Trend and Prediction Analysis for Enhanced Road Safety

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CERTIFCATE OF APPROVAL

It is certified that the contents and form of Final Report entitled **"Trend and Prediction Analysis for Enhanced Road Safety"** submitted by *Sumaira Shamim* (2009-NUST-BE-BICSE-212), *Hamza Shahzad* (2009-NUST-BE-BICSE-153) and *Muneza Naeem* (2009-NUST-BE-BICSE-198) has been found satisfactory for the requirement of the degree.

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DEDICATION

To Allah Almighty

&

To our Parents and Faculty

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List of Abbreviations:

- > NH & MP: National Highways and Motorways Police
- ➢ ITP: Islamabad Traffic Police
- BI: Business Intelligence
- CSV: Comma Separated Variables
- > WHO: World Health Organization
- OLAP: Online Analytical Processing

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ABSTRACT

Most organizations are realizing the value of data; and Business Intelligence (BI) is the key to analyze data in today's information world. BI is used to investigate real world scenarios to identify potential problems. One such scenario is the ever increasing rate of auto-mobile accidents in Pakistan. Most of the countries are gaining benefits by analyzing the past road accidents data and forecasting a lot of information about possible future accidents. But unfortunately in Pakistan, data is collected to a limited extent, and the standard format of collecting data is not truly followed. There is no central department for accident prevention neither is there any data repository. Hence the number of accidents is steadily on the rise.

The aim of this project is to improve road safety and reduce accidents by identifying the root causes of these accidents. In order to achieve this goal, we have analyzed the traffic accidents data of Pakistan's Motor-ways and National highways. The data was collected from National Highways and Motorway Police, cleaned and modeled on different dimensions for analysis and prediction. The major recommendations concluded from the results are as follows.

- Sunder (near Lahore) had the most number of accidents due to non-availability of pedestrian overhead bridge; and over speeding of drivers. A solution is to install overhead bridge and to have police patrol the area round the clock to keep drivers in check.
- 2. Trucks and trollies are involved in the most *dozing at the wheel* accidents; hence we need to plant road bumps on locations such as Sahiwal, where these accidents are more prominent.
- 3. Motorcycle riders have the most frequent fatal accidents among all vehicle types; the reason being disregard of safety by not wearing the helmet. A solution is to impose heavy fines on driving without helmet.

Chapter 1

INTRODUCTION

This chapter gives an overview of our project and explains the motivation, scope, problem statement and project domain.

Defining the Domain: Business Intelligence

The domain of the project is Data warehousing and Business intelligence.

"Data warehouse is a system that retrieves and consolidates data periodically from the source systems into a dimensional or normalized data store. It usually keeps years of history and is queried for business intelligence or other analytical activities". [1]

"Business intelligence (BI) is a set of theories, methodologies, processes, architectures, and technologies that transform raw data into meaningful and useful information for business purposes. BI can handle large amounts of information to help identify and develop new opportunities. Making use of new opportunities and implementing an effective strategy can provide a competitive market advantage and long-term stability". [2]

Data warehousing helps us to provide information on the techniques involved in designing, building, maintaining and retrieving information, from a data warehouse. [3] [4]

Business Intelligence is a method that extracts useful and meaningful information from the raw data which is used to enable more effective strategic, calculated, and operational insights and assist in decision-making. [2] [5]

Often data warehouse or data mart are basis to extract information using BI. The aim is to define and specify useful management reports from warehouse data, which will be used to make fast, accurate and intelligent business decisions. [5]

Problem Statement:

Traffic Accidents is the cause of a considerable number of casualities and injuries to such an extent that now the issue of Road security is being raised internationally. Accidents are being recorded according to the World Health Organizatin (WHO) standard internationally, which unluckily cannot be implemented in Pakistan as recording of accidents data is not followed comprehensively. So the need for devising intelligent system for road security by prediction analysis of traffic accidents in Pakistan has become more prevalent.

Thus, the problems statement of the project is:

"Reduce the severity i.e. fatalities caused by traffic accidents by analyzing the trends of auto-accidents in Pakistan".

Project Motivation

• BI and Dataware housing – A technology in Trend:

With the increasing data, there is a greater trend in making data warehouses and BI using them for especially in the area of predictive analytics, in which data is analyzed to predict future trends.

• Traffic status in Pakistan:

In Pakistan traffic is increasing annually because the public transport available to general public is grossly insufficient. Hence the only outcome is for people to buy their own vehicles, which is leading to an alarming increase in traffic. Hence the need to better control traffic. • Lack of any concrete work on Road safety in Pakistan:

There has been some work done in individual capacity before to address the issue of road safety in Pakistan but nothing concrete has been achieved. Hence the motivation for us to explore this field and possibly come up with a solution that will benefit Pakistan and its people.

Project Description

The project focuses on the trend and prediction analysis of traffic accidents in Pakistan for which Traffic Accidents Data is collected from National Highway and Motorways Authority.

Central database of accidents is created using IBM DB2, which is acting as the primary data source for metadata modeling and reporting. Prior to creating the database, data cleaning was performed using Google Refine.

Data Modeling and Reporting was performed using IBM Cognos Framework Manager and IBM Cognos Report Studio. The aim is to provide better strategy making for traffic management. The analysis will provide for information regarding trends in traffic accidents and aid in reducing the severity and occurrences of auto-accidents in Pakistan.

Areas that are identified for having higher number of accidents are ranked as red zones. Those areas are marked on Pakistan Map using R – Language.

Further, using the same accidents data from National Highways and Motorways, a prediction model is made using IBM SPSS to predict the severity of accident cases on National Highways and Motorways.

Aims and Objectives

Overall project has been divided into several milestones over the timeline, which are listed below:

• Trend Analysis:

Identify the trends of traffic accidents on National Highways and Motorways.

• Predictions Analysis:

Predicting the severity of accidents and ranking them as fatal and non-fatal.

• Efficient access of information through dashboards:

Customized dashboards created using the reported trends provide ease of access of information.

• Mapping of locations of accidents on maps:

Zones having maximum number of accidents are identified so that steps can be taken for enhanced road security.

Project Plan

.

The project has been divided into different phases. The phases are incremental and sub-tasks can be dependent on each other. There are comprehensive requirements of setting up servers of IBM COGNOS BI and IBM SPSS that have been catered to meet and support the overall project plan

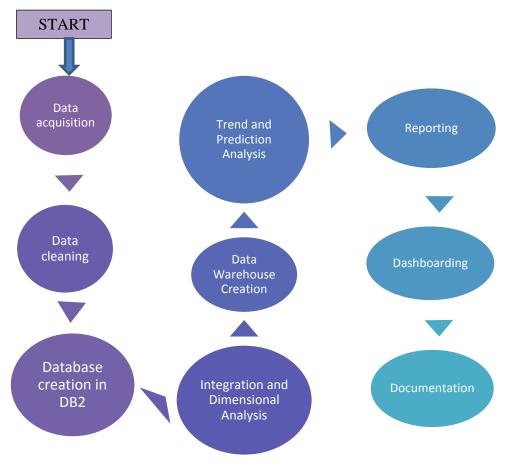


Figure 1: Work Flow of Project

Project Deliverables

Our project consists of the following deliverables:

- 1. Data acquisition.
- 2. Data Cleaning.
- 3. Integration and Dimensional Analysis.
- 4. Data Warehouse creation.
- 5. Dimensional Modeling
- 6. Trend and Prediction analysis.

- 7. Dash boarding.
- 8. Final Test case Reporting.
- 9. Documentation

Project Timeline

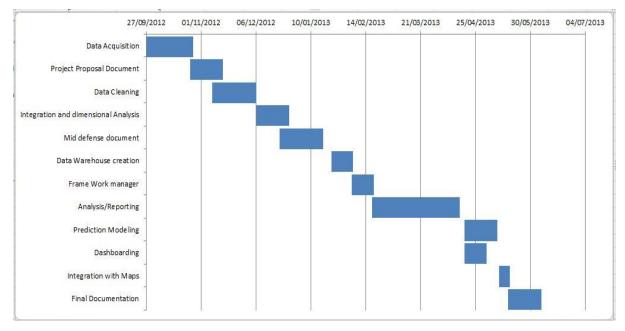


Table 1: Project Timeline

Figure 2: Gantt chart

Chapter 2

LITERATURE REVIEW

In this chapter the important topics and research related to this project will be highlighted.

Related Work:

This section contains review of the research papers, reports and projects related to Road safety.

a) Intelligent System for Road Safety:

This project was done in 2011-12, by Mr. Hamza Iftikhar and Mr. Abdul Ghaffar of BIT 10 SEECS NUST. The abstract of their project is as follows.

The problem of deaths and injuries as a result of road accidents is now acknowledged to be a Global phenomenon with authorities in virtually all countries of the world concerned about the growth in the number of people killed and seriously injured on their roads.

World Health Organization has defined a standard for recording the road accidents data. Most of the countries follow that standard for recording their causalities and injuries caused by the accidents. The data is stored in bulk form, so that the government or the concerned authority can analyze the data and take proper measures in the future. Most of the countries are doing the same thing and using that data to reduce future accidents. But unfortunately in Pakistan, that data is collected to limited extent, but the standard format of surveying/collecting the data is not followed. So, the data is of no use for anyone. [12]

b) Road Traffic Injury Research and Prevention Centre:

Ex-minister for health Dr. Jumma Khan started an awareness campaign in Karachi focused on road safety and accident prevention. He also collected data of accidents and related information like ambulance arrival time etc. His work in Karachi has been paramount in reducing the accidents there and creating awareness among people.¹

Business Intelligence: Introduction and Benefits

Business intelligence is an assortment of applications that analyzes numerous aspects of data and facilitates decision making. It is a host of modules such as projecting, Online Analytical Processing (OLAP), predictive modeling, data management, data mining and optimization [6]. Using these tools, corporations can accurately judge trends in their organization and businesses and identify future trends to maximize profits. The various benefits of BI depend on the organization, but a number of general benefits are [7]

- Find out useful relationships between data events
- Provide visual representations of hypothetical scenarios, helping companies to see which option works best
- Meet or exceed customer expectations by using factual information
- Use multiple sources of data for strategic decisions
- Efficient distribution of statistics
- Central accessibility to departments and end users

¹ <u>www.roadsafety.pk</u>

Data Mining

The process of identifying and detecting patterns in the data is called data mining. Data mining techniques aim to provide insight that allows for a better understanding of data and its essential features. Companies and organizations can employ two basic types of data mining: [4]

• Validation of Hypothesis

It is used to validate an idea or a hypothesis about a relation between data.

