

Usability of Data Warehousing and Data Mining for Interactive Decision Making in Textile Sector

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Abstract

Data warehouse is one of the most rapidly growing areas in management information system. With this approach, data for Executive Information System (EIS) and Decision Support System (DSS) applications are separated from operational data and stored in a separate database. This process is called data warehousing. The major advantages of this approach are: improved in performance, better data quality, and the ability to consolidate and summarize data from heterogeneous systems. A data warehouse is part of a larger infrastructure that includes legacy data sources, external data sources, a repository, data acquisition software, and user interface and related analytical tools. The aim of this research work is to elaborate that how the textile industry can manage and improve their production capacity and resources at optimum level to produce a good quality result using data warehousing and data mining techniques. This research work is conducted in Masood Textile Mills Limited, Faisalabad, Pakistan (MTML). The results may hopefully open up an era of research and methodology that could further benefit the Industry to support in decision support system.

Index terms— Executive Information System, Decision Support System, data warehouse, data mining, production capacity, heterogeneous systems.

1 I. INTRODUCTION

storing summarized integrated business information in a central repository used by analytical application with different user perspective. It is a process of merging data in a centralized location. The data warehouses' databases usually store the complete organizational business history from start to end. The frequently used procedure for fetching decision making information from data warehouses is based on executing queries and tools which perform online analytical processing (OLAP) [1]. Data mining is a rising methodology used for identifying, extracting and evaluating variables to find useful information [2]. Data Mining is a process of extracting previously stored, valid, potential by useful and hidden patterns from large data sets as data volume is increasing rapidly day by day [3]. It allows the users to analyze data from different dimensions, categorize and summarize the relationships, identified during the mining process [4]. Different data mining Author ? : Dept of Computer Science, University of Agriculture, Faisalabad, 38000, Pakistan. E-mail : shakeelfaridi@gmail.com Author ? : Dept of Computer Science, University of Agriculture, Faisalabad, 38000, Pakistan. E-mail : tasleem_mustafa@uaf.edu.pk techniques are used in various organizations and institutions like pharmaceutical, telecommunication, engineering, education, banking, marketing, sale, etc [5]. It is very difficult to normalize data without using any sort of technique. To get required benefits and useful information from such large amount of stored data, you must use some data mining techniques like Classical Techniques: Statistics, Neighborhoods and Clustering and Next Generation Techniques: Trees, Networks and Rules [4]. These sorts of techniques facilitate the user to get useful and accurate information on time for interactive decision.

42 The objective of this study is to elaborate that how the textile industries can manage and improve their
43 production capacity and resources at optimum level to produce good quality results comparing from the historic
44 information using data warehousing and data mining techniques.

45 2 II.

46 3 THE USABILITY CONCEPT

47 The definition of usability or usability engineering by international standardization organization (ISO) and
48 usability-experts also defined it as "the effectiveness, efficiency and satisfaction with which specified user can
49 achieve the specified goals in particular environment" [10,11,12,13,14,15]. The basic purpose of usability is to
50 make sure that the work is smoothly going done by the end user by using his average ability without any
51 frustration [16, ???]. It not only discusses but also applies to the design of software system, applications and
52 product but also includes users" interface, supporting documents to run the software efficiently ??18,19]. There
53 is surety for unsuccessfulness of a fully functional software or application, if it has poor usability. As functionality
54 and

55 4 D

56 Usability is defined as the capability in human functional terms that to be used easily, effectively and satisfactorily
57 by specific users, performing specific responsibilities, in specific environments [23]. It can be measured analytically
58 through the use of inspection method ???; 8]. The analytical approach to measuring usability attempts to identify
59 actual usability issues that are practically faced by the end users ???; 9]. Usability inspection is regarded as a
60 cost-effective approach (to measuring usability) which relies on the review of usability experts [7]. Inspection
61 techniques differ from empirical techniques by identifying potential usability issues, as opposed to actual usability
62 issues. usability both are tasks and user related terms [20]. Functions need to match task requirements while
63 the users need to understand the exact functionality of the system to meet their requirements ??21]. "Grouping
64 related commands into menus, other examples of good interface design and documentation help overcome the
65 conflict between power and ease of use to enhance both usability and functionality [22]." Evaluator"s experience,
66 available resources like time, budget, and labor are the multiple methods to evaluate the usability. It also depends
67 on the stage of development tools.

