 1.3.

2. Methodology

To answer the research questions raised in Section 1, this paper first conducts a systematic review (Guo et al., 2019) of social

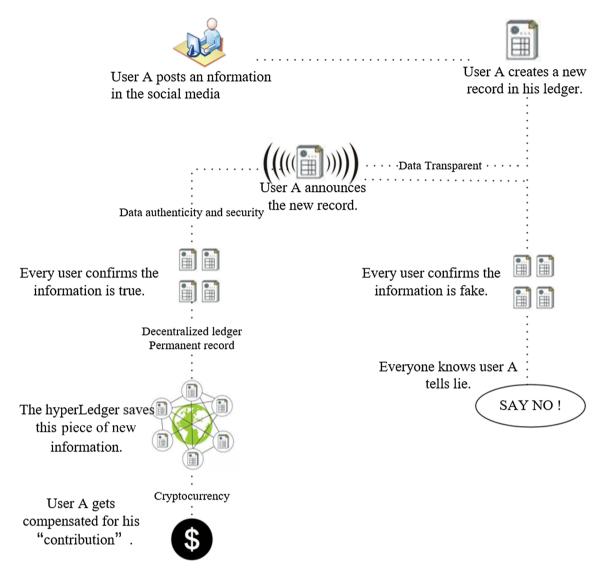


Fig. 1.2. How the blockchain technology is used in a social media platform.

Table 1.1 Features of the blockchain technology which can be applied for social media platforms.

Features	Details
Permanent record	The information cannot be edited, erased, or changed without the consent of all related stakeholders.
Data authenticity and security	The blockchain technology can help keep and share the authentic and secure data publicly which facilitate checking and verification.
Decentralized ledger	Supports parallel computing and computation.
Cryptocurrency	Provides incentive to encourage users' active participation and engagement.
Data transparency	The blockchain technology enhances data transparency.

media analytics related literature in leading SCOM journals. To be specific, for the selection of SCOM journals, we make reference to the SCM journal list, the UK's ABS journal list (levels 4 and 4*), the USA's UTD journal list, and the Australian ABDC A* journals related to logistics and supply chain management. Thus, the following journals are hence included: Decision Sciences (DSJ), European Journal of Operational Research (EJOR), International Journal of Operations and Production Management (IJOPM), Journal of

⁴ https://jindal.utdallas.edu/the-utd-top-100-business-school-research-rankings/index.php (accessed: 26 November 2019).

⁵ http://www.scmlist.com/ (accessed: 26 November 2019).

Table 1.2
The TSM and BSM counterparts.

Traditional Social Media Platforms (TSM)	The Blockchain Technology Supported Social Media Platform (BSM) Counterparts
Facebook	Foresting
Twitter	Peepeth
Youtube	Dtube
Lindedin	Indorse
Instagram	Steepshot

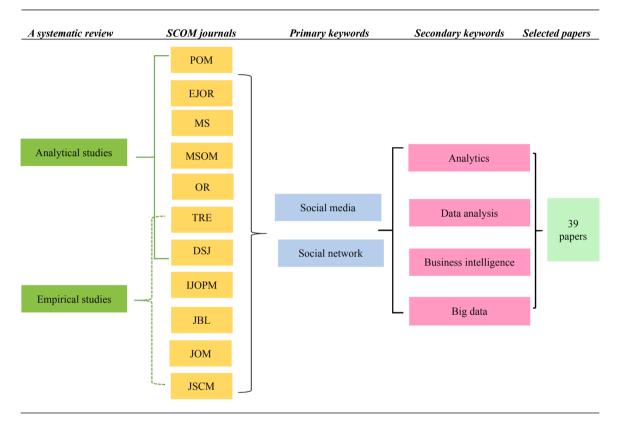


Fig. 2.1. Process of selecting the research papers in this review.

Table 1.3 A list of abbreviations.

Abbreviations	Full Forms	
SMA	Social media analytics	
BCT	Blockchain technology	
SCOM	Supply chain operations management	
TSM	Traditional social media	
BSM	Blockchain technology supported social media	
AI	Artificial intelligence	
RBV	Resource based view	
IPW	Inverse probability weighting	
FS	Flash sales	
API	Application programming interface	
JSON	Javascript object notation	

Business Logistics (JBL), Journal of Operations Management (JOM), Journal of Supply Chain Management (JSCM), Management Science (MS), Manufacturing & Service Operations Management (MSOM), Operations Research (OR), Production and Operations Management (POM), and Transportation Research – Part E (TRE).

Based on this target journal list, a deep searching process was then conducted in summer 2019 through Web of Science. In the searching, we employed the primary keywords of "social media", and "social network" in paper's titles, keywords, and abstracts. This ensures social media analytics is the major focus of the selected papers. In the meantime, no constraint on publication year is imposed throughout the whole searching process, since the objective of this paper is to provide a comprehensive review on the use of SMA for SCOM. Then we examined the definitions of "analytics" as all activities related to the data management. Accordingly, we supplemented the searching process by secondary keywords such as "analytics", "data analysis", "business intelligence", and "big data". We focused on the published original research and filtered out those papers which are literature reviews or discussion based. Finally, 39 papers were selected. Note that we do not find any prior review papers on the topic that we are examining and hence this paper positions itself as the pioneering review and research paper in this new and timely topic.

The distribution of filtered papers is shown in Table 2.1. It is found that most publications appear in DSJ, EJOR, POM and MS, in which DSJ publishes both analytical and empirical studies, and EJOR, POM and MS are more "analytical-oriented". Fewer than half of the collected papers are published in the traditional empirical based journals such as IJOPM, JBL, JOM and JSCM. This finding is rather reasonable as SMA is computational in nature and hence more reviewed studies are in the analytical domain. Of course, it also means there is a big room for having more empirical based studies for SMA in SCOM.

From the collected papers, we identify the reported applications of social media analytics in the current SCOM literature. We uncover the values and problems of SMA with respect to the current social media platforms.

After that, we study and explore how BCT can overcome these problems. To be specific, we conduct original research and examine the BCT supported social media platforms in the real world. We identify their properties and then discuss how the real-world blockchain technology supported social media (BSM) platforms would enhance SMA for SCOM.

3. Literature review

A detailed literature review on the selected papers, which are selected based on the research method mentioned in Section 2, is reported in Section 3.1.

3.1. Social media analytics in SCOM

Social media has been critical in SCOM. Over the past decade, a lot of studies have been reported to explore how the proper use of data from the social media can help enhance real world operations. In the following, from the 35 selected papers published in leading SCOM journals (see Table 2.1), we further check and review the studies which have conducted technical social media data based analytics research in SCOM.⁷

Social influence: As users of social media platforms interact with one another and naturally influence one another, one critical application of SMA is to uncover the social influence in the market, which will have good implications for SCOM. In the literature, Aral and Walker (2014) conduct behavioral experiments on Facebook to explore how marketing is affected by messages sent by users in the social media. They focus on revealing the effects of social influence. Guo et al. (2015) examine social influences, which are critical for marketing strategies, by studying social media like Facebook. The authors categorize six different kinds of social interactions on online social media, in two major categories. They employ the support vector machines approach to explore these two major categories. Gunnec and Raghavan (2017) investigate the share-of-choice (SOC) strategic optimization problem. The authors highlight how to incorporate the critical social network effect, e.g., of Facebook, into SOC analysis formally. They employ a genetic algorithm based method to solve the problem. Srinivasan et al. (2017) study the broadcasting of important life events on online social networks such as Facebook, and Twitter. The authors find that the one-time-life events (e.g., weddings) have a bigger impact than other events. Chen et al. (2019) develop a method to conduct individual-level inference to identify social influences in social media platforms like Facebook, Twitter, and Flickr. They call their proposed method the "learn-simulation-approach", which is based on well-established machine learning models. The authors computationally test the performance of the method with respect to the adoption of new products as well as repeat-purchases.

Online reviews: In addition to social influence, there is no doubt that what users of social media posted online would have important effects on others (Kumar et al., 2019). In particular, online reviews are important reference materials for consumers, e.g., when they plan to visit a restaurant or buy from a retailer. Pekgün et al. (2018) investigate the effects of consumer reviews in social media like blogs and discussion boards, as well as uncertainty in experience. The authors uncover that if consumers put more weights on negative reviews, when the consumer awareness is higher (for the product with a lower quality), the firm will charge higher prices and achieve higher profits than its higher quality competitor. Ramanathan et al. (2017) explore how social media based online customer reviews can be used in marketing to improve the satisfaction of customers. The authors examine this topic from the perspective of retail network designs. They highlight that online reviews on social media (either in Twitter or Facebook) would significantly affect customer satisfaction. Motivated by the common concerns on fake reviews posted online, Dong et al. (in press) study features of online reviews. The authors argue that consumer trust in online reviews is critically important. With the use of

⁶ As a remark, Web of Science is chosen as the searching engine for this paper, which is very popular in the academic domain (e.g., Neely (2005), and Guo et al. (2018)) because of its SCI/SSCI database's completeness. Besides, note that all the target journals are listed in Web of Science.

⁷ Some theoretical studies which examine the role played by SMA in the companies and the respective influences are not reviewed in this section as they only treat SMA as a resource or dynamic capability.

Table 2.1No. of SMA related papers collected from each journal (listed in an alphabetical order).

Journal	Mainly: Empirical (E) or Analytical (A)?	Number of SMA Related Papers
DSJ	Е, А	6
EJOR	A	4
IJOPM	E	3
JBL	E	3
JOM	E	2
JSCM	E	1
MS	A	6
MSOM	A	1
OR	A	3
POM	A	6
TRE	E, A	4

behavioral experiments, the authors uncover that reviews which are positive, factual, and appearing on social networks (like in Facebook posts) are more important and perceived as more trustworthy than the reviews which are negative, emotional, and showing on the seller's website.

Advertising and marketing promotion: One most natural use of SMA is for marketing activities such as advertising and promotion. This is probably also the most common application area of SMA. In the literature, Bimpikis et al. (2016) explore via gametheory the optimal advertising strategy to individuals in social networks. The work is motivated by the fact that social media provides lots of data of consumer preferences which can facilitate target marketing. The authors establish the optimal advertising policy and show that it depends on the structure of the respective social network. Lobel et al. (2017) examine the use of social media's referral, such as a link or a Facebook post, to attract new customers. They derive the optimal payment scheme and explore how a linear functional form might serve as an approximation. Lee et al. (2018) use Facebook and the respective data to reveal the effect of social media advertisement on the level of consumer engagement. The authors make use of natural language processing algorithms to study the data. The authors uncover that direct-information content, such as the quotation of price, yields a lower level of consumer engagement when included in isolated messages while it leads to a higher level of consumer engagement level when provided together with brand specific and associated attributes. Ghose et al. (2019) study the consumer footprints in the online search-engine portals. They build an econometric model to investigate how consumer preferences can enhance their experience. The authors put the consumer preferences and searching costs together in the presence of both the online search engine and social media platform. Gu and Ye (2014) study the company's responses to customers via social media. The authors employ data retrieved from a Chinese largescale online travel agency. They find that online social media management responses are effective especially for the customers with low satisfaction. Chen et al. (2015) develop a hierarchical-ensemble-learning (HEL) model to study customer response, given the large amount of user data from social media nowadays (like Twitter or other social media sites). The authors use a micro-blog dataset to prove the performance of the HEL. Besbes et al. (2016) study the online-content-recommendation system, which directs online surfers who read an article to the related content. The authors explore more than the click-through rate but also the engagement.

