COMPREHENSIVE STUDY OF METHODS TECHNIQUES AND TOOLS USED FOR DATA ANALYTICS IN FIELDS OF FINANCE

by

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Table of Contents

Chapter	Title	Page	
	Title Page Table of Contents List of Figures	i ii iii	
	List of Tables	iv	
1	Introduction	1	
	1.1 Overview1.2 Why Analytics for financial institutions?1.3 State of analytics in financial companies	1 4 5	
	1.4 Objectives	6	
2	Descriptive Techniques	7	
	2.1 Social media analytics2.2 Log analytics2.3 Text analytics2.4 Location analytics	8 8 8 8	
3	Predictive Techniques	9	
	 3.1 Geo fencing 3.2 Artificial intelligence 3.3 Deep learning 3.4 Voice analytics 3.5 Image recognition 3.6 Video analytics 	9 9 10 10 10	
4	Prescriptive Techniques	11	
5	Optimization Techniques		
6	Visualization Techniques		
7	Conclusion and Recommendations	14	
	7.1 Conclusion7.2 Recommendations	14 14	
	References	15	
Referer	nces	15	

List of Figures

Figure	Title	Page
1.1	Data Analytics	3
1.2	Reprinted from the Morgan Stanley Digitization Index ranks	4
1.3	Da capablities	5

List of Tables

Table	Title	Page
1.1	Data mining use cases	2

Introduction

1.1 Overview

Data analysis is an very robust topic in the field of data science and encompasses the various mathematical functions. The functions are statistical in nature and are performed on the data obtained. The goal of data analytics is to support (or reject) the hypothesis which the data scientist postulates. By processing a steady stream of real-time data, organizations can make time-sensitive decisions faster than ever before, monitor emerging trends, course-correct rapidly and jump on new business opportunities. [BIG Data Analytics: A Framework for Unstructured Data Analysis pdf]

This paper tries to enlist most of the up-to-date techniques used by researchers and mathematicians to make sense of the data. Also the paper presents them in three groups of analytics. But then there arises a question which is, why the need for data analytics? Well, to answer that the paper proposes the literature from an article of Kdnuggets (Balm, 2015).

According to Paul following are use cases of data analytics: 1. Analytics powers our decisions we do not need to guess while making bold new decisions, we should use the information from data at hand. 2. Your data analysis weighs down your opponent's argument. 3. Cut down on loss making ventures with data analytics. 4. Can be applied to all domains be it health-care, banking, marketing, sales, operations etc. On the scenario when describing Data analytics it is very important to put the focus on Hypothesis.

Shown in the table 1.1 is some business areas

Application area	Applications	Specifics
Insurance	Fraud detection	Identify claims meriting investigation
Telecom	Churn	Identify likely customer turnover
Telemarketing	On-line information	Aid telemarketers with easy data access
Human resource man-	Churn	Identify potential employee turnover
agement		
Retail	Affinity positioning	Position product effectively
Retail	Cross-selling	Find more products for customers
Dankina	Customer relationship	Identify customer value
Banking	management	
		Develop programs to maximize revenue
Credit card manage-	Lift	Identify effective market segments
ment		
	Churn	Identify likely customer turnover

 Table 1.1: Data mining use cases

Figure 1.1 is a view of the data analytics with respect to the main fields.

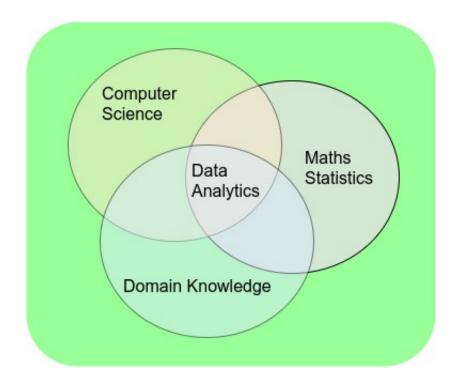


Figure 1.1: Data Analytics

1.2 Why Analytics for financial institutions?

According to a joint research study, by Boston Consulting Group and Morgan Stanley with analytics professionals, it was reveled that the financial institutions lagged behind other verticals in the use of data analytics (Malhotra et al., 2017) and it is shown in figure 1.2. One of the findings of the research was that FI's are investing a lot of capital, an estimated total of about \$1bn. In addition it was found that for near term value creation the FI's expected data analytics to optimize customer acquisition, customer retention, operational efficiency, and risk mitigation.