68 5 III. COMMON USABILITY EVALUATION PRINCIPLES

69 There are some common usability evaluations principles are used to evaluate the requirements. These evaluation
70 principles are; ease of use, usefulness, customization, task support, flexibility, navigation, guidance, memorability,
71 system reliability, user interface, output presentation, learn ability, system responsiveness, accuracy and
72 completeness [6,24,25].

73 The usability evaluation principles like usefulness, guidance, customization, accuracy and completeness are
74 showing the amazing results in my work.

75 IV.

76 6 HOW TO IMPROVE DECISION MAKING BY USING 77 DATA MINING

78 Hence, data mining uses predictive techniques to expose patterns in the data. These patterns of a data playing
79 a vital role in the process of decision making since they expose the areas where improvements is required in the
80 whole process. Organizations can use data mining in such a way as to improve profitableness and effectiveness
81 of their interactions with their customers, improve the management of risk, and detect fraud. In other words,
82 these patterns that are expose by using data mining techniques to assist business organizations and stakeholders
83 of the organizations to make timely, accurately and better decisions [26].

84 7 Prediction

85 Forecasting, like estimation, is everywhere in business. Accurate prediction can reduce the cost, optimize your
86 sales and better utilization of available resources. If you can predict the future, you will definitely boost your
87 business and bring more profits ??17].

88 8 ii. Estimation

89 This process is useful in just about every surface of business. From finance to marketing to sales, the better you
90 can estimate your expenses, product mix optimization, or potential customer value, the better off you will be.
91 This and the next use are fairly self-evident if you have ever spent a day at a business ??17].

92 iii. Regression This is the most widely known and the oldest statistical technique that is utilized by the data
93 mining community. Essentially, regression makes use of a dataset to develop a mathematical formula which fits
94 the data. So whenever you want to use the results for predicting future behavioral patterns, all you need to do

95 is just take the new data, and apply it to the model that has been developed, and you will get your prediction.
96 But if you need to work with data that is categorical, where there is no significant order, such as gender, name,
97 or color, it is better to use a different technique.

98 [26] iv. Classification If you need to work with categorical data, or a combination of categorical data and
99 continuous numeric, classification analysis will meet your requirements. This technique has the capability to
100 process a more extensive variety of data compared to regression and is therefore increasingly popular. The
101 complex mathematical formula that the regression technique provides, in this you will be provided a decision
102 tree which requires a sequence of binary decisions. MTM have their complete and fully in-house developed
103 ERP system. It contains all important applications that provide the absolute solutions for its operations and
104 management as well. This well designed ERP system makes MTM competitive advantages over the others in this
105 region.

106 9 VII. CONCLUSION

107 The value of data warehousing and data mining can be seen in effective decision making based on concrete evidence
108 from the old data. The purpose of this research was basically to highlight the core area of business issues. These
109 core areas, where top management needs to be focused to make the organization more profitable by using latest
110 technologies. In order to continue aggressive data mining techniques must be used to save money, achieve better
111 operational performance and mitigate risk. Cost saving will be achieved by maintaining or improving production
112 performance. It would be difficult to achieve more profitability by ignoring these loop holes. The present study
113 was focused only one area of the business by using data mining for decision analysis. In future the author will
try to discuss the other core area of business that can directly affect the organization's objectives. ^{1 2}

