

# **INTRODUCTION**

## **1.1 OVERVIEW :**

An insurance policy is an arrangement by which a company undertakes to provide a guarantee of compensation for specified loss, damage, illness, or death in return for the payment of a specified premium. A premium is a sum of money that the customer needs to pay regularly to an insurance company for this guarantee. Building a model to predict whether a customer would be interested in Vehicle Insurance is extremely helpful for the company because it can then accordingly plan its communication strategy to reach out to those customers and optimize its business model and revenue

## **1.2 PURPOSE**

The main aim of this use case is to predict, whether the customer would be interested in Vehicle insurance provided information about demographics (gender, age, region code type), Vehicles (Vehicle Age, Damage), Policy (Premium, sourcing channel) etc

# **LITERATURE SURVEY**

## **2.1 EXISTING PROBLEM**

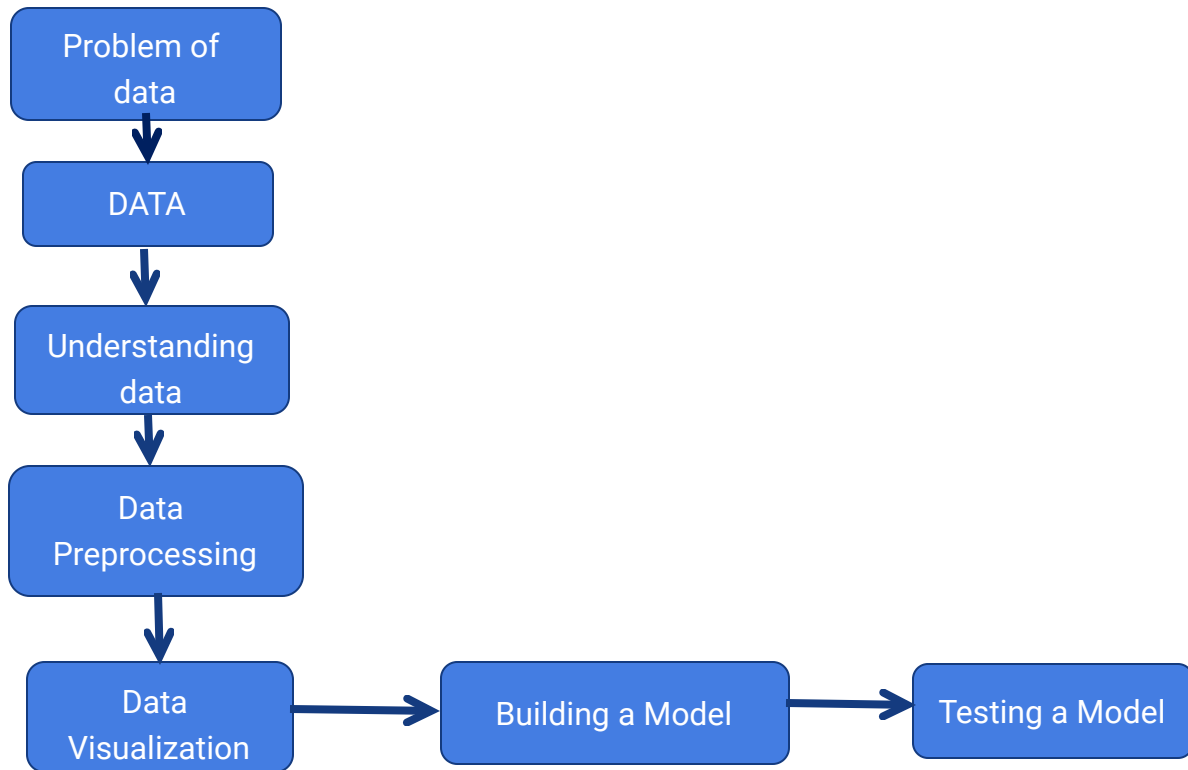
The customers don't know much about the process of insurance and the documentation what they have to submit while applying to Insurance.

## **2.2 Proposed Solution**

Here, we predict the vehicle insurance on the given details and we will give the explanation about the insurance. If any incident occurs what claims did he get and we tell about the difference between the premium account and normal account. In this we describe the benefits of the accounts.

