INTEGRATED MODELLING TOOL OF GEOSPATIAL CRITERIA FOR ECO-INDUSTRIAL PARK SITE SELECTION

STEVEN KUBA NUHU

A thesis submitted in fulfilment of the requirements for the award of the degree of Doctor of Philosophy

School of Chemical and Energy Engineering Faculty of Engineering Universiti Teknologi Malaysia

APRIL 2022

DEDICATION

This thesis is dedicated to my mother (Nah Uhoman Nuhu) of blessed memory, who taught me that even the largest task can be accomplished if it is done one step at a time. It is also dedicated to my father, who taught me that the best kind of knowledge to have is that which is learned for its own sake. And to my wife, children and siblings who stood beside me in prayers, with courage, understanding and love.

ACKNOWLEDGEMENT

In preparing this thesis, I was in contact with many people, researchers, academicians, and practitioners. They have contributed to my understanding and thoughts. I wish to express my sincere appreciation to my main thesis supervisor, Professor Dr Ir. Zainuddin Bin Abd Manan, for encouragement, guidance, critics, and friendship. I am also very thankful to my co-supervisor Professor Dr Ir. Sharifah Rafidah Wan Alwi and Dr Mohd Nadzri Md Reba for their guidance, advice, and motivation. Without their continued support and interest, this thesis would not have been the same as presented here.

I am also indebted to Universiti Teknologi Malaysia (UTM) for their sound knowledge imparted to me during my PhD study. The Librarians of UTM also deserve special thanks for their assistance in supplying the relevant literature. I am thankful to Plateau State Polytechnic, Barkin-Ladi, Jos, Nigeria for allowing me to further my education to the PhD level. I am also grateful to Tertiary Education Trust Fund (TETFund), Abuja, Nigeria for their financial support.

My unquantifiable appreciation to Fatin Nabihah Syahira Ridzuan for assisting me with the fundamentals of GIS. My profound appreciation goes to my wife, children and siblings for their support and understanding during my absence. My fellow postgraduate student should also be recognised for their support. My sincere appreciation also extends to all my colleagues and others who have assisted on various occasions. Their views and tips are useful indeed. Unfortunately, it is not possible to list all of them in this limited space. I am grateful to all my family members.

ABSTRACT

Single multi-criteria decision-making (SMCDM) approaches are limited by inconsistencies in the evaluation of criteria weights, making it unreliable for industrial park (IP) site selection. This led to wrong industrial site choice, difficulty to attract industry symbiosis clusters and often resulted in brownfield industrial parks (BFIP) with excessive greenhouse gas emissions. Many BFIPs are being phased out in favour of eco-industrial parks (EIP) with favourable locations for industrial clusters to synergise and manage materials efficiently. Industrial site selection heavily depends on criteria weighting and ranking. This study aimed to develop an integrated multi-criteria decision-making (IMCDM) tool and MCDM-GIS model that would enable researchers to consolidate the advantages and eliminate the weaknesses of SMCDM. To address the mentioned limitations, the analytic hierarchy process (AHP), analytic network process (ANP), and fuzzy-analytic hierarchy process (F-AHP) tools were constructed using the eigenvalue, limit supermatrix and triangular fuzzy numbers. The spatial criteria weights and ranking of water bodies, roads, residential areas, existing industries, land surface temperature and slope were evaluated. The SMCDM priority vectors were alternately integrated to produce the IMCDM methods which were also used in assessing the criteria weights. All criteria weights were subjected to sensitivity analyses and standard deviation. To test the weighting consistency of the SMCDM, IMCDM and the model efficiency, the spatial criteria data of 2009 and 2019 of Tanjung Langsat Industrial Area (TLIA) were collected using the geographic information system (GIS) and screened by the Boolean logic. The Landsat-7 enhanced thematic mapper and the kompsat-3 imager obtained the land use land cover data through PLANMalaysia. The GIS prepared the Euclidean distance and reclassified raster layers. The single and integrated weights percent were separately overlaid in the MCDM-GIS model with the 2009 criteria dataset. The SMCDM and IMCDM approach identified the water bodies as suitable brownfield eco-industrial park (BF-EIP) sites. This shows tool inconsistency using sparse criteria because industries cannot be built inside water bodies. Using the 2019 data, the AHP, ANP and F-AHP identified 5%, 2% and 3% as very-highly-suitable sites all in the northern part of the TLIA. The small spots were found away from the existing industries' location when superimposed with the criteria layers. The integrated hierarchy network-fuzzy analytic process and hierarchy network analytic process methods identified vast sites of different suitability but included 12% part of the water bodies as low-suitable, hence considered as inconsistent. The hierarchy fuzzy-analytic process (H-FAP) and network fuzzy hierarchy-analytic process (NFh-AP) measured large different suitable sites with explicit identification not including water bodies, hence consistent and reliable tools. When overlaid with the criteria layer, the veryhighly-suitable site was identified in the centre of the TLIA, falling in place with the existing industries. The integrated H-FAP and NFh-AP algorithms become consistent and the best because of the interplay of hierarchical, geometric ratio and networking tools coming from different groupings of paired comparison and uncertainty, as well as their weights being close to the averages of the criteria set as evaluated by the standard deviation. The IMCDM tools are consistent only with concentrated criteria. However, the SMCDM tools are weak with both the sparse and concentrated criteria. This can lead to the wrong choice of an industrial site. Both SMCDM and IMCDM measured the economic, environmental, and social attributes as the most important in supporting the criteria to achieve the BF-EIP site selection. The MCDM-GIS model is efficient as the outputs of suitable EIP site layers under different criteria weights and distinguished spatial data. The H-FAP, NFh-AP have been proven to be the consistent criteria weight assessment algorithms and a flexible MCDM-GIS is hereby presented to support the government, EIP investors/developers, and researchers. This is to achieve an easy 4IR-driven modelling process to select brownfields for EIPs.

ABSTRAK

Pendekatan membuat keputusan pelbagai kriteria tunggal (SMCDM) adalah terhad oleh penilaian pemberat kriteria yang tidak konsisten, menjadikannya kurang berkesan untuk pemilihan tapak taman industri (IP). Ini membawa kepada pilihan tapak perindustrian yang kurang sesuai, sekaligus menjadikan kelompo simbiosis industri kurang menarik dan mengakibatkan pelepasan gas rumah hijau yang berlebihan dalam taman perindustrian brownfield (BFIP). Banyak BFIP sedang dilupuskan secara berperingkat dan digantikan dengan taman eko-industri (EIP) pada lokasi yang sesuai bagi membolehkan kelompok industri bersinergi dan mengurus bahan dengan cekap. Pemilihan tapak industri sangat bergantung pada pemberat dan kriteria kedudukan. Matlamat kajian adalah untuk membangunkan kaedah membuat keputusan berbilang kriteria bersepadu (IMCDM) dan model MCDM-GIS yang membolehkan penyelidik menggabungkan kelebihan dan menghapuskan kelemahan SMCDM. Untuk menangani kelemahan yang dinyatakan, proses hierarki analitik (AHP), proses rangkaian analitik (ANP) dan proses hierarki analitik-kabur (F-AHP) telah dibina menggunakan nilai eigen, had supermatriks dan nombor kabur segi tiga. Berat kriteria spatial dan kedudukan sumber air, jalan raya, kawasan perumahan, industri sedia ada, suhu permukaan tanah, dan cerun telah dinilai. Vektor keutamaan SMCDM disepadukan secara bergilirgilir dan menghasilkan kaedah IMCDM yang juga digunakan dalam menilai wajaran kriteria. Semua berat kriteria tertakluk kepada analisis sensitiviti dan sisihan piawai. Untuk menguji keseragaman pemberat SMCDM, IMCDM dan kecekapan model, data kriteria spatial 2009 dan 2019 kawasan perindustrian Tanjung Langsat (TLIA) telah dikumpulkan menggunakan sistem maklumat geografi (GIS) dan disaring oleh logik Boolean. Landsat-7 pemeta tematik dipertingkatkan dan pengimej kompsat-3 dan kompsat-3 imager memperoleh data tutupan tanah dan guna tanah melalui PLANMalaysia. GIS menyediakan jarak Euclidean dan mengklasifikasikan semula lapisan raster. Peratus pemberat tunggal dan bersepadu telah ditindih secara berasingan dalam model MCDM-GIS dengan set data kriteria 2009. Pendekatan SMCDM dan IMCDM telah dikenal pasti termasuk sumber air sebagai tapak taman eko-industri brownfield (BF-EIP) yang sesuai. Ini menunjukkan tidak konsisten nya kaedah menggunakan kriteria jarang oleh kerana industri tidak boleh dibina di dalam air. Menggunakan data 2019, AHP, ANP dan F-AHP mengenal pasti 5%, 2% dan 3% sebagai tapak yang sangat sesuai semuanya di bahagian utara TLIA. Bintikbintik kecil ditemui jauh dari lokasi industri sedia ada apabila ditindih dengan lapisan kriteria. Kaedah rangkaian hierarki proses analitik kabur dan proses analitik rangkaian hierarki bersepadu mengenal pasti tapak yang luas dengan kesesuaian yang berbeza tetapi termasuk 12% bahagian sumber air sebagai kurang sesuai, oleh itu dianggap tidak konsisten. Proses analitik kabur hierarki (H-FAP) dan proses hierarki-analitik rangkaian kabur (NFh-AP) mengukur tapak besar berbeza yang sesuai dengan pengenalan yang jelas tanpa memasukkan sumber air, justeru mendapati alat yang konsisten dan boleh dipercayai. Apabila ditindih dengan lapisan kriteria, ianya adalah tapak yang sangat sesuai yang dikenal pasti di tengah-tengah TLIA, sesuai dengan industri sedia ada algoritma H-FAP dan NFh-AP bersepadu menjadi konsisten dan terbaik kerana interaksi hierarki, nisbah geometri dan alatan rangkaian yang datang daripada kumpulan berbeza perbandingan berpasangan dan ketidakpastian, serta beratnya hampir dengan purata kriteria yang ditetapkan seperti yang dinilai oleh sisihan piawai. Alat IMCDM hanya konsisten dengan kriteria tertumpu. Walau bagaimanapun, alatan SMCDM adalah lemah dengan kedua-dua kriteria yang jarang dan tertumpu. Ini boleh menyebabkan pilihan tapak perindustrian yang salah. Kedua-dua SMCDM dan IMCDM mengukur sifat ekonomi, alam sekitar dan sosial sebagai yang paling penting untuk menyokong kriteria untuk mencapai pemilihan tapak BF-EIP. Model MCDM-GIS adalah cekap kerana keluara lapisan tapak EIP yang sesuai di bawah berat kriteria yang berbeza dan data spatial dibezakan. H-FAP, NFh-AP telah dibuktikan sebagai algoritma penilaian berat kriteria yang konsisten dan MCDM-GIS yang fleksibel telah dihasilkan bagi membantu pihak kerajaan, pelabur/pemaju EIP serta penyelidik, sekaligus menghasilkan proses pemodelan dipacu 4IR yang mudah untuk memilih medan brownfileds untuk EIP.

TABLE OF CONTENTS

TITLE

13

iii	CLARATION	DEC
iv	DICATION	DED
v	KNOWLEDGEMENT	АСК
vi	TRACT	ABS
vii	TRAK	ABS
viii	BLE OF CONTENTS	TAB
XV	Г OF TABLES	LIST
xviii	Γ OF FIGURES	LIST
xxii	Γ OF ABBREVIATIONS	LIST
XXV	Г OF SYMBOLS	LIST
xxvi	Γ OF APPENDICES	LIST
1	INTRODUCTION	CHAPTER 1
1	Background	1.1
5	Problem Statement	1.2
7	Research Goal	1.3

1.4	Scope of Work	8
1.5	Significance of the Study	9
1.6	Thesis Outline	10

CHAPTER 2 LITERATURE REVIEW

2.1	Introd	uction	13
2.2	The Planni	Eco-Industrial Park Site Concepts/Goals, ing, Development, and Challenges	13
	2.2.1	The Eco-Industrial Park Concept and Objectives	13
	2.2.2	Eco-Industrial Park Planning	17
	2.2.3	The Development of Eco-Industrial Park	19
	2.2.4	Challenges of Eco-Industrial Park	21

2.3	The C	ategories and Types of Industrial Parks	22
	2.3.1	Greenfield Industrial Park	22
	2.3.2	Brownfield Industrial Park	22
	2.3.3	Types of Eco-Industrial Parks	25
2.4	Eco-II	ndustrial Parks Development Projects	25
	2.4.1	Eco-Industrial Park Symbiosis in Kalundborg, Denmark	25
	2.4.2	The United Kingdom National Eco-Industrial Park Symbiosis Program	28
	2.4.3	Eco-Industrial Park Developmental Projects Around the World	29
2.5	The Geogr	Multi-Criteria Decision-Making and the caphic Information System	34
2.6	Overv Multi- Select	riew of Geographic Information System and -criteria Decision-Making in Industrial Site ion	35
	2.6.1	Application of GIS and MCDM by Country/Region	35
	2.6.2	Published Journals	36
	2.6.3	Yearly Publications and Citations	38
	2.6.4	Number of Authors and Tools Used	40
2.7	Proble Asses	ems of Single MCDM Tools in Criteria Weight sment	43
	2.7.1	Challenges of GIS Application for EIP Site Selection	45
	2.7.2	Application of Multi-Criteria Decision-Making Issues on EIP Site Selection	46
2.8	Impac Use an	et of Geographic Information System on Land nd Eco-Industrial Parks	47
2.9	Hieran Decisi	rchy, Network and Fuzzy Multi-Criteria ion-Making Tools Overview	48
	2.9.1	The Analytic Hierarchy Process	48
	2.9.2	Analytic Network Process	51
	2.9.3	Fuzzy Analytic Hierarchy Process	54
2.10	Addre MCD Select	essing the Research Gap of the Inconsistent M Tools for Brownfield Eco-Industrial Park Site ion	57

CHAPTER 3	RESEARCH METHODOLOGY	59
3.1	Introduction	59
3.2	Criteria Identified by the GIS for the Brownfield Eco- Industrial Park Site Selection Study	60
3.3	Software Used	60
3.4	The Analytic Hierarchy Process	63
	Step 1 Analytic Hierarchy Structure Construction	63
	Step 2 Pairwise Comparison Matrix Formation and Evaluation	63
	Step 3 Weight Normalisation Determination	64
	Step 4 Priority Vector Process	64
	Step 5 Consistency Confirmation Evaluation	64
	Step 6 Overall Priority Ranking	66
	Step 7 Sensitivity Analysis Calculation	66
3.5	The Analytic Network Process Structure	66
	Step 1 Supermatrix Synthesis	66
	Step 2 Weighted Supermatrix Assessment	67
	Step 3 Limit Supermatrix Computations	68
	Step 4 Criteria Weight Standardization	68
	Step 5 Overall Priority Ranking	68
3.6	Fuzzy Analytic Hierarchy Process	69
	Step 1 Construction of Fuzzy Analytic Hierarchy Process Structure	69
	Step 2 Decomposition and Pair-wise Comparison Matrix and Evaluation	69
	Step 3 Fuzzy Geometric Mean Ratio Evaluation	70
	Step 4 Relative Fuzzy Weights Determination	70
	Step 5 Fuzzy Weights Defuzzification	71
	Step 6 Weights Normalization	71
	Step 7 Overall Priority Ranking	71
3.7	Multi-Criteria Decision-Making Eigenvectors Integration	72
3.8	Standard Deviation	72