Logistics management: Traditionally, logistics management is a fundamental part of SCOM. The use of SMA also plays a role in supporting some activities in logistics management. Fry and Binner (2016) explore how social media (like Twitter, and other news websites) can be used for emergency logistics and evacuations. The authors propose that human behaviors are critical and develop a Bayesian statistical model based algorithm to achieve the optimal evacuation with the consideration of worsening situations. Yoo et al. (2016) study the application of social media for quick information dissemination for humanitarian events. The authors conduct an empirical field study to reveal how effectively information can spread over the social media network. They also use a dataset from Twitter for the Hurricane Sandy event for the investigation. Lam et al. (2016) argue that the social media adoption by companies (e.g., the Delta Air Lines' social media adoption on Twitter) may enhance information sharing and knowledge creation in the supply chain. Using the resource-based-view (RBV) theory on information capability of companies, they study how social media initiatives affect operational innovativeness and performance. Service recovery is important in operations. Fan and Niu (2016) study via exploring Tweeter the critical factors which may influence the effectiveness of service recovery. The authors highlight how the speed of recovery and the degree of severity of failure in service recovery relate to the use of social network. Hou et al. (2018) explore the use of social media data to enhance "delivery service" operations for e-tailers. The authors employ a game-theoretical approach to conduct the analysis. Singh et al. (2018) investigate the use of social media data from twitter to identify critical issues in supply chain logistics management for the food industry. The authors apply the "support vector machine (SVM)" to conduct text mining.

Forecasting: Demand prediction and sales forecasting are both critical in SCOM. In order to do a good job in demand/sales forecasting, companies have to learn from the market and collect the timeliest information. The social media hence provides the important channel for collecting valuable market data. Ballings and Van den Poel (2015) study social media based customer relationship management. The authors study with respect to Facebook and highlight whether in terms of demand forecasting, their proposed algorithm can predict increases in Facebook usage frequency. Lau et al. (2018) explore how big data can be uses to enhance sales forecasting. The authors develop an efficient sentiment analysis scheme to analyze online comments (e.g., retrieved from Twitter, or Weibo) by consumers on products. They apply co-evolutionary extreme learning machines to explore the sales forecasting performance. Choi (2018) studies the case in which there is a "boundedly-rational" fashion retailer implementing quick response (Choi et al., 2018b). Building the model based on the Bayesian theoretical framework, the author analytically uncovers when

manipulating "social media comments" is beneficial to the fashion manufacturer with the use of a "surplus sharing contract". Cui et al. (2018) study how the use of social media data can help to improve sales forecasting. The authors work with a fashion e-tailer to investigate how daily sales forecasting can be improved by using Facebook data. They argue that the accuracy of sales forecasting is improved with a mean absolute percentage error (MAPE) of around 12%.

<u>Risk analysis:</u> Risk relates to uncertainty and negative events. For risk related to human, the social media platforms provide some hints regarding the occurrence of some events. In the literature, <u>Klausen et al.</u> (2018) study how to use social media data for identifying extremists such as cyber-bullies and terrorists. The authors make use of Twitter accounts associated with terrorists to help search for the extremist users and develop the optimal searching policy.

Product introduction and design: Introducing new products and establishing new product designs are two important yet challenging topics in SCOM. Usually, companies do not know what customers actually want and hence there is no guarantee for the success of new product introduction or the launching of a newly designed product. SMA does provide a way for us to know the market and companies may even "test the water" by observing market responses towards the new product or design via the social media. In the literature, Aral and Walker (2011) study how firms can employ word-of-mouth as a way to influence peers and affect product design. The authors conduct a field study involving over one million peers of over 9000 experimental users of Facebook. The authors build a model for using randomized trials to uncover peer influences in social media. Abedi et al. (2014) study the optimal design of the retail network and marketing measures for new product introduction with the consideration of word of mouth. The authors derive the solution algorithms and characterize the optimal solution.

<u>Medical treatment:</u> SMA also plays a role in healthcare related activities. In particular, social media facilitates information exchange among individuals. Yan et al. (2019) study the scenario when patients have access to the health data of other patients and share experiences. The authors employ the theory of social-information-processing to show that positive comments and experiences, which may be shared through Twitter or other popular social sites, would affect the perceived treatment outcomes of patients.

For other applications of SMA, papers on SMA methods and remarks on the relationship between SMA and big data, refer to Appendix B.

Table 3.1 shows the summary of the papers reviewed under each topic/area, the corresponding methods employed for the social media analytics related analyses, and some insights developed in terms of the methods. From Table 3.1, we can see that the majority of studies either employ machine learning approaches or the statistical based methods in conducting SMA studies. Traditional optimization methods are also commonly seen. These three categories of scientific methods are hence the most commonly used and important ones for SMA. In addition, most topics/areas are highly related to the "market". This is intuitive as social media platforms are the places for consumers in the market to express their views and for sellers to observe their behaviors. As modern SCOM is highly market driven (i.e. adopting the "pull process"), it makes full sense that the most commonly seen applications of SMA include forecasting, online reviews, product introduction and other marketing related operations analysis.

3.2. Blockchain technologies supported operations

Section 3.1 reports the systematic review on SMA. As this paper proposes the use of blockchain technology which is critical to enhance SMA, we review some related studies as follows. Note that as the field is very new, only very limited studies are currently published in SCOM. In the context of blockchain technology, Babich and Hilary (in press) comprehensively propose how the blockchain based technologies could be applied in SCOM. The authors propose many areas which they believe to be promising to explore. Chod et al. (2018) investigate the use of blockchain technology in supporting the financial aspects in the supply chain. The authors include features of blockchain operations (such as cryptocurrency) in their model. Wang et al., 2019b propose in the form of a discussion paper how blockchain may affect future operations of supply chains. The authors highlight the application of the theory of sense-making in their analysis. They conduct various personal interviews with several practitioners and experts related to traditional supply chain operations. They focus on revealing how the theory of sense-making can explain managers' behaviors with the use of blockchain. Saberi et al. (2019) investigate one important functionality of blockchain, namely the smart contracting mechanism. The authors focus on how the blockchain technology may help to develop sustainable supply chains. Most recently, Choi (2019) studies the diamond authentication issue in the supply chain with the use of blockchain. The author analytically examines the impact of blockchain technology based diamond authentication schemes under different supply chain structures. This paper also examines the blockchain technology but the focal point is on its impacts on social media and SMA. This is totally different from the current reported studies in the literature.

4. Traditional social media (TSM) without blockchain technology

4.1. Enhancing SCOM by social media analytics

From the literature review, we have identified various areas where SMA has been explored in SCOM. In particular, we can easily see that most studies focus on sales/demand forecasting, promotion and advertisement, and logistics management. Other areas such as risk analysis (Asian and Nie, 2014; Choi and Lambert, 2017; Niu and Zou, 2017; Chiu et al., 2018; Choi et al., 2018b; Choi and Liu, 2019; Tian et al., 2019) for terrorists, stock returns forecasting, and product design are relatively under-studied.

In addition, as expected, the two major social media platforms that have been utilized for SMA are Facebook and Twitter. Table 4.1 summarizes the social media platforms which have been employed for studying SMA in the SCOM literature.

Table 3.1 Methods employed in the reviewed SMA studies for business operations.

