

Optimal Solution of Classification (Prediction) Problem

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Summary

Classification or prediction problem is how to solve it using a specific feature to obtain the predicted class. A wheat seeds specifications 4 3 classes of seeds will be used in a prediction process. A multi linear regression will be built, and a prediction error ratio will be calculated. To enhance the prediction ratio an ANN model will be built and trained. The obtained results will be examined to show how to make a prediction tool capable to compute a predicted class number very close to the target class number.

Keywords:

Classification, prediction, ANN, FFANN, training, MLR, prediction error ratio, features, regression coefficients, regression equation .

1. Introduction

As you Classification is the study of methods that are used to categorize data based on distinct classes. We might call classes targets, labels, or categories. Categorization can be done by three or more methods. These methods can be listed as follows. Distinct labeling of data or what is known as supervised learning. Data division into classes as in unsupervised learning. Distinguished features selection, and combinations of these stated methods

Based on the input of features or patterns [4], [5], [6]. The classifier can be used to distinguish the inputted data by performing needed actions to output a predicted class. Thereafter it's possible to use the calculated classifier to implement an action [7], [8], [9]. In classification without the label, the data is inputted to the model, the model should return a class in a specific place [10], [11]. There are two forms of data mining that can be used to extract models. And these forms are classification descriptions and predication. To predict future data direction, the following is of concern:-

- 1- Find the missing elements in the datasheet.
- 2- Predicting the outcome by the classification model.
- 3- No dependence on the label of the class
- 4- Predication is based on both the label and the class model.

1.1 Classification by Regression

In general, classification involves the prediction of labeling, while regression involves the prediction of

quantity. Multiple Linear Regression is an estimator for the relationship between two independent variables or several variables, used as an input, and a single dependent variable used as an output. Input variable can be categorical, which contain a finite number of categories or distinct groups. Also, the input variables could be continuous variables.

To describe relationship between variables, we usually fit a line in the observed data. This will allow us to estimate how the output variable changes in relation to the inputted variables. To perform multiple line regression, we must use a formula as shown in figure (1) where y is the observed dependent variable [4], [5].

$$Y = a_0 + a_1x_1 + a_2x_2 + \dots + a_px_p + \epsilon$$

Figure 1: MLR model

To find a class value using MLR we have to follow up the following phases (see figure 2)

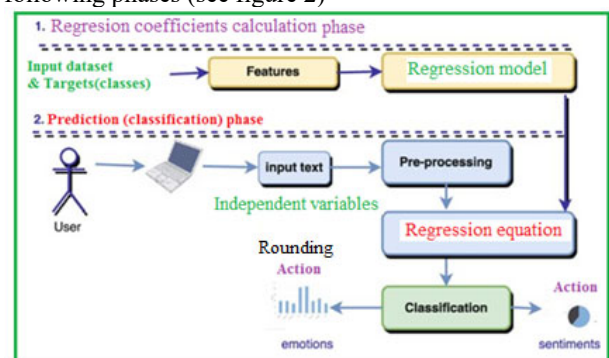


Figure 2: Classification using MLR model

Phase 1:

Use the input dataset(features) to build MLR model, then apply this model to obtain the regression coefficients.

Phase 2:

The following steps were used to implement this phase:

- Obtain the features values(independent variables values).

- Use the regression coefficients to build the regression equation.
- Apply the regression equation to find the class value.
- Round the result to get the class value.

1.2 Classification using artificial neural networks

Artificial neural network (ANN) [16], [17] is a powerful computational model used in various fetal applications such as curve fitting (regression analysis), pattern recognition (classification), clustering and time series. One variant of ANN is a feed forward ANN (FFANN) [18], [19]. FFANN is a set of fully connected neurons, arranged in layers as shown in figure 3, each neuron acts as a computational cell and performs two main functions as shown in figure 4. The first function is the summation of products of the inputs and the associated weights, the second function is computing the neuron outputs depending on the activation function (logsig, tansig or linear) selected for the neuron within a specified layer [20], [21].

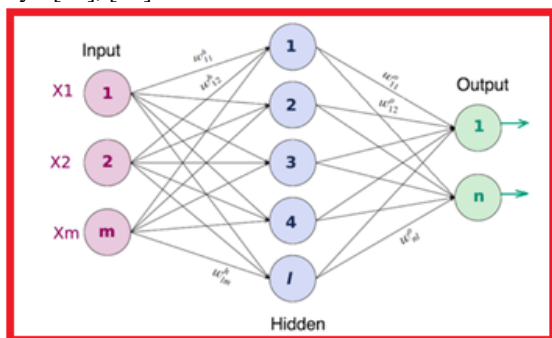


Figure 3: FFANN architecture example

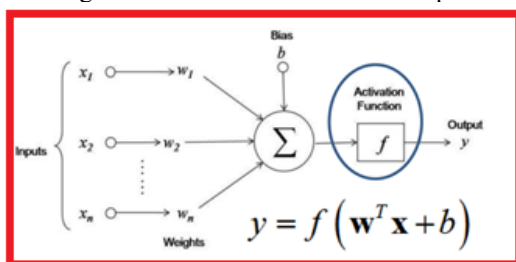


Figure 4: Neuron operations

To use FFANN as a classification tool we have to follow up the following procedures (see figure 5) [22], [23], [24]:

Phase 1: Training

In this phase we have to apply the following steps:

1. Select the input dataset (features) and targets (classes values), if required normalize the data (preprocessing).
2. Create FFANN (select FFANN architecture) by the definition of
 - a) number of layers,

- b) Number of neurons in each individual layer
- c) The function of activation for all layers.
3. Initialize FFANN.
4. Define some parameters for the net, such as the goal (the error between the target and the calculated output must equal or closed to zero), the number of training cycles (epochs).
 5. Train the net using the features and classes.
 6. Check the error value, if the error is acceptable save the net to be used later as a recognition tool, else increase the number of training cycle or adjust the net architecture and train it again.

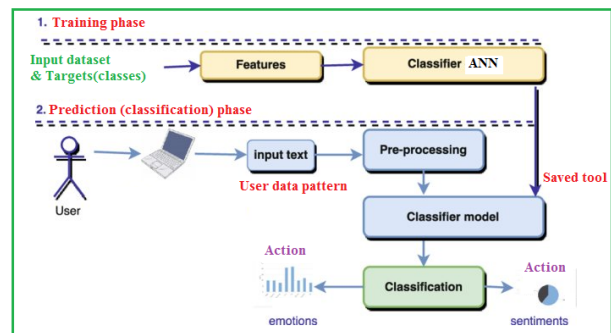


Figure 5: Using ANN as a classifier

Each training cycle contains two phases as shown in figures 6 and 7. The feedforward phase by calculating the neurons outputs starting from the input layer, and backward phase starting from the output layer to adjust the neurons weight.

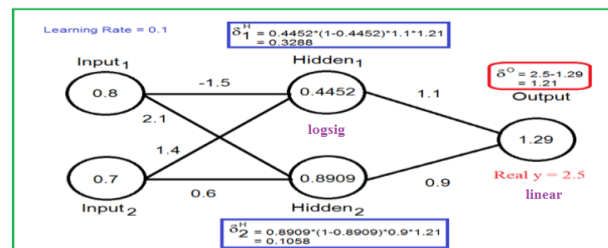


Figure 6: Calculating the outputs (Forward phase)

2. Implementation and experimental results

The wheat seeds dataset was used as an input to the created models to generate classification. When the dataset for the seeds includes the prediction of species. The measurement of seeds was given from different kinds of wheat. This was a two-class classification binary case. The observation for the individual classes was balanced. There exist 210 observations with seven input variables and one output variable. The variables names are as follows:

- 1- Area
- 2- Compactness
- 3- Perimeter
- 4- Length of kernel
- 5- Width of kernel
- 6- Asymmetric Coefficient
- 7- Length kernel groove
- 8- Class (1,2,3).

Nine samples are shown below in figure 7.

Word or equivalent Word Processors and justify to block. The heading of each section should be printed in small, 12pt, left justified, bold, serif. You must use the Arabic numbers 1, 2, 3, for the sections numbering and not the Roman numbers (I, II, III). Please, follow the paragraph indentation that is used in this template.

X1	Area	15.2600	14.8800	14.2900	13.8400	16.1400	14.3800	14.6900	14.1100	16.6300
X2	Perimeter	14.8400	14.5700	14.0900	13.9400	14.9900	14.2100	14.4900	14.1000	15.4600
X3	Compactness	0.8710	0.8811	0.9050	0.8955	0.9034	0.8951	0.8799	0.8911	0.8747
X4	Length of kernel	5.7630	5.5540	5.2910	5.3240	5.6580	5.3860	5.5630	5.4200	6.0530
X5	Width of kernel	3.3120	3.3330	3.3370	3.3790	3.5620	3.3120	3.2590	3.3020	3.4650
X6	Asymmetry coefficient	2.2210	1.0180	2.6990	2.2590	1.3550	2.4620	3.5860	2.7000	2.0400
X7	Length of kernel groove	5.2200	4.9560	4.8250	4.8050	5.1750	4.9560	5.2190	5.0000	5.8770

Figure 7: Some samples of the input dataset

First we apply MLR , the coefficients of the regression outputs are shown in table 1:

Table 1: Regression coefficients

Coefficient	Value
a0	53.4436
a1	1.4891
a2	-3.2204
a3	-30.6774
a4	-2.3151
a5	0.2460
a6	0.1149
a7	2.1926

Applying the regression equation using the obtained regression coefficients we found the predicted values of the classes, rounding these values we can obtain the predicted class number for any given values of the independent variables. From the used 210 samples, 36 classes were wrong calculated, with prediction error ratio equal 17.14 %., figure 8 shows the target and the predicted classes, while figure 9 shows the error between them (before rounding).