• Knowledge Discovery:

It is used to find hidden and unknown relations to get statistical results which may be significant for the data elements.

Different tasks in data mining are

- Classification: A model classifies samples to a particular classified group
- **Clustering:** Groups samples that have common characteristics
- Association: Finds relations between events at a given time
- Sequencing: Similar to association. But relations are sorted according to time period
- **Regression:** Linear and nonlinear techniques to display information according to predictive value
- Forecasting: Prediction of future values based on history

Predictive Analysis

The success of an organization depends on the capability to analyze the future possibilities and understand the trends, so that correct action can be taken at the right time. Predictive analytics play an essential role to help and plan future actions according to risks and seize opportunities. The result of predictive analysis is used to identify patterns and trends and accuracy of the analysis depends on the complexity of interdependency of data [8]. Some predictive analysis methods are

• Linear regression

- Binomial Logistic regression
- Multinomial Logistic regression
- Decision trees
- Probity regression
- Neural networks

Data Analysis Tools:

There are many different tools available for data warehousing and Business Intelligence. Some of them are

- For ETL: Informatica and Kettle are available
- For reporting/analytics: Jasper, Pentaho and Cognos are available
- For Benchmarking: Informatic, Kettle and Cognos
- For enterprise Management: BO, Cognos and opensource tools

IBM Cognos for trend analysis:

It Provides services such as financial performace, strategy management, analytics application. [9]

IBM Cognos offers the following tools:

• IBM Cognos Metric Designer:

It is used to create extracts for use in IBM Cognos scorecarding applications. The created extracts are used to map and transfer information from existing metadata sources.

• IBM Cognos Query Studio:

Reporting tool for creating simple queries and reports in IBM Cognos Business Intelligence.

• IBM Cognos Framework Manager:

It is a metadata model development environment i.e. it assists in creating and managing business related metadata for use in BI analysis.

• IBM Cognos Connection:

It is a portal to IBM Cognos software and provides a single acces point to all corporate data available in IBM Cognos software.

• IBM Cognos Report Studio:

It is a report authoring tool used to create sophisticated and managed reports in short it is a full bloown report generatin tool.

• IBM Cognos Analysis Studio:

Wih analysis studio, it is possible to see trends and understand strange condition or situation or variances that may not be evident with other types of reporting.

• IBM Cognos Business Insight:

It is a web-based tool that allows you to use IBM Cognos content and external data sources like HTML/text sources to build interactive dashboards for better decision making.

• IBM Cognos for Microsoft Office:

It can be used to work with secure IBM Cognos BI content in Microsoft Office environment.

IBM SPSS for predictive analysis:

Given a dataset, many types of data analysis can be done. IBM SPSS is a powerful statistical tool that aids in building of accurate predictive models intuitively [10]. It enables the user to:

- Discover patterns and trends in data more easily using a visual interface and supported by advanced analytical techniques.
- The outcome of the model shows the influence of various factors on the considered problem. This helps the user to make better statistical decisions.

R-Language for statistical computing:

R is a free open source software programming language and a software environment for statistical computing and graphics. The R language is widely used in statistics and data mining for developing statistical software and data analysis. Polls and surveys of data miners are showing R's popularity has increased substantially in recent years. [11]

R language has a library called RgoogleMaps that can be used to do integration of your data set to Google maps using R.

Chapter 3

METHODOLGY AND IMPLEMENTATION

This chapter explains the detailed structure of the proposed solution along with the brief enlightenment.

Proposed Methodology

The methodology consists of the following mentioned points with their definitions.

• Data collection

The most critical part of any project in the field of BI is the collection of authentic data for reporting. It is often the first stage, due to the fact that data is available in different formats and it is necessary to make a standard data set.

• Data cleaning

It is possible that the gathered data is not up to the mark and contains some inconsistencies like duplication or null values, therefore data cleaning is an important task to make the data consistent.

• Database Creation

This is the first proper stage of data warehousing. Database needs to be created so that the gathered data may be unified and arranged in a central location to offer ease of access.

• Metadata Modeling

It is also important as this provides the crucial link between the raw data and the highly organized reports; it is a sample of data on which you model your design to help in the reporting process.

• Reporting and Analysis

The most crucial component is BI is the reporting and analysis because in the end that is why business intelligence exists so that you may be able to make sense of data and answer important business questions, reporting and analysis is the key to that. • Prediction Analysis

The next stage in BI is to predict what happens next or what if this scenario occurs what then, therefore prediction is the usage of statistical models for successful answers of what if questions.

• Location Mapping

To make BI more appealing visually to the customer mapping is used so that you get an easy to understand interface of your business.

Methodology Implementation

As we discussed in the proposed methodology the different concepts and stages of BI, let us now explain how they are implemented in our project.

Now, all the steps will be explained in detail.

Data collection:

It is a really hectic process as data collection of accidents in Pakistan is not up to the international standards on which effective reporting is done and further recommendations are made. We visited the following offices for data acquisition.

- National Highways & Motorway Police: Meeting with Mr. Farooq Azam (AIG-R&D)
- Rescue 15: Meeting with SP Mr. Sajid Kiyani
- Islamabad Traffic Police: Meeting with SSP Mr. Moin who provided us with statistical data.
- Road Traffic Injury Research and Prevention Centre Karachi: contacted Dr. Jooma Khan
- Kashmir: SP Mr. Sajid Kiyani from RESCUE 15 assisted us in getting statistical data from Kashmir.
- 1122: Meeting with Mr. Tauqir Khan (Assistant Director Cares)

We were able get data from National Highways & Motorway Police with the assistance from Mr. Farooq Azam, who was kind enough to lend us the data from end-

2008 to mid-2012 with the hope to improve the Traffic System of NH & MP with our road safety recommendations.

We acquired Accidents data in the form of excel sheets.

1- Accident Wise

2- Vehicle Wise

Accident Wise:

No of accident data entries = 2044

Column Names:

| Column Names | Column Names | Column Names |
|---------------------|-------------------------|-----------------------|
| CaseNo_ID | Accident Year | Zone Value |
| Sector Value | No Passenger Killed | No Passenger Injured |
| Beat Value | No Driver Injured | No Pedestrian Killed |
| Accident Month | No Driver Killed | No Pedestrian Injured |
| Place Name Value | Accident Severity Value | Collision Type Value |
| Accident Time | Travel Direction | Weather Value |
| Accident Date | Kilometer Marker | Light Condition Value |
| Road Geometry Value | Accident Cause Value | Accident Day |
| Network Value | Motorway Name Value | |

Table 2: Accident Wise Column names

Following are screenshots of Accident Wise data:

| 1 | CaseNo_ID 💌 | AccidentYear 💌 | ZoneValue 💌 | SectorValue 💌 | BeatValue 💌 | AccidentDate 💌 | AccidentTime 💌 | AccidentMonth 💌 |
|----|-------------|----------------|-------------|---------------|-------------|----------------|----------------|-----------------|
| 2 | 62 | 2008 | MOTORWAY | M-1 | Beat 1 | 7/6/2008 | 1/1/1900 1:32 | July |
| 3 | 63 | 2008 | MOTORWAY | M-1 | Beat 4 | 7/14/2008 | 1/1/1900 23:25 | July |
| 4 | 64 | 2008 | MOTORWAY | M-1 | Beat 2 | 7/17/2008 | 1/1/1900 15:55 | July |
| 5 | 65 | 2008 | MOTORWAY | M-1 | Beat 1 | 7/24/2008 | 1/1/1900 7:45 | July |
| 6 | 66 | 2008 | MOTORWAY | M-1 | Beat 1 | 7/26/2008 | 1/1/1900 20:04 | July |
| 7 | 67 | 2008 | MOTORWAY | M-1 | Beat 1 | 7/14/2008 | 1/1/1900 12:24 | July |
| 8 | 68 | 2008 | MOTORWAY | M-2(S) | Beat 10 | 7/1/2008 | 1/1/1900 1:52 | July |
| 9 | 69 | 2008 | MOTORWAY | M-2(S) | Beat 11 | 7/3/2008 | 1/1/1900 12:14 | July |
| 10 | 70 | 2009 | MOTORWAY | M-1 | Beat 2 | 1/1/2009 | 1/1/1900 17:27 | January |
| 11 | 71 | 2009 | MOTORWAY | M-1 | Beat 3 | 1/5/2009 | 1/1/1900 13:04 | January |
| 12 | 73 | 2009 | MOTORWAY | M-1 | Beat 4 | 1/18/2009 | 1/1/1900 20:06 | January |
| 13 | 74 | 2009 | MOTORWAY | M-1 | Beat 1 | 1/26/2009 | 1/1/1900 9:56 | January |
| 14 | 75 | 2009 | MOTORWAY | M-1 | Beat 4 | 1/11/2009 | 1/1/1900 14:45 | January |
| 15 | 76 | 2009 | MOTORWAY | M-2(N) | Beat 8 | 1/17/2009 | 1/1/1900 22:54 | January |
| 16 | 77 | 2009 | MOTORWAY | M-2(S) | Beat 12 | 1/10/2009 | 1/1/1900 8:33 | January |
| 17 | 78 | 2009 | MOTORWAY | M-2(S) | Beat 11 | 1/21/2009 | 1/1/1900 21:07 | January |
| 18 | 80 | 2009 | North | North-I | Beat 1 | 1/15/2009 | 1/1/1900 11:45 | January |
| 19 | 81 | 2009 | North | North-I | Beat 4 | 1/25/2009 | 1/1/1900 8:30 | January |
| 20 | 82 | 2009 | North | North-II | Beat 5 | 1/16/2009 | 1/1/1900 16:55 | January |
| 21 | 83 | 2009 | North | North-II | Beat 5 | 1/11/2009 | 1/1/1900 17:00 | January |
| 22 | 84 | 2009 | North | North-II | Beat 5 | 1/12/2009 | 1/1/1900 12:38 | January |
| 23 | 85 | 2009 | North | North-II | Beat 5 | 1/26/2009 | 1/1/1900 12:15 | January |
| 24 | 86 | 2009 | North | North-III | Beat 11 | 1/23/2009 | 1/1/1900 3:12 | January |