Gune et al. (2015) Gune et al. (2015) Gune et al. (2015) Gune et al. (2015) Gune et al. (2017) Srinivasan et al. (2017) Chen et al. (2019) Machine learning Chen et al. (2019) Machine learning Chen et al. (2019) Optimization and game theory Chen et al. (2018) Optimization and game theory Chen et al. (2018) Optimization and game theory Chen et al. (2018) Optimization and game theory Advertising and marketing promotion Advertising and marketing promotion Chen et al. (2017) Doing et al. (2017) Doing et al. (2018) Chen et al. (2019) Behavioral experiments and statistics Gune et al. (2019) Behavioral experiments and statistics Chen et al. (2015) Adorntising and marketing promotion Advertising and comparation and game theory Chen et al. (2019) Adorntising and marketing promotion Chen et al. (2019) Behavioral experiments and statistics Chen et al. (2019) Behavioral experiments and stat	Topics/Areas	Papers	Methods and Tools	Insights
Gunec and Gunec and Raghavan (2017) Srinivasan et al. (2017) Straivasan et al. (2019) Chen et al. (2019) Machine learning Chen et al. (2019) Machine learning Chen et al. (2019) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Online reviews Online reviews Pekgün et al. (2018) Optimization and game theory Optimization and game theory Online reviews optimization and game theory Optimization and game theory Optimization and game theory Advertising and Markine learning Optimization and game theory Lobel et al. (2019) Dong et al. (in press) Optimization and game theory Lobel et al. (2019) Dong et al. (in press) Optimization and game theory Lobel et al. (2019) Dong et al. (in press) Optimization and game theory Lobel et al. (2017) Optimization and game theory Lobel et al. (2018) Markine learning Optimization and game theory Lobel et al. (2019) Behavioral experiments and statistics Chen et al. (2018) Markine learning algorithms and computational experiments and statistics Online review and a proper product review to product review to product review to product review to product review operator generators, marking and replaced to help a long train behavior and properators operators, facult reviews, facult reviews, and revie	Social influence		Behavioral experiments and statistics	
Gunne and Raghavan (2017) Srinivasan et al. (2017) Chen et al. (2018) Optimization and genetic algorithm Raghavan (2017) Machine learning Chen et al. (2019) Machine learning Chen et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Online reviews Pekgün et al. (2018) Optimization and game theory Dong et al. (in press) Dong et al. (in press) Behavioral experiments and statistics Advertising and marketing promotion Advertising and Bimpikis et al. (2017) Lobel et al. (2019) Behavioral experiments Gu and Ye (2014) Panel data and statistics Chen et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning algorithms and computational experiments and statistics Chen et al. (2017) Advertising and Gu and Ye (2014) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning algorithms and computational experiments and computational experiments and statistics Advertising and Gu and Ye (2014) Chen et al. (2017) Advertising and Gu and Ye (2014) And Ye (2014) Another the Year Action of The		Guo et al. (2015)	Support vector machine	The method of support vector machines (SVM) is used to build a classifier for virtual community membership. Then a "two-stage probit least squares" (2SPLS) method is used to estimate of the
Srinivasan et al. (2017) Chen et al. (2019) Machine learning Machine learning Chen et al. (2018) Optimization and game theory Chen et al. (2016) Ramanathan et al. (2016) Dong et al. (in press) Dong et al. (in press) Dong et al. (2017) Lobel et al. (2017) Lobel et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Chen et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning Machine learning Machine learning Machine learning algorithms and computational experiments Advertising and Machine learning algorithms and computational experiments and statistics Chen et al. (2018) Machine learning algorithms and computational experiments and statistics Machine learning algorithms and computational experiments and computational experiments and statistics Machine learning algorithms and computational experiments and computational experiments Machine learning algorithms and computational experiments Machine learning alg			Optimization and genetic algorithm	Explicitly incorporating social network effects into the product design phase brings benefit. The genetic algorithm can yield
Chen et al. (2019) Machine learning Machine learning and simulation methods can be coording perform "individual-level causal inference" and "social influence identification" in social media. Machine learning models are used to predict "user adoption behaviour". Simulation is used to predict "user adoption behaviour". Simulation is used to predict "user adoption behaviour". Simulation is used to identify the "individual-leval social influence." Negative and positive reviews to be weighted differently consumer's assessment of the valence of the posted review Relative weighting of user reviews substantially affects on pricing decisions and profits. Chen et al. (2016) Ramanathan et al. (2017) Dong et al. (in press) Behavioral experiments and statistics Advertising and marketing promotion Advertising and Bimpliks et al. (2018) Lobel et al. (2017) Optimization and game theory Lee et al. (2018) Surveys and machine learning Ghose et al. (2019) Behavioral experiments and statistics Go and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning algorithms and computational experiments Machine learning and simulation is used to inference and sinducation influence. Machine learning algorithms and computational experiments Machine learning and machine learning algorithms and computational experiments			Statistics	Understanding the structure and nature of communication patterns of users in social networks is critically important for users, online social networking firms, and advertisers alike. Knowledge regarding the social network and network structure is needed in order to better understand the dynamics of
Online reviews Pekgün et al. (2018) Optimization and game theory Chen et al. (2016) Ramanathan et al. (2017) Dong et al. (in press) Behavioral experiments and statistics Advertising and marketing promotion marketing promotion Ghose et al. (2018) Ghose et al. (2019) Ghose et al. (2019) Behavioral experiments and statistics Ghose et al. (2019) Behavioral experiments and statistics Ghose et al. (2019) Chen et al. (2019) Behavioral experiments and statistics Ghose et al. (2019) Advertising and Bimpikis et al. Chen et al. (2019) Behavioral experiments and statistics Ghose et al. (2019) Behavioral experiments and statistics Chen et al. (2015) Machine learning algorithms and computational experiments Advertising and Chen et al. (2015) Machine learning algorithms and computational experiments Advertising and Advertising and Advertising and Advertising and Bimpikis et al. Coptimization and game theory Chen et al. (2019) Behavioral experiments and statistics Chen et al. (2019) Machine learning algorithms and computational experiments Chen et al. (2015) Machine learning algorithms and computational experiments An improved in under the further vector and computational experiments An improved in under the influence of online management reveposes coustomers under the further exposes of customers attifaction. An improved in under the influence of online management responses of customers attifaction. An improved in under the influence of online management responses of customers subsistated for processing unstructured measurement responses of customers attifaction. An improved in under the influence of online management responses of customers attifaction. An improved in other three review substantial to a pricing decisions and provides advertising our series and costs on search engines to in user experience and costs on search engines to in user experience and costs on search engines to in user experience and costs on search engines to in user experience and costs on search engines to in user		Chen et al. (2019)	Machine learning	Machine learning and simulation methods can be coordinated to perform "individual-level causal inference" and "social influence identification" in social media. Machine learning models are used to predict "user adoption behaviour".
Chen et al. (2016) Ramanathan et al. (2017) Ramanathan et al. (2017) Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Advertising and marketing promotion Advertising and marketing promotion Bimpikis et al. (2016) Lobel et al. (2017) Coptimization and game theory Lee et al. (2018) Surveys and machine learning Ghose et al. (2019) Behavioral experiments and statistics Gin and Ye (2014) Chen et al. (2015) Machine learning algorithms and computational experiments Machine learning algorithms and computational experiments An in probability weighting" approach can be applied to help in probability and value-based network model is used to either operations, marketing and feedback through social media reviews. Social media reviews and unique service operat represent a significant influence on customer satisfaction Positive reviews, factual reviews, and reviews appearing or refailer sites, respectively. Anal covariance is used to examine the individual and joint eff the review attributes on the "three review trustworthine dimensions". Effects of the network structure and the "word of mouth" outcome of marketing competition between the firms is a by a game-theoretic model. A symmetric Bayes-Nash equilibrium is used to study ho optimally attract new customers using a referral program. The content-tagging methodology combines surveys implemented on Amazon Mechanical Turk with machine learning algorithms is useful for processing unstructured (such as advertising content or product reviews). The structural econometric model can be used to unders consumers' preferences and costs on search engines to in user experience under unstructured social media. A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online managemen	Online reviews	Pekgün et al. (2018)	Optimization and game theory	"individual-level social influence". Negative and positive reviews to be weighted differently in a consumer's assessment of the valence of the posted reviews. Relative weighting of user reviews substantially affects optimal
Ramanathan et al. (2017) Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Dong et al. (in press) Behavioral experiments and statistics Advertising and machine learning Behavioral experiments and statistics Doptimization and game theory Doptimizat		Chen et al. (2016)		It is important to capture the silent users' comments. An "inverse
Dong et al. (in press) Behavioral experiments and statistics Positive reviews, factual reviews, and reviews appearing social networks lead to "perceptions of greater benevoler ability, and integrity" than negative reviews, embtoinal re and reviews appearing on retailer sites, respectively. Anal covariance is used to examine the individual and joint eff the review attributes on the "three review trustworthine dimensions". Advertising and Bimpikis et al. (2016) Dotimization and game theory (2016) Lobel et al. (2017) Lee et al. (2018) Surveys and machine learning Ghose et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning algorithms and computational experiments Chen et al. (2015) Machine learning algorithms and computational experiments Machine learning algorithms and computational experiments A pinch data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. An improved hierarchical multiple kernel support vector machine (H-MK-SVM) is developed to integrate related of feature selection from multiple correlated attributes and ensemble learning in user response modeling.			Survey questionnaire and statistics	A loyalty and value-based network model is used to integrate operations, marketing and feedback through social media reviews. Social media reviews and unique service operations
Advertising and marketing promotion (2016) Bimpikis et al. (2017) (2016) Lobel et al. (2017) Optimization and game theory (2018) Lee et al. (2018) Surveys and machine learning The content-tagging methodology combines surveys implemented on Amazon Mechanical Turk with machine learning algorithms is useful for processing unstructured (such as advertising content or product reviews). Ghose et al. (2019) Behavioral experiments and statistics The structural econometric model can be used to underst consumers' preferences and costs on search engines to in user experience under unstructured social media. Gu and Ye (2014) Panel data and statistics A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. Chen et al. (2015) Machine learning algorithms and computational experiments machine (H-MK-SVM) is developed to integrate related defeature selection from multiple correlated attributes and ensemble learning in user response modeling.		Dong et al. (in press)	Behavioral experiments and statistics	Positive reviews, factual reviews, and reviews appearing on social networks lead to "perceptions of greater benevolence, ability, and integrity" than negative reviews, emotional reviews and reviews appearing on retailer sites, respectively. Analysis of covariance is used to examine the individual and joint effects of the review attributes on the "three review trustworthiness"
Loe et al. (2017) Optimization and game theory Lee et al. (2018) Surveys and machine learning Chose et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Chen et al. (2015) Machine learning algorithms and computational experiments Chen et al. (2015) Machine learning algorithms and computational experiments A symmetric Bayes-Nash equilibrium is used to study ho optimally attract new customers using a referral program The content-tagging methodology combines surveys implemented on Amazon Mechanical Turk with machine learning algorithms is useful for processing unstructured (such as advertising content or product reviews). The structural econometric model can be used to underst consumers' preferences and costs on search engines to in user experience under unstructured social media. A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. An improved hierarchical multiple kernel support vector machine (H-MK-SVM) is developed to integrate related of feature selection from multiple correlated attributes and ensemble learning in user response modeling.	•		Optimization and game theory	Effects of the network structure and the "word of mouth" on the outcome of marketing competition between the firms is studied
Lee et al. (2018) Surveys and machine learning The content-tagging methodology combines surveys implemented on Amazon Mechanical Turk with machine learning algorithms is useful for processing unstructured (such as advertising content or product reviews). The structural econometric model can be used to underst consumers' preferences and costs on search engines to in user experience under unstructured social media. Gu and Ye (2014) Panel data and statistics A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. Chen et al. (2015) Machine learning algorithms and computational experiments Machine (H-MK-SVM) is developed to integrate related defeature selection from multiple correlated attributes and ensemble learning in user response modeling.		Lobel et al. (2017)	Optimization and game theory	A symmetric Bayes-Nash equilibrium is used to study how to
Ghose et al. (2019) Behavioral experiments and statistics Gu and Ye (2014) Panel data and statistics Panel data and statistics The structural econometric model can be used to underst consumers' preferences and costs on search engines to in user experience under unstructured social media. A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. Chen et al. (2015) Machine learning algorithms and computational experiments Machine learning algorithms and ensemble learning in user response modeling.		Lee et al. (2018)	Surveys and machine learning	The content-tagging methodology combines surveys implemented on Amazon Mechanical Turk with machine learning algorithms is useful for processing unstructured data
Gu and Ye (2014) Panel data and statistics A panel data model that controls for regression toward the and heterogeneity in individual preference is effective to measure the influence of online management responses of customer satisfaction. Chen et al. (2015) Machine learning algorithms and computational experiments Machine (H-MK-SVM) is developed to integrate related defeature selection from multiple correlated attributes and ensemble learning in user response modeling.		Ghose et al. (2019)	Behavioral experiments and statistics	The structural econometric model can be used to understand consumers' preferences and costs on search engines to improve
Chen et al. (2015) Machine learning algorithms and computational experiments An improved hierarchical multiple kernel support vector machine (H-MK-SVM) is developed to integrate related d feature selection from multiple correlated attributes and ensemble learning in user response modeling.		Gu and Ye (2014)	Panel data and statistics	A panel data model that controls for regression toward the mean and heterogeneity in individual preference is effective to measure the influence of online management responses on
		Chen et al. (2015)	0 0	An improved hierarchical multiple kernel support vector machine (H-MK-SVM) is developed to integrate related data fo feature selection from multiple correlated attributes and for
Besbes et al. (2016) Optimization and live pilot experiment The optimization problem that leverages both clickability engageability is built and solved.		Besbes et al. (2016)	Optimization and live pilot experiment	The optimization problem that leverages both clickability and

Table 3.1 (continued)

