

**ENHANCED ARTIFICIAL BEE COLONY-LEAST SQUARES
SUPPORT VECTOR MACHINES ALGORITHM FOR TIME SERIES
PREDICTION**

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Abstrak

Sejak beberapa dekad yang lalu, Mesin Sokongan Vektor Kuasa Dua Terkecil (LSSVM) telah digunakan secara meluas dalam masalah ramalan di pelbagai domain aplikasi. Walaubagaimanapun, literatur sedia ada menunjukkan keupayaan LSSVM bergantung kepada nilai parameter hiper, iaitu parameter regularisasi dan parameter kernel, di mana iaanya akan mempengaruhi generalisasi LSSVM dalam tugas ramalan. Kajian ini mencadangkan algoritma hibrid, berdasarkan Koloni Lebah Buatan (ABC) dan LSSVM yang terdiri dari tiga algoritma; ABC-LSSVM, *lvABC*-LSSVM and *cmABC*-LSSVM. Algoritma *lvABC* diperkenalkan untuk mengatasi masalah minimum setempat dengan menambah baik proses carian menggunakan mutasi Levy. Dalam pada itu, algoritma *cmABC* yang menggunakan mutasi konvensional dapat mengatasi masalah penyesuaian terlebih atau penyesuaian terkurang. Kombinasi algoritma *lvABC* dan *cmABC*, yang kemudiannya dikenali sebagai algoritma Koloni Lebah Buatan Dipertingkat-Mesin Sokongan Vektor Kuasa Dua Terkecil (*eABC*-LSSVM) telah direalisasikan pada ramalan harga komoditi sumber asli yang tidak boleh diperbaharui. Setelah tugas pengumpulan data dan pra pemprosesan data siap dilakukan, algoritma *eABC*-LSSVM direkabentuk dan dibangunkan. Keupayaan *eABC*-LSSVM dinilai berdasarkan lima metrik statistik, iaitu Min Peratusan Ralat Mutlak (MAPE), ramalan ketepatan, simetri Min Peratusan Ralat Mutlak (sMAPE), Peratusan Ralat Punca Kuasa Min (RMSPE) dan Theil's U. Keputusan menunjukkan *eABC*-LSSVM mempunyai kadar ralat ramalan yang lebih rendah berbanding dengan lapan model hibrid antara LSSVM dan algoritma Evolusi Pengkomputan (EC). Tambahan pula, algoritma yang dicadangkan juga telah dibandingkan dengan teknik ramalan tunggal iaitu Mesin Sokongan Vektor (SVM) dan Rangkaian Neural dengan Rambatan ke Belakang (BPNN). Secara umumnya, *eABC*-LSSVM telah menghasilkan ramalan ketepatan melebihi 90%. Ini menunjukkan *eABC*-LSSVM berkeupayaan dalam menyelesaikan masalah optimisasi terutamanya dalam bidang ramalan. Algoritma *eABC*-LSSVM dijangka dapat memberi manfaat kepada para pelabur dan pedagang komoditi dalam perancangan pelaburan dan pengunjuran keuntungan.

Kata kunci: Koloni Lebah Buatan, Mesin Sokongan Vektor Kuasa Dua Terkecil, Ramalan Siri Masa

Abstract

Over the past decades, the Least Squares Support Vector Machines (LSSVM) has been widely utilized in prediction task of various application domains. Nevertheless, existing literature showed that the capability of LSSVM is highly dependent on the value of its hyper-parameters, namely regularization parameter and kernel parameter, where this would greatly affect the generalization of LSSVM in prediction task. This study proposed a hybrid algorithm, based on Artificial Bee Colony (ABC) and LSSVM, that consists of three algorithms; ABC-LSSVM, *lv*ABC-LSSVM and *cm*ABC-LSSVM. The *lv*ABC algorithm is introduced to overcome the local optima problem by enriching the searching behaviour using Levy mutation. On the other hand, the *cm*ABC algorithm that incorporates conventional mutation addresses the over-fitting or under-fitting problem. The combination of *lv*ABC and *cm*ABC algorithm, which is later introduced as Enhanced Artificial Bee Colony – Least Squares Support Vector Machine (*e*ABC-LSSVM), is realized in prediction of non renewable natural resources commodity price. Upon the completion of data collection and data pre processing, the *e*ABC-LSSVM algorithm is designed and developed. The predictability of *e*ABC-LSSVM is measured based on five statistical metrics which include Mean Absolute Percentage Error (MAPE), prediction accuracy, symmetric MAPE (sMAPE), Root Mean Square Percentage Error (RMSPE) and Theils' U. Results showed that the *e*ABC-LSSVM possess lower prediction error rate as compared to eight hybridization models of LSSVM and Evolutionary Computation (EC) algorithms. In addition, the proposed algorithm is compared to single prediction techniques, namely, Support Vector Machines (SVM) and Back Propagation Neural Network (BPNN). In general, the *e*ABC-LSSVM produced more than 90% prediction accuracy. This indicates that the proposed *e*ABC-LSSVM is capable of solving optimization problem, specifically in the prediction task. The *e*ABC-LSSVM is hoped to be useful to investors and commodities traders in planning their investment and projecting their profit.