Figure 3: Data

| 1 | AccidentMonth 💌 | PlaceNameValue 💌 | KilometerMarker 💌 | TravelDirection 💌 | AccidentSeverityValue |
|----|-----------------|------------------|-------------------|-------------------|-----------------------|
| 2 | July | Peshawar | 9 | Bravo | Minor |
| 3 | July | Islamabad | 1 | Alpha | Major |
| 4 | July | Peshawar | 75 | Alpha | Major |
| 5 | July | Rashakai | 6 | Alpha | Major |
| 6 | July | Rashakai | 6 | Alpha | Minor |
| 7 | July | Peshawar | 18 | Bravo | Fatal |
| 8 | July | Islamabad | 1 | Bravo | Major |
| 9 | July | Islamabad | 63 | Alpha | Minor |
| 10 | January | Rashakai | 49 | Bravo | Major |
| 11 | January | 99 (Beat 03) | 99 | Bravo | Minor |
| 12 | January | Burhan | 361 | Alpha | Minor |
| 13 | January | Peshawar | 34 | Alpha | Major |
| 14 | January | Burhan | 359 | Bravo | Major |
| 15 | January | Burhan | 195 | Bravo | Fatal |
| 16 | January | ATTOCK | 320 | Bravo | Minor |
| 17 | January | Khankadogrra | 31 | Alpha | Major |
| 18 | January | Burhan | 18 | Bravo | Fatal |
| 19 | January | TAXILA | 1 | Alpha | Fatal |
| 20 | January | Lahore | 25 | Alpha | Fatal |
| 21 | January | Channi Bridge | 1 | Bravo | Major |
| 22 | January | Radio Pakistan | 1 | Alpha | Major |
| 23 | January | LRBT Hospital | 1 | Alpha | Fatal |
| 24 | January | Lahore | 25 | Alpha | Major |
| 25 | January | CITY KAMONKE | 1 | Alpha | Fatal |

Figure 4: Data-2

Vehicle Wise:

No of accident data entries = 3471 (i.e. more vehicles are involved in one accident)

Column Names:

| Column Names | Column Names | Column Names |
|-----------------------|-------------------------|-----------------------|
| CaseNo_ID | Accident Year | Zone Value |
| Sector Value | No Passenger Killed | No Passenger Injured |
| Beat Value | No Driver Injured | No Pedestrian Killed |
| Accident Month | No Driver Killed | No Pedestrian Injured |
| Place Name Value | Accident Severity Value | Collision Type Value |
| Accident Time | Travel Direction | Weather Value |
| Accident Date | Kilometer Marker | Light Condition Value |
| Road Geometry Value | Accident Cause Value | Accident Day |
| Vehicle Company Value | Motorway Name Value | Vehicle Type Value |

Table 3: Vehicle Wise Column Names

Following are screenshots of Vehicle Wise data:

| 1 | CaseNo_ID 🚽 | AccidentYear 💌 | AccidentMonth 💌 | AccidentDate 💌 | AccidentDay 💌 | AccidentTime 💌 | AccidentSeverityValue | Trav |
|----|-------------|----------------|-----------------|----------------|---------------|----------------|-----------------------|------|
| 2 | 62 | 2008 | July | 7/6/2008 | Sunday | 1/1/1900 1:32 | Minor | Brav |
| 3 | 62 | 2008 | July | 7/6/2008 | Sunday | 1/1/1900 1:32 | Minor | Brav |
| 4 | 63 | 2008 | July | 7/14/2008 | Monday | 1/1/1900 23:25 | Major | Alpł |
| 5 | 63 | 2008 | July | 7/14/2008 | Monday | 1/1/1900 23:25 | Major | Alpł |
| 6 | 63 | 2008 | July | 7/14/2008 | Monday | 1/1/1900 23:25 | Major | Alpl |
| 7 | 63 | 2008 | July | 7/14/2008 | Monday | 1/1/1900 23:25 | Major | Alpl |
| 8 | 64 | 2008 | July | 7/17/2008 | Thursday | 1/1/1900 15:55 | Major | Alpł |
| 9 | 65 | 2008 | July | 7/24/2008 | Thursday | 1/1/1900 7:45 | Major | Alpł |
| 10 | 66 | 2008 | July | 7/26/2008 | Saturday | 1/1/1900 20:04 | Minor | Alpł |
| 11 | 67 | 2008 | July | 7/14/2008 | Monday | 1/1/1900 12:24 | Fatal | Brav |
| 12 | 68 | 2008 | July | 7/1/2008 | Tuesday | 1/1/1900 1:52 | Major | Brav |
| 13 | 68 | 2008 | July | 7/1/2008 | Tuesday | 1/1/1900 1:52 | Major | Brav |
| 14 | 68 | 2008 | July | 7/1/2008 | Tuesday | 1/1/1900 1:52 | Major | Brav |
| 15 | 69 | 2008 | July | 7/3/2008 | Thursday | 1/1/1900 12:14 | Minor | Alpł |
| 16 | 70 | 2009 | January | 1/1/2009 | Thursday | 1/1/1900 17:27 | Major | Brav |
| 17 | 71 | 2009 | January | 1/5/2009 | Monday | 1/1/1900 13:04 | Minor | Brav |
| 18 | 71 | 2009 | January | 1/5/2009 | Monday | 1/1/1900 13:04 | Minor | Brav |
| 19 | 73 | 2009 | January | 1/18/2009 | Sunday | 1/1/1900 20:06 | Minor | Alpł |
| 20 | 73 | 2009 | January | 1/18/2009 | Sunday | 1/1/1900 20:06 | Minor | Alpł |
| 21 | 73 | 2009 | January | 1/18/2009 | Sunday | 1/1/1900 20:06 | Minor | Alpł |
| 22 | 74 | 2009 | January | 1/26/2009 | Monday | 1/1/1900 9:56 | Major | Alpł |
| 23 | 75 | 2009 | January | 1/11/2009 | Sunday | 1/1/1900 14:45 | Major | Brav |
| 24 | 76 | 2009 | January | 1/17/2009 | Saturday | 1/1/1900 22:54 | Fatal | Brav |
| 25 | 76 | 2009 | January | 1/17/2009 | Saturday | 1/1/1900 22:54 | Fatal | Brav |
| 26 | 77 | | January | 1/10/2009 | Saturday | 1/1/1900 8:33 | | Brav |
| | | | | | - · · | | | _ |

Figure 5: Vehicle wise data

| 1 | TravelDirection 💌 | PlaceNameValue 💌 | BeatValue 💌 | SectorValue 💌 | ZoneValue 💌 | NoDriverKilled 💌 |
|----|-------------------|------------------|-------------|---------------|-------------|------------------|
| 2 | Bravo | Peshawar | Beat 1 | M-1 | MOTORWAY | (|
| 3 | Bravo | Peshawar | Beat 1 | M-1 | MOTORWAY | (|
| 4 | Alpha | Islamabad | Beat 4 | M-1 | MOTORWAY | (|
| 5 | Alpha | Islamabad | Beat 4 | M-1 | MOTORWAY | (|
| 6 | Alpha | Islamabad | Beat 4 | M-1 | MOTORWAY | (|
| 7 | Alpha | Islamabad | Beat 4 | M-1 | MOTORWAY | (|
| 8 | Alpha | Peshawar | Beat 2 | M-1 | MOTORWAY | (|
| 9 | Alpha | Rashakai | Beat 1 | M-1 | MOTORWAY | (|
| 10 | Alpha | Rashakai | Beat 1 | M-1 | MOTORWAY | (|
| 11 | Bravo | Peshawar | Beat 1 | M-1 | MOTORWAY | (|
| 12 | Bravo | Islamabad | Beat 10 | M-2(S) | MOTORWAY | (|
| 13 | Bravo | Islamabad | Beat 10 | M-2(S) | MOTORWAY | (|
| 14 | Bravo | Islamabad | Beat 10 | M-2(S) | MOTORWAY | (|
| 15 | Alpha | Islamabad | Beat 11 | M-2(S) | MOTORWAY | (|
| 16 | Bravo | Rashakai | Beat 2 | M-1 | MOTORWAY | (|
| 17 | Bravo | 99 (Beat 03) | Beat 3 | M-1 | MOTORWAY | (|
| 18 | Bravo | 99 (Beat 03) | Beat 3 | M-1 | MOTORWAY | (|
| 19 | Alpha | Burhan | Beat 4 | M-1 | MOTORWAY | (|
| 20 | Alpha | Burhan | Beat 4 | M-1 | MOTORWAY | (|
| 21 | Alpha | Burhan | Beat 4 | M-1 | MOTORWAY | (|
| 22 | Alpha | Peshawar | Beat 1 | M-1 | MOTORWAY | (|
| 23 | Bravo | Burhan | Beat 4 | M-1 | MOTORWAY | (|
| 24 | Bravo | Burhan | Beat 8 | M-2(N) | MOTORWAY | 1 |
| 25 | Bravo | Burhan | Beat 8 | M-2(N) | MOTORWAY | 1 |

Figure 6: Vehicle wise-2

Data cleaning:

Google Refine was used as a tool for data cleaning.