## THEORITICAL ANALYSIS

### 3.1 BLOCK DIAGRAM



## HARDWARE / SOFTWARE DESIGNING

### SOFTWARE REQUIRED

- 1.WEKA SOFTWARE
- 2.ECLIPSE IDE
- 3.INSTALL TABLE SAW AND WEKA PACKAGES

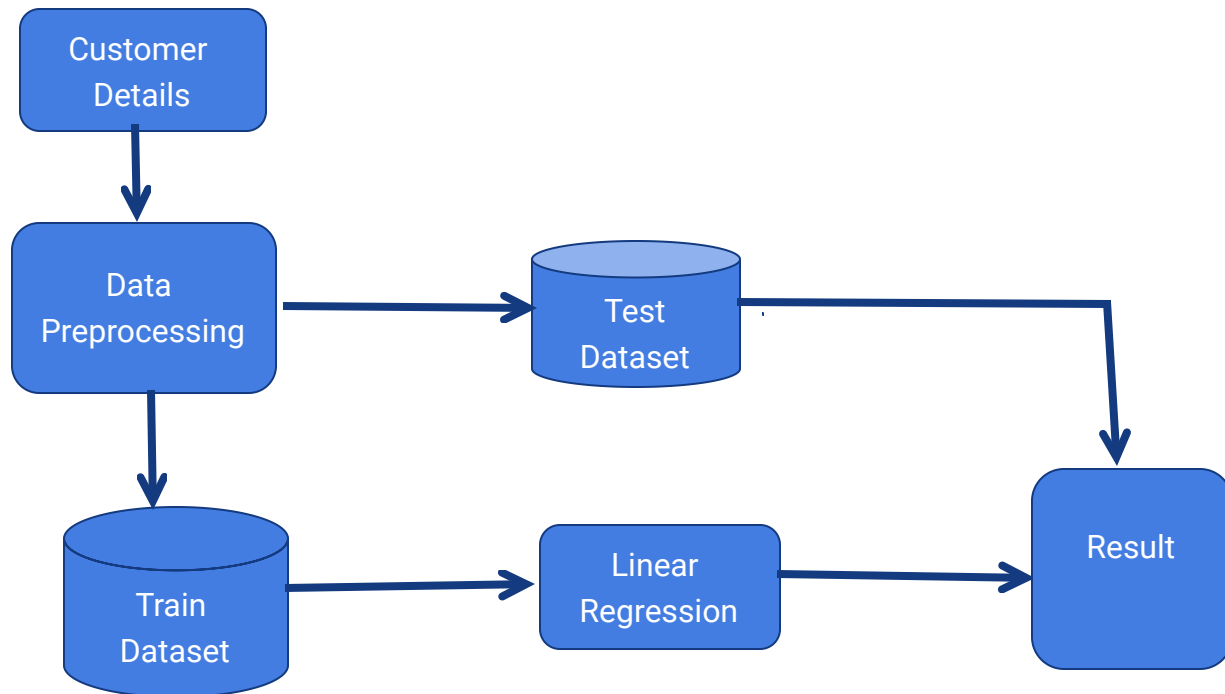
### EXPERIMENTAL INVESTIGATION

- People are not much familiar with the insurance .some people may not apply for the renewal of insurance.
- In earlier , there are less number of insurance companies .There are only

government insurance company

- In insurance there will be two types one is normal and other is premium.
- Normal only some benefits compare to premium.

## FLOWCHART



## RESULT

Correlation coefficient	1
Mean absolute error	6.6518
Root mean squared error	8.1034
Relative absolute error	0.032 %
Root relative squared error	0.0307 %
Total Number of Instances	780

Time taken to test model on supplied test set: 0.06 seconds

=== Summary ===

Correctly Classified Instances	636	81.5385 %
Incorrectly Classified Instances	144	18.4615 %
Kappa statistic	0.5162	
Mean absolute error	0.2708	
Root mean squared error	0.368	
Relative absolute error	73.1586 %	
Root relative squared error	85.5712 %	
Total Number of Instances	780	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.670	0.138	0.612	0.670	0.640	0.517	0.772	0.509	Y
	0.862	0.330	0.890	0.862	0.876	0.517	0.772	0.875	N
Weighted Avg.	0.815	0.283	0.822	0.815	0.818	0.517	0.772	0.785	

=== Confusion Matrix ===

```
a  b  <-- classified as
128 63 | a = Y
81 508 | b = N
```

Preprocess Classify Cluster Associate Select attributes Visualize

## Classifier

Choose LinearRegression -S 0 -R 1.0E-8 -num-decimal-places 4

## Test options

- ☐ Use training set
- ☒ Supplied test set
- ☐ Cross-validation Folds 10
- ☐ Percentage split % 66
- 