Topics/Areas	Papers	Methods and Tools	Insights
Logistics management	Fry and Binner	Optimization and algorithm	A Bayesian algorithm which enables the optimal evacuation to
	(2016)		be conducted sequentially under "worsening conditions".
	Yoo et al. (2016)	Field study and statistics	A field study that examines how information diffuses through social media networks embedded.
	Lam et al. (2016)	Panel data and statistics	Social media initiatives improve operational efficiency and
			innovativeness. A "system generalized method of moments"
	E 1 Min (0016)	C	statistical approach can be applied to help.
	Fan and Niu (2016)	Grounded theory approach and statistics	The grounded theory approach is used to investigating the roles of recovery speed and failure severity in the service recovery
		statistics	process using social network platform.
	Hou et al. (2018)	Game theory	Using game theory to explore different competition scenarios
	()		(namely, the "Nash game" and "Stackelberg game") for delivery
			service operations with the use of "social media".
	Singh et al. (2018)	Support vector machine	Conducting text analysis via SVM to uncover the "cluster of
			words" which can indicate the major issues (e.g., product
			quality and flows) in the food supply chain.
Forecasting	Ballings and Van	Logistics regression, random forest,	The "Stochastic Adaptive Boosting" method outperforms the
	den Poel (2015)	stochastic adaptive boosting, neural	other methods in predicting increases in Facebook usage
	Lau et al. (2018)	networks and support vector machines Extreme learning machine	frequency. A parallel co-evolutionary extreme learning machine is used to
	Lau et al. (2016)	Extreme learning machine	enhance sales forecasting, consumer sentiments mined from big
			data can improve the accuracy of sales forecasting.
	Choi (2018)	Bayesian statistics	The use of social media data in fashion quick response supply
		•	chains is analysed.
	Cui et al. (2018)	Machine learning	Nonlinear boosting models outperform traditional linear models
			with the use of public social media data for daily sales
			forecasting.
New product	Chan et al. (2016)	Mixed-methods including fuzzy AHP,	Multiple methods are needed to properly extract values of social
development	Vlausen et al. (2010)	TOPSIS Statistics	media data.
Risk analysis	Klausen et al. (2018)	Statistics	Behavioral models can be used to track existing "extremist users" by identifying pairs of accounts belonging to the same
			user.
Product introduction and	Aral and Walker	Behavioral experiments and statistics	"Randomized trials" can identify peer influences in social
design	(2011)	•	networks.
	Abedi et al. (2014)	Optimization and algorithm	The "generalized proportional hazard rate (GPHR)" model,
			describes demand diffusion across multiple markets and/or
			segments. The method can identify the optimal mix of marketing
3.6 12 1	1 (0010)	0	channels and expenditures in them over time.
Medical treatment	Yan et al. (2019)	Statistics	A "latent response model" is built to separately test the effects of
			each shared attribute information—effectiveness, side effects, compliance, and cost on patients' treatment ratings.
Others	Huang et al. (2015)	Statistics and counterfactual	"Counterfactual experiments", with statistical analyses,
o anero	11dang et an (2010)	experiments	demonstrate that leisure-related blogging has positive spillovers
			for work-related blogging.
	Sul et al. (2017)	Statistics and content analysis	The sentiment in social media postings can predict stock returns
			on future trading days by statistical methods. Three separate
			measures are used to better model sentiment. (1) the ratio of the
			"amount of negative sentiment to the total communication". (2)
			a normalized ratio of the "overall positive or negative sentiment
			expressed". (3) an unstandardized ratio of "positive to negative sentiment".
	Ma et al. (2019)	Big data analytics	Using big data from Twitter to investigate the "overbooking
	ma ct ai. (2017)	Dig data marytics	crisis" in an airline.
	Count at al. (2010)	Ameliatic historials, massessing (ALID)	
	Syed et al. (2019)	Analytic hierarchy processing (AHP)	An AHP based "Identity Management Value Model" helps

4.2. TSM examples and features

From the review, Facebook and Twitter are the two dominating TSM platforms used for social media analytics in SCOM analysis (see Table 4.1). In the following, we examine them.

As we can see from the literature review, SMA is especially prominent with Facebook and Twitter. This finding can partially be explained by the availability of their Application-programming-interfaces (APIs) which are publicly accessible online. A lot of software applications hence arise aiming to help get data from them and conduct further analyses. However, there are also challenges.

Twitter supports social media analytics as it works with Gnip to provide Twitter's data available to companies all around the world to use. For general public, Twitter provides access to some data via its API. To be specific, by default of Twitter's accounts, the

Table 4.1The social media platforms used for SMA.

Social Media Platforms	Applications	Related References
Facebook	Sales/demand forecasting	Ballings and Van den Poel (2015), Cui et al. (2018)
	Social influence identification	Aral and Walker (2014), Guo et al. (2015), Gunnec and Raghavan (2017), Srinivasan et al. (2017) ¹ , Chen et al. (2019)
	Online reviews	Ramanathan et al. (2017), Dong et al. (in press)
	Product design	Aral and Walker (2011)
	Advertisement and marketing	Lobel et al. (2017), Lee et al. (2018)
	promotion	
Twitter	Sales/demand forecasting	Lau et al. (2018)
	Social influence identification	Srinivasan et al. (2017), Chen et al. (2019)
	Online reviews	Ramanathan et al. (2017)
	Advertisement and marketing	Chen et al. (2015)
	promotion	
	Stock returns forecasting	Sul et al. (2017)
	Logistics management	Fry and Binner (2016), Lam et al. (2016), Yoo et al. (2016), Fan and Niu (2016), Singh
		et al. (2018), Ma et al. (2019)
	Medical treatment	Yan et al. (2019)
	Risk analysis	Klausen et al. (2018)
Blogs	Online reviews	Pekgün et al. (2018)

¹ Some references use both Facebook and Twitter, e.g., Srinivasan et al. (2017) and Chen et al. (2019).

users' posts, mentions and replies (i.e. "Tweets") are public (even though individual users can modify to make some of their "Tweets" invisible for privacy reasons). It is noted that over 90% of Twitter's accounts are public and less than 10% are private (Batrinca and Treleaven, 2015). In addition, for those public accounts, Tweets are stored in a data exchange format called "JSON" (Javascript object notation) and they are available via Twitter's "Search API" which makes them available for social media analytics. Batch data (past) and real-time data are available via "Search API" and "Streaming API", respectively. It is important to note that APIs in Twitter are not fully open for any users and developers. There are restrictions, such as an upper bound on the API usage rate for individuals (Brooker et al., 2016).

Facebook, similar to Twitter, also provides its data via API. However, its privacy policy makes it more complicated for social media analytics. To be specific, Facebook stores data in the object format and provides different kinds of APIs, e.g., the "Keyword Insight API", "Search API", etc. It is known that for Facebook, many status messages are more difficult to collect compared to Twitter's Tweets because we need to obtain the authorization from users (or the users set it to be open) (Batrinca and Treleaven, 2015).

4.3. Problems of TSM platforms

From Section 4.2, we can clearly see the basic features of Facebook and Twitter. There are problems which are associated with them and other traditional social media platforms for SMA. We discuss them in more details as follows:

<u>Data collection:</u> Web crawling for data from social media platforms is not fully open and perfectly public. For example, Twitter imposes a rate limit for individual user (i.e. account)'s usage of the APIs for searching and getting data. Facebook has its own privacy control system which makes it even more complicated than Twitter for users to obtain data.

<u>Incentive:</u> At the very beginning, traditional social media platforms provide a free place for users to get together and have fun. As such, in most cases, traditional social media platforms would not provide very strong incentive to entice engagement and faithful participation of users. Nowadays, Youtube and Facebook Live would have some incentive schemes but arguably the sharing of benefit is relatively limited and lacks transparency (Mire, 2018).

Missing data: Social media data may be incomplete with many missing details. Some data are of poor quality owing to the veracity nature of social media data. All these create non-trivial challenges. Some traditional methods of ignoring the missing data would simply lead to biases. Some advanced methods are hence proposed to help (Chen et al., 2016). However, missing data is still missing and this problem is critical.

<u>Data authentication problem:</u> There are "bot" problems in which information posted online in social media need not be real. Companies can use machines to add comments or create posts by paying some individuals. As a result, fake information and falsified data are rather common (Holsapple et al., 2014). This is an important issue and Forbes reports that Facebook has recently been very busy dealing with the fake account issues.⁸

<u>High volume and velocity of data:</u> Social media data is a typical example of big data, with which the most fundamental characteristics include high volume in size and high velocity in applications requirement (Choi et al., 2018a). As a result, traditionally, the right software architecture and storage technology must be present to overcome this challenge.

⁸ https://www.forbes.com/sites/andrewarnold/2018/05/29/how-social-media-can-benefit-from-blockchain-technology/#3a85124b1018 (accessed: 16 November 2019).

Table 4.2 Problems associated with the social media platforms (in Table 3.1).

Social Media Platforms	Problems	Related References
Facebook	Privacy issues of Facebook are more complicated than Twitter, making it even	Batrinca and Treleaven
	more difficult to get the data.	(2015)
Twitter	Not totally open and available without restrictions to users. There is a rate limit	Brooker et al. (2016)
	for individual entity (i.e. account)'s usage of APIs.	
Traditional Social Media Platforms (including	Incentive: In most cases, traditional social media does not provide strong	Stieglitz et al. (2018), Mire
Facebook and Twitter)	enough incentive to entice engagement and faithful participation of users.	(2018)
	Unstructured data: Social media data are in different formats, including texts,	Cui et al. (2018)
	photos, videos, sounds, etc.	
	Missing data: Data may be incomplete or of low quality owing to the veracity	Stieglitz et al. (2018)
	nature of social media data	-
	High volume and velocity of data: It is necessary to find the right software	Choi et al. (2018a)
	architecture and storage technology	
	Data authentication problem: Fake information (the "bot problem"), falsified	Holsapple et al. (2014)
	data.	

<u>Unstructured data:</u> Social media is a source of different kinds of data, including texts, pictures, videos, and sounds. In almost all cases, data are not given in a structured manner. This makes the use of data more difficult, especially for scientifically sound analysis for operations management. In fact, <u>Leetaru (2019b)</u> discovers that most currently used social media analytical tools focus mainly on texts but not on graphics or videos. Even for those which examine graphics or photos, they only examine logos. This means the full picture is missing from the scene.

Table 4.2 summarizes the problems associated with the use of Facebook, Twitter, as well as other traditional social media platforms for social media analytics.

5. Blockchain-technology-supported social media (BSM)

5.1. Real BSM platforms

After exploring the deficiencies of the traditional social media platforms in Section 4, we now examine the blockchain technology supported social media platforms in the real world. In fact, for each major TSM platform, there is usually a counterpart which is widely and commonly known.

Foresting is usually treated as the BSM platform counterpart of Facebook. It has functions very similar to Facebook by allowing users to post their own content materials for sharing on the platform. Most importantly, after posting the materials online, when other users express "like" towards the materials or content, the posters will receive rewards in the form of cryptocurrency, which is a flagship application of the blockchain technology (from its gene in bitcoin). Foresting also features the Foresting crypto-bank, which can help users to create content with the needed financial support and assistance.