Data Cleaning in Google Refine included

> Finding/ Removing the inconsistencies in the data

An inconsistency was found in 'Accident Year' column where three rows had

the value of 1.

| Facet / Filter Undo / Redo o | | | 3 matching rows (2044 total) | | | | | | | | | |
|------------------------------|---------------------|----|--------------------------------------|------|------------|--------------|----|-----------|-------------|---------|--------------------------|--|
| resh Reset All Remove All | | Sh | Show as: rows records Show C to come | | | | | | | | | |
| ccidentYear | change invert reset | | All | | Casello_ID | AccidentYear | C | ZoneValue | SectorValue | 💌 BeatV | e 💌 AccidentDa | |
| noices Sort by: na | me count Cluster | 岔 | | 729. | 820 | 1 | Þ | | North-II | Beat 7 | 2010-02- 15T00:00:00Z | |
| 03 | | 슔 | | 730. | 821 | 1 | \$ | h | South-III | Beat 34 | 2010-01- 18T00:00:00Z | |
| 0 0 0.800 9 | exclude | ☆ | | 731. | 822 | 1 | S | h | South-III | Beat 34 | 2010-01- 28T00:00:00Z | |
| 09.0 687 | edit include | | | | | | | | | | _ | |
| 10.0 556 11.0 457 | edit include | | | | | | | | | | | |
| 12.0 332 | | | | | | | | | | | | |
| cet by choice counts | | | | | | | | | | | | |

Figure 7: Google Refine

After the data was closely observed, it was found form the 'Accident Date' column that these entries belong to 'Accident Year' 2010. And so this inconsistency was removed.

Google refine Data Accidents xlsx Permalink

| 0 | | | | | | | | | | | |
|--------------------------|---------------|------------|----|-------|-------|-------------|---------------|-----------|-------------|-----------|--------------------------|
| Facet / Filter | Undo / Redo 5 | | 20 | 44 | rov | VS | | | | | |
| Refresh | Reset All | Remove All | Sh | ow a: | s: ro | ows records | Show: 5 10 25 | 50 rows | | | |
| × AccidentYear | | change | | All | | CaseNo_ID | AccidentYear | ZoneValue | SectorValue | BeatValue | Accident |
| 5 choices Sort by: | name count | Cluster | | | 1. | 62 | 2008 | MOTORWAY | M-1 | Beat 1 | 2008-07- 06T00:00:00Z |
| 2009.0 687 2010.0 559 | | | ☆ | | 2. | 63 | 2008 | MOTORWAY | M-1 | Beat 4 | 2008-07- 14T00:00:00Z |
| 2011.0 457 2012.0 332 | | | | | 3. | 64 | 2008 | MOTORWAY | M-1 | Beat 2 | 2008-07- 17T00:00:00Z |
| 2008.0 9 | | | ☆ | | 4. | 65 | 2008 | MOTORWAY | M-1 | Beat 1 | 2008-07- 24T00:00:00Z |
| Facet by choice coun | ts | | | | 5. | 66 | 2008 | MOTORWAY | M-1 | Beat 1 | 2008-07- 26T00:00:00Z |
| | | | ☆ | | 6. | 67 | 2008 | MOTORWAY | M-1 | Beat 1 | 2008-07- 14T00:00:00Z |
| | | | 岔 | | 7. | 68 | 2008 | MOTORWAY | M-2(S) | Beat 10 | 2008-07- 01T00:00:00Z |
| | _ | | A. | | ۰ | 20 | 2000 | HOTODWAY | H 9/61 | Deat 11 | 2000 07 |

Figure 8: Google Refine-2

> Validation

1. Duplicate Facet

It was found that data has 2044 unique entries by checking with Duplicate Facet.

| - |
|---|
| |

Figure 9: Duplicate facet

2. Numerical Facet

The rows which should have text values present e.g. Accident Month, were checked with numeric facet to check that they do not have any text value present.

| ×AccidentMonth | change reset |
|-----------------------------|---------------|
| | unange reser |
| No numeric value present. | |
| × AccidentDay | change reset |
| No numeric value present. | |
| | |
| × AccidentSeverityValue | change reset |
| No numeric value present. | |
| × TravelDirection | change reset |
| No numeric value present. | |
| ine inerite terres presents | |
| × ZoneValue | change reset |
| No numeric value present. | |
| × CollisionTypeValue | change reset |
| | energie reest |
| No numeric value present. | |
| × VehicleTypeValue | change reset |
| No numeric value present. | |
| | |
| × VehicleCompanyValue | change reset |
| No numeric value present. | |
| × LightConditionValue | change reset |
| No numeric value present. | |
| the manufactor present. | |
| × WeatherValue | change reset |
| No numeric value present. | |
| | change reset |
| × AccidentCauseValue | unange reser |
| No numeric value present. | |
| | |

Figure 10: Numerical Facet

3. Blank Facets

The whole data i.e. for all 26 columns was verified for null values using Blank Facet. 'False 2044' indicates that none of the 2044 rows has any null value present.

| * KilometerMarker | LightConditionValue | change |
|-----------------------------|-----------------------------------|---|
| I choices Sort by, same cou | 1 choices Sort by: name count | |
| false 2044 | false 2044 | - |
| Facet by choice counts | Facet by choice counts | - |
| * TravelDirection | KordGeometryValue | change |
| 1 choices Sort by: same cou | 1 chokes Sort by: name count | |
| false 2044 | false 2044 | |
| Facet by choice counts | Facet by choice counts | - |
| AccidentSeverityValue | × AccidentCauseValue | change |
| | | Control of the |
| | 1 choises Soft by: name count | |
| | false 2044 | - |
| Facet by choice counts | Facet by choice counts | |
| × CollisionTypeValue | * AccidentDay | change |
| 1 choices Sort by: some cou | 1 choices Sort by: name count | |
| false 2044 | talse 2044 | |
| Facet by choice counts | Facet by choice counts | |
| × WeatherValue | × MotorwayNameValue | change |
| 1 choices Sort by: name cou | | |
| false 2044 | | 12 |
| Facet by choice counts | 10100 2044 | - |
| | false 2944 Facet by choice counts | 1 choices Sort by, same coult 1 choices Sort by; same coult failse 2044 Facet by choice counts K TravelDirection 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices Sort by; same coult failse 2044 Facet by choice counts K TravelDirection 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices Sort by; same coult failse 2044 Facet by choice counts K AccidentSeverityValue 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices Sort by; same coult failse 2044 Facet by choice counts K CollisionTypeValue 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices Sort by; same coult failse 2044 Facet by choice counts K WeatherValue 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices counts K WeatherValue 1 choices Sort by; same coult 1 choices Sort by; same coult 1 choices counts K MotorwayNameValue 1 choices Sort by; same count 1 choices Sort by; same coult 1 choices Sort by same count failse 2044 Facet by choice counts |

Figure 11: Blank facets

Database Creation:

IBM DB2 Enterprise Server was chosen as a database server. All the refined data was loaded into the database from csv files. A snapshot is as follows

| 0 | bject View |
|----|--------------|
| Co | ntrol Center |

| - 🛅 All Databases |
|---------------------|
| ACCIDENT |
| 🥟 Tables |
| 🛅 Views |
| 🛅 Triggers |
| Application Objects |
| # 🗍 COGNOS |
| 😐 🗍 GSDB |
| 0 SAMPLE |
| 🚊 🗍 TOOLSDB |
| |

| HAMZA-77CC36518 - DB2 - A | CCIDENT - Tables | | | | | | | |
|---------------------------|------------------|-------------|--------|------------------------|--------------------------|-------|--------------|----------------|
| Name 🗢 | Schema | Table space | Commer | to Index table space o | Large data table space 🖨 | Type¢ | Cardinality¢ | Statistics tir |
| ACCIDENT_CAUSE | ACCIDENT | USERSPACE1 | | | | T | 40 | 1/28/13 5:2 |
| ACCIDENT_FACT | ACCIDENT | USERSPACE1 | | | | T | 3470 | 4/16/13 4:5 |
| ACCIDENT_SEVERITY | ACCIDENT | USERSPACE1 | | | | T | 5 | 1/28/13 5:2 |
| BEATS | ACCIDENT | USERSPACE1 | | | | T | 57 | 4/14/13 3:1 |
| COLLISION_TYPE | ACCIDENT | USERSPACE1 | | | | т | 9 | 1/28/13 5:2 |
| E LOCATION | ACCIDENT | USERSPACE1 | | | | Т | 1751 | 4/17/1312 |
| MOTORWAY_NAMES | ACCIDENT | USERSPACE1 | | | | T | 3 | 4/14/13 3.5 |
| ROAD_GEOMETRY | ACCIDENT | USERSPACE1 | | | | т | 5 | 1/28/13 5.2 |
| ECTORS | ACCIDENT | USERSPACE1 | | | | T | 16 | 4/14/13 3:1 |
| TIME_DIM | ACCIDENT | USERSPACE1 | | | | T | 2556 | 4/11/131:5 |
| TRAVEL_DIRECTIONS | ACCIDENT | USERSPACE1 | | | | T | 2 | 4/16/13 4.5 |
| VEHICLE | ACCIDENT | USERSPACE1 | | | | Т | 80 | 4/14/13 4:0 |
| VEHICLE_COMPANIES | ACCIDENT | USERSPACE1 | | | | T | 22 | 4/13/13 10 |
| VEHICLE_TYPES | ACCIDENT | USERSPACE1 | | | | T | -1 | |
| WEATHER | ACCIDENT | USERSPACE1 | | | | T | 6 | 1/28/13 5:2 |
| T ZONES | ACCIDENT | USERSPACE1 | | | | т | 5 | 4/14/13 3.1 |

