By Surajit Das, Mohsen Adam, Jamal Derrick.

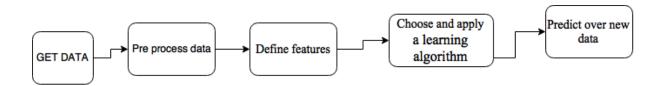
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FLIGHT DELAY PREDICTION DOCUMENTATION

Can an app be created to predict flight delays due to weather related issues?

The following paper tries to model solutions to predict flight delays due to weather related issues. The proposed models for flight delay prediction uses supervised machine learning techniques. The goal of this research is to be able to predict departure delays 2-24 hrs in advance. The paper analyzes the performance of the following algorithms on the flight delay prediction model namely – Simple Logistic Regression, Naive Bayyes classification and Support Vector Machines. The end goal is to choose the best performing algorithm and implement it to predict the delay of flights in advance. The objective of this research is to answer the question mentioned above with concrete results and explanation of the chosen solution.

The working model of the paper is explained with the aid of the following diagram



Step 1:- Get Data

The dataset for this model was collected from the following websites for the month (April – June) for the year 2013.

- 1. http://gallery.cortanaintelligence.com/Experiment/Binary-Classification-Flight-delay-prediction-3?share=1.
- 2. https://www.ncdc.noaa.gov/cdo-web/datasets.
- 3. http://www.transtats.bts.gov/DL_SelectFields.asp?Table_ID=236&DB_Short_Name=On-Time.
- 4. https://www.wunderground.com/.

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In order to maintain the accuracy of the dataset for the flight and weather, the datasets were cross checked for accuracy by validating the information from the above mentioned websites.

Step 2:- Pre-Process Data

The dataset obtained from the above sources was combined and a fresh dataset was created. The initial dataset contained 18 features combined of (flight + weather). We have chosen to work on a small cluster of data to optimize accuracy and get a better feel of the performance of the algorithm. Since, our main objective is to apply machine learning and the saying goes "There is no fixed size of dataset for better performance in Machine learning. The more, the merrier". So, our initial goal is to work on 344 datapoints and analyze the results. We can always add more data in our dataset in case of redundancy or conflicting results.

To select the most useful features for the prediction model, we used the Weka software.

• Check for Missing values in the database using the Weka Software.

To check for missing values in the dataset, the dataset is loaded in the Weka software. Then each column is checked in the Weka to identify any missing values. The dataset used in the current experiment doesn't contain any missing values.

Step 3:- Get Features

• Identify the most important features for the classification algorithms using Weka.

Algorithm 1:- Simple Logistic Regression

Available Features:-

- 1. Month
- 2. Day
- 3. Time
- 4. Timegroup
- 5. Airportid
- 6. Temperature (temperature of the origin airport).
- 7. DewPoint (of the origin airport)
- 8. humidity (of the origin airport)
- 9. Pressure (of the origin airport)
- 10. WindSpeed (of the origin airport)

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- 11. destination airport ID
- 12. desttemperature (of the destination airport)
- 13. destdewpoint (of the destination airport)
- 14. destinationPressure (of the destination airport)
- 15. destwindspeed (of the destination airport)
- 16. Prediction (Yes/No).

Our first algorithm that we wanted to experiment was Simple Logistic Regression classification. We wanted to Weka to help us select the best features available for the Simple Logistic Regression Classification Model. Weka returned us with the following results.

```
=== Run information ===
```

Evaluator: weka.attributeSelection.ClassifierSubsetEval -B

weka.classifiers.functions.SimpleLogistic -T -H "Click to set hold out or test

instances" -E acc -- -I 0 -M 500 -H 50 -W 0.0

Search: weka.attributeSelection.BestFirst -D 1 -N 5

Relation: r-weka.filters.unsupervised.attribute.Remove-R14-

weka.filters.unsupervised.attribute.Remove-R10-

weka.filters.unsupervised.attribute.Remove-R11

Instances: 344 Attributes: 16

Month

Day

Time

Timegroup

Airportid

Temperature

DewPoint

humidity

Pressure

WindSpeed

destination airport ID

desttemperature

destdewpoint

destinationPressure

destwindspeed

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Prediction

Evaluation mode: evaluate on all training data

=== Attribute Selection on all input data ===

Search Method:

Best first.

