UNIVERSITY OF MORATUWA

Predictive Model for Gap Reduction Between Web Analytics and Business Strategy

by

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Declaration

| I declare that this is my own work and contains no material that has been published previously in whole or in part for the fulfillment of any degree program. All the referenced materials have been acknowledged in text. | | | | | |
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Student:

Supervisor:

Abstract

Digital marketing and web analytics are two distinct areas that have captured the attention of many industrial firms. There are a lot of tools developed and a lot of studies carried out in each area separately. But still, a firms ability to harness web analytics to optimize digital marketing elements is limited. This work focuses on evaluating previous work in each of these areas and combine them to build a model that would define the relationship between digital marketing and web analytics. Data captured through each area is expected to be analyzed in the form of a time series forecasting problem. Time series forecasting is a very popular area that captured a lot of firms attention in recent years. This is due to the fact that most real-world problems are linked to a temporal component, and thus can be considered as a time series. Furthermore, this work utilizes cloud services for building and running the learning models.

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Abbreviations

ANN Artificial Neural Network

ARMA Auto Regressive Moving Average

ARIMA Auto Regressive Integrated Moving Average

AWS Amazon Web Services

Deep Auto Regressive

GCP Google Cloud Platform

GUI Graphical User Interface

IaaS Infrastructure as a Service

K-NN K-Nearest Neighbour

LSTM Long Short-Term Memory

MAD Mean Absolute Deviation

MLMVN Multi Layer Multi Valued Neurons

MSE Mean Squared Error

PaaS Platform as a Service

QRF Quantile Random Forest

RNN Recurrent Neural Network

Simple Storage Service

SDK Software Development Kit

TS Time Series