Keywords: Artificial Bee Colony, Least Squares Support Vector Machines, Time series prediction

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&
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Table of Contents

Permission to Use	ii
Abstrak.....	iii
Abstract.....	iv
Acknowledgement	v
Table of Contents.....	vii
List of Tables	xii
List of Figures	xvi
List of Appendices	xx
List of Abbreviations	xxi
CHAPTER ONE INTRODUCTION	1
1.1 Background Study.....	1
1.2 Time Series Prediction.....	3
1.3 Problem Statement.....	10
1.4 Research Question.....	12
1.5 Objectives of Study.....	13
1.6 Limitation of Study.....	13
1.7 Scope of Study.....	14
1.8 Significance of Study.....	15
1.9 Thesis Outline.....	15
CHAPTER TWO LITERATURE REVIEW	17
2.1 Optimization of LSSVM Hyper Parameters.....	17
2.1.1 Optimization of LSSVM using Cross Validation.....	17
2.1.2 Optimization of LSSVM using Evolutionary Computation Algorithm.....	24
2.2 Methods to Enhance Searching Behavior in ABC Algorithm.....	39
2.3 Methods to Prevent Over fitting and Under fitting in LSSVM and SVM.....	44
2.4 Evaluation in Time Series Predictions.....	50
2.4.1 Metrics.....	50
2.4.2 Methods.....	56

2.4.2.1 Statistical Approach.....	56
2.4.2.2 Computational Intelligence Approach.....	59
2.5 Least Squares Support Vector Machines.....	68
2.6 Artificial Bee Colony.....	72
2.7 Levy Probability Distribution.....	77
2.7.1 Levy Probability Distribution in Evolutionary Computation.....	80
2.8 Summary.....	83
CHAPTER THREE METHODOLOGY	84
3.1 Introduction.....	84
3.2 Data Collection and Preparation.....	85
3.2.1 Test for Correlation.....	88
3.2.2 Test for Non linearity.....	90
3.2.3 Data Normalization.....	91
3.3 Algorithm Design.....	92
3.3.1 ABC-LSSVM.....	93
3.3.1.1Flow of ABC-LSSVM.....	93
3.3.1.1.1 Initialization Phase.....	93
3.3.1.1.2 EB Phase.....	95
3.3.1.1.3 OB Phase.....	96
3.3.1.1.4 SB Phase.....	97
3.3.2 Enrich the Exploitation Process in Search Space (<i>lvABC</i>).....	98
3.3.3 Preventing Over fitting and Under fitting (<i>cmABC</i>).....	100
3.3.4 <i>e</i> ABC-LSSVM.....	101
3.3.4.1 Flow of <i>e</i> ABC-LSSVM.....	102
3.3.4.1.1 Initialization Phase	103
3.3.4.1.2 EB Phase	104
3.3.4.1.3 OB Phase.....	106
3.3.4.1.4 SB Phase.....	107
3.4 Algorithm Development.....	107
3.5 Evaluation.....	108
3.5.1 Experimental Setup.....	108