Start set: no attributes Search direction: forward

Stale search after 5 node expansions Total number of subsets evaluated: 127 Merit of best subset found: 0.299

Attribute Subset Evaluator (supervised, Class (nominal): 16 Prediction):

Classifier Subset Evaluator

Learning scheme: weka.classifiers.functions.SimpleLogistic

Scheme options: -I 0 -M 500 -H 50 -W 0.0

Hold out/test set: Training data

Subset evaluation: classification error

Selected attributes: 1,3,4,7,8,13,14:7

Month

Time

Timegroup DewPoint

humidity

destdewpoint

destinationPressure

The next step was to run the Simple Logistic Regression on the selected attributes as suggested by Weka. We ran the Simple Logistic Regression on the whole dataset first and then wanted to compare the results with the Simple Logistic Regression on the selected attributes of the dataset.

Results are displayed as below:-

1. Whole dataset (70% training set; 30% test set; 16 features)

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```
=== Run information ===
Scheme:
            weka.classifiers.functions.SimpleLogistic -I 0 -M 500 -H 50 -W 0.0
           r-weka.filters.unsupervised.attribute.Remove-R14-
Relation:
weka.filters.unsupervised.attribute.Remove-R10-
weka.filters.unsupervised.attribute.Remove-R11
Instances:
           344
Attributes: 16
        Month
        Day
        Time
        Timegroup
        Airportid
        Temperature
        DewPoint
        humidity
        Pressure
        WindSpeed
        destination airport ID
        desttemperature
        destdewpoint
        destinationPressure
        destwindspeed
        Prediction
Test mode: split 70.0% train, remainder test
=== Classifier model (full training set) ===
SimpleLogistic:
Class 0:
0.84 +
[Timegroup] * -0.46
Class 1:
-0.84 +
[Timegroup] * 0.46
```

Time taken to build model: 0.19 seconds
=== Evaluation on test split ===
Time taken to test model on training split: 0 seconds
=== Summary ===
Correctly Classified Instances 63 61.165 % Incorrectly Classified Instances 40 38.835 % Kappa statistic 0.2429 Mean absolute error 0.4498 Root mean squared error 0.4802 Relative absolute error 88.9816 % Root relative squared error 94.5167 %
Total Number of Instances 103
=== Detailed Accuracy By Class ===
TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class 0.761 0.509 0.547 0.761 0.636 0.258 0.733 0.701 no 0.491 0.239 0.718 0.491 0.583 0.258 0.733 0.697 yes Weighted Avg. 0.612 0.360 0.642 0.612 0.607 0.258 0.733 0.699
=== Confusion Matrix ===
a b < classified as 35 11 a = no 29 28 b = yes
2. Selected attributes (70% training set; 30% test set; 7 features)
=== Run information ===
Scheme: weka.classifiers.functions.SimpleLogistic -I 0 -M 500 -H 50 -W 0.0

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```
Relation:
           r-weka.filters.unsupervised.attribute.Remove-R14-
weka.filters.unsupervised.attribute.Remove-R10-
weka.filters.unsupervised.attribute.Remove-R11-
weka.filters.unsupervised.attribute.Remove-R2,5-6,9-12,15
Instances: 344
Attributes: 8
        Month
        Time
        Timegroup
        DewPoint
        humidity
        destdewpoint
        destinationPressure
        Prediction
Test mode: split 70.0% train, remainder test
=== Classifier model (full training set) ===
SimpleLogistic:
Class 0:
3.04 +
[Month] * -0.11 +
[Time] * 0 +
[Timegroup] * -0.58 +
[DewPoint] * -0 +
[humidity] * 0.12 +
[dest dewpoint] * -0.02 +
[destinationPressure] * -0.1
Class 1:
-3.04 +
[Month] * 0.11 +
[Time] * -0 +
[Timegroup] * 0.58 +
[DewPoint] * 0 +
[humidity] * -0.12 +
[dest dewpoint] * 0.02 +
[destinationPressure] * 0.1
```

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Time taken to build model: 0.14 seconds

=== Evaluation on test split ===

Time taken to test model on training split: 0 seconds

=== Summary ===

Correctly Classified Instances 69 66.9903 %
Incorrectly Classified Instances 34 33.0097 %
Kappa statistic 0.3512
Mean absolute error 0.4598

