

GEOSPATIAL APPROACH FOR FLOOD VULNERABILITY ASSESSMENT IN  
KELANTAN RIVER BASIN, MALAYSIA

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GEOSPATIAL APPROACH FOR FLOOD VULNERABILITY ASSESSMENT IN  
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## **DEDICATION**

This thesis is dedicated to my beloved family, who have supported me throughout my academic life.

Also, this work is dedicated to my dear Ab Latif Ibrahim, who serve as my enlightenment teacher.

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## ABSTRACT

Flood is the most destructive natural disaster in Malaysia, causing significant economic and human life losses. Recent flood management paid more attention to flood risk assessment to provide information on risk of flooding. It consists of three components i.e., flood hazard, flood exposure, and flood vulnerability, which can be supported directly and indirectly by remotely sensed data. The research aims to utilise a geospatial approach for flood vulnerability assessment in the Kelantan River Basin (KRB), Malaysia. For flood hazard modelling, the flood event in December 2014 over Kelantan has been simulated using the Rainfall-Runoff-Inundation (RRI) model. The model is supported by four different Satellite Rainfall Products (SRPs) (i.e., Integrated Multi-satellitE Retrievals-Late, -Early (IMERG-L, -E), Precipitation Estimation from Remotely Sensed Information using Artificial Neural Networks-Cloud Classification System (PERSIANN-CCS) and Global Satellite Mapping of Precipitation (GSMaP)) and rain gauge data. The simulation results were compared to the observed discharge flow. For the flood exposure analysis, a framework to extract detailed element-at-risk information was developed by classifying detailed land use and land cover in heterogeneous urban areas of Kota Bharu using Very-High-Resolution (VHR) satellite imagery and Light Detection and Ranging (LiDAR) data. The Feature Selection (FS) algorithm was also adopted to achieve an efficient classification framework. Reference points extracted from existing land use was used to validate the classification results. Finally, a geospatial approach for flood vulnerability assessment of buildings has been developed, which combines outputs from flood hazard and flood exposure. The framework incorporates relevant building parameters derived from the VHR satellite image and LiDAR data, and flood hazard for a vulnerability analysis of building using Machine Learning (ML) approach. The geospatial approach-based vulnerability results were validated using in-situ flood damage data derived from questionnaire method. The results of flood inundation showed that the GSMaP has the best performance in simulating hourly runoff with the lowest relative bias (RB) and the highest Nash-Sutcliffe efficiency (NSE) of 4.9% and 0.79, respectively. The results of image classification showed that there was a significant difference between classification accuracies using two datasets (50 features and 107 features) in object-based approach. The overall accuracy increased to 93.7% from 79.2% using the Random Forest (RF) classifier. Nevertheless, the RF classifier have achieved significantly better classification results compared to other classifiers (k-NN and Support Vector Machine (SVM)). The extreme gradient boosting (xgbTree) FS method improved the classification accuracy from 93.7% to 94.1% using 26 features. However, there is a statistically significant difference between the results produced by the Recursive Eliminate Feature (RFE) and Simulated Annealing (SA). The result of vulnerability showed that the geospatial approach achieved a good prediction result with an overall accuracy of 81.8%. In conclusion, the SRPs can be used to support spatially distributed flood hazard assessment in a scarcity rain gauge station area. High resolution remotely sensed data can be used to extract information of element-at-risk in highly heterogeneous location in Malaysia, which allows ML method for flood vulnerability assessment.

## ABSTRAK

Banjir adalah bencana alam yang paling merosakkan di Malaysia, menyebabkan kerugian ekonomi dan kehilangan nyawa manusia yang ketara. Pengurusan banjir baru-baru ini memberi lebih perhatian kepada penilaian risiko banjir untuk memberikan maklumat tentang risiko banjir. Ia terdiri daripada tiga komponen: komponen bahaya, komponen pendedahan dan komponen kerentanan. Data penderiaan jauh boleh, secara langsung dan tidak langsung, menyokong komponen bahaya dan pendedahan, dan komponen kerentanan, masing-masing. Kajian ini bertujuan untuk menggunakan pendekatan geospasial untuk penilaian kerentanan banjir di Lembangan Sungai Kelantan (KRB), Malaysia. Untuk pemodelan bahaya banjir, kejadian hujan melampau Disember 2014 di Kelantan telah disimulasikan menggunakan model RRI dengan empat Produk Hujan Satelit (SRP) berbeza (Pendapatan Berbilang Satelit Bersepadu-Lewat, -Awal (IMERG-L, -E); Anggaran Kerpasan daripada Maklumat Penderia Jauh menggunakan Rangkaian Neural Tiruan-Sistem Klasifikasi Awan (PERSIANN-CCS), Global Satelit Permetaan Kerpasan (GSMaP) dan data tolok hujan. Hasil simulasi telah dibandingkan dengan aliran luahan dicerap. Untuk analisis pendedahan banjir, rangka kerja untuk mengklasifikasikan penggunaan tanah terperinci dan pemetaan tutupan tanah di kawasan bandar heterogen di Kota Bharu menggunakan Resolusi Sangat Tinggi (VHR) imej satelit dan data pengesanan dan julat cahaya (LiDAR) untuk pengenalpastian elemen banjir yang berisiko telah dicadangkan. Sehubungan itu, algoritma Pemilihan Ciri (FS) telah diterima pakai untuk mencapai rangka kerja pengelasan yang cekap. Titik rujukan yang diekstrak daripada penggunaan tanah sedia ada telah digunakan untuk mengesahkan keputusan pengelasan. Akhirnya, pendekatan geospasial untuk penilaian kerentanan banjir bangunan telah dibangunkan, yang menggabungkan output daripada bahaya banjir dan pendedahan banjir. Rangka kerja ini menggabungkan parameter bangunan berkaitan yang diperolehi daripada imej satelit VHR dan data LiDAR, dan bahaya banjir untuk analisis kerentanan bangunan menggunakan pendekatan Pembelajaran Mesin (ML). Keputusan kerentanan berasaskan pendekatan geospasial telah disahkan menggunakan data kerosakan banjir in-situ yang diperolehi daripada kaedah soal selidik. Keputusan bahaya menunjukkan bahawa GSMaP mempunyai prestasi terbaik dalam mensimulasikan luahan setiap jam dengan bias relatif terendah (RB) dan kecekapan *Nash-Sutcliffe* (NSE) tertinggi masing-masing sebanyak 4.9% dan 0.79. Keputusan menunjukkan terdapat perbezaan yang signifikan secara statistik antara klasifikasi tentang ketepatan menggunakan dua set data (50 ciri berbanding 107 ciri) dalam pendekatan berasaskan objek. Ketepatan keseluruhan meningkat kepada 93.7% daripada 79.2% menggunakan pengelas RF. Namun begitu, pengelas RF telah mencapai keputusan pengelasan yang jauh lebih baik berbanding dengan pengelas lain (*k*-NN dan SVM). Walaupun ketepatan klasifikasi telah bertambah baik, pengelasan telah mengambil masa yang lebih lama. Peningkatan kecerunan ekstrim (XgbTree) meningkatkan ketepatan kepada 94.1% daripada 93.7% menggunakan 26 ciri, manakala masa pemrosesan dikurangkan kepada 56s daripada 634s. Walau bagaimanapun, terdapat perbezaan yang signifikan secara statistik antara keputusan yang dihasilkan oleh Ciri Penyingkiran Rekursif (RFE) dan Penyepuhlindapan Simulasi (SA). Kedua-dua kaedah telah dilatih oleh algoritma pembelajaran, dengan kata lain, lebih berhati-hati harus diambil untuk menentukan algoritma pembelajaran yang disesuaikan dengan pambalut berasaskan RFE dan SA. Keputusan menunjukkan bahawa rangka kerja berasaskan penderiaan jauh mencapai keputusan ramalan yang baik dengan ketepatan keseluruhan 0.81 dengan menggunakan pendigitalan poligon dan ciri geometri. Kesimpulannya, SRP boleh digunakan untuk menyokong penilaian bahaya banjir yang diedarkan secara ruang di kawasan stesen tolok hujan kekurangan. Data deria jarak jauh resolusi tinggi boleh digunakan untuk mengekstrak maklumat elemen berisiko di lokasi yang sangat heterogen di Malaysia, yang membolehkan kaedah ML untuk penilaian kerentanan banjir.

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## LIST OF ABBREVIATIONS

1D	-	one dimensional
1D2D	-	integrated one-dimensional and two-dimensional
2D	-	two dimensional
ACC	-	Flow accumulation
AMS	-	annual maximum discharge
ASCII	-	U.S. Standard Code for Information Interchange
ASEAN	-	Association of Southeast Asian Nations
AWGWRM	-	ASEAN Working Group on Water Resource Management
bil	-	band interleaved by line
CART	-	classification and regression of trees
CC/ <i>r</i>	-	correlation coefficient
CCA	-	Canonical Correlation Analysis
CCI	-	Coping Capacity Index
CFS	-	Correlation-Based Feature Selection
CLC	-	CORINE Land Cover
CREST	-	Core Research for Evolutional Science and Technology
CREST	-	Coupled Routing and Excess Storage
CS	-	Chi-Squared
DEM	-	Digital Elevation Model
DID	-	Department of Irrigation and Drainage
DIR	-	Drainage directions
DRR	-	Disaster Risk Reduction
DSM	-	Digital Surface Model
DT	-	Decision Tree
EAD	-	expected annual damage
EC	-	European Commission
EEA	-	European Environment Agency
EM-DAT	-	Emergency Events Database

ENVI	-	Environment for Visualizing Images
EO	-	Earth observation
EO-1	-	Earth Observation One
ESA	-	European Space Agency
ESP	-	Estimation of Scale Parameter
EU	-	European Union
FAO	-	Food and Agriculture Organisation
FE	-	feature extraction
FNEA	-	Fractal Net Evolution Approach
FRM	-	Flood Risk Management
FS	-	Feature Selection
FSO	-	Feature Space Optimisation
FVI	-	Flood Vulnerability Index
GA	-	Genetic Algorithm
GEO	-	geostationary
GIS	-	Geographical Information System
GLCM	-	grey-level co-occurrence matrix
GLCNMO	-	Global Land Cover by National Mapping Organisations
GPM	-	Global Precipitation Mission
GR	-	Gain Ratio
GSMaP_Gauge	-	GSMaP standard_gauge
GSMaP_Gauge_NRT)	-	GSMaP realtime_gauge
GSMaP_Gauge_RNL	-	GSMaP reanalysis_gague
GSMaP_MVK	-	GSMaP standard
GSMaP_NOW	-	GSMaP now
GSMaP_NRT	-	GSMaP realtime
GSMaP_RNC	-	GSMaP riken_nowcast
GSMaP_RNL	-	GSMaP reanalysis
GSMaP-NRT	-	Global Satellite Mapping of Precipitation-Near- Real-Time
GWP	-	Global Water Partnership
HEC	-	Hydrologic Engineering Center

HWSD	-	Harmonized World Soil Database
HydroSHEDS	-	Hydrological data and maps based on Shuttle Elevation Derivatives at multiple Scales
ICAMS	-	Image Characterization and Modeling System
IFM	-	Integrated Flood Management
IG	-	Information Gain
IMERG	-	Integrated Multisatellite Retrievals for GPM
IMERG-E	-	Early Run
IMERG-F	-	Final Run
IOs	-	image objects
IR	-	infrared
ISCGM	-	International Steering Committee for Global Mapping
ISODATA	-	Iterative Self-Organizing Data Analysis Techniques
IWRM	-	Integrated Water Resources Management
JICA	-	Japan International Cooperation Agency
JST	-	Japan Science and Technology Agency
<i>k</i> -NN	-	<i>k</i> -Nearest Neighbor
KRB	-	Kelantan River Basin
LCCS	-	Land Cover Classification System
LDA	-	Linear Discriminant Analysis
LEO	-	low-earth orbit
LiDAR	-	Light Detection and Ranging
LULC	-	land use land cover
LV	-	local variance
MacGDI	-	National Geospatial Centre
maxdepth	-	Maximum depth
MERG-L	-	Late Run
minsplit	-	Minimum number of observations in a node
ML	-	Machine Learning
ml	-	Maximum Likelihood
MODIS	-	Moderate Resolution Imaging Spectroradiometer
MSAN	-	National Water Resources Council

MSMA	-	Storm Water Management Manual
mtry	-	Number of variables
MW	-	Mann-Whitney
NB	-	Naïve Bayes
nDSM	-	normalized Digital Surface Model
NetCDF	-	Network Common Data Form
NLCD	-	National Land Cover Database
NOAA	-	National Oceanic and Atmospheric Administration
NRT	-	Near-Real-Time
NSCE	-	Nash-Sutcliffe coefficient efficiency
NSE	-	Nash-Sutcliffe Efficiency
ntree	-	Number of trees
OA	-	overall accuracy
OBIA	-	object-based image analysis
OOB	-	out-of-bag
PBIA	-	pixel-based image analysis
PCA	-	Principle Component Analysis
		Precipitation Estimation from Remotely Sensed
PERSIANN-CCS	-	Information using Artificial Neural Networks-Cloud Classification System
PMM	-	Precipitation Measuring Mission
PMW	-	passive microwave sensors
PR	-	active microwave sensors
R <sup>2</sup>	-	coefficient of determination
RB	-	relative bias
RF	-	Random Forest
RFE	-	Recursive Feature Elimination
ROC-LV	-	rates of change of LV
RRI	-	Rainfall-Runoff-Inundation
RVS	-	Rapid Visual Screening
SA	-	Simulated Annealing
SAR	-	Synthetic Aperture Radar
SCS-CN	-	Soil Conservation Service – Curve Number

SFS	-	Sequential Forward Selection
SFVI	-	social flood vulnerability index
SOFM	-	self-organizing features map
SoVI	-	social vulnerability index
SRPs	-	Satellite rainfall products
SRTM	-	Shuttle Radar Topography Mission
SSI	-	social susceptibility index
SVM	-	Support Vector Machine
SVM <sub>L</sub>	-	Support Vector Machine with the linear kernel
SVM <sub>RBF</sub>	-	Support Vector Machine with radial basis function kernel
SVM-RFE	-	Support Vector Machine – Recursive Feature Elimination
SWAT	-	Soil & Water Assessment
tiff	-	Tagged Image File Format
TRMM	-	Tropical Rainfall Measuring Mission
USGS	-	States Geological Survey
UTM	-	Universiti Teknologi Malaysia
VHR	-	very-high-resolution
WFIUH	-	Width Function Instantaneous Unit Hydrograph
WT	-	Watershed Transformation
xgbTree	-	eXtreme Gradient Boosting
RSO	-	Rectified Skew Orthomorphic



## LIST OF SYMBOLS

$h$	-	height of water from the local surface
$q_x$	-	width discharges in $x$ direction
$q_y$	-	width discharges in $y$ direction
$u$	-	flow velocities in $x$ direction
$v$	-	flow velocities in $y$ direction
$r$	-	rainfall intensity
$f$	-	infiltration rate
$H$	-	height of water from the datum
$\rho_w$	-	density of water
$g$	-	gravitational acceleration
$\tau_x$	-	shear stresses in $x$ direction
$\tau_y$	-	shear stresses in $y$ direction
$n$	-	Manning's roughness parameter
$\text{sgn}$	-	signum function
$q_x^{i,j}$	-	$x$ direction discharges from a grid cell at $(i,j)$
$q_y^{i,j}$	-	$y$ direction discharges from a grid cell at $(i,j)$
$k_a$	-	lateral saturated hydraulic conductivity
$da$	-	soil depth times the effect porosity
$k_m$	-	unsaturated zone
$k_v$	-	vertical saturated hydraulic conductivity
$\phi$	-	soil porosity
$\theta_i$	-	initial water volume content
$S_f$	-	suction at the vertical wetting front
$F$	-	cumulative infiltration depth
$f$	-	infiltration loss
$C$	-	regularization parameter / penalty parameter
$\gamma$	-	kernel width / gamma parameter
$d$	-	degree of polynomial kernel
$D$	-	depth of river
$W$	-	width of river

A - Area basin

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# CHAPTER 1

## INTRODUCTION

### 1.1 Introduction

Every year, floods cause massive damage throughout the world. As a result, many areas, particularly low-lying regions, are at severe risk of flooding. The effects of global warming will exacerbate the flooding in these areas as sea levels rise (Le, Nguyen, Wolanski et al., 2007; Mousavi, Irish, Frey et al., 2011). However, floods also have beneficial effects, whereby fine sediments and nutrients floating along with the floodwater can fertilise floodplains and aquatic inhabitants (Hofer and Messerli, 1997; Bonyongo, Mubyana, Totolo et al., 2002). Such benefits of flooding are important to regions that depend heavily on agriculture and fisheries, such as the Mekong delta and the Ganges–Brahmaputra delta. Nevertheless, the negative consequences of floods are still immense. Flooding was the most common form of natural disasters (47%) between 1995 and 2015, affecting 2.3 billion people, resulting in financial losses worth USD 662 billion (UNISDR, 2015). Approximately 800 million people live in flood-prone areas, of whom 70 million encounter annual flooding (UNISDR, 2011). The effects of urbanisation and climate change are the key factors that could raise the risk of flooding (Kundzewicz, Kanae, Seneviratne et al., 2013; Arnell and Gosling, 2016), making 2 billion people vulnerable to annual flooding by 2050 (UHU, 2004).

Developed areas are extremely complicated environments that contain complex components in terms of their social structure, built environment and natural environment. As a consequence, flood damage affects multidimensional components. In general, flood damage has been categorised into direct or indirect damage types, while it can be further defined as tangible or intangible damage (Chan, 2015b; Jonkman, Bockarjova, Kok et al., 2008; Messner and Meyer, 2006). Direct damage is caused by losses due to direct interaction between the floodwaters and the human body,

physical structures or physical objects and their contents; indirect damage refers losses that arise from flood occurrence without interaction with the floodwater. For instance, this could be a loss of production among businesses that are outside the affected area or a loss of trust in the relevant authorities. Tangible losses are reflected in monetary value, whereas the intangible losses are not measured in such terms, they include anxiety and nervousness.

The goal of flood management is to reduce flood damage and achieve sustainable development (Chan and Parker, 1996). Conventional flood management, which focuses pivotally on managing flood hazards (Merz, Kreibich, Schwarze et al., 2010), has concentrated on controlling or minimising the flood hazard, such as by decreasing the frequency of such events and the severity of inundation. However, past experiences have demonstrated that absolute protection or structural measures (i.e., river enhancement, and the construction of dykes, levees and dams) are unsuccessful against recurrent floods (Chan, 1997a; Kundzewicz and Takeuchi, 1999). In addition, a dam break will aggravate the flood conditions in downstream areas (Fread, 1993). On the other hand, non-structural measures (e.g., land use planning, flood forecasting, insurances and flood mapping) are an indispensable tools for providing useful information to decision makers or engineers so they can implement effective structural measures to cope with the floods (Chan, 1997a; Simonovic, 2002; Zakaria, Zin, Mohamad et al., 2017).

Flood Risk Management (FRM), a modern concept in the flood management framework, incorporates structural and non-structural measures. FRM is a synthesis of two elements, flood risk reduction and flood risk assessment (Schanze, 2006). The aims of flood risk assessment are to provide information on the risk of flooding and identify areas of unacceptably high risk. The flood risk reduction approach focuses on identifying effective measures based on a risk assessment output and reducing the risk of flooding to an acceptable level. In Malaysia, the Department of Irrigation and Drainage (DID) has implemented the Flood Risk Management concept for flood control (Chan, 2015b). Understanding flood risk is the priority process established in the Flood Directive (2007/60/EC) and the Sendai Framework for Disaster Risk Reduction 2015-2030. A comprehensive understanding of flood risk enables more

efficient and effective flood management in terms of reducing flood damage to an acceptable level. Flood risk comprises three components: hazard, exposure, and vulnerability. Flood hazard is defined as the probability of an extreme event occurring in a given area during a specific timeframe. An example of a flood hazard is river discharge and water levels that have corresponding exceedance probability. Flood exposure/elements-at-risk comprises any environments, assets, and people at risk of flooding. Flood vulnerability refers to the degree to which those exposed elements are susceptible to damage. Flood risk is defined as the likelihood that floods of a particular intensity and loss (monetary damage) will occur in a specific area within a given period (Merz, Thielen and Goch, 2007).

Earth observation (EO) satellites play a key role in Disaster Risk Reduction (DRR) by providing information on the various risk management processes, including preparedness, the emergency response, mitigation, and recovery (Tralli, Blom, Zlotnicki et al., 2005; Gillespie, Chu, Frankenberg et al., 2007; CEOS, 2003; Deichmann, Ehrlich, Small et al., 2011; Voigt, Giulio-Tonolo, Lyons et al., 2016; Denis, de Boissezon, Hosford et al., 2016). The use of satellite technology in natural disasters management is rising, especially during the response stage. Between 2006 and 2007, satellite use increased dramatically due to technological advancements such as internal mechanism enhancements and virtual online globes, which increased the awareness and adoption of geographical data (Voigt et al., 2016). The key reasons for this are the disaster community's awareness and appreciation of the use of geospatial data (Voigt et al., 2016), as well as the EO satellites' capable to deliver timely and geospatial information on areas affected by an event (Denis et al., 2016). Using EO satellites as a form of disaster support has various advantages: satellite are not vulnerable the disaster itself, satellite images are consistently collected on multiple scales, remotely sensed hazardous areas can be incorporated all disaster management phases, and the data is free (CEOS, 2015).

In the context of flood risk components, remote sensing data can contribute directly to the hazard and exposure components (CEOS, 2015; Deichmann et al., 2011; Taubenböck, Wurm, Netzband et al., 2011), and indirectly to the vulnerability component (Mueller, Segl, Heiden et al., 2006; Panagiota, Erwan, Philippe et al.,

2012a). For example, satellite imagery has been used extensively in flood modelling to parameterise hydrological or hydraulic models, as well as delineate flood maps to validate flood models (Bates, Horritt, Smith et al., 1997; Bates, 2004; Bates and De Roo, 2000). Furthermore, the provision of rainfall data can be an alternative source in areas where there is no rain gauge, while the generation of a Digital Elevation Model (DEM) uses stereo satellite imagery, as does flooded area delineation. In mapping flood exposure, analysing the earth observation data is an important method for distinguishing built-up areas, buildings and their characteristics (Ehrlich, Kemper, Blaes et al., 2013; Deichmann et al., 2011), which serve as key inputs with which the flood vulnerability assessment estimates vulnerability level of a building in a flood hazard area. The current study provides an operational context for the use of remote sensing data to promote the reduction of flood risk in Malaysia.