3.5.1.1 Input and Output Variables.....	109
3.5.1.2 Training, Validation and Testing.....	111
3.5.1.3 Properties Setting.....	112
3.5.2 Performance Evaluation Metrics.....	113
3.5.3 Benchmarking Techniques.....	115
3.6 Summary.....	116
CHAPTER FOUR ARTIFICIAL BEE COLONY-LEAST SQUARES SUPPORT VECTOR MACHINES	117
4.1 Introduction.....	117
4.2 Summary.....	121
CHAPTER FIVE ENHANCED ARTIFICIAL BEE COLONY-LEAST SQUARES SUPPORT VECTOR MACHINES	122
5.1 Introduction.....	122
5.2 Summary.....	128
CHAPTER SIX RESULTS AND DISCUSSION	129
6.1 Introduction.....	129
6.2 <i>e</i> ABC-LSSVM on Energy Fuels.....	129
6.2.1 Data Set A.....	129
6.2.1.1 <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	129
6.2.1.1.1 Crude Oil Price.....	130
6.2.1.1.2 Heating Oil Price Prices.....	135
6.2.1.1.3 Gasoline Price.....	139
6.2.1.1.4 Propane Prices.....	143
6.2.1.1.5 Results Analysis for Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	147
6.2.1.2 <i>e</i> ABC-LSSVM vs Other Techniques.....	148
6.2.1.2.1 Crude Oil Price.....	149
6.2.1.2.2 Heating Oil Prices.....	152
6.2.1.2.3 Gasoline Prices.....	155
6.2.1.2.4 Propane Prices.....	158

6.2.1.2.5 Results Analysis for Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	161
6.2.2 Data Set B.....	162
6.2.2.1 <i>e</i> ABC-LSSVM vs ABC-LSSVM and Its Variants	162
6.2.2.1.1 Crude Oil Prices.....	163
6.2.2.1.2 Heating Oil Prices.....	167
6.2.2.1.3 Gasoline Prices.....	171
6.2.2.1.4 Propane Prices.....	175
6.2.2.1.5 Results Analysis for Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	179
6.2.2.2 <i>e</i> ABC-LSSVM vs Other Techniques.....	180
6.2.2.2.1 Crude Oil Prices.....	181
6.2.2.2.2 Heating Oil Prices.....	184
6.2.2.2.3 Gasoline Prices.....	187
6.2.2.2.4 Propane Prices.....	190
6.2.2.2.5 Results Analysis for Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	193
6.3 <i>e</i> ABC-LSSVM on Metal.....	194
6.3.1 Data Set C.....	194
6.3.1.1 <i>e</i> ABC-LSSVM vsABC-LSSVM and Its Variants.....	194
6.3.1.1.1 Gold.....	195
6.3.1.1.2 Results Analysis for Data Set C: <i>e</i> ABC-LSSVM vs ABC-LSSVM and Its Variants.....	199
6.3.1.2 <i>e</i> ABC-LSSVM vs Other Techniques.....	199
6.3.1.2.1 Gold.....	200
6.3.1.2.2 Results Analysis for Data Set C: <i>e</i> ABC-LSSVM vs Other Techniques.....	203
6.4 Summary.....	203
CHAPTER SEVEN CONCLUCIONS AND RECOMMENDATION FOR FUTURE WORKS	205
7.1 Conclusion.....	205

7.2 Contribution.....	205
7.2.1 Knowledge Contribution.....	207
7.2.2 Practical Contribution.....	208
7.3 Recommendation for Future Works.....	209
REFERENCES.....	211
APPENDICES.....	225

List of Tables

Table 1.1: Correlation among Energy Fuels from 22 December 1992–30 November 2007 ...	6
Table 2.1: Summaries of Time Series Prediction Technique using LSSVM Optimized by CV.....	22
Table 2.2: Summaries of Time Series Prediction Technique using LSSVM Optimized by EC Algorithm.....	36
Table 2.3:Summaries of Methods to Overcome the Searching Behavior in ABC Algorithm	43
Table 2.4: Summaries of Methods to Prevent Over fitting and Under fitting in LSSVM and SVM.....	48
Table 2.5: Summaries of metrics utilized in Evaluating of Time Series Prediction.....	54
Table 2.6: Summaries of Time Series Prediction Technique using Statistical Approach.....	58
Table 2.7: Summaries of Time Series Prediction Technique using CI Approach.....	66
Table 2.8: Data Representation of <i>e</i> ABC-LSSVM Algorithm.....	77
Table 2.9: Summaries of Levy Probability Distribution in Evolutionary Computation Algorithm.....	82
Table 3.1: Data Set A.....	85
Table 3.2: Sample of Data Set A.....	85
Table 3.3: Data Set B.....	86
Table 3.4: Sample Data Set B.....	86
Table 3.5: Data Set C.....	87
Table 3.6: Sample of Data Set C.....	87
Table 3.7: Price Correlation among Input Data Set A.....	89
Table 3.8: Price Correlation among Input Data Set B.....	89
Table 3.9: Price Correlation among Input Data Set C.....	89
Table 3.10: Samples of Normalized Input for Data Set A.....	92
Table 3.11: Samples of Normalized Input for Data Set B.....	92
Table 3.12: Samples of Normalized Input for Data Set C.....	92
Table 3.13: Initial Fitness Function.....	94
Table 3.14: EB Phase.....	95
Table 3.15: Probability.....	96
Table 3.16: OB Phase	97