Figure 12: IBM DB2

| | editing. | | | | |
|-----------------|----------|----------------|---------------------------------------|------------|------------|
| COLLISION_ID \$ | | LOCATION_ID \$ | DATEID 👙 | TIME_ID 🔶 | Add Row |
| 1 | | 1 | 188 | 20080706 📩 | Delete Roy |
| 1 | - 11 | 1 | 188 | 20080706 | Delete How |
| 8 | | 2 | 196 | 20080714 | |
| 8 | 1 | 2 | | 20080714 | |
| 8 | (77.64 | 2 | | 20080714 | |
| 7 | | 3 | 21.02.02 | 20080717 | |
| 6 | | 4 | 206 | 20080724 | |
| 4 | | 5 | | 20080714 | |
| 1 | | 6 | 10.000 | 20080701 | |
| 1 | | 6 | 183 | 20080701 | |
| 7 | | 7 | 185 | 20080703 | |
| 7 | | 8 | 367 | 20090101 | |
| 7777 | | 9 | 371 | 20090105 | |
| 1 | | 9 | 371 | 20090105 | |
| 1 | 10.00 | 10 | i i i i i i i i i i i i i i i i i i i | 20090118 | |
| 1 | 2.2.24 | 10 | 10 Sec. 1 | 20090118 | |
| 1 | | 10 | 384 | 20090118 | |
| 7 | | 11 | 392 | 20090126 | |
| 7 | | 12 | 377 | 20090111 | |
| 1 | | 13 | 0 202 | 20090117 | |
| 1 | | 2.6 | | 20090110 | |
| 1 | | 14 | | 20090110 | |
| 1 | | 14 | 577270 | 20090110 | |
| 1 | | 14 | | 20090110 | |
| 1 | 12 | 14 | 376 | 20090110 | |
| < | | | | 3 | |

Figure 13: DB2

Installation of COGNOS BI:

After database creation, the next step was to configure an IBM Cognos BI server, in order to move to metadata model development and BI reporting and analysis.

We installed the IBM Cognos BI 10.1.1. The server was setup up using the standard procedure, available in the Official Installation Guide published by IBM Corporation, USA. The following components of IBM Cognos BI were available for use, after the successful deployment:

- Framework Manager
- Report Studio
- Query Studio

Framework Manager:

Metadata model was created for the purpose of reporting in which facts and Dimensions related to Auto Accidents were identified. The information given comes from the data mart.

Metadata model hides the structural complexity of the underlying data sources and makes reporting faster and easier. Advantages of framework manager include predictable results, an easy to use interface and a dimensional model in the form of Star Schema.

Star schema:

It is the dimensional view necessary for data warehousing and for intelligent reporting. Following are individual dimensions in the star schema followed by the star schema.

Accident Fact

- No of drivers killed
- No of drivers injured
- No of passengers killed
- No of passengers injured
- No of pedestrians killed
- No of pedestrians injured

Weather

- Weather ID
- Weather Type

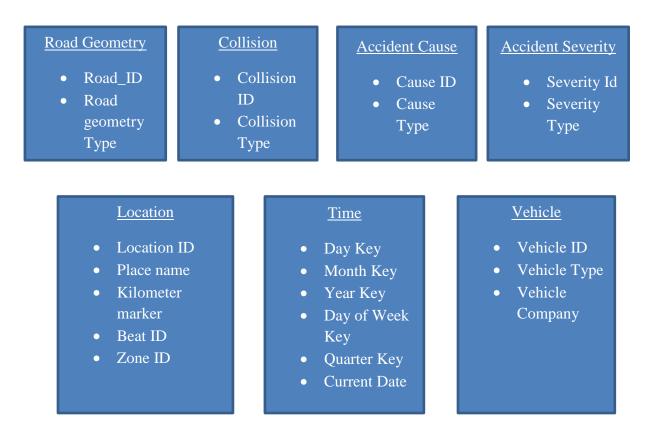




Figure 14:Star Schema

Reporting:

After the metadata model was completed the next step was to publish it on Cognos content so that we may form reports and perform analysis on the data. We made different reports which we will mention in detail in the next chapter. A couple of examples are as follows.

| Cause Name | Number Of Acidents |
|--|-----------------------|
| Fatal | |
| Careless Driving (Negligence of Driver) | 268 |
| Improper Crossing By Pedistrain | 155 |
| Dozing at Wheel | 137 |
| Tyre Burst | 73 |
| Taking Dangerous U-Turn / U-Turn where Prohibited | 68 |
| Over Speeding | 59 |
| Dangerous Overtaking / Overtaking where prohibited | 58 |
| Improper stopping/Turning or Changing Direction including Sudden Lane Changing by Motor vehicles and converging to Highway | 37 |

Accident Cause With most severity

Figure 15: Report Example

| Number o | f Accidents | Fatal | Major | Minor | Property Damage | Total |
|----------------|------------------|-------|-------|-------|-----------------|-------|
| Animal Vehicle | Info Required | 11 | 3 | | | 14 |
| | Total | 11 | 3 | | | 14 |
| Bus | BUS FASAIL MOVER | 11 | 9 | 4 | | 24 |
| | Deawoo | 7 | 2 | 4 | 1 | 14 |
| | HINO | 79 | 41 | 32 | 1 | 153 |
| | Info Required | 61 | 28 | 24 | 1 | 114 |
| | Mazda | 1 | | | | 1 |
| | Toyota | 2 | 3 | | | 5 |
| | Total | 161 | 83 | 64 | 3 | 311 |
| Car | DAIHATSU | 8 | 15 | 3 | | 26 |
| | Honda | 33 | 36 | 14 | 1 | 84 |
| | HYUNDAI | 3 | 2 | 1 | 2 | 8 |
| | Info Required | 55 | 42 | 10 | 3 | 110 |
| | KIA | 2 | 2 | | | 4 |
| | Mazda | 1 | | | | 1 |
| | Mitsubishi | 1 | 4 | 2 | | 7 |
| | SUZUKI | 119 | 140 | 29 | 3 | 291 |

Figure 16: Report example 2

Statistical Analysis:

Most important step of the project is to statistically analyze the data to understand the impact of different variables that are affecting the particular data. For this purpose IBM SPSS statistics tool is used.

IBM SPSS:

The different files are loaded into IBM SPSS statistics tool and one comprehensive file containing all the required variables is saved in .sav format. A validation check is performed in the end to verify variable data types and measures.

| Туре | Width | Decimals | Label | Values | Missing | Columns | Align |
|---------|-------|----------|--------------|--------|---------|---------|---------|
| Numeric | 12 | 0 | 1.1.00010000 | None | None | 12 | Right |
| String | 12 | 0 | | None | None | 12 | E Left |
| String | 7 | 0 | | None | None | 7 | E Left |
| Numeric | 12 | 0 | | None | None | 12 | Right |
| Date | 11 | 0 | | None | None | 11 | Right |
| Numeric | 12 | 4 | | None | None | 12 | Right |
| Numeric | 12 | 0 | | None | None | 12 | Right |
| Numeric | 12 | 0 | | None | None | 12 | Right |
| Numeric | 12 | 1 | | None | None | 12 | Right |
| String | 7 | 0 | | None | None | 7 | ME Left |
| Numeric | 12 | 4 | | None | None | 12 | Right |
| String | 60 | 0 | | None | None | 50 | E Left |
| Numeric | 12 | 1 | | None | None | 12 | Right |

Figure 17: SPSS Example

Solution Architecture

After the detailed analysis of the problem statement and deeper insight into the project, an approach to address the problem statement was devised.

A layered plan of action from the first step i.e. data sources to the last step i.e. Predictive Modeling, is adapted to achieve all the objectives smoothly and timely.

The diagram of the detailed 6-layered solution architecture of the project is given below.

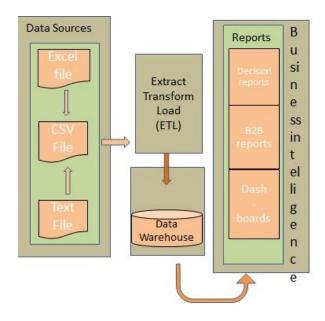


Figure 18: Architecture Diagram

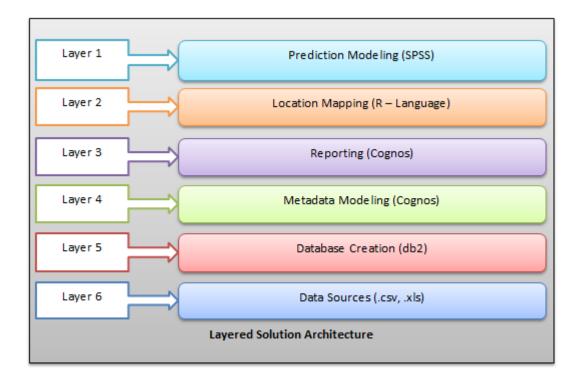


Figure 3: Layered Solution Architecture

From the figure, we can see that layer 6 and layer 5 are related to data and database. In these two layers the data has been loaded into the database after collection from different data sources and performing data cleaning. Database is created in IBM DB2.

In layer 4, a metadata model is created which would store the star schema groupings of the dimension tables and a fact table.