## **1.2 Problem Statement**

The development of an official flood vulnerability map in Malaysia continues to be relatively lacking compared to flood hazard mapping (Zakaria et al., 2017). Numerous studies on flood vulnerability assessment in Malaysia have been conducted (Ho, 2009; Tam, Ibrahim, Rahman et al., 2014; Elsheikh, Ouerghi and Elhag, 2015; Romali, 2018; Mojaddadi, Pradhan, Nampak et al., 2017; Udin and Malek, 2018b). However, none of these have focused on assessing micro-scale flood building vulnerability or used the remote sensing approach to facilitate micro-scale flood building vulnerability assessment. The term "micro-scale" refers to a study in which the unit of analysis is a single object or building (Gerl, Kreibich, Franco et al., 2016; Jongman, Kreibich, Apel et al., 2012a). Flood hazard and exposure (elements-at-risk) are the two components that must exist before conducting a micro-scale flood building vulnerability assessment as vulnerability does not apply if the building is located outside a flood hazard area.

The scarcity of rain gauge stations in basins is a major issue for the flood hazard component (Croke, Islam, Ghosh et al., 2011; Long, Zhang and Ma, 2016; Li, Yang and Hong, 2013; Dile and Srinivasan, 2014) because these stations represent the most

important input variables in hydrological modelling (Croke et al., 2011). Malaysia faces this problem (Osman and Abustan, 2011; Patrick, Mah, Putuhena et al., 2017). Therefore, due to their high spatial-temporal characteristics and broad coverage, satellite rainfall products (SRPs) have become an alternative source of rainfall data with which to characterise rainfall patterns in data-sparse conditions (Stisen and Sandholt, 2010; Bui, Ishidaira and Shaowei, 2019). The reliability of SRPs must be examined before their use for hydrological applications. As such, studies have been conducted in Malaysia to assess SRPs (Semire, Mohd-Mokhtar, Ismail et al., 2012; Tan, Ibrahim, Duan et al., 2015; Mohd Zad, Zulkafli and Muharram, 2018; Tan and Santo, 2018b; Varikoden, Samah and Babu, 2010; Tan, Gassman and Cracknell, 2017). These studies did not focus on flood hazard modelling nor investigate the performance of near-real-time products (temporal resolution up to 30 minutes to one hour), such as Global Satellite Mapping of Precipitation-Near-Real-Time (GSMaP-NRT) and Precipitation Estimation from Remotely Sensed Information using Artificial Neural Networks-Cloud Classification System (PERSIANN-CCS). These have not yet been evaluated in Malaysia, particularly in characterising an extreme rainfall event. Hence, this study attempted to examine how these products performed in characterising extreme rainfall-induced flooding events.

The results of the detailed land use/land cover classification used Very-High-Resolution (VHR) satellite imagery and LiDAR data, and the findings were overlaid on flood inundation maps to identify flooded buildings as elements-at-risk. The highly varied building rooftops colours increased the variation of the intra- and inter-class spectral responses caused by renovation or degradation, consequently posing a considerable challenge for image classification. In Malaysia, several studies have focused on using Very-High-Resolution (VHR) satellite imagery and LiDAR data to classify detailed land use/land cover (Rizeei, Pradhan and Saharkhiz, 2018; Ziaei, Pradhan and Mansor, 2014; Su, Li, Chen et al., 2008) as well as conducted intra-urban mapping (Gibril, Shafri and Hamedianfar, 2017; Hamedianfar and Shafri, 2015). Most of these studies utilised the standard image features (i.e., the grey level co-occurrence matrix, indices, spectral bands, and topography). Hypothetically, these features are inadequate for differentiating the building types. To address this issue, more spatial features were considered in the classification process used in this study.



Using large image features can lead to model overfitting due to redundant and insignificant information, which creates a complex model that eventually requires a longer computation time to complete a classification task (Georganos, Grippa, Vanhuysse et al., 2017a). Feature Selection (FS) is one method that has been used to address these issues, and its value has been proven in previous studies (Yu, Gong, Clinton et al., 2006; Laliberte, Browning and Rango, 2012; Zhou, Zhang, Wang et al., 2018; Cánovas-García and Alonso-Sarría, 2015). However, these researchers did not thoroughly evaluate the wrapper method to select a subset of features by training various machine learning algorithms on them. Hence, the present study attempted to compare the performance of two wrapper methods (Recursive Elimination Feature and Simulated Annealing) used to wrap different machine learning algorithms and select a group of features. Hence, this study attempted to investigate a feature selection approach to improve the operational framework of detailed land use/land cover classification.

Flood vulnerability assessment has widely been conducted in Malaysia (Mojaddadi et al., 2017; Ibrahim and Asmawi, 2018; Roslee, Tongkul, Mariappan et al., 2018; Wahab and Muhamad Ludin, 2018; Lawal, Matori, Yusof et al., 2014a; Matori, Lawal, Yusof et al., 2014b). However, none of these studies focused on micro-scale flood vulnerability assessment, which had been reported by de Ruiter, Ward, Daniell et al. (2017) when conducting a comprehensive review to compare flood and earthquake vulnerability assessments. Furthermore, the traditional method of determining the vulnerability level of a building is time-consuming and labour-intensive, particularly for large areas (Anbazhagan, Giridhar, Ganesha Raj et al., 2010; Panagiota et al., 2012a). Hence, several studies have demonstrated that Very-High-Resolution (VHR) remote sensing data can provide the building parameters be used to determine flood vulnerability assessment (Taubenböck, Post, Roth et al., 2008; Mueller et al., 2006; Mück, Taubenböck, Post et al., 2013). Nevertheless, this research is still at an early stages (Geiß and Taubenböck, 2013). To date, no such study has been conducted in Malaysia, especially on a micro-scale. Hence, this study attempted to provide a geospatial approach to flood vulnerability assessment by considering flood hazards. Table 1.1 summarises of the research problem and gaps, as well as the contribution of this study.

Table 1.1 Summarises of issues, gaps, and contributions of this research.

No.	Issues	Gaps	Contributions
1.	<ul style="list-style-type: none"> <li>• The number of rain gauge stations in the basin to characterise rainfall events, which is an important input to hydrological modelling for flood simulation, is inadequate for hazard modelling.</li> </ul>	<ul style="list-style-type: none"> <li>➤ Different types of satellite rainfall products (SRPs) have been used to address this issue but there is no local study using near-real-time SRPs to characterise extreme rainfall events</li> </ul>	<ul style="list-style-type: none"> <li>❖ Contribute to the improvement of satellite retrieval algorithms that used to characterise extreme rainfall event-induced flooding</li> </ul>
2.	<ul style="list-style-type: none"> <li>• Land use is widely used to provide information on elements-at-risk. The advent of VHR satellite imagery enables detailed land use and land cover classification, but the condition of rooftops has changed over time due to deterioration and restoration. As a result, using ordinary image features to characterise building types is relatively difficult task.</li> <li>• Using large image features lead to overfitting of classification, classification model become more complex, and requiring longer computational times.</li> </ul>	<ul style="list-style-type: none"> <li>➤ Numerous classifications of detailed land use and land cover studies have been conducted in Malaysia but none of these studies attempt to address the issue of heterogenous urban areas, particularly heterogenous rooftop of a building.</li> <li>➤ Various feature selection methods (i.e., filter, wrapper, and embedded) have been applied to object-based classification but few studies focused on examine the advantages of wrapper approaches trained by different machine learning algorithms.</li> </ul>	<ul style="list-style-type: none"> <li>❖ Contribute to the improvement of an operational framework for detailed land use and land cover classification for a highly heterogenous rooftop of building types in an urban area</li> </ul>
3.	<ul style="list-style-type: none"> <li>• Conventional method based on field measurement to obtain building vulnerability is time consuming, tedious and labour-intensive, especially for large-scale area</li> </ul>	<ul style="list-style-type: none"> <li>➤ Flood vulnerability assessment have been conducted in Malaysia but most of these studies less focus on micro-scale and none of the studies using remote sensing technology to derive building parameters to characterise building vulnerability</li> </ul>	<ul style="list-style-type: none"> <li>❖ Contribute to the enhancement of the framework for assessing building flood vulnerability using a geospatial approach. Unlike previous studies, this framework took hazard intensity into account</li> </ul>

### **1.3 Research Aim and Objectives**

The aim of the current study is to utilise a geospatial approach for flood vulnerability assessment in the Kelantan River Basin (KRB), Malaysia. The specific objectives are outlined below followed by the research questions and hypotheses:

- (a) To improve flood hazard modelling in a basin that lacks rain gauges by employing Near-Real-Time (NRT) Satellite Rainfall Products (SRPs) in the Rainfall-Runoff-Inundation (RRI) model to undertake flood inundation mapping of the 2014 event in Kota Bharu.
- (b) To improve an operational framework of detailed land use/land cover classification for the highly heterogeneous rooftops of the buildings in Kota Bharu using Very-High-Resolution (VHR) satellite imagery and LiDAR data, before overlaying this with the flood inundation map to identify flooded buildings as elements-at-risk.
- (c) To improve a micro-scale flood vulnerability assessment in Kota Bharu based on flooded buildings and using a geospatial approach.

#### **1.3.1 Research Questions**

Two questions address the first objective:

- i. How reliable is the RRI model in predicting streamflow and flood inundation map for the 2014 Kelantan flood event?
- ii. How reliable are the Near-Real-Time (NRT) satellite rainfall products for capturing extreme rainfall events for hydrological modelling and flood simulation?

Four questions address the second objective:

- i. Which classification methods (pixel-based or object-based) produce better classification accuracy?
- ii. Would the advanced classifiers (random forest, the support vector machine and decision trees) yield better classification accuracy than the traditional classifier ( $k$ -nearest neighbour)?
- iii. Is it adequate to employ detailed land use/land cover classification for highly heterogeneous rooftops of buildings using common image features?
- iv. Which feature selection algorithms produce efficient workflow in terms of classification accuracy and computation time?

Two questions address the third objective:

- i. How reliable is the geospatial approach for assessing building flood vulnerability?
- ii. Which remote sensing data-derived building parameters are the most important for determining the building flood vulnerability assessment?

### **1.3.2 Research Hypotheses**

One hypothesis was tested to address the first objective:

- i. NRT satellite rainfall products with high spatial and temporal resolution are more likely to capture extreme rainfall events for hydrological modelling and flood simulation than rain gauges are.

Four hypotheses were tested to address the second objective:

- i. Hypothesis: An object-based classification is more likely to improve the overall accuracy than a pixel-based classification.
- ii. Hypothesis: The advanced classifiers are more likely to outperform the traditional classifier.
- iii. Hypothesis: The addition of spatial features to the classification process is more likely to increase the classification accuracy.
- iv. Hypothesis: The use of feature selection in the classification process is more likely to maintain the classification accuracy and reduce the classification computational time.

One hypothesis was tested to address the third objective:

- i. The geospatial approach is likely to be useful in determining the building flood vulnerability assessment.

#### **1.4 Significance of the Study**

This study provides three outputs. The first is a framework for a large-scale assessment of flood hazards. The first breakthrough was the evaluation of the applicability of hourly satellite rainfall products using the Rainfall-Runoff-Inundation (RRI) model to facilitate a large-scale flood hazard assessment in Malaysia. The outcomes of this research could provide useful guidelines for the related agencies (e.g., the Department of Irrigation Drainage, the Department of Mineral and Geoscience, the Malaysian National Hydraulic Research Institute, National Disaster Management Agency) to use in selecting suitable SRPs, especially for flood modelling or runoff

estimation in cases where rain gauge stations are extremely scarce or unavailable during the flood season.

The second output is the technical framework for extracting elements-at-risk from the high-resolution remotely sensed and LiDAR data. The novelty of the second output focuses on improving the existing method for elements-at-risk extraction over a highly heterogeneous area by combining feature selection methods and machine learning classifiers. This framework acts as an exposure aspect to address what the exposed elements are. This technical framework can be a useful guideline for those agencies (e.g., the Department of Irrigation Drainage, the Department of Mineral and Geoscience, the Construction Industry Development Board, National Disaster Management Agency) could use to extract related elements-at-risk for specific hazard types, such as flooding.

The third output is the remote sensing-based approach for micro-scale flood building vulnerability assessment. The innovative third output focuses on improving the remote sensing-based approach for assessing flood building vulnerability by considering the hazard component using the machine learning method. In a developing country such as Malaysia, which is still in the early stages of establishing a flood risk management policy, this study can be a starting point for encouraging the implementation of flood vulnerability assessment. This approach of micro-scale decision making in flood vulnerability assessment can form a guideline for different agencies, such as the Department of Irrigation Drainage, the Department of Mineral and Geoscience, the Construction Industry Development Board, and the National Disaster Management Agency.

## **1.5 Study Area**

Kelantan is situated in the north of Peninsular Malaysia. It is surrounded by Terengganu (east), Pahang (south), Perak (west), and Thailand (north). Kelantan has 10 districts, the main one being Kota Bharu, the capital city of Kelantan state.

Kelantan has a population of approximately 1.9 million (DOS, 2019), with a population density of 124.8 persons per square kilometre, as shown in Table 1.2. However, Kota Bharu has a larger population density of 1,481.1 persons/km<sup>2</sup>. Its district size, economic opportunities and topography are the primary determinants of the population distribution in Kelantan (DID, 2011). As the state capacity city, Kota Bharu has the largest commercial centre and state government office. However, its geographical characteristics and unplanned urbanisation make Kota Bharu vulnerable to floods every year, which are mainly caused by the northeast monsoon.

Correspondingly, the study area included two parts: the Kelantan River Basin for the flood hazard modelling, and Kota Bharu town for the detailed land use/land cover classification as well as the extraction of elements-at-risk.

Table 1.2 The population of Kelantan by districts.

<b>District</b>	<b>Land (km<sup>2</sup>)</b>	<b>Population (*000)</b>	<b>Population Density (Persons/km<sup>2</sup>)</b>
Bachok	280	162.5	580.4
Kota Bharu	403	596.9	1,481.1
Machang	530	113.6	214.3
Pasir Mas	572	231.8	405.2
Pasir Puteh	425	143.1	336.7
Tanah Merah	884	149.2	168.8
Tumpat	179	187.3	1,046.4
Gua Musang	8,213	113.9	13.9
Kuala Krai	2,287	135.2	59.1
Jeli	1,325	50.8	38.3
<b>Total</b>	<b>15,099</b>	<b>1,884.3</b>	<b>124.8</b>

### 1.5.1 The Kelantan River Basin

The state of Kelantan has a total area of 15,099 km<sup>2</sup>, of which the Kelantan, Kemasin, Semarak, and Golok Rivers are the four major river basins. The Kelantan River Basin covers approximately 85% of the state area, while the other three basins cover only 15%. All the rivers' basins flow in a northerly direction, eventually into the South China Sea.

The Kelantan River Basin covers a total area of approximately 12,981 km<sup>2</sup> and consists of four main tributaries, namely the Nenggiri, Galas, Lebir, and Pergau Rivers (Figure 1.1). The Galas River is formed by a junction between the Nenggiri River and the Pergau River. The former originates in the south-west of the central mountain range (Main Range). The latter comes from the Tahan Mountains. About 95% of the catchment area is steep, mountainous and rises to a height of 2,183 m, while the remaining area is flat. The land use/land cover in the Kelantan River Basin are dominated by tropical forests (71.5%), followed by rubber (11.1%), oil palm (6.3%), urban (1.2%), paddy (1.8%), and other agriculture (6.8%). The Kelantan River Basin is characterised by a tropical monsoon climate, with an average annual rainfall of  $\geq 2500$  mm, most of which falls from November to January. The average annual temperature of the basin is about 27.5 °C (Tan et al., 2017). The Kelantan River Basin is regularly affected by monsoon flooding events during the northeast monsoon season.

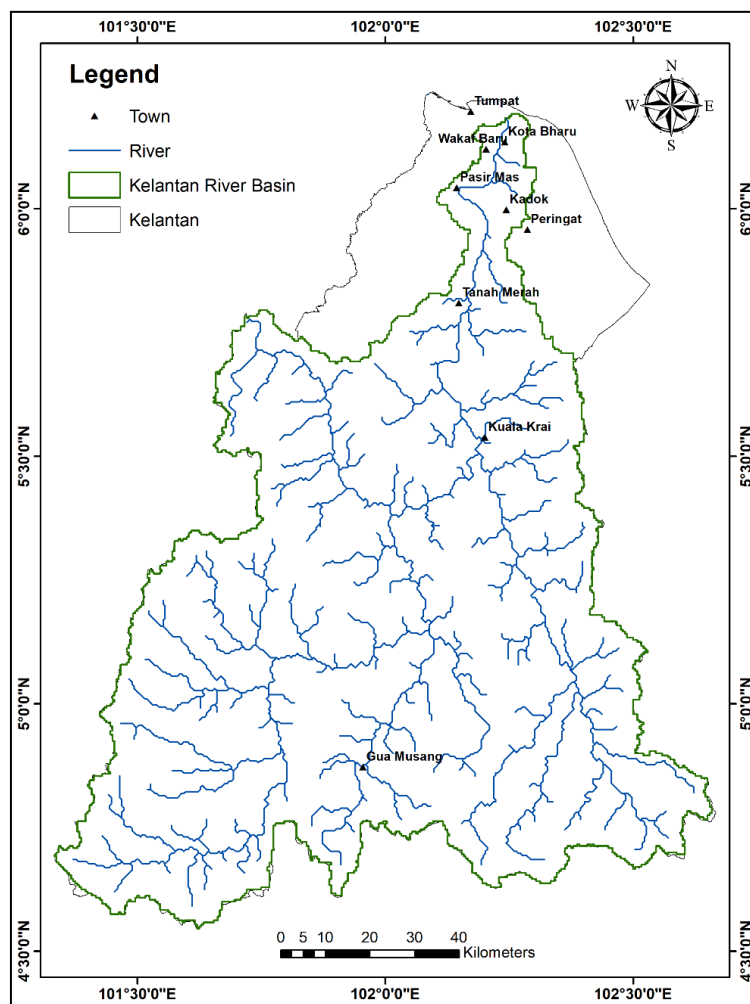


Figure 1.1 The Kelantan River Basin.



### 1.5.2 Kota Bharu Town

The growth of Kota Bharu began around the new palace, the Istana Balai Besar, which was built in 1845 by Sultan Muhammad II. Since then, Kota Bharu has been the centre of administrative and economic activity for the state of Kelantan and has acted as the capital city. Kota Bharu was dominated by the British in the 19<sup>th</sup> century; the overall landscape of Kota Bharu is therefore based on two eras of growth, with the development of urban patterns remaining unplanned until the 19<sup>th</sup> century. Thereafter, however, the urban fabric was designed according to the form of the box (Aiman Abdullah, Mohd Noor and Abdullah, 2018). Both patterns remain to this day, as seen in Figure 1.2. The only variations are in the building designs and façade.

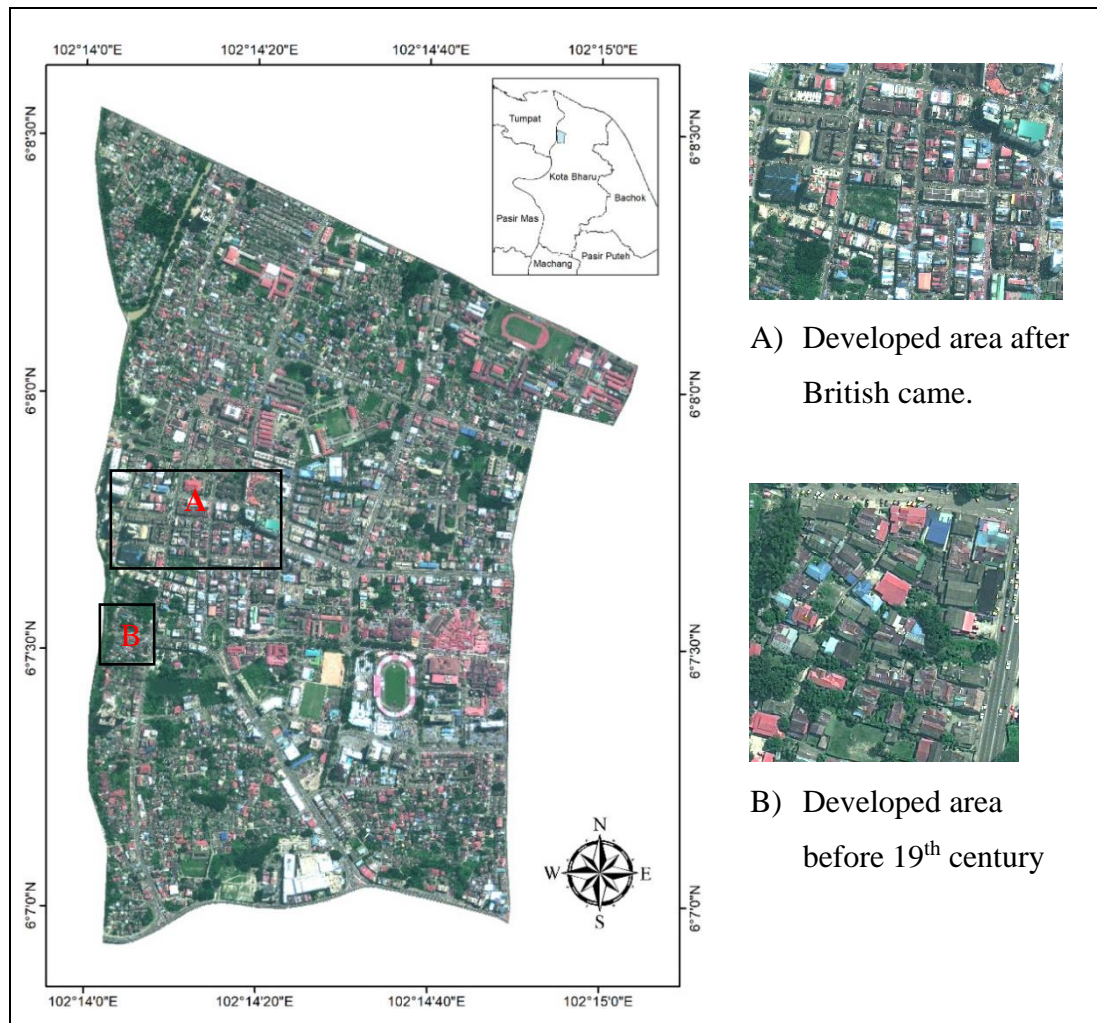


Figure 1.2 Study Area (WorldView-4 image with 0.5 m) (a) developed area after British came, (b) developed area before 19<sup>th</sup> century.

