



UNIVERSITI PUTRA MALAYSIA

USING GENETIC ALGORITHMS TO OPTIMISE LAND USE SUITABILITY

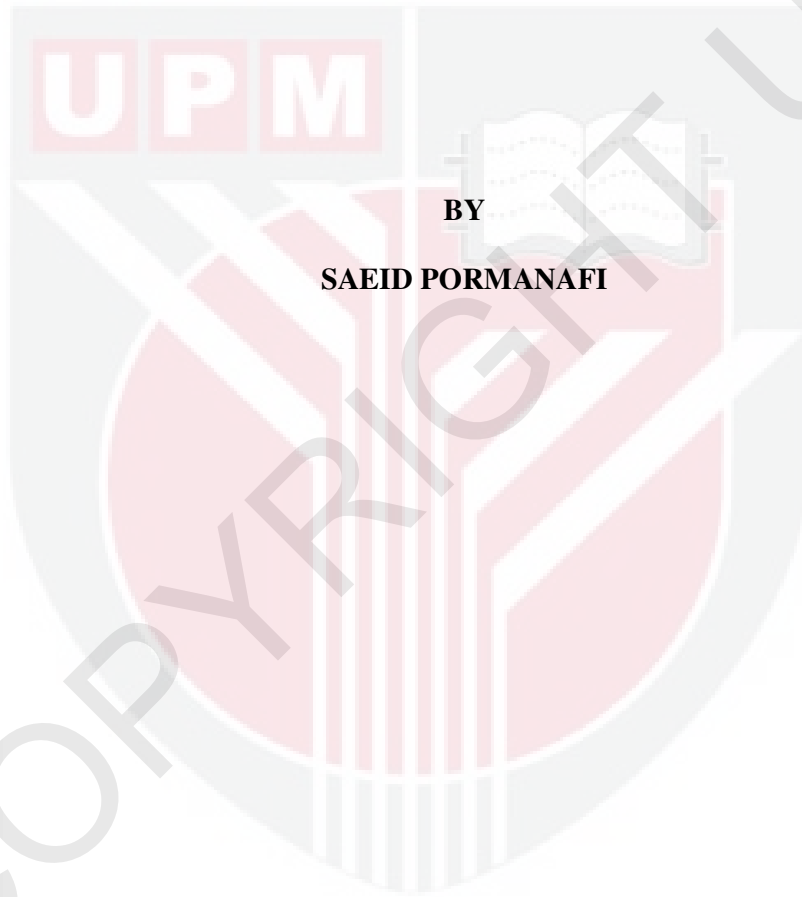
SAEID PORMANAFI

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USING GENETIC ALGORITHMS TO OPTIMISE LAND USE SUITABILITY

BY

SAEID PORMANAFI



**Thesis submitted to the School of Graduate Studies, Universiti Putra Malaysia
in fulfilment of the Requirement for the Degree of Doctor of Philosophy**

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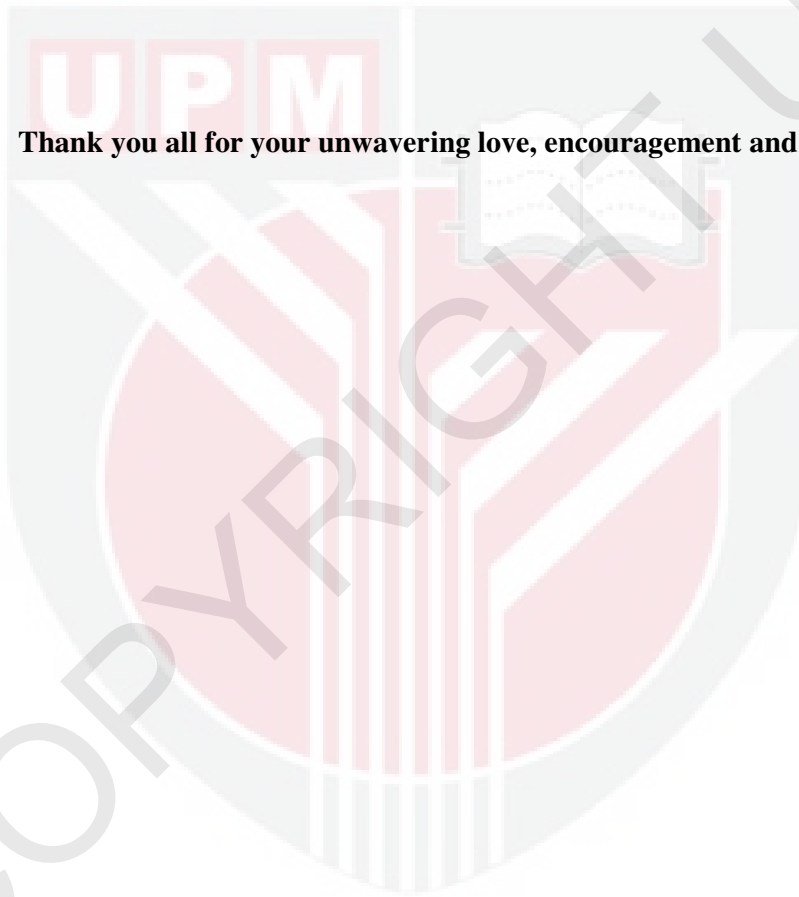
DEDICATION

Dedicated:

To my beloved wife

To my dearest parent

Thank you all for your unwavering love, encouragement and support



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment
of the requirements for the degree of Doctor of Philosophy