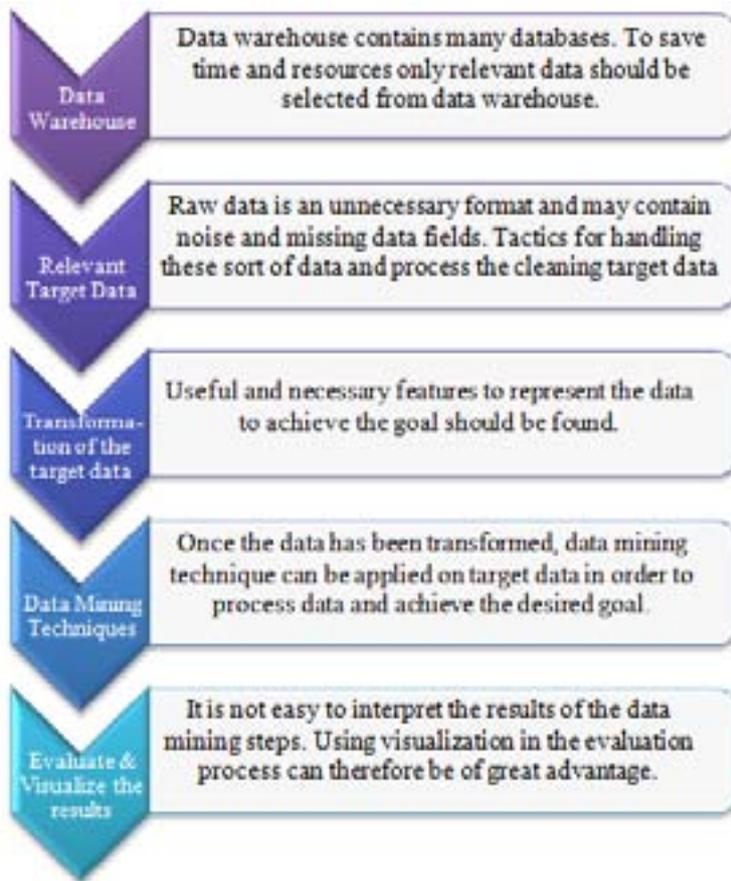


Figure 1: © 1 2012

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Figure 2: Figure . 1 :