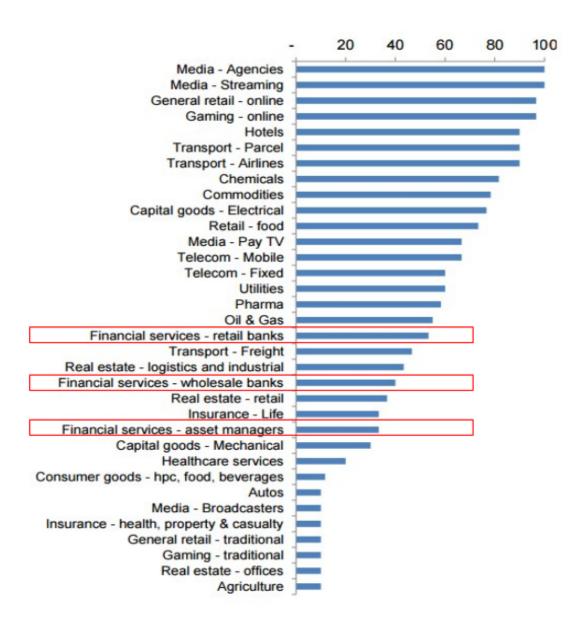


Figure 1.2: Reprinted from the Morgan Stanley Digitization Index ranks

1.3 State of analytics in financial companies

From the survey of FI's, a mix of interviewees representing payment companies, service providers, insurance, commercial banks, BCG made some interesting discoveries. It was found that most organizations have invested in analytics techniques to generate market perceptions. Some of the most used were those of social media, log, text and location analytics as shown in figure 1.3.

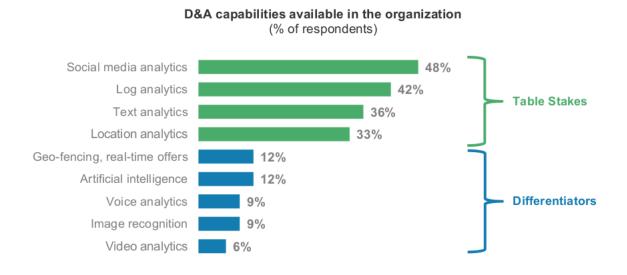


Figure 1.3: Reprinted from the work of Boston consulting group

Researchers have highlighted that most of the interviewees claimed that institutions are yet to make substantial gain from investments in analytics. They identified that companies can make gains from automation and digitization of manual processes. Additionally, it was noted that financial institutions are adopting digital processes to automate data collection for KYC (Know Your Customer) service and Anti-money laundering Customers are loyal consumers of products, which give value for their monetary investments. Big institutions tend to over look the need for value creation and focus their analytics and metrics to improve profit (Reichheld, 1996a). In the HBR report Reichheld makes certain observations that customers of large companies witness degrading value standards and secondly increasing rate of customer churn is a good variable to predict cash flow from consumer to company. He also noted that companies can replace old customer with new ones but the profits are lower as cost of induction is high.

1.4 Objectives

The overall objective of the study report is to understand the major methods, techniques and tools adopted by financial institutions to perform data analytics.

Descriptive Techniques

The sole purpose of using the methods in this category would be to describe the Hypothesis. Descriptive analytics or data mining are at the bottom of the big data value chain, but they can be valuable for uncovering patterns that offer insight. A simple example of descriptive analytics would be assessing credit risk; using past financial performance to predict a customers likely financial performance. Descriptive analytics can be useful in the sales cycle, for example, to categorize customers by their likely product preferences and sales cycle.