Total Number of Instances 103

Root mean squared error

Root relative squared error

Relative absolute error

=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class

0.4846

95.3704 %

90.9621 %

0.783 0.421 0.600 0.783 0.679 0.364 0.723 0.725 no 0.579 0.217 0.767 0.579 0.660 0.364 0.723 0.673 yes Weighted Avg. 0.670 0.308 0.693 0.670 0.669 0.364 0.723 0.696

=== Confusion Matrix ===

a b <-- classified as 36 10 | a = no 24 33 | b = yes

So, comparing the results from the classification model, we can see that using the features selected by Weka gives us a ~5% boost in the prediction accuracy. We shall be using the 7 features as suggested by Weka for the Simple Logistic Regression classification program which will be mentioned later.

Algorithm 2:- Naive Bayyes Classification

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Our second proposed classification algorithm is the Naive Bayyes algorithm. As mentioned above, we wanted to get the best features for our Naive Bayyes model and Weka provided us with the following suggestions.

=== Run information === Evaluator: weka.attributeSelection.ClassifierSubsetEval -B weka.classifiers.bayes.NaiveBayes -T -H "Click to set hold out or test instances" -E acc weka.attributeSelection.BestFirst -D 1 -N 5 Search: Relation: r-weka.filters.unsupervised.attribute.Remove-R14weka.filters.unsupervised.attribute.Remove-R10weka.filters.unsupervised.attribute.Remove-R11weka.filters.unsupervised.attribute.Remove-R2,5-6,9-12,15 Instances: 344 Attributes: 8 Month Time Timegroup **DewPoint** humidity destdewpoint destinationPressure Prediction Evaluation mode: evaluate on all training data

Search Method:

Best first.

Start set: no attributes Search direction: forward

=== Attribute Selection on all input data ===

Stale search after 5 node expansions Total number of subsets evaluated: 43 Merit of best subset found: 0.305

Class

no

yes

Attribute

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Attribute Subset Evaluator (supervised, Class (nominal): 8 Prediction): Classifier Subset Evaluator Learning scheme: weka.classifiers.bayes.NaiveBayes Scheme options: Hold out/test set: Training data Subset evaluation: classification error Selected attributes: 3,4,6:3 Timegroup **DewPoint** Destdewpoint So, our chosen attributes for the Naive Bayyes Classification Model are Timegroup, DewPoint and Destdewpoint. We wanted to run a test on the Naive Bayyes classification model on the selected attributes. The results are mentioned below. === Run information === Scheme: weka.classifiers.bayes.NaiveBayes Relation: r-weka.filters.unsupervised.attribute.Remove-R14weka.filters.unsupervised.attribute.Remove-R10weka.filters.unsupervised.attribute.Remove-R11weka.filters.unsupervised.attribute.Remove-R2,5-6,9-12,15weka.filters.unsupervised.attribute.Remove-R1-2,5,7 Instances: 344 Attributes: 4 Timegroup **DewPoint** destdewpoint Prediction Test mode: split 70.0% train, remainder test === Classifier model (full training set) === Naive Bayes Classifier

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(0.52) (0.48)
=======================================
Timegroup
mean 1.5363 1.9394
std. dev. 0.6789 0.5689
weight sum 179 165
precision 1 1
DewPoint
mean -5.8049 -5.4939
std. dev. 6.1858 6.314
weight sum 179 165
precision 0.6167 0.6167

destdewpoint

mean 13.1389 14.8783 std. dev. 7.6673 8.2039 weight sum 179 165 precision 0.678 0.678

Time taken to build model: 0 seconds

=== Evaluation on test split ===

Time taken to test model on training split: 0 seconds

=== Summary ===

Correctly Classified Instances	70	67.9612 %
Incorrectly Classified Instance	es 33	32.0388 %
Kappa statistic	0.3664	
Mean absolute error	0.4594	
Root mean squared error	0.4733	
Relative absolute error	90.8908 %	
Root relative squared error	93.1475 %	
Total Number of Instances	103	