**USING GENETIC ALGORITHMS TO OPTIMISE LAND USE
SUITABILITY**

By

SAEID PORMANAFI

February 2012

Chairman: Professor Shattri B. Mansor, PhD

Faculty: Institute of advanced Technology

Land-use planning is defined as the most appropriate utilization that would achieve the paramount benefit of protecting the resources. In this study, under environment-friendliness objective, based on multi-agent genetic algorithms, was developed a geospatial model for the land use allocation. The model applied to solve the practical multi-objective spatial optimization allocation problems of land use in the core region of Menderjan Basin in Iran. The first task was studying the dominant of crops and economic suitability evaluation of land with the land evaluation framework developed by FAO, (1976-2007) using GIS. Second task is to determine the fitness function for the genetic algorithms. The third objective is to optimize the land use map using economic benefits. In the socioeconomic assessment of the Menderjan watershed; consultation with experts and the interview with local residents implemented. Different scenarios then arranged according to the land suitability classes. The Erosion Potential Method (EPM) used in erosion estimation and sediment yield of the study. The highest annual erosion rate belongs to the potato agricultural land use. Third scenario suggested in comparison to the economic views.

In this research, based on both irrigation managements of the crops and water demands' model of crops would be developed and calculated which they integrated in RS and GIS environment. In the GAs Model, parent selected among the initial population. In fact, the initial population includes the land suitability analysis, land use/ land cover, which is extracted from RS and scenarios of land evaluation and crop suitability. To sum up, coding is remarkably based on objective function, which it has been great in cost/ benefit from all cultivating activities and obtained costs of land erosion. After calculating the fitness function, which it includes, cost and benefit matrix, cost of changing land uses together, offspring (the next generation) which are importantly generated. Selecting the offspring during the research has been based on their capability of elitism. This selection implemented according to the percentage of progressing, comparison and replacing in GAs programming. Finally, the land use and defined scenarios obtained as optimized output, which is a dynamic model in this study. The results shows; the major limitations regarding to wheat in this region is related to the topography. 28.6% of the land has severe topographic limitations. The most suitable class is S₂ for Potato. The limitation of this suitability class majorly is soil properties. Results of Almond land suitability analysis shows, the most extensive land is in the moderate limitation class. The main limitation is properties of the soil and climate. After doing the related analyses, it has been achieved that the water consumption (water demand) for wheat in May had the most consumption of water and April and June comes afterward. Potato in July has more water consumption and after that August, September, June and May. The erosion potential categories determined that heavy and severe class covered 35% of the area. Land use/ Land Cover is obtained by satellite image processing that the overall kappa of the classification is 87.4% and the overall accuracy is 89.6%. As it has mapped, the

Irrigated area is 4689 ha. According to the results of the GAs Programme and the produced graphs in evaluating the best solutions, it has been recognized that after 25 frequencies there is not any intensify change, which it happened in the optimized beneficial value, so, extra reiteration has not influence in the possible better answer. The final optimized benefit is $12 \cdot 10^{11}$.



Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan Ijazah Doktor Falsafah

**GENETIK UNTUK MENGOPTIMUMKAN KESESUAIAN PENGGUNAAN
TANAH**

Oleh

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Perancangan penggunaan tanah didefinisikan sebagai pendayagunaan yang paling wajar yang dapat mencapai faedah yang paling berkesan bagi melindungi sumber. Dalam kajian ini, disebabkan oleh objektif mesra alam persekitaran, berdasarkan algoritma genetic multiagen, model geospasial bagi alokasi penggunaan tanah telah dihasilkan. Model ini diaplikasikan bagi menyelesaikan masalah pengalokasian optima spatial multiobjektif yang praktikal terhadap penggunaan tanah di daerah utama di Lembangan Menderjan di Iran. Tugas pertama ialah mengkaji tanaman yang dominan dan penilaian kesesuaian ekonomi tanah yang dimajukan oleh FAO, (1976-2007). Tugas kedua adalah untuk menentukan fungsi fitness bagi algoritma genetik. Objektif ketiga adalah untuk mengoptimumkan peta penggunaan tanah menggunakan faedah ekonomi. Dalam penilaian sosioekonomi legeh Menderjan; konsultasi dengan pakar dan temubual dengan penduduk. Senario yang berbeza telah diatur berdasarkan kelas kesesuaian tanah. Kaedah Potensi Hakisan (EPM) digunakan dalam menganggarkan hakisan dan hasil sedimen dalam kajian ini. Dari sudut ekonomi, kos penggunaan tanah yang efektif ialah. Senario ketiga dicadangkan