The Malay village in Peninsular Malaysia is known as the *kampung*, which traditionally comprises many wooden houses. In Kelantan, Malay villages can easily be found in Kota Bharu and other districts (Mohd Nasir, 2011). The urban design of Kota Bharu is therefore relatively unique compared to other cities in the Peninsular. The *kampung* is composed of several Malay houses and their compounds (delimited by vegetation but without physical boundaries, such as fences). Furthermore, a *kampung* does not develop according to a plan (Chen, 1998). As a result, a number of traditional houses may be in one lot of land, while different types of buildings may share a land parcel. For example, schools, dwellings, and commercial lots might coexist in a parcel of land. This landscape is therefore characterised as highly heterogeneous.

As previously mentioned, Kota Bharu is denser than the state's districts. As a result, most lands in the Kota Bharu district, particularly the Kota Bharu sub-district, has been developed. Residential areas are the dominant form of land use (1.27 km<sup>2</sup>), followed by the area allocated to infrastructure and community facilities (0.67 km<sup>2</sup>), as well as commercial activities (0.55 km<sup>2</sup>). The remaining land use classes occupy minimal portions of land. Of the types of buildings, Malay village occupy the largest area, followed by retail store, schools and government offices, with the remaining building types occupying only a small area.

## **1.6 Scope of the Study**

- i. Flood exposure and vulnerability mapping only covers the Kota Bharu area, which has a high density of elements-at-risk. To obtain the flood exposure, a December 2014 flood event simulation was simulated, while this study attempted to assess the reliability of SRPs during an emergency response, so no bias adjustment was made.
- ii. The satellite imagery was captured in 2017, while the LiDAR data was obtained in 2008 so the datasets have a nine-year time gap. However, minor changes were detected after cross-checking with Google Earth

software and using the historical imagery tool. Consequently, these changes would have impacted on the accuracy of the classification.

- iii. Due to the scarcity of building vulnerability field data, only a small number of buildings were evaluated in the building vulnerability assessment.

## **1.7 Structure of this Thesis**

This thesis consists of five chapters, with each section focusing on various aspects. Chapter 1 provides an overview of the study, which includes a brief introduction, the problem statement, the research objectives and questions, the location of the study area, as well as study's scope and significance.

The core of this thesis is set outlined in Chapter 2, which analyses the relevant studies that informed this research. This chapter initially offers an overview of the flood situations across Malaysia before defining flood risk and other terminologies. The most significant aspect of this chapter is the discussion regarding the contributions of remote sensing data to the hazard and exposure dimensions. Finally, the chapter outlines the current applications of remote sensing techniques in terms of the mapping of exposure and hazard components.

Chapter 3 provides the detailed methodology and explains how this study conducted, including the data acquisition, pre-processing, post-processing, and the results assessment method.

Chapter 4 focuses on the presentation and interpretation of the findings. In addition, an insight is provided into the findings. Lastly,

Chapter 5 provides thorough conclusion compassing all the study objectives and outlines several recommendations.

## REFERENCES

- Ab Razak, N. H., Aris, A. Z., Ramli, M. F., Looi, L. J. & Juahir, H. 2018. Temporal flood incidence forecasting for Segamat River (Malaysia) using autoregressive integrated moving average modelling. *Journal of Flood Risk Management*, 11, S794-S804.
- Abbaspour, K. C., Rouholahnejad, E., Vaghefi, S., Srinivasan, R., Yang, H. & Kløve, B. 2015. A continental-scale hydrology and water quality model for Europe: Calibration and uncertainty of a high-resolution large-scale SWAT model. *Journal of Hydrology*, 524, 733-752.
- Abdi, A. M. 2019. Land cover and land use classification performance of machine learning algorithms in a boreal landscape using Sentinel-2 data. *GIScience & Remote Sensing*, 1-20.
- Abdi, A. M. 2020. Land cover and land use classification performance of machine learning algorithms in a boreal landscape using Sentinel-2 data. *GIScience & Remote Sensing*, 57, 1-20.
- Abdikan, S. & Bayik, C. 2017. *Assessment of ALOS PALSAR 25-m mosaic data for land cover mapping*.
- Abdullah, A. F., Vojinovic, Z., Price, R. K. & Aziz, N. a. A. 2012. Improved methodology for processing raw LiDAR data to support urban flood modelling-accounting for elevated roads and bridges. *Journal of Hydroinformatics*, 14, 253-269.
- Abdullah, A. F., Vojinovic, Z. & Rahman, A. A. 2013. A methodology for processing raw LiDAR data to support urban flood modelling framework: Case study-Kuala Lumpur Malaysia. *Lecture Notes in Geoinformation and Cartography*.
- Abdullah, J., Muhammad, N. S., Julien, P. Y., Ariffin, J. & Shafie, A. 2018. Flood flow simulations and return period calculation for the Kota Tinggi watershed, Malaysia. *Journal of Flood Risk Management*, 11, S766-S782.
- Abe, B. T., Olugbara, O. O. & Marwala, T. 2014. Experimental comparison of support vector machines with random forests for hyperspectral image land cover classification. *Journal of Earth System Science*, 123, 779-790.
- Abu Alfeilat, H. A., Hassanat, A. B. A., Lasassmeh, O., Tarawneh, A. S., Alhasanat, M. B., Eyal Salman, H. S. & Prasath, V. B. S. 2019. Effects of Distance Measure Choice on K-Nearest Neighbor Classifier Performance: A Review. *Big Data*, 7, 221-248.
- Adam, E., Mutanga, O., Odindi, J. & Abdel-Rahman, E. M. 2014. Land-use/cover classification in a heterogeneous coastal landscape using RapidEye imagery: evaluating the performance of random forest and support vector machines classifiers. *International Journal of Remote Sensing*, 35, 3440-3458.
- Adiat, K. a. N., Nawawi, M. N. M. & Abdullah, K. 2012. Integration of geographic information system and 2D imaging to investigate the effects of subsurface conditions on flood occurrence. *Modern Applied Science*, 6, 11-21.
- Adler, R. F., Huffman, G. J., Chang, A., Ferraro, R., Xie, P.-P., Janowiak, J., Rudolf, B., Schneider, U., Curtis, S., Bolvin, D., Gruber, A., Susskind, J., Arkin, P. & Nelkin, E. 2003. The Version-2 Global Precipitation Climatology Project

- (GPCP) Monthly Precipitation Analysis (1979–Present). *Journal of Hydrometeorology*, 4, 1147-1167.
- Adnan, N. A. & Atkinson, P. M. 2011. Exploring the impact of climate and land use changes on streamflow trends in a monsoon catchment. *International Journal of Climatology*, 31, 815-831.
- Adnan, N. A. & Atkinson, P. M. Remote sensing of river bathymetry for use in hydraulic model prediction of flood inundation. Signal Processing and its Applications (CSPA), 2012 IEEE 8th International Colloquium on, 23-25 March 2012 2012a. 159-163.
- Adnan, R., Ruslan, F. A., Samad, A. M. & Md. Zain, Z. Flood water level modelling and prediction using artificial neural network: Case study of Sungai Batu Pahat in Johor. Proceedings - 2012 IEEE Control and System Graduate Research Colloquium, ICSGRC 2012, 2012b. 22-25.
- Adorada, A., Permatasari, R., Wirawan, P. W., Wibowo, A. & Sujiwo, A. Support Vector Machine - Recursive Feature Elimination (SVM - RFE) for Selection of MicroRNA Expression Features of Breast Cancer. 2018 2nd International Conference on Informatics and Computational Sciences (ICICoS), 30-31 Oct. 2018 2018. 1-4.
- Adrien, N. G. 2003. *Computational hydraulics and hydrology: an illustrated dictionary*, CRC Press.
- Aggarwal, A. 2016. Exposure, hazard and risk mapping during a flood event using open source geospatial technology. *Geomatics, Natural Hazards and Risk*, 7, 1426-1441.
- Aguilar, M. A., Saldaña, M. M. & Aguilar, F. J. 2013. GeoEye-1 and WorldView-2 pan-sharpened imagery for object-based classification in urban environments. *International Journal of Remote Sensing*, 34, 2583-2606.
- Ahmad, M., Zani, N. M. & Hashim, K. F. 2015. Knowledge sharing behavior among flood victims in Malaysia. *ARPN Journal of Engineering and Applied Sciences*, 10, 968-976.
- Ahmad, U. N., Shabri, A. & Zakaria, Z. A. 2011. Flood frequency analysis of annual maximum stream flows using L-Moments and TL-Moments approach. *Applied Mathematical Sciences*, 5, 243-253.
- Ai, J., Gao, W., Gao, Z., Shi, R. & Zhang, C. 2017. Phenology-based *Spartina alterniflora* mapping in coastal wetland of the Yangtze Estuary using time series of GaoFen satellite no. 1 wide field of view imagery. *Journal of Applied Remote Sensing*, 11, 026020.
- Aiman Abdullah, A. A., Mohd Noor, N. & Abdullah, A. Drone 3D mapping in identifying Malay urban form: case study of Kota Bharu. IOP Conference Series: Earth and Environmental Science, 2018. IOP Publishing, 012084.
- Akbari, A., Samah, A. A. & Daryabor, F. 2016. Raster-based derivation of a flood runoff susceptibility map using the revised runoff curve number (CN) for the Kuantan watershed, Malaysia. *Environmental Earth Sciences*, 75.
- Akbulut, Y., Sengur, A., Guo, Y. & Smarandache, F. 2017. NS-k-NN: Neutrosophic set-based k-nearest neighbors classifier. *Symmetry*, 9, 179.
- Alaghmand, S., Abdullah, R., Abustan, I., Said, M. a. M. & Vosoogh, B. 2012. Gis-based river basin flood modelling using HEC-HMS and MIKE11 - Kayu Ara river basin, Malaysia. *Journal of Environmental Hydrology*, 20, 1-16.
- Alaghmand, S., Abdullah, R. B., Abustan, I. & Vosoogh, B. 2010. GIS-based river flood hazard mapping in urban area (a case study in Kayu Ara river basin, Malaysia). *International Journal of Engineering and Technology*, 2, 488-500.

- Alaghmand, S., Bin Abdullah, R., Abustan, I. & Eslamian, S. 2012. Comparison between capabilities of HEC-RAS and MIKE11 hydraulic models in river flood risk modeling (a case study of Sungai Kayu Ara River basin, Malaysia). *International Journal of Hydrology Science and Technology*, 2, 270-291.
- Alho, P. & Aaltonen, J. 2008. Comparing a 1D hydraulic model with a 2D hydraulic model for the simulation of extreme glacial outburst floods. *Hydrological Processes*, 22, 1537-1547.
- Ali, A. N. A. & Ariffin, J. 2011. Model reliability assessment: A hydrodynamic modeling approach for flood simulation in Damansara Catchment using InfoWorks RS. *Advanced Materials Research*.
- Ali, N., Neagu, D. & Trundle, P. 2019. Evaluation of k-nearest neighbour classifier performance for heterogeneous data sets. *SN Applied Sciences*, 1, 1559.
- Aliagha, U. G., Jin, T. E., Choong, W. W., Nadzri Jaafar, M. & Ali, H. M. 2014. Factors affecting flood insurance purchase in residential properties in Johor, Malaysia. *Natural Hazards and Earth System Sciences*, 14, 3297-3310.
- Alias, N. E., Mohamad, H., Chin, W. Y. & Yusop, Z. 2016. Rainfall analysis of the Kelantan big yellow flood 2014. *Jurnal Teknologi*, 78, 83-90.
- Allen, K. 2003. Vulnerability reduction and the community-based approach. *Natural disasters and development in a globalizing world*, 170.
- Amadio, M., Mysiak, J., Carrera, L. & Koks, E. 2016. Improving flood damage assessment models in Italy. *Natural Hazards*, 82, 2075-2088.
- Amin, N. F. M. & Othman, F. 2018. Generation of flood map using infoworks for Sungai Johor. *International Journal of Integrated Engineering*, 10, 142-145.
- Aminah Shakirah, J., Sidek, L. M., Hidayah, B., Nazirul, M. Z., Jajarmizadeh, M., Ros, F. C. & Roseli, Z. A. A Review on Flood Events for Kelantan River Watershed in Malaysia for Last Decade (2001-2010). IOP Conference Series: Earth and Environmental Science, 2016.
- Amini, S., Homayouni, S., Safari, A. & Darvishsefat, A. A. 2018. Object-based classification of hyperspectral data using Random Forest algorithm. *Geospatial Information Science*, 21, 127-138.
- Anbazhagan, P., Giridhar, V., Ganesha Raj, K. & Shreedhara, V. Seismic Vulnerability Assessment Using High Resolution Satellite Data and Field Studies. 14th Symposium on Earthquake Engineering (14SEE), Department of Earthquake Engineering, IIT Roorkee, 2010. 267-276.
- Anees, M. T., Abdullah, K., Nawawi, M. N. M. & Kadir, M. O. B. A. Morphometric analysis for delineation flood hazard zone of Batang Padang catchment, Perak, Peninsular Malaysia. International Conference on Space Science and Communication, IconSpace, 2015. 231-236.
- Aonashi, K., Awaka, J., Hirose, M., Kozu, T., Kubota, T., Liu, G., Shige, S., Kida, S., Seto, S., Takahashi, N. & Takayabu, Y. N. 2009. GSMaP Passive Microwave Precipitation Retrieval Algorithm : Algorithm Description and Validation. *Journal of the Meteorological Society of Japan. Ser. II*, 87A, 119-136.
- Arnell, N. W. & Gosling, S. N. 2016. The impacts of climate change on river flood risk at the global scale. *Climatic Change*, 134, 387-401.
- Asean, A. O. S. a. N. 2019. *IWRM in Malaysia* [Online]. Available: <https://aseaniwrm.water.gov.my/iwrm-in-malaysia/> [Accessed 10 September 2019].
- Ashouri, H., Hsu, K.-L., Sorooshian, S., Braithwaite, D. K., Knapp, K. R., Cecil, L. D., Nelson, B. R. & Prat, O. P. 2015. PERSIANN-CDR: Daily precipitation

- climate data record from multisatellite observations for hydrological and climate studies. *Bulletin of the American Meteorological Society*, 96, 69-83.
- Asmara, T. a. T. & Ludin, A. N. M. Mapping perception of community preparedness towards flood in Muar River, Johor Malaysia. IOP Conference Series: Earth and Environmental Science, 2014.
- Asmara, W. a. H. W. M. & Aziz, N. H. A. SMS flood alert system. Proceedings - 2011 IEEE Control and System Graduate Research Colloquium, ICSGRC 2011, 2011. 18-22.
- Atikah, S. 2009. *Extreme flood event: a case study on floods of 2006 and 2007 in Johor*. Master thesis, Colorado State University.
- Avelar, S., Zah, R. & Tavares-Corrêa, C. 2009. Linking socioeconomic classes and land cover data in Lima, Peru: Assessment through the application of remote sensing and GIS. *International Journal of Applied Earth Observation and Geoinformation*, 11, 27-37.
- Awange, J. L., Ferreira, V. G., Forootan, E., Khandu, Andam-Akorful, S. A., Agutu, N. O. & He, X. F. 2016. Uncertainties in remotely sensed precipitation data over Africa. *International Journal of Climatology*, 36, 303-323.
- Ayob, N. H. & Sakdan, M. F. 2016. Preliminary validation on the factors influence political belief of flood victims in Malaysia. *International Review of Management and Marketing*, 6, 67-73.
- Ayog, J. L., Tongkul, F., Mirasa, A. K., Roslee, R. & Dullah, S. Flood Risk Assessment on Selected Critical Infrastructure in Kota Marudu Town, Sabah, Malaysia. MATEC Web of Conferences, 2017.
- Aziz, A. A., Barman, A., Amin, R. M., Kadarman, N., Marican, A. F., Wahab, A., Shukor, A. S. A. & Sulung, S. 2016. Coping among flood affected traditional village residents in Kuala Nerus, Terengganu. *Malaysian Journal of Public Health Medicine*, 16, 8-14.
- Azizat, N. & Wan Omar, W. M. S. Assessment of Three Flood Hazard Mapping Methods: A Case Study of Perlis. E3S Web of Conferences, 2018.
- Baatz, M. & Schäpe, A. Multiresolution Segmentation: an optimization approach for high quality multi-scale image segmentation. Proceedings of the Beiträge zum AGIT-Symposium, 2000. 12-23.
- Badyalina, B. & Shabri, A. 2015. Flood estimation at ungauged sites using group method of data handling in Peninsular Malaysia. *Jurnal Teknologi*, 76, 373-380.
- Baharum, M. S., Awang, R. A. & Baba, N. H. Flood monitoring system (MyFMS). Proceedings - 2011 IEEE International Conference on System Engineering and Technology, ICSET 2011, 2011. 204-208.
- Balaguer, A., Ruiz, L. A., Hermosilla, T. & Recio, J. A. 2010. Definition of a comprehensive set of texture semivariogram features and their evaluation for object-oriented image classification. *Computers & Geosciences*, 36, 231-240.
- Balica, S. & Wright, N. G. 2010. Reducing the complexity of the flood vulnerability index. *Environmental Hazards*, 9, 321-339.
- Balica, S. F., Popescu, I., Beevers, L. & Wright, N. G. 2013. Parametric and physically based modelling techniques for flood risk and vulnerability assessment: A comparison. *Environmental Modelling & Software*, 41, 84-92.
- Banzhaf, E. & Hofer, R. 2008. Monitoring Urban Structure Types as Spatial Indicators With CIR Aerial Photographs for a More Effective Urban Environmental Management. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 1, 129-138.

- Barasa, B. N. & Perera, E. D. P. 2018. Analysis of land use change impacts on flash flood occurrences in the Sosiani River basin Kenya. *International Journal of River Basin Management*, 16, 179-188.
- Basri, A. B., Ismail, A. F., Khairolanuar, M. H., Sobli, N. H. M., Badron, K. & Hasan, M. K. 2016. Analyses of rainfall rate during Malaysian 2014 flood event. *International Journal of Multimedia and Ubiquitous Engineering*, 11, 237-246.
- Basukala, A. K., Oldenburg, C., Schellberg, J., Sultanov, M. & Dubovyk, O. 2017. Towards improved land use mapping of irrigated croplands: performance assessment of different image classification algorithms and approaches. *European Journal of Remote Sensing*, 50, 187-201.
- Bates, P. D. 2004. Remote sensing and flood inundation modelling. *Hydrological Processes*, 18, 2593-2597.
- Bates, P. D. & De Roo, A. P. J. 2000. A simple raster-based model for flood inundation simulation. *Journal of Hydrology*, 236, 54-77.
- Bates, P. D., Horritt, M. S., Smith, C. N. & Mason, D. 1997. Integrating remote sensing observations of flood hydrology and hydraulic modelling. *Hydrological Processes*, 11, 1777-1795.
- Bayable, G., Amare, G., Alemu, G. & Gashaw, T. 2021. Spatiotemporal variability and trends of rainfall and its association with Pacific Ocean Sea surface temperature in West Harege Zone, Eastern Ethiopia. *Environmental Systems Research*, 10, 7.
- Beck, H. E., Van Dijk, A. I., Levizzani, V., Schellekens, J., Gonzalez Miralles, D., Martens, B. & De Roo, A. 2017. MSWEP: 3-hourly 0.25 global gridded precipitation (1979-2015) by merging gauge, satellite, and reanalysis data. *Hydrology and Earth System Sciences*, 21, 589-615.
- Beirami, B. A. & Mokhtarzade, M. SVM classification of hyperspectral images using the combination of spectral bands and Moran's I features. 2017 10th Iranian Conference on Machine Vision and Image Processing (MVIP), 2017. IEEE, 139-144.
- Beirami, B. A. & Mokhtarzade, M. 2020. Spatial-spectral classification of hyperspectral images based on multiple fractal-based features. *Geocarto International*, 37, 231-245.
- Belgiu, M. & Drăgut, L. 2014. Comparing supervised and unsupervised multiresolution segmentation approaches for extracting buildings from very high resolution imagery. *ISPRS Journal of Photogrammetry and Remote Sensing*, 96, 67-75.
- Belgiu, M. & Drăgut, L. 2016. Random forest in remote sensing: A review of applications and future directions. *ISPRS Journal of Photogrammetry and Remote Sensing*, 114, 24-31.
- Belousov, A. I., Verzakov, S. A. & Von Frese, J. 2002. A flexible classification approach with optimal generalisation performance: support vector machines. *Chemometrics and Intelligent Laboratory Systems*, 64, 15-25.
- Benediktsson, J. A. & Sveinsson, J. R. 1997. Feature extraction for multisource data classification with artificial neural networks. *International Journal of Remote Sensing*, 18, 727-740.
- Bengio, Y., Courville, A. & Vincent, P. 2013. Representation learning: A review and new perspectives. *IEEE transactions on pattern analysis and machine intelligence*, 35, 1798-1828.
- Beven, K. J. 2011. *Rainfall-runoff modelling: the primer*, John Wiley & Sons.



