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The Role of Data-Driven Decision Making in Business Strategy

Chantal Emmanuel Sylvestre

Faculty of Education Kampala International University Uganda

ABSTRACT

In the era of Big Data, businesses face an unprecedented volume, variety, and velocity of data. The effective use of this data is becoming crucial for strategic decision-making. This paper explores the role of data-driven decision-making (DDDM) in shaping business strategies, emphasizing the transition from intuition-based to evidence-based approaches. It examines the foundational principles of DDDM, including the significance of data quality, the integration of big data analytics, and the balance between human intuition and data insights. The paper also discusses the implementation challenges, including technological, ethical, and organizational factors, while highlighting the benefits such as enhanced operational efficiency, improved risk management, and increased innovation. Finally, it addresses future trends in DDDM, particularly the growing importance of analytics in corporate governance and strategic planning, and the implications for business strategy in a data-centric world.

Keywords: Data-Driven Decision Making (DDDM), Big Data, Business Strategy, Data Analytics, Evidence-Based Decision Making.

INTRODUCTION

With the transformation of information technology space, firms have an unprecedented digital and diversified data explosion with exceedingly growing size and complexity that include all kinds of unstructured, semi-structured and structured info from diverse continuously changing sources such as social media, emails, internet articles, images, open government site, and databases Banks and telecommunication firms). This explosive growth of data is often referred to as Big Data (BD). Massive structured and unstructured digital info is continually produced and shared at an unprecedented scale and pace in the world. Over 90% of the presently data available worldwide has been created in the last several years and more than 2.5 quintillion bytes of data is generated every day. In early stage, social networking services like Face book, Twitter, linked in, banking transactions and telecommunication messages exponentially generated a large amount of unstructured data and turned web 2.0 era into a data-centric era. BD is a term used to express digital data that is huge in size, come from various sources, media and digital, and keep on increasing in size. The three dimensions of BD are referred as volume, variety and velocity. With the advancement of billions of sensors such as mobile phones, advanced equipment's, and smart devices, firms built on the massive sources of data streams into their organizations [1]. One of the most important challenges that BD poses to business decision making is the growth of potentially erroneous info and noise that often emerges simultaneously. To wisely make decisions using BD, such noise and misunderstanding must be filtered off. Potentially erroneous info/knowledge needs to be judged to take into account the level of uncertainty of the info. Since the first growth of BD, organizations have been implementing BD technologies and tools. Some of these technologies and tools have been successfully used while some of them have not been so effectively implemented. There is an increasing recognition that such technologies and tools and data (large/semi-complex) differently influence the strategic decision making process in the organization. Despite that very little attention has been paid to how BD technologies assist decision makers in developing strategic decisions [2].

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FOUNDATIONS OF DATA-DRIVEN DECISION MAKING

This section delves into the fundamental concepts and principles that underpin data-driven decision making. It explores the theoretical and practical foundations that are essential for understanding the role of data in shaping business strategy. The section may cover topics such as data sources, types of data, and the relevance of data quality in decision making. A look at the potential of big data in nurturing intuition in organisational decision makers. As big data (BD) and data analytics have gained significance, the industry expects helping executives will eventually move towards evidence-based decision making. A key question is whether executives make decisions by intuition and whether big data would ever substitute human intuition. This research investigates the 'mind-set' of executives about the application and limitations of big data by considering their decision-making behaviour. The aim is to look into how BD technologies facilitate greater intuitiveness in executives, leading to faster and sustainable business growth. Big Data (BD) refers to digital data that is huge in size, usually in terabits, comes from various sources, and continues to increase. Organisations have used this data to generate insights and grow their businesses at an unprecedented rate. Initially, the success of BD started from internet-based companies that mined data to increase their market share or differentiate themselves. BD-based IT technologies for strategic decision making have resulted in gaining insights that lead to building a sustainable competitive advantage and can help executives develop competitive institutions that stimulate innovation [3]. The Role of Big Data in Influencing Strategic Decision-Making for Organizations: A Review. In recent years, many scholars have become interested in the increasingly important role of Big Data in strategic management. However, somewhat of a research gap exists on the role of Big Data in influencing strategic decision-making. This paper provides a focused review of literature on the increasingly important role of Big Data in strategic management, with special attention to the role of Big Data in strategic decision-making. Big Data represents one of the most important assets available to strategic decision-makers. Big Data refers to incredibly large datasets that are either well-structured or unstructured, and the value of Big Data increases over time. Extracting and analyzing information from Big Data enables managers and leaders to make informed predictions and decisions about critical business issues. Big Data offers many opportunities for companies to make better decisions and to increase the value of their products and services. The major questions investigated were: What kind of decisions is being supported by Big Data in organizations today? Does Big Data help organizations make better decisions? How can organizations best make use of Big Data to support strategic decision-making? This review of literature includes articles from well-respected English-language business journals that discussed the connection between Big Data and strategic decision-making [4].