73	3.9 MCDM-GIS Model Design for Eco-Industrial Park Site Selection
74	3.10 Geographic Information System
74	Step 1 Location and Collection of Brownfield Spatial Criteria Data
74	Step 2 Criteria Screening and Classification
75	Step 3 Conversion, Euclidean Distances Raster Layers Determination and Reclassification
75	3.11 Weighted Overlay Analyses
10	5.11 Weighted Overlag Thild 565
77	CHAPTER 4 RESULTS AND DISCUSSION
77	4.1 The Importance of Criteria and Alternatives for Eco- Industrial Park Site Selection
79	4.2 The Analytic Hierarchy Process Hierarchy Structure
80	4.2.1 Criteria Pairwise Comparison
81	4.2.2 Normalised Criteria Weight
82	4.2.3 Criteria Priority Vector and Consistency Ratio Based on Goal
83	4.2.4 Alternatives Weights Based on Roads
Q 1	4.2.5 Alternatives Weights Based on Existing Industries
86	126 Alternatives Weights Based on Water Bodies
87	4.2.7 Attributes Weights Based on Residential Areas
88	4.2.8 Attributes Weights with Respect to Slope
00	4.2.0 Alternatives Pairwise Comparison Matrix
90	Based on Land Surface Temperature
01	4.2.10 Analytic Hierarchy Process Overall Priority Vector
91	
93	4.2.11 The Analytic Hierarchy Process Overall Priority Sensitivity Analysis
94	4.3 Analytic Network Process
	4.3.1 Unweighted Supermatrix of Analytic Network Process
95	1100035
96	4.3.2 Limit Supermatrix
	 4.2.11 The Analytic Hierarchy Process Overall Priority Sensitivity Analysis 4.3 Analytic Network Process 4.3.1 Unweighted Supermatrix of Analytic Network Process 4.3.2 Limit Supermatrix

	4.3.3	Weights Normalisation	97
	4.3.4	The Analytic Network Process Sensitivity Analysis	
			99
4.4	The Fu	uzzy Analytical Hierarchy Process	100
	4.4.1	The Triangular Numbers Pairwise Comparison Matrix	101
			101
	4.4.2	The Geometric Ratio	102
	4.4.3	The Fuzzy Relative Weights	102
	4.4.4	Defuzzified Fuzzy Weights	102
	4.4.5	Normalised Criteria Weights	102
	4.4.6	Alternatives Weights Based on Roads	104
	4.4.7	Attributes Weights Based on Existing Industries	
			105
	4.4.8	Attributes Comparison Matrix Based on Water Bodies	
			106
	4.4.9	Alternatives Comparison Matrix Based on Slope	107
			107
	4.4.10	Alternatives Comparison Matrix Based on Residential Areas	108
	4.4.11	Attributes Comparison Matrix Based on Land Surface Temperature	109
4.5	The In	tegration of the Pairwise, Network and Fuzzy	
	Metho	ds	113
4.6	Standa	rd Deviation and Criteria Weight Consistency	117
4.7	The St	udy Site	119
4.8	Data C	Collected	120
4.9	Screen	ed Data by the Boolean Logic	121
4.10	Land U	Jse Land Cover Layers	121
4.11	Criteri	a Euclidean Distance	123
	4.11.1	Roads Euclidean Distance	123
	4.11.2	Existing Industries Euclidean Distance	125
	4.11.3	Water Bodies Euclidean Distance	126

	4.11.4 Residential Areas Euclidean Distance	127
4.12	Reclassification of Criteria Raster Layers	129
	4.12.1 Roads Layers Reclassified	130
	4.12.2 Existing Industries Layers Reclassified	131
	4.12.3 Waterbodies Layers Reclassified	133
	4.12.4 Residential Area Layers Reclassified	134
	4.12.5 Slope Reclassified Layer	136
	4.12.6 Land Surface Temperature Layer Reclassified	138
4.13	The Multi-Criteria Decision Making–Geographic Information System Model	139
4.14	Designed Model Output of Single Multi-Criteria Decision-Making with the Tanjung Langsat Industrial Area Criteria Data	140
	4.14.1 Analytic Hierarchy Process with 2009 Spatial Criteria Data	140
	4.14.2 Analytic Network Process with 2009 Spatial Criteria Data	142
	4.14.3 Fuzzy Analytic Hierarchy Process with 2009 Spatial Criteria Data	144
	4.14.4 Model Validation	145
	4.14.5 Analytic Hierarchy Process with 2019 Spatial Criteria Data	145
	4.14.6 Analytic Network Process with 2019 Spatial Criteria Data	147
	4.14.7 Fuzzy Analytic Hierarchy Process with 2019 Spatial Criteria Data	149
4.15	Integrated Algorithms Weights Percent with Tanjung Langsat Industrial Area Criteria Data	151
	4.15.1 HN-FAP Weights Percent with 2009 Spatial Criteria Data	152
	4.15.2 HNAP Weights with 2009 Spatial Criteria Data	153
	4.15.3 H-FAP Eigenvectors with 2009 Spatial Criteria Data	154
	4.15.4 NFh-AP Priority Vectors with 2009 Spatial Criteria Data	155
	4.15.5 HN-FAP Weights Percent with 2019 Spatial Criteria Data	156

	4.15.6 HNAP Weights with 2019 Spatial Criteria Data	150
		158
	4.15.7 H-FAP Eigenvectors with 2019 Spatial Criteria Data	159
	4.15.8 NFh-AP Priority Vectors with 2019 Spatial Criteria Data	161
4.16	Proposed Brownfield Eco-Industrial Park Site Selection Guidelines	167
CHAPTER 5	CONCLUSION AND RECOMMENDATIONS	171
5.1	Research Outcomes	171
5.2	Contributions to Knowledge	173
5.2 5.3	Contributions to Knowledge Future Works	173 174
5.2 5.3 REFERENCES	Contributions to Knowledge Future Works	173 174 175

LIST OF PUBLICATIONS	234

LIST OF TABLES

TABLE NO.	TITLE	PAGE
Table 2.1	Eco-Industrial Park Targets	16
Table 2.2	EIP Development Benefits	21
Table 2.3	Differences Between Greenfield and Brownfield EIPs	24
Table 2.4	The Main Benefits of Eco-Industrial Park	28
Table 2.5	The Earliest and Prominent EIPs and Their Present State	33
Table 2.6	Characteristics of Some Successful EIPs	33
Table 2.7	Grouping of Single Traditional Multi-Criteria Decision- Making Tools	45
Table 3.1	2009 and 2019 Criteria Data for Tanjung Langsat Industrial Area	60
Table 4.1	Pairwise Comparison Matrix of Criteria Base on Goal	81
Table 4.2	Normalised Values for Criteria Based on Goal	81
Table 4.3	Priority Vector, Eigenvalue and Consistency Ratio of Criteria Based on Goal	83
Table 4.4	Comparison Matrix for Alternatives Based on Roads	83
Table 4.5	AHP Normalised Values for Attributes Based on Roads	84
Table 4.6	Attributes Priority Vector, Eigenvalue and Consistency Ratio Based on Roads	84
Table 4.7	Alternatives Pairwise Comparison Matrix Based on Existing Industries	85
Table 4.8	Normalised Attributes Based on Existing Industries	85
Table 4.9	Alternatives Priority Vector, Eigenvalue and Consistency Ratio Based on Existing Industries	85
Table 4.10	Alternatives Pairwise Comparison Matrix Based on Water Bodies	86
Table 4.11	Attributes Standardised Pairwise Comparison Based on Water Bodies	86
Table 4.12	Alternatives Priority Vector, Eigenvalue and Consistency Ratio Based on Water Bodies	87

Table 4.13	Attributes Pairwise Comparison Matrix Based on Residential Areas	87
Table 4.14	Standardised Pairwise Comparison and Eigenvector	88
Table 4.15	Attributes Weight Based on Residential Areas	88
Table 4.16	Attributes Pairwise Comparison Matrix Based on Slope	89
Table 4.17	Attributes Standardised Values Based on Slope	89
Table 4.18	Attributes weights, Eigenvalue and Consistency Ratio Based on Slope	89
Table 4.19	Alternatives Pairwise Comparison Matrix Based on Land Surface Temperature	90
Table 4.20	Attributes Normalised Pairwise Comparison Based on Land Surface Temperature	90
Table 4.21	Attributes Priority Vector, Eigenvalue and Consistency Ratio Based on Land Surface Temperature	91
Table 4.22	Overall AHP Criteria and Alternative Weights of Importance	91
Table 4.23	AHP Sensitivity Analysis Percent Error	93
Table 4.24	Unweighted Supermatrix	96
Table 4.25	Weighted Supermatrix	96
Table 4.26	Limit Supermatrix	97
Table 4.27	Overall ANP Normalised Priority Vector	97
Table 4.28	ANP Sensitivity Analysis and Percent Error	99
Table 4.29	F-AHP Criteria Pairwise Comparison Matrix	101
Table 4.30	Geometric Mean, Relative Weights, Defuzzified, and Priority Vectors for Goal Based on Criteria	103
Table 4.31	F-AHP Comparison Matrix for Alternatives Based on Roads	104
Table 4.32	Alternatives Geometric Ratio, Relative Weights, Defuzzified and Priority Vectors Based on Roads	104
Table 4.33	F-AHP Matrix for Alternatives Based on Existing Industries	105
Table 4.34	Alternatives Geometric Ratio, Relative Weights, defuzzified and Priority Vectors Based on Existing Industries	106

Table 4.35	F-AHP Matrix for Alternatives Based on Waterbodies				
Table 4.36	F-AHP Geometric Ratio, Relative Weights, Defuzzified and Priority Vectors for Alternatives Based on Water Bodies	107			
Table 4.37	F-AHP Matrix for Alternatives Based on Slope	107			
Table 4.38	Alternatives Geometric Mean, Fuzzy Weights, Defuzzified, and Priority Vectors Based on Slope	108			
Table 4.39	F-AHP Matrix for Alternatives Based on Residential				
Table 4.40	Alternatives Geometric Ratio, Fuzzy Weights, Defuzzified and Weight of Importance Based on Residential Areas	109			
Table 4.41	Alternatives Weightage Based on Land Surface Temperature	109			
Table 4.42	Alternatives Geometric Mean, Relative Weights, Defuzzified and Priority Vectors Based on Land Surface Temperature	110			
Table 4.43	F-AHP Overall Priority Vector	110			
Table 4.44	F-AHP Sensitivity Analysis Percent Error	112			
Table 4.45	Single Overall Criteria Weight Importance	113			
Table 4.46	Integrated Overall Criteria Weight Importance	114			
Table 4.47	Integrated Attributes Weight Importance	117			
Table 4.48	Single and Integrated Algorithms Weights Standard Deviations	118			
Table 4.49	Roads Designated Scale Factor	130			
Table 4.50	Existing Industries Designated Scale Factor	131			
Table 4.51	Waterbodies Designated Scale Factor	133			
Table 4.52	Residential Area Designated Scale Factor	135			
Table 4.53	Slope Percentage and Designated Scale Factor	136			
Table 4.54	Land Surface Temperature and Designated Scale Factor	138			
Table 4.55	Summary of TLIA EIP Site Suitability Layers Produced by the Single MCDM Tools	151			
Table 4.56	Summary of TLIA EIP Site Suitability Layers Produced by the Integrated MCDMs	163			

LIST OF FIGURES

FIGURE NO	D. TITLE	PAGE
Figure 2.1	Overall Concept of EIP	15
Figure 2.2	Supply and Demand (Value Chain) for EIP Symbiosis	20
Figure 2.3	The Kalundborg EIP	26
Figure 2.4	EIP Projects in Developing Countries	32
Figure 2.5	Application of GIS and MCDM for EIP Site Selection Study by Regions/Countries	36
Figure 2.6	Use of GIS and MCDM in EIP Site Selection Case Studies Journals by Publications	37
Figure 2.7	Yearly Number of Publications	39
Figure 2.8	Yearly Number of Citations	40
Figure 2.9	Number of Authors and Tool Applied	41
Figure 2.10	Spatial Multi-Criteria Analysis Structure	46
Figure 2.11	Analytic Hierarchy Process Structure	49
Figure 2.12	Network Framework	52
Figure 2.13	Triangular Membership Function of Fuzzy Number	55
Figure 2.14	Trapezoid Membership Function of Fuzzy Number	55
Figure 2.15	Bell-Shape/Gaussian Membership Function of Fuzzy Number	56
Figure 2.16	Membership Functions Linguistic Variables for Criteria Comparison	56
Figure 3.1	Flowchart of SMCDM, IMCDM and GIS Processes	61
Figure 3.2	Flow Diagram of the Methodology	62
Figure 4.1	AHP and F-AHP Framework for Criteria Weight Importance	80
Figure 4.2	Overall AHP Criteria Weight Importance	92
Figure 4.3	AHP Alternative Weight Importance	92
Figure 4.4	AHP Sensitivity Analysis Error Bar	94
Figure 4.5	The Analytic Network Process Structure	95

Figure 4.6	ANP Criteria Weight of Importance	98
Figure 4.7	ANP Alternative Weights	99
Figure 4.8	ANP Sensitivity Analysis Error Bar	100
Figure 4.9	F-AHP Overall Criteria Weight Importance	111
Figure 4.10	F-AHP Alternative Weight Percent	112
Figure 4.11	F-AHP Sensitivity Analysis Error Bar	113
Figure 4.12	The Integrated Algorithms Criteria Weight Importance	115
Figure 4.13	HN-FAP Error Bar	115
Figure 4.14	HNAP Error Bar	115
Figure 4.15	H-FAP Error Bar	116
Figure 4.16	NFh-AP Error Bar	116
Figure 4.17	Integrated Algorithm Attribute Weight Importance	117
Figure 4.18	(a) Extended Macro Location (Johor, Malaysia), (b) Narrow Macro Location (Johor Bahru, Johor), (c) Micro	
	Location (Tanjung Langsat Industrial Area in Johor Bahru)	120
Figure 1 10	Land Use of Tanjung Langsat Industrial Area in 2009	120
Figure 4.20	Land Use of Tanjung Langsat Industrial Area in 2009	122
Figure 4.20	Roads Network in 2009	123
Figure 4.22	Roads Network in 2019	124
Figure 4.22	Existing Industries in 2009	124
Figure 4.23	Existing Industries in 2009	125
Figure 4.25	Water Bodies in 2009	120
Figure 4 26	Water Bodies in 2019	127
Figure 4.27	Residential Areas in 2009	128
Figure 4.28	Residential Areas in 2019	128
Figure 4.29	Reclassified Laver of Roads of 2009 Data	130
Figure 4.30	Reclassified Layer of Roads of 2019 data	131
Figure 4.31	Reclassified Laver of Existing Industries of 2009 Data	132
Figure 4.32	Reclassified Layer of Existing Industries of 2019 Data	132
Figure 4.33	Reclassified Laver of Water Bodies of 2009 Data	133
-0		-00