untuk membandingkannya dengan sudut pandangan ekonomi. Dalam penyelidikan ini, berdasarkan kedua-dua pengurusan pengairan tanaman dan model permintaan terhadap air bagi tanaman pada musim pertumbuhan, sepatutnya dihasilkan dan dikira yang diintegrasikan dalam persekitaran RS dan GIS. Dalam model GAs, Sebenarnya, populasi awal termasuk penilaian kesesuaian tanah, penggunaan tanah/ permukaan tanah, yang diekstrak daripada RS dan senario penilaian tanah dan kesesuaian tanaman. Ringkasnya, koding berdasarkan fungsi yang objektif, yang di dalam kos/manfaat daripada semua aktiviti penanaman dan kos hakisan tanah.. Selaepas mengambil kira fungsi fitness yang termasuk kos dan matrik faedah, kos penukaran penggunaan tanah bersama, keturunan (generasi seterusnya) yang penting dijana. Pemilihan keturunan ketika penyelidikan berdasarkan kapabiliti elitisme mereka. Pemilihan ini dilaksanakan berdasarkan peratusan dari segi kemajuan, perbandingan dan penggantian dalam program GA. Akhirnya, penggunaan tanah dan senario yang telah dikena diperoleh sebagai output yang merupakan model yang dinamik dalam kajian ini. Keputusan menunjukkan; limitasi utamaberkaitan dengan gandum di daerah ini ialah yang berkaitan dengan topografi. 28.6% tanah mempunyai limitasi topografi yang teruk. Kelas yang paling sesuai ialah S₂ untuk kentang. Limitasi kesesuaian kelas yang paling utama ialah sifat tanah. Keputusan analisis kesesuaian tanah untuk badam menunjukkan tanah yang paling ekstensif ialah kelas limitasi yang sederhana. Limitasi utama ialah sifat fizikal tanah dan iklim. Selepas menjalankan analisis keputusan, didapati bahawa konsumsi air (permintaan untuk air) bagi gandum dalam bulan Mei menunjukkan konsumsi air yang paling ketara dan April dan seterusnya bulan Jun. Kentang pula menunjukkan konsumsi air yang banyak pada bulan Julai dan diikuti bulan Ogos, September, Jun dan Mei. Kategori potensi hakisan menentukan kelas berat dan teruk meliputi 35% kawasan.

Penggunaan tanah/ permukaan tanah diperoleh melalui pemrosesan imej satelit yang menunjukkan bahawa keseluruhan klasifikasi kappa ialah 87.4% dan keseluruhan ketepatan ialah 89.6%. Disebabkan pemetaan, kawasan berpengairan ialah 4689 hektar. Berdasarkan keputusan yang diperoleh daripada Program GAs dan graf yang dihasilkan dalam menilai penyelesaian yang terbaik, didapati bahawa selepas 25 frekuensi tidak terdapat perubahan ketara yang berlaku pada nilai bermanfaat yang optimum, oleh itu, tambahan reiterasi tidak mempengaruhi jawapan yang lebih baik setakat ini. Faedah optimum terakhir ialah 12×10^{11} .

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Saeid Pormanafi

APPROVAL

I certify that a Thesis Examination Committee has met on (24 February 2012) to conduct the final examination of (SAEID PORMANAFI) on his (or her) thesis entitled "**Using Genetic Algorithms to Optimise Landuse Suitability**" in accordance with the Universities and University Colleges Act 1971 and the Constitution of the Universiti Putra Malaysia [P.U.(A) 106] 15 March 1998. The Committee recommends that the student be awarded the (Doctor of Philosophy).

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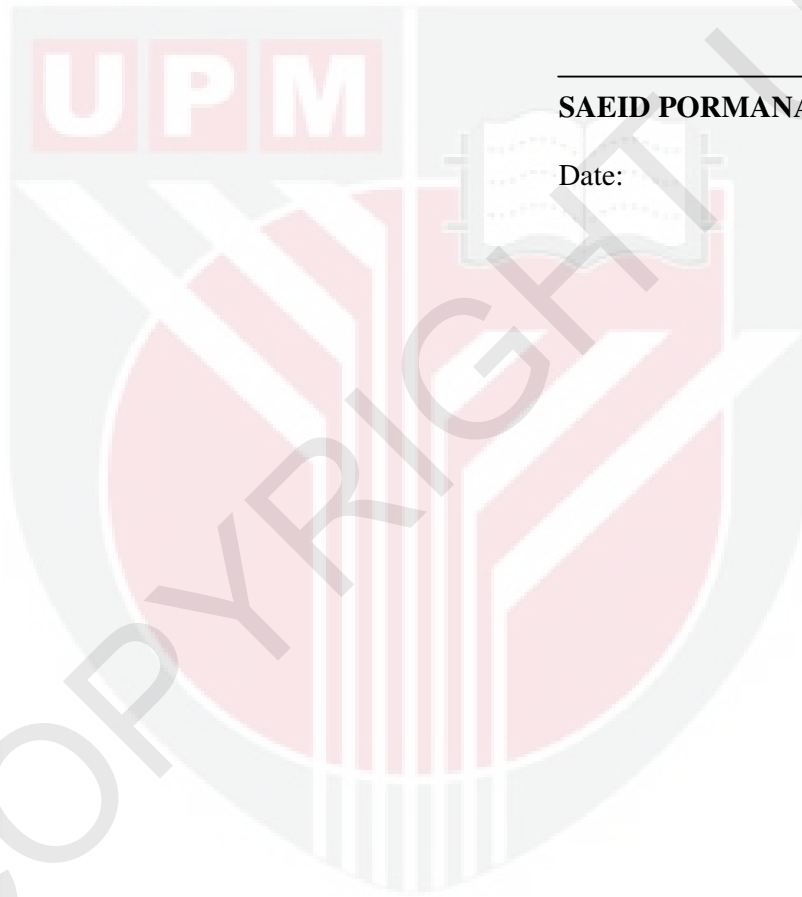
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DECLARATION

I declare that this thesis is my original work except for quotations and citations, which have been duly acknowledged. I also declare that it has not been previously and is not concurrently, submitted for any other degree at Universiti Putra Malaysia or other institutions.