For the TSM Twitter, we have Peepeth as the BSM counterpart. It is known that Peepeth employs the Ethereum blockchain to support its platform. For Twitter, we have Tweets whereas for Peepeth, we use "Peeps" to represent the posts. With the blockchain technology, "Peeps" exhibit the feature that they cannot be amended, edited, changed or deleted. As a result, permanent data as well as trustworthy data records are guaranteed. This is critically important to ensure data authenticity and quality.

For video sharing, Youtube is probably the most well-established TSM platform. With the blockchain technology, a BSM called Dtube emerges. Dtube is a decentralized video sharing platform supported by the STEEM Blockchain. It applies the IPFS P2P network technological model. It is similar to Youtube in its major applications. However, one of its strengths is to allow users to share the financial benefit in the form of "Steem currency". The amount of financial benefit received is directly proportional to the user engagement. It also supports the choice of "advertisement-free-platform" set by users.

The BSM platform Indorse is similar to the TSM platform Lindedin. The major difference is that Indorse provides cryptocurrency as rewards when users expand their networks, endorse other users' skills, etc. It is reported that the Indorse network tends to be more reliable and honest as "reputation" counts. Observe that there is an artificial intelligence (AI) chatbot in Indorse to validate details of the users in real time. This enhances authenticity.

For sharing photos, Instagram is the most popular TSM platform. The BSM platform Steepshot also aims to offer photo sharing functions. The major difference is that, Steepshot provides incentives to users who post photos in the form of cryptocurrency. Again, the incentive relates to the engagement of users.

Table 5.1 summarizes the BSM platform counterparts of the major TSM platforms. ⁹ The roles played by blockchain technology are also concisely shown.

⁹ Note that the existence of these BSM counterparts is well-reported and noticed in practice, see Mire (2018) and Radar (2018).

Table 5.1The blockchain technology supported social media platforms and roles of blockchain.

TSM	The BSM Counterparts	Roles of Blockchain
Facebook	Foresting	Foresting has functions similar to Facebook by allowing users to post their own materials for sharing. Interestingly, when other users like the materials or content, the posters will receive rewards in the form of cryptocurrency. There is even a Foresting crypto-bank which provides financial supports to content creators if needed.
Twitter	Peepeth	Peepeth employs blockchain technology (P.S.: The Ethereum blockchain) and the posts called "Peeps" cannot be altered, changed or deleted. So, permanent data as well as trustworthy data records are guaranteed.
Youtube	Dtube	DTube is a decentralized video sharing platform supported by blockchain (P.S.: the STEEM Blockchain). It is similar to Youtube and it allows users to share the financial benefit of in the form of "Steem currency". It supports the choice of advertisement free by users.
Lindedin	Indorse	Indorse is similar to Lindedin but it provides cryptocurrency as rewards when users expand their networks, endorse others' skills, etc. The network tends to be more reliable and honest as "reputation" counts. There is an artificial intelligence (AI) chatbot to validate details in real time.
Instagram	Steepshot	Steepshot has a similar function as Instagram while there are incentives to users who post photos in the form of cryptocurrency.

Table 5.2 How blockchain technology helps social media platforms to support SCOM.

Topics/Areas	Papers	Requirements	How BCT Can Help?
Social influence	Srinivasan et al. (2017)	Better understand the dynamics of communication in these networks	BTC enhances the data transparency.
	Dong et al. (in press)	Acquire Online review characteristics and trust	People can feel free to trust the information in the blockchain.
Advertising and marketing promotion	Bimpikis et al. (2016)	Gather vast amounts of data regarding consumers' preferences and the structure and intensity of their social interactions for targeted advertising	BCT enables advertiser acquired the consumers' preferences and their social interactions more easily.
	Lee et al. (2018)	Utilize unstructured data	BCT can organize data in a more structured way which makes the respective data usage and analysis easier.
Logistics management	Fry and Binner (2016)	Emergency evacuations using social media	BCT structure can increases information diffusion efficiency.
	Yoo et al. (2016)	Examine how effectively information diffuses through social media networks embedded in these platforms	BCT increases information diffusion efficiency.
Risk analysis	Klausen et al. (2018)	Track existing extremist users by identifying pairs of accounts belonging to the same user	BCT helps identify the user's information and guarantees the authenticity.
Product introduction and design	Aral and Walker (2011)	Create social contagion	BCT can increase social influence by decentralized ledger. Every node will be informed.
C	Abedi et al. (2014)	Identify the available retail facilities to carry the product.	BCT enables users to get the updated information timely.
Medical treatment	Yan et al. (2019)	Eliminate health information asymmetry and improve patients self-care engagement	BCT guarantees data transparency and enable patient to access others experiences more easily.
Others	Sul et al. (2017)	Prediction of stock returns on future trading days based on the sentiment in social media postings	BCT saves permanent data record and enhances data transparency. The distributed decentralized platform facilitates information sharing.
	Syed et al. (in press)	Identity management objectives	BCT can solve the problem of data authentication and fake information issue.

5.2. How blockchain technology helps

After noting that in the real world, virtually every TSM platform has a BSM platform counterpart, in which some additional and critical functions supported by the blockchain technology are present. In the following, we discuss how the blockchain technology can overcome or ease the problems raised in Section 4.

First of all, in Section 3, we present a summary of the methods and insights derived from the reviewed SMA papers in Table 3.1. In Table 4.2, we show the problems associated with each major social media platform. From them, we further explore how blockchain technology can play a role to enhance some of the discussed methods. Table 5.2 shows the results.

From Table 5.2, it is crystal clear that the blockchain technology can overcome many problems, including those associated with data authentication, data transparency, and information diffusion, etc. All are important and critical for SCOM. In addition to these reviewed methods related to SMA that blockchain can help, in the following, we discuss some general areas in which the BSM, with its features supported by blockchain, can enhance SMA.

Regarding difficulty in getting the data: The blockchain technology helps create a permanent data record which can be publicly available. This directly enhances data transparency, supports SMA, and also facilitates the use of information along the supply chain. This is crucial for improving operations with the use of social media. Sola is a BSM platform, which advocates the freedom of information and its distributed decentralized platform facilitates information sharing (Mire, 2018). It hence helps address

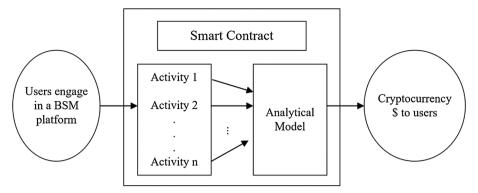


Fig. 5.1. The use of smart contracting for granting cryptocurrency to BSM users.

the data collection problem associated with the TSM platforms.

For lack of incentive to entice engagement and faithful participation of users: This problem can be fully addressed by the blockchain technology because it supports cryptocurrency. Many BSM platforms offer cryptocurrency incentives to users, in which the amounts are related to their levels of participation and engagement. BSM platforms such as Indorse, Dtube, Steepshot, Killi, Sapien, Steemit, etc. all provide incentives in the form of cryptocurrency to ensure users get compensated for their data and content. Fig. 5.1 shows how the blockchain-based smart contract (a computer program) can help to support this cryptocurrency incentive scheme to entice users to faithfully participate in the BSM platform related activities. Note that in Fig. 5.1, the analytical model represents the scientific model with formulas to help determine the right amount of cryptocurrency to be granted to each user based on her corresponding participation levels in different activities.

Regarding the unstructured data: When datasets are kept in the blockchain, it is easier for social media analysts to think of a way of exploring them by the right approach. If learning is needed, the amount of data provided by the BSM platform will help to improve the effectiveness with the use of data. Plus, most blockchain technology based databases organize data in a more structured way which makes the respective data usage and analysis easier.

For the problem of missing data: The blockchain technology can help to provide the long history of data for checking and verification. The amount of data available will be more sufficient and this helps with dampening the missing data problem.

Addressing the "high volume and high velocity" challenge: This requires the adoption of techniques in big data analytics (Choi et al., 2018a; Guha and Kumar, 2018). Having said that, the BSM platform may also provide a way to enhance the respective SMA because blockchain can facilitate information dissemination in a more transparent way (Sodhi and Tang, in press). This may facilitate the collection and retrieval of data for further processing in SMA.

For the data authentication and fake information issue: This challenging issue is well tackled by the blockchain technology. To be specific, for many BSM platforms, user identity is verified and is real. For instance, Civic is a BSM platform offering digital user identity verification services. The data and contents created by users are also recorded by the transparent blockchain system with permanent records. All these features would make data authentic and trustworthy. This will lead to more accurate SMA results to support business operations decisions. It is reported that SoMee is a BSM platform which adopts the blockchain technology to secure user data cryptographically. Moreover, the BSM platform Sapien uses blockchain to lift the level of veracity and authenticity of its posted content (Mire, 2018).

<u>For the problem of data island:</u> Each enterprise has its own data. Through the use of blockchain technology, data among members can be shared in a more transparent way. This directly leads to the more accurate characterization of customer features (e.g., the users of Facebook). It helps to make many market driven activities, from demand forecasting to advertising, more accurate for the targeted customers.

Guarantee the freedom of speech: In many traditional social media platforms, what we read from the posts by users may actually be censored. When the published content damages the interests of a party or affects the image of a party, even if the content is authentic and credible, it will be deleted in many cases. As a result, the freedom of speech is not well-supported and this would lead to biased data source for SMA. The blockchain technology supported social media platforms can enhance the freedom of speech by ensuring the details are real, uncensored and clear. It hence allows users of social media to truly express themselves without the central control of the platform. Using the respective data can improve the quality of recommendations by SMA.

<u>Privacy protection:</u> Traditional social media platforms adopt the centralized database approach. When they are being attacked by cyber criminals, users' personal data will be lost. This creates some doubts to whether real data should be kept on the social media platforms. The blockchain technology is a decentralized ledger which adopts the distributed network to store data. It is known to be very secure and it is basically impossible for cyber criminals to attack. With user privacy being well-protected, users are willing to put more real data that they want to share on the social media and the data that can be used for SMA will be of a higher quality.

5.3. Graphical illustrations and further discussions

Based on the above discussions, Figs. 5.2 and 5.3 depict the scenarios with TSM and BSM, respectively.

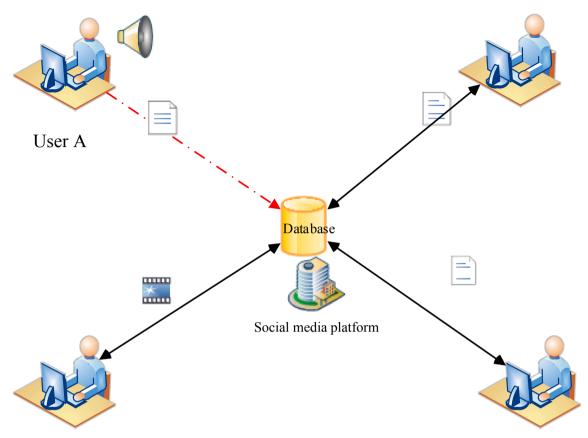


Fig. 5.2. The traditional social media (TSM) platform.