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```
=== Detailed Accuracy By Class ===
         TP Rate FP Rate Precision Recall F-Measure MCC
                                                            ROC Area PRC
Area Class
         0.761
                0.386 0.614
                               0.761
                                      0.680
                                               0.375
                                                     0.719
                                                             0.730
                                                                     no
         0.614 0.239 0.761
                               0.614
                                      0.680
                                               0.375
                                                     0.719
                                                             0.670
                                                                     yes
Weighted Avg. 0.680 0.305 0.695
                                      0.680 0.680
                                                     0.375 0.719
                                                                   0.696
=== Confusion Matrix ===
 a b <-- classified as
35 \ 11 \mid a = no
```

Algorithm 3:- Support Vector Machine Classification

Our third proposed classification algorithm is the Support Vector Machine classification algorithm. We ran our dataset through Weka and wanted to come up with the best available features for our Support Vector Machine Classification Model. Results are displayed as below:-

```
Evaluator: weka.attributeSelection.ClassifierSubsetEval -B
weka.classifiers.functions.SMO -T -H "Click to set hold out or test instances" -E acc -
- -C 1.0 -L 0.001 -P 1.0E-12 -N 0 -V -1 -W 1 -K
"weka.classifiers.functions.supportVector.PolyKernel -E 1.0 -C 250007" -calibrator
"weka.classifiers.functions.Logistic -R 1.0E-8 -M -1 -num-decimal-places 4"
Search:
           weka.attributeSelection.BestFirst -D 1 -N 5
Relation:
           r-weka.filters.unsupervised.attribute.Remove-R14-
```

weka.filters.unsupervised.attribute.Remove-R10weka.filters.unsupervised.attribute.Remove-R11weka.filters.unsupervised.attribute.Remove-R2,5-6,9-12,15

Instances: 344 Attributes: 8 Month Time Timegroup

 $22\ 35 \mid b = yes$

=== Run information ===

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DewPoint humidity destdewpoint destinationPressure Prediction

Evaluation mode: evaluate on all training data

=== Attribute Selection on all input data ===

Search Method:

Best first.

Start set: no attributes Search direction: forward

Stale search after 5 node expansions
Total number of subsets evaluated: 31
Merit of best subset found: 0.308

Attribute Subset Evaluator (supervised, Class (nominal): 8 Prediction):

Classifier Subset Evaluator

Learning scheme: weka.classifiers.functions.SMO

Scheme options: -C 1.0 -L 0.001 -P 1.0E-12 -N 0 -V -1 -W 1 -K

weka.classifiers.functions.supportVector.PolyKernel -E 1.0 -C 250007 -calibrator weka.classifiers.functions.Logistic -R 1.0E-8 -M -1 -num-decimal-places 4

Hold out/test set: Training data

Subset evaluation: classification error

Selected attributes: 1,2,3,5,6,7 : 6

Month

Time

Timegroup humidity

destdewpoint

destinationPressure

The next step was to run the Support Vector Classifier on the dataset based on the selected attributes. The results are displayed as below:-

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```
=== Run information ===
            weka.classifiers.functions.SMO -C 1.0 -L 0.001 -P 1.0E-12 -N 0 -V -1 -
Scheme:
W 1 -K "weka.classifiers.functions.supportVector.PolyKernel -E 1.0 -C 250007" -
calibrator "weka.classifiers.functions.Logistic -R 1.0E-8 -M -1 -num-decimal-places
4"
Relation:
           r-weka.filters.unsupervised.attribute.Remove-R14-
weka.filters.unsupervised.attribute.Remove-R10-
weka.filters.unsupervised.attribute.Remove-R11-
weka.filters.unsupervised.attribute.Remove-R2,5-6,9-12,15-
weka.filters.unsupervised.attribute.Remove-R4
Instances: 344
Attributes: 7
        Month
        Time
        Timegroup
        humidity
        destdewpoint
         destinationPressure
        Prediction
Test mode: split 70.0% train, remainder test
=== Classifier model (full training set) ===
SMO
Kernel used:
 Linear Kernel: K(x,y) = \langle x,y \rangle
Classifier for classes: no, yes
BinarySMO
Machine linear: showing attribute weights, not support vectors.
     0.3218 * (normalized) Month
     0.0813 * (normalized) Time
+
     2.8137 * (normalized) Timegroup
+
     -0.1539 * (normalized) humidity
```

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- + 1.1437 * (normalized) destdewpoint
- + 0.9435 * (normalized) destinationPressure
- 2.6875

Number of kernel evaluations: 10243 (73.184% cached)

Time taken to build model: 0.03 seconds

=== Evaluation on test split ===

Time taken to test model on training split: 0 seconds

=== Summary ===

Correctly Classified Instances 69 66.9903 % Incorrectly Classified Instances 34 33.0097 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.3405