SAEID PORMANAFI

Date:

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LIST OF ABBREVIATIONS

GIS	Geographical Information System
MCDM	Multi Criteria Decision Making
MCA	Multi Criteria Analysis
GA	Genetic Algorithm
GPS	Global Positioning System
LADSS	Land Decision Support System
DSS	Decision Support System
FAO	Food and Agriculture Organization
LUT	Land Utilization Types
LESA	Land-Evaluation and Site Assessment
AEZ	Agro-Ecological Zoning
ALES	Automated Land Evaluation System
AI	Artificial Intelligence
RS	Remote Sensing
ET	Actual Evapotranspiration
ET _o	Reference Evapotranspiration
ET _c	Crop Evapotranspiration
WBM	Water Balance Model
USLE	Universal Soil Loss Equation
MUSLE	Modified Universal Soil Loss Equation
WEPP	Water Erosion Prediction Project
RUSLE	Revised Universal Soil Loss Equation
PSIAC	Pacific Southwest Interagency Committee
EPM	Erosion Potential Method

MQCE	Method for the Quantitative Classification of Erosion
EA	Evolutionary Algorithms
ES	Evolutionary Strategies
GP	Genetic Programming
LADSS	Land Allocation Decision Support System
LP	Linear Programming
AWEC	Available Water Exchange Capacity
PH	Soil Reaction
CEC	Cation Exchange Capacity
EC	Exchange Capacity
ESP	Soil Alkalinity
CaCo ₃	Carbonates
LQ	Land Qualities
LC	Land Characteristics
U.S.D.A	United State of Drainage and Agriculture
AWHC	Available Water Holding Capacity
ISODATA	Interactive Self-Organizing Data Analysis
MLC	Maximum Likelihood Classifier
RMS	Root Mean Square
GCP	Ground Control Points
RMSE	Root Mean Square Error
DN	Digital Number
DOS	Dark Object Subtraction
UTM	Universal Transverse Mercator
OIF	Optimum Index Factor

PCA	Principle Component Analysis
ER	Effective Rainfall
WSP	Waste Special
IRS	Indian remote sensing
DEM	Digital Elevation Model
OM	Organic Matter
SOM	Soil Organic Matter
GCP	Ground Control Point
ANN	A Neural Network
CA	Cellular Automata
EAs	Evolutionary Algorithms
IP	Integer Programming

CHAPTER 1

INTRODUCTION

1.1 Background

With the increase in population, as well as human activities, pressure on land has been intensified (Bandyopadhyay, Jaiswal, Hegde, & Jayaraman, 2009).

Degradation of farmlands is happening with farming activities in the watersheds without proper management practices, such as measures to reduce soil losses due to the soil erosion, rainwater harvesting, the replacement of nutrients using organic matter the applications, etc. Different approaches of land evaluation have been developed, and each has a particular methodological procedure (FAO, 1976; Storie, 1933). The factor based land evaluation uses either single or multiple parameters converted to an integrated index (Guo et al., 2005). The (FAO, 1976) has recommended a framework for land suitability evaluation for crops in terms of suitability ratings ranging from highly suitable to not suitable based on climatic and terrain data and soil properties which is used in the methodology of current research. The success of the FAO framework (1976) and subsequent guidelines for application in diverse types of land uses and land areas (FAO, 1983, 1984, 1985a, 1991; Rossiter, 1996) is an encouraging development.

Land suitability is the fitness of a given type of land for a defined use. The process of land suitability classification is the appraisal and grouping of specific areas of land in terms of their suitability for defining uses (FAO, 1976). Land suitability could be assessed for present condition or after improvement (Potential Land Suitability). In land and crop suitability, information is based on physical environment data

generated from soil or land resources surveys. The information is based on soil characteristics and climate data related to the growth requirements of crops being evaluated.

1.2 GIS Methods for Land Suitability Evaluation

Improvements in information and communication technology that have caused to develop decision support models by the computer for land evaluation. Geographical information system (GIS) is one of the tools to implement to improvement land evaluation through the map analysis techniques. In land suitability evaluation, geospatial data (GIS layers, Global Positioning System (GPS) data and Remote Sensing (RS) imagery) and analysis significantly help in facilitating the decision-making process. GIS can be used in specific applications, ranging from the evaluation of land resource assessment and land-use planning using tools (FAO, 2007), such as components (GIS functions, data models and sources), software, operations (Malczewski, 2004) and evaluation of scenarios. Researchers can build geographic databases and GPS or other source such as RS technology (FAO, 2007) can import new research data. According to (Malczewski, 2004), there are three main groups of methods for the GIS land-use suitability analysis: 1- computer-assisted overlay, 2- multi-criteria assessment and; 3- Artificial intelligence (soft computing or Geo-computation) methods.

The computer-assisted techniques are developed as one the greatest response toward the manual method's limitations of mapping and combining of large datasets. Rather than manually mapping the values of a series of suitability factors in gray or colour scales, the models stored in a numerical forms as matrices in the computer. The integration of MCDM techniques with GIS has significantly been progressed the

conventional map overlay approaches to the land-use suitability analysis (Malczewski, 2004). GIS-based MCDA can assumed of as a process that combines and transforms spatial data (input) into a resultant decision (output). The MCDM procedures (or decision rules) define a relationship between the input maps and the output map. The procedures heavily consisted the utilization of geographical data, the decision maker's preferences and the manipulation of the data and preferences according to specified decision rules (Malczewski, 2006).