Table 3.17: Initial Fitness Functions.....	104
Table 3.18: EB Phase.....	105
Table 3.19: Probability.....	105
Table 3.20: OB Phase.....	107
Table 3.21: Assigning Input and Output Variable for Data Set A.....	109
Table 3.22: Assigning Input and Output Variable for Data Set B.....	110
Table 3.23: Assigning Input and Output Variable for Data Set C.....	110
Table 3.24: Training, Validation and Testing.....	111
Table 3.25 Example of data arrangement for Data Set A.....	112
Table 3.26: Properties of Prediction Techniques Utilized.....	113
Table 3.27: Range of Percentage by MAPE.....	114
Table 6.1: CL Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	131
Table 6.2: Significant Test for CL Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	131
Table 6.3: HO Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	136
Table 6.4: Significant Test for HO Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	136
Table 6.5: HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	140
Table 6.6: Significant Test for HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	140
Table 6.7: PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	144
Table 6.8: Significant Test for PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	144
Table 6.9: CL Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	150
Table 6.10: Significant Test for CL Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	150
Table 6.11: HO Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	153
Table 6.12: Significant Test for HO Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	153

Table 6.13: HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	156
Table 6.14: Significant Test for HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	156
Table 6.15: PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	159
Table 6.16: Significant Test for PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	159
Table 6.17: CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	164
Table 6.18: Significant Test for CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	164
Table 6.19: HO Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	168
Table 6.20: Significant Test for HO Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	168
Table 6.21: HU Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	172
Table 6.22: Significant Test for HU Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	172
Table 6.23: PN Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	176
Table 6.24: Significant Test for PN Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	176
Table 6.25: CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	182
Table 6.26: Significant Test for CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	182
Table 6.27: HO Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	185
Table 6.28: HO Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	185
Table 6.29: HU Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	188
Table 6.30: Significant Test for HU Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	188

Table 6.31: PN Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	191
Table 6.32: Significant Test for PN Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	191
Table 6.33 Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	196
Table 6.34: Significant Test for Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	196
Table 6.35: Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. Other Techniques.....	201
Table 6.36: Significant Test for Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. Other Techniques.....	201

List of Figures

Figure 1.1: Classes of Commodities	4
Figure 2.1: Class of Evolutionary Computation.....	24
Figure 2.2: Mapping of the Input Space to a High Dimensional Feature Space.....	69
Figure 2.3: LPD with Different Value of α	78
Figure 3.1: Methodology.....	84
Figure 3.2: Representation of Food Source Position as Possible Solution in X	94
Figure 3.3: Representation of Food Source Position as Possible Solution in X	103
Figure 4.1: Flow of ABC-LSSVM in Initialization and EB Phase.....	118
Figure 4.2: Flow of ABC-LSSVM in OB and SB Phase.....	119
Algorithm 4.1: ABC-LSSVM algorithm.....	120
Algorithm 5.1: <i>lv</i> ABC-LSSVM.....	123
Algorithm 5.2: <i>cm</i> ABC-LSSVM.....	124
Figure 5.1: Flow of <i>e</i> ABC-LSSVM in Initialization and EB Phase.....	125
Figure 5.2: Flow of <i>e</i> ABC-LSSVM in OB and SB Phase.....	126
Algorithm 5.1: <i>e</i> ABC-LSSVM algorithm.....	127
Figure 6.1: Comparison of CL Price Predictions Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	133
Figure 6.2: Comparison of Convergence for CL Price Prediction Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	133
Figure 6.3: Exploitation of Search Space by <i>e</i> ABC-LSSVM in CL Price Prediction Data Set A.....	134
Figure 6.4: Exploitation of Search Space by ABC-LSSVM in CL Price Prediction Data Set A.....	134
Figure 6.5: Comparison of HO Price Predictions Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	137
Figure 6.6: Comparison of Convergence for HO Price Prediction Data Set A: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	137
Figure 6.7: Exploitation of search space by <i>e</i> ABC-LSSVM in HO Price Prediction Data Set A.....	138
Figure 6.8: Exploitation of search space by ABC-LSSVM in HO Price Prediction Data Set A.....	138