(Nom) fraud\_reported

Start

Stop

## Result list (right-click for options)

18:08:52 - trees.J48

18:09:48 - misc.InputMappedClassifier

23:03:02 - misc.InputMappedClassifier

23:04:07 - misc.InputMappedClassifier

23:07:12 - misc.InputMappedClassifier

23:10:51 - misc.InputMappedClassifier

## Classifier output

Time taken to test model on supplied test set: 0.06 seconds

=== Summary ===

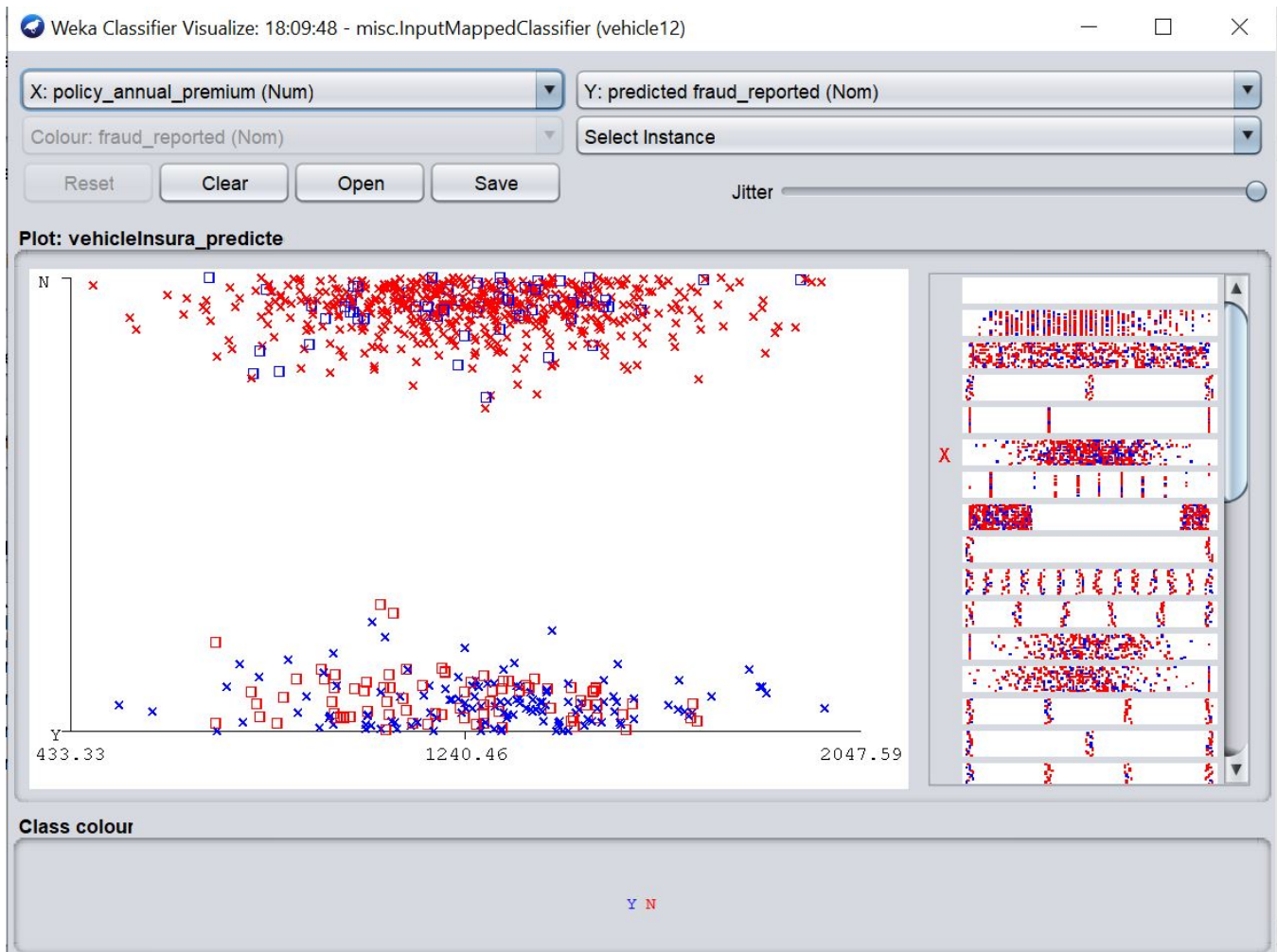
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## Advantages and Disadvantages

### Advantages

- Its Efficiency is 80%
- It Predicts accurately

### Applications

- It is used in insurance company

### Conclusion

- The customer can know whether he can apply for premium or normal. The result of insurance is given in few Minutes

### Future Scope

- It works very efficiently .