To summarize data into meaningful charts and reports, for example, about budgets, sales, revenues, or cost. They allow managers to obtain standard and customized reports, and drill down into the data and to make queries to understand the impact of an advertising campaign, for example, review business performance to find problems or areas of opportunity, and identify patterns and trends in data. Typical questions that descriptive analytics help answer are: How much did we sell in each region? What was our revenue and profit last quarter? How many and what types of complaints did we resolve? Which factory has the lowest productivity? Descriptive analytics also help companies to classify customers into different segments, which enable them to develop specific marketing campaigns and advertising strategies. [1]

There are two main approaches to apply in this topic Data warehousing and Visual analytics with reporting.

Descriptive analytics is very common and basic form of data mining technique to derive meaning and trend from data. Almost all of the companies in financial sector utilize the tools based on these techniques. The techniques in this are relatively basic compared with those of predictive models and prescriptive models. Institutions generally use tools which can generate reports regularly, some are printed monthly while some others are generated as a yearly routine. Such demands for reports describe the status quo of the balance sheets, organizational debts and cash flows. Daly transactions and updates of balances are maintained in the books or accounts and reports are generated daily for accountants and auditors. Thus it is imperative that vital functions of reporting, updating and auditing are functionalities of some business analytics.

There are several techniques employed by businesses to analyze and display data. They are as follows:

- 1. OLAP
- 2. Datawarehouse
- 3. Graphical views
- 4. Performance dashboard and KPI's

- 2.1 Social media analytics
- 2.2 Log analytics
- 2.3 Text analytics
- 2.4 Location analytics

Predictive Techniques

Predictive analytics use big data to identify past patterns to predict the future. For example, some companies are using predictive analytics for sales lead scoring. Some companies have gone one step further use predictive analytics for the entire sales process, analyzing lead source, number of communications, types of communications, social media, documents, CRM data, etc. Properly tuned predictive analytics can be used to support sales, marketing, or for other types of complex forecasts.

The basis of Predictive modeling is the use of data mining techniques to forecast future results. Data mining is the process of sorting data to find patterns or infer relationships. Thus a formulation of a statistical model with relevant variables is essential for prediction. ref from site (http://searchdatamanagement.techtarget modeling) and (http://searchsqlserver.techtarget.com/definition/data-mining)

Data mining techniques are used in credit card systems to detect fraud, in loan approval systems, identifying customer types and targeting specific promotional schemes. It is used to model customer behavior and detect churning to certain financial products. Insurance schemes and bank deposits are volatile instruments due to the uncertainty of the free economy. site (https://hbr.org/1996/03/learning-from-customer-defections).

There are many techniques to modeling predictive analytics

- 1.
- 2. Regession methods
- 3.
- 4.

e)

3.1 Geo fencing

3.2 Artificial intelligence

AI was developed a long time ago, but it has for recent years gained much traction and entered our lives. AI was already present in According to an article by PwC (Reichheld, 1996b)

- 3.3 Deep learning
- 3.4 Voice analytics
- 3.5 Image recognition
- 3.6 Video analytics

Prescriptive Techniques

Prescriptive analytics is really valuable, but largely not used. According to Gartner, 13 percent of organizations are using predictive but only 3 percent are using prescriptive analytics. Where big data analytics in general sheds light on a subject, prescriptive analytics gives you a laser-like focus to answer specific questions. For example, in the healthcare industry, you can better manage the patient population by using prescriptive analytics to measure the number of patients who are clinically obese, then add filters for factors like diabetes and LDL cholesterol levels to determine where to focus treatment. The same prescriptive model can be applied to almost any industry target group or problem.

Optimization Techniques

Visualization Techniques

Conclusion and Recommendations

This section presents the conclusion of study report.

7.1 Conclusion

In this paper we have tried to present a picture of the methods and tools of data analytics. They are summed up under three categories. Our motivation was to study the various methods and practices used by researchers and businesses. We hope this paper encourage new companies to apply softwares and schemes to analyze data. The analytic results could give them insights about To achieve what was not perceived, to deduce what was not understood is all made possible with data analytics. Thus it is our effort that the topics discussed in this paper would be of use to those who step into the world of Analytics without knowledge of background. References:

http://www.cbpp.uaa.alaska.edu/afef/business_analytics.htm

7.2 Recommendations

Text..

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