0.3301

0.5745

65.3019 %

113.082 %

=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class

0.696 0.351 0.615 0.696 0.653 0.343 0.672 0.564 no 0.649 0.304 0.725 0.685 0.672 0.649 0.343 0.665 yes 0.343 0.672 Weighted Avg. 0.670 0.325 0.676 0.670 0.671 0.620

=== Confusion Matrix ===

a b <-- classified as

 $32 \ 14 \mid a = no$

 $20\ 37 \mid b = yes$

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Step 4:- Choose and Apply a Learning Algorithm

The results of the classification algorithms are mentioned below. Each of the algorithms were run on 70% training set and 30% test set in Weka.

Algorithm	Accuracy (%)	Error (%)	Completion Time
Simple Logistic	66.9903	32.0388	0.14 seconds
Regression			
Naive Bayyes	67.9612	32.0388	0 seconds
Support Vector	66.9903	33.0097	0.03 seconds
Machines			

The above results suggest that Naive Bayyes is our best but we wanted to have an open mind and create a prediction program for all the three algorithms. Our preferred programming language is Java and we are building a simple classifier program for the above algorithms.

We chose java and decided to come up with desktop software for now. Java has a write once, run everywhere policy. This makes our program platform independent. Our preferred IDE is eclipse.

The code for each of the above algorithms is in a separate folder in the GITHUB account of the team and also is attached below at the end of the paper.

Conclusion:-

The aim of this research was to answer the question if an app can be created to predict flight delays in advance based on weather related issues. Based on our above research, we can firmly say that Yes, an app can be created to predict flight delays based on weather data of the past. We applied Machine learning and data mining for this project. Machine learning is one of the fastest growing areas of research in our modern era and is being implemented by some of the top notch companies namely- Google, Microsoft, Facebook, NASA.

We wanted to go a step further and come up with a way to estimate the time of delay along with the prediction (yes/no). To achieve that we first identify if a flight is likely to be delayed based on the weather data. After that we wanted to apply K-means clustering to group together data of the similar type of that specific weather data. We look at the amount of time in departure delays in the dataset of the previous data. Then we take the mean of the

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departure delay times of the group formed by K-means clustering and suggest it as an approximate time of delay for that particular date.

Java code for Naive Bayes Classifier algorithm

Jar files required – weka.jar

```
import java.io.FileNotFoundException;
import java.io.FileReader;
import java.io.BufferedReader;
import java.io.IOException;
import java.util.Random;
import weka.classifiers.evaluation.*;
import weka.classifiers.bayes.NaiveBayes;
import weka.classifiers.bayes.net.*;
import weka.core.Instance;
import weka.core.Instances;
import weka.core.converters.ConverterUtils;
public class JavaWeka {
      public static void main(String[] args) {
             // TODO Auto-generated method stub
               ConverterUtils.DataSource source1;
                   try {
                          source1 = new
ConverterUtils.DataSource("traindata.arff");
                           Instances train = source1.getDataSet();
                            // setting class attribute if the data format does not
provide this information
                            // For example, the XRFF format saves the class
attribute information as well
                            if (train.classIndex() == -1)
                                train.setClassIndex(train.numAttributes() - 1);
                            ConverterUtils.DataSource source2 = new
ConverterUtils.DataSource("testdata.arff");
                            Instances test = source2.getDataSet();
                            // setting class attribute if the data format does not
provide this information
                            // For example, the XRFF format saves the class
attribute information as well
                            if (test.classIndex() == -1)
                                test.setClassIndex(train.numAttributes() - 1);
                            NaiveBayes naiveBayes = new NaiveBayes();
                            naiveBayes.buildClassifier(train);
```