1.3 Artificial Intelligence Methods

Developments in spatial analysis showed that Artificial intelligence (AI) offers new opportunities to land use suitability analysis and planning (Fotheringham, Openshaw, & Abrahart, 2000) includes modern techniques of calculation that may help the modeling and description of complex systems for inference and decision making (Malczewski, 2006). AI is a generic term that covers several methods such as Evolutionary Algorithms (EAs), Genetic Programming (GP), Artificial Neural Networks (ANN), Cellular Automata (CA) and Fuzzy systems (Zadeh, 1994). The scope of Geo-Computation got sometimes used to cover these new computer-based techniques for analysis and modelling geographic data and solving spatial problems (De Smith, Goodchild, & Longley, 2007; Fotheringham et al., 2000). Table 1.1 shows comparative characteristics of the components of Soft Computing.

1.3.1 Genetic Algorithms (GAs)

GAs is a search method, originally developed by John Holland in the 1970s (Holland, 1975) and one part of the intelligence and evolutionary computation.

Table 1.1. Comparative characteristics of the components of Soft Computing, after,(Kennedy & Krouse, 1999; ogly Aliev & Aliev, 2001).

Strategy	Fuzzy Sets	Artificial Neural Networks	Evolutionary Computing, GAs
Weaknesses	<ul style="list-style-type: none"> • Knowledge Acquisition • Learning • needs a set of rules to be devised which can be difficult • rules can conflict with each other 	<ul style="list-style-type: none"> • Black Box Interpretability • not tolerant of missing data • can be 'overtrained' • can be inappropriately applied • duplicate results not dealt which well 	<ul style="list-style-type: none"> • Coding • Computational speed
Strengths	<ul style="list-style-type: none"> • Interpretability • Transparency • Plausibility • Graduality • Modeling • Reasoning • Very tolerant of noisy or highly variable input data and missing data. • Tolerance to imprecision 	<ul style="list-style-type: none"> • Learning • Adaptation • Fault tolerance • Curve fitting • Generalization ability • Approximation ability 	<ul style="list-style-type: none"> • Computational Efficiency • Global optimization • medium steadily improves • new medium automatically selected
Concept	Multivalued logic which executes a series of rules	Model that mimics the learning ability of the brain	Uses evolutionary natural selection process

Figure 1.1 shows the division of into Artificial Intelligence vs. Computational Intelligence (Fotheringham et al., 2000). GAs are optimization methods inspired by the biological processes of natural selection and survival of the fittest that allows a population composed of many individuals to evolve under specified selection rules to a state that maximizes the "fitness" (i.e., Minimizes the cost function) (Haupt, Haupt, & Wiley, 2004; Koomen & Stillwell, 2007; Malczewski, 2004). It is based on the principles of Darwin's evolutionary theory (Koomen & Stillwell, 2007).

1.3.2 Linking a Geographic Information System and GAs

GAs have been applied to spatial optimization problems (Aerts & Heuvelink, 2002; Matthews, 2001; (K. Matthews, Buchan, Sibbald, & Craw, 2002) By representing each geographical feature as a "layer" which it has been using in a different GIS

software so as to perform various sorts of spatial analysis of huge bulk number of geographic accurate data.

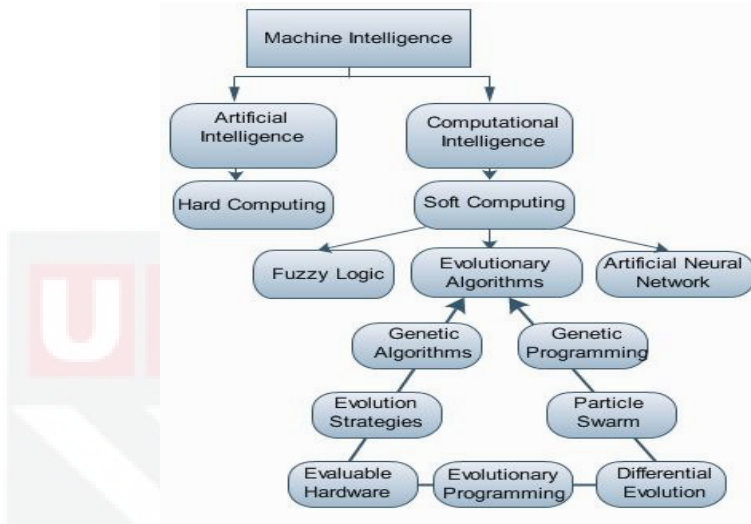


Figure 1.1. Artificial Intelligence vs. Computational Intelligence

All databases that come out from descriptive data will link to the GAs as an input and GAs population (Park, Choi, Wang, & Park, 2006). However, by combining GIs and integer programming (IP) with each other, they can significantly solve complexity in decision-making, and increase participation of stakeholders (Datta, Deb, Fonseca, Lobo, & Condado, 2007).

1.4 Problem Statement

Majority geographical problems are not directly solvable through the straightforward application of a specific methodology. Such these problems often require the participation of a variety of stakeholders with Varsity and often conflicting objectives (Branke, Deb, & Miettinen, 2008; Xiao, Bennett, & Armstrong, 2007). Allocating the Land evaluation at land use planning; for instance, requiring decision makers

minimize its economic cost, and minimize negative environmental effects. In the land use management, the incorporation of multiple objectives into decision-making and the search for suitable land use policies are known as the critical to the sustainable regional development. These and other types of multi-objective problems present a significant challenge to researchers for three main reasons. First, they are combinatorial optimization problems that often require a large amount of computation time to solve. Second, the search for solutions to these problems often involves the participation of stakeholders who have different backgrounds and view the problem from different perspectives. Finally, a solution that meets all criteria may not exist. Instead, stakeholders are required to examine trade-off among competing alternatives before a final solution can be reached (Xiao, Bennett, & Armstrong, 2007). As a consequence, it is important to develop some solution approaches that are (1) efficient in terms of their time complexity, (2) effective in terms of their ability to find a variety of high quality solutions, and (3) interactive so that decision makers can experiment with criteria, visually explore alternatives, and learn about a problem as they search for its solution.