From Fig. 5.2, we can see that in the TSM, every user posts messages or gets information through the platform. All the user data are stored in and manipulated by the central database. For example, when User A posts a new message, it may actually be censored by the platform. The other users can only access the new message through the platform.

From Fig. 5.3, we can observe that in the BSM, datasets of users are stored by a decentralized ledger. When User B posts a new message, everyone in the blockchain system is informed, as shown by the red dot lines. The blockchain technology helps create a permanent data record which can be publicly available and achieve data transparency as shown in the blue dot lines. The information cannot be edited by the social media platform without the consent of all the users. It hence guarantees the freedom of speech. It also helps to solve the problem of missing data and ensure data authenticity and security. What is more, BSM provides an incentive to entice engagement and faithful participation of all users. The users can get compensated for their "contribution" (e.g., in providing data and creating content) in the form of cryptocurrency.

Figs. 5.4 and 5.5 illustrate the major steps involved in SMA with the use of TSM and BSM, respectively.

From Fig. 5.4, it is crystal clear that in the case with TSM, companies which conduct data analytics on the social media can crawl data from the platform. Since the problems of missing data and unstructured data may occur, companies should rebuild the missing data and standardize the data, and then de-duplicate the data. In order to solve the data authentication and fake information issue, they also need to verify the data before conducting the feature extraction. Finally, the well-processed data could be used for data analytics.

From Fig. 5.5, we can observe that in the case with BSM, since blockchain advocates the freedom of information and its distributed decentralized platform facilitates information sharing, the data collection problems associated with the TSM platform are well-addressed. In addition, the blockchain technology helps to create a permanent data record which can be publicly available; this addresses the problem of missing data. Most blockchain technology based databases organize data in a more structured way than the traditional databases, which makes the respective data usage and analysis easier. The data authentication challenge and fake information issue are well tackled by the blockchain. So, the BSM based SMA can substantially improve the efficiency of data analytics. Comparing the required steps in the TSM-based SMA and the BSM-based SMA (i.e., Figs. 5.4 and 5.5), we can find that the whole SMA process is more streamlined with the help of blockchain.

Fig. 5.6 further shows how the blockchain technology can enhance social media platforms to support SMA for SCOM. In particular, the benefits for SMA by using BSM are highlighted one by one.

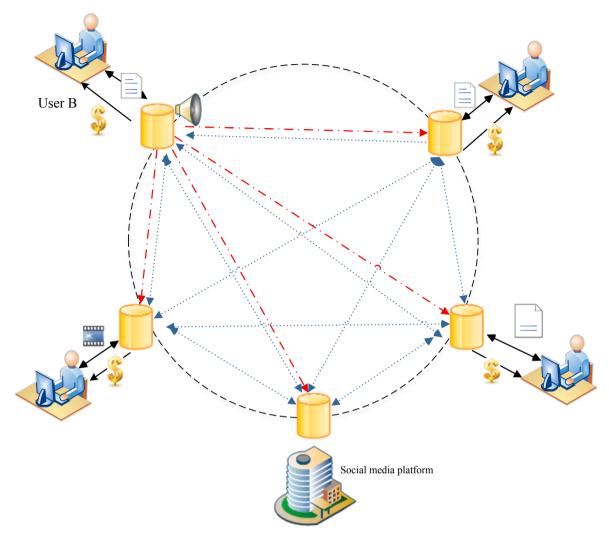


Fig. 5.3. The blockchain-based social media (BSM) platform.

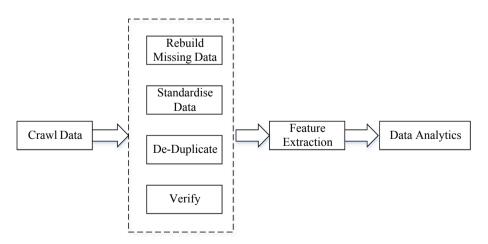


Fig. 5.4. TSM based SMA.



Fig. 5.5. BSM based SMA.

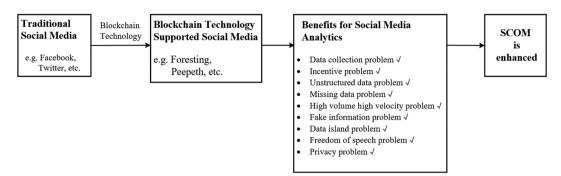


Fig. 5.6. How the blockchain technology enhances SMA for SCOM.

6. Future research agenda

From the literature review and the above detailed analyses, we identify a few areas which deserve future deeper explorations. Studying them in the future would involve different methods and approaches (e.g., analytical, computational, empirical, etc.), as detailed below.

More empirical SCOM research in SMA: As evidenced from the systematic literature review results, currently, there are much more analytical studies on SMA published in top mainstream analytical SCOM journals than the empirical counterparts. It directly means that more empirical SCOM research in SMA should be conducted as there are obviously lots of opportunities for further empirical studies in the future. In particular, currently, there are very few SMA studies with relevance to and emphasis on empirical SCOM theories. However, obviously, from the organization perspective, SMA is associated with many classic SCOM theories (Anand and Gray, 2017) such as the resource view based theory (Hitt et al., 2016), and the transactions cost theory (Grover and Malhotra, 2003). These hence provide the theoretical ground for future empirical SCOM research on SMA.

Employing BSM platforms for SMA: For applications of SMA, currently, almost all studies focus on TSM platforms such as Facebook and Twitter. It will be critically important to study the use of BSM platforms for SMA. In particular, it will be interesting to explore how the cryptocurrency incentive schemes affect the user participation in social media and how that will affect SMA, etc. Even though we argue that the blockchain technology can bring many benefits, whether there are any problems or negative impacts should be examined. Moreover, to conduct SMA on the BSM platforms may involve some new technological challenges, which also deserves deeper investigation.

Quantifying the benefits of blockchain technology for SMA: In this paper, we propose the use of BSM and argue that it is useful for SMA. However, for real world supply chain operations, conducting SMA is not without cost. It is thus important to quantitatively explore the benefits of blockchain technology for SMA, and the associated factors. The result will support new investment decisions of companies when they plan to conduct SMA on BSM platforms.

<u>Innovative business models:</u> SMA supports many innovative business models. Among them Groupon is one area which is highly related to social media analytics (Wang et al., 2018). How blockchain technology supported social media platforms will affect innovative business models such as Groupon and others is under-explored and deserves further explorations.

<u>Methods of analysis:</u> It is reported that currently, the managerial recommendations based on SMA for the same dataset can be totally different (<u>Leetaru</u>, 2019a). This is a direct result of the different algorithms employed in different practical SMA tools. Many tools currently employed simply put "speed" as the priority but they overlook the importance of accuracy and data quality. In future research, more emphasis should be put on exploring how the methods of analysis can be chosen with respect to quality and accuracy of the recommendations. Moreover, how the blockchain technology may play a role also deserves deeper investigation.

7. Conclusion

7.1. Summary and concluding remarks

Nowadays, social media analytics (SMA) is a part of many business operations. There is no doubt that it also helps to support many activities in the supply chain. However, the use of SMA for SCOM is far from perfect. Problems such as data accuracy (e.g., fake data), user privacy, data security, etc are all present.

Motivated by the fact that (i) SMA is critical for SCOM, (ii) the traditional social media (TSM) platforms are far from perfect, and (iii) the "blockchain technology supported social media" (BSM) platforms have emerged, we have studied in this paper whether and how the BSM would enhance social media analytics for SCOM. To be specific, by reviewing papers published in leading SCOM

journals, we have found the applications and limitations of SMA for SCOM. Then, we have conducted real case studies to examine the shortcomings of the TSM platforms and uncovered features of their BSM counterparts. We investigate how the blockchain technology would potentially improve the use of SMA for SCOM. Finally, we have proposed a few important future research directions.

This paper positions itself as the pioneering study, which combines literature review and original real case research, to uncover how the use of blockchain technology would enhance SMA for SCOM. We believe that the derived findings are valuable to both the practitioners and academicians. It also contributes to the SCOM literature, especially in the SCOM-technology interface.

7.2. Limitations

We admit a few limitations. First of all, for the systematic review, even though we have tried our best to be accurate, some papers might still be missing. Second, this is an exploratory study and many BSM platforms have not yet been fully developed. As a result, our proposals and discussions are based on our own observations and other reported studies. In fact, we have searched those BSM platforms and tried to see if major enterprises like Apple, LV, Walmart have marked their presence. However, we cannot find any at the moment. Readers should thus interpret our findings with care and we believe that more industrial observations will be needed to double check the results in future studies. Third, for the data that we have considered in this paper, we refer to those "public data" that can available from the social media platforms in which companies can "crawl". However, SMA may also refer to those "private data" that the companies possess from their own social media accounts. We argue that the difference in the availability and usage of this kind of "private data" should be similar for the cases with BSM and TSM. But further studies may still be needed. Last but not least, social media analytics with the use of blockchain can support many important operations in the sharing economy, such as rental service platforms (Sun et al., in press), mobile-app-web dual channel retailing (Choi, in press), fighting counterfeiting (Choi, 2019; Wang et al., in press), and enhance sustainable operations (Shi et al., in press; Cai et al., 2019; Choi and Luo, 2019; Li et al., in press; Fung et al., in press; Guo et al., 2020). These should deserve deeper explorations in the future. Moreover, a multi-methodological research approach (Chiu et al., 2011, in press; Choi et al., 2016, 2019a; Sheu and Choi, 2019; Guo et al., 2020) can be employed during the research process so as to reveal more scientifically sound results and insights.

CRediT authorship contribution statement

Tsan-Ming Choi: Conceptualization, Investigation, Methodology, Validation, Formal analysis, Writing - original draft, Visualization, Writing - review & editing. **Shu Guo:** Investigation, Writing - review & editing. **Suyuan Luo:** Investigation, Writing - review & editing, Validation, Visualization.

Acknowledgements

Tsan-Ming Choi's research is partially supported by The Hong Kong Polytechnic University (Project ID: P0009613). Suyuan Luo acknowledges the support from China Postdoctoral Science Foundation, [grant number 2019M653049] and the MOE (Ministry of Education in China) Project of Humanities and Social Sciences.