- Bhadoria, P., Agrawal, S. & Pandey, R. Image Segmentation Techniques for Remote Sensing Satellite Images. IOP Conference Series: Materials Science and Engineering, 2020. IOP Publishing, 012050.
- Bhagabati, S. S. & Kawasaki, A. 2017. Consideration of the rainfall-runoff-inundation (RRI) model for flood mapping in a deltaic area of Myanmar. *Hydrological Research Letters*, 11, 155-160.
- Bhaskaran, S., Paramananda, S. & Ramnarayan, M. 2010. Per-pixel and object-oriented classification methods for mapping urban features using Ikonos satellite data. *Applied Geography*, 30, 650-665.
- Bhuiyan, T. R., Reza, M. I. H., Choy, E. A. & Pereira, J. J. 2018. Facts and trends of urban exposure to flash flood: A case of Kuala Lumpur city. *Community, Environment and Disaster Risk Management*.
- Bidin, M. S. & Wahab, A. F. Modeling of type-2 fuzzy cubic B-spline surface for flood data problem in Malaysia. AIP Conference Proceedings, 2017.
- Billa, L. 2004. Spatial information technology in flood early warning systems: an overview of theory, application and latest developments in Malaysia. *Disaster Prevention and Management: An International Journal*, 13, 356-363.
- Billa, L., Mansor, S., Mahmud, A. & Ghazali, A. 2005. AVHRR Data for Real-Time Operational Flood Forecasting in Malaysia. In: OOSTEROM, P., ZLATANOVA, S. & FENDEL, E. (eds.) *Geo-information for Disaster Management*. Springer Berlin Heidelberg.
- Billa, L., Mansor, S. & Mahmud, A. R. 2011. Pre-flood inundation mapping for flood early warning. *Journal of Flood Risk Management*, 4, 318-327.
- Billa, L., Mansor, S., Mahmud, A. R. & Ghazali, A. H. 2006a. Modelling rainfall intensity from NOAA AVHRR data for operational flood forecasting in Malaysia. *International Journal of Remote Sensing*, 27, 5225-5234.
- Billa, L., Shattri, M., Mahmud, A. R. & Ghazali, A. H. 2006b. Comprehensive planning and the role of SDSS in flood disaster management in Malaysia. *Disaster Prevention and Management*, 15, 233-240.
- Bins, L. S. A., Fonseca, L. G., Erthal, G. J. & Ii, F. M. 1996. Satellite imagery segmentation: a region growing approach. *Simpósio Brasileiro de Sensoriamento Remoto*, 8, 677-680.
- Birkinshaw, S. J., Moore, P., Kilsby, C. G., O'donnell, G. M., Hardy, A. J. & Berry, P. a. M. 2014. Daily discharge estimation at ungauged river sites using remote sensing. *Hydrological Processes*, 28, 1043-1054.
- Bitew, M. M. & Gebremichael, M. 2011. Evaluation of satellite rainfall products through hydrologic simulation in a fully distributed hydrologic model. *Water Resources Research*, 47.
- Bitew, M. M., Gebremichael, M., Ghebremichael, L. T. & Bayissa, Y. A. 2012. Evaluation of High-Resolution Satellite Rainfall Products through Streamflow Simulation in a Hydrological Modeling of a Small Mountainous Watershed in Ethiopia. *Journal of Hydrometeorology*, 13, 338-350.
- Blanco-Vogt, Á., Haala, N. & Schanze, J. Building extraction from remote sensing data for parameterising a building typology: A contribution to flood vulnerability assessment. Joint Urban Remote Sensing Event 2013, 21-23 April 2013. 147-150.
- Blanco-Vogt, Á., Haala, N. & Schanze, J. 2015. Building parameters extraction from remote-sensing data and GIS analysis for the derivation of a building taxonomy of settlements – a contribution to flood building susceptibility assessment. *International Journal of Image and Data Fusion*, 6, 22-41.

- Blanco-Vogt, A. & Schanze, J. 2014. Assessment of the physical flood susceptibility of buildings on a large scale—conceptual and methodological frameworks. *Natural Hazards and Earth System Sciences*, 14, 2105-2117.
- Blaschke, T. 2010. Object based image analysis for remote sensing. *ISPRS Journal of Photogrammetry and Remote Sensing*, 65, 2-16.
- Blaschke, T., Lang, S., Lorup, E., Strobl, J. & Zeil, P. 2000. Object-oriented image processing in an integrated GIS/remote sensing environment and perspectives for environmental applications. In: CREMERS, A. & GERVE, K. (eds.) *Environmental information for planning, politics and the public*. Metropolis-Verlag, Marburg.
- Bohachevsky, I. O., Johnson, M. E. & Stein, M. L. 1986. Generalized Simulated Annealing for Function Optimization. *Technometrics*, 28, 209-217.
- Bolón-Canedo, V., Sánchez-Marroño, N. & Alonso-Betanzos, A. 2013. A review of feature selection methods on synthetic data. *Knowledge and Information Systems*, 34, 483-519.
- Bonnifait, L., Delrieu, G., Lay, M. L., Boudevillain, B., Masson, A., Belleudy, P., Gaume, E. & Saulnier, G.-M. 2009. Distributed hydrologic and hydraulic modelling with radar rainfall input: Reconstruction of the 8–9 September 2002 catastrophic flood event in the Gard region, France. *Advances in Water Resources*, 32, 1077-1089.
- Bonyongo, M. C., Mubyana, T., Totolo, O. & Veenandaal, E. M. 2002. Flooding and soil nutrient status in the Okavango Delta's seasonal floodplains. *Botswana Notes and Records*, 34, 123-130.
- Borden, K. A., Schmidlein, M. C., Emrich, C. T., Piegorsch, W. W. & Cutter, S. L. 2007. Vulnerability of US cities to environmental hazards. *Journal of Homeland Security and Emergency Management*, 4.
- Boren, L., Mao, P. & Zixing, W. An improved segmentation of high spatial resolution remote sensing image using Marker-based Watershed Algorithm. 2012 20th International Conference on Geoinformatics, 15-17 June 2012. 1-5.
- Borfecchia, F., De Cecco, L., Pollino, M., La Porta, L., Lugari, A., Martini, S., Ristatore, E. & Pascale, C. 2010. Active and passive remote sensing for supporting the evaluation of the urban seismic vulnerability. *Italian Journal of Remote Sensing*, 42, 129-141.
- Bormudo, A. & Nagai, M. 2017. Perception of risk and coping capacity: A study in Jiadhal Basin, India. *International Journal of Disaster Risk Reduction*, 21, 376-383.
- Bravo, J. M., Allasia, D., Paz, A. R., Collischonn, W. & Tucci, C. E. M. 2012. Coupled Hydrologic-Hydraulic Modeling of the Upper Paraguay River Basin. *Journal of Hydrologic Engineering*, 17, 635-646.
- Breiman, L. 2001. Random forests. *Machine learning*, 45, 5-32.
- Breiman, L., Friedman, J. H., Olshen, R. A. & Stone, C. J. 1984. *Classification and regression trees*, London, Chapman & Hall/CRC.
- Brocca, L., Pellarin, T., Crow, W. T., Ciabatta, L., Massari, C., Ryu, D., Su, C.-H., Rüdiger, C. & Kerr, Y. 2016. Rainfall estimation by inverting SMOS soil moisture estimates: A comparison of different methods over Australia. *Journal of Geophysical Research: Atmospheres*, 121, 12,062-12,079.
- Budiyono, Y., Aerts, J., Brinkman, J., Marfai, M. & Ward, P. 2014. Flood risk assessment for delta mega-cities: a case study of Jakarta. *Natural Hazards*, 75, 389-413.

- Bui, H. T., Ishidaira, H. & Shaowei, N. 2019. Evaluation of the use of global satellite–gauge and satellite-only precipitation products in stream flow simulations. *Applied Water Science*, 9, 53.
- Bukari, S. M., Ahmad, M. A., Wai, T. L., Kaamin, M. & Alimin, N. Spatial Analysis in Determination of Flood Prone Areas Using Geographic Information System and Analytical Hierarchy Process at Sungai Sembrong's Catchment. IOP Conference Series: Materials Science and Engineering, 2016.
- Buslima, F. S., Omar, R. C., Jamaluddin, T. A. & Taha, H. 2018. Flood and flash flood geo-hazards in Malaysia. *International Journal of Engineering and Technology(UAE)*, 7, 760-764.
- Buytaert, W., Celleri, R., Willems, P., Bièvre, B. D. & Wyseure, G. 2006. Spatial and temporal rainfall variability in mountainous areas: A case study from the south Ecuadorian Andes. *Journal of Hydrology*, 329, 413-421.
- Cai, G., Ren, H., Yang, L., Zhang, N., Du, M. & Wu, C. 2019. Detailed Urban Land Use Land Cover Classification at the Metropolitan Scale Using a Three-Layer Classification Scheme. *Sensors*, 19, 3120.
- Cai, L., Shi, W., He, P., Miao, Z., Hao, M. & Zhang, H. 2015. Fusion of multiple features to produce a segmentation algorithm for remote sensing images. *Remote Sensing Letters*, 6, 390-398.
- Cammerer, H., Thieken, A. H. & Verburg, P. H. 2013. Spatio-temporal dynamics in the flood exposure due to land use changes in the Alpine Lech Valley in Tyrol (Austria). *Natural Hazards*, 68, 1243-1270.
- Cánovas-García, F. & Alonso-Sarría, F. 2015. Optimal Combination of Classification Algorithms and Feature Ranking Methods for Object-Based Classification of Submeter Resolution Z/I-Imaging DMC Imagery. *Remote Sensing*, 7, 4651-4677.
- Carleer, A. P., Debeir, O. & Wolff, E. 2005. Assessment of Very High Spatial Resolution Satellite Image Segmentations. *Photogrammetric Engineering & Remote Sensing*, 71, 1285-1294.
- Cattani, E., Merino, A. & Levizzani, V. 2016. Evaluation of Monthly Satellite-Derived Precipitation Products over East Africa. *Journal of Hydrometeorology*, 17, 2555-2573.
- Ceos, C. O. E. O. S. 2003. The Use of Earth Observing Satellites for Hazard Support: Assessments and Scenarios. Final Report of the CEOS Disaster Management Support Group (DMSG).
- Ceos, C. O. E. O. S. 2015. Satellite Earth Observations in Support of Disaster Risk Reduction - Special 2015 WCDRR Edition.
- Chan, J. C.-W. & Paelinckx, D. 2008. Evaluation of Random Forest and Adaboost tree-based ensemble classification and spectral band selection for ecotope mapping using airborne hyperspectral imagery. *Remote Sensing of Environment*, 112, 2999-3011.
- Chan, N. 2015a. Impacts of Disasters and Disaster Risk Management in Malaysia: The Case of Floods. In: ALDRICH, D. P., OUM, S. & SAWADA, Y. (eds.) *Resilience and Recovery in Asian Disasters*. Springer Japan.
- Chan, N. W. 1995. Flood disaster management in Malaysia: an evaluation of the effectiveness of government resettlement schemes. *Disaster Prevention and Management*, 4, 22-29.
- Chan, N. W. 1997a. Increasing flood risk in Malaysia: causes and solutions. *Disaster Prevention and Management: An International Journal*, 6, 72-86.

- Chan, N. W. 1997b. Institutional arrangements for flood hazard management in Malaysia: An evaluation using the criteria approach. *Disasters*, 21, 206-222.
- Chan, N. W. 2015b. Impacts of Disasters and Disaster Risk Management in Malaysia: The Case of Floods. In: ALDRICH, D., OUM, S. & SAWADA, Y. (eds.) *Resilience and Recovery in Asian Disasters*. Japan: Springer
- Chan, N. W., Mahamud, K. R. B. K. & Karim, M. Z. A. 2016. Assessing different types of flood losses in Kelantan state in Malaysia during the December 2014 flood. *1st International Conference on Society, Space & Environment 2016*. Bali, Indonesia.
- Chan, N. W. & Parker, D. J. 1996. Response to dynamic flood hazard factors in peninsular Malaysia. *Geographical Journal*, 162, 313-325.
- Chandrashekar, G. & Sahin, F. 2014. A survey on feature selection methods. *Computers & Electrical Engineering*, 40, 16-28.
- Chen, J., Du, P., Wu, C., Xia, J. & Chanussot, J. 2018a. Mapping Urban Land Cover of a Large Area Using Multiple Sensors Multiple Features. *Remote Sensing*, 10, 872.
- Chen, L. & Cheng, X. 2016a. Classification of High-resolution Remotely Sensed Images Based on Random Forests. *J. Softw. Eng*, 10, 318-327.
- Chen, S., Hong, Y., Cao, Q., Kirstetter, P.-E., Gourley, J. J., Qi, Y., Zhang, J., Howard, K., Hu, J. & Wang, J. 2013. Performance evaluation of radar and satellite rainfalls for Typhoon Morakot over Taiwan: Are remote-sensing products ready for gauge denial scenario of extreme events? *Journal of Hydrology*, 506, 4-13.
- Chen, S., Hu, J., Zhang, A., Min, C., Huang, C. & Liang, Z. 2019. Performance of near real-time Global Satellite Mapping of Precipitation estimates during heavy precipitation events over northern China. *Theoretical and Applied Climatology*, 135, 877-891.
- Chen, T. & Guestrin, C. Xgboost: A scalable tree boosting system. KDD '16: Proceedings of the 22nd ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, 2016b. 785-794.
- Chen, V. F. 1998. The Encyclopaedia of Malaysia: Architecture. Archipelago Press.
- Chen, Y., Ge, Y., Heuvelink, G. B. M., An, R. & Chen, Y. 2018b. Object-Based Superresolution Land-Cover Mapping From Remotely Sensed Imagery. *IEEE Transactions on Geoscience and Remote Sensing*, 56, 328-340.
- Chen, Y., Su, W., Li, J. & Sun, Z. 2009. Hierarchical object oriented classification using very high resolution imagery and LIDAR data over urban areas. *Advances in Space Research*, 43, 1101-1110.
- Chen, Y. R., Yeh, C. H. & Yu, B. 2016c. Flood damage assessment of an urban area in Taiwan. *Natural Hazards*, 83, 1045-1055.
- Chen, Z., Wang, G. & Liu, J. 2012. A modified object-oriented classification algorithm and its application in high-resolution remote-sensing imagery. *International Journal of Remote Sensing*, 33, 3048-3062.
- Chen, Z., Zhao, Z., Gong, P. & Zeng, B. 2006. A new process for the segmentation of high resolution remote sensing imagery. *International Journal of Remote Sensing*, 27, 4991-5001.
- Cheng, G. & Han, J. 2016. A survey on object detection in optical remote sensing images. *ISPRS Journal of Photogrammetry and Remote Sensing*, 117, 11-28.
- Chia, C. 2004. Managing flood problems in Malaysia. *Bulletin Ingenieur*.
- Chong, T., Leong, C., Pan, S. L., Bahri, S. & Khan, A. F. A. Use of social media in disaster relief during the Kuantan (Malaysia) flood. 35th International

- Conference on Information Systems "Building a Better World Through Information Systems", ICIS 2014, 2014.
- Chow, M. F. & Jamil, M. M. 2017. Review of development and applications of Integrated Flood Analysis System (IFAS) for flood forecasting in insufficiently-gauged catchments. *Journal of Engineering and Applied Sciences*, 12, 9210-9215.
- Chu, C., Hong, L., Liu, C. & Chen, J. A new regional shape index for classification of high resolution remote sensing images. 2014 Third International Workshop on Earth Observation and Remote Sensing Applications (EORSA), 11-14 June 2014. 156-160.
- Cleve, C., Kelly, M., Kearns, F. R. & Moritz, M. 2008. Classification of the wildland–urban interface: A comparison of pixel- and object-based classifications using high-resolution aerial photography. *Computers, Environment and Urban Systems*, 32, 317-326.
- Colkesen, I. & Kavzoglu, T. 2018. Selection of Optimal Object Features in Object-Based Image Analysis Using Filter-Based Algorithms. *Journal of the Indian Society of Remote Sensing*, 46, 1233-1242.
- Colkesen, I. & Kavzoglu, T. 2019. Comparative Evaluation of Decision-Forest Algorithms in Object-Based Land Use and Land Cover Mapping. In: POURGHASEMI, H. R. & GOKCEOGLU, C. (eds.) *Spatial Modeling in GIS and R for Earth and Environmental Sciences*. Elsevier.
- Congalton, R. G. 1991. A review of assessing the accuracy of classifications of remotely sensed data. *Remote Sensing of Environment*, 37, 35-46.
- Congalton, R. G. & Green, K. 2019. *Assessing the accuracy of remotely sensed data: principles and practices*, CRC press.
- Cortes, C. & Vapnik, V. 1995. Support-vector networks. *Machine Learning*, 20, 273-297.
- Cover, T. & Hart, P. 1967. Nearest neighbor pattern classification. *IEEE transactions on information theory*, 13, 21-27.
- Cream 2021. Flood Risk Assessment and Flood Vulnerability Index for Critical Infrastructure (CI) in Malaysia.
- Croke, B. F. W., Islam, A., Ghosh, J. & Khan, M. A. 2011. Evaluation of approaches for estimation of rainfall and the unit hydrograph. *Hydrology Research*, 42, 372-385.
- Crow, W. T., Van Den Berg, M. J., Huffman, G. J. & Pellarin, T. 2011. Correcting rainfall using satellite-based surface soil moisture retrievals: The Soil Moisture Analysis Rainfall Tool (SMART). *Water Resources Research*, 47.
- Custer, R. & Nishijima, K. 2015. Flood vulnerability assessment of residential buildings by explicit damage process modelling. *Natural Hazards*, 78, 461-496.
- Cutler, D. R., Edwards Jr., T. C., Beard, K. H., Cutler, A., Hess, K. T., Gibson, J. & Lawler, J. J. 2007. Random Forests for Classification In Ecology. *Ecology*, 88, 2783-2792.
- Cutter, S. L., Boruff, B. J. & Shirley, W. L. 2003. Social Vulnerability to Environmental Hazards\*. *Social Science Quarterly*, 84, 242-261.
- Dai, Q., Bray, M., Zhuo, L., Islam, T. & Han, D. 2017. A Scheme for Rain Gauge Network Design Based on Remotely Sensed Rainfall Measurements. *Journal of Hydrometeorology*, 18, 363-379.

- Damm, M., Fekete, A. & Bogardi, J. J. Intersectoral vulnerability indices as tools for framing risk mitigation measures and spatial planning. Conference Proc. Hydro Predict, 2010.
- Damodaran, B. B., Höhle, J. & Lefèvre, S. 2017. Attribute Profiles on Derived Features for Urban Land Cover Classification. *Photogrammetric Engineering & Remote Sensing*, 83, 183-193.
- De Brito, M. M., Evers, M. & Höllermann, B. 2017. Prioritization of flood vulnerability, coping capacity and exposure indicators through the Delphi technique: A case study in Taquari-Antas basin, Brazil. *International Journal of Disaster Risk Reduction*, 24, 119-128.
- De Groeve, T. 2010. Flood monitoring and mapping using passive microwave remote sensing in Namibia. *Geomatics, Natural Hazards and Risk*, 1, 19-35.
- De Jong, S. M. & Burrough, P. A. 1995. A fractal approach to the classification of Mediterranean vegetation types in remotely sensed images. *Photogrammetric Engineering and Remote Sensing*, 61, 1041-1053.
- De Jong, S. M., Pebesma, E. J. & Meer, F. D. 2004. Spatial variability, mapping methods, image analysis and pixels. *Remote sensing image analysis: Including the spatial domain*. Springer.
- De Moel, H. & Aerts, J. C. J. H. 2011a. Effect of uncertainty in land use, damage models and inundation depth on flood damage estimates. *Natural Hazards*, 58, 407-425.
- De Moel, H., Aerts, J. C. J. H. & Koomen, E. 2011b. Development of flood exposure in the Netherlands during the 20th and 21st century. *Global Environmental Change*, 21, 620-627.
- De Moel, H., Jongman, B., Kreibich, H., Merz, B., Penning-Rowsell, E. & Ward, P. J. 2015. Flood risk assessments at different spatial scales. *Mitigation and Adaptation Strategies for Global Change*, 1-26.
- De Pinho, C. M. D., Fonseca, L. M. G., Korting, T. S., De Almeida, C. M. & Kux, H. J. H. 2012. Land-cover classification of an intra-urban environment using high-resolution images and object-based image analysis. *International Journal of Remote Sensing*, 33, 5973-5995.
- De Ruiter, M. C., Ward, P. J., Daniell, J. E. & Aerts, J. C. J. H. 2017. Review Article: A comparison of flood and earthquake vulnerability assessment indicators. *Nat. Hazards Earth Syst. Sci.*, 17, 1231-1251.
- Debba, P., Van Ruitenbeek, F. J. A., Van Der Meer, F. D., Carranza, E. J. M. & Stein, A. 2005. Optimal field sampling for targeting minerals using hyperspectral data. *Remote Sensing of Environment*, 99, 373-386.
- Defries, R. S. & Chan, J. C.-W. 2000. Multiple Criteria for Evaluating Machine Learning Algorithms for Land Cover Classification from Satellite Data. *Remote Sensing of Environment*, 74, 503-515.
- Dehvari, A. & Heck, R. 2007. Comparison of object-based and pixel based infrared airborne image classification methods using DEM thematic layer. *Journal of Geography and Regional Planning*, 2, 086-096.
- Deichmann, U., Ehrlich, D., Small, C. & Zeug, G. 2011. Using high resolution satellite data for the identification of urban natural disaster risk.
- Denis, G., De Boissezon, H., Hosford, S., Pasco, X., Montfort, B. & Ranera, F. 2016. The evolution of Earth Observation satellites in Europe and its impact on the performance of emergency response services. *Acta Astronautica*, 127, 619-633.