Now, metadata model is ready for business intelligence reports to be made upon it. Thus in layer 3, business intelligence reports are made with suggestion and recommendation alongside.

Also, the locations of accident spots are marked on map using R-language.

Finally, comes the part of predictive analytics in which the objective is to predict the severity category of the accidents depending upon various independent factors.

Tools and Technologies

The tools and technologies being used to achieve the stated objectives by following the proposed strategy are as follows:

| | Tools | Purposes |
|---------------------|--------------------------------------|---|
| Database | Google Refine | Data Cleaning and |
| Database | | Refining |
| | IBM DB2 Enterprise Server Edition | Database Creation |
| | Cognos: Framework Manager | Metadata Modeling for |
| BI Modeling and | | quick reporting and trend |
| Reporting | | |
| | Cognos: Report Studio | For making intelligent reports and dashboards |
| Predictive Analysis | IBM SPSS Statistics | Prediction Modeling |
| Mapping | R-Language | Plotting locations on maps |

Table 2: Tools and Technologies

Chapter 4

RESULTS

Our results are divided into two main categories, Analysis results and Prediction results. We will discuss analysis results first.

Analysis Results:

These results include the different reports we created to answer different questions about accidents, their causes and how to reduce them. These reports are divided into the following categories.

- 1. Drill reports.
- 2. Exploratory reports.
- 3. Descriptive Reports.
- 4. Active reports.

Drill reports:

Drill reports include those reports which involve drill through process, meaning that you can go from one report to the other just by clicking on it. We made two drill reports, as follows.

Year Wise Report:

This first drill report is basically a year wise analysis of accidents which is further drilled to quarters, then months and finally to days. As you can see, we cannot say that accidents are decreasing or increasing with the passage of time. For 2009 the accidents are large whereas for 2010 and 2011 they are less, for 2012 the data is till mid of august and if we extrapolate the analysis we see the accidents roughly equal to 2009.

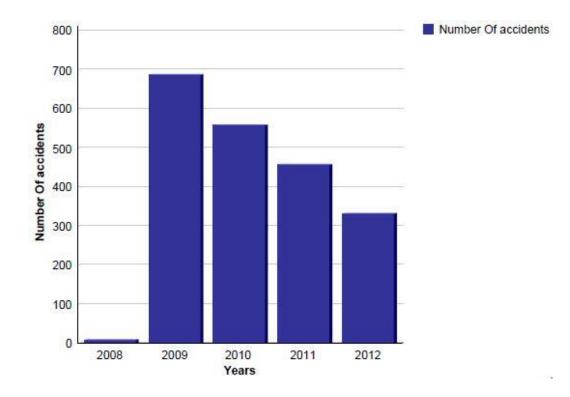


Figure 19: Year Wise

Location Wise Report:

This drill report is about locations, first off is the general distribution of accidents with respect to locations, which then drills down to zones and sectors. The location analysis tells us that the accidents on motorway are very less as compared to national highway this is because of certain extra precautions that the motorway implies while are absent on national highway.

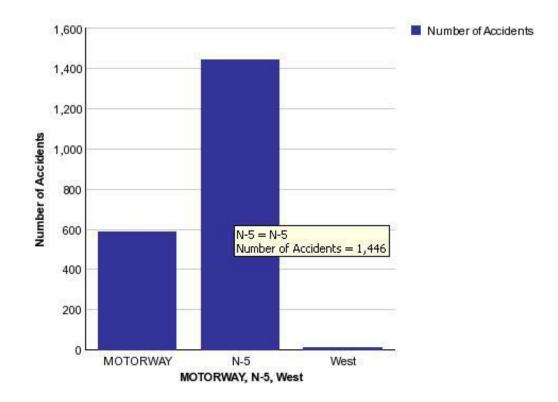


Figure 20: Location Wise

Exploratory Analysis:

It is used when one result lets us explore additional results and in the end it leads us to an answer to the problem. The following reports were made for this purpose.

Dozing at Wheel:

We see from the following report that when the road geometry is straight, dozing at wheel is contributing as an accident cause. This warranted further study and hence we further explored what role dozing at wheel plays.

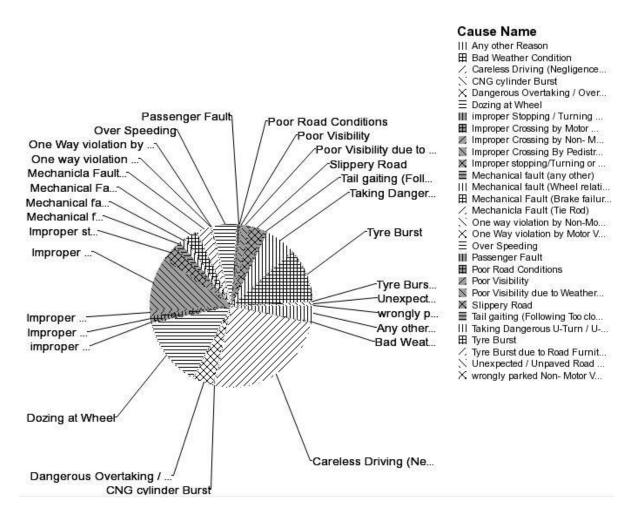


Figure 21: Road Geometry and Cause

Vehicle types and dozing at wheel:

The next stage of exploration was to see if different vehicle types had more dozing at wheel incidents than others, and we found out that heavy transport vehicles like trailers trucks etc. have more cases of dozing at wheel than other vehicle types. We then further explored to see which locations had a large number of accidents due to dozing at the wheel.

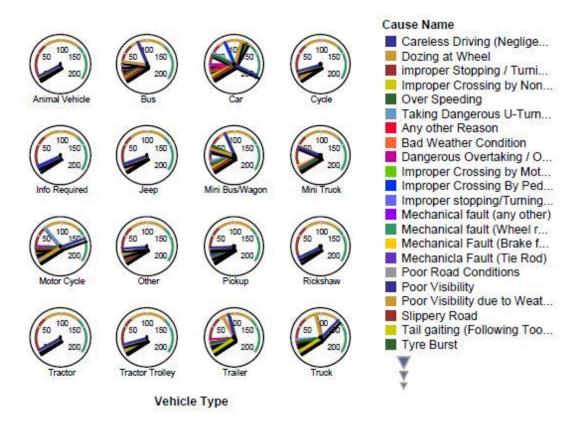


Figure 22: Vehicle Type and Cause

Locations and Dozing at Wheel:

The third report we made was location wise analysis of Dozing at Wheel accidents. So that we might have an idea about the locations which involve dozing at wheel accidents and we found out that a few places as seen from the image have more frequency of these accidents.

| Place Name | CASENO_ID |
|------------------|-----------|
| Dozing at Wheel | |
| Sahiwal | 6 |
| Burhan | 5 |
| Chakri | 5 |
| 149 Beat 09 | 4 |
| 246 (Beat-07) | 4 |
| 54 | 4 |
| 88 Beat 10 | 4 |
| ATTOCK | 4 |
| Khankadogrra | 4 |
| Muridke Bus Stop | 4 |
| Rashakai | 4 |
| 100 (Beat 10) | 3 |
| 1059 (Beat 16) | 3 |
| 239 | 3 |
| 329 | 3 |
| 55 (Beat 11) | 3 |
| 66 Beat 10 | 3 |
| Gujranwala | 3 |

Figure 23: Location Wise dozing

Mechanical Faults:

A report we made was on mechanical faults and the effects it has on different vehicles and if any vehicles have more faults than others. We found that brake failure was a major factor in some of the heavy transport vehicles and hence this warranted further exploration.

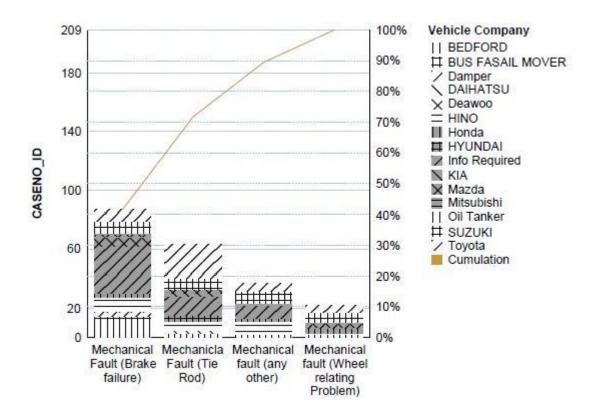


Figure 24: Mechanical Faults in Vehicle Types

HTV vs. LTV:

This report further clarified to us that mechanical faults are more prevalent in heavy vehicles than in light vehicles. The brake failure especially is greater in heavy vehicles hence we came to our second exploratory result.

| CASENO_ID | Any other Reason | CONTRACTOR OF STREET | chanica It (any er) | | | anical fault el relating em) | | anical (Brake e) | Mech Fault Rod) | | Tyre Burst | Total |
|------------------------|--------------------------------|----------------------|---------------------------|----|----|---|----|------------------------------------|-----------------------|-----------------------------|---------------|-------|
| BEDFORD | 9 | | 24.1 | 2 | | 2 | | 11 | | 3 | 8 | 35 |
| BUS FASAIL MOVER | 1 | | | | | | | 3 | | | 1 | 5 |
| Damper | 2 | | | | | | | 3 | | | 3 | 8 |
| Deawoo | 1 | | | | | | | 1 | | | 1 | 3 |
| HINO | 7 | | | 9 | | 1 | | 9 | | 6 | 14 | 46 |
| Oil Tanker | 2 | | | 1 | | | | 3 | | | 5 | 11 |
| Total | 22 | | | 12 | | 3 | | 30 | | 9 | 32 | 108 |
| CASENO_ID | Mechanic Fault (Tie Rod) | | Tyre Burst | | | Mechanical fau (Wheel relating Problem) | | Mechanic Fault (Bra failure) | | Mecha fault (a other) | | Total |
| DAIHATSU | | 2 | 7 | | | | | | | | 1 | 9 |
| DATSUN | | | 3 | | | | | | | | | 3 |
| Honda | | 2 | 19 | | 19 | | 2 | | 3 | | | 45 |
| HYUNDAI | | 2 | 7 | | 2 | | 1 | | | | 2 | 14 |
| Mitsubishi | | | 4 | | 1 | | | | 1 | | 1 | 7 |
| SUZUKI | | 7 | 49 | | 7 | | 7 | | 5 | | 7 | 82 |
| Toyota | | 23 | 100 | | 18 | | 5 | | 8 | | 6 | 160 |
| Total | | 36 | 189 | | 47 | | 15 | | 17 | | 16 | 320 |