Recognizing the links among agricultural/environmental policies, human decision making through land use choices, and environmental outcome can help design policies that directly affect incentives pertaining to land use and management. Consequently, in this study an agent-based model within a watershed-planning context is used to analyse the trade-off involved in producing a number of ecosystem services and agricultural commodities given a number of price and policy scenarios while assuming three different types of scenarios in terms of their goals. Most of the previous simulation studies used traditional mathematical programming methods lack the capability of modelling complex, human-decision- making process of

feedback and interaction of agents with the environment and among themselves, and they also lack in spatial specificity (Berger, Schreinemachers, & Woelcke, 2006).

In this research a multi-agent-based model that capture land suitability scenario, current land use and economic issues are developed by using a historic search and optimization technique called genetic algorithms (GAs) (Holland, 1975), which belongs to a broader class of Evolutionary Algorithms (EA).

To investigate regional scale and watershed, this required to collect the data and information from library resources and fieldwork. GIS has made a robust help in storing and analysing the data, as data layers. Using GIS and RS in evaluating and suitability of land and crop, not only improve the accuracy of the study, but also decrease the cost of evaluating by accelerating the investigation. In addition, majority of the GIS analysis is based on the overlade layers. By developing various sciences and increasing applications of the other sciences in GIS have been advancing the integrating of layers in this system and science relegated GIS, or inverse. Boolean Method was one of the popular methods. Boolean method observes zero and one value only. This method and the others such as index overlay, weight overlay, Multi Criteria Decision Making (MCDM) are one of the fundamental new developing methods in Geo-statistic and mathematic methods which they have been utilizing in GIS. Nowadays those mentioned methods indicate more accurate and precise results to users. New scientific methods in GIS and land-use suitability analysis have their origin backed in applications from hand-drawn overlay technique that used by landscape architects in the United State in the late nineteenth century and the beginning of modern 20th.

The important limitation of the classical overlay methods is lackness as to define mechanisms to judge (e.g. the decision-makers preference) the GIS, which, based on

legal procedures. This limitation is able to remove by integrating GIS and MCDM methods. The main problems related to the choice of method to combining different evaluation criteria, standardization of criterion maps, and the specification of criterion weights. MCDM depends on the experts and has some limitations in the numbers of the criteria, while the number of factors increase, the management and comparisons of the factors would be difficult, and may influence results. Different methods may produce different type results of in expert systems methods, different weighting methods would result in different overall land-use patterns. Some researchers suggested that these problems could be, at least partially, resolved by using the AI based methods (Malczewski, 2004).

All traditional artificial intelligence systems have been implemented using the Hard Computing technology, which restricts considerably the abilities of those systems. Moreover, the traditional artificial intelligence, does not accept the numerical methods, which are important for accounting for uncertainty and imprecision. Due to the above limitations, the Machine Intelligence Proportion for traditional artificial intelligence systems is not sufficiently high. Soft Computing methodology implies cooperative activity rather than autonomous one for such new computing paradigms as fuzzy logic, neural networks, evolutionary computation and others. This approach allows solving many important real-world problems, which were impossible to solve using traditional artificial intelligence methods (ogly Aliev & Aliev, 2001).

A number of optimization techniques have been proposed for the computation of the optimal allocation of land use within an area. However, most of these techniques are aimed at selecting optimal sites for a single land use type within an area. Heuristic algorithms have also been applied to predominantly single site allocation problems (T. J. Stewart, Janssen, & van Herwijnen, 2004).

Therefore, the AI methods have been used to eliminate possibility of enigmatic gaps and to improve results considerably. However, mentioned methods, (Overlay methods, Expert systems, etc) also used in AI methods in GIS environment. AI methods consist of Fuzzy, ANN, GA, and CA that used in set of spatial data complex, which eluded as “Geo-Computation”. Therefore, Geo-computation is an innovative research area within the field of GIS and geospatial analysis. For this reason, it is strongly influenced by recent developments in programming, data processing and interface design. Nowhere is this more obvious than in the concern for modelling of dynamic process (De Smith et al., 2007).

As comprehended before, it is worth mentioning that all previously mentioned methods have some gaps, which caused some restrictions for users. Accordingly, it is possible to integrate GAs with GIS and apply them as matrix base in solving related problems in the real world. In this method, mathematical and spatial analysis used in related to GAs analysis and its operators. It seems that integration of GA and GIS can remarkably decrease those gaps among previously mentioned methods. GIS are as the efficient and effective way of storing and presenting geo-referenced information. Both vector-based and grid-based systems can provide the input for optimization algorithms and can be used to present the results generated by these algorithms. If, however, the planning problem involves a large area and/or activities need to be allocated to small spatial units, then the amount of data to be used can be enormous. This requires that the algorithm be able to handle a large amount of data and that there be good communication between the algorithm and the GIS (T. J. Stewart et al., 2004). A few studies have done in land use planning by using of GA all around world. Specifically, according to the Iranian educational organization, there have not done any research thesis in related subject so far. To sum up, this method attempts to

use, more capabilities of the land suitability evaluation. Table 1.2 presents a comparison of their abilities at different fields of application, along with those of control theory and AI.