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```
Evaluation eval = new Evaluation(train);
                           eval.evaluateModel(naiveBayes,test);
System.out.println(eval.toSummaryString("\nResults\n###\n",true));
                           System.out.println(eval.fMeasure(1)+ "" +
eval.precision(1)+ "" +eval.recall(1));
                   } catch (Exception e) {
                          // TODO Auto-generated catch block
                          e.printStackTrace();
                   }
      }
}
Output-
Results
####
                                                                %
Correctly Classified Instances
                                       65
                                                        62.5
Incorrectly Classified Instances
                                       39
                                                        37.5
                                                                %
Kappa statistic
                                        0.245
K&B Relative Info Score
                                     1013.5341 %
                                      10.1229 bits
K&B Information Score
                                                         0.0973 bits/instance
Class complexity | order 0
                                      103.8896 bits
                                                         0.9989 bits/instance
Class complexity | scheme
                                      101.6568 bits
                                                         0.9775 bits/instance
                                        2.2328 bits
Complexity improvement
                                                         0.0215 bits/instance
Mean absolute error
                                        0.4585
Root mean squared error
                                        0.4909
Relative absolute error
                                       91.8479 %
                                       98.2601 %
Root relative squared error
Total Number of Instances
0.58064516129032260.6279069767441860.54
```

Java code for Support Vector Machine

Jar files required – JAVAML, WEKA, LibSVM

```
import java.io.File;
import java.io.IOException;

import libsvm.LibSVM;
import net.sf.javaml.classification.Classifier;
import net.sf.javaml.core.Dataset;
```

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```
import net.sf.javaml.core.Instance;
import net.sf.javaml.tools.data.FileHandler;
public class SVMweka {
      public static void main(String[] args) throws IOException {
             // TODO Auto-generated method stub
              Dataset data = FileHandler.LoadDataset(new
File("traindata.csv"),3,",");
               * Contruct a LibSVM classifier with default settings.
              Classifier svm = new LibSVM();
              svm.buildClassifier(data);
               * Load a data set, this can be a different one, but we will use
the
               * same one.
               */
              Dataset dataForClassification = FileHandler.LoadDataset(new
File("testdata.csv"),3,",");
              /* Counters for correct and wrong predictions. */
              int correct = 0, wrong = 0;
              /* Classify all instances and check with the correct class values
*/
              for (Instance inst : dataForClassification) {
                  Object predictedClassValue = svm.classify(inst);
                  Object realClassValue = inst.classValue();
                   if (predictedClassValue.equals(realClassValue))
                       correct++;
                  else
                      wrong++;
              System.out.println("Correct predictions " + correct);
              System.out.println("Wrong predictions " + wrong);
      }
}
      Output:-
```

Correct predictions 72 Wrong predictions 33

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Java code for Simple Logistic Regression

```
import java.io.BufferedReader;
import java.io.File;
import java.io.FileNotFoundException;
import java.io.FileReader;
import java.io.IOException;
import java.util.ArrayList;
import java.util.Arrays;
import java.util.List;
import java.util.Scanner;
/**
public class SLR {
      /** the learning rate */
      private double rate;
      /** the weight to learn */
      private double[] weights;
      /** the number of iterations */
      private int ITERATIONS = 3000;
      public SLR(int n) {
             this.rate = 0.0001;
```

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```
weights = new double[n];
             }
             private static double sigmoid(double z) {
                    return 1.0 / (1.0 + Math.exp(-z));
             }
             public void train(List<Instance> instances) {
                    for (int n=0; n<ITERATIONS; n++) {</pre>
                           double lik = 0.0;
                           for (int i=0; i<instances.size(); i++) {</pre>
                                  double[] x = instances.get(i).x;
                                  double predicted = classify(x);
                                  int label = instances.get(i).label;
                                  for (int j=0; j<weights.length; j++) {</pre>
                                         weights[j] = weights[j] + rate * (label -
predicted) * x[j];
                                  }
                                  // not necessary for learning
                                  lik += label * Math.log(classify(x)) + (1-label)
* Math.log(1- classify(x));
                           }
                           System.out.println("iteration: " + n + " " +
Arrays.toString(weights) + " mle: " + lik);
                    }
             }
             private double classify(double[] x) {
```

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double logit = .0;

