A STATISTICAL COMPARISON BETWEEN GENETIC ALGORITHM AND LOGISTIC REGRESSION FOR A CLINICAL STUDY

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DECLARATION

I declare that this is my own work and this dissertation does not incorporate without acknowledgement any material previously submitted for a Degree or Diploma in any other University or institute of higher learning and to the best of my knowledge and belief it does not contain any material previously published or written by another person except where the acknowledgement is made in the text.

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ABSTRACT

A Statistical Comparison between Genetic Algorithm and Logistic Regression for a Clinical Study

Identifying a combination of variables causing infections or infectious diseases is one of the main tasks in clinical models in medicine. Forward and backward variable selection techniques in Logistic Regression (LR) are widely used in such situations, where it assumes linearity of independent variables and the absence of multi-collinearity. More often, the observed data do not satisfy these assumptions and thus, LR is not applicable. Hence, the Genetic Algorithm (GA), which does not depend on pre-defined assumptions, has proven to be better under such circumstances. By evaluating prediction rates of LR and GA techniques, the objective of this study was to perform binary LR and GA to reduce the number of variables on a sample of clinical data and compare the goodness of fit statistics to identify the better variable reduction method. Three models were built using 40 independent variables (3 non-categorical and 37 categorical) for a sample of 497 observations collected from suspected respiratory syncytial virus (RSV) infected children under 5 years of age, who were hospitalized to the Kegalle Base Hospital from May 2016 to July 2018. The binary dependent variable indicates whether the suspected child is infected with RSV positive or negative. Log-likelihood and Area Under Curve (AUC) represent the fitness functions of two GAs. The goodness of fits on the three models was compared using statistical measurements: -2log-likelihood, Psudo R-square values, Correctly Classified Percentage, Specificity, and Sensitivity. Results shown that Log-likelihood GA produces better goodness of fit measurements compared to other the two methods. However, LR reduces 40 variables into 8 with lower number of iterations while two GAs reduced into 17 variables to predict the status of RSV infection. This study suggests that the LR has a better prediction power with the most associated combination of variables. However, GA indicated better in analysing when the predefined assumptions were not satisfied and solving high dimensional classification problems in a large or complex searching space in the background of the study.

Keywords: Clinical Data, Fitness Function, Genetic Algorithm, Logistic Regression

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LIST OF ABBREVIATIONS

AIC Akaike Information Criterion

ARIMA Auto-Regressive Integrated Moving Average

ARTI Acute Respiratory Tract Infections

AUC Area Under the Curve

AUC-ROC Area Under the Receiver Operating Characteristics Curve

BIC Bayesian Information Criterion

DIB Difficulty in Breathing

FDA Fisher Discriminant Analysis

GA Genetic Algorithm

GARI Genetic Algorithm Rainfall Intensity

GARS Genetic Algorithm for Regressors Selection

GARST Genetic Algorithm for Regressors Selection with the

Transformation

LSM Least Square Method

ROC Receiver Operating Characteristics Curve

RSV Respiratory Syncytial Virus

SIC Schwarz Information Criterion

SOB Shortness of Breath

VIF Variance Inflation Factor