- Díaz-Uriarte, R. & Alvarez De Andrés, S. 2006. Gene selection and classification of microarray data using random forest. *BMC Bioinformatics*, 7, 3.
- Dickson, E., Baker, J. L., Hoornweg, D. & Asmita, T. 2012. Urban risk assessments: an approach for understanding disaster and climate risk in cities. Washington, D.C.: The World Bank.
- Did 2015. Generation of Flood Hazard for Sungai Linggi Basin, in the State of Melaka.
- Did, D. O. I. a. D. 2009. Volume 1 - Flood Management. Kuala Lumpur.
- Did, D. O. I. a. D. 2011. Review of The National Water Resources Study (2000-2050) and Formulation of National Water Resources Policy. Final Report, Volume 10 - Kelantan. Kuala Lumpur, Malaysia.
- Did, D. O. I. a. D. 2013. Peta Hazard Banjir. Department of Irrigation and Drainage.
- Dile, Y. T. & Srinivasan, R. 2014. Evaluation of CFSR climate data for hydrologic prediction in data-scarce watersheds: an application in the Blue Nile River Basin. *JAWRA Journal of the American Water Resources Association*, 50, 1226-1241.
- Dinku, T., Ruiz, F., Connor, S. J. & Ceccato, P. 2010. Validation and Intercomparison of Satellite Rainfall Estimates over Colombia. *Journal of Applied Meteorology and Climatology*, 49, 1004-1014.
- Do-Tu, H. & Installé, M. A fast clustering procedure based on isodata algorithm with application to remote sensing. Proceedings of the 4th International Joint Conference on Pattern Recognition, 1978.
- Dos, D. O. S. 2019. Poket Stats Negeri Kelantan ST3 2019.
- Douben, N. & Ratnayake, R. 2006. Characteristic data on river floods and flooding; facts and figures. In: VAN ALPHEN, J., VAN BEEK, E. & TAAL, M. (eds.) *Floods, from Defence to Management*. London: Taylor & Francis.
- Douville, H., Chauvin, F. & Broqua, H. 2001. Influence of Soil Moisture on the Asian and African Monsoons. Part I: Mean Monsoon and Daily Precipitation. *Journal of Climate*, 14, 2381-2403.
- Drăguț, L., Csillik, O., Eisank, C. & Tiede, D. 2014. Automated parameterisation for multi-scale image segmentation on multiple layers. *ISPRS Journal of Photogrammetry and Remote Sensing*, 88, 119-127.
- Drăguț, L., Tiede, D. & Levick, S. R. 2010. ESP: a tool to estimate scale parameter for multiresolution image segmentation of remotely sensed data AU - Drăguț, Lucian. *International Journal of Geographical Information Science*, 24, 859-871.
- Duncan, J. M. A. & Biggs, E. M. 2012. Assessing the accuracy and applied use of satellite-derived precipitation estimates over Nepal. *Applied Geography*, 34, 626-638.
- Duro, D. C., Franklin, S. E. & Dubé, M. G. 2012a. A comparison of pixel-based and object-based image analysis with selected machine learning algorithms for the classification of agricultural landscapes using SPOT-5 HRG imagery. *Remote Sensing of Environment*, 118, 259-272.
- Duro, D. C., Franklin, S. E. & Dubé, M. G. 2012b. Multi-scale object-based image analysis and feature selection of multi-sensor earth observation imagery using random forests. *International Journal of Remote Sensing*, 33, 4502-4526.
- Dutsenwai, H. S., Bin Ahmad, B., Mijinyawa, A. & Tanko, A. I. 2016. Fusion of SAR images for flood extent mapping in northern peninsula Malaysia. *INTERNATIONAL JOURNAL OF ADVANCED AND APPLIED SCIENCES*, 3, 37-48.

- Dwarakish, G. S., Devia, G. K., Ganasri, B. P. & Dwarakish, G. S. 2015. A Review on Hydrological Models. *Aquatic Procedia*, 4, 1001-1007.
- Ebert, A., Kerle, N. & Stein, A. 2009. Urban social vulnerability assessment with physical proxies and spatial metrics derived from air- and spaceborne imagery and GIS data. *Natural Hazards*, 48, 275-294.
- Echeverribar, I., Morales-Hernández, M., Brufau, P. & García-Navarro, P. 2019. Use of internal boundary conditions for levees representation: application to river flood management. *Environmental Fluid Mechanics*, 19, 1253-1271.
- Eckert, S., Jelinek, R., Zeug, G. & Krausmann, E. 2012. Remote sensing-based assessment of tsunami vulnerability and risk in Alexandria, Egypt. *Applied Geography*, 32, 714-723.
- Ecognition Developer, T. 2014. 9.0 User Guide. *Trimble Germany GmbH: Munich, Germany*.
- Ehrlich, D., Kemper, T., Blaes, X. & Soille, P. 2013. Extracting building stock information from optical satellite imagery for mapping earthquake exposure and its vulnerability. *Natural Hazards*, 68, 79-95.
- El Merabet, Y., Meurie, C., Ruichek, Y., Sbihi, A. & Touahni, R. 2015. Building Roof Segmentation from Aerial Images Using a Line and Region-Based Watershed Segmentation Technique. *Sensors*, 15, 3172-3203.
- Elboshy, B., Kanae, S., Gamaleldin, M., Ayad, H., Osaragi, T. & Elbarki, W. 2018. A framework for pluvial flood risk assessment in Alexandria considering the coping capacity. *Environment Systems and Decisions*.
- Elsheikh, R. F. A., Ouerghi, S. & Elhag, A. R. 2015. Flood risk map based on GIS, and multi criteria techniques (case study Terengganu Malaysia). *Journal of Geographic Information System*, 7, 348.
- Em-Dat 2021. [www.emdat.be](http://www.emdat.be): CRED / UCLouvain, Brussels, Belgium
- Emerson, C. W., Lam, N. S.-N. & Quattrochi, D. A. 1999. Multi-Scale Fractal Analysis of Image Texture and Pattern. *Photogrammetric Engineering and Remote Sensing*, 65, 51-61.
- Emerson, C. W., Lam, N. S. N. & Quattrochi, D. A. 2005. A comparison of local variance, fractal dimension, and Moran's I as aids to multispectral image classification. *International Journal of Remote Sensing*, 26, 1575-1588.
- Espindola, G. M., Camara, G., Reis, I. A., Bins, L. S. & Monteiro, A. M. 2006. Parameter selection for region - growing image segmentation algorithms using spatial autocorrelation. *International Journal of Remote Sensing*, 27, 3035-3040.
- Estoque, R. C., Murayama, Y. & Akiyama, C. M. 2015. Pixel-based and object-based classifications using high- and medium-spatial-resolution imageries in the urban and suburban landscapes. *Geocarto International*, 30, 1113-1129.
- Ettinger, S., Mounaud, L., Magill, C., Yao-Lafourcade, A.-F., Thouret, J.-C., Manville, V., Negulescu, C., Zuccaro, G., De Gregorio, D., Nardone, S., Uchuchoque, J. a. L., Arguedas, A., Macedo, L. & Manrique Llerena, N. 2016. Building vulnerability to hydro-geomorphic hazards: Estimating damage probability from qualitative vulnerability assessment using logistic regression. *Journal of Hydrology*, 541, 563-581.
- Fischer, G., Nachtergaele, F., Prieler, S., Van Velthuisen, H., Verelst, L. & Wiberg, D. 2008. Global agro-ecological zones assessment for agriculture (GAEZ 2008). *IIASA, Laxenburg, Austria and FAO, Rome, Italy*.



- Fix, E. & Hodges, J. L. 1951. Discriminatory Analysis, Nonparametric Discrimination: Consistency Properties. *Technical Report 4*. USAF school of Aviation Medicine.
- Fizri, F. F. A., Rahim, A. A., Sibly, S., Koshy, K. C. & Nor, N. M. 2014. Strengthening the capacity of flood-affected rural communities in Padang Terap, State of Kedah, Malaysia. *Sustainable Living with Environmental Risks*.
- Foody, G. M. & Mathur, A. 2004. Toward intelligent training of supervised image classifications: directing training data acquisition for SVM classification. *Remote Sensing of Environment*, 93, 107-117.
- Foudi, S., Osés-Eraso, N. & Tamayo, I. 2015. Integrated spatial flood risk assessment: The case of Zaragoza. *Land Use Policy*, 42, 278-292.
- Fread, D. L. 1993. Flow routing. In: MAIDMENT, D. R. (ed.) *Handbook of hydrology*. New York: McGraw-Hill Professional.
- Friedl, M. A. & Brodley, C. E. 1997. Decision tree classification of land cover from remotely sensed data. *Remote Sensing of Environment*, 61, 399-409.
- Fugura, A. A., Billa, L., Pradhan, B., Mohamed, T. A. & Rawashdeh, S. 2011. Coupling of hydrodynamic modeling and aerial photogrammetry-derived digital surface model for flood simulation scenarios using GIS: Kuala Lumpur flood, Malaysia. *Disaster Advances*, 4, 20-28.
- Fukuda, S., Yasunaga, E., Nagle, M., Yuge, K., Sardud, V., Spreer, W. & Müller, J. 2014. Modelling the relationship between peel colour and the quality of fresh mango fruit using Random Forests. *Journal of Food Engineering*, 131, 7-17.
- Gain, A. K., Mojtahed, V., Biscaro, C., Balbi, S. & Giupponi, C. 2015. An integrated approach of flood risk assessment in the eastern part of Dhaka City. *Natural Hazards*, 79, 1499-1530.
- Gao, J. 2009. *Digital analysis of remotely sensed imagery*, McGraw-Hill Education.
- Gao, Y., Mas, J. F., Kerle, N. & Navarrete Pacheco, J. A. 2011. Optimal region growing segmentation and its effect on classification accuracy. *International Journal of Remote Sensing*, 32, 3747-3763.
- Gaona, M. F. R., Overeem, A., Brasjen, A. M., Meirink, J. F., Leijnse, H. & Uijlenhoet, R. 2017. Evaluation of Rainfall Products Derived From Satellites and Microwave Links for The Netherlands. *IEEE Transactions on Geoscience and Remote Sensing*, 55, 6849-6859.
- Gasim, M. B., Mokhtar, M., Surif, S., Toriman, M. E., Rahim, S. A. & Lun, P. I. 2012. Analysis of thirty years recurrent floods of the Pahang River, Malaysia. *Asian Journal of Earth Sciences*, 5, 25-35.
- Gasim, M. B., Surif, S., Mokhtar, M., Toriman, M. E. H., Rahim, S. A. & Bee, C. H. 2010. Flood analysis of December 2006: Focus at Segamat town, Johor. *Sains Malaysiana*, 39, 353-361.
- Gasim, M. B., Toriman, M. E. & Abdullahi, M. G. 2014. Floods in Malaysia historical reviews, causes, effects and mitigations approach. *International Journal of Interdisciplinary Research and Innovations*, 2, 59-65.
- Gebere, S. B., Alamirew, T., Merkel, B. J. & Melesse, A. M. 2015. Performance of High Resolution Satellite Rainfall Products over Data Scarce Parts of Eastern Ethiopia. *Remote Sensing*, 7, 11639-11663.
- Geiß, C., Klotz, M. & Taubenböck, H. Remote sensing for seismic building vulnerability assessment. Second European Conference on Earthquake Engineering and Seismology Aug 25-29 2014 Istanbul.
- Geiß, C. & Taubenböck, H. 2013. Remote sensing contributing to assess earthquake risk: from a literature review towards a roadmap. *Natural Hazards*, 68, 7-48.

- Geiß, C., Taubenböck, H., Tyagunov, S., Tisch, A., Post, J. & Lakes, T. 2014. Assessment of Seismic Building Vulnerability from Space. *Earthquake Spectra*, 30, 1553-1583.
- Georganos, S., Grippa, T., Vanhuysse, S., Lennert, M., Shimoni, M., Kalogirou, S. & Wolff, E. 2017a. Less is more: optimizing classification performance through feature selection in a very-high-resolution remote sensing object-based urban application. *GIScience & Remote Sensing*, 55, 221-242.
- Georganos, S., Grippa, T., Vanhuysse, S., Lennert, M., Shimoni, M. & Wolff, E. 2017b. *Optimizing classification performance in an object-based very-high-resolution land use-land cover urban application*, SPIE.
- Georganos, S., Grippa, T., Vanhuysse, S., Lennert, M., Shimoni, M. & Wolff, E. 2018. Very High Resolution Object-Based Land Use–Land Cover Urban Classification Using Extreme Gradient Boosting. *IEEE Geoscience and Remote Sensing Letters*, 15, 607-611.
- Gerl, T., Bochow, M. & Kreibich, H. 2014. Flood Damage Modeling on the Basis of Urban Structure Mapping Using High-Resolution Remote Sensing Data. *Water*, 6, 2367.
- Gerl, T., Kreibich, H., Franco, G., Marechal, D. & Schröter, K. 2016. A Review of Flood Loss Models as Basis for Harmonization and Benchmarking. *PLoS ONE*, 11, e0159791.
- Ghani, A. A., Chang, C. K., Leow, C. S. & Zakaria, N. A. 2012. Sungai Pahang digital flood mapping: 2007 flood. *International Journal of River Basin Management*, 10, 139-148.
- Ghani, N. Z. C., Hasan, Z. A. & Liang, L. T. 2014. Estimation of missing rainfall data using GEP: case study of raja river, Alor Setar, Kedah. *Advances in Artificial Intelligence*, 2014, 6.
- Ghazali, J. N. & Kamsin, A. A Real Time Simulation of Flood Hazard. 2008 Fifth International Conference on Computer Graphics, Imaging and Visualisation, 26-28 Aug. 2008. 393-397.
- Ghorbani, K., Wayayok, A. & Abdullah, A. F. 2016. Simulation of flood risk area in Kelantan watershed, Malaysia using numerical model. *Jurnal Teknologi*, 78, 51-57.
- Ghosh, A. & Joshi, P. K. 2014. A comparison of selected classification algorithms for mapping bamboo patches in lower Gangetic plains using very high resolution WorldView 2 imagery. *International Journal of Applied Earth Observation and Geoinformation*, 26, 298-311.
- Gibril, M. B. A., Idrees, M. O., Shafri, H. Z. M. & Yao, K. 2018. Integrative image segmentation optimization and machine learning approach for high quality land-use and land-cover mapping using multisource remote sensing data. *Journal of Applied Remote Sensing*, 12, 1-16, 16.
- Gibril, M. B. A., Shafri, H. Z. M. & Hamedianfar, A. 2017. New semi-automated mapping of asbestos cement roofs using rule-based object-based image analysis and Taguchi optimization technique from WorldView-2 images. *International Journal of Remote Sensing*, 38, 467-491.
- Gillespie, T. W., Chu, J., Frankenberg, E. & Thomas, D. 2007. Assessment and Prediction of Natural Hazards from Satellite Imagery. *Progress in physical geography*, 31, 459-470.
- Gissing, A. & Blong, R. 2004. Accounting for variability in commercial flood damage estimation. *Australian Geographer*, 35, 209-222.

- Glas, H., Deruyter, G., De Maeyer, P., Mandal, A. & James-Williamson, S. 2016. Analyzing the sensitivity of a flood risk assessment model towards its input data. *Nat. Hazards Earth Syst. Sci.*, 16, 2529-2542.
- Gleason, C. J. & Smith, L. C. 2014. Toward global mapping of river discharge using satellite images and at-many-stations hydraulic geometry. *Proceedings of the National Academy of Sciences*.
- Goodin, D. G., Anibas, K. L. & Bezymennyi, M. 2015. Mapping land cover and land use from object-based classification: an example from a complex agricultural landscape. *International Journal of Remote Sensing*, 36, 4702-4723.
- Gosain, A. K., Rao, S. & Basuray, D. 2006. Climate change impact assessment on hydrology of Indian river basins. *Current Science*, 90, 346-353.
- Gosset, M., Viarre, J., Quantin, G. & Alcoba, M. 2013. Evaluation of several rainfall products used for hydrological applications over West Africa using two high-resolution gauge networks. *Quarterly Journal of the Royal Meteorological Society*, 139, 923-940.
- Green, C. H. & Penning-Rowsell, E. C. 1989. Flooding and the Quantification of 'Intangibles'. *Water and Environment Journal*, 3, 27-30.
- Green, W. H. & Ampt, G. A. 1911. Studies on soil physics. *Journal of Agricultural Science*, 4, 1-24.
- Grimaldi, S., Petroselli, A., Arcangeletti, E. & Nardi, F. 2013. Flood mapping in ungauged basins using fully continuous hydrologic-hydraulic modeling. *Journal of Hydrology*, 487, 39-47.
- Grippa, T., Georganos, S., Zarougui, S., Bognounou, P., Diboulo, E., Forget, Y., Lennert, M., Vanhuysse, S., Mboga, N. & Wolff, E. 2018. Mapping Urban Land Use at Street Block Level Using OpenStreetMap, Remote Sensing Data, and Spatial Metrics. *ISPRS International Journal of Geo-Information*, 7, 246.
- Grünthal, G. & Levret, A. 1998. European macroseismic scale 1998 (EMS-98) cahiers du centre Européen de géodynamique et de séismologie 15. *Centre Européen de géodynamique et de séismologie, Luxembourg*.
- Grybas, H., Melendy, L. & Congalton, R. G. 2017. A comparison of unsupervised segmentation parameter optimization approaches using moderate- and high-resolution imagery. *GIScience & Remote Sensing*, 54, 515-533.
- Guan, H., Li, J., Chapman, M., Deng, F., Ji, Z. & Yang, X. 2013. Integration of orthoimagery and lidar data for object-based urban thematic mapping using random forests. *International Journal of Remote Sensing*, 34, 5166-5186.
- Guha-Sapir, D., Below, R. & Hoyois, P. 2015. EM-DAT: International disaster database. *Catholic University of Louvain: Brussels, Belgium*, 27, 57-58.
- Guyon, I., Weston, J., Barnhill, S. & Vapnik, V. 2002. Gene Selection for Cancer Classification using Support Vector Machines. *Machine Learning*, 46, 389-422.
- Gxumisa, A. & Breytenbach, A. 2017. Evaluating pixel vs. segmentation based classifiers with height differentiation on SPOT 6 imagery for urban land cover mapping. *South African Journal of Geomatics*, 6, 436-448.
- Habib, E., Haile, A. T., Sazib, N., Zhang, Y. & Rientjes, T. 2014. Effect of Bias Correction of Satellite-Rainfall Estimates on Runoff Simulations at the Source of the Upper Blue Nile. *Remote Sensing*, 6, 6688-6708.
- Hadi, L., Naim, W., Adnan, N., Nisa, A. & Said, E. 2017a. GIS based multi-criteria decision making for flood vulnerability index assessment. *Journal of Telecommunication, Electronic and Computer Engineering (JTEC)*, 9, 7-11.

- Hadi, L. A., Naim, W. M., Adnan, N. A., Nisa, A. & Said, E. S. 2017b. GIS based multi-criteria decision making for flood vulnerability index assessment. *Journal of Telecommunication, Electronic and Computer Engineering*, 9, 7-11.
- Hafiz, I., Nor, M. D., Sidek, L. M., Basri, H., Fukami, K., Hanapi, M. N. & Livia, L. Flood forecasting and early warning system for Dungun River Basin. IOP Conference Series: Earth and Environmental Science, 2013a.
- Hafiz, I., Nor, N. D. M., Sidek, L. M., Basri, H., Hanapi, M. N. & Livia, L. Application of Integrated Flood Analysis System (IFAS) for Dungun river basin. IOP Conference Series: Earth and Environmental Science, 2013b.
- Hafiz, I., Sidek, L. M., Basri, H., Fukami, K., Hanapi, M. N., Livia, L. & Jaafar, A. S. Integrated flood analysis system (IFAS) for Kelantan river basin. ISTT 2014 - 2014 IEEE 2nd International Symposium on Telecommunication Technologies, 2015. 159-162.
- Hair Zaki, U. H., Ibrahim, R., Abd Halim, S. & Yokoi, T. 2017. A review on service oriented architecture approach in flood disaster management framework for sentiment analysis: Malaysia context. *Frontiers in Artificial Intelligence and Applications*.
- Halim, M. K. A., Ahmad, A., Rahman, M. Z. A., Amin, Z. M., Khanan, M. F. A., Musliman, I. A., Kadir, W. H. W., Jamal, M. H., Maimunah, D. S., Wahab, A. K. A., Zabidi, M. M. A., Suaib, N. M. & Zain, R. M. Land use/land cover mapping for conservation of UNESCO Global Geopark using object and pixel-based approaches. IOP Conference Series: Earth and Environmental Science, 2018.
- Hall-Beyer, M. 2017. Practical guidelines for choosing GLCM textures to use in landscape classification tasks over a range of moderate spatial scales. *International Journal of Remote Sensing*, 38, 1312-1338.
- Hall, M. A. 1999. Correlation-based feature selection for machine learning.
- Hall, M. A. & Holmes, G. 2003. Benchmarking feature selection techniques for discrete class data mining. *IEEE Transactions on Data Engineering*, 15, 1-16.
- Hall, M. A. & Smith, L. A. Feature subset selection: a correlation based filter approach. International Conference on Neural Information Processing and Intelligent Information Systems, 1997. 855-858.
- Hamedianfar, A., Shafri, H. Z., Mansor, S. & Ahmad, N. 2014a. Combining data mining algorithm and object-based image analysis for detailed urban mapping of hyperspectral images. *Journal of Applied Remote Sensing*, 8, 085091.
- Hamedianfar, A. & Shafri, H. Z. M. 2014b. Development of fuzzy rule-based parameters for urban object-oriented classification using very high resolution imagery. *Geocarto International*, 29, 268-292.
- Hamedianfar, A. & Shafri, H. Z. M. 2015. Detailed intra-urban mapping through transferable OBIA rule sets using WorldView-2 very-high-resolution satellite images. *International Journal of Remote Sensing*, 36, 3380-3396.
- Hamedianfar, A. & Shafri, H. Z. M. 2016. Integrated approach using data mining-based decision tree and object-based image analysis for high-resolution urban mapping of WorldView-2 satellite sensor data. *Journal of Applied Remote Sensing*, 10, 025001.
- Hamedianfar, A., Shafri, H. Z. M., Mansor, S. & Ahmad, N. 2014c. Improving detailed rule-based feature extraction of urban areas from WorldView-2 image and lidar data. *International Journal of Remote Sensing*, 35, 1876-1899.