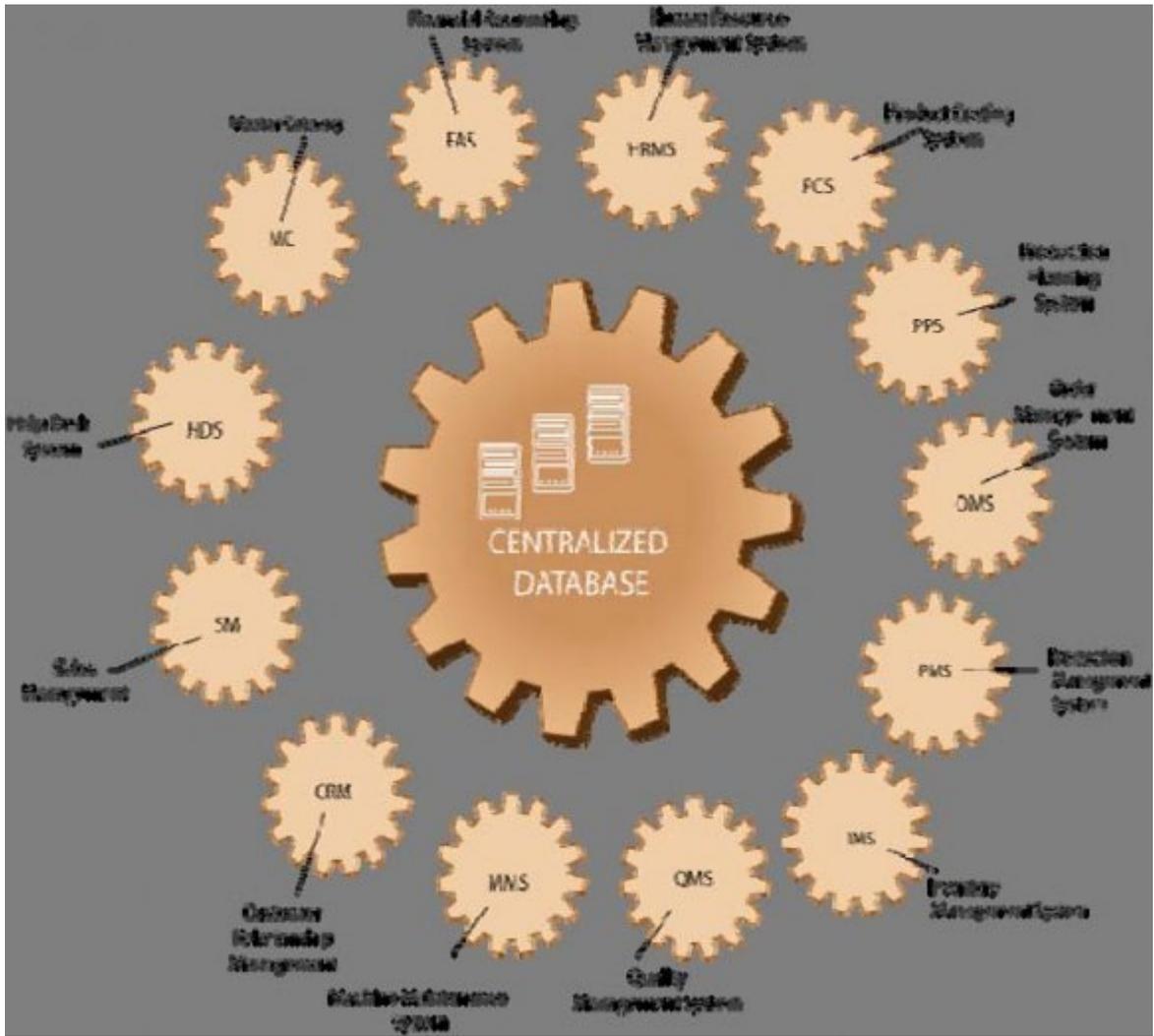
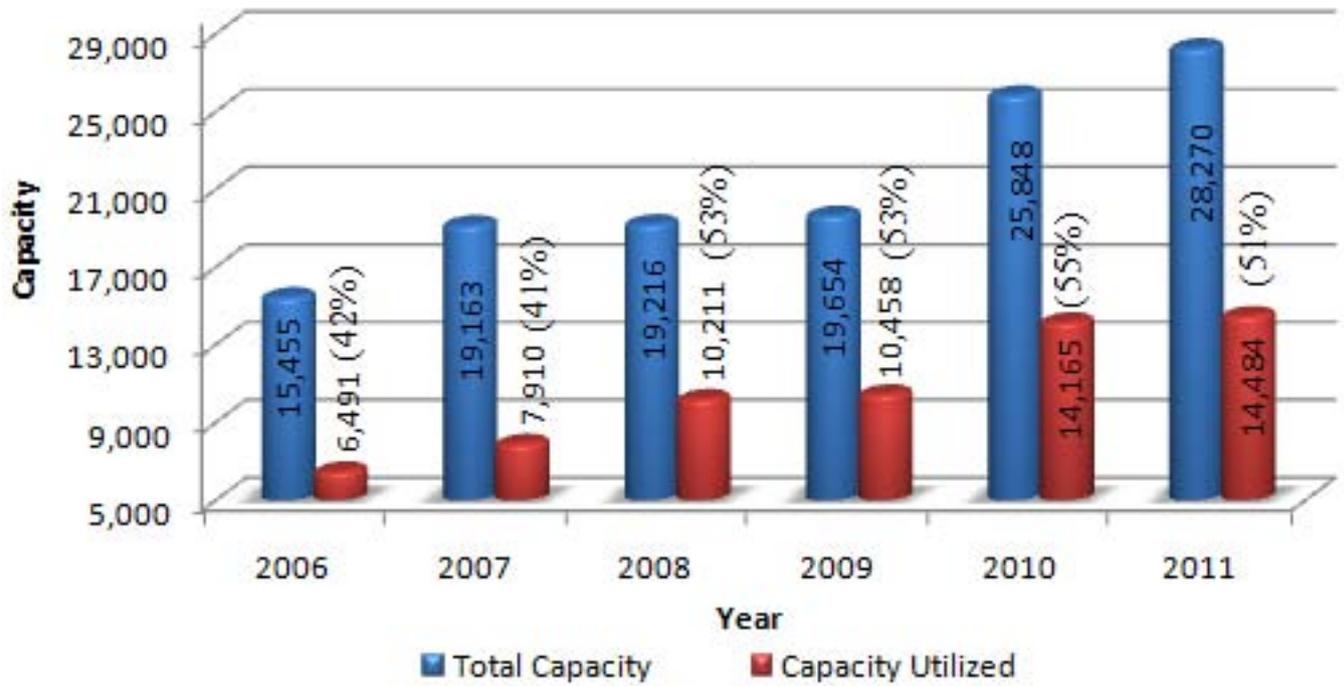