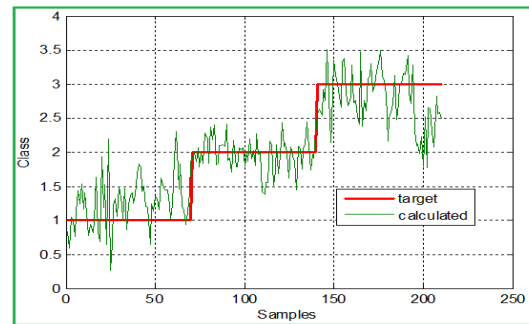


Figure 8: Targets and predicted classes (regression model)

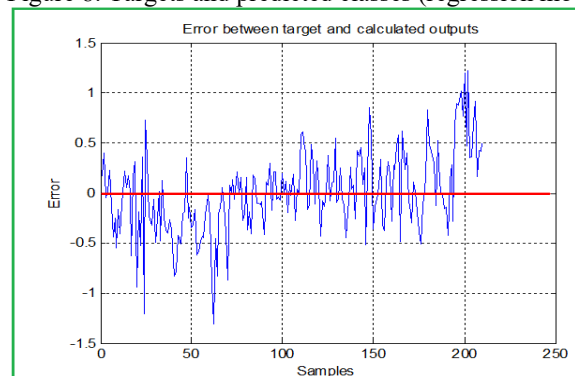


Figure 9: Error between the target and predicted classes(regression model)

Second we apply ANN mode:

ANN was created using the following features:

- Number of inputs 7.
- Number of outputs 1.
- ANN contains 2 layers, the first layer with 7 neurons and tansig activation function, the second layer with 1 neuron and linear activation function.
- The goal (error) was set to zero.
- The number of training cycles was set to 2000.

ANN was trained using the same data set and targets, the trained ANN was used to predict the class number using a specific features, here the max error obtained was equal $1.7979e-005$, which means that the prediction ratio is very closed to 100%, figure 10 shows the target and the predicted classes, while figure 11 shows the error between them (before rounding).

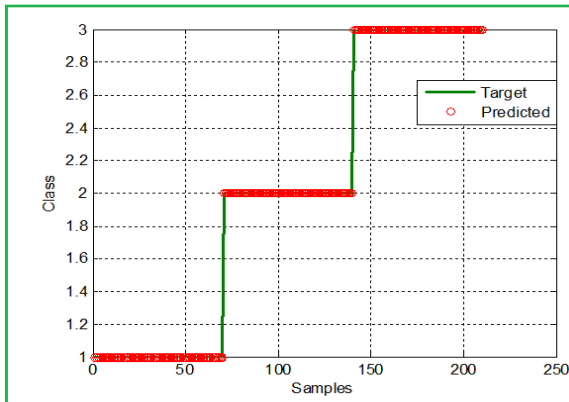


Figure 10: Targets and predicted classes(ANN model)

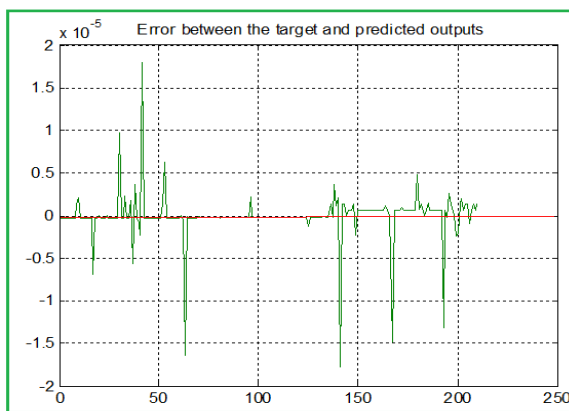


Figure 11: Error between the target and predicted classes(ANN model)

From the obtained experimental results, we can raise the following facts:

- MLR model is restricted to a specific prediction problem.
- Using MLR model for prediction causes a high regression error ratio.
- To rapidly decrease the prediction error ratio, we recommend using ANN model.
- Building ANN model is a very simple process, this model can be easily updated to suit any input data set with any target classes.

3. Conclusion

A wheat seeds dataset with various specifications for 3 classes of wheats were used as an input data set for a classification problem, a MLR model was built and implemented to calculate the wheat class number depending on a selected specification, the obtained results was acceptable but with prediction error equal 17.14 %. To minimize the prediction error ration, and to obtain a predicted class number closed to the target class, we recommend using ANN model. The ANN model decreases

rapidly the prediction error, it is very simple to build and update.

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