Appendix A. Details of Fig. 1.1

In Fig. 1.1, we describe the common features of blockchain through four layers. Usually, users interact with the system through the top layer, called the Application Layer, which is similar to various software interfaces in the computer. The Application Layer is the BTC supported platform that people can actually use directly, e.g., the diamond authentication platform, social media platform, financial platform, etc. When people use this layer, cryptocurrency or other transaction will easily be involved. For example, people can earn or use cryptocurrency in a financial platform and a third party will also help to record the transaction. The cryptocurrency or other transactions will spread or used in other layers. The second layer is the Extension Layer which is similar to an extended application with analytical power. It is designed to make blockchain products more powerful and useful. For example, people can use the smart contract or the BSM to share and store data. For the smart contract, if certain conditions are met, the contract will automatically execute, such as automatic transfer of securities, automatic payment, and so on. The third layer is the Network Layer. By BCT, the data is stored in a decentralized mode. The node and the shared news are guaranteed to be authentic and secure through crypto mining, voting and other consensus algorithms. The bottom layer is the Storage Layer, in which information is stored in blocks and establish the chain. The data is verified and recorded by every node in the chain, which guarantees the permanent and transparent ledger of data is kept.

Appendix B. More reviews on SMA

Other applications: Other related studies on SMA's applications include the following works. Huang et al. (2015) explore the blogging behaviors of workers in a company. The authors examine the workers' consumption behaviors as well as content creation in social media. They interestingly show that if companies impose restriction on leisure blogging, the employee performance will be negatively affected. Sul et al. (2017) demonstrate empirically, by using millions of Twitter posts (on individual companies), how Twitter postings would affect the stock returns of the corresponding companies. They prove that for companies with a sufficiently small number of followers, the Twitter postings have a significant impact on the respective companies' stock-returns within a 20-day

period. Ma et al. (2019) employ Twitter data on an airline to conduct analysis on the overbooking challenge. The authors propose how the big data from social media can effectively enhance revenue management for airline operations. Motivated by the fact that identities of individuals on social media can be faked, Syed et al. (in press) study the identity management issue on social media. The authors argue that on social media, proper identity management needs a value perspective. They conduct a multiple-objectives decision making analysis. Using Facebook as an example, they quantify the achievement of their proposed method.

Methods for SMA: In addition to the applications we reviewed in Section 3.1, since SMA relates to the use of social media to extract, manipulate and process data to generate important insights and knowledge, many different approaches and methods for SMA are discussed in the literature. For instance, in terms of methods, Chan et al. (2016) propose a hybrid approach to manipulate social media data which are unstructured, subjective, and diverse. The authors present a case study (based on Samsung Mobile's Facebook page) to show the effectiveness of the proposed method. Chen et al. (2016) develop a method, called IPW (inverse-probability-weighting) to reduce the bias derived from the incomplete nature of many online review data. They conduct a real case analysis on the online reviews in Blockbuster.com and show that their proposed IPW method is effective and practical. Massimino (2016) explores techniques on web-crawling. The author first defines important concepts for web-crawling. They then give specific examples and highlight some probable problematic areas. Sodero and Rabinovich (2017) explore flash sales (FS) business operations on the Internet. FS refers to the case when retailers purchase excessive inventories of a product from the spot market with predetermined time duration. The authors establish a demand prediction model which considers the points made by consumers in social media (including forum posts) associated with different deals.

SMA and big data analytics: As a remark, SMA is also related to big data analytics and emerging technologies in the Industry 4.0 era (Tang and Veelenturf, 2019). In this domain, Waller and Fawcett (2014) study how emerging technologies such as additive manufacturing (also called "3D printing"), which are connected to both the Internet and the social media online communities, would affect supply chain operations. The authors focus on the reduction of cycle time upon the implementation of these technologies. Schoenherr and Speier-Pero (2015) examine big data analytics in supply chain systems by conducting a survey among professionals in supply chain management. The authors discuss future research on the topic. Zhan and Tan (in press) investigate how the use of big data can improve supply chain operations. The authors establish a well-integrated analytical infrastructure for this purpose. Klumpp and Zijm (in press) study social issues with humans when logistics operations are under the transition to artificial intelligence (AI) automation. The authors propose a theoretical framework and discuss how human trust to AI may affect technological innovation.

References

Abedi, V.S., Berman, O., Krass, D., 2014. Supporting new product or service introductions: location, marketing, and word of mouth. Oper. Res. 62 (5), 994–1013. Anand, G., Gray, J.V., 2017. Strategy and organization research in operations management. J. Oper. Manage. 53, 1–8.

Aral, S., Walker, D., 2011. Creating social contagion through viral product design: a randomized trial of peer influence in networks. Manage. Sci. 57 (9), 1623–1639.

Aral, S., Walker, D., 2014. Tie strength, embeddedness, and social influence: a large-scale networked experiment. Manage. Sci. 60 (6), 1352–1370. Asian, S., Nie, X., 2014. Coordination in supply chains with uncertain demand and disruption risks: existence, analysis, and insights. IEEE Trans. Syst., Man, Cybernet.:

Syst. 44 (9), 1139–1154.
Babich, V., Hilary, G., 2019Babich and Hilary, in press. What OM researchers should know about blockchain technology. Manuf. Service Operat. Manage (in press).

Ballings, M., Van den Poel, D., 2015. CRM in social media: predicting increases in Facebook usage frequency. Eur. J. Oper. Res. 244 (1), 248–260. Batrinca, B., Treleaven, P.C., 2015. Social media analytics: a survey of techniques, tools and platforms. AI & Soc. 30 (1), 89–116.

Besbes, O., Gur, Y., Zeevi, A., 2016. Optimization in online content recommendation services: Beyond click-through rates. Manuf. Service Operat. Manage. 18 (1), 15–33.

Bimpikis, K., Ozdaglar, A., Yildiz, E., 2016. Competitive targeted advertising over networks. Oper. Res. 64 (3), 705–720.

Brooker, P., Barnett, J., Cribbin, T., Sharma, S., 2016. Have we even solved the first 'big data challenge?' Practical issues concerning data collection and visual representation for social media analytics. In: Digital Methods for Social Science, Springer, pp. 34–50.

Cai, Y., Chen, Y., Siqin, T., Choi, T.M., Chung, S.H., 2019. Pay upfront or pay later? Fixed royal payment in sustainable fashion brand franchising. Int. J. Prod. Econ. 214, 95–105.

Chan, H.K., Wang, X., Lacka, E., Zhang, M., 2016. A mixed-method approach to extracting the value of social media data. Prod. Operat. Manage. 25 (3), 568–583. Chen, H., Zheng, Z., Ceran, Y., 2016. De-biasing the reporting bias in social media analytics. Prod. Operat. Manage. 25 (5), 849–865.

Chen, Z.-Y., Fan, Z.-P., Sun, M., 2015. Behavior-aware user response modeling in social media: learning from diverse heterogeneous data. Eur. J. Oper. Res. 241 (2), 422–434.

Chen, Z.-Y., Fan, Z.-P., Sun, M., 2019. Individual-level social influence identification in social media: A learning-simulation coordinated method. Eur. J. Oper. Res. 273 (3), 1005–1015.

Chiu, C.H., Chan, H.L., Choi, T.M., 2019;al., in press. Risk minimizing price-rebate-return contracts in supply chains with ordering and pricing decisions: a multimethodological analysis. IEEE Trans. Eng. Manage (in press).

Chiu, C.H., Choi, T.M., Dai, X., Shen, B., Zheng, J.H., 2018. Optimal advertising budget allocation in luxury fashion markets with social influences: a mean-variance analysis. Prod. Operat. Manage. 27 (8), 1611–1629.

Chiu, C.H., Choi, T.M., Tang, C.S., 2011. Price, rebate, and returns supply contracts for coordinating supply chains with price-dependent demands. Prod. Operat. Manage. 20 (1), 81–91.

Chod, J., Trichakis, N., Tsoukalas, G., Aspegren, H., Weber, M., 2018. Blockchain and the value of operational transparency for supply chain finance. Working paper, Mack Institute for Innovation Management, Boston College. 25 Nov.

Choi, T.M., 2018. Incorporating social media observations and bounded rationality into fashion quick response supply chains in the big data era. Transp. Res. Part E 114, 386–397.

Choi, T.M., 2019. Blockchain-technology-supported platforms for diamond authentication and certification in luxury supply chains. Transp. Res. – Part E 128, 17–29. Choi, T.M., 2020Choi, in press. Mobile-app-online-website dual channel strategies: Privacy concerns, e-payment convenience, channel relationship, and coordination. IEEE Trans. Syst., Man, Cybernet.: Syst (in press).

Choi, T.M., Cai, Y., Shen, B., 2019a. Sustainable fashion supply chain management: a system of systems analysis. IEEE Trans. Eng. Manage. 66 (4), 730-745.

Choi, T.M., Cheng, T.C.E., Zhao, X., 2016. Multi-methodological research in operations management. Prod. Operat. Manage. 25 (3), 379–389.

Choi, T.M., Feng, L., Li, R., 2019b. Information disclosure structure in supply chains with rental service platforms in the Bblockchain technology era. Int. J. Prod. Econ (in press).

Choi, T.M., Lambert, J.H., 2017. Advances in risk analysis with big data. Risk Anal. 37 (8), 1435–1442.

Choi, T.M., Liu, N., 2019. Optimal advertisement budget allocation and coordination in luxury fashion supply chains with multiple brand-tier products. Transport. Res.