```
for (int i=0; i<weights.length;i++) {</pre>
                          logit += weights[i] * x[i];
                    }
                    return sigmoid(logit);
             }
             public static class Instance {
                    public int label;
                    public double[] x;
                    public Instance(int label, double[] x) {
                          this.label = label;
                          this.x = x;
                    }
             }
             public static List<Instance> readDataSet(String file) throws
FileNotFoundException {
                    List<Instance> dataset = new ArrayList<Instance>();
                    Scanner scanner = null;
                    try {
                          scanner = new Scanner(new File(file));
                          while(scanner.hasNextLine()) {
                                 String line = scanner.nextLine();
                                 String[] columns = line.split("\\s+");
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```
// skip first column and last column is the label
                                  int i = 1;
                                  double[] data = new double[columns.length-2];
                                  for (i=1; i<columns.length-2; i++) {</pre>
                                        data[i-1] =
Double.parseDouble(columns[i]);
                                  }
                                  int label = Integer.parseInt(columns[i]);
                                  Instance instance = new Instance(label, data);
                                  dataset.add(instance);
                           }
                    } finally {
                           if (scanner != null)
                                  scanner.close();
                    }
                    return dataset;
             }
             public static void main(String... args) throws FileNotFoundException
{
                    List<Instance> instances = readDataSet("editedata.txt");
                    SLR logistic = new SLR(3);
                    logistic.train(instances);
                    String month, day, time, origin, destination;
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```
Scanner input = new Scanner(System.in);
                    System.out.println("Enter the month of travel");
                   month = input.nextLine();
                    System.out.println("Enter the day of travel");
                    day = input.nextLine();
                   System.out.println("Enter the time of travel");
                   time = input.nextLine();
                    System.out.println("Enter the origin airport code of travel");
                    origin = input.nextLine();
                    System.out.println("Enter the destination airport code of
travel");
                    destination = input.nextLine();
                   String csvFile = "fdata.txt";
                    BufferedReader br = null;
                   String line = "";
                    String cvsSplitBy = ",";
                   try {
```

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```
Scanner x = new Scanner(new File(csvFile));
                         while(x.hasNext())
                           {
                                 String
month1,day1,time1,timegroup,airportid,temperature,dewpoint,humidity,windspeed,dest
inationairportid;
                                 month1 = x.next();
                                 day1 = x.next();
                                 time1 = x.next();
                                 timegroup = x.next();
                            airportid = x.next();
                               temperature = x.next();
                               dewpoint = x.next();
                               humidity = x.next();
                               windspeed = x.next();
                               destinationairportid = x.next();
                      if(month.compareTo(month1)==0)
                      {
                       if(day.compareTo(day1)==0)
                       {
                              if(time.compareTo(time1)==0)
                              {
                              if(origin.compareTo(airportid)==0)
                              {
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```
if(destination.compareTo(destinationairportid)==0)
                                    {
                                           System.out.println(month1 + "\t" + day1
+"\t"+ time1 + "\t"+ timegroup+"\t"+ airportid+"\t"+temperature
+"\t"+dewpoint+"\t"+humidity+
      "\t"+windspeed+"\t"+destinationairportid);
                                           int t1 = Integer.parseInt(timegroup);
                                           int m1 = Integer.parseInt(month1);
                                           Double tmp =
Double.parseDouble(temperature);
                                           System.out.println(t1 + "" + m1 + " " +
tmp);
                                           double[] z = \{m1, tmp, t1\};
                                               System.out.println("Done");
                                               System.out.println("prob(1|x) = " +
logistic.classify(z));
                                            double res = logistic.classify(z);
                                            if(res>0.5)
                                            {
                                               System.out.println("Flight might
delay");
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```
}
                                            else
                                              System.out.println("Flight is on
time");
                                    }
                             }
                             }
                      }
                      }
                          }
                          x.close();
                   }
                          catch (FileNotFoundException e) {
```

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```
e.printStackTrace();
                    } catch (IOException e) {
                           e.printStackTrace();
                    } finally {
                           if (br != null) {
                                 try {
                                        br.close();
                                 } catch (IOException e) {
                                        e.printStackTrace();
                                 }
                           }
                    }
                    System.out.println("Done");
}
}
```

```
Output from the Simple Logistic Regression program

Enter the month of travel

4

Enter the day of travel

7

Enter the time of travel
```

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```
1552
Enter the origin airport code of travel
10140
Enter the destination airport code of travel
11259
4 7 1552 1 10140 23.3 -11.1 0.09 7.2
                                                     11259
14 23.3
Done
prob(1|x) = 0.9606277435328205
Flight might delay***
Done
Enter the month of travel
Enter the day of travel
3
Enter the time of travel
852
Enter the origin airport code of travel
10140
Enter the destination airport code of travel
11259
4 3 852 1 10140 6.1 -1.7 0.58 3.6
                                                      11259
14 6.1
```

Done

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prob(1|x) = 0.02986981914996907

Flight is on time***

References:- https://github.com/tpeng/logistic-regression