Figure 4.34	Reclassified Layer of Water Bodies of 2019 Data	134
Figure 4.35	Reclassified Layer of Residential Areas of 2009 Data	135
Figure 4.36	Reclassified Layer of Residential Areas of 2019 Data	136
Figure 4.37	Reclassified Layer of Slope of 2009 Data	137
Figure 4.38	Reclassified Layer of Slope of 2019 Data	137
Figure 4.39	Reclassified Layer of Land Surface Temperature of 2009 Data	139
Figure 4.40	Reclassified Layer of Land Surface Temperature of 2019 Data	139
Figure 4.41	MCDM - GIS-Based EIP Site Selection Suitability Model	140
Figure 4.42	EIP Suitability Layer from AHP Weights with 2009 Criteria Data	141
Figure 4.43	EIP Suitability Layer from AHP Weights with 2009 Data Overlaid	142
Figure 4.44	EIP Suitability Map from ANP Weights with 2009 Data	143
Figure 4.45	EIP Suitability Map from ANP Weights with 2009 Data Imposed	143
Figure 4.46	EIP Suitability Layer from F-AHP Weights with 2009 Data	144
Figure 4.47	EIP Suitability Map from F-AHP Weights with 2009 Data Overlaid	145
Figure 4.48	EIP Suitability Layer from AHP Weights with 2019 Data	146
Figure 4.49	EIP Suitability Layer from AHP Weights with 2019 Data Overlaid	147
Figure 4.50	EIP Suitability Map from ANP Weights with 2019 Data	148
Figure 4.51	EIP Suitability Map from ANP Weights with 2019 Data Imposed	148
Figure 4.52	EIP Suitability Layer from F-AHP Weights with 2019 Data	149
Figure 4.53	EIP Site Suitability Map from F-AHP Weights with 2019 Data Overlaid	150
Figure 4.54	EIP Suitability Layer from HN-FAP Weights with 2009 Data	152
Figure 4.55	EIP Suitability Layer from HN-FAP Weights with 2009 Data Overlaid	153

Figure 4.56	EIP Suitability Layer from HNAP Weights with 2009 Data	153
Figure 4.57	EIP Suitability Layer from HNAP Weights with 2009 Data Overlaid	154
Figure 4.58	EIP Suitability Layer from H-FAP Weights with 2009 Data	154
Figure 4.59	EIP Suitability Layer from H-FAP Weights with 2009 Data Overlaid	155
Figure 4.60	EIP Suitability Layer from NFh-AP Weights with 2009 Data	155
Figure 4.61	EIP Suitability Layer from NFh-AP Weights with 2009 Data Overlaid	156
Figure 4.62	EIP Suitability Layer from HN-FAP Weights with 2019 Data	157
Figure 4.63	EIP Suitability Layer from HN-FAP Weights with 2019 Data Overlaid	157
Figure 4.64	EIP Suitability Layer from HNAP Weights with 2019 Data	158
Figure 4.65	EIP Suitability Layer from HNAP Weights with 2019 Data Overlaid	159
Figure 4.66	EIP Suitability Layer from H-FAP Weights with 2019 Data	160
Figure 4.67	EIP Suitability Layer from H-FAP Weights with 2019 Data Overlaid with Roads, Residential and Existing Industries	160
Figure 4.68	EIP Suitability Layer from NFh-AP Weights with 2019 Data	161
Figure 4.69	EIP Suitability Layer from NFh-AP Weights with 2019 Data Overlaid	162

LIST OF ABBREVIATIONS

AHP	-	Analytic Hierarchy Process
AI	-	Artificial Intelligence
ANN	-	Artificial Neural Network
ANP	-	Analytical Network Process
ArcGIS	-	Aeronautical Reconnaissance Coverage Geographic Information System
ArcMap	-	Aeronautical Reconnaissance Coverage Map
BF-EIP	-	Brownfield Eco-Industrial Park
BFIP	-	Brownfield Industrial Park
BIP	-	Brownfield Industrial Park
BOCR	-	Benefits, Opportunities, Costs, And Risks
CI	-	Consistency Index
CR	-	Consistency Ratio
DEFRA	-	Department for Environment, Food and Rural Affairs
DELPHI	-	Documentation and Exchange of Lively and Pure Homeopathic Information
DEMATEL	-	Decision Making Trial and Evaluation Laboratory
DSS	-	Decision Support System
EIA	-	Environmental Impact Assessment
ELECTRE	-	Elimination of Choice Translating Reality
ETM	-	Enhanced Thematic Mapper
EVec	-	Eigenvector
EZ	-	Economic Zone
F-AHP	-	Fuzzy Analytic Hierarchy Process
FDI	-	Foreign Direct Investment
FTN	-	Fuzzy Trapezium Numbers
GA	-	Genetic Algorithm
GF-EIP	-	Greenfield Eco-Industrial Park
GFIP	-	Greenfield Industrial Park
GIS	-	Geographic Information System
GIZ	-	German Development Cooperation

GNU	-	GNU's Not Unix
GPS	-	Global Positioning System
GRA	-	Grey Relational Analysis
GT	-	Geospatial Technology
GTP	-	Green Technology Park
H-FAP	-	Hierarchy-Fuzzy Analytic Process
HMCDM	-	Hybrid Multi-Criteria Decision-Making
HNAP		Hierarchy Network Analytic Process
HN-FAP	-	Hierarchy Network-Fuzzy Analytic Process
IC	-	Industrial Corridor
IP	-	Industrial Park
IZ	-	Industrial Zones
KIS	-	Kalundborg Industrial Symbiosis
LULC	-	Land Use Land Cover
MADM	-	Multi-Attribute Decision Making
MAUT	-	Multi-Attribute Utility Theory
MCDA	-	Multi-criteria Decision Attribute
MCDM	-	Multi-criteria Decision-making
MODM	-	Multi-objective Decision making
MOO	-	Multi-objective Optimisation
MOORA	-	Multi-objective Optimization Ratio Analysis
NFh-AP	-	Network Fuzzy hierarchy Analytic Process
NISP	-	British National Industrial Symbiosis Program
NVec	-	New Vector
OLI	-	Operational Land Imager
OPVec	-	Overall Priority Vector
PROMETHEE	-	Preference Ranking Organization Method for Enrichment Evaluations
PVec	-	Priority Vector
R & D	-	Research and Development
RE	-	Renewable Energy
RI	-	Random Index
SA	-	Sensitivity Analysis
SAW	-	Simple Additive Weighting

-	Sustainable Development Goals
-	State Environmental Protection Agency
-	Single Multi-Criteria Decision-Making
-	Shuttle Radar Terrain Mission
-	Triangular Fuzzy Number
-	Traditional Industry
-	Tanjung Langsat Brownfield Industrial Area
-	Tanjung Langsat Industrial Area
-	Technique for Order Preference by Similarity to the Ideal Solution
-	United Nations Industrial Development Organisation
-	United States Geological Survey
-	Viekriterijumsko Kompromisno Rangiranje
-	World Bank Group
-	Weighted Linear Combination
-	Weighted Overlay Analysis
-	Weighted Sum Model

LIST OF SYMBOLS

\tilde{r}_1	-	Geometric Mean
\widetilde{W}_1	-	Fuzzy Weight
M _i	-	Crisp Value
$\tilde{\alpha}_{in}$	-	Fuzzy Dimension Number
.shp	-	Shapefile
.tif	-	Image files
dpi	-	Dots Per Inch
k	-	Number of successive powers a weighted supermatrix is raised
		Tubbu
k→∞	-	Number of Successive Powers to Infinity
k→∞ km	-	Number of Successive Powers to Infinity Kilometre
k→∞ km kth	- - -	Number of Successive Powers to Infinity Kilometre Steady State Raised to a Power
$k \rightarrow \infty$ km kth <i>l</i>	- - -	Number of Successive Powers to Infinity Kilometre Steady State Raised to a Power Lower Value
$k \rightarrow \infty$ km kth l m	- - -	Number of Successive Powers to Infinity Kilometre Steady State Raised to a Power Lower Value Medium value
$k \rightarrow \infty$ km kth l m u	- - - -	Number of Successive Powers to Infinity Kilometre Steady State Raised to a Power Lower Value Medium value Upper Value
$k \rightarrow \infty$ km kth l m u N_i	- - - -	Number of Successive Powers to Infinity Kilometre Steady State Raised to a Power Lower Value Medium value Upper Value Normalised Weight

LIST OF APPENDICES

APPENDIX	TITLE	PAGE
Appendix A	Fundamental Numerical Scale for AHP and F-AHP Criteria Selection Index	193
Appendix B	Random Index Number of Criteria	194
Appendix C	Analytical Hierarchy Process	195
Appendix D	Analytical Network Process	210
Appendix E	Fuzzy Analytical Hierarchy Process	214
Appendix F	Integration of Analytic Hierarchy Process, Analytic Network Process and Fuzzy Analytic Hierarchy Process	226
Appendix G	Standard Deviation	232

CHAPTER 1

INTRODUCTION

1.1 Background

A Traditional Industry (TI) is a production method of the pre-industrial period which was carried out on a small scale within a family in a small workshop or space and located in cities, characterised by spillages, wastes and noise pollution (Moreau et al., 2017). Due to these problems, the need for TIs to be placed in one area out of the city was planned and referred to as Industrial Park (IP) where all factories were moved. As a result of the lack of suitable locations, pollution management from isolated traditional industries, regulatory challenges arise, and industrial parks are eventually abandoned (Neves et al., 2020). IP started in Trafford, Manchester, England in mid-1896 (Beers et al., 2019). IP is described as an area of land out of a city partitioned and formed into plots with or without constructed factories and basic amenities by a team of captains of industry (UNIDO, 2016). Based on product demand and economic expansion, IP gradually spread to Naples, Italy, in 1904, Clearing Industrial District near Chicago, the United States of America in 1907, Singapore in 1951, Germany, the Netherlands, Austria, and cities in the Nordic countries of Sweden, Denmark and Norway, and to other parts of the world (Beers et al., 2019). The Environmental Impact Assessment (EIA) is the common practice of assessing areas for industrial development (Sarmiento & Vargas-Berrones, 2018). EIA is used for the identification and evaluation of the potential effects and consequences of proposed projects, programs or policy actions relative to the physical, cultural and economic components of the total environment (Loomis & Dziedzic, 2018). The EIA process is timeconsuming and produces inaccurate results because of the number of dependents, independents, manual, and incomplete acquisition of variables associated with industrial locations (Loomis & Dziedzic, 2018). As a result, most IPs were sited in unsuitable locations due to a lack of scientific, geospatial, or multi-criteria decisionmaking criteria weight assessment methods for conducive industrial sites, resulting in

brownfield industrial park (BFIP) (Kolhoff et al., 2018). A BFIP is "an abandoned or underutilised existing industrial site where industries, resources and services are disconnected and therefore lacks industrial resource exchange" (Klusacek *et al.*, 2018); (Giamalaki & Tsoutsos, 2019); (Massard et al., 2018) and (UNIDO, 2021).

Most IPs lacked suitable locations for groups of industry synergy, water and wastewater treatment methods, waste and pollution control systems, process automation, energy and material efficiency, and infrastructure, therefore, they emit carbon. To drive industrial dynamics, IPs did not attract Foreign Direct Investment (FDI) due to the absence of these elements (Torabi-Kaveh et al., 2016). IP brought about negative environmental and social impacts including greenhouse gas (GHG) emissions, related public interference, and high operating costs (Doorga et al., 2019). At that stage, since the traditional industries cannot manage the industrial and environmental guidelines geared towards abating greenhouse gas (GHG), the search for Economic Zones (EZ) or Industrial Corridors (IC) began. An EZ is designed to be a top-down and carefully selected industrial district, which can provide economic and regulatory advantages to companies located in its site, protect the environment and the social wellbeing (Stucki et al., 2019). The EZ is divided into several types one of which is the Eco-Industrial Park (EIP). An EIP is a new type of industrial organisation based on the circular economy for optimisation and sustainable development in which byproducts and waste are recycled as raw materials to another company in the park and optimised for sustainable development. UNIDO, (2016) defined an EIP as "a concentration of clusters or interconnected manufacturing, engineering, and mutual service companies or industries located in a favourable site and linked by sharing products, by-products and a common management in the pursuit for green, profitable and social activity through a partnership in handling environmental and resource issues". EIPs must meet some environmental, social, economic, and technical conditions.

The EIP location is heavily influenced by spatial criteria, which necessitate strong decision-making criteria weight assessments and ranking (Giamalaki & Tsoutsos, 2019). As a result, the challenges of socio-economic, technical, and environmental issues caused by segregated factories in an IP or brownfield can be

addressed through a carefully selected area that carries the features for EIP development (Maiolo & Pantusa, 2018). The industrial clusters constraint inspires GHG and the resultant global warming (Sarmiento & Vargas-Berrones, 2018). The key EIP site feature is a suitable location for industry cluster synergy and symbiosis for cleaner production which the IP lacked and failed (Asadabadi et al., 2019). EIP site selection is most effective when many criteria are used to investigate a wide variety of information about the area (Piengang et al., 2019). The suitable location reflects on water bodies, scalable available land, proven infrastructural development (for example, roads, railways, airports, electric grid, and seaports), utilities/amenities (such as electricity, portable water and telecommunication facilities), and existing industries for training (Belaud et al., 2019). Other features include proximity to raw materials, restricted areas (such as mining camps, agricultural farms, wetlands, slopes and mountains), institutions such as religious, health, financial, and academic for research and development (R&D) (Chumaidiyah et al., 2020). Amongst others are the proximity to urban settlement for the search of skilled and unskilled labour, closeness to the market for raw materials, by-products, and finished products supplies (Ajibade et al., 2019). Other crucial factors are the presence and proximity to coastal areas, favourable climatic conditions which can supply sufficient annual rainfall, wind and solar radiation to supplement the generation of clean energy (Geng et al., 2016). The EIP site selection and design are a strategic economic growth problem-solving and the initial process to an industrial carbon emission control and reduction from brownfields. The systematic site selection connects the bridge between favourable locations and several separate industry clusters to synergise and resolve resource management and pollution problems and produce solutions to abate carbon emissions. It links location, innovation, technology, and research to provide competitive advantage across environmental, economic, social and technical aspects through cleaner production and services (Neves et al., 2020). Therefore, the EIP site selection procedure becomes an intricate multi-criteria study.