- Part E: Logistics Transport. Rev. 130, 95-107.
- Choi, T.M., Luo, S., 2019. Data quality challenges for sustainable fashion supply chain operations in emerging markets: roles of blockchain, government sponsors and environment taxes. Transport. Res. Part E 131, 139–152.
- Choi, T.M., Wallace, S.W., Wang, Y., 2018a. Big data analytics in operations management. Prod. Operat. Manage. 27 (10), 1868-1883.
- Choi, T.M., Wen, X., Sun, S.T., Chung, S.H., 2019c. The mean-variance approach for global supply chain risk analysis with air logistics in the blockchain technology era. Transport. Res. Part E 127, 178–191.
- Choi, T.M., Zhang, J., Cheng, T.C.E., 2018b. Quick response in supply chains with stochastically risk sensitive retailers. Decision Sci. 49 (5), 932-957.
- Cui, R., Gallino, S., Moreno, A., Zhang, D.J., 2018. The operational value of social media information. Prod. Operat. Manage. 27 (10), 1749-1769.
- DHL Trend Research, 2018. Blockchain in logistics, pp. 1-22.
- Dong, B., Li, M., Sivakumar, K., 2019; al., in press. Online review characteristics and trust: a cross-country examination. Decision Sci (in press).
- Fan, Y., Niu, R.H., 2016. To tweet or not to tweet? Exploring the effectiveness of service recovery strategies using social media. Int. J. Operat. Prod. Manage. 36 (9), 1014–1036.
- Fry, J., Binner, J.M., 2016. Elementary modelling and behavioural analysis for emergency evacuations using social media. Eur. J. Oper. Res. 249 (3), 1014–1023. Fung, Y.N., Choi, T.M., Liu, R., 2020. Sustainable planning strategies in supply chain systems: Proposal and applications with a real case study in fashion. Prod. Plan. Control (in press).
- Ghose, A., Ipeirotis, P.G., Li, B., 2019. Modeling consumer footprints on search engines: an interplay with social media. Manage. Sci. 65 (3), 1363-1385.
- Grover, V., Malhotra, M.K., 2003. Transaction cost framework in operations and supply chain management research: theory and measurement. J. Oper. Manage. 21 (4), 457–473.
- Gu, B., Ye, Q., 2014. First step in social media: measuring the influence of online management responses on customer satisfaction. Prod. Operat. Manage. 23 (4), 570–584.
- Guha, S., Kumar, S., 2018. Emergence of big data research in operations management, information systems, and healthcare: Past contributions and future roadmap. Prod. Operat. Manage. 27 (9), 1724–1735.
- Gunnec, D., Raghavan, S., 2017. Integrating social network effects in the share-of-choice problem. Decision Sci. 48 (6), 1098-1131.
- Guo, H., Pathak, P., Cheng, H.K., 2015. Estimating social influences from social networking sites—articulated friendships versus communication interactions. Decision Sci. 46 (1), 135–163.
- Guo, S., Choi, T.M., Shen, B., 2020. Green product development under competition: a study of the fashion apparel industry. Eur. J. Oper. Res. 280 (2), 523-538.
- Guo, S., Choi, T.M., Shen, B., Jung, S., 2019. Inventory management in mass customization operations: a review. IEEE Trans. Eng. Manage. 66 (3), 412–428.
- Hitt, M.A., Xu, K., Carnes, C.M., 2016. Resource based theory in operations management research. J. Oper. Manage. 41, 77-94.
- Holsapple, C., Hsiao, S.-H., Pakath, R., 2014. Business social media analytics: Definition, benefits, and challenges. In: The 20th Americas Conference on Information Systems, Savannah, pp. 1–12.
- Hou, R., de Koster, R., Yu, Y., 2018. Service investment for online retailers with social media—Does it pay off? Transp. Res. Part E 118, 606-628.
- Huang, Y., Singh, P.V., Ghose, A., 2015. A structural model of employee behavioral dynamics in enterprise social media. Manage. Sci. 61 (12), 2825–2844.
- Ismail, L., Maerwala, H., 2019. A review of blockchain architecture and consensus protocols: Use cases, challenges, and solutions. Symmetry 11, 1198.
- Kalyan, S., Feng, Q., Ganeshan, R., Sanders, N.R., Shanthikumar, J.G., 2018. Introduction to the special issue on perspectives on big data. Prod. Operat. Manage. 27 (9), 1639–1641
- Klausen, J., Marks, C.E., Zaman, T., 2018. Finding extremists in online social networks. Oper. Res. 66 (4), 957-976.
- Klumpp, M., Zijm, H., 2019Klumpp and Zijm, in press. Logistics innovation and social sustainability: how to prevent an artificial divide in human–computer interaction. J. Bus. Logistics (in press).
- Kumar, S., Mookerjee, V., Shubham, A., 2018. Research in operations management and information systems interface. Prod. Operat. Manage. 27 (11), 1893–1905. Kumar, N., Venugopal, D., Qiu, L., Kumar, S., 2019. Detecting anomalous online reviewers: an unsupervised approach using mixture models. J. Manage. Inform. Syst. 36 (4), 1313–1346.
- Lam, H.K., Yeung, A.C.L., Cheng, T.C.E., 2016. The impact of firms' social media initiatives on operational efficiency and innovativeness. J. Oper. Manage. 47, 28–43. Lau, R.Y.K., Zhang, W., Xu, W., 2018. Parallel aspect-oriented sentiment analysis for sales forecasting with big data. Prod. Operat. Manage. 27 (10), 1775–1794.
- Lee, D., Hosanagar, K., Nair, H.S., 2018. Advertising content and consumer engagement on social media: evidence from Facebook. Manage. Sci. 64 (11), 5105-5131.
- Leetaru, K., 2019a. Social media analytics is a disaster: Why can't we fix it?. Available at https://www.forbes.com/sites/kalevleetaru/2019/02/16/social-media-analytics-is-a-disaster-why-cant-we-fix-it/ (accessed 25 October 2019).
- Leetaru, K., 2019b. Social media analytics is failing as social media becomes more visual. available at https://www.forbes.com/sites/kalevleetaru/2019/01/25/social-media-analytics-is-failing-as-social-media-becomes-more-visual/ (accessed 25 October 2019).
- Li, G., Li, L., Choi, T.M., Sethi, S.P., 2019;al., in press. Green supply chain management in Chinese firms: Innovative measures and the moderating role of quick response technology. J. Oper. Manage (in press).
- Lobel, I., Sadler, E., Varshney, L.R., 2017. Customer referral incentives and social media. Manage. Sci. 63 (10), 3514-3529.
- Luna, S., Pennock, M.J., 2018. Social media applications and emergency management: a literature review and research agenda. Int. J. Disaster Risk Reduct. 28, 565–577.
- Ma, J., Tse, Y.K., Wang, X., Zhang, M., 2019. Examining customer perception and behaviour through social media research an empirical study of the United Airlines overbooking crisis. Transp. Res. Part E 127, 192–205.
- Massimino, B., 2016. Accessing online data: web-crawling and information-scraping techniques to automate the assembly of research data. J. Bus. Logistics 37 (1), 34–42.
- Mire, S., 2018. Blockchain for social media: 11 possible use cases. available at https://www.disruptordaily.com/blockchain-use-cases-social-media/ (accessed 21 October 2019).
- Mishra, N., Singh, A., 2018. Use of twitter data for waste minimisation in beef supply chain. Ann. Oper. Res. 270 (1-2), 337-359.
- Neely, A., 2005. The evolution of performance measurement research: developments in the last decade and a research agenda for the next. Int. J. Operat. Prod. Manage. 25 (12), 1264–1277.
- Niu, B., Zou, Z., 2017. Better demand signal, better decisions? Evaluation of big data in a licensed remanufacturing supply chain with environmental risk considerations. Risk Anal. 37 (8), 1550–1565.
- Pekgün, P., Galbreth, M.R., Ghosh, B.P., 2018. How unequal perceptions of user reviews impact price competition. Decision Sci. 49 (2), 250-274.
- Radar, S., 2018. Top 10 Blockchain social media alternatives to watch out for. available at http://fintechnews.sg/20994/blockchain/top-10-blockchain-social-media-alternatives-to-watch-out-for/ (accessed 21 October 2019).
 Ramanathan, U., Subramanian, N., Parrott, G., 2017. Role of social media in retail network operations and marketing to enhance customer satisfaction. Int. J. Operat.
- Prod. Manage. 37 (1), 105–123.
- Saberi, S., Kouhİzadeh, M., Sarkis, J., Shen, L., 2019. Blockchain technology and its relationships to sustainable supply chain management. Int. J. Prod. Res. 57 (7), 2117–2135.
- Schoenherr, T., Speier-Pero, C., 2015. Data science, predictive analytics, and big data in supply chain management: current state and future potential. J. Bus. Logistics 36 (1), 120–132.
- Sheu, J.B., Choi, T.M., 2019. Extended consumer responsibility: syncretic value-oriented pricing strategies for trade-in-for-upgrade programs. Transport. Res. Part E: Logistics Transport. Rev. 122, 350–367.
- Shi, X., Chan, H.L., Dong, C., 2018;al., in press. Value of bargaining contract in a supply chain system with sustainability investment: an incentive analysis. IEEE Trans. Syst., Man, Cybernet.: Syst (in press).
- Singh, A., Shukla, N., Mishra, N., 2018. Social media data analytics to improve supply chain management in food industries. Transp. Res. Part E 114, 398–415. Sodero, A.C., Rabinovich, E., 2017. Demand and revenue management of deteriorating inventory on the internet: an empirical study of flash sales markets. J. Bus.

Logistics 38 (3), 170-183.

Sodhi, M.S., Tang, C.S., 2018Sodhi and Tang, in press. Research opportunities in supply chain transparency". Prod. Operat. Manage (in press).

Srinivasan, A., Guo, H., Devaraj, S., 2017. Who cares about your big day? Impact of life events on dynamics of social networks. Decision Sci. 48 (6), 1062–1097. Stieglitz, S., Mirbabaie, M., Ross, B., Neuberger, C., 2018. Social media analytics–Challenges in topic discovery, data collection, and data preparation. Int. J. Inf. Manage. 39, 156–168.

Sul, H.K., Dennis, A.R., Yuan, L., 2017. Trading on twitter: Using social media sentiment to predict stock returns. Decision Sciences 48 (3), 454-488.

Sun, L., Teunter, R.H., Babai, M.Z., Hua, G., 2019;al., in press. Optimal pricing for ride-sourcing platforms. Eur. J. Oper. Res (in press).

Syed, R., Dhillon, G., Merrick, J., 2019;al., in press. The identity management value model: A design science approach to assess value gaps on social media. Decision Sci (in press).

Tang, C.S., Veelenturf, L.P., 2019. The strategic role of logistics in the industry 4.0 era. Transp. Res. Part E 129, 1-11.

Tian, Y., Choi, T.M., Ding, X., Xing, R., Zhao, J., 2019. A grid cumulative probability localization based industrial risk monitoring system. IEEE Trans. Autom. Sci. Eng. 16 (2), 557–569.

Waller, M.A., Fawcett, S.E., 2014. Click here to print a maker movement supply chain: how invention and entrepreneurship will disrupt supply chain design. J. Bus. Logistics 35 (2), 99–102.

Wang, C.-Y., Chen, Y., Liu, K.J.R., 2018. Game-theoretic cross social media analytic: how yelp ratings affect deal selection on groupon? IEEE Trans. Knowledge Data Eng. 30 (5), 908–921.

Wang, Y., Lin, J., Choi, T.M., 2020;al., in press. Gray market and counterfeiting in supply chains: a review of the operations literature and implications to luxury industries. Transport. Res. Part E: Logistics Transport. Rev (in press).

Wang, J.-C., Wang, Y.-Y., Che, T., 2019a. Information sharing and the impact of shutdown policy in a supply chain with market disruption risk in the social media era. Inform. Manage. 56 (2), 280–293.

Wang, Y., Singgih, M., Wang, J., Rit, M., 2019b. Making sense of blockchain technology: How will it transform supply chains? Int. J. Prod. Econ. 211, 221–236.

Yan, L., Yan, X., Tan, Y., Sun, S.X., 2019. Shared minds: How patients use collaborative information sharing via social media platforms. Prod. Operat. Manage. 28 (1), 9–26.

Yoo, E., Rand, W., Eftekhar, M., Rabinovich, E., 2016. Evaluating information diffusion speed and its determinants in social media networks during humanitarian crises. J. Operat. Manage. 45, 123–133.

Zeng, D., Chen, H., Lusch, R., Li, S.-H., 2010. Social media analytics and intelligence. IEEE Intelligent Syst. 25 (6), 13–16.

Zhan, Y., Tan, K.H., 2018Zhan and Tan, in press. An analytic infrastructure for harvesting big data to enhance supply chain performance. Eur. J. Operat. Res (in press).