- Han, N., Wu, J., Tahmassebi, A. R. S., Xu, H.-W. & Wang, K. 2011. NDVI-Based Lacunarity Texture for Improving Identification of *Torreya* Using Object-Oriented Method. *Agricultural Sciences in China*, 10, 1431-1444.
- Han, Y., Kim, H., Choi, J. & Kim, Y. 2012. A shape-size index extraction for classification of high resolution multispectral satellite images. *International Journal of Remote Sensing*, 33, 1682-1700.
- Harb, M., De Vecchi, D. & Dell'acqua, F. 2015. Physical Vulnerability Proxies from Remotes Sensing: Reviewing, Implementing and Disseminating Selected Techniques. *IEEE Geoscience and Remote Sensing Magazine*, 3, 20-33.
- Hardin, P. J. & Thomson, C. N. 1992. Fast nearest neighbor classification methods for multispectral imagery. *The Professional Geographer*, 44, 191-202.
- Hashemi, H., Nordin, M., Lakshmi, V., Huffman, G. J. & Knight, R. 2017. Bias Correction of Long-Term Satellite Monthly Precipitation Product (TRMM 3B43) over the Conterminous United States. *Journal of Hydrometeorology*, 18, 2491-2509.
- Hashim, N., Hamid, J. R. A., Saraf, N. M., Naharudin, N., Halim, M. A. & Razali, M. H. Spectral Information Extraction from Worldview-2 Image for Urban Features Identification. 2019 IEEE 10th Control and System Graduate Research Colloquium (ICSGRC), 2-3 Aug. 2019 2019. 76-81.
- Hassan, A., Ghani, A. A. & Abdullah, R. Development of flood risk map using GIS for Sg. Selangor Basin. Proceeding of the Sixth International Conference on ASIA GIS, 9-10 Mar 2006, UTM Skudai, Johor, Malaysia, 2006.
- He, X., Wu, Y. & Wu, Y. 2013. Texture feature extraction method combining nonsubsampling contour transformation with gray level co-occurrence matrix. *Journal of multimedia*, 8, 675.
- Heiden, U., Heldens, W., Roessner, S., Segl, K., Esch, T. & Mueller, A. 2012. Urban structure type characterization using hyperspectral remote sensing and height information. *Landscape and Urban Planning*, 105, 361-375.
- Herath, S., Dutta, D. & Musiaka, K. Flood damage estimation of an urban catchment using remote sensing and GIS. Proc. the Eighth International Conference on Urban Storm Drainage, 1999.
- Herold, M., Liu, X. & Clarke, K. C. 2003. Spatial metrics and image texture for mapping urban land use. *Photogrammetric Engineering & Remote Sensing*, 69, 991-1001.
- Herold, M. & Roberts, D. A. 2010. The spectral dimension in urban remote sensing. *Remote sensing of urban and suburban areas*. Springer.
- Hirpa, F. A., Gebremichael, M. & Hopson, T. 2010. Evaluation of High-Resolution Satellite Precipitation Products over Very Complex Terrain in Ethiopia. *Journal of Applied Meteorology and Climatology*, 49, 1044-1051.
- Hirpa, F. A., Hopson, T. M., De Groeve, T., Brakenridge, G. R., Gebremichael, M. & Restrepo, P. J. 2013. Upstream satellite remote sensing for river discharge forecasting: Application to major rivers in South Asia. *Remote Sensing of Environment*, 131, 140-151.
- Ho, J. C. 2009. *Coastal flood risk assessment and coastal zone management: A case study of Seberang Perai and Kuantan Pekan in Malaysia*. Msc Thesis, University of Southampton.
- Hofer, T. & Messerli, B. 1997. Floods in Bangladesh; Process Understanding and Development Strategies. Institute of Geography, Univ. of Beren, Switzerland.
- Hong, Y., Adler, R. F., Huffman, G. J. & Pierce, H. 2010. Applications of TRMM-Based Multi-Satellite Precipitation Estimation for Global Runoff Prediction:

- Prototyping a Global Flood Modeling System. *In: GEBREMICHAEL, M. & HOSSAIN, F. (eds.) Satellite Rainfall Applications for Surface Hydrology.* Dordrecht: Springer Netherlands.
- Hong, Y., Gochis, D., Cheng, J.-T., Hsu, K.-L. & Sorooshian, S. 2007. Evaluation of PERSIANN-CCS Rainfall Measurement Using the NAME Event Rain Gauge Network. *Journal of Hydrometeorology*, 8, 469-482.
- Hong, Y., Hsu, K.-L., Sorooshian, S. & Gao, X. 2004. Precipitation Estimation from Remotely Sensed Imagery Using an Artificial Neural Network Cloud Classification System. *Journal of Applied Meteorology*, 43, 1834-1853.
- Hooli, L. J. 2016. Resilience of the poorest: coping strategies and indigenous knowledge of living with the floods in Northern Namibia. *Regional Environmental Change*, 16, 695-707.
- Hoque, R., Nakayama, D., Matsuyama, H. & Matsumoto, J. 2011. Flood monitoring, mapping and assessing capabilities using RADARSAT remote sensing, GIS and ground data for Bangladesh. *Natural Hazards*, 57, 525-548.
- Hossain, M. D. & Chen, D. 2019. Segmentation for Object-Based Image Analysis (OBIA): A review of algorithms and challenges from remote sensing perspective. *ISPRS Journal of Photogrammetry and Remote Sensing*, 150, 115-134.
- Hou, A. Y., Kakar, R. K., Neeck, S., Azarbarzin, A. A., Kummerow, C. D., Kojima, M., Oki, R., Nakamura, K. & Iguchi, T. 2014. The Global Precipitation Measurement Mission. *Bulletin of the American Meteorological Society*, 95, 701-722.
- Hsu, K.-L., Behrangi, A., Imam, B. & Sorooshian, S. 2010. Extreme Precipitation Estimation Using Satellite-Based PERSIANN-CCS Algorithm. *In: GEBREMICHAEL, M. & HOSSAIN, F. (eds.) Satellite Rainfall Applications for Surface Hydrology.* Dordrecht: Springer Netherlands.
- Huan, L. & Setiono, R. Chi2: feature selection and discretization of numeric attributes. Proceedings of 7th IEEE International Conference on Tools with Artificial Intelligence, 5-8 Nov. 1995 1995. 388-391.
- Huang, C., Davis, L. S. & Townshend, J. R. G. 2002. An assessment of support vector machines for land cover classification. *International Journal of Remote Sensing*, 23, 725-749.
- Huang, M. J., Shyue, S. W., Lee, L. H. & Kao, C. C. 2008. A knowledge-based approach to urban feature classification using aerial imagery with lidar data. *Photogrammetric Engineering and Remote Sensing*, 74, 1473-1485.
- Huang, S., Tang, L., Hupy, J. P., Wang, Y. & Shao, G. 2021. A commentary review on the use of normalized difference vegetation index (NDVI) in the era of popular remote sensing. *Journal of Forestry Research*, 32, 1-6.
- Huang, X., Weng, C., Lu, Q., Feng, T. & Zhang, L. 2015. Automatic Labelling and Selection of Training Samples for High-Resolution Remote Sensing Image Classification over Urban Areas. *Remote Sensing*, 7, 16024-16044.
- Huffman, G. J. & Bolvin, D. T. 2013. Version 1.2 GPCP one-degree daily precipitation data set documentation. NASA, Goddard Space Flight Center, Greenbelt, MD, USA <https://rda.ucar.edu/datasets/ds728>, 3.
- Huffman, G. J., Bolvin, D. T., Braithwaite, D., Hsu, K., Joyce, R., Xie, P. & Yoo, S.-H. 2015. NASA global precipitation measurement (GPM) integrated multi-satellite retrievals for GPM (IMERG). *Algorithm theoretical basis document, version.*

- Huffman, G. J., Bolvin, D. T., Nelkin, E. J., Wolff, D. B., Adler, R. F., Gu, G., Hong, Y., Bowman, K. P. & Stocker, E. F. 2007. The TRMM Multisatellite Precipitation Analysis (TMPA): Quasi-Global, Multiyear, Combined-Sensor Precipitation Estimates at Fine Scales. *Journal of Hydrometeorology*, 8, 38-55.
- Hughes, G. 1968. On the mean accuracy of statistical pattern recognizers. *IEEE transactions on information theory*, 14, 55-63.
- Huong, H. T. L. & Pathirana, A. 2013. Urbanization and climate change impacts on future urban flooding in Can Tho city, Vietnam. *Hydrol. Earth Syst. Sci.*, 17, 379-394.
- Hussain, E. & Shan, J. 2016. Urban building extraction through object-based image classification assisted by digital surface model and zoning map. *International Journal of Image and Data Fusion*, 7, 63-82.
- Hussain, T. P. R. S., Nor, A. R. M. & Ismail, H. 2014. The level of satisfaction towards flood management system in Kelantan, Malaysia. *Pertanika Journal of Social Science and Humanities*, 22, 257-269.
- Huth, J., Kuenzer, C., Wehrmann, T., Gebhardt, S., Tuan, V. Q. & Dech, S. 2012. Land Cover and Land Use Classification with TWOPAC: towards Automated Processing for Pixel- and Object-Based Image Classification. *Remote Sensing*, 4, 2530-2553.
- Huyck, C., Verrucci, E. & Bevington, J. 2014. Chapter 1 - Remote Sensing for Disaster Response: A Rapid, Image-Based Perspective. In: SHRODER, J. F. & WYSS, M. (eds.) *Earthquake Hazard, Risk and Disasters*. Boston: Academic Press.
- Ibrahim, I. & Asmawi, M. Z. Multicriteria analysis for flood mapping of Sungai Pahang. Proceedings - 39th Asian Conference on Remote Sensing: Remote Sensing Enabling Prosperity, ACRS 2018, 2018. 267-277.
- Ince, F. 1987. Maximum likelihood classification, optimal or problematic? A comparison with the nearest neighbour classification. *International Journal of Remote Sensing*, 8, 1829-1838.
- Inglada, J. 2007. Automatic recognition of man-made objects in high resolution optical remote sensing images by SVM classification of geometric image features. *ISPRS Journal of Photogrammetry and Remote Sensing*, 62, 236-248.
- Ippcc 2012. Managing the risks of extreme events and disasters to advance climate change adaptation: special report of the intergovernmental panel on climate change. In: FIELD, C. B., BARROS, V., STOCKER, T. F., QIN, D., DOKKEN, D. J., EBI, K. L., MASTRANDREA, M. D., MACH, K. J., PLATTNER, G.-K., ALLEN, S. K., TIGNOR, M. & MIDGLEY, P. M. (eds.) *A Special Report of Working Groups I and II of the Intergovernmental Panel on Climate Change* Cambridge, United Kingdom and New York, NY, USA: Cambridge University Press.
- Isahak, A., Reza, M. I. H., Siwar, C., Ismail, S. M., Sulaiman, N., Hanafi, Z., Zainuddin, M. S. & Taha, M. R. 2018. Delineating risk zones and evaluation of shelter centres for flood disaster management along the Pahang River Basin, Malaysia. *Jamba: Journal of Disaster Risk Studies*, 10.
- Ito, A. 2005. Issues in the implementation of the International Charter on Space and Major Disasters. *Space Policy*, 21, 141-149.
- Iwami, Y., Hasegawa, A., Miyamoto, M., Kudo, S., Yamazaki, Y., Ushiyama, T. & Koike, T. 2017. Comparative study on climate change impact on precipitation and floods in Asian river basins. *Hydrological Research Letters*, 11, 24-30.

- Jain, A. K., Duin, R. P. W. & Mao, J. 2000. Statistical pattern recognition: a review. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 22, 4-37.
- Jajarmizadeh, M., Mohd Sidek, L., Basri, H. B. & Jaffar, A. S. Flood Forecasting via Time Lag Forward Network; Kelantan, Malaysia. IOP Conference Series: Earth and Environmental Science, 2016.
- Jakomulska, A. & Clarke, K. 2001. Variogram-derived measures of textural image classification. *geoENV III—Geostatistics for Environmental Applications*. Springer.
- Janecek, A., Gansterer, W., Demel, M. & Ecker, G. On the relationship between feature selection and classification accuracy. New challenges for feature selection in data mining and knowledge discovery, 2008. 90-105.
- Jebur, M. N., Mohd Shafri, H. Z., Pradhan, B. & Tehrany, M. S. 2014. Per-pixel and object-oriented classification methods for mapping urban land cover extraction using SPOT 5 imagery. *Geocarto International*, 29, 792-806.
- Jensen, J. R. 2015. *Introductory digital image processing: a remote sensing perspective*, Liverpool, UK, Prentice Hall Press.
- Jia, J., Yang, N., Zhang, C., Yue, A., Yang, J. & Zhu, D. 2013. Object-oriented feature selection of high spatial resolution images using an improved Relief algorithm. *Mathematical and Computer Modelling*, 58, 619-626.
- Jia, K., Liu, J., Tu, Y., Li, Q., Sun, Z., Wei, X., Yao, Y. & Zhang, X. 2018. Land use and land cover classification using Chinese GF-2 multispectral data in a region of the North China Plain. *Frontiers of Earth Science*.
- Jiao, L., Liu, Y. & Li, H. 2012. Characterizing land-use classes in remote sensing imagery by shape metrics. *ISPRS Journal of Photogrammetry and Remote Sensing*, 72, 46-55.
- Jobin, B., Labrecque, S., Grenier, M. & Falardeau, G. 2008. Object-Based Classification as an Alternative Approach to the Traditional Pixel-Based Classification to Identify Potential Habitat of the Grasshopper Sparrow. *Environmental Management*, 41, 20-31.
- Johansen, K., Coops, N. C., Gergel, S. E. & Stange, Y. 2007. Application of high spatial resolution satellite imagery for riparian and forest ecosystem classification. *Remote Sensing of Environment*, 110, 29-44.
- Johnson, B., Bragais, M., Endo, I., Magcale-Macandog, D. & Macandog, P. 2015. Image Segmentation Parameter Optimization Considering Within- and Between-Segment Heterogeneity at Multiple Scale Levels: Test Case for Mapping Residential Areas Using Landsat Imagery. *ISPRS International Journal of Geo-Information*, 4, 2292.
- Johnson, B. & Xie, Z. 2013. Classifying a high resolution image of an urban area using super-object information. *ISPRS Journal of Photogrammetry and Remote Sensing*, 83, 40-49.
- Johnson, M. S., Coon, W. F., Mehta, V. K., Steenhuis, T. S., Brooks, E. S. & Boll, J. 2003. Application of two hydrologic models with different runoff mechanisms to a hillslope dominated watershed in the northeastern US: a comparison of HSPF and SMR. *Journal of Hydrology*, 284, 57-76.
- Jongman, B., Koks, E. E., Husby, T. G. & Ward, P. J. 2014. Increasing flood exposure in the Netherlands: implications for risk financing *Natural Hazards and Earth System Sciences*, 14, 1429-1429.
- Jongman, B., Kreibich, H., Apel, H., Barredo, J. I., Bates, P. D., Feyen, L., Gericke, A., Neal, J., Aerts, J. C. J. H. & Ward, P. J. 2012a. Comparative flood damage



- model assessment: towards a European approach. *Nat. Hazards Earth Syst. Sci.*, 12, 3733-3752.
- Jongman, B., Ward, P. J. & Aerts, J. C. J. H. 2012b. Global exposure to river and coastal flooding: Long term trends and changes. *Global Environmental Change*, 22, 823-835.
- Jonkman, S. N., Bockarjova, M., Kok, M. & Bernardini, P. 2008. Integrated hydrodynamic and economic modelling of flood damage in the Netherlands. *Ecological Economics*, 66, 77-90.
- Jović, A., Brkić, K. & Bogunović, N. A review of feature selection methods with applications. 2015 38th International Convention on Information and Communication Technology, Electronics and Microelectronics (MIPRO), 25-29 May 2015 2015. 1200-1205.
- Joyce, R. J., Janowiak, J. E., Arkin, P. A. & Xie, P. 2004. CMORPH: A Method that Produces Global Precipitation Estimates from Passive Microwave and Infrared Data at High Spatial and Temporal Resolution. *Journal of Hydrometeorology*, 5, 487-503.
- Julien, P., Ghani, A., Zakaria, N., Abdullah, R. & Chang, C. 2010. Case Study: Flood Mitigation of the Muda River, Malaysia. *Journal of Hydraulic Engineering*, 136, 251-261.
- Kabir, S., He, D. C., Sanusi, M. A. & Wan Hussina, W. M. A. 2010. Texture analysis of IKONOS satellite imagery for urban land use and land cover classification. *The Imaging Science Journal*, 58, 163-170.
- Kai, L. & Muller, J.-P. Segmenting satellite imagery: a region growing scheme. Remote sensing: global monitoring for earth management, Espoo, June 3-6, 1991, 1990.
- Kakhani, N., Mokhtarzade, M. & Valadan Zouj, M. J. 2019. Classification of very high-resolution remote sensing images by applying a new edge-based marker-controlled watershed segmentation method. *Signal, Image and Video Processing*, 13, 1319-1327.
- Kamaludin, M., Azlina, A. A., Padli, J. & Alipiah, R. M. 2018. Assessing public's preferences and willingness to pay for flood prevention program in Kuala Krai, Malaysia. *Disaster Advances*, 11, 1-10.
- Kamin, M., Ahmad, N. F. A., Razali, S. N. M., Hilaham, M. M., Rahman, M. A., Ngadiman, N. & Sahat, S. 2017. Flood prediction using geographical information system (GIS) application at sungai sembrong. *Journal of Telecommunication, Electronic and Computer Engineering*, 9, 113-117.
- Karegowda, A. G., Jayaram, M. & Manjunath, A. 2010. Feature subset selection problem using wrapper approach in supervised learning. *International journal of Computer applications*, 1, 13-17.
- Kaszta, Ž., Van De Kerchove, R., Ramoelo, A., Cho, M. A., Madonsela, S., Mathieu, R. & Wolff, E. 2016. Seasonal Separation of African Savanna Components Using Worldview-2 Imagery: A Comparison of Pixel- and Object-Based Approaches and Selected Classification Algorithms. *Remote Sensing*, 8, 763.
- Katuk, N., Ku-Mahamud, K. R., Norwawi, N. & Deris, S. 2009. Web-based support system for flood response operation in Malaysia. *Disaster Prevention and Management*, 18, 327-337.
- Katuk, N., Mahamud, K., Ruhana, K., Norwawi, N. & Deris, S. Web-Based Support System for Flood Response Operation. 2006 IEEE/WIC/ACM International Conference on Web Intelligence and Intelligent Agent Technology Workshops, 18-22 Dec. 2006 2006. 169-171.

- Kaur, B. & Garg, A. Mathematical morphological edge detection for remote sensing images. 2011 3rd International Conference on Electronics Computer Technology, 8-10 April 2011 2011. 324-327.
- Kavzoglu, T. 2017. Object-Oriented Random Forest for High Resolution Land Cover Mapping Using Quickbird-2 Imagery. In: SAMUI, P., SEKHAR, S. & BALAS, V. E. (eds.) *Handbook of Neural Computation*. Academic Press.
- Kavzoglu, T. & Colkesen, I. 2009. A kernel functions analysis for support vector machines for land cover classification. *International Journal of Applied Earth Observation and Geoinformation*, 11, 352-359.
- Kavzoglu, T., Colkesen, I. & Yomralioglu, T. 2015. Object-based classification with rotation forest ensemble learning algorithm using very-high-resolution WorldView-2 image. *Remote Sensing Letters*, 6, 834-843.
- Keshtkar, H., Voigt, W. & Alizadeh, E. 2017. Land-cover classification and analysis of change using machine-learning classifiers and multi-temporal remote sensing imagery. *Arabian Journal of Geosciences*, 10, 154.
- Khan, M. M. A., Shaari, N. A., Bahar, A. M. A. & Baten, M. A. 2014a. Impact of the flood occurrence in Kota Bharu, Kelantan using statistical analysis. *Journal of Applied Sciences*, 14, 1944-1951.
- Khan, M. M. A., Shaari, N. a. B., Bahar, A. M. A., Baten, M. A. & Nazaruddin, D. a. B. 2014b. Flood impact assessment in Kota Bharu, Malaysia: a statistical analysis. *World Applied Sciences Journal*, 32, 626-634.
- Khosravi, I., Momeni, M. & Rahnemoonfar, M. 2014. Performance Evaluation of Object-based and Pixel-based Building Detection Algorithms from Very High Spatial Resolution Imagery. *Photogrammetric Engineering & Remote Sensing*, 80, 519-528.
- Kia, M., Pirasteh, S., Pradhan, B., Mahmud, A., Sulaiman, W. & Moradi, A. 2012. An artificial neural network model for flood simulation using GIS: Johor River Basin, Malaysia. *Environmental Earth Sciences*, 67, 251-264.
- Kiala, Z., Mutanga, O., Odindi, J. & Peerbhay, K. 2019. Feature Selection on Sentinel-2 Multispectral Imagery for Mapping a Landscape Infested by Parthenium Weed. *Remote Sensing*, 11, 1892.
- Kidd, C., Bauer, P., Turk, J., Huffman, G. J., Joyce, R., Hsu, K.-L. & Braithwaite, D. 2012. Intercomparison of High-Resolution Precipitation Products over Northwest Europe. *Journal of Hydrometeorology*, 13, 67-83.
- Kienberger, S., Lang, S. & Zeil, P. 2009. Spatial vulnerability units – expert-based spatial modelling of socio-economic vulnerability in the Salzach catchment, Austria. *Nat. Hazards Earth Syst. Sci.*, 9, 767-778.
- Kim, H., Lakes, T., Kleinschmit, B. & Kenneweg, H. Different approaches for urban habitat type mapping–The case study of Berlin and Seoul. 3rd International symposium on remote sensing and data fusion over urban areas 14-16 March 2005 AZ, USA: Tempe.
- Kim, M., Warner, T. A., Madden, M. & Atkinson, D. S. 2011. Multi-scale GEOBIA with very high spatial resolution digital aerial imagery: scale, texture and image objects. *International Journal of Remote Sensing*, 32, 2825-2850.
- Kimani, M., Hoedjes, J. & Su, Z. 2018. Bayesian Bias Correction of Satellite Rainfall Estimates for Climate Studies. *Remote sensing*, 10, 1074.
- Kirkpatrick, S., Gelatt, C. D. & Vecchi, M. P. 1983. Optimization by simulated annealing. *science*, 220, 671-680.
- Kitada, K. & Fukuyama, K. 2012. Land-Use and Land-Cover Mapping Using a Gradable Classification Method. *Remote Sensing*, 4, 1544-1558.