Table 1.2. comparison of some methods (Xiaoli, Chen, & Daoliang, 2009)

Method	Control Theory	Neural Network	Fuzzy Logic	Artificial Intelligence	Genetic Algorithm
Mathematical Model	Good	X	Fair	Need	X
Learning Data	X	Good	X	X	Good
Operator Knowledge	Needs	X	Good	Good	X
Real Time	Good	Good	Good	X	Needs
Knowledge representation	X	X	Needs	Good	X
Nonlinearity	X	Good	Good	Needs	X
Optimization	X	Fair	X	X	Good

Explanation of Symbols: Good=Good or suitable, Needs=Needs some other knowledge or techniques, X=Unsuitable or does not require.

Based on the aim of research, important and strategic crops (wheat, potato and almond) investigated. Land evaluation in this current study includes; land (irrigated and rein-fed) and crops suitability evaluation, in both rain-fed and irrigated land type use, and crop type, which has the capability to implant, regarding environmental and ecologic land potential. Reviewing related literatures revealed that, in the most aspects of them, only evaluation of land use or sort of crops was being considered.

In this study, the experts and residents' attitudes considered using the FAO framework in the field of land evaluation and land suitability for implanting each kind of rain-fed and irrigated crops. Therefore, it is hopeful to achieve, considered goals in the conservation of the soil and other ecological potential of the region, and lead them to an economic benefit to all residents. Socioeconomic assessments remarkably have been considered in the majority of land suitability investigations and biophysical land evaluation model. This fact is a gap in the results. On the other word, evaluating of the crop evaluation without socioeconomic assessments is a

problem for planners and decision makers. In land evaluation FAO model (1976), the socioeconomic assessments are considered. According to this research, both land suitability and crop suitability evaluations considered.

1.5 The objectives of research

This study analyzes the multi-functional-based model, in the decision making process, on the possible economic and environmental outcome for policy scenarios and change in agricultural/ environmental policies such as soil conservation. This study tests the robustness of the developed agent-model in accurately capturing the variations in the decision-making process of various scenario defined and three main crop (wheat, potato and almond) due to variations in endogenous (e.g. agents value) and exogenous factors (e.g. market price) compared to the actual land use land cover map. This research has to provide an overview of recent developments in multi-objective problem solving. This believed that borrowing ideas from other fields will benefit research on spatial decision-making. Evolutionary algorithms such as GAs are particularly important because they can be used to solve multiobjective geographical problems efficiently, effectively, and often interactively. In this study uses the GAs coding to optimizing land use/ land cover obtained from remote sensing processing of the study area and determining the best current economic benefit of the region.

The main purpose of this research is; to optimizing the developed FAO's Evaluation Model (1976) and land use/ land cover in Menderjan Watershed using GIS, RS and GAs; These Objectives are mentioned as follows:

- To determine Land Suitability Analysis for three main crops (wheat, potato and almond).

- To determine Fitness Function by Presenting Water Erosion spatial Pattern and estimating economical amount of cost/ benefit crops and soil loss.
- To investigation the use of GAs to optimize land use allocation.

1.6 Significance of the Study

The most significant part of this research divided into two paramount aspects, firstly matters of great moments, which are existed in the case study from different points of economical and agricultural issues which all have to meticulously consider important in the study, and secondly using different methods both technical or skilful procedures to solve above problems. It is valuable to state that the land potential and agriculture in Menderjan basin, which it is one of the important catchments in “Zayandeh Rood” Watershed, is vital for residents and planners. Further comprehensive study has not implemented in Menderjan Watershed yet. Therefore, this is an innovative study and it will be both significant for sustainable development of Menderjan and planners. The technical and methodological significant in this study uses of the FAO model (1976-2007) as basic. Author attempts to contribute the expert’s advice using GIS and RS to optimize the FAO framework for Menderjan Basin. In addition, in this research, the mathematical methods, statistical and spatial analysis, and programming used to achieve technical significant. GAs as an optimization method, which is using in this study has widely used to solve one or more problems of optimization goals (Câmara, Monteiro, 2001). As the result of surveys in this research we can include that the significant current research method, which is a matrix based in terms of using capabilities of GAs, operators and elements of GAs in a dynamic ambiances is being innovated; meanwhile these GIS layers

introduce as an input of GAs. The contributions and novelty of current research are as below:

- According to the literature review, most GAs studies have been done in both the industries and electronics fields, and in the Geomatic fields also; surveys toward GAs studies has an enormous worth of further researching and studying as a novelty issue.

- Capability in handling the great numbers of pixels.

1.7 Study Area

“Zayandeh Rood” river is one of the most important rivers of Iran which is vital for the city of Esfahan as drinking water and the alluvial plain for agriculture. The river has a very vast watershed and extends over Esfahan and Charmahal& Bakhtiari Provinces.