- If we give the predictions the growth of insurance will be increased

## BIBLIOGRAPHY

<https://www.kaggle.com/c/auto-insurance-fall-2017/data>

<https://www.synthesized.io/data-template-pages/vehicle-insurance-claim-prediction>

## APPENDIX

package org.dl;

```
import java.io.IOException;

import java.util.Arrays;

import tech.tablesaw.api.Table;
import tech.tablesaw.plotly.Plot;
import tech.tablesaw.plotly.components.Figure;
import tech.tablesaw.plotly.components.Layout;
import tech.tablesaw.plotly.traces.BoxTrace;
import tech.tablesaw.plotly.traces.ScatterTrace;
import weka.classifiers.Classifier;
import weka.classifiers.evaluation.Evaluation;
import weka.core.Instance;
import weka.core.Instances;
import weka.core.converters.ConverterUtils.DataSource;

public class DataAnalysis{

    public static Instances getInstances (String filename)
    {

        DataSource source;
        Instances dataset = null;
        try {
            source = new DataSource(filename);
```

```

        dataset = source.getDataSet();
        dataset.setClassIndex(dataset.numAttributes()-1);

    } catch (Exception e) {
        // TODO Auto-generated catch block
        e.printStackTrace();
    }

    return dataset;
}

@SuppressWarnings("static-access")
public static void main(String args[]) {
    System.out.println("data analytics");
    try {
        Table
insurance_data=Table.read().csv("C:\\Users\\hp\\Desktop\\1214\\org.dl\\src\\main\\java\\org\\dl\\V
ehicle12.csv");

        System.out.println(insurance_data.shape());
        System.out.println(insurance_data.structure());
        System.out.println(insurance_data.summary());
        Layout l1=Layout.builder().title("age distribution").build();
        //box-plot
        BoxTrace
t1=BoxTrace.builder(insurance_data.categoricalColumn("fraud_reported"),
insurance_data.nCol("vehicle_claim")).build();
        Plot.show(new Figure(l1,t1));
        //scatter-plot
        ScatterTrace
s=ScatterTrace.builder(insurance_data.nCol("policy_annual_premium"),insurance_data.nCol("vehicle_cl
aim")).build();
        Plot.show(new Figure(l1,s));
    }
    catch(IOException e) {
        e.printStackTrace();
    }
    Instances train_data =

```



```

getInstances("C:\\Users\\hp\\Desktop\\Train.arff");
        Instances test_data =
getInstances("C:\\Users\\hp\\Desktop\\Test.arff");
        System.out.println("The train data size is "+train_data.size());

        /** Classifier here is Linear Regression */
        Classifier classifier = new weka.classifiers.functions.Logistic();
        /** */
        try {
            classifier.buildClassifier(train_data);
        } catch (Exception e1) {
            // TODO Auto-generated catch block
            e1.printStackTrace();
        }

        /**
         * train the algorithm with the training data and evaluate the
         * algorithm with testing data
         */
        Evaluation eval;
        try {
            eval = new Evaluation(train_data);
            eval.evaluateModel(classifier, test_data);
            System.out.println("** Logistic Regression Evaluation

with Datasets **");

            System.out.println(eval.toSummaryString());
            System.out.print(" the expression for the input data as

per alogorithm is ");

            System.out.println(classifier);

            double confusion[][] = eval.confusionMatrix();
            System.out.println("Confusion matrix:");
            for (double[] row : confusion)
                System.out.println(    Arrays.toString(row));
            System.out.println("-----");

            System.out.println("Area under the curve");
            System.out.println( eval.areaUnderROC(0));

```

```

        System.out.println("-----");

System.out.println(eval.getAllEvaluationMetricNames());

        System.out.print("Recall :");

System.out.println(Math.round(eval.recall(1)*100.0)/100.0);

        System.out.print("Precision:");

System.out.println(Math.round(eval.precision(1)*100.0)/100.0);
        System.out.print("F1 score:");

System.out.println(Math.round(eval.fMeasure(1)*100.0)/100.0);

        System.out.print("Accuracy:");
        double acc = eval.correct()/(eval.correct()+
eval.incorrect());

        System.out.println(Math.round(acc*100.0)/100.0);

        System.out.println("-----");
    } catch (Exception e1) {
        // TODO Auto-generated catch block
        e1.printStackTrace();
    }

    /** Print the algorithm summary */

    Instance predicationDataSet = test_data.get(2);
    double value;
    try {
        value = classifier.classifyInstance(predicationDataSet);
        System.out.println("Predicted label:");
        System.out.print(value);
    } catch (Exception e) {
        // TODO Auto-generated catch block
        e.printStackTrace();
    }

```

```
}  
/** Prediction Output */
```

```
}
```

```
}
```