

Chapter 65

Big Data, Data Management, and Business Intelligence

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ABSTRACT

This paper examines big data and the opportunities it presents for improved business intelligence and decision making. Big data comes in multiple forms. It can be structured, semi-structured, or unstructured. The opportunity it presents is that there is so much of it and it is readily available to organizations. Organizations use big data for business intelligence (BI). They can apply analytics in BI activities to assess big data in order to gain new insights and opportunities for decision making. The problem is that oftentimes the data is of poor quality and it contains personal information. This paper explores these issues and examines the importance of effective data management in facilitating sound business intelligence. The Master Data Management methodology is reviewed and the importance of management support in its deployment is emphasized. With the advent of new sources of big data from IoT devices, the need for even more management involvement is stressed to ensure that organizational BI yield sound decisions and that use of data are in compliance with new regulations.

INTRODUCTION

Big data is one of the most commonly written about topics in today's press. In fact, today it has become a ubiquitous term. It is continuously transforming business, finance, healthcare, medicine, engineering, science, and society itself. Routinely we are provided with new information telling us much more data there is, how much more is now able to be captured, how many new sources it comes from, new ways to process it, and how it can be used in new and novel for decision making. It is a topic that is regularly discussed in boardrooms, business publications, and the mainstream media, because big data provides new insights into everything. Big data encompasses traditional sources of structured transaction data that is now supplemented by mass quantities of unstructured data. This data is processed by new, inexpensive, and faster hardware that is then scrutinized by new and more advanced analytics that provide

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organizations with more in-depth insight into their operational environment than ever before. And now, role of decision maker is increasingly being assumed by smart machines.

The goal of business intelligence [BI] is to extract value from data to better inform decision making. BI combines text, video, voice, location data, social media, and any other new source of data with traditional data sets in order to learn about, interact with, and predict what is happening so that the organization can respond as fast as possible to whatever it perceives is the opportunity that the data reveals. BI deals with imperfect data that is oftentimes ambiguous, but which is available on a vast scale. As a result, Mayer-Shonberger and Cukier (2014) assert that the effect is that the extraction of value from big data is analogous to a treasure hunt. That is, organizations are scrutinizing big data to learn what is happening, without necessarily needing to understand why. They argue that in a big data world, correlations supersede causality, because the data is simply used to discover patterns and correlations in the data that offer novel and invaluable insights. The more data you have the better the insights. The underlying premise for BI then becomes this: the more data an organization can capture, the better the data-driven probability of understanding what is happening, and the faster you can respond to this insight. This means then that actions taken in BI are often based on an organizational confidence level in the analytic assessment of what the data suggests without the necessity of a clear understanding of the root cause.

Big data would appear to many to be more about systems, and less about people. Certainly, people are important because they are themselves a major source of big data fodder and it is oftentimes people's behavior that big data is trying to affect. Nevertheless, big data also is dependent on people because people must inevitably be responsible for how data is used, how it is managed, and for the consequences of the decisions made when using it.

This paper is intended then to remind us that big data is not simply something that data, systems, and analytics make happen and that we are somehow divorced from it and not responsible for unintended consequences. Instead, prudence would require that since we have unleashed big data, we have to somehow insure to the best of our ability that if we can't control big data, we can at least use common sense in how we approach it and manage it. And, despite new technologies enabling machines to use big data to increasingly make decisions, people will not be able to divorce themselves from their decision-making responsibilities.

BACKGROUND

The amount of data in our world has exploded exponentially such that data, especially unstructured data, is now referred to as "big data". Where measures of data were once gradually evolving from megabytes to terabytes, the sudden phenomena of big data accelerated these measures to volumes expressed in petabytes (1,024 terabytes) or exabytes (1,024 petabytes). The new influx of data is derived from billions to trillions of records of millions of people—all from different sources (e.g. Web, sales, customer contact center, social media, mobile data and so on). The data is typically loosely structured and often incomplete.

Petrov (2019) states that in 2019, the big data market is expected to grow by 20 percent, with every person generating 1.7 megabytes of data in just a second. By 2020, he claims, there will be approximately 40 trillion gigabytes of data. He notes that 97.2 percent of organizations are investing in big data and artificial intelligence (AI), because automated analytics will become increasingly vital to big data by 2020.

Big Data is the natural result of four major global trends:

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