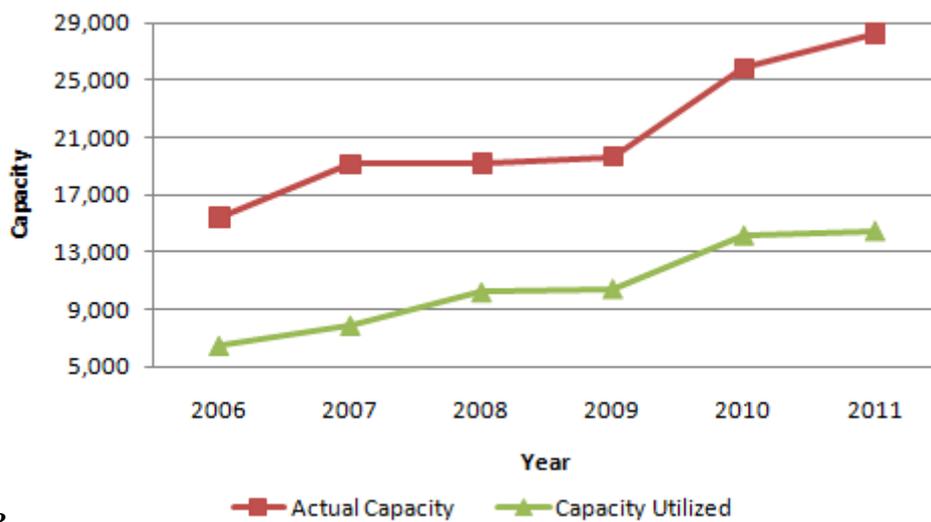