UNDERSTANDING DATA-DRIVEN DECISION MAKING

Within the broader context of the foundations, this subsection provides a detailed exploration of the concept of data-driven decision making. It may discuss the process, methodologies, and frameworks involved in leveraging data for decision making within the business domain. Additionally, it may address the significance of data literacy and the role of technology in understanding and implementing data-driven strategies [5]. Making decisions based on data is no longer merely a proposed concept; it is now a necessity for survival in the business world and society in general. Data-driven decision making is no longer a visionary goal; it is an immediate path that organizations, businesses, and societies must take to ensure the purposeful and timely use of all data on hand. However, various terms such as business intelligence, big data, open data, data-driven innovation, analytical culture, data-driven science, and machine learning have been used inconsistently or interchangeably. Therefore, to avoid convoluted, confused, and conflicting perceptions within the extensive discussion, it is imperative to first address the understanding of data-driven decision making before exploring its use, quality, conditions, challenges, or any other closely related aspects [6]. Data can be comprised of textual, numeric, audio, graphical, or other materials that, on their own, have no significance. Data become information when data are meaningful and interpretable (e.g., showing a time series of prices); information becomes knowledge when properly understood (i.e., having been structured in comparison with other data, information, and knowledge). Knowledge becomes wisdom when understood at various levels of abstraction (e.g., that the price tends to follow an upward or downward trend), helping answer the why. The heart of decision making within organizations of any kind, from science and public office to life and death, is wisdom. If decision making is driven by anything but wisdom, the odds for purposes being achieved are comically slight or worse [7].

IMPLEMENTING DATA-DRIVEN DECISION MAKING

More and more organizations are using data to help with their decisions. By doing this, they can make sure that the plans they come up with will actually lead to the results they want. Nowadays there are

many techniques that can help organizations plan ahead. These techniques fit well with data mining, which involves getting useful information from large amounts of facts. Data mining needs to be used wisely; it may give many suggestions but only some of those will fit well with a specific organization [8]. If an organization does not have the right technology or acceptance model, it will be difficult to make use of data mining techniques. People need to understand that these techniques will only operate better than the traditional planning models when they are used by specialists who know how to interpret and use the results from the analysis. Organizations also need to realize that data mining cannot replace the knowledge of experts. Data mining techniques can efficiently identify potential plans but relying solely on them will not lead to the desired results. Every identified plan still needs to be carefully scrutinized [9].

DATA COLLECTION AND ANALYSIS

Under the implementation framework, data collection and analysis constitute one critical component. First, data collection methods will be considered in response to the objective of the study (e.g. quantitative, qualitative, or both), the source of data (e.g. primary, secondary), and the time dimension (cross-sectional, longitudinal). To justify these characteristics of the methodology, the advantages and disadvantages of the approaches are elaborated based on prior studies. Second, tools for data analysis in line with the chosen collection methods will be proposed, and the subsequent process will be outlined. The sampling strategy will also be described, along with the number of business practitioners targeted and justifications based on prior research. As a part of trustworthiness, the evaluation criteria and means of achieving them will be discussed. Lastly, data collection ethics will be addressed [10]. Responding to the study objective, a qualitative approach will be used to collect empirical data focusing on cause-and-effect relationships in terms of the choices business practitioners have made, and the reasons and outcomes of these choices. Addressing previous calls for more qualitative research on strategy processes and how they are affected by contextual factors, the qualitative approach will add valuable knowledge to the strategy-making discourse. The approach is in line with the interpretivist paradigm that recognizes subjectivity and attempts to explain social reality based on participants' understanding of their contexts. The understanding of business practitioners will be the main focus, which is a prerequisite for building a theory of the context-specific descriptions, experiences, interpretations, or perceptions of social realities of these participants. A qualitative approach is further considered appropriate because decisions should be examined in detail to identify the baseline responses that subsequently influence successful implementation [11].

