

Chapter 3

Business Intelligence and Data Mining

Zsolt T. Kardkovács
UI Research, Hungary

ABSTRACT

Whenever decision makers find out that they want to know more about how the business works and progresses, or why customers do what they do, then data miners are summoned, and business intelligence is to be built or altered. Data mining aims at retrieving valid, interesting, explicable connection between key factors for either operative reporting or supporting strategic planning. While data mining discovers static connections between factors, business intelligence visualizes relevant data for decision makers in order to make them identify fast changes and analyze precisely business states. In this chapter, the authors give a short introduction for data oriented decision support systems with data mining and business intelligence in it. While these techniques are widely used in business processes, there are much more bad practices than good ones. We try to make an attempt to demystify and clear the myths about these technologies, and determine who should and how (not) to use them.

INTRODUCTION

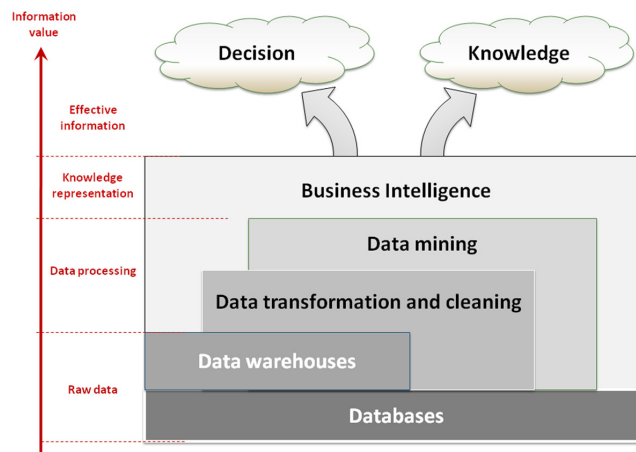
Data-driven decision support systems (DSS) are the state-of-the-art technology for an information era company to analyze current states, to discover changes in customer behaviors, or to get a picture about present and future business. DSS consist of operative databases (current state of business), data

warehouses (historic business data) for aggregations, some analytics (including data cleaning and customer anonymization), and visualization platforms (See Figure 1).

Data mining processes deployed on the technical infrastructure of an enterprise is called analytics. Its purpose is either to detect some discrepancies in the data that may indicate frauds, changes

DOI: 10.4018/978-1-4666-4181-5.ch003

Figure 1. General overview of data driven decision support systems



in the business or the business environment, or to group customers, products, and so forth, using unique signatures in order to treat them properly. Data mining techniques are able to discover relevant features in business data, and determine latent semantics between these data, which allows for data mining algorithms to give extrapolations and predictions for future changes of these data. Once these latent semantics are discovered, a business analyst can analyze the proper features, and see whether a situation is in place or not.

Business intelligence is a process in which decision makers use interactive diagrams on time series data to retrieve information. Visualization helps analysts understand the large amount of information better as human visual processing are faster, more comprehensive, and more precise than any other human communication forms. Simple visualization techniques can only show fix states without letting know what it is behind the numbers. Business intelligence is essential to discover sudden changes, and to react faster for such changes.

Data mining and business intelligence involves a certain human factor as well. It does not only observe changes in the data, but the knowledge workers involved in the observation process also influence its output. Data mining and business intelligence trigger a learning cycle: first data

miners learn, then their algorithms learn, then data analysts learn, then reporting systems learn, and finally decision makers learn. That is, the success of a data mining or a business intelligence project largely depends on proper human interactions and intuitions: on the ones who observe trends, select and refine algorithms or choose data representations adequately, and on the ones who derive further implications and conclusions according to the findings. Data mining and business intelligence algorithms themselves neither understand business, nor make implications, but discover statistically relevant correlations. Since the human factor is an essential part in this process, even a well-founded project can end up in a disaster, if not properly controlled or verified.

This chapter presents the fundamental steps towards building up business intelligence based on data mining, and in each step we highlight the most typical human mistakes and misbelieves we found in the over 40 data mining projects we audited or attended as a consultant in the last 10 years. These projects cover various sectors including health care, bank and insurance businesses, telecommunications, and public services.

In this chapter, we use the term data mining for structured data only (i.e., data stored in relational schemas or as a stream of records, e.g., time series) without lack of generality. Literature distinguishes

12 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:

www.igi-global.com/chapter/business-intelligence-data-mining/78081

Related Content

A Techno-Economic Perspective of Constrained Application Protocol

Tapio Levä, Mahya Ilaghi, Vilen Looga, Miika Komu, Nicklas Beijarand Oleksiy Mazhelis (2017). *Driving Innovation and Business Success in the Digital Economy* (pp. 251-267).

www.irma-international.org/chapter/a-techno-economic-perspective-of-constrained-application-protocol/173198

On Some Misconceptions Concerning Digital Banking and Alternative Delivery Channels

Aijaz A. Shaikhand Heikki Karjaluo (2016). *International Journal of E-Business Research* (pp. 1-16).

www.irma-international.org/article/on-some-misconceptions-concerning-digital-banking-and-alternative-delivery-channels/157390

Towards a Virtual Enterprise Architecture for the Environmental Sector

Ioannis N. Athanasiadis (2008). *Agent and Web Service Technologies in Virtual Enterprises* (pp. 256-266).

www.irma-international.org/chapter/towards-virtual-enterprise-architecture-environmental/5003

Mobile Multi-Brand Loyalty Programs: Elaborating Customer Value and Satisfaction

Gokhan Aydin (2022). *International Journal of E-Business Research* (pp. 1-25).

www.irma-international.org/article/mobile-multi-brand-loyalty-programs/309397

What Makes Consumers Adopt a Wearable Fitness Device?: The Roles of Cognitive, Affective, and Motivational Factors

Jing Zhang and En Mao (2023). *International Journal of E-Business Research* (pp. 1-17).

www.irma-international.org/article/what-makes-consumers-adopt-a-wearable-fitness-device/323204