Figure 25: HTV vs LTV

Maximum accidents location wise:

The third exploratory report we made was on the maximum number of accidents to occur in a location. We came to this report via the map results which showed that Sunder has the most number of accidents.

| Cause Name | CASENO_ID |
|--|-----------|
| Sunder | |
| Careless Driving (Negligence of Driver) | 33 |
| Improper Crossing By Pedistrain | 8 |
| Over Speeding | 6 |
| Slippery Road | 3 |
| Dangerous Overtaking / Overtaking where prohibited | 2 |
| Improper stopping/Turning or Changing Direction including Sudden Lane Changing by Motor vehicles and converging to Highway | 2 |
| Mechanical fault (any other) | 2 |
| Unexpected / Unpaved Road or Poor Road Condition i.e Rutting etc | 2 |
| Any other Reason | 1 |
| Dozing at Wheel | 1 |
| Improper Crossing by Motor Vehicles | 1 |
| Mechanical Fault (Brake failure) | 1 |
| Mechanicla Fault (Tie Rod) | 1 |
| One way violation by Non-Motor Vehicles (Bicycles, Animals / Animal Drawn Vehilces etc) | 1 |
| One Way violation by Motor Vehicles | 1 |
| Poor Visibility due to Weather Condition | 1 |
| Tail gaiting (Following Too closely) | 1 |
| Taking Dangerous U-Turn / U-Turn where Prohibited | 1 |

Figure 26: Accident Causes at Sunder

Pedestrian Accidents in Sunder:

Upon further exploration we found out that pedestrian related accidents were the most frequent in Sunder, the reason being that there is no overhead pedestrian bridge, hence a large number of pedestrians walk on the road.

| Place Name | Collision Type | CASENO_ID |
|---------------------|----------------|-----------|
| Mandra | Pedestrian | 13 |
| Sunder | Pedestrian | 13 |
| Gujar Khan | Pedestrian | 8 |
| TAXILA | Pedestrian | 6 |
| 23 (M-2(S)) Beat-12 | Pedestrian | 5 |
| 1516 (Beat 05) | Pedestrian | 4 |
| Beat 5 Rawat(1517) | Pedestrian | 4 |
| Dina Bridge | Pedestrian | 4 |
| Kamonke | Pedestrian | 4 |
| 1506 | Pedestria 🛱 | |
| CITY KAMONKE | Pedestrian | 3 |

Figure 27: Pedestrian accidents vs Locations

Over speeding incidents in Sunder:

Another result we found out was that over speeding accidents were also the most frequent in Sunder, due to the vehicles there driving very fast. Hence a strict speed check should be enforced in that place to reduce the accidents.

| Place Name | CASENO_ID |
|------------------------|-----------|
| Over Speeding | |
| Sunder | 6 |
| 1575 | 3 |
| 1124 (Beat 15) | 2 |
| 1451 | 2 |
| 1507 | 2 |
| 1568 | 2 |
| 400 | 2 |
| 8 Beat 01 (M-1) | 2 |
| AzaKhel | 2 |
| Barriel no 2 wah cantt | 2 |
| Burhan | 2 |

Figure 28: Overspeeding vs Location

Descriptive Reports:

These reports are the basic reports which basically answer a simple question or prove some fact. They can also be made in simple databases. Following are some descriptive reports.

Motorcycle fatality ratio:

During our research we found a survey that motorcycle riders have the most number of fatal accidents, using the data we thought to see whether this was in fact true or not. And we found out that it was true. Motorcycle riders have a fatality rate of 71% higher than any other transport type.

| Motor Cycle | Severity Category | CASENO_ID |
|---------------|-------------------|-----------|
| Motor Cycle | Fatal | 292,201 |
| Motor Cycle | Major | 127,461 |
| Motor Cycle | Minor | 25,247 |
| Motor Cycle | Property Damage | 1,403 |
| Overall - Tot | 446,312 | |

Figure 29: Motorcycle fatalities

Pedestrian related accidents:

Another survey we read stated that pedestrians here are not aware of road crossing and safety rules and hence get involved in accidents. We checked this with our data and found out that yes indeed this was the case.

| Pedestrian | Fatal | CASENO_ID | Cause Name |
|--------------|-------|-----------|--|
| Pedestrian | Fatal | 3 | Any other Reason |
| Pedestrian | Fatal | 25 | Careless Driving (Negligence of Driver) |
| Pedestrian | Fatal | 3 | Dangerous Overtaking / Overtaking where prohibited |
| Pedestrian | Fatal | 4 | Dozing at Wheel |
| Pedestrian | Fatal | 1 | Improper Crossing by Motor Vehicles |
| Pedestrian | Fatal | 132 | Improper Crossing By Pedistrain |
| Pedestrian | Fatal | 2 | Mechanical Fault (Brake failure) |
| Pedestrian | Fatal | 4 | Over Speeding |
| Pedestrian | Fatal | 1 | Passenger Fault |
| Pedestrian | Fatal | 1 | Poor Visibility due to Weather Condition |
| Pedestrian | Fatal | 1 | Tyre Burst due to Road Furniture i.e Cat Eyes etc |
| Overall - To | otal | 177 | |

Figure 30: Pedestrian Accidents Causes

Days of Week report:

Another descriptive report we made was about the comparison of whether the number of accidents change during weekdays or weekends. We found out that Thursday and Friday have the most accidents due to the fact that most travelers of highways are businessmen who get Friday off so they travel on these two days. Then Saturday and Sunday have greater accidents due to government employees traveling mostly on these days and Wednesday has the fewest.

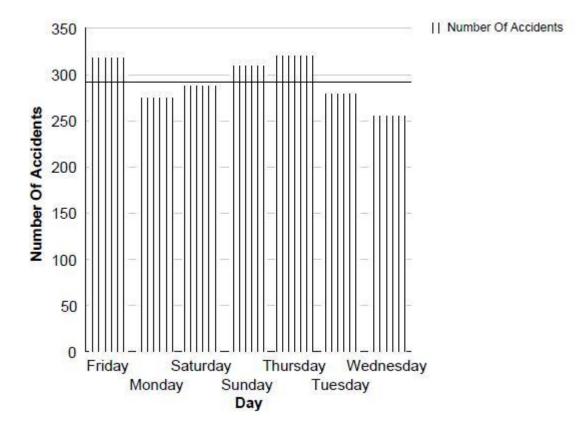


Figure 31: Day of Week Analysis

Active Reports:

These reports are called such because they are interactive, they allow the user to make changes to them and see their result in real time. This report can be used by people to show results and consequences in real time and to do what if analysis. Following report is an active report allowing the user to filter the search according to what he/she feels is best.

| Mini Truck | (i | Y Side Swipe | 71 | ✓ Fatal | 16 | ✓ Dry | V Straight | |
|-----------------|--------------------|-------------------|----------------------|-----------------|--------------------------|-------------------|---|-------------------|
| Vehicle Type | Vehicle Company | Collision Type | Severity Category | Weather Type | Road Geometry Type | Place Name | Cause Name | Total Injuries |
| Mini Truck | BEDFORD | Side Swipe | Fatal | Dry | Straight | Qadirabad | Dangerous Overtaking / Overtaking where prohibited | 0 |
| Mini Truck | Info Required | Side Swipe | Fatal | Dry | Straight | Manga pull | Tail gaiting (Following Too closely) | 1 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | 254(M-2 North) | Careless Driving (Negligence of Driver) | 1 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | Dina Bridge | Careless Driving (Negligence of Driver) | 0 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | EME | Careless Driving (Negligence of Driver) | 0 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | 780 | Dozing at Wheel | 1 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | Rehmannia | Improper stopping/Turning or Changing Direction including Sudden Lane Changing by Motor vehicles and converging to Highway | 0 |
| Mini Truck | Mazda | Side Swipe | Fatal | Dry | Straight | Hino U-Turn | Taking Dangerous U-Turn / U-Turn where Prohibited | 2 |

Figure 32: Active Report

Prediction Results:

This was done using binary logistic regression. The data we collected was totally categorical. Our dependent variable was categorical and independent variables were categorical as well. Therefore the only statistical model suitable for prediction was binary logistic regression.

Our dependent category was accident severity, our prediction was on whether an accident will be fatal or not. Initially it had multiple categories of fatal major minor property damage. This was giving us less accuracy in prediction so we grouped it into two major categories.

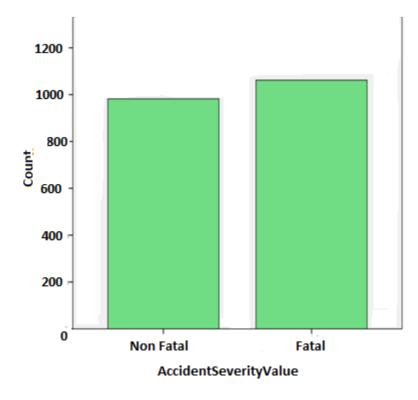


Figure 33: Accident Severity

We also had to group multi category variables like accident causes and collision type. We grouped them as follows.