According to the Isfahan Directorate of Watershed Management of Jihad Agriculture Organization reports, more than 100 families have left their village for the cities. Insufficient income was found to be the main reason for immigration. Isfahan, Gom and Tehran are attracted cities for the immigrants. The most important problem of Menderjan Basin is land destruction whereas the most principle economic activities are agriculture and ranching. Consequently, superfluity to agriculture lands and natural resources has destructed the soil and its fertility and has decreased the economic benefits of the residents. In the long time, residents have been engaged in irrigated and rain-fed agriculture to crop wheat, barley, potato and almond, grazing have been a long history in this region. The watershed has been divided into 28 basins or sub-catchments; each of them is called a parcel. The research area is one of these parcels, named B2 or Menderjan, which is selected. Zayandeh Rood Dam is

located above the Zayandeh Rood River in Chadegan City in Isfahan Province and is important in terms of agriculture. Figure 1.2 shows the location of Menderjan watershed. A considerable amount of information is available for the area and therefore the research can be run more fluently. It is located west of Esfahan, on the western side of Chadegan city and north of the Zayandeh Rood dam. Its geographic location is $50^{\circ} 7' 16''$ to $50^{\circ} 40' 34''$ E and $32^{\circ} 45' 12''$ to $32^{\circ} 56' 48''$ N. The slope of thirty percent of the area is between 30 to 60 percent and nearly half of the area is located between 2200 to 2400 meters elevation. The most important physical characteristics of the study area are defined in Table 1.3. According to Coppen's method, the area falls in the continental moderate or cold agro-ecological zone with cool summers and very cold winters. The average monthly maximum temperature is 10°C and the average monthly minimum temperature minus 5°C , with the absolute maximum temperature of 38.5°C . The area has a Mediterranean type rainfall, characterized by rainy winters and dry summers. Annual rainfall concentrated in winter is about 332 mm. The rainy season starts in October and ends in March. Fifty percent of the precipitation is snowy; and the other scope is rainy which falls in winter and spring. Stats Total Days of Ice Station Chadegan months of that December, January and February in all those days of the freezing order freeze the average number of days in a year is 103 days. Agricultural practices in Menderjan can be divided into two sectors, irrigated and rain-fed.

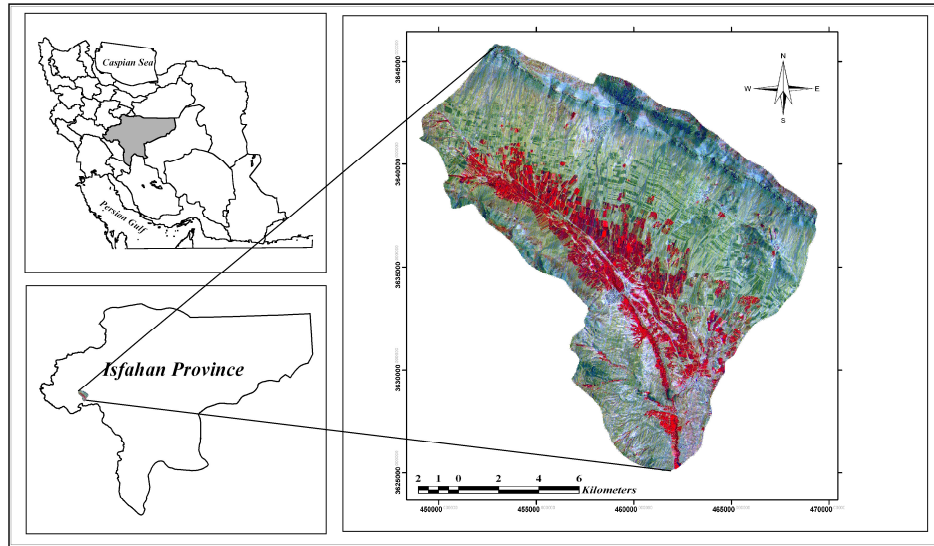


Figure 1.2. Location of the study area (Color Composite RGB 432, IRS- P6)

Table 1.3. Physical characteristics of the study area

Land Characteristic	Description
General Aspect of Slope	South
Area	226.8 km ²
Perimeter	69.5 km
Mean altitude	2430 m
Maximum altitude	3642 m
Minimum altitude	2100 m
Length of the longest channel	22.8 km
Rough slope of the main channel	2.2%
Net slope of the main channel	14.3%
Average slope of the basin	13.3%
Compactness coefficient	1.51
Gravellus coefficient	1.29
Form factor	0.436
Time of concentration	4.67 hr
Stream frequency	1.87 km/km ²

Rain-fed agriculture: is a cause of erosion in the study area. Fertile soil washed away rapidly, and land productivity decreases because of rain-fed agriculture. Principal crops are winter wheat and winter barley. Crops harvested in good years and grazed in bad years.

Irrigated agriculture: The principle-irrigated crops are wheat, barley, potato and forage such as alfalfa. Sources of water are rivers, wells and small streams. The Esfahan Water Organization reports that 121 wells, 35 streams, and 66 Ghanats are available in the Menderjan watershed. The Province Agricultural Organization has reported that the surface areas occupied by gardens are 1036.3 hectares, and other crops 7680 hectares.

1.8 Layout of the Thesis

This thesis organized into five chapters. The first chapter is the introduction, which gives a background of the Land Evaluation, Land Use Planning, Land Crop Suitability, GIS and Genetic Algorithms. It also brings forward the idea of Genetic Algorithms that is one of the parts in Geo-computation, Artificial Intelligence and Water Demand Model. The objectives and significance of the study mentioned in this chapter. Chapter 2 is the literature review, which gives an overview of the FAO Land Evaluation, Land Use Planning, Land Crop Suitability and GIS and Genetic Algorithm. A water demand model discussed, as it is the important aspect of the research work. Chapter three present discussions of Land Evaluation, Land Use Planning, Land Crop Suitability in the Menderjan basin using GIS, RS and Genetic Algorithms. The study area and the used method and material highlighted. The results and discussion presented in Chapter 4. In chapter five, summaries of the work presented, suggestions of possible areas of improvement are given and the research concluded.

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