Figure 6.9: Comparison of HU Price Predictions Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	141
Figure 6.10: Comparison of Convergence for HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Variants ABC-LSSVM.....	141
Figure 6.11: Exploitation of Search Space by <i>eABC-LSSVM</i> in HU Price Prediction Data Set A.....	142
Figure 6.12: Exploitation of Search Space by ABC-LSSVM in HU Price Prediction Data Set A.....	142
Figure 6.13: Comparison of PN Price Predictions Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	145
Figure 6.14: Comparison of Convergence for PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	145
Figure 6.15: Exploitation of Search Space by <i>eABC-LSSVM</i> in PN Price Prediction Data Set A.....	146
Figure 6.16: Exploitation of Search Space by ABC-LSSVM in PN Price Prediction Data Set A.....	146
Figure 6.17: Comparison of CL price predictions Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	151
Figure 6.18: Comparison of Convergence for CL Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	151
Figure 6.19: Comparison of HO price predictions Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	154
Figure 6.20: Comparison of Convergence for HO Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	154
Figure 6.21 Comparison of HU Price Predictions Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	157
Figure 6.22 Comparison of Convergence for HU Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	157
Figure 6.23: Comparison of PN Price Predictions Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	160
Figure 6.24: Comparison of convergence rate for PN Price Prediction Data Set A: <i>eABC-LSSVM</i> vs. Other Techniques.....	160
Figure 6.25: Comparison of CL Price Predictions Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	165

Figure 6.26: Comparison of Convergence for CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	165
Figure 6.27: Exploitation of Search Space by <i>eABC-LSSVM</i> in CL Price Prediction Data Set B.....	166
Figure 6.28: Exploitation of Search Space by ABC-LSSVM in CL Price Prediction Data Set B.....	166
Figure 6.29: Comparison of HO Price Predictions Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	169
Figure 6.30: Comparison of Convergence for HO Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	169
Figure 6.31: Exploitation of Search Space by <i>eABC-LSSVM</i> in HO Price Prediction Data Set B.....	170
Figure 6.32 Exploitation of Search Space by ABC-LSSVM in HO Price Prediction Data Set B.....	170
Figure 6.33: Comparison of HU Price Predictions Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	173
Figure 6.34: Comparison of Convergence for HU Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	173
Figure 6.35: Exploitation of Search Space by <i>eABC-LSSVM</i> in HU Price Prediction Data Set B.....	174
Figure 6.36: Exploitation of Search Space by ABC-LSSVM in HU Price Prediction Data Set B.....	174
Figure 6.37: Comparison of PN Price Predictions Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	177
Figure 6.38 Comparison of Convergence Rate for PN Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. ABC-LSSVM and Its Variants.....	177
Figure 6.39: Exploitation of Search Space by <i>eABC-LSSVM</i> in PN Price Prediction Data Set B.....	178
Figure 6.40: Exploitation of Search Space by ABC-LSSVM in PN Price Prediction Data Set B.....	178
Figure 6.41: Comparison of CL Price Predictions Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	183
Figure 6.42: Comparison of Convergence for CL Price Prediction Data Set B: <i>eABC-LSSVM</i> vs. Other Techniques.....	183