- Knorn, J., Rabe, A., Radeloff, V. C., Kuemmerle, T., Kozak, J. & Hostert, P. 2009. Land cover mapping of large areas using chain classification of neighboring Landsat satellite images. *Remote Sensing of Environment*, 113, 957-964.
- Kohavi, R. & John, G. H. 1997. Wrappers for feature subset selection. *Artificial Intelligence*, 97, 273-324.
- Komolafe, A. A., Herath, S. & Avtar, R. 2018. Sensitivity of flood damage estimation to spatial resolution. *Journal of Flood Risk Management*, 11, S370-S381.
- Krakauer, N. Y., Pradhanang, S. M., Lakhankar, T. & Jha, A. K. 2013. Evaluating Satellite Products for Precipitation Estimation in Mountain Regions: A Case Study for Nepal. *Remote Sensing*, 5, 4107-4123.
- Kreibich, H., Thielen, A. H., Petrow, T., Muller, M. & Merz, B. 2005. Flood loss reduction of private households due to building precautionary measures - lessons learned from the Elbe flood in August 2002. *Natural Hazards and Earth System Sciences*, 5, 117-126.
- Kron, A. 2007. Flood Damage Estimation and Flood Risk Mapping. In: A . VASSILOPOULOS , R. A., C . ZEVENBERGEN , E . PASCHE , AND S . GARVIN (ed.) *Advances in Urban Flood Management*. London, UK: Taylor & Francis.
- Krzysztofowicz, R. & Davis, D. R. 1983. Category - unit loss functions for flood forecast - response system evaluation. *Water resources research*, 19, 1476-1480.
- Kta Tenaga Sdn Bhd 2003. Flood Damage Assessment of 26 April 2001 Flooding Affecting the Klang Valley and The Generalised Procedures and Guidelines for Assessment of Flood Damages. Volume 2 - Guidelines and Procedures for the Assessment of Flood Damages in Malaysia.
- Ku-Mahamud, K. R., Zakaria, N., Katuk, N. & Shbier, M. Flood pattern detection using sliding window technique. Proceedings - 2009 3rd Asia International Conference on Modelling and Simulation, AMS 2009, 2009. 45-50.
- Kubota, T., Shige, S., Hashizume, H., Aonashi, K., Takahashi, N., Seto, S., Hirose, M., Takayabu, Y. N., Ushio, T., Nakagawa, K., Iwanami, K., Kachi, M. & Okamoto, K. 2007. Global Precipitation Map Using Satellite-Borne Microwave Radiometers by the GSMaP Project: Production and Validation. *IEEE Transactions on Geoscience and Remote Sensing*, 45, 2259-2275.
- Kuffer, M., Barros, J. & Sliuzas, R. V. 2014. The development of a morphological unplanned settlement index using very-high-resolution (VHR) imagery. *Computers, Environment and Urban Systems*, 48, 138-152.
- Kuhn, M. & Johnson, K. 2013. *Applied Predictive Modeling*, New York, Springer.
- Kuhn, M. & Johnson, K. 2019. *Feature engineering and selection: A practical approach for predictive models*, CRC Press.
- Kuhn, M., Wing, J., Weston, S., Williams, A., Keefer, C., Engelhardt, A., Cooper, T., Mayer, Z., Kenkel, B. & Team, R. C. 2020. Package 'caret'. *The R Journal*, 223, 7.
- Kumar, T. G., Murugan, D. & Manish, T. I. 2018. An Analysis on Road Extraction from Satellite Image Using Otsu Method and Genetic Algorithm Techniques. *WSEAS TRANSACTIONS on COMPUTERS*, 17.
- Kundzewicz, Z. W., Kanae, S., Seneviratne, S. I., Handmer, J., Nicholls, N., Peduzzi, P., Mechler, R., Bouwer, L. M., Arnell, N., Mach, K., Muir-Wood, R., Brakenridge, G. R., Kron, W., Benito, G., Honda, Y., Takahashi, K. &

- Sherstyukov, B. 2013. Flood risk and climate change: global and regional perspectives. *Hydrological Sciences Journal*, 1-28.
- Kundzewicz, Z. W., Luger, N., Dankers, R., Hirabayashi, Y., Döll, P., Pińskwar, I., Dysarz, T., Hochrainer, S. & Matczak, P. 2010. Assessing river flood risk and adaptation in Europe—review of projections for the future. *Mitigation and Adaptation Strategies for Global Change*, 15, 641-656.
- Kundzewicz, Z. W. & Takeuchi, K. 1999. Flood protection and management: quo vadimus? *Hydrological Sciences Journal-Journal Des Sciences Hydrologiques*, 44, 417-432.
- Kuok, K. K., Liew, Z. Z. & Chiu, P. 2013. Flood map development by coupling satellite maps and three-dimensional drafting software: Case study of the Sarawak River Basin. *Water SA*, 39, 175-182.
- Laban, N., Abdellatif, B., Ebeid, H. M., Shedeed, H. A. & Tolba, M. F. Improving Land-Cover and Crop-Types Classification of Sentinel-2 Satellite Images. 2018 Cham. Springer International Publishing, 449-458.
- Laliberte, A. S., Browning, D. M. & Rango, A. 2012. A comparison of three feature selection methods for object-based classification of sub-decimeter resolution UltraCam-L imagery. *International Journal of Applied Earth Observation and Geoinformation*, 15, 70-78.
- Latif, A. A. & Arshad, N. H. A review of flood management governance framework in Malaysia and selected countries. Conference Proceedings - 6th International Conference on Information Technology and Multimedia at UNITEN: Cultivating Creativity and Enabling Technology Through the Internet of Things, ICIMU 2014, 2015. 178-183.
- Lawal, D. U., Matori, A. N., Yusof, K. W., Hashim, A. M., Aminu, M., Sabri, S., Balogun, A. L., Chandio, I. A. & Mokhtar, M. R. M. 2014a. group-based decision support for flood hazard forecasting: A geospatial technology-based group analytic hierarchy process approach. *Research Journal of Applied Sciences, Engineering and Technology*, 7, 4838-4850.
- Lawal, D. U., Matori, A. N., Yusuf, K. W., Hashim, A. M. & Balogun, A. L. Analysis of the flood extent extraction model and the natural flood influencing factors: A GIS-based and remote sensing analysis. IOP Conference Series: Earth and Environmental Science, 2014b.
- Lawrence, R. L. & Moran, C. J. 2015. The AmericaView classification methods accuracy comparison project: A rigorous approach for model selection. *Remote Sensing of Environment*, 170, 115-120.
- Le, T. V. H., Nguyen, H. N., Wolanski, E., Tran, T. C. & Haruyama, S. 2007. The combined impact on the flooding in Vietnam's Mekong River delta of local man-made structures, sea level rise, and dams upstream in the river catchment. *Estuarine, Coastal and Shelf Science*, 71, 110-116.
- Lehner, A. & Blaschke, T. 2019. A Generic Classification Scheme for Urban Structure Types. *Remote Sensing*, 11, 173.
- Lehner, B., Verdin, K. & Jarvis, A. 2006. HydroSHEDS technical documentation, version 1.0. *World Wildlife Fund US, Washington, DC*, 1-27.
- Lekuthai, A. & Vongvisessomjai, S. 2001. Intangible Flood Damage Quantification. *Water Resources Management*, 15, 343-362.
- Leman, A. M., Rahman, K. A., Salleh, M. N. M., Baba, I., Feriyanto, D., Johnson, L. S. C. & Hidayah, S. N. 2016a. A review of flood catastrophic management in Malaysia. *ARPJ Journal of Engineering and Applied Sciences*, 11, 8738-8742.

- Leman, A. M., Rahman, K. A., Salleh, M. N. M., Baba, I., Johnson, L. S. C., Feriyanto, D., Mumamad, S. N. H. & Kassim, S. N. H. 2016b. Development of inter agency information system for flood catastrophic preparedness in Malaysia. *ARPN Journal of Engineering and Applied Sciences*, 11, 8733-8737.
- Li, D., Zhang, G., Wu, Z. & Yi, L. 2010. An Edge Embedded Marker-Based Watershed Algorithm for High Spatial Resolution Remote Sensing Image Segmentation. *IEEE Transactions on Image Processing*, 19, 2781-2787.
- Li, M., Ma, L., Blaschke, T., Cheng, L. & Tiede, D. 2016. A systematic comparison of different object-based classification techniques using high spatial resolution imagery in agricultural environments. *International Journal of Applied Earth Observation and Geoinformation*, 49, 87-98.
- Li, M., Stein, A., Bijker, W. & Zhan, Q. 2016. Urban land use extraction from Very High Resolution remote sensing imagery using a Bayesian network. *ISPRS Journal of Photogrammetry and Remote Sensing*, 122, 192-205.
- Li, P., Cheng, T. & Guo, J. 2009. Multivariate Image Texture by Multivariate Variogram for Multispectral Image Classification. *Photogrammetric Engineering & Remote Sensing*, 75, 147-157.
- Li, P. & Xiao, X. 2007. Multispectral image segmentation by a multichannel watershed - based approach. *International Journal of Remote Sensing*, 28, 4429-4452.
- Li, Q., Wong, F. K. K. & Fung, T. Comparison Feature Selection Methods for Subtropical Vegetation Classification with Hyperspectral Data. IGARSS 2019 - 2019 IEEE International Geoscience and Remote Sensing Symposium, 28 July-2 Aug. 2019 2019. 3693-3696.
- Li, Z., Yang, D. & Hong, Y. 2013. Multi-scale evaluation of high-resolution multi-sensor blended global precipitation products over the Yangtze River. *Journal of Hydrology*, 500, 157-169.
- Liang, B. & Weng, Q. 2018. Characterizing Urban Landscape by Using Fractal-based Texture Information. *Photogrammetric Engineering & Remote Sensing*, 84, 695-710.
- Lim, S. P. & Cheok, H. S. 2009. Two-dimensional flood modelling of the Damansara river. *Proceedings of the Institution of Civil Engineers: Water Management*, 162, 13-24.
- Lim, Y. H. & Lye, L. M. 2003. Regional flood estimation for ungauged basins in Sarawak, Malaysia. *Hydrological Sciences Journal*, 48, 79-94.
- Liu, D. & Xia, F. 2010. Assessing object-based classification: advantages and limitations. *Remote Sensing Letters*, 1, 187-194.
- Liu, H. & Setiono, R. Chi2: Feature selection and discretization of numeric attributes. Proceedings of 7th IEEE International Conference on Tools with Artificial Intelligence, 1995. IEEE, 388-391.
- Liu, J., Duan, Z., Jiang, J. & Zhu, A.-X. 2015. Evaluation of Three Satellite Precipitation Products TRMM 3B42, CMORPH, and PERSIANN over a Subtropical Watershed in China. *Advances in Meteorology*, 2015, 13.
- Liu, J., Li, P. & Wang, X. 2015. A new segmentation method for very high resolution imagery using spectral and morphological information. *ISPRS Journal of Photogrammetry and Remote Sensing*, 101, 145-162.
- Liu, J. G. & Mason, P. J. 2016. *Image processing and GIS for remote sensing: Techniques and applications*, John Wiley & Sons.
- Liu, Y., Li, X. & Wu, Z. 2003. The feature subset selection algorithm. *Journal of Electronics (China)*, 20, 57-61.

- Liu, Z., Merwade, V. & Jafarzadegan, K. 2019. Investigating the role of model structure and surface roughness in generating flood inundation extents using one- and two-dimensional hydraulic models. *Journal of Flood Risk Management*, 12, e12347.
- Livingston, F. 2005. Implementation of Breiman's random forest machine learning algorithm. *ECE591Q Machine Learning Journal Paper*, 1-13.
- Lo Conti, F., Hsu, K.-L., Noto, L. V. & Sorooshian, S. 2014. Evaluation and comparison of satellite precipitation estimates with reference to a local area in the Mediterranean Sea. *Atmospheric Research*, 138, 189-204.
- Long, Y., Zhang, Y. & Ma, Q. 2016. A Merging Framework for Rainfall Estimation at High Spatiotemporal Resolution for Distributed Hydrological Modeling in a Data-Scarce Area. *Remote Sensing*, 8, 599.
- Löw, F., Michel, U., Dech, S. & Conrad, C. 2013. Impact of feature selection on the accuracy and spatial uncertainty of per-field crop classification using support vector machines. *ISPRS journal of photogrammetry and remote sensing*, 85, 102-119.
- Lu, D., Batistella, M., Li, G., Moran, E., Hetrick, S., Freitas, C. D. C., Dutra, L. V. & Sant'anna, S. J. S. 2012. Land use/cover classification in the Brazilian Amazon using satellite images. *Pesquisa Agropecuária Brasileira*, 47, 1185-1208.
- Lu, D. & Weng, Q. 2007. A survey of image classification methods and techniques for improving classification performance. *International Journal of Remote Sensing*, 28, 823-870.
- Lu, X., Tang, G., Wei, M., Yang, L. & Zhang, Y. 2018. Evaluation of multi-satellite precipitation products in Xinjiang, China. *International Journal of Remote Sensing*, 39, 7437-7462.
- Lu, Z., Im, J., Rhee, J. & Hodgson, M. 2014. Building type classification using spatial and landscape attributes derived from LiDAR remote sensing data. *Landscape and Urban Planning*, 130, 134-148.
- Luu, C. & Von Meding, J. 2018. A Flood Risk Assessment of Quang Nam, Vietnam Using Spatial Multicriteria Decision Analysis. *Water*, 10, 461.
- Lv, Z. Y., Zhang, P., Benediktsson, J. A. & Shi, W. Z. 2014. Morphological Profiles Based on Differently Shaped Structuring Elements for Classification of Images With Very High Spatial Resolution. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 7, 4644-4652.
- Ma, J., Sun, W., Yang, G. & Zhang, D. 2018. Hydrological Analysis Using Satellite Remote Sensing Big Data and CREST Model. *IEEE Access*, 6, 9006-9016.
- Ma, L., Cheng, L., Li, M., Liu, Y. & Ma, X. 2015a. Training set size, scale, and features in Geographic Object-Based Image Analysis of very high resolution unmanned aerial vehicle imagery. *ISPRS Journal of Photogrammetry and Remote Sensing*, 102, 14-27.
- Ma, L., Fu, T., Blaschke, T., Li, M., Tiede, D., Zhou, Z., Ma, X. & Chen, D. 2017a. Evaluation of Feature Selection Methods for Object-Based Land Cover Mapping of Unmanned Aerial Vehicle Imagery Using Random Forest and Support Vector Machine Classifiers. *ISPRS International Journal of Geo-Information*, 6, 51.
- Ma, L., Li, M., Ma, X., Cheng, L., Du, P. & Liu, Y. 2017b. A review of supervised object-based land-cover image classification. *ISPRS Journal of Photogrammetry and Remote Sensing*, 130, 277-293.
- Ma, L., Ma, F., Ji, Z., Gu, Q., Wu, D., Deng, J. & Ding, J. 2015b. Urban Land Use Classification Using LiDAR Geometric, Spatial Autocorrelation and

- Lacunarity Features Combined with Postclassification Processing Method. *Canadian Journal of Remote Sensing*, 41, 334-345.
- Mah, D. Y. S., Putuhena, F. J. & Lai, S. H. 2011. Modelling the flood vulnerability of deltaic Kuching City, Malaysia. *Natural Hazards*, 58, 865-875.
- Maidin, S. S., Othman, M. & Ahmad, M. N. Information sharing in governance of flood management in malaysia: Cobit based framework. 2014 International Conference on IT Convergence and Security, ICITCS 2014, 2014.
- Maidment, D. R. GIS and hydrologic modeling-an assessment of progress. Third International Conference on GIS and Environmental Modeling, Santa Fe, New Mexico, 1996.
- Mamun, A. A., Hashim, A. & Amir, Z. 2012. Regional statistical models for the estimation of flood peak values at ungauged catchments: Peninsular Malaysia. *Journal of Hydrologic Engineering*, 17, 547-553.
- Man, Q., Dong, P. & Guo, H. 2015. Pixel- and feature-level fusion of hyperspectral and lidar data for urban land-use classification. *International Journal of Remote Sensing*, 36, 1618-1644.
- Manakos, I., Schneider, T. & Ammer, U. 2000. A comparison between the ISODATA and the eCognition classification methods on basis of field data. *IAPRS*, 33, 133-139.
- Mariz, C., Gianelle, D., Bruzzone, L. & Vescovo, L. 2009. Fusion of multi - spectral SPOT - 5 images and very high resolution texture information extracted from digital orthophotos for automatic classification of complex Alpine areas. *International Journal of Remote Sensing*, 30, 2859-2873.
- Martín-Valdivia, M. T., Díaz-Galiano, M. C., Montejo-Raez, A. & Ureña-López, L. A. 2008. Using information gain to improve multi-modal information retrieval systems. *Information Processing & Management*, 44, 1146-1158.
- Martinis, S. & Twele, A. 2010. A Hierarchical Spatio-Temporal Markov Model for Improved Flood Mapping Using Multi-Temporal X-Band SAR Data. *Remote Sensing*, 2, 2240.
- Maruti, S. F., Amerudin, S., Wan Kadir, W. H., Abd Rahman, M. Z., Mohamed Yusof, Z., Ariffin, A. & Tze Huey, T. 2018. HYDRODYNAMIC MODELLING OF A PROPOSED LEBIR AND GALAS DAM FOR FLOOD HAZARD ANALYSIS. *Malaysian Journal of Civil Engineering*, 28.
- Masood, M. & Takeuchi, K. 2012. Assessment of flood hazard, vulnerability and risk of mid-eastern Dhaka using DEM and 1D hydrodynamic model. *Natural Hazards*, 61, 757-770.
- Mastrantonas, N., Bhattacharya, B., Shibuo, Y., Rasmy, M., Espinoza-Dávalos, G. & Solomatine, D. 2019. Evaluating the benefits of merging near-real-time satellite precipitation products: A case study in the Kinu basin region, Japan. *Journal of Hydrometeorology*, 20, 1213-1233.
- Masud, M. M., Sackor, A. S., Ferdous Alam, A. S. A., Al-Amin, A. Q. & Abdul Ghani, A. B. 2018. Community responses to flood risk management – An empirical Investigation of the Marine Protected Areas (MPAs) in Malaysia. *Marine Policy*, 97, 119-126.
- Mather, P. & Tso, B. 2016. *Classification methods for remotely sensed data*, CRC press.
- Mathieu, R., Freeman, C. & Aryal, J. 2007. Mapping private gardens in urban areas using object-oriented techniques and very high-resolution satellite imagery. *Landscape and Urban Planning*, 81, 179-192.

- Matori, A. N. & Lawal, D. U. 2014a. Flood disaster forecasting: A GIS-based group analytic hierarchy process approach. *Applied Mechanics and Materials*.
- Matori, A. N., Lawal, D. U., Yusof, K. W., Hashim, M. A. & Balogun, A. L. Spatial analytic hierarchy process model for flood forecasting: An integrated approach. IOP Conference Series: Earth and Environmental Science, 2014b.
- Mavhura, E., Manyena, S. B., Collins, A. E. & Manatsa, D. 2013. Indigenous knowledge, coping strategies and resilience to floods in Muzarabani, Zimbabwe. *International Journal of Disaster Risk Reduction*, 5, 38-48.
- Maxwell, A. E., Warner, T. A. & Fang, F. 2018. Implementation of machine-learning classification in remote sensing: an applied review. *International Journal of Remote Sensing*, 39, 2784-2817.
- Mccollum, J. R., Gruber, A. & Ba, M. B. 2000. Discrepancy between Gauges and Satellite Estimates of Rainfall in Equatorial Africa. *Journal of Applied Meteorology*, 39, 666-679.
- Mcgarigal, K. 2002. FRAGSTATS: Spatial Pattern Analysis Program for Categorical Maps. Computer software program produced by the authors at the University of Massachusetts, Amherst. <http://www.umass.edu/landeco/research/fragstats/fragstats.html>.
- Md Ali, A., Solomatine, D. P. & Di Baldassarre, G. 2015. Assessing the impact of different sources of topographic data on 1-D hydraulic modelling of floods. *Hydrology and Earth System Sciences*, 19, 631-643.
- Meesuk, V., Vojinovic, Z., Mynett, A. E. & Abdullah, A. F. 2015. Urban flood modelling combining top-view LiDAR data with ground-view SfM observations. *Advances in Water Resources*, 75, 105-117.
- Meng, Q., Cieszewski, C. J., Madden, M. & Borders, B. E. 2007. K Nearest Neighbor Method for Forest Inventory Using Remote Sensing Data. *GIScience & Remote Sensing*, 44, 149-165.
- Merz, B., Kreibich, H., Schwarze, R. & Thielen, A. 2010. Review article "Assessment of economic flood damage". *Natural Hazards and Earth System Science*, 10, 1697-1724.
- Merz, B., Kreibich, H., Thielen, A. & Schmidtke, R. 2004. Estimation uncertainty of direct monetary flood damage to buildings. *Natural Hazards and Earth System Sciences*, 4, 153-163.
- Merz, B., Thielen, A. H. & Gocht, M. 2007. Flood Risk Mapping At The Local Scale: Concepts and Challenges. In: BEGUM, S., STIVE, M. F. & HALL, J. (eds.) *Flood Risk Management in Europe*. AA Dordrecht, The Netherlands: Springer Netherlands.
- Messner, F. & Meyer, V. 2006. Flood damage, vulnerability and risk perception – challenges for flood damage research. In: SCHANZE, J., ZEMAN, E. & MARSALEK, J. (eds.) *Flood Risk Management: Hazards, Vulnerability and Mitigation Measures*. Springer Netherlands.
- Messner, F., Penning-Rowsell, E., Green, C., Meyer, V., Tunstall, S. & Van Der Veen, A. 2007. Evaluating flood damages: guidance and recommendations on principles and methods. *FLOODsite-Report T09-06-01*.
- Meyer, V., Haase, D. & Scheuer, S. 2009. Flood Risk Assessment in European River Basins—Concept, Methods, and Challenges Exemplified at the Mulde River. *Integrated Environmental Assessment and Management*, 5, 17-26.
- Middleton, E. M., Ungar, S. G., Mandl, D. J., Ong, L., Frye, S. W., Campbell, P. E., Landis, D. R., Young, J. P. & Pollack, N. H. 2013. The Earth Observing One



- (EO-1) Satellite Mission: Over a Decade in Space. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 6, 243-256.
- Mo, Y., Zhong, R., Sun, H., Wu, Q., Du, L., Geng, Y. & Cao, S. 2019. Integrated Airborne LiDAR Data and Imagery for Suburban Land Cover Classification Using Machine Learning Methods. *Sensors*, 19, 1996.
- Moazami, S., Golian, S., Kavianpour, M. R. & Hong, Y. 2013. Comparison of PERSIANN and V7 TRMM Multi-satellite Precipitation Analysis (TMPA) products with rain gauge data over Iran. *International Journal of Remote Sensing*, 34, 8156-8171.
- Mohamad, S., Hashim, N. M., Aiyub, K. & Toriman, M. E. 2012. Flash flood and community's response at Sg. Lembing, Pahang. *Advances in Natural and Applied Sciences*, 6, 19-25.
- Mohammed, T. A., Al-Hassoun, S. & Ghazali, A. H. 2011a. Prediction of flood levels along a stretch of the langat river with insufficient hydrological data. *Pertanika Journal of Science and Technology*, 19, 237-248.
- Mohammed, T. A., Said, S., Bardaie, M. Z. & Basri, S. N. Numerical simulation of flood levels for tropical rivers. IOP Conference Series: Materials Science and Engineering, 2011b.
- Mohd Nasir, M. R. 2011. *A study of the Kelantan Malay cultural landscape: recent trends and future prospect*. Doctor of Philosophy, University of Sheffield.
- Mohd Zad, S., Zulkafli, Z. & Muharram, F. 2018. Satellite Rainfall (TRMM 3B42-V7) Performance Assessment and Adjustment over Pahang River Basin, Malaysia. *Remote Sensing*, 10, 388.
- Mohit, M. A. & Sellu, G. M. 2013. Mitigation of climate change effects through non-structural flood disaster management in Pekan Town, Malaysia. *Procedia-Social and Behavioral Sciences*, 85, 564-573.
- Mojaddadi, H., Pradhan, B., Nampak, H., Ahmad, N. & Ghazali, A. H. B. 2017. Ensemble machine-learning-based geospatial approach for flood risk assessment using multi-sensor remote-sensing data and GIS. *Geomatics, Natural Hazards and Risk*, 8, 1080-1102.
- Moradkhani, H. & Sorooshian, S. 2008. General Review of Rainfall-Runoff Modeling: Model Calibration, Data Assimilation, and Uncertainty Analysis. In: SOROOSHIAN, S., HSU, K.-L., COPPOLA, E., TOMASSETTI, B., VERDECCHIA, M. & VISCONTI, G. (eds.) *Hydrological Modelling and the Water Cycle*. Springer Berlin Heidelberg.
- Moriasi, D. N., Arnold, J. G., Van Liew, M. W., Bingner, R. L., Harmel, R. D. & Veith, T. L. 2007. Model Evaluation Guidelines for Systematic Quantification of Accuracy in Watershed Simulations. *Transactions of the ASABE*, 50, 885-900.
- Mountrakis, G., Im, J. & Ogole, C. 2011. Support vector machines in remote sensing: A review. *ISPRS Journal of Photogrammetry and Remote Sensing*, 66, 247-259.
- Mousavi, M. E., Irish, J. L., Frey, A. E., Olivera, F. & Edge, B. L. 2011. Global warming and hurricanes: the potential impact of hurricane intensification and sea level rise on coastal flooding. *Climatic Change*, 104, 575-597.
- Mucherino, A., Papajorgji, P. J. & Pardalos, P. M. 2009. k-Nearest Neighbor Classification. *Data Mining in Agriculture*. New York, NY: Springer New York.
- Mück, M., Taubenböck, H., Post, J., Wegscheider, S., Strunz, G., Sumaryono, S. & Ismail, F. A. 2013. Assessing building vulnerability to earthquake and tsunami hazard using remotely sensed data. *Natural Hazards*, 68, 97-114.