It is estimated that around 80% of the data used for EIP site selection decisionmaking is spatial and the rest is 20% non-spatial (Das & Gupta, 2021). This was not taken into account when industrial parks were selected (Donni et al., 2017). This meant that there was limited research available on suitable industrial sites, resulting in poor site selection and the emergence of brownfields. The high percentage of spatial criteria required for suitable EIP site selection makes it a complex multi-criteria study that demands the use of geospatial and strong multi-criteria decision-making (MCDM) technologies for criteria assessment and suitability selection (Bansal *et al.*, 2017). Geospatial technology is "the range of modern machinery that is used to obtain, stock, and operate geographic data that is positioned to the earth. The data is used to analyse, model, simulate and imagine the location data for human, environment and the earth" (Avtar *et al.*, 2019). The technology gives well-informed choices based on the status and precedence of sites (Yatim, Ngan, & Lam, 2017). Geospatial technology takes forms such as the Geographic Information System (GIS), which gives a completely different way in which maps are formed to manage our societies and industry's suitability locations (Loomis & Dziedzic, 2018).

The MCDM approach employs decision support systems (DSS) tools that quantitatively analyse, weighs and ranks the importance of a criterion for a specific project site selection (Rahmat et al., 2017). MCDM applies to different procedures that aid decision-makers to discover improved answers, where the purpose is to use the value-oriented method and generate the decision choices of criteria and/or attributes as the essential component in the industrial site selection decision study. MCDM estimates normalised data which achieves evaluations in suitability sites including industrial park location issues where it is crucial to correlate a few qualitative and quantitative measures in an extremely unspecified and unclear location (Zarin et al., 2021a).

MCDM tools are numerous which include the Analytical Hierarchy Process (AHP) (Wind & Saaty, 1980), Fuzzy Analytic Hierarchy Process (F-AHP) (Ohnishi *et al.*, 2017), Analytical Network Process (ANP) (Gnanasekaran & Venkatachalam, 2019), Weighted Linear Combination (WLC), Technique for Order Preference by Similarity to the Ideal Solution (TOPSIS) (Feyzi *et al.*, 2019), and Elimination of Choice Translating Reality (ELECTRE) (Ohnishi *et al.*, 2017) as few examples. MCDM tools have been widely employed to explore site selection results and to establish the ideal choice of project locations. However, previous studies have established that traditional single multi-criteria decision-making (SMCDM) methods have limitations (Donni et al., 2017; Qin et al., 2020) that often give inconsistent

criteria weight and ranking for site suitability selection. Most MCDM tools struggle in the assessment of criteria that are near in weight, or dependent (Liu & Ma, 2021), for example, TOPSIS and F-AHP do not link criteria, while AHP and ANP do not resolve the uncertainty among criteria (Tavana et al., 2017).

When the number of criteria exceeds three, the consistency ratio frequently goes beyond the threshold of 0.1 (Paul, 2015), making its reliability uncertain and reasonable ranking difficult (Rahmat et al., 2017). These have shown that SMCDM techniques in the criteria weight assessment for EIP site suitability selection each have their capabilities and weaknesses. To overcome these, therefore, integrating the SMCDM approaches (Ahmed et al., 2020) can work out the limitations to eliminate the weaknesses and consolidate their strengths (Qin et al., 2020). When an SMCDM technique is combined with two or more methods, integrated multi-criteria decision-making (IMCDM) method is created (Walls & Paquin, 2015; Chumaidiyah et al., 2020;). The integration utilises the strength of SMCDM from different groupings to effectively and objectively evaluate consistent criteria and attribute eigenvectors for EIP site suitability selection. Osra & Kajjumba, (2019); Chumaidiyah et al., (2020) reiterated that the academic study of EIP site suitability selection using a robust approach for consistent criteria weight assessment should commence preventing any EIP site from being a brownfield.

1.2 Problem Statement

Brownfield industrial parks (BFIP) as manufacturing parks that are partially inhabited, underutilised, or derelict, are found all over the world which emit carbon to the environment. BFIPs suffer a variety of issues, such as unsuitable location, insufficient expansion area, shortage of existing industries, isolation, waste and wastewater management challenges, material and energy inefficiencies, and a lack of information (Qin et al., 2020). BFIPs generate GHG that pollute the environment, risk human health, destroy flora and fauna, and climate change and contribute to global warming. Unsuitable locations (Luthra et al., 2020) are the leading causes of BFIP, which emerge mostly as a result of insufficient or absent site criteria evaluation or the use of single traditional assessment procedures that struggle with spatial and consistent decision-making abilities. Many researchers have reported that single techniques have site criteria weights assessment consistency problems, which can result in wrong industrial sites suitability choices (Paul, 2015); (Rahmat et al., 2017); (Ahmed et al., 2020). It is a fact that the application of MCDM in the site criteria weighting and ranking has a significant impact on the selection of suitable industrial locations (Asadabadi et al., 2019).

The 2015 Paris Agreement committed governments to keep global temperature below 1.5°C by embarking on green manufacturing techniques to reduce GHG emissions and reduce global warming. Since BFIP emit carbon dioxide, methane, and nitrous oxides, which contribute to 28% of global GHG emissions (IPCC, 2019), it was recommended that they be transformed into eco-industrial parks (EIP). The mission of the EIP is to bring together industries in a strategic location for a circular economy, which is a new model that strives to systematically emulate natural symbiotic concepts of reducing, reusing, and recycling resources for cleaner manufacturing. Symbiosis minimises raw material consumption in EIPs by encouraging energy and waste reuse while also improving material efficiency and industry competitiveness among clusters. Synergy in EIPs promotes the sites' economic, environmental, social, and technological advantages, all of which contribute to minimizing the overall carbon footprint from industrial activities.

As BFIPs are being mapped to be converted to EIPs for industrial symbiosis, a detailed investigation of the spatial brownfield sites criteria using GIS and integrated MCDM methods, which are currently lacking, is required to determine the suitability of the location. Criteria such as favourable geographic proximity to urban/residential and industrial locations, a suitable climate for renewable energy resources, and accessible transportation and utility infrastructure are necessary. Other factors include the availability of water bodies, labour and markets, stable political areas, and available land for industry and development.

Since each SMCDM tool has a specific goal, it may not be suitable to be used to evaluate goals that they are not designed for. These make SMCDM tools constrained in assessing spatial variables that are nearly identical in rank and/or connected, and inconsistency threshold index when several criteria are employed. In this study, three MCDM procedures (AHP, ANP, and F-AHP) are to be assessed and integrated to build a consistent multi-criteria decision-making tool that can produce dependable criteria weights for selecting BFIP to a suitable EIP site. The integrated multi-criteria decision-making (IMCDM) algorithm would overcome the shortcomings of each SMCDM tool and improve its strengths to accurately assess brownfield spatial criteria weight for EIP site selection. The GIS with its power of collecting, evaluating, producing, and storing spatial criteria will be used to capture the spatial criteria of a selected brownfield. An MCDM-GIS model will be developed to run the results of the SMCDM and IMCDM methods to compare the criteria weight consistencies by the suitability layers of the BFIP for decision-making to EIP site conversion.

1.3 Research Goal

The study aims to develop an IMCDM algorithm and an MCDM-GIS model that can assess consistent criteria weights and accurately select brownfields for suitable and sustainable EIP sites.

The objectives of the research are:

- (a) To establish the weightage for ranking for the selection of BF-EIP site using the SMCDM (AHP, ANP and F-AHP). To subsequently integrate the SMCDM methods to create an IMCDM algorithm weighting process for the assessment of a consistent criteria weight for the selection of a BF-EIP site.
- (b) To design an MCDM-GIS model for the weighted overlay analysis of the SMCMD and IMCDM weights and spatial criteria.
- (c) To use GIS to collect and prepare the spatial criteria to test the SMCDM and IMCDM methods weight assessment consistencies and ranking, and the

resilience of the model in the selection of brownfield for conversion to an EIP site.

1.4 Scope of Work

To achieve the specific objectives of the research, the scope of the work is as follows:

- (a) Review and identify the brownfield industrial area spatial criteria.
- (b) Use the AHP, ANP, and F-AHP SMCDM to evaluate the weight percent of each criterion.
- (c) Integrate the SMCDM methods to create an IMCDM algorithm and use them to assess the criteria/alternative weights of importance.
- (d) Evaluate the sensitivity analyses and standard deviations of both the SMCDM and IMCDM weights.
- (e) Design an MCDM-GIS model for the overlay of the spatial criteria and criteria weights assessed by the SMCDM and IMCDM methods to simplify and accurately select brownfield sites for EIP.
- (f) Use the GIS to obtain 2009 and 2019 (a ten-year interval) spatial criteria data of Tanjung Langsat Brownfield Industrial Area (TLBIA), analyse, classify, and store them. Obtain the land use land cover (LULC) data of the TLBIA from PLANMalaysia.
- (g) Prepare the Euclidean distance by assigning the desired distances (km), and non-distance criteria in percent (%) and degree Celsius (°C). Reclassified the Euclidean distance raster layers by considering a scale of 1 to 5 (5 being the preferred whether near or far, and 1 unpreferred whether close or farther).

- (h) Perform weighted overlay analysis (WOA) of the SMCDM and IMCDM separately with the 2009 spatial criteria data and LULC in the MCDM-GIS model. Examine the consistencies of the SMCDM and IMCDM techniques and the performance of the model in each case.
- (i) Further, perform WOA using the SMCDM and IMCDM algorithm weights with the 2019 spatial data. Compare the weight assessment consistencies and the resilience of the model in all cases for the conversion of a brownfield site to EIP.

1.5 Significance of the Study

The importance of this study is

- (a) Since SMCDM methods have limitations in the assessment of consistent criteria weights for industrial site selection, as reported by many studies which have brought about the emergence of abandoned and/or underutilised industrial parks, integrated MCDM algorithms are required to provide assessments of consistent criteria weights upon which the EIP site selection depends.
- (b) The IMCDM algorithm will provide reliable weights and the MCDM GISbased model will make the EIP site selection easy to stimulate the redevelopment of brownfield to EIP. This will promote EIP development to mitigate industrial emissions for the global reduction of GHG to 1.5°C or less by 2030 as mandated by the 2015 Paris Agreement.
- (c) The design of the integrated algorithm and the model will help spur governments, brownfield-EIP developers, and research students in brownfield EIP site selection activities to attract investors in brownfield-EIP development.

1.6 Thesis Outline

This thesis comprises five chapters as follows:

Chapter 1 begins with an overview of the study, which is the background to study, problem statement, research objectives, the scope of work, and significance of the study.

Chapter 2 presents the critical review of the previous literature on geographic information system and multi-criteria decision-making technologies used in the selection of industrial sites. This chapter also highlights the problems of SMCMD tools in criteria weight assessment, the challenges of the GIS application for EIP site selection, the application of MCDM issues on EIP site selection, the impact of GIS on land use and EIPs. The chapter also discusses the AHP, ANP, and F-AHP weighting tools. The EIP concept, objectives, planning, development, challenges, the categories, and types of EIPs, EIP development projects in Kalundborg, the United Kingdom and around the world are also discussed. The chapter finally addressed the research gap based on the literature review.

Chapter 3 explains the research method for the EIP site selection. This consists of outlining the software and tools used and data collection. There is the criteria construction of the structures, formation of the pairwise comparison matrices, supermatrix, triangular fuzzy numbers and criteria/attribute weight assessments of the single/traditional AHP, ANP, and F-AHP tools. The chapter also performs normalisation, consistency confirmation evaluation, the overall priority ranking and sensitivity analysis of the weight outcomes. This chapter also deals with the integration of the AHP, ANP, and F-AHP, and evaluates the standard deviation of the criteria weight outcomes. The processing of the criteria spatial data by GIS, which includes Boolean logic criteria screening and classification, conversion and preparation of Euclidean distance raster layers, reclassification of the raster layers, and land use land cover layers acquisition. Finally, the chapter explains the development of the MCDM GIS-based algorithm model, testing the model, generating, and selecting the suitable EIP site. Chapter 4 presents and discusses all the outcomes from the application of the methods described in chapter 3 and provides the best-integrated algorithms for the evaluation of the consistent criteria weight of importance for use and a guide to a suitable BF-EIP site selection.

Chapter 5 summarises the research findings, enumerates the contributions to knowledge and recommendations for future works.

REFERENCES

- Abdulhasan, M. J., Hanafiah, M. M., Satchet, M. S., Abdulaali, H. S., Toriman, M. E., & Al-Raad, A. A. (2019). Combining gis, fuzzy logic, and AHP models for solid waste disposal site selection in Nasiriyah, Iraq. *Applied Ecology and Environmental Research*, 17(3), 6701–6722. https://doi.org/10.15666/aeer/1703_67016722
- Abdullah, L., & Adawiyah, C. W. (2014). Simple Additive Weighting Methods of Multi-criteria Decision Making and Applications: A Decade Review. *International Journal of Information Processing and Management(IJIPM)*, 5(1), 39–49. http://link.springer.com/10.1007/s10764-012-9584-5
- Adair, A., Berry, J., McGreal, S., Deddis, B. and Hirst, S. (2000). The Financing of Urban Regeneration. *Journal of Land Use Policy*, *17*(2), 147–156. https://doi.org/10.1016/S0264-8377(00)00004-1
- Afshari, H., Farel, R., & Peng, Q. (2018). Challenges of value creation in Eco-Industrial Parks (EIPs): A stakeholder perspective for optimizing energy exchanges. *Resources, Conservation and Recycling*, 139, 315–325. https://doi.org/10.1016/j.resconrec.2018.09.002
- Ahmad, N.; Zhu, Y.; Ibrahim, M.; Waqas, M.; Waheed, A. (2018). Development of a Standard Brownfield Definition, Guidelines, and Evaluation Index System for Brownfield Redevelopment in Developing Countries: The Case of Pakistan. *Journal of Sustainability*, 10, 4347–4359. https://doi.org/10.3390/su10124347
- Ahmadipari, M., Hoveidi, H., Jafari, H. R., & Pazoki, M. (2018). An integrated environmental management approach to industrial site selection by genetic algorithm and fuzzy analytic hierarchy process in a geographical information system. In *Global Journal of Environmental Science and Management*, 4(3), 339–350). https://doi.org/10.22034/GJESM.2018.03.007
- Ahmed, W., Tan, Q., Solangi, Y. A., & Ali, S. (2020). Sustainable and special economic zone selection under fuzzy environment: A case of Pakistan. *Symmetry*, *12*(2). https://doi.org/10.3390/sym12020242
- Ajibade, F. O., Olajire, O. O., Ajibade, T. F., Nwogwu, N. A., Lasisi, K. H., Alo, A. B., Owolabi, T. A., & Adewumi, J. R. (2019). Combining multicriteria decision analysis with GIS for suitably siting landfills in a Nigerian state. *Environmental and Sustainability Indicators*, 3(4), 100010. https://doi.org/10.1016/j.indic.2019.100010
- Aktaş, E., & Türkyılmaz Tahta, B. (2018). Investigation of suitable land use potential for industrial sites: the case of Kemalpaşa. *Environmental Monitoring and Assessment*, 190(11). https://doi.org/10.1007/s10661-018-7007-6
- Akther, A., Ahamed, T., Noguchi, R., Genkawa, T., & Takigawa, T. (2019). Site suitability analysis of biogas digester plant for municipal waste using GIS and multi-criteria analysis. *Asia-Pacific Journal of Regional Science*, 3(1), 61–93. https://doi.org/10.1007/s41685-018-0084-2
- Algohary, S., & Aly, A. L. Y. I. M. (2018). A Proposal for Using Zafarana Area for

Siting of a Hybrid Wind and Nuclear Power Plant in Egypt. *International Journal of Renewable Energy Sources*, 3, 10–19.

- Alqaderi, M. B., Emar, W., & Saraereh, O. A. (2018). Concentrated solar power site suitability using GIS-MCDM technique taken UAE as a case study. *International Journal of Advanced Computer Science and Applications*, 9(4), 261–268. https://doi.org/10.14569/IJACSA.2018.090440
- Arabsheibani, R., Kanani Sadat, Y., & Abedini, A. (2016). Land suitability assessment for locating industrial parks: a hybrid multi-criteria decision-making approach using Geographical Information System. *Journal of Geographical Research*, 54(4), 446–460. https://doi.org/10.1111/1745-5871.12176
- Arsovski, S., Todorovic, G., Lazić, Z., Arsovski, Z., Ljepava, N., & Aleksic, A. (2017).
 Model for Selection of the Best Location-Based on Fuzzy AHP and Hurwitz Methods. *Mathematical Problems in Engineering*, 2017. https://doi.org/10.1155/2017/2803461
- Asadabadi, M. R., Chang, E., & Saberi, M. (2019). Are MCDM methods useful? A critical review of the Analytic Hierarchy Process (AHP) and Analytic Network Process (ANP). *Cogent Engineering*, 6(1), 1–11. https://doi.org/10.1080/23311916.2019.1623153
- Avtar, R., Sahu, N., Aggarwal, A. K., Chakraborty, S., Kharrazi, A., Yunus, A. P., Dou, J., & Kurniawan, T. A. (2019). *Remote Sensing and GIS — A Review*. *Resources* 8(3), 149. https://doi.org/10.3390/resources8030149
- Aydi, A., Abichou, T., Nasr, I. H., Louati, M., & Zairi, M. (2016). Assessment of land suitability for olive mill wastewater disposal site selection by integrating fuzzy logic, AHP, and WLC in a GIS. *Environmental Monitoring and Assessment*, 188(1), 1–13. https://doi.org/10.1007/s10661-015-5076-3
- Azizi, A., Malekmohammadi, B., Jafari, H.R., Nasiri, H. and Parsa, V. A. (2014). Land suitability assessment for wind power plant site selection using ANP-DEMATEL in a GIS environment: a case study of Ardabil province, Iran. *Environmental Monitoring and Assessmental*, 86(10), 6695–6709. https://doi.org/10.1007/s10661-014-3883-6
- Babalola, M. (2018). Application of GIS-Based Multi-Criteria Decision Technique in Exploration of Suitable Site Options for Anaerobic Digestion of Food and Biodegradable Waste in Oita City, Japan. *Environments*, 5(7), 77. https://doi.org/10.3390/environments5070077
- Badri Ahmadi, H., Hashemi Petrudi, S. H., & Wang, X. (2017). Integrating sustainability into supplier selection with analytical hierarchy process and improved grey relational analysis: a case of the telecom industry. *International Journal of Advanced Manufacturing Technology*, 90(9–12), 2413–2427. https://doi.org/10.1007/s00170-016-9518-z
- Bansal, A., Kumar, B., & Garg, R. (2017). Multi-criteria decision-making approach for the selection of software effort estimation model. *Management Science Letters*, 7(6), 285–296. https://doi.org/10.5267/j.msl.2017.3.003
- Barzehkar, M., Dinan, N. M., Mazaheri, S., Tayebi, R. M., & Brodie, G. I. (2019). Landfill site selection using GIS-based multi-criteria evaluation (case study: SaharKhiz Region located in Gilan Province in Iran). SN Applied Sciences, 1(9),

1-11. https://doi.org/10.1007/s42452-019-1109-9

- Beames, A., Broekx, S., Schneidewind, U., Landuyt, D., van der Meulen, M., Heijungs, R., & Seuntjens, P. (2018). Amenity proximity analysis for sustainable brownfield redevelopment planning. *Landscape and Urban Planning*, 171, 68– 79. https://doi.org/10.1016/j.landurbplan.2017.12.003
- Beers, D. van, Flammini, A., Meylan, F. D., & Stucki, J. (2019). Lessons learned from the application of the UNIDO eco-industrial park toolbox in Viet Nam and other countries. *Sustainability* (*Switzerland*), *11*(17), 1–18. https://doi.org/10.3390/su11174687
- Behera, S. K., Kim, J. H., Lee, S. Y., Suh, S., Park, H. S. (2012). Evolution of 'designed' industrial symbiosis networks in the Ulsan Eco-industrial Park: 'research and development into businesses as the enabling framework. *Journal of Cleaner Production*, 29(30), 103–112. https://doi.org/10.1016/j.jclepro.2012.02.009
- Belaud, J. P., Adoue, C., Vialle, C., Chorro, A., & Sablayrolles, C. (2019). A circular economy and industrial ecology toolbox for developing an eco-industrial park: perspectives from French policy. *Clean Technologies and Environmental Policy*, 21(5), 967–985. https://doi.org/10.1007/s10098-019-01677-1
- Brans, J.P., Mareschal, B., Vincke, P. (1984). PROMETHEE: A new family of outranking methods in multicriteria analysis. *Oper. Res.*, *3*, 408–421.
- Brunelli, M. (2015). *Introduction to the Analytic Hierarchy Process* (2nd Ed). Springer, U. S. A. https://doi.org/10.1007/978-3-319-12502-1
- Buckley, J. J. (1985). Fuzzy hierarchical analysis. *Fuzzy Sets and Systems*, 17(3), 233–247. https://doi.org/10.1016/0165-0114(85)90090-9
- Büyüközkan, G., & Feyzioğlu, O. (2004). A new approach based on soft computing to accelerate the selection of new product ideas. *Computers in Industry*, 54(2), 151–167. https://doi.org/10.1016/j.compind.2003.09.007
- Ceballos, B., Lamata, M. T., & Pelta, D. A. (2016). A comparative analysis of multicriteria decision-making methods. *Progress in Artificial Intelligence*, *5*(4), 315– 322. https://doi.org/10.1007/s13748-016-0093-1
- Chang, D. Y. (1996). Applications of the extent analysis method on fuzzy AHP. *European Journal of Operational Research*, 95(3), 649–655. https://doi.org/10.1016/0377-2217(95)00300-2
- Chang, P., & Lin, H. (2015). Manufacturing Plant Location Selection in Logistics Network Using Analytic Hierarchy Process. 8(5), 1547–1575. doi:10.3926/jiem.1456
- Chang, P. Y., Lin, H. Y., Fataei, E., erdi, M. A., Farhadi, H., Mohammadian, A., Rikalovic, A., Cosic, I., Labati, R. D., Piuri, V., Chang, P. Y., Lin, H. Y., Feiz, R., Rikalovic, A., Cosic, I., Labati, R. D., & Piuri, V. (2015). A comprehensive method for industrial site selection: The macro-location analysis. *IEEE Systems Journal*, 8(5), 1547–1575. https://doi.org/10.3384/lic.diva-105942
- Chen, S. J., & Chen, S. M. (2007). Fuzzy risk analysis based on the ranking of generalized trapezoidal fuzzy numbers. *Applied Intelligence*, 26(1), 1–11. https://doi.org/10.1007/s10489-006-0003-5

- Chen, W., Zhu, Y., Yang, M., & Yuan, J. (2017). Optimal site selection of wind-solar complementary power generation project for a large-scale plug-in charging station. *Sustainability (Switzerland)*, 9(11). https://doi.org/10.3390/su9111994
- Chertow, M. (2013). Taking Action on Materials: Global Examples of By-Product Synergy. US Business Council for Sustainable Development. 87-116. https://doi.org/10.1007/978-3-319-20571-7_5
- Chou, S. Y., Chang, Y. H., & Shen, C. Y. (2008). A fuzzy simple additive weighting system under group decision-making for facility location selection with objective/subjective attributes. *European Journal of Operational Research*, 189(1), 132–145. https://doi.org/10.1016/j.ejor.2007.05.006
- Chumaidiyah, E., Dewantoro, M. D. R., Hakimah, D. A., Arffan, Z., & Robbi, R. M. N. (2020). Measurement of Criterion Weight to Determine Industrial Area Location Using AHP for Economic Growth. *IOP Conference Series: Materials Science and Engineering*, 1, 1003. https://doi.org/10.1088/1757-899X/1003/1/012154
- Cotic, B. (2021). Industrial Symbiosis in Brownfields in Kranj, Slovenia. 18th IOP Conf. Series: Materials Science and Engineering, May 30 - June 3 2021, 471. https://doi.org/10.1088/1757-899X/471/11/112073
- Cruz, R. B. C. da, Marins, K. R. de C., & Kurokawa, F. A. (2021). Multicriteria methodological-rational model to evaluate urban areas: A case study of the São Paulo City/Brazil. Sustainable Cities and Society, 67. https://doi.org/10.1016/j.scs.2021.102718
- Cui, D., Deng, Z., & Liu, Z. (2019). China's non-fossil fuel CO2 emissions from industrial processes. *Applied Energy*, 254(August), 113537. https://doi.org/10.1016/j.apenergy.2019.113537
- Danesh, D., Ryan, M. J., & Abbasi, A. (2017). A systematic comparison of multicriteria decision-making methods for the improvement of project portfolio management in complex organisations. *International Journal of Management* and Decision Making, 16(3), 280–320. https://doi.org/10.1504/IJMDM.2017.085638
- Das, S., & Gupta, A. (2021). Geoscience Frontiers Multi-criteria decision based geospatial mapping of flood susceptibility and temporal hydro-geomorphic changes in the Subarnarekha basin. *Geoscience Frontiers*, 12(5), 101206. https://doi.org/10.1016/j.gsf.2021.101206
- de Souza, F. F., Ferreira, M. B., Saraceni, A. V., Betim, L. M., Pereira, T. L., Petter, R. R. H., Pagani, R. N., de Resende, L. M. M., Pontes, J., & Piekarski, C. M. (2020). Temporal comparative analysis of industrial symbiosis in a business network: Opportunities of a circular economy. *Sustainability (Switzerland)*, 12(5), 1–18. https://doi.org/10.3390/su12051832
- Deswal, M., & Laura, J. S. (2018). GIS-based modelling using Analytic Hierarchy Process (AHP) for optimization of landfill site selection of Rohtak city, Haryana (India). *Journal of Applied and Natural Science*, 10(2), 633–642. https://doi.org/10.31018/jans.v10i2.1753
- Di Ludovico, D., & Fabietti, V. (2018). Strategic Environmental Assessment, key issues of its effectiveness. The results of the Speedy Project. *Environmental*

Impact Assessment Review, 68, 19–28. https://doi.org/10.1016/j.eiar.2017.10.007

- Ding, Z., Zhu, M., Wang, Y., & Zhu, J. (2018). An AHP GIS-based model of C&D waste landfill site selection: A triangulation of critical factors. *Proceedings of the* 21st International Symposium on Advancement of Construction Management and Real Estate, 2016, 209889, 163–174. https://doi.org/10.1007/978-981-10-6190-5_16
- Dixon, T., Otsuka, N. and Abe, H. (2011). Critical Success Factors in Urban Brownfield Regeneration: An Analysis of 'Hardcore' Sites in Manchester and Osaka during the Economic Recession. *Journal of Environment and Planning*, 43(4), 961–980. https://doi.org/10.1068/a43468.
- Domenech, T. & Davies, M. (2011). Structure and morphology of industrial symbiosis networks: The case of Kalundborg. *Procedia Social and Behavioural Sciences*, *10*, 79–88. https://doi.org/10.1016/j.sbspro.2011.01.011.
- Dong, H., Ohnishi, S., Fujita, T., Geng, Y., Fujii, M., & Dong, L. (2014). Achieving carbon emission reduction through industrial & urban symbiosis: A case of Kawasaki. *Energy*, 64, 277–286. https://doi.org/10.1016/j.energy.2013.11.005
- Dong, L., Fujita, T., Dai, M., Geng, Y., Ren, J., Fujii, M., Wang, Y., & Ohnishi, S. (2016). Towards preventative eco-industrial development: An industrial and urban symbiosis case in one typical industrial city in China. *Journal of Cleaner Production*, 114, 387–400. https://doi.org/10.1016/j.jclepro.2015.05.015
- Dong, L., Zhang, H., Fujita, T., Ohnishi, S., Li, H., Fujii, M., & Dong, H. (2013). Environmental and economic gains of industrial symbiosis for the Chinese iron/steel industry: Kawasaki's experience and practice in Liuzhou and Jinan. *Journal of Cleaner Production*, 59, 226–238. https://doi.org/10.1016/j.jclepro.2013.06.048
- Donni, M., Siahaan, L., Surbakti, A. B., & Lubis, A. H. (2017). Implementation of Simple Additive Weighting Algorithm in Particular Instance. *International Journal of Scientific Research in Science and Technology*, 3(6), 442–447.
- Dos Santos, P. H., Neves, S. M., Sant'Anna, D. O., Oliveira, C. H. de, & Carvalho, H. D. (2019). The analytic hierarchy process supporting decision making for sustainable development: An overview of applications. *Journal of Cleaner Production*, 212, 119–138. https://doi.org/10.1016/j.jclepro.2018.11.270
- El-Massah, S. (2018). Achieving sustainable industrialisation in Egypt: assessment of the potential for EIPs. *Interdisciplinary Environmental Review*, *19*(1), 31–43.
- Eldrandaly, K. (2013). Developing a GIS-based MCE site selection tool in ArcGIS using COM technology. *International Arab Journal of Information Technology*, *10*(3), 276–282.
- ElMassah, S. (2018). Industrial symbiosis within eco-industrial parks: Sustainable development for Borg El-Arab in Egypt. *Business Strategy and the Environment*, 27(7), 884–892. https://doi.org/10.1002/bse.2039
- Emrouznejad, A., and Ho, W. (2018). *Fuzzy Analytic Hierarchy Process* (Chong, E. A. and Ho, W. (2nd ed.); CRC Press, Taylor & Prancis, Florida, U.S.A.
- Erdogan, S. A., Šaparauskas, J., & Turskis, Z. (2019). A multi-criteria decisionmaking model to choose the best option for sustainable construction management. *Sustainability (Switzerland)*, 11(8). https://doi.org/10.3390/su11082239

- Fang, J., & Partovi, F. Y. (2021). Criteria determination of analytic hierarchy process using a topic model. *Expert Systems with Applications*, 169, 1–13. https://doi.org/10.1016/j.eswa.2020.114306
- Fataei, E., erdi, M. A., Farhadi, H., & Mohammadian, A. (2015). Industrial State Site Selection Using MCDM Method and GIS in Germi, Ardabil, Iran. *Journal of Industrial and Intelligent Information*, 3(4), 324–329. https://doi.org/10.12720/jiii.3.4.324-329
- Feyzi, S., Khanmohammadi, M., Abedinzadeh, N., & Aalipour, M. (2019). Multicriteria decision analysis FANP based on GIS for siting municipal solid waste incineration power plant in the north of Iran. Sustainable Cities and Society, 47(October 2018), 101513. https://doi.org/10.1016/j.scs.2019.101513
- Fischer, T. B., Jha-Thakur, U., Hayes, S. (2015). Environmental impact assessment and strategic environmental assessment research in the UK. *Journal of Environmental Assessment Policy Management*, 17, 150–169.
- Fishburn, P. C. (1968). Utility theory. *Management Science*, 14(5), 335–378. https://doi.org/doi:10.1287/mnsc.14.5.335
- Gao, C., Gao, C., Song, K., Ye, Z., & Dong, J. (2019). Regional water ecosystem risk assessment based on GIS and pollutant diffusion model: A case study of Shenzhen eco-industrial park. *Process Safety and Environmental Protection*, 130, 182–189. https://doi.org/10.1016/j.psep.2019.08.004
- Geng, Y., Fujita, T., Park, H. S., Chiu, A. S. F., & Huisingh, D. (2016). Recent progress on innovative eco-industrial development. *Journal of Cleaner Production*, 114, 1–10. https://doi.org/10.1016/j.jclepro.2015.09.051
- Ghobadi, M., & Ahmadipari, M. (2018). Environmental Planning for Wind Power Plant Site Selection using a Fuzzy PROMETHEE-Based Outranking Method in Geographical Information System. 2(2), 75–87. https://doi.org/10.22097/eeer.2018.148760.1041
- Gnanasekaran, S., & Venkatachalam, N. (2019). A review on applications of multicriteria decision making (MCDM) for solar panel selection. *International Journal* of Mechanical and Production Engineering Research and Development, 9(2), 11–20. https://doi.org/10.24247/ijmperdapr20192
- Harris, P., & Viliani, F. (2018). Strategic health assessment for large scale industry development activities: An introduction. *Environmental Impact Assessment Review*, 68(June 2017), 59–65. https://doi.org/10.1016/j.eiar.2017.10.002
- Hassaan, M. A., Hassan, A., & Al-Dashti, H. (2020). GIS-based suitability analysis for siting solar power plants in Kuwait. *Egyptian Journal of Remote Sensing and Space Science*, 24(3), 453-461. https://doi.org/10.1016/j.ejrs.2020.11.004
- Hayek, M. (2010). Mapping industrial legacies: Building a comprehensive brownfield database in geographic information systems. Planning Practice and Research. 25(4), 461–475. https://doi.org/10.1080/02697459.2010.511018
- Hishammuddin, M. A. H., Ling, G. H. T., Chau, L. W., Ho, C. S., Ho, W. S., & Idris, A. M. (2018). Circular economy (CE): A framework towards sustainable low carbon development in Pengerang, Johor, Malaysia. *Chemical Engineering Transactions*, 63, 481–486. https://doi.org/10.3303/CET1863081
- Hu, W., Tian, J., Zang, N., Gao, Y., & Chen, L. (2019). Study of the development and

performance of centralized wastewater treatment plants in Chinese industrial parks. *Journal of Cleaner Production*, 214, 939–951. https://doi.org/10.1016/j.jclepro.2018.12.247

- Huang, J. Y., & Wey, W. M. (2019). Application of Big Data and Analytic Network Process for the Adaptive Reuse Strategies of School Land. *Social Indicators Research*, 142(3), 1075–1102. https://doi.org/10.1007/s11205-018-1951-y
- Hwang, C.-L., & Yoon, K. (1981). Methods for Multiple Attribute Decision Making. In Multiple Attribute Decision Making. Springer, Berlin, Heidelberg, 58–191. https://doi.org/10.1007/978-3-642-48318-9_3
- Ibrahim, M. F., Hod, R., Toha, H. R., Mohammed Nawi, A., Idris, I. B., Mohd Yusoff, H., Sahani, M. (2021). The Impacts of Illegal Toxic Waste Dumping on Children's Health: A Review and Case Study from Pasir Gudang, Malaysia. Int. J. Environ. Res. Public Health, 18(2221), 2-16. https://doi.org/10.3390/ ijerph18052221
- Indahingwati, A., Barid, M., Wajdi, N., Susilo, D. E., Kurniasih, N. and. Rahim, R. (2017). Comparison Analysis of TOPSIS and Fuzzy Logic Methods on Fertilizer Selection. *International Journal of Engineering Technology*, 23, 109–114.
- IPCC. (2019). Climate Change: Mitigation of Climate Change. Contribution of Working Group III to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Edenhofer, O., R. Pichs-Madruga, Y. Sokona, E. Farahani, S. Kadner, K. Seyboth, A. Adler, I. B. www.cambridge.org / 9781107654815
- Jangid, J., Bera, A. K., Joseph, M., Singh, V., Singh, T. P., Pradhan, B. K., & Das, S. (2016). Potential zones identification for harvesting wind energy resources in a desert region of India – A multi-criteria evaluation approach using remote sensing and GIS. In *Renewable and Sustainable Energy Reviews*. 65, 1-10. https://doi.org/10.1016/j.rser.2016.06.078
- Jozaghi, A., Alizadeh, B., Hatami, M., Flood, I., Khorrami, M., Khodaei, N., & Tousi, E. G. (2018). A comparative study of the AHP and TOPSIS techniques for dam site selection using GIS: A case study of Sistan and Baluchestan Province, Iran. *Geosciences* (*Switzerland*), 8(12), 1–23. https://doi.org/10.3390/geosciences8120494
- Kamali, M., Alesheikh, A. A., Alavi Borazjani, S. A., Jahanshahi, A., Khodaparast, Z., & Khalaj, M. (2017). Delphi-AHP and weighted index Overlay-GIS approaches for industrial site selection case study: Large extractive industrial units in Iran. *Journal of Settlements and Spatial Planning*, 8(2), 99–105. https://doi.org/10.24193/JSSP.2017.2.03
- Kamdar, I., Ali, S., Bennui, A., Techato, K., & Jutidamrongphan, W. (2019). Municipal solid waste landfill siting using an integrated GIS-AHP approach: A case study from Songkhla, Thailand. *Resources, Conservation and Recycling*, 149, 220–235. https://doi.org/10.1016/j.resconrec.2019.05.027
- Kanniah, K. D., Sheikhi, A., Cracknell, A. P., Goh, H. C. Tan, K. P., Ho, C. S., Rasli, F. N. (2015). Satellite Images for Monitoring Mangrove Cover Changes in a Fast Growing Economic Region in Southern Peninsular Malaysia. *Remote Sensing*, 7, 14360–14385. https://doi.org/10.3390/rs71114360

Karimi, H., Herki, B. M. A., Gardi, S. Q., Galali, S., Mirzaei, K., & Pirsaheb, M.

(2020). Site selection and environmental risks assessment of medical solid waste landfill for the City of. *International Journal of Environmental Health Research*, 00(00), 1–13. https://doi.org/10.1080/09603123.2020.1742876

- Khan, M. U. H., Vaezi, M., & Kumar, A. (2018). Optimal siting of solid waste-tovalue-added facilities through a GIS-based assessment. *Science of the Total Environment*, 610(611), 1065–1075. https://doi.org/10.1016/j.scitotenv.2017.08.169
- Khavarian-Garmsir, A. R., & Rezaei, M. R. (2015). Selection of appropriate locations for industrial areas using GIS-fuzzy methods. A case study of Yazd township, Iran. *Journal of Settlements and Spatial Planning*, 6(1), 19–25.
- Klusáček, P., Alexandrescu, F., Osman, R., Malý, J., Kunc, J., Dvořák, P., Frantál, B., Havlíček, M., Krejčí, T., Martinát, S., Skokanová, H., & Trojan, J. (2018). Good governance as a strategic choice in brownfield regeneration: Regional dynamics from the Czech Republic. *Land Use Policy*, 73, 29–39. https://doi.org/10.1016/j.landusepol.2018.01.007
- Kolhoff, A. J., Driessen, P. P. J., & Runhaar, H. A. C. (2018). Overcoming low EIA performance - A diagnostic tool for the deliberate development of EIA system capacities in low and middle-income countries. *Environmental Impact* Assessment Review, 68, 98–108. https://doi.org/10.1016/j.eiar.2017.11.001
- Kubler, S., Robert, J., Derigent, W., Voisin, A., & Le Traon, Y. (2016). A state-of-theart survey & testbed of Fuzzy AHP (FAHP) applications. *Expert Systems with Applications*, 65, 398–422. https://doi.org/10.1016/j.eswa.2016.08.064
- Kucukvar, M., Onat, N. C., & Haider, M. A. (2018). Material dependence of national energy development plans: The case for Turkey and United Kingdom. *Journal of Cleaner Production*, 200, 490–500. https://doi.org/10.1016/j.jclepro.2018.07.245
- Kunc, J., Navrátil, J., Tonev, P., Frantál, B., Klusáček, P., Martinát, S., Havlíček, M. and Černík, J. (2014). Perception of Urban Renewal: Reflexions and Coherences of Socio-Spatial Patterns (Brno, Czech Republic). *Geographia Technica*, 9(1), 66–77.
- Lambert, A. J. D., & Boons, F. A. (2002). Eco-industrial parks: Stimulating sustainable development in mixed industrial parks. *Technovation*, 22(8), 471–484. https://doi.org/10.1016/S0166-4972(01)00040-2
- Lawal, M., Wan Alwi, S. R., Manan, Z. A., & Ho, W. S. (2021). Industrial symbiosis tools—A review. *Journal of Cleaner Production*, 280. https://doi.org/10.1016/j.jclepro.2020.124327
- Leong, Y. T., Tan, R. R., Aviso, K. B., & Chew, I. M. L. (2016). Fuzzy analytic hierarchy process and targeting for inter-plant chilled and cooling water network synthesis. *Journal of Cleaner Production*, 110, 40–53. https://doi.org/10.1016/j.jclepro.2015.02.036
- Li, Y., Hu, Y., Zhang, X., Deng, Y. and Mahadevan, S. (2014). An Evidential DEMATEL Method to Identify Critical Success Factors in Emergency Management. *Journal of Applied Soft Computing*, 5(2), 1235–1251.
- Li, C., Wu, J., & Zeng, T. (2020). Global industrial park research trends: a bibliometric analysis from 1987 to 2016. *Environmental Monitoring and Assessment*, 192(1). https://doi.org/10.1007/s10661-019-7993-z

- Li, H., Dong, L., & Ren, J. (2015). Industrial symbiosis as a countermeasure for resource-dependent city: A case study of Guiyang, China. *Journal of Cleaner Production*, 107, 252–266. https://doi.org/10.1016/j.jclepro.2015.04.089
- Li, X., Bardos, P., Cundy, A. B., Harder, M. K., Doick, K. J., Norrman, J., Williams, S., & Chen, W. (2019). Using a conceptual site model for assessing the sustainability of brownfield regeneration for soft reuse: A case study of Port Sunlight River Park (U.K.). *Science of the Total Environment*, 652, 810–821. https://doi.org/10.1016/j.scitotenv.2018.10.278
- Li, Y., Lin, C., Wang, Y., Gao, X., Xie, T., Hai, R., Wang, X., & Zhang, X. (2017). Multi-criteria evaluation method for site selection of industrial wastewater discharge in coastal regions. *Journal of Cleaner Production*, 161, 1143–1152. https://doi.org/10.1016/j.jclepro.2017.05.030
- Lin, B. and Wang, X. (2015). Carbon Emissions from Energy Intensive Industry in China: Evidence from the Iron and Steel Industry. *Renewable Sustainable Energy Review*, 47(9), 746–754. https://doi.org/10.1016/j.rser.2015.03.056
- Liu, L., Wang, J., Song, H., Du, J., & Yang, F. (2017). Multi-period water network management for industrial parks considering predictable variations. *Computers* and *Chemical Engineering*, 104, 172–184. https://doi.org/10.1016/j.compchemeng.2017.04.015
- Liu, X., & Ma, Y. (2021). A method to analyze the rank reversal problem in the ELECTRE II method. *Omega* (*United Kingdom*), 102. https://doi.org/10.1016/j.omega.2020.102317
- Liu, Z., Adams, M., Wen, Z., Massard, G., & Dong, H. (2017). Eco-industrial development around the globe: recent progress and continuing challenges. *Resources, Conservation and Recycling*, 127, A1–A2. https://doi.org/10.1016/j.resconrec.2017.09.018
- Loomis, J. J., & Dziedzic, M. (2018). Evaluating EIA systems' effectiveness: A state of the art. *Environmental Impact Assessment Review*, 68, 29–37. https://doi.org/10.1016/j.eiar.2017.10.005
- Lowe, E. A. (2014). *Eco-Industrial Handbook for Asian Developing Countries* (Ernest A. Lowe (2nd ed). John Wiley & Sons, Inc., New Jersey, U.S.A.
- Lowe, Ernest A. (2001). Eco-industrial Park Handbook for Asian Developing Countries. A Report to Asian Development Bank, Environment Department, Indigo Development. Oakland, CA http://www.adb.org/environment/aeo/pub/index.html
- Luthra, S., Kumar, A., Zavadskas, E. K., Mangla, S. K., & Garza-Reyes, J. A. (2020). Industry 4.0 as an enabler of sustainability diffusion in the supply chain: an analysis of influential strength of drivers in an emerging economy. *International Journal of Production Research*, 58(5), 1505–1521. https://doi.org/10.1080/00207543.2019.1660828
- MacCrimon, K. R. (1968). *Decision Marking Among Multiple–Attribute Alternatives: A Survey and Consolidated Approach, RAND Memorandum* (2nd ed). The Rand Corporation. Santa Monica, Califonia, U.S.A.
- Maiolo, M., & Pantusa, D. (2018). Infrastructure Vulnerability Index of drinking water systems to terrorist attacks. *Cogent Engineering*, 5(1).

https://doi.org/10.1080/23311916.2018.1456710

- Makaba, L. P., & Munyati, C. (2018). Strategic Environmental Assessment implementation and effectiveness bottlenecks: Lessons from Botswana. *Environmental Development*, 26, 86–99. https://doi.org/10.1016/j.envdev.2018.05.001
- Malczewski, J. (2000). On the use of weighted linear combination method in GIS: Common and best practice approaches. *Transactions in GIS*, 4(1), 5–22. https://doi.org/10.1111/1467-9671.00035
- Mardani, A., Jusoh, A., Nor, K. M. D., Khalifah, Z., Zakwan, N., & Valipour, A. (2015). Multiple criteria decision-making techniques and their applications - A review of the literature from 2000 to 2014. *Economic Research-Ekonomska Istrazivanja*, 28(1), 516–571. https://doi.org/10.1080/1331677X.2015.1075139
- Martin, M., & Harris, S. (2018). Prospecting the sustainability implications of an emerging industrial symbiosis network. *Resources, Conservation and Recycling*, 138, 246–256. https://doi.org/10.1016/j.resconrec.2018.07.026
- Massard, G., Jacquat O., Z. D. (2014). International survey on eco-innovation parks. Learning from experiences on the spatial dimension of eco-innovation. Federal Office for the Environment and the Eranet Eco-Innovera, Bern. *Environmental Studies*, 310–321.
- Massard, G., Leuenberger, H., & Dong, T. D. (2018). Standards requirements and a roadmap for developing eco-industrial parks in Vietnam. *Journal of Cleaner Production*, 188, 80–91. https://doi.org/10.1016/j.jclepro.2018.03.137
- Maurice, J. (2015). UN set to change the world with new development goals. *The Lancet*, 386(9999), 1121–1124. https://doi.org/10.1016/S0140-6736(15)00251-2
- Meng, X., Yang, F., Bao, Z., Deng, J., Serge, N. N., Zhang, Z., Chauhan, A., Saini, R. P. P., Summary, E., Fu, Z., Yang, J., Zou, T., Brown, S. V., Nderitu, D. G., Preckel, P. V., Gotham, D. J., Allen, B. W., Akella, a. K., Sharma, M. P., ... Shin, M. Y. (2014). Linking Renewable Energy to Rural Development. *Renewable and Sustainable Energy Reviews*, 39(2), 580–603. https://doi.org/10.1016/j.esd.2014.01.007
- Messaoudi, D., Settou, N., Negrou, B., Rahmouni, S., Settou, B., & Mayou, I. (2019). Site selection methodology for the wind-powered hydrogen refuelling station based on AHP-GIS in Adrar, Algeria. *Energy Procedia*, 162, 67–76. https://doi.org/10.1016/j.egypro.2019.04.008
- Michael Hayek, G. A. & J. G. (2010). No TitleAssessing London, Ontario's brownfield redevelopment effort to promote urban intensification, Local Environment. *The International Journal of Justice and Sustainability*, 15(4), 389–402. https://doi.org/10.1080/13549831003677712
- Monsef, A. H., & Smith, S. E. (2019). Integrating remote sensing, geographic information system, and analytical hierarchy process for hazardous waste landfill site selection. *Arabian Journal of Geosciences*, 12(5), 155. https://doi.org/10.1177/0734242X20932213
- Montevali, A. and Koloor, R. T. (2017). A comparison between pollutants and greenhouse gas emissions from the operation of different dryers based on energy consumption of power plants. *Journal of Cleaner Production*, 15(4), 445–461.

https://doi.org/10.1016/j.jclepro.2017.03.219

- Moreau, V., Sahakian, M., van Griethuysen, P., Vuille, F. (2017). Coming full circle: Why social and institutional dimensions matter for the circular economy. *Journal* of *Industrial Ecology*, 21(497–506). https://doi.org/10.1111/jiec.12598
- Moses, A., Iwara, E., Gbadebo, N., Olubukola, O., & Omin, I. (2018). Geographic Foci of Industries: A Suitability Analysis. Journal of Geography, Environment and Earth Science International, 14(1), 1–14. https://doi.org/10.9734/jgeesi/2018/39994
- Mousqué, F., Boix, M., Montastruc, L., Domenech, S., & Négny, S. (2020). Optimal Design of Eco-Industrial Parks with coupled energy networks addressing Complexity bottleneck through an Interdependence analysis. *Journal of Clean Energy Technologies*, 138, 106859. https://doi.org/10.1016/j.compchemeng.2020.106859
- Muhsin, N., Ahamed, T., & Noguchi, R. (2018). GIS-based multi-criteria analysis modelling used to locate suitable sites for industries in suburban areas in Bangladesh to ensure the sustainability of agricultural lands. *Asia-Pacific Journal* of Regional Science, 2(1), 35–64. https://doi.org/10.1007/s41685-017-0046-0
- Nabernegg, S.; Bednar-Friedl, B.; Wagner, F.; Schinko, T.; Cofala, J. (2017). The deployment of low carbon technologies in energy-intensive industries: A macroeconomic analysis for Europe. *Chi. In. Ener*, 10, 360–371. https://doi.org/10.3390/en10030360
- Naghdi, F., Monavvari, S. M., Hosseini, S. M., Gharagozlu, A. (2017). Industrial Zoning of East Azerbaijan Province of Iran Using Multi-criteria Evaluation Modelling. *Journal of Applied Ecology and Environmental Research*, 15(3), 1565–1576. http://dx.doi.org/10.15666/aeer/1503_15651576
- Neves, A., Godina, R., Azevedo, S. G., & Matias, J. C. O. (2020). A comprehensive review of industrial symbiosis. *Journal of Cleaner Production*, 247. https://doi.org/10.1016/j.jclepro.2019.119113
- Nguyen, H. T., Md Dawal, S. Z., Nukman, Y., Rifai, A. P., & Aoyama, H. (2016). An integrated MCDM model for conveyor equipment evaluation and selection in an FMC based on a Fuzzy AHP and Fuzzy ARAS in the presence of vagueness. *PLoS ONE*, 11(4), 1–26. https://doi.org/10.1371/journal.pone.0153222
- Noorollahi, Y., Yousefi, H., & Mohammadi, M. (2016). Multi-criteria decision support system for wind farm site selection using GIS. Sustainable Energy Technologies and Assessments. 23. 38-50. https://doi.org/10.1016/j.seta.2015.11.007
- Norgate, T., Johanshahi, S., Rankin, W. (2007). Assessing the Environmental Impact of Metal Production Process. *Journal of Cleaner Production*, *11*(3), 838–848.
- Nuhu, S. K., Manan, Z. A., Wan Alwi, S. R., & Md Reba, M. N. (2021). Roles of geospatial technology in eco-industrial park site selection: State-of-the-art review. *Journal of Cleaner Production*, 309, 127361. https://doi.org/10.1016/j.jclepro.2021.127361
- Ohnishi, S., Dong, H., Geng, Y., Fujii, M., & Fujita, T. (2017). A comprehensive evaluation on industrial & urban symbiosis by combining MFA, carbon footprint and emergy methods—Case of Kawasaki, Japan. In *Ecological Indicators*, 73, 315–324. https://doi.org/10.1016/j.ecolind.2016.10.016

- Okada, A., & Siddharthan, N. (2007). Industrial clusters in India: Evidence from automobile clusters in Chennai and the national capital region. *Institute of Developing Economies*, 1–107.
- Osra, F. A., & Kajjumba, G. W. (2019). Landfill site selection in Makkah using geographic information system and analytical hierarchy process. *Waste Management and Research*. 38(3), 245-253. https://doi.org/10.1177/0734242X19833153
- Pacheco, A. G. C., & Krohling, R. A. (2018). Ranking of Classification Algorithms in Terms of Mean–Standard Deviation Using A-TOPSIS. *Annals of Data Science*, 5(1), 93–110. https://doi.org/10.1007/s40745-018-0136-5
- Park, J. M., Park, J. Y., & Park, H. S. (2016). A review of the National Eco-Industrial Park Development Program in Korea: Progress and achievements in the first phase, 2005-2010. *Journal of Cleaner Production*, 114, 33–44. https://doi.org/10.1016/j.jclepro.2015.08.115
- Paul, S. K. (2015). Supplier selection for managing supply risks in the supply chain: a fuzzy approach. *International Journal of Advanced Manufacturing Technology*, 79(1–4), 657–664. https://doi.org/10.1007/s00170-015-6867-y
- Penadés-Plà, V., García-Segura, T., Martí, J. V., & Yepes, V. (2016). A review of multi-criteria decision-making methods applied to the sustainable bridge design. *Sustainability (Switzerland)*, 8(12). https://doi.org/10.3390/su8121295
- Piengang, F. C. N., Beauregard, Y., & Kenné, J. P. (2019). An APS software selection methodology integrating experts and decision-makers opinions on selection criteria: A case study. *Cogent Engineering*, 6(1). https://doi.org/10.1080/23311916.2019.1594509
- Pramanik, M. K. (2016). Site suitability analysis for agricultural land use of Darjeeling district using AHP and GIS techniques. *Modeling Earth Systems and Environment*, 2(2), 1–22. https://doi.org/10.1007/s40808-016-0116-8
- Qin, R., Cheng, C., Thompson, R. G., Peck, M., Chipman, R., Parks, F. O. R. E., Zhu, Q., Geng, Y., Sarkis, J., Lai, K., Rock, B., Guiqin, W., Li, Q., Guoxue, L., Van Berkel, R. (2020). Criteria of Eco-Industrial Park Location and their Prioritization with Using Fuzzy AHP and Triangular Fuzzy Number. *Journal of Cleaner Production*, 12(1), 1–12. https://doi.org/https://doi.org/10.31235/osf.io/jt6r7
- Qin, R., Cheng, C., Thompson, R. G., Nations, U., Peck, M., Chipman, R., Parks, F. O. R. E., Sumathi, V. R., Service, G., Portal, P. R., Commission, C. A., Zhu, Q., Geng, Y., Sarkis, J., Lai, K., Rock, B., Guiqin, W., Li, Q., Guoxue, L., ... Van Berkel, R. (2020). Criteria of Eco-Industrial Park Location and their Prioritization with Using Fuzzy AHP and Triangular Fuzzy Number. *Journal of Cleaner Production*, 12(1), 1–12. https://doi.org/10.31235/osf.io/jt6r7
- Rahim, R. (2017). Study Approach Technique for Order of Preference by Similarity to Ideal Solution (TOPSIS). *International Journal of Recent Trends in Engineering and Research*, 3(3), 268–285. https://doi.org/10.23883/ijrter.2017.3077.gzxdl
- Rahimi, S., Ashournejad, Q., Moore, A. B., & Ghorbani, H. (2020). Optimal siting of edutainment energy parks through the modelling of weighted spatial decision criteria. *Journal of Cleaner Production*, 255, 120279.

https://doi.org/10.1016/j.jclepro.2020.120279

- Rahmat, Z. G., Niri, M. V., Alavi, N., Goudarzi, G., Babaei, A. A., Baboli, Z., & Hosseinzadeh, M. (2017). Landfill site selection using GIS and AHP: a case study: Behbahan, Iran. *KSCE Journal of Civil Engineering*, 21(1), 111–118. https://doi.org/10.1007/s12205-016-0296-9
- Ramos, A. R., Ferreira, J. C. E., Kumar, V., Garza-Reyes, J. A., & Cherrafi, A. (2018). A lean and cleaner production benchmarking method for sustainability assessment: A study of manufacturing companies in Brazil. *Journal of Cleaner Production*, 177, 218–231. https://doi.org/10.1016/j.jclepro.2017.12.145
- Rasim, E. F. Rahman, N. F. Dewi, and Riza, L. S. (2017). Decision Support Systems for Performance and Evaluation of Teachers in General-English Course by Using the SMARTER and TOPSIS Methods. *IOP Conference of Service, Matter, Science and Engineering*, 180(1), 012283.
- Register, R. (2002). *Eco-cities: building cities in balance with nature*. Berkeley Hills Books, Berkeley. Berkeley, San Francisco, USA and Social Science Literature Press, Beijing
- Reisi, M., Afzali, A., & Aye, L. (2018a). Applications of analytical hierarchy process (AHP) and analytical network process (ANP) for industrial site selections in Isfahan, Iran. *Environmental Earth Sciences*, 77(14), 0. https://doi.org/10.1007/s12665-018-7702-1
- Reisi, M., Afzali, A., & Aye, L. (2018b). Applications of analytical hierarchy process (AHP) and analytical network process (ANP) for industrial site selections in Isfahan, Iran. *Environmental Earth Sciences*, 77(14), 1–13. https://doi.org/10.1007/s12665-018-7702-1
- Rikalovic, A., Cosic, I., Labati, R. D., & Piuri, V. (2018). Intelligent decision support system for industrial site classification: A GIS-based hierarchical neuro-fuzzy approach. *IEEE Systems Journal*, 12(3), 2970–2981. https://doi.org/10.1109/JSYST.2017.2697043
- Rikalovic, A., Cosic, I., & Lazarevic, D. (2014). GIS-Based Multi-Criteria Analysis for Industrial Site Selection GIS-Based Multi-Criteria Analysis for Industrial Site Selection. *Procedia Engineering*, 69, 1054–1063. https://doi.org/10.1016/j.proeng.2014.03.090
- Rizzo, E., Pesce, M., Pizzol, L., Alexandrescu., F. M., Giubilato, E., Critto, A., Marcomini, A., & Bartke, S. (2015). Brownfield regeneration in Europe: Identifying stakeholder perceptions, concerns, attitudes and information needs. *Land Use Policy*, 48, 437–453. https://doi.org/10.1016/j.landusepol.2015.06.012
- Robinson, J., and Amirtharaj, H. (2014). MADM Problems with Correlation Coefficient of Trapezoidal Fuzzy Intuitionistic Fuzzy Sets. Advancement In Decision Science, 3(2), 1. https://doi.org/10.1155/2014/159126
- Saaty, L.T. and Peniwati, K. (2008). *Group decision making drowning out and reconciling differences*. RWS.
- Saaty, T. L., Vargas, L. G. (2013). *Decision Making with the Analytic Network Process* (Hillier, C. C., Frederick, S., Price, C. (2nd ed). Springer, New York. https://doi.org/10.1007/978-1-4614-7279-7
- Saaty, T. L. (1977). A scaling method for priorities in hierarchical structures. Journal

of Mathematical Psychology, 15(3), 234–281. https://doi.org/10.1016/0022-2496(77)90033-5

- Saaty, T. L. (2004). Decision making the Analytic Hierarchy and Network Processes (AHP/ANP). *Journal of Systems Science and Systems Engineering*, 13(1), 1–35. https://doi.org/10.1007/s11518-006-0151-5
- Saaty, T. L., & Vargas, L. G. (2013). Sensitivity analysis in the analytic hierarchy process. In *International Series in Operations Research and Management Science*, 195, 345-360. https://doi.org/10.1007/978-1-4614-7279-7_15
- Sakr, D., Baas, L., El-Haggar, S., & Huisingh, D. (2011). Critical success and limiting factors for eco-industrial parks: Global trends and Egyptian context. *Journal of Cleaner Production*, 19(11), 1158–1169. https://doi.org/10.1016/j.jclepro.2011.01.001
- Sánchez-Lozano, J. M., Henggeler Antunes, C., García-Cascales, M. S., & Dias, L. C. (2014). GIS-based photovoltaic solar farms site selection using ELECTRE-TRI: Evaluating the case for Torre Pacheco, Murcia, Southeast of Spain. *Renewable Energy*. 66, 478-494. https://doi.org/10.1016/j.renene.2013.12.038
- Sarmiento, R., & Vargas-Berrones, K. X. (2018). Modelling the implementation of green initiatives: An AHP-BOCR approach. *Cogent Engineering*, 5(1), 4–17. https://doi.org/10.1080/23311916.2018.1432120
- Schlarb, M. (2001). Eco-Industrial Deelopment: A Strategy for Building Sustainable Communities. *Review of Economic Development Interaction and Practice*, 8.
- Sedrati, M., Maanan, M., & Rhinane, H. (2019). PV power plants sites selection using GIS-FAHP based approach in north-western Morocco. *International Archives of the Photogrammetry, Remote Sensing and Spatial Information Sciences - ISPRS Archives*, 42(4/W19), 385–392. https://doi.org/10.5194/isprs-archives-XLII-4-W19-385-2019
- Sellitto, M. A. & Murakami, K. F. (2018). Industrial symbiosis: a case study involving steelmaking, cement manufacturing, and a zinc smelting plant. *Chemical Engineering Transactions*, 70, 211–216. https://doi.org/10.3303/CET1870036
- Sertyesilisik, B., & Sertyesilisik, E. (2016). Eco-Industrial Development : As a Way of Enhancing Sustainable Development. *Journal of Economic Development, Environment and People*. 5(1), 6–27. http://dx.doi.org/10.26458/jedep.v5i1.133
- Seyedmohammadi, J., Sarmadian, F., Jafarzadeh, A. A., Ghorbani, M. A., & Shahbazi, F. (2018). Application of SAW, TOPSIS and fuzzy TOPSIS models in cultivation priority planning for maize, rapeseed and soybean crops. *Geoderma*, 310, 178– 190. https://doi.org/10.1016/j.geoderma.2017.09.012
- Shine, T. A. L., James, J., Dolly, D. E., Mathew, M. (2020). Design Concepts of Green and Sustainable Industrial Park. *International Journal of Scientific Research & Engineering Trends*, 6(2), 734–738.
- Singh, L. K., Jha, M. K., & Chowdary, V. M. (2018). Assessing the accuracy of GISbased Multi-Criteria Decision Analysis approaches for mapping groundwater potential. *Ecological Indicators*, 91, 24–37. https://doi.org/10.1016/j.ecolind.2018.03.070
- Stojcic, M., Zavadskas, E. K., Pamucar, D., Stevic, Z. and Mardani, A. (2019). Application of MCDM Methods in Sustainability Engineering: A Literature

Review 2008 – 2018. *Symmetry*, 11(350), 2–24. https://doi.org/10.3390/sym11030350

- Stucki, J., Flammini, A., van Beers, D., Phuong, T. T., Anh, N. T., Dong, T. D., Huy, V. Q., & Hieu, V. T. M. (2019). Eco-industrial park (EIP) development in Viet Nam: Results and key insights from UNIDO's EIP project (2014-2019). *Sustainability (Switzerland)*, 11(17), 1–23. https://doi.org/10.3390/su11174667
- Sun, C., Sun, C., Yang, Z., Zhang, J., & Deng, Y. (2016). Urban land development for industrial and commercial use: A case study of Beijing. *Sustainability* (*Switzerland*), 8(12). https://doi.org/10.3390/su8121323
- Susur, E., Hidalgo, A., & Chiaroni, D. (2019). A strategic niche management perspective on transitions to eco-industrial park development: A systematic review of case studies. *Resources, Conservation and Recycling*, 140, 338–359. https://doi.org/10.1016/j.resconrec.2018.06.002
- Susur, E., Martin-Carrillo, D., Chiaroni, D., & Hidalgo, A. (2019). Unfolding ecoindustrial parks through niche experimentation: Insights from three Italian cases. *Journal of Cleaner Production*, 239, 118069. https://doi.org/10.1016/j.jclepro.2019.118069
- Tadić, S., Zečević, S. and Krstić, M. (2014). A novel hybrid MCDM model based on fuzzy DEMATEL, fuzzy ANP and fuzzy VIKOR for city logistics concept selection. Expert Systems with Applications. *Journal of Applied Ecology and Environmental Research*, 41(10), 8112–8128. https://doi.org/10.1016/j.eswa.2014.07.021
- Taibi, A., Atmani, B., Britain, G., Avenue, E., Saaty, T. L. T. L., Vargas, L. G., Reisi, M., Afzali, A., Aye, L., Saaty, T. L. T. L., Ohri, A., & Singh, P. K. (2013). Sensitivity analysis in the analytic hierarchy process. *International Series in Operations Research and Management Science*, 195(3), 345–360. https://doi.org/10.1007/978-1-4614-7279-7_15
- Tavana, M., Yazdani, M., & Di Caprio, D. (2017). An application of an integrated ANP–QFD framework for sustainable supplier selection. *International Journal* of Logistics Research and Applications, 20(3), 254–275. https://doi.org/10.1080/13675567.2016.1219702
- Taye, B., Gebre, S. L., Gemeda, D. O., & Getahun, K. (2019). Using Geospatial Techniques in the Selection of Potential Ecotourism Sites in Menz-Geramidir District, Ethiopia. *Ghana Journal of Geography*, 11(1), 201–227.
- Techno, D. E. (2016). EIA for Production of Extender Oils at PLO 13, Jalan Tengar, Tanjung Langsat Industrial Area, Malaysia. www.detec.my
- Tessitore, S., Daddi, T., & Iraldo, F. (2015). Eco-industrial parks development and integrated management challenges: Findings from Italy. *Sustainability* (*Switzerland*), 7(8), 10036–10051. https://doi.org/10.3390/su70810036
- Thieriot, H., & Sawyer, D. (2015). Development of Eco-Efficient Industrial Parks in China: A review. *IISD Report*, 1–24. https://www.iisd.org/sites/default/files/publications/development-eco-efficientindustrial-parks-china-review-en.pdf
- Tolstykh, T., Shmeleva, N., & Gamidullaeva, L. (2020). Evaluation of circular and integration potentials of innovation ecosystems for industrial sustainability.

Sustainability (Switzerland), 12(11), 1–17. https://doi.org/10.3390/su12114574

- Torabi-Kaveh, M., Babazadeh, R., Mohammadi, S. D., & Zaresefat, M. (2016). Landfill site selection using a combination of GIS and fuzzy AHP, a case study: Iranshahr, Iran. Waste Management and Research, 34(5), 438–448. https://doi.org/10.1177/0734242X16633777
- Ubando, A. T., Felix, C. B., Gue, I. H. V., Promentilla, M. A. B., & Culaba, A. B. (2020). A fuzzy analytic hierarchy process for the site selection of the Philippine algal industry. *Clean Technologies and Environmental Policy*, 22(1), 171–185. https://doi.org/10.1007/s10098-019-01775-0
- UNIDO. (2016). Global assessment of eco-industrial parks in developing and emerging countries: Achievements, good practices and lessons learned from thirty-three industrial parks in twelve selected emerging and developing countries. *World Evaluation of Eco-Industrial Parks Development*, 351–385.
- UNIDO. (2021). An International Framework for Eco-Industrial Parks, Version 2.0. World Bank. https://openknowledge.worldbank.org/handle/10986/35110 License: CC BY-NC 3.0 IGO
- Valenzuela-Venegas, G., Vera-Hofmann, G., & Díaz-Alvarado, F. A. (2020). Design of sustainable and resilient eco-industrial parks: Planning the flows integration network through multi-objective optimization. *Journal of Cleaner Production*, 243, 118610. https://doi.org/10.1016/j.jclepro.2019.118610
- Valverde, J. P. B., Blank, C., Roidt, M., Schneider, L., & Stefan, C. (2016). Application of a GIS multi-criteria decision analysis for the identification of intrinsic suitable sites in Costa Rica for the application of Managed Aquifer Recharge (MAR) through spreading methods. *Water (Switzerland)*, 8(9). https://doi.org/10.3390/w8090391
- Van Laarhoven, P. J. M., & Pedrycz, W. (1983). A fuzzy extension of Saaty. *Fuzzy* Sets and Systems, 11(1-3), 229–241. https://doi.org/10.1016/S0165-0114(83)80082-7
- Walls, J. L., & Paquin, R. L. (2015). Organizational Perspectives of Industrial Symbiosis: A Review and Synthesis. *Organization and Environment*, 28(1), 32– 53. https://doi.org/10.1177/1086026615575333
- Wang, S., Li, G., Fang, C., Li, H. Qiang, Wang, L. mao, Shen, L., Chen, F. nan, Liu, Z., Guan, D., Wei, W., Davis, S. J., Ciais, P., Bai, J., Peng, S., Zhang, Q., Hubacek, K., Marland, G., Andres, R. J., Crawford-Brown, D., Cui, Z. (2018). Combining Fuzzy AHP with GIS and Decision Rules for Industrial Site Selection. *Chemical Engineering Transactions*, 4(6), 60. https://doi.org/10.9781/ijimai.2017.06.001
- Wind, Y., & Saaty, T. L. (1980). Marketing Applications of the Analytic Hierarchy Process. In *Management Science*, 26(7), 641–658). https://doi.org/10.1287/mnsc.26.7.641
- Yang, C. L., Chuang, S. P., Huang, R. H., & Tai, C. C. (2008). Location selection based on AHP/ANP approach. 2008 IEEE International Conference on Industrial Engineering and Engineering Management, IEEM 2008, 1148–1153. https://doi.org/10.1109/IEEM.2008.4738050
- Yap, J., Ho, C., & Ting, C. (2019). A systematic review of the applications of multi-

criteria decision-making methods in site selection problems, J., Ho, C., & Ting, C. *Built Environment Project and Asset Management*, 9(4), 548–563. https://doi.org/10.1108/BEPAM-05-2018-0078

- Yatim, P., Ngan, S. L., & Lam, H. L. (2017). Financing green growth in Malaysia: Enabling conditions and challenges. *Chemical Engineering Transactions*, 61(27567), 1579–1584. https://doi.org/10.3303/CET1761261
- Yousefi, E.; Esmail, Salehi; Seyed, Hamid, Zahiri; Ahmadreza, yavar. (2016). Green Space Suitability Analysis Using Evolutionary Algorithm and Weighted Linear Combination (WLC) Method. *Space Ontology International Journal*, 5(4), 51– 60.. https://doi.org/20.1001.1.23456450.2016.5.4.5.5
- Yu, C., Dijkema, G. P. J., De Jong, M., & Shi, H. (2015). From an eco-industrial park towards an eco-city: A case study in Suzhou, China. *Journal of Cleaner Production*, 102, 264–274. https://doi.org/10.1016/j.jclepro.2015.04.021
- Yuen, K. K. F. (2012). Membership maximization prioritization methods for fuzzy analytic hierarchy process. *Fuzzy Optimization and Decision Making*, 11(2), 113– 133. https://doi.org/10.1007/s10700-012-9119-8
- Zadeh, L. A. (1975). The concept of a linguistic variable and its application to approximate reasoning-I. *Information Sciences*, 8(3), 199–249. https://doi.org/10.1016/0020-0255(75)90036-5
- Zailan, R. (2020). Towards Eco-Industrial Park in Malaysia: Promising Opportunities, Challenges and Regulator Roles. *International Journal of Psychosocial Rehabilitation*, 24(1), 562–570. https://doi.org/10.37200/ijpr/v24i1/pr200162
- Zamani, R., and Yousefi, P. (2013). Optimal decision-making approach for selecting effort estimation model. *International Journal of Machine Learning and Computing*, 3(1), 1–4.
- Zarin, R., Azmat, M., Naqvi, S. R., Saddique, Q., & Ullah, S. (2021a). Landfill site selection by integrating fuzzy logic, AHP, and WLC method based on multicriteria decision analysis. *Environmental Science and Pollution Research*, 28(16), 19726–19741. https://doi.org/10.1007/s11356-020-11975-7
- Zarin, R., Azmat, M., Naqvi, S. R., Saddique, Q., & Ullah, S. (2021b). Landfill site selection by integrating fuzzy logic, AHP, and WLC method based on multicriteria decision analysis. *Environmental Science and Pollution Research*, 28, 9726–9741. https://doi.org/10.1007/s11356-020-11975-7
- Zeleny, M. (2012). Multiple criteria decision making Kyoto 1975 (Topcu, Z. M., Ozaydin, O., Kabak, O. (1st ed). Springer Science & Business Media, Cham, Sweitzerland. eBook ISBN 978-3-030-52406-7
- Zhang, L., Geng, Y., Dong, H., Zhong, Y., Fujita, T., Xue, B., & Park, H. S. (2016). Emergy-based assessment on the brownfield redevelopment of one old industrial area: A case of Tiexi in China. *Journal of Cleaner Production*, 114, 150–159. https://doi.org/10.1016/j.jclepro.2015.05.065
- Zimmermann, H. J. (2000). An application-oriented view of modelling uncertainty. *European Journal of Operational Research*, 122(2), 190–198. https://doi.org/doi:10.1016/S0377-2217(99)00228-3

LIST OF PUBLICATIONS

Journal with Impact Factor

- Nuhu, S. K., Manan, Z. A., Wan Alwi, S. R., & Md Reba, M. N. (2021). Roles of geospatial technology in eco-industrial park site selection: State-of-the-art review. *Journal of Cleaner Production*, 309, 127361. https://doi.org/10.1016/j.jclepro.2021.127361 (Q1, IF: 9.297).
- Nuhu, S. K., Manan, Z. A., Wan Alwi, S., Reba, N. M. (2021). A New Hybrid Modelling Approach for an Eco-Industrial Park Site Selection. *Chemical Engineering Transactions*. 89, 343-348 DOI:10.3303/CET2189058 (Q3, IF: 0.683).
- 3) **Nuhu, S. K.**, Manan, Z. A., Wan Alwi, S. R., & Md Reba, M. N. (2021). Integration of the Modelling Tools Used for Site Selection of an Eco-Industrial Park. *Journal of Cleaner Production* (**IF 9.297: Q1**), (Under review).

Non-Indexed Conference Proceedings

- Nuhu, S. K., Manan, Z. A., Wan Alwi, S. R., & Md Reba, M. N. (2020). Improvement of the Site Selection Guide for Brownfield Eco-Industrial Park: The Fuzzy Analytic Hierarchy Process. 3rd Scientia Academia International Conference (SAICon-2020), Scientia Academia Malaysia, 26 – 27 December 2020.
- 2) Nuhu, S. K., Reba, N. M., Manan, Z. A., Wan Alwi, S. (2021). Assessing the Criteria of Eco-Industrial Park Site Selection for the SDG Initiatives. *Regional Conference in Civil Engineering and Sustainable development Goals in Higher Education Institutions 2020*, Universiti Teknologi Malaysia, 23 – 24 January 2021.
- 3) Nuhu, S. K., Manan, Z. A., Wan Alwi, S., Reba, N. M. (2021). A New Hybrid Modelling Approach for an Eco-Industrial Park Site Selection. 7th International Conference on Low Carbon on Asia & Beyond. "Circular Economy Towards Sustainable Development" 18th & 19th October 2021, Johor Bahru, Malaysia.

Book Chapter (Indexed)

- Nuhu, S. K., Manan, Z. A., Wan Alwi, S., Reba, N. M. (2021). Development and Design of Eco-Industrial Park Toward Circular Economy; (Chapter 11) in the book Process Design and Optimisation for Circular Economy. First Edition 2021, Edited by Azizul Azri Mustaffa & Peng Yen Liew. Published by Penerbit UTM, Malaysia. *ISBN 978-983-52-1790-6*.
- 2) **Nuhu, S. K.**, Reba, N. M., Manan, Z. A., Wan Alwi, S. (2021). Assessing the Criteria of Eco-Industrial Park Site Selection for the SDG Initiatives; (Book

Chapter) in a Book "Sustainability Management Strategies and Impact in Developing Countries. By *Community, Environment and Disaster Risk Management* (CEDRM) Book Series Volume. Emerald Publisher (indexed by MDPI) (accepted, in press)