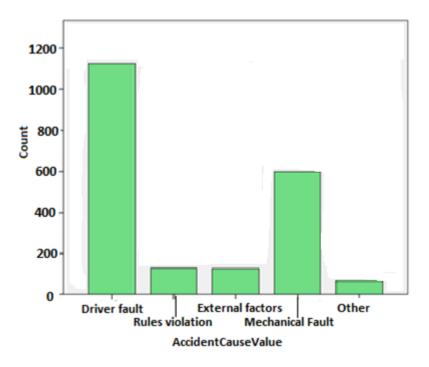


Figure 34: Accident Cause

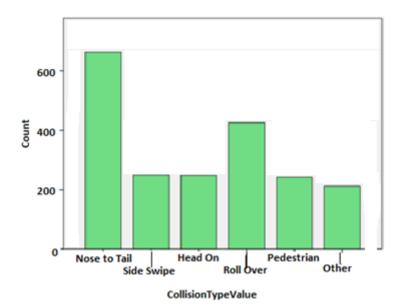


Figure 35: Collision Type

Independent Variable Impact on Accident Severity:

We then check and see whether these variables have any effect on accident severity or not.

Accident Cause Value:

We found that most of the accidents are caused due to the Driver Fault. Further, under the category of driver fault, we see that the bar for fatal accidents is a little higher than for non-fatal. Moreover there are fewer accidents caused due to external factors which include light conditions and weather factors.

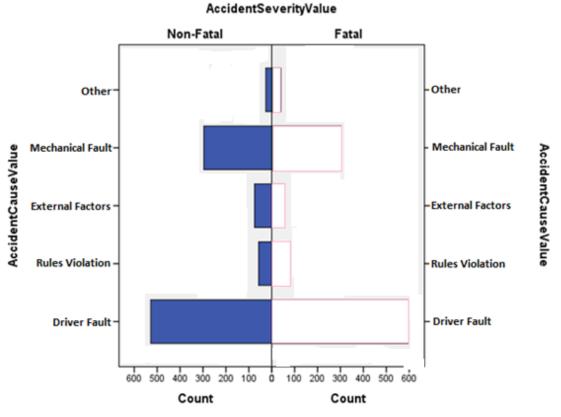


Figure 36: Cause Value

Collision Type Value:

We see that most of the accidents are caused due to the Nose to Tail Collision. Further, more Non-fatal accidents are caused by Roll over collision type than fatal. Additionally, when the vehicles hit the pedestrians, the result is mostly a fatal accident.

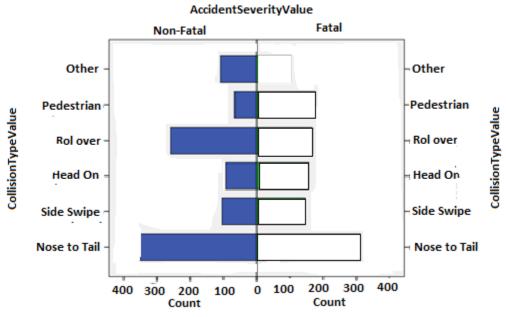


Figure 37: Collision value

Location Value:

We had three cases to consider, motorway name, sector or zone. We found out that motorway name was very generic while sector had too much information to be a good predictor hence we select zone value as a good predictor.

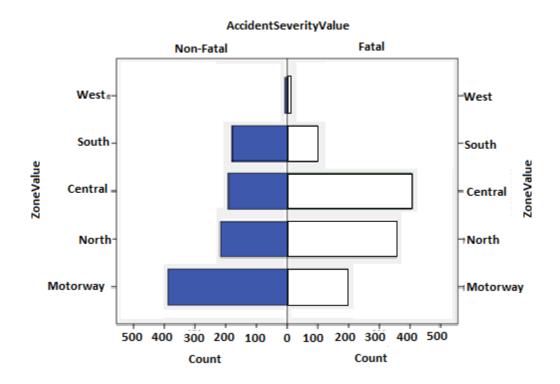


Figure 38: Zone Value

Training Model:

We then trained our model on 80% of data and came to this result.

| Variable | 2 Log likelihood | Difference | P - value | Accuracy | Decision |
|--------------|------------------|------------|-----------|----------|----------|
| First Model | 2284.802 | - | - | 53.5% | - |
| Second Model | 2193.343 | 91.459 | 0.00000 | 60.1% | Reject |
| Third Model | 2073.181 | 120.162 | 0.00000 | 67.8% | Accept |
| Fourth Model | 2072.798 | 0.392 | 0.53125 | 67.9% | Reject |
| Fifth Model | 2065.252 | 7.929 | 0.00486 | 67.4% | Reject |
| Sixth Model | 2072.849 | 0.668 | 1.00000 | 67.8% | Reject |

Figure 39: Training Set

We then verified it by using three test models.

| Model 1 | Training Set | First set + Second Set | 66.4 % |
|---------|--------------|------------------------|--------|
| Model 2 | TrainingSet | First Set + Third Set | 66.1 % |
| Model 3 | TrainingSet | Second Set + Third Set | 68.2% |

Average Accuracy of the Training Sets = 66.9%

Figure 40: Three Set

Model Testing:

We then tested our model for both the training sets. First is the result of 80 % sample model. We tested on 20 % remaining data.

| | | Accuracy = e * Predicted V | | bulation | |
|-------------------------|---|-------------------------------|----------------|----------|--------|
| | and the second secon | | PredictedValue | | Total |
| | | | .00 | 1.00 | |
| | Non-Fatal Fatal | Count | 102 | 87 | 189 |
| A | | % of Total | 26.4% | 22.5% | 49.0% |
| Accident Severity Value | | Count | 56 | 141 | 197 |
| | | % of Total | 14.5% | 36.5% | 51.0% |
| Tabal | | Count | 158 | 228 | 386 |
| Total | | % of Total | 40.9% | 59.1% | 100.0% |

Figure 41: Prediction Accuracy-20% test

Below is the actual graph predicting about accident severity.

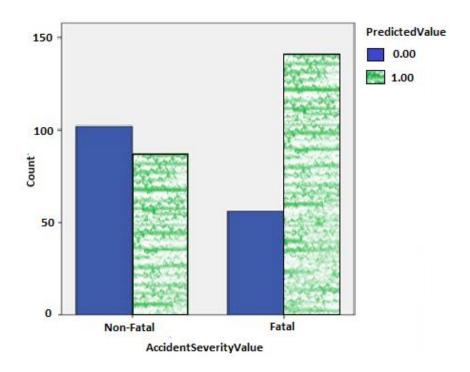


Figure 42: Accuracy of Prediction

Combination of three dataset model testing is just to verify that the results we got were in fact in approximation of these results.

Average accuracy of Test Sets = 65%

| | Data Set | | Percentage Accuracy |
|--------------|--------------|------------------------|---------------------|
| | Training Set | First Set + Second Set | 66.4% |
| Test Model 1 | Test Set | Third Set | 64.1% |
| | Training Set | First Set + Third Set | 66.1% |
| Test Model 2 | Test Set | Second Set | 66.9% |
| | Training Set | Second Set + Third Set | 68.2% |
| Test Model 3 | Test Set | First Set | 64% |

Chapter 5

CONCLUSIONS & RECOMMENDATIONS

Conclusion:

Automotive accidents are a global phenomenon which is increasing day by day. Pakistan is similarly affected by these accidents and a major portion of deaths are due to automobiles. Therefore a system needs to be in place to keep a check and balance on these accidents so that their frequency may decrease. We took data from National Highways and Motorway police and for the first time in Pakistan concluded reports and facts based on actual data instead of surveys or guesswork.

We identified places and locations where accidents are more frequent. We proved that motorcycle drivers are involved in the most number of fatal accidents, while a large portion of pedestrian accidents occur due to ignorance on road crossing and safety. In fact a large reason of accidents in Pakistan is the unawareness about road safety and driving etiquettes.

Our hope is that these analysis and reports will be a stepping stone in reducing the number of accidents in Pakistan. The scope of this project is unlimited. Our data was limited to motorway and national highway. This project can easily be extended to other parts of the country or cities where the numbers of accidents are even greater like Lahore, Rawalpindi etc.

There is no substitute for life and we should all try to work together in reducing the deaths caused by auto accidents.

Recommendations:

Following recommendations are made so that it might help people in the future.

Change data from categorical to Scale:

This will specially help in prediction as if the data collected is according to a scale then it will form a continuous distribution which will give even better results in prediction analysis.

A standard data gathering format:

World Health Organization has defined specific data gathering parameters which need to be filled when an accident takes place. It comprises of nearly 80 questions but unfortunately for Pakistan that measure cannot be implemented has a whole. So a compromise has to be made and custom based data gathering parameters must be defined which can be utilized here and which are acceptable to W.H.O as well.

Following are the recommendations we prepared for the motorway so that they may incorporate them and help reduce the number of accidents.

Pedestrian Crossings:

Overhead bridges at Mandra and Sunder need to be built and guard rails need to be implemented all along the road to stop pedestrians from crossing and only use the bridges. This would bring reduction in the pedestrian deaths.

Over speeding:

Speed cameras and at least one police car must be present in those locations where over speeding is frequent like Sunder. This would force drivers to be careful and reduce their speeds.

Dozing at Wheel:

Rumble Strips need to be implied on locations like Sahiwal where dozing at the wheel is a prominent cause of accidents. The strips will shake the vehicle strongly waking up the drivers and hopefully avoiding the accident.

Brake checks:

Heavy transport vehicles must be checked for any brake problems before being allowed to travel on an inclined road. If they check their brakes before climbing or descending then it would reduce the accidents.

Chapter 6

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