BENEFITS AND CHALLENGES OF DATA-DRIVEN DECISION MAKING

Data-driven decisions have become very important in today's business environment where more information keeps coming in due to technological improvements. This availability of knowledge allows people to make rational and direct decisions that can help their businesses grow. Numerous studies have been done to see what kinds of benefits come from using evidence in decision-making processes. The general agreement is that a data-driven approach improves performance up to 5% and increases the chances of innovation. A European Commission survey done linking competitiveness index to statistical benchmarks of evidence-based decision making showed that, on average, a data-driven strategy could increase performance up to 15%. In addition, it was demonstrated that adopting data-driven decision making in business could increase competitiveness index up to 10%. Since the costs of investment in analytics tools are considered small in comparison to these benefits, it is expected that by sharing its knowledge on data-driven decisions, the EU could establish a competitive advantage in the twenty-first century [12]. The implications of consumer privacy, cyber-security, and transparency issues when using big data technologies for decision making are discussed at the end. Current media stories indicate that there has been a lot of misuse of personal data often without their agreement, keeping the actual value of that data hidden. These events raise doubts regarding the reliability of big data technologies that allow uncovering information that otherwise would be hidden. Investigations of big data decision-making practices used by the US in military operations developed many questions regarding their fairness. There is evidence that the policy of the Department of Defense does not assure transparent processes of how algorithmic decisions are made nor provides instruction on how to challenge them. These uncertainty and unfairness issues, amongst others, have been raised by scientists and brought to the attention of the public in numerous media stories. Given the limitations of big data, there is a doubt regarding the capability of these technologies to substitute human decision-makers when it is believed that their decisions might deviate from what is believed to be right [13].

BENEFITS

Organizations can reap numerous advantages by embracing data-driven decision-making, developing and implementing data-driven strategies supported by data analytics technologies. Several potential benefits

include the following: Organizations can achieve improved business insights by enhancing their data visibility, accessibility, understanding, and analysis, transforming data into valuable assets and core competencies. A greater awareness of organizational performance is gained by establishing a data-driven culture that encourages employees at all levels to collect and examine data using advanced data analytics technologies. These technologies reveal patterns, trends, and insights, illuminating areas for improvement and uncovering opportunities that might have been overlooked by traditional operational methods [14]. Data-driven decision-making increases operational efficiency by optimizing operational processes, reducing production costs, and improving customer service and satisfaction through personalized experiences and communication. Data analytics technologies can provide better customer relationship management, form regime analytics to identify risk mitigation techniques, and evaluate the creditworthiness of potential customers and clients. By recognizing fraudulent or erroneous activities, security and fraud detection can be enhanced. Addressing any problems raises vigilance levels, while the increased detection speeds become a protective measure [15]. Data analytics technologies enhance risk management, enabling organizations to make accurate and reliable decisions by predicting the likelihood and associated costs of alternative scenarios. Data-driven decision-making leads to informed strategic planning, where growth and expansion decisions are based on factual data analysis, detecting market trends and client behavior alterations before implementation. Strategies will be more realistic and accurate, resulting in fewer resources wasted in failed attempts at growth [16].

FUTURE TRENDS IN DATA-DRIVEN DECISION MAKING DECISION MAKING: IMPLICATIONS FOR BUSINESS STRATEGY

There are a number of industry forces expected to affect the maturation of decision making. As the demand for more rigorous and disciplined decision management processes increases, this demand will be met in part by the gradual diffusion of the more sophisticated types of decision management practices that are now limited largely to a few leading businesses. Eventually decision management will be more firmly established as an integral part of corporate governance that combines both formal and informal processes. Society, particularly in the Western world, is moving toward an era of more governance of all organizations including businesses, NGOs, and government agencies. In their drive for more control over decision making, boards of directors who are accountable to stakeholders will be far more involved than they are today in setting business strategies, approving key decision rules, and sanctioning make-or-buy decisions. Linked to this, there is a growing expectation of transparency and accountability to stakeholders across a range of issues including business strategies and decision processes. This raises questions regarding how the business strategy development and decision making processes of companies can be monitored by stakeholders. Companies will be developing innovative processes to monitor the formal and informal elements on which decision making is based and how both these interact to produce strategic outcomes [17]. Emerging into this uncertain environment at the present time for business strategy development are a number of technology and social media trends including social networks, blogs, wikis, and advances in analytics of big data. There are opportunities for companies to leverage these developments for improved decision making. The most powerful of these opportunities for change is analytics, which offers the potential significantly to enhance the evidential basis of business strategy development. The ever-increasing size and complexity of the data available will lead decision making in the direction of ever-increasing sophistication and sophistication. Better decision results are likely to accrue to those organizations able to make best use of their data analytics capabilities, particularly the forecasting and worst-case simulation capabilities on which a few groups currently focus and invest [18].

CONCLUSION

The integration of data-driven decision-making into business strategy is no longer an option but a necessity in the modern digital landscape. By leveraging the vast amounts of data available, organizations can make more informed, accurate, and strategic decisions that drive growth, innovation, and competitive advantage. However, successful implementation requires overcoming challenges related to data quality, technological infrastructure, and the human factors involved in decision-making. As the business environment continues to evolve, the ability to harness data effectively will become a key determinant of organizational success. Embracing DDDM will not only enhance operational efficiency and risk management but also ensure that businesses remain agile and responsive to market changes. The future of business strategy lies in the continuous refinement of decision-making processes through advanced analytics, ensuring that organizations can navigate the complexities of the data-centric world.

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