- Mueller, M., Segl, K., Heiden, U. & Kaufmann, H. 2006. Potential of High-Resolution Satellite Data in the Context of Vulnerability of Buildings. *Natural Hazards*, 38, 247-258.
- Mugiraneza, T., Nascetti, A. & Ban, Y. 2019. WorldView-2 Data for Hierarchical Object-Based Urban Land Cover Classification in Kigali: Integrating Rule-Based Approach with Urban Density and Greenness Indices. *Remote Sensing*, 11, 2128.
- Muhadi, N. A. & Abdullah, A. F. 2015. Flood damage assessment in agricultural area in Selangor river basin. *Jurnal Teknologi*, 76, 111-117.
- Muhadi, N. A., Abdullah, A. F. & Vojinovic, Z. Estimating Agricultural Losses using Flood Modeling for Rural Area. MATEC Web of Conferences, 2017.
- Muller, A., Reiter, J. & Weiland, U. 2011. Assessment of urban vulnerability towards floods using an indicator-based approach - a case study for Santiago de Chile. *Natural Hazards and Earth System Sciences*, 11, 2107-2123.
- Murray, H., Lucieer, A. & Williams, R. 2010. Texture-based classification of sub-Antarctic vegetation communities on Heard Island. *International Journal of Applied Earth Observation and Geoinformation*, 12, 138-149.
- Mustaffa, A. A., Rasib, A. W., Rosli, M. I., Razi, M. a. M., Adnan, M. S. & Tan, L. W. 2015. Identification of flood-prone areas by integrated remote sensing model. *Lowland Technology International*, 17, 105-110.
- Mustaffa, A. A., Rosli, M. F., Abustan, M. S., Adib, R., Rosli, M. I., Masiri, K. & Saifullizan, B. A Study of Flood Evacuation Center Using GIS and Remote Sensing Technique. IOP Conference Series: Materials Science and Engineering, 2016a.
- Mustaffa, C. S., Marzuki, N. A. & Rahaman, N. H. 2016b. Relationship between communication competence and psychological well-being of flood victims in Malaysia. *International Review of Management and Marketing*, 6, 1-7.
- Myburgh, G. & Niekerk, A. V. 2014. Impact of Training Set Size on Object-Based Land Cover Classification: A Comparison of Three Classifiers. *Int. J. Appl. Geosp. Res.*, 5, 49-67.
- Myint, S. W. 2003. Fractal approaches in texture analysis and classification of remotely sensed data: Comparisons with spatial autocorrelation techniques and simple descriptive statistics. *International Journal of Remote Sensing*, 24, 1925-1947.
- Myint, S. W., Galletti, C. S., Kaplan, S. & Kim, W. K. 2013. Object vs. pixel: a systematic evaluation in urban environments. *Geocarto International*, 28, 657-678.
- Myint, S. W., Gober, P., Brazel, A., Grossman-Clarke, S. & Weng, Q. 2011. Per-pixel vs. object-based classification of urban land cover extraction using high spatial resolution imagery. *Remote Sensing of Environment*, 115, 1145-1161.
- Myint, S. W. & Lam, N. 2005a. Examining lacunarity approaches in comparison with fractal and spatial autocorrelation techniques for urban mapping. *Photogrammetric Engineering & Remote Sensing*, 71, 927-937.
- Myint, S. W. & Lam, N. 2005b. A study of lacunarity-based texture analysis approaches to improve urban image classification. *Computers, Environment and Urban Systems*, 29, 501-523.
- Myint, S. W., Lam, N. & Tyler, J. 2002. An Evaluation of Four Different Wavelet Decomposition Procedures for Spatial Feature Discrimination in Urban Areas. *Transactions in GIS*, 6, 403-429.

- Myint, S. W., Mesev, V. & Lam, N. 2006. Urban Textural Analysis from Remote Sensor Data: Lacunarity Measurements Based on the Differential Box Counting Method. *Geographical Analysis*, 38, 371-390.
- Myint, S. W., Mesev, V., Quattrochi, D. A. & Wentz, E. A. 2015. Urban image classification: Per-pixel classifiers, subpixel analysis, object-based image analysis, and geospatial methods. *Remotely Sensed Data Characterization, Classification, and Accuracies*. CRC Press.
- Nashwan, M. S., Ismail, T. & Ahmed, K. 2018. Flood susceptibility assessment in Kelantan river basin using copula. *International Journal of Engineering and Technology(UAE)*, 7, 584-590.
- Nastiti, K. D., An, H., Kim, Y. & Jung, K. 2018. Large-scale rainfall–runoff–inundation modeling for upper Citarum River watershed, Indonesia. *Environmental Earth Sciences*, 77, 640.
- Negreiros, J. G., Painho, M. T., Aguilar, F. J. & Aguilar, M. A. 2010. A comprehensive framework for exploratory spatial data analysis: Moran location and variance scatterplots. *International Journal of Digital Earth*, 3, 157-186.
- Ng, Z. F., Gisen, J. I. & Akbari, A. Flood Inundation Modelling in the Kuantan River Basin using 1D-2D Flood Modeller coupled with ASTER-GDEM. IOP Conference Series: Materials Science and Engineering, 2018.
- Nguyen, P., Thorstensen, A., Sorooshian, S., Hsu, K. & Aghakouchak, A. 2015. Flood Forecasting and Inundation Mapping Using HiResFlood-UCI and Near-Real-Time Satellite Precipitation Data: The 2008 Iowa Flood. *Journal of Hydrometeorology*, 16, 1171-1183.
- Nguyen, P., Thorstensen, A., Sorooshian, S., Hsu, K., Aghakouchak, A., Sanders, B., Koren, V., Cui, Z. & Smith, M. 2016. A high resolution coupled hydrologic–hydraulic model (HiResFlood-UCI) for flash flood modeling. *Journal of Hydrology*, 541, 401-420.
- Nicholls, R. J., Hoozemans, F. M. & Marchand, M. 1999. Increasing flood risk and wetland losses due to global sea-level rise: regional and global analyses. *Global Environmental Change*, 9, S69-S87.
- Nikolopoulos, E. I., Anagnostou, E. N. & Borga, M. 2013. Using High-Resolution Satellite Rainfall Products to Simulate a Major Flash Flood Event in Northern Italy. *Journal of Hydrometeorology*, 14, 171-185.
- Nirupama, N. & Simonovic, S. P. 2007. Increase of Flood Risk due to Urbanisation: A Canadian Example. *Natural Hazards*, 40, 25.
- Noor, M. S. F. M., Sidek, L. M., Basri, H., Husni, M. M. M., Jaafar, A. S., Kamaluddin, M. H., Majid, W. H. a. W. A., Mohammad, A. H. & Osman, S. Development of Flood Forecasting Using Statistical Method in Four River Basins in Terengganu, Malaysia. IOP Conference Series: Earth and Environmental Science, 2016.
- Novack, T., Esch, T., Kux, H. & Stilla, U. 2011. Machine Learning Comparison between WorldView-2 and QuickBird-2-Simulated Imagery Regarding Object-Based Urban Land Cover Classification. *Remote Sensing*, 3, 2263-2282.
- Okamoto, K. I., Ushio, T., Iguchi, T., Takahashi, N. & Iwanami, K. The global satellite mapping of precipitation (GSMaP) project. Proceedings. 2005 IEEE International Geoscience and Remote Sensing Symposium, 2005. IGARSS '05., 29-29 July 2005 2005. 3414-3416.
- Olanrewaju, A. T. & Ahmad, R. 2018. Examining the information dissemination process on social media during the Malaysia 2014 floods using Social Network

- Analysis (SNA). *Journal of Information and Communication Technology*, 17, 141-166.
- Omran, A., Schwarz-Herion, O. & Abu Bakar, A. 2018. Factors contributing to the catastrophic flood in Malaysia. *The Impact of Climate Change on Our Life: The Questions of Sustainability*.
- Oo, S. S., Amin, R. B. M., Aziz, A. B. A., Aung, M. M. T. & Husain, R. B. 2016. Prevalence and perceived severity of post-traumatic stress disorder among flood victims in Kuala Terengganu, Malaysia. *Malaysian Journal of Public Health Medicine*, 16, 30-40.
- Osanaiye, O., Cai, H., Choo, K.-K. R., Dehghantanha, A., Xu, Z. & Dlodlo, M. 2016. Ensemble-based multi-filter feature selection method for DDoS detection in cloud computing. *EURASIP Journal on Wireless Communications and Networking*, 2016, 130.
- Osman, S. & Abustan, I. Estimating the Clark Instantaneous Unit Hydrograph Parameters for Selected Gauged Catchments in The West Coast of Peninsular Malaysia. *ASEAN Engineering Journal Part C*, 2011. 128-143.
- Otsu, N. 1979. A Threshold Selection Method from Gray-Level Histograms. *IEEE Transactions on Systems, Man, and Cybernetics*, 9, 62-66.
- Özdoğan, M. 2016. Image Classification Methods in Land Cover and Land Use. In: THENKABAIL, P. (ed.) *Remote Sensing Handbook, Volume I, Remotely Sensed Data Characterization, Classification, and Accuracies*. New York: CRC Press.
- Pacifici, F., Chini, M. & Emery, W. J. 2009. A neural network approach using multi-scale textural metrics from very high-resolution panchromatic imagery for urban land-use classification. *Remote Sensing of Environment*, 113, 1276-1292.
- Padlee, S. F., Razali, N. N. H. N., Zulkiffli, S. N. A. & Hussin, N. Z. I. 2018. An assessment of the perception and satisfaction with flood evacuation centre service quality in East Coast states of Peninsular Malaysia. *Journal of Sustainability Science and Management*, 65-77.
- Paiva, R. C. D., Collischonn, W. & Tucci, C. E. M. 2011. Large scale hydrologic and hydrodynamic modeling using limited data and a GIS based approach. *Journal of Hydrology*, 406, 170-181.
- Pakoksung, K. & Takagi, M. 2016. Effect of satellite based rainfall products on river basin responses of runoff simulation on flood event. *Modeling Earth Systems and Environment*, 2, 143.
- Pal, M. 2005. Random forest classifier for remote sensing classification. *International Journal of Remote Sensing*, 26, 217-222.
- Pal, M. 2006. Support vector machine - based feature selection for land cover classification: a case study with DAIS hyperspectral data. *International Journal of Remote Sensing*, 27, 2877-2894.
- Pal, M. 2013. Hybrid genetic algorithm for feature selection with hyperspectral data. *Remote Sensing Letters*, 4, 619-628.
- Pal, M. & Foody, G. M. 2010. Feature selection for classification of hyperspectral data by SVM. *IEEE Transactions on Geoscience and Remote Sensing*, 48, 2297-2307.
- Pal, M. & Mather, P. M. 2003. An assessment of the effectiveness of decision tree methods for land cover classification. *Remote Sensing of Environment*, 86, 554-565.

- Pal, M. & Mather, P. M. 2005. Support vector machines for classification in remote sensing. *International Journal of Remote Sensing*, 26, 1007-1011.
- Pal, N. R. & Pal, S. K. 1993. A review on image segmentation techniques. *Pattern Recognition*, 26, 1277-1294.
- Palsson, F., Sveinsson, J. R., Benediktsson, J. A. & Aanaes, H. 2012. Classification of Pansharpened Urban Satellite Images. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 5, 281-297.
- Panagiota, M., Erwan, P., Philippe, G. & Jocelyn, C. Seismic vulnerability assessment using Support Vector Machine classification for remote sensing and in-situ Data. 15th World Conference on Earthquake Engineering 2012, 24-28 September 2012a Lisbon, Portugal. Curran Associates, Inc.
- Panagiota, M., Jocelyn, C., Erwan, P. & Philippe, G. A support vector regression approach for building seismic vulnerability assessment and evaluation from remote sensing and in-situ data. 2012 IEEE International Geoscience and Remote Sensing Symposium, 22-27 July 2012 2012b. 7533-7536.
- Papathoma-Köhle, M. 2016. Vulnerability curves vs. vulnerability indicators: application of an indicator-based methodology for debris-flow hazards. *Nat. Hazards Earth Syst. Sci.*, 16, 1771-1790.
- Papathoma-Köhle, M., Gems, B., Sturm, M. & Fuchs, S. 2017. Matrices, curves and indicators: A review of approaches to assess physical vulnerability to debris flows. *Earth-Science Reviews*, 171, 272-288.
- Parry, M., Parry, M. L., Canziani, O., Palutikof, J., Van Der Linden, P. & Hanson, C. 2007. *Climate change 2007-impacts, adaptation and vulnerability: Working group II contribution to the fourth assessment report of the IPCC*, Cambridge University Press.
- Paska, J., Lau, A. M. S., Tan, M. L. & Tan, K. C. Evaluation of TRMM 3B42V7 product on extreme precipitation measurements over peninsular Malaysia. *SPIE Remote Sensing*, 2017. SPIE, 6.
- Patrick, M., Mah, Y. S., Putuhena, F. J., Wang, Y. C. & Selaman, O. S. 2017. TRMM Satellite Algorithm Estimates to Represent the Spatial Distribution of Rainstorms. *MATEC Web Conf.*, 87, 01006.
- Penning-Rowsell, E. C. & Chatterton, J. B. 1977. *The benefits of flood alleviation: A manual of assessment techniques*, Saxon House Farnborough.
- Penning - Rowsell, E. & Green, C. 2000. New insights into the appraisal of flood - alleviation benefits:(1) Flood damage and flood loss information. *Water and Environment Journal*, 14, 347-353.
- Perera, E. D. P. & Lahat, L. 2015. Fuzzy logic based flood forecasting model for the Kelantan River basin, Malaysia. *Journal of Hydro-Environment Research*, 9, 542-553.
- Pittore, M. & Wieland, M. 2013. Toward a rapid probabilistic seismic vulnerability assessment using satellite and ground-based remote sensing. *Natural Hazards*, 68, 115-145.
- Poona, N. K. & Ismail, R. Reducing hyperspectral data dimensionality using random forest based wrappers. 2013 IEEE International Geoscience and Remote Sensing Symposium - IGARSS, 21-26 July 2013 2013. 1470-1473.
- Pradhan, B. 2009. Effective flood monitoring system using GIS Tools and remote sensing data. In: BEHR, F.-J., SCHRÖDER, D. & PRADEEPKUMAR, A. P. (eds.) *Taking the Benefits of Geographic Information Technologies: Applied*

- Geoinformatics for Society and Environment*. Druckerei Walter Stolz, Kirchheim, Germany: AGSE Publishing.
- Pradhan, B. 2010. Flood susceptible mapping and risk area delineation using logistic regression, GIS and remote sensing. *Journal of Spatial Hydrology*, 9.
- Pradhan, B., Hagemann, U., Shafapour Tehrany, M. & Prechtel, N. 2014. An easy to use ArcMap based texture analysis program for extraction of flooded areas from TerraSAR-X satellite image. *Computers and Geosciences*, 63, 34-43.
- Pradhan, B., Sameen, M. I. & Kalantar, B. 2017. Optimized Rule-Based Flood Mapping Technique Using Multitemporal RADARSAT-2 Images in the Tropical Region. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 10, 3190-3199.
- Pradhan, B. & Shafiee, M. 2009a. Flood Hazard Assessment for Cloud Prone Rainy Areas in a Typical Tropical Environment. *Disaster Advances*, 2, 7-15.
- Pradhan, B., Shafiee, M. & Pirasteh, S. 2009b. Maximum flood prone area mapping using RADARSAT images and GIS: Kelantan river Basin. *International Journal of Geoinformatics*, 5, 11-23.
- Pradhan, B., Tehrany, M. S. & Jebur, M. N. 2016. A New Semiautomated Detection Mapping of Flood Extent from TerraSAR-X Satellite Image Using Rule-Based Classification and Taguchi Optimization Techniques. *IEEE Transactions on Geoscience and Remote Sensing*, 54, 4331-4342.
- Pradhan, B. & Youssef, A. M. 2011. A 100-year maximum flood susceptibility mapping using integrated hydrological and hydrodynamic models: Kelantan River Corridor, Malaysia. *Journal of Flood Risk Management*, 4, 189-202.
- Price, K. P., Guo, X. & Stiles, J. M. 2002. Optimal Landsat TM band combinations and vegetation indices for discrimination of six grassland types in eastern Kansas. *International Journal of Remote Sensing*, 23, 5031-5042.
- Prudhomme, C., Reynard, N. & Crooks, S. 2002. Downscaling of global climate models for flood frequency analysis: where are we now? *Hydrological Processes*, 16, 1137-1150.
- Pu, R., Landry, S. & Yu, Q. 2011. Object-based urban detailed land cover classification with high spatial resolution IKONOS imagery. *International Journal of Remote Sensing*, 32, 3285-3308.
- Qi, C. & Peng, G. 2004. Automatic variogram parameter extraction for textural classification of the panchromatic IKONOS imagery. *IEEE Transactions on Geoscience and Remote Sensing*, 42, 1106-1115.
- Qi, Z., Yeh, A. G. O. & Li, X. 2016. Land Use and Land Cover Mapping and Change Detection and Monitoring Using Radar Remote Sensing. In: PRASAD, T. S. (ed.) *Remote Sensing Handbook-Volume II-Land Resources Monitoring, Modelling, and Mapping with Remote Sensing*. New York: CRC Press.
- Qian, Y., Zhou, W., Yan, J., Li, W. & Han, L. 2015. Comparing Machine Learning Classifiers for Object-Based Land Cover Classification Using Very High Resolution Imagery. *Remote Sensing*, 7, 153-168.
- Qin, Y., Li, S., Vu, T.-T., Niu, Z. & Ban, Y. 2015. Synergistic application of geometric and radiometric features of LiDAR data for urban land cover mapping. *Optics express*, 23, 13761-13775.
- Quattrochi, D., Lam, N. S.-N., Qiu, H. & Zhao, W. 1997. Image characterization and modeling system (ICAMS): a geographic information system for the characterization and modeling of multiscale remote sensing data. CRC Press, Boca Raton, Florida.

- Radhi, A. M., Rohasliney, H. & Zarul, H. 2017. Fish composition and diversity in Perak, Galas and Kelantan rivers (Malaysia) after the major flood of 2014. *Transylvanian Review of Systematical and Ecological Research*, 19, 41-56.
- Rahman, I. I. A. & Alias, N. M. A. Rainfall forecasting using an artificial neural network model to prevent flash floods. 8th International Conference on High-Capacity Optical Networks and Emerging Technologies, HONET 2011, 2011. 323-328.
- Rajbhandari, S., Aryal, J., Osborn, J., Lucieer, A. & Musk, R. 2019. Leveraging Machine Learning to Extend Ontology-Driven Geographic Object-Based Image Analysis (O-GEOBIA): A Case Study in Forest-Type Mapping. *Remote Sensing*, 11, 503.
- Ramakrishnan, S., Hishan, S. S., Shahabuddin, A. S. M. & Kanjanapathy, M. 2016. The role of corporate social responsibility in flood mitigation among the listed insurance companies in Malaysia. *International Review of Management and Marketing*, 6, 86-90.
- Ramezan, C. A., Warner, T. A., Maxwell, A. E. & Price, B. S. 2021. Effects of Training Set Size on Supervised Machine-Learning Land-Cover Classification of Large-Area High-Resolution Remotely Sensed Data. *Remote Sensing*, 13, 368.
- Ramo, R., