"If we meticulously keep on working upon the known things, Allah will grant us knowledge of the unknown things."

----Abba

To my Father, a gem of a person, a true philanthropist, and above all a very affectionate Father.

This thesis is dedicated to my father, who have encouraged me to do Ph.D. and always inspired me to be honest and not to take hardships seriously and keep going.

#### DECLARATION BY THE CANDIDATE

The candidate certifies that the thesis entitled "Near-Infrared based Solutions for Quality Assessment during Manufacturing and Storage of a Ready to Eat Rice" submitted to the School of Engineering, Tezpur University in partial fulfillment for the award of the degree of Doctor of Philosophy in the Department of Food Engineering and Technology is a record of research work carried out by me under the supervision of Prof. Manuj Kumar Hazarika.

All assistance received from various sources has been appropriately acknowledged. No part of this thesis has been submitted elsewhere for the award of any degree.

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This is to certify that the thesis entitled Near Infrared based Solutions for Quality Assessment during Manufacturing and Storage of a Ready to Eat Rice submitted to the School of Engineering, Tezpur University in part fulfillment for the award of the degree of Doctor of Philosophy in Food Engineering and Technology is a record of research work carried out by Ms. Shagufta Rizwana under my supervision and guidance.

All help received by her from various sources has been duly acknowledged.

No part of this thesis has been submitted elsewhere for award of any other degree.

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# List of notations

$ ho_b$	Bulk density
$ ho_t$	True density
M	Moisture
$k_1$	Peleg's rate constant
$k_2$	Peleg's capacity constant
$M_i$	Initial moisture content
$M_e$	Equilibrium moisture content
MR	Moisture ratio
$C_A$	Volumetric concentration
$C_{As}$	Volumetric concentration in saturation
$C_{A0}$	Initial volumetric concentration
β	Rate constant estimated using experimental data
$ ho_{\scriptscriptstyle W}$	Volume of pure water
$ ho_s$	Volume of pure solid
$d_p$	Equivalent diameter
$D_{ ho_A}$	Diffusion coefficient
κ	Arbitrary constants used in Fick's diffusion
$k_f$	Arbitrary constants used in Fick's diffusion
k	First order rate constant
$D_G$	Degree of gelatinization

<sup>\*\*</sup> A few arbitrary constants for the empirical drying model are kept as same as literature and those have been reported accordingly.

#### List of abbreviations

ANFIS Adaptive Neuro Fuzzy Inference System

ANP Aged Non-Parboiled AP Aged Parboiled

ANN Artificial Neural Network BP-ANN Back propagation-ANN

BiPLS Backward PLS

CFTRI Central Food Technological Research Institute

CART Classification and Regression Tree
CNN Convolutional Neural Network
DSC Differential Scanning Calorimetry

db Dry basis

EMC Equilibrium moisture content EDM Euclidean distance measure

FiPLS Forward PLS
GI Glycemic index
KNN K-Nearest Neighbors

LS-SVM Least-squares support vector machines

LDA Linear Discriminant Analysis

LR Logistic Regression
LSTM Long short-term memory
ML Machine Learning
m.c. Moisture content

MLP-ANN Multilayer Perceptron ANN

NB Naïve Bayes NIR Near Infrared

NIRS Near Infrared Spectroscopy

PLS Partial Least Square PLSR PLS regression

PCA Principal component analysis
PAT Process Analytical Technology

RF Random Forest
RVA Rapid Visco Analysis
RNN Recurrent Neural Network

SIMCA Soft Independent Modeling Class Analogy

SAM Spectral angle measure
SCM Spectral correlation measure
SVM Support Vector Machines
UANP Unaged Non-Parboiled
UP Unaged Parboiled

UVE-SPA-LS-SVM Uninformative variable elimination (UVE) and successive

projections algorithm (SPA) LA-SVM

wb Wet basis