Figure 6.43: Comparison of HO Price Predictions Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	186
Figure 6.44: Comparison of Convergence HO Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	186
Figure 6.45: Comparison of HU Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	189
Figure 6.46: Comparison of Convergence for HU Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	189
Figure 6.47: Comparison of PN Price Predictions Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	192
Figure 6.48: Comparison of Convergence for PN Price Prediction Data Set B: <i>e</i> ABC-LSSVM vs. Other Techniques.....	192
Figure 6.49: Comparison of Data Set C price prediction: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	197
Figure 6.50: Comparison of convergence for Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. ABC-LSSVM and Its Variants.....	197
Figure 6.51: Exploitation of Search Space by <i>e</i> ABC-LSSVM in Data Set C Price Prediction	198
Figure 6.52: Exploitation of search space by ABC-LSSVM in Data Set C Price Prediction	198
Figure 6.53: Comparison of Data Set C price predictions: <i>e</i> ABC-LSSVM vs. Other Techniques.....	202
Figure 6.54: Comparison of Convergence for Data Set C Price Prediction: <i>e</i> ABC-LSSVM vs. Other Techniques.....	202

List of Appendices

Appendix A Correlation Among Input.....	225
Appendix B BDS Test Results.....	228
Appendix C Testing Prediction Results.....	229
Appendix D Sample of Raw Data.....	233
Appendix E Publications during the Doctorate Study.....	234

List of Abbreviations

ABC	-	Artificial Bee Colony Algorithm
ACO	-	Ant Colony Optimization
AFSA	-	Artificial Fish Swarm Algorithm
AI	-	Artificial Intelligent
ANFIS	-	Adaptive Neuro Fuzzy Inference
ANN	-	Artificial Neural Network
ARIMA	-	Autoregressive Integrated Moving Average
ARE	-	Average Relative Error
BCC	-	Bacteria Colony Chemotaxis
BPNN	-	Back Propagation Neural Network
CI	-	Computational Intelligence
CV	-	Cross Validation
DE	-	Differential Evolution
EB	-	Employed Bee
EC	-	Evolutionary Computation
EMD	-	Empirical Mode Decomposition
EP	-	Evolutionary Programming
ERM	-	Empirical Risk Minimization
GA	-	Genetic Algorithm
GDP	-	Gross Domestic Product
GMSE	-	Generalization Mean Square Error
GSA	-	Gravitational Search Algorithm
IMF	-	Intrinsic Mode Function
LMSE	-	Learning Mean Square Error
LSSVM	-	Least Squares Support Vector Machines
MAE	-	Mean Absolute Error
MAPE	-	Mean Absoluter Percentage Error
MATLAB	-	Matrix Laboratory Software
MLP	-	Multilayer Perceptron

MRE	-	Mean Relative Error
MSE	-	Mean Square Error
NLL	-	Negative Log Likelihood
OB	-	Onlooker Bee
OECD	-	Organization for Economic Co-operation and Development
PCA	-	Principal Component Analysis
PNN	-	Probabilistic Neural Network
PSO	-	Particle Swarm Optimization
QP	-	Quadratic Programming
RBF	-	Radial Basis Function
RE	-	Relative Error
RMSE	-	Root Mean Square Error
SB	-	Scout Bee
SOM	-	Self Organizing Maps
SRM	-	Structural Risk Minimization
SVM	-	Support Vector Machines
SVR	-	Support Vector Regression
WTI	-	West Texas Intermediate

CHAPTER ONE

INTRODUCTION

1.1 Background Study

Since its emergence in the past decades, Least Squares Support Vector Machines (LSSVM) (Suykens, Van Gestel, De Brabanter, De Moor, & Vandewalle, 2002) which is an extension of Support Vector Machines (SVM) (Vapnik, 1995) has contributed significant impact in machine learning community. As a powerful algorithm, it has been recognized as one of the standard tools in solving various data mining tasks which include prediction, classification and many others (Cheng, Guo, & Wu, 2010, Li, 2009). Nonetheless, besides its diversity in application, it is worth noting that the capability of LSSVM is highly dependent on the value of its hyperparameters, namely regularization parameter, γ and kernel parameter, σ^2 (Jiang & Zhao, 2013).

In this regard, this study proposes a hybridization of LSSVM with a relatively new optimization algorithm namely Artificial Bee Colony (ABC) (Karaboga, 2005). The ABC algorithm which has been introduced by Dervis Karaboga is enlightened from the intelligent foraging behavior of honey bees swarm (Karaboga, 2005). In 2008, an extensive review and comparative analysis regarding its performance efficiency was examined which concluded that the ABC algorithm is comparable to the other existing optimization algorithms including Differential Evolution (DE), Particle Swarm Optimization (PSO) and Genetic Algorithm (GA) (El-Abd, 2012; Karaboga & Basturk, 2008).

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