Figure 3:



2

Figure 4: Figure 2 :



3

Figure 5: Figure 3 :

Year	Actual Capacity	Capacity Utilized	% age Utilized
2006	15,455	6,491	42
2007	19,163	7,910	41
2008	19,216	10,211	53
2009	19,654	10,458	53
2010	25,848	14,165	55
2011	28,270	14,484	51

3

Figure 6: Figure 3 and

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Figure 7: Table 1 April

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[Note: knitting Production Capacity & Utilization]

Figure 8: Table 1 :

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- 115 [Jordan ()] *An introduction to usability, Taylor and Francis Ltd*, P W Jordan . 1998. p. .
- 116 [Chaudhuri and Dayal ()] ‘An overview of data warehousing and OLAP Technology’. S Chaudhuri , U Dayal .
- 117 *SIGMOD Record*, 1997. 26 p. .
- 118 [Connolly et al. ()] T Connolly , C Begg , A Strachan . *Database Systems: A Practical Approach to Design,*
- 119 *Implementation, and Management*, 1999. Addison-Wesley. 687. (3rd Ed)
- 120 [Hornbaek ()] ‘Current practice in measuring usability: Challenges to usability studies and research’. K Hornbaek
- 121 . *International Journal of Human-Computer Studies* 2006. 64 (2) p. .
- 122 [Data mining techniques and applications: Available] *Data mining techniques and applications: Available,*
- 123 [Zhaohui and Maclennan ()] *Data Mining with SQL Server*, Zhaohui , Maclennan . 2005. 2005. Wihely Publish-
- 124 ing, Inc.
- 125 [Benyon et al. ()] *Designing the interactive system*, D Benyon , Phil Tuner , Susan Tuner . 2005. p. .
- 126 [Herbert et al. ()] L Herbert , J Ragsdale , E Gaynor . *Put Business Applications to the Usability Test*, 2006.
- 127 Forrester Research.
- 128 [Preece ()] *Interaction design*, Preece . 2002. New York: jhon Wiley & sons. 230 p. .
- 129 [Schmettow and Vietz ()] ‘Introducing Item Response Theory for Measuring Usability Inspection Processes’. M
- 130 Schmettow , W Vietz . *Proceedings of The Twenty-Sixth Annual SIGCHI Conference on Human Factors*
- 131 *in Computing Systems*, (The Twenty-Sixth Annual SIGCHI Conference on Human Factors in Computing
- 132 SystemsFlorence, Italy) 2008. ACM Press. p. .
- 133 [El-Halees ()] *Mining Students Data to Analyze e-Learning Behavior: A Case Study*, Alaa El-Halees . 2009.
- 134 [Davis ()] ‘Perceived usefulness, perceived ease of use, and user acceptance of information technology’. F D Davis
- 135 . *MIS Quarterly* 1989. 13 p. .
- 136 [Sap ()] Sap . http://www.sapdesignguild.org/resources/user_day_toolkit/index.htm *User*
- 137 *Day Toolkit*, 2004. (Accessed on March)
- 138 [Krug and Don ()] *t Make Me Think: A Common Sens Approach to Web Usability*, S Krug , Don . 2000.
- 139 Indianapolis, IN: New Riders Publishing.
- 140 [Han and Kamber (ed.) ()] *The Morgan Kaufmann Series in Data Management Systems*, J Han , M Kamber .
- 141 Jim Gray, Series Editor (ed.) 2006. (Data Mining: Concepts and Techniques)
- 142 [Calisir and Calisir ()] ‘The relation of interface usability characteristics, perceived usefulness, and perceived
- 143 ease of use to end-user satisfaction with enterprise resource planning (ERP) systems’. F Calisir , F Calisir .
- 144 *Computers in Behavior* 2004. p. 20.
- 145 [Jokela and Iivari ()] ‘The Standard of User-Centered Design and the Standard Definition of Usability: Analyzing
- 146 ISO 13407 against ISO9241-11’. T Jokela , N Iivari . *Proceedings of the Latin American conference on Human-*
- 147 *computer interaction* 2003. 46 p. .
- 148 [Shackel ()] ‘Usability -Context, Framework, Definition, Design and Evaluation’. B Shackel . *Human Factors for*
- 149 *Informatics Usability*, B Shackel, S Richardson (ed.) (Cambridge) 1991. Cambridge University Press. p. .
- 150 [Nielsen ()] *Usability Engineering*, J Nielsen . 1993. Academic Press. p. .
- 151 [Faulkner ()] *Usability Engineering*, X Faulkner . 2000. Macmillan press Ltd. p. 7.
- 152 [Shakeel ()] ‘Usability Evaluation of In-Housed Developed ERP System’. Muhammad Shakeel , Faridi
- 153 . 10.1117/12.91321221. doi: 10. 1145/1400214. 1400239. [http://spiedigitallibrary.org/](http://spiedigitallibrary.org/proceedings/resource/2/psisdg/8285/1/82850C_1?isAuthorized=no)
- 154 [proceedings/resource/2/psisdg/8285/1/82850C_1?isAuthorized=no](http://spiedigitallibrary.org/proceedings/resource/2/psisdg/8285/1/82850C_1?isAuthorized=no) *Proc. SPIE*, 2011. 8285 p.
- 155 82850C.
- 156 [Hollingsed and Novick ()] ‘Usability inspection methods after 15 years of research and practice’. T Hollingsed ,
- 157 D G Novick . *Proceedings of The 25th Annual ACM international Conference on Design of Communication,*
- 158 (The 25th Annual ACM international Conference on Design of CommunicationEl Paso, Texas, USA) 2007.
- 159 ACM Press. p. .
- 160 [Hohmann ()] ‘Usability: Happier users mean greater profits’. L Hohmann . *Information Systems Management*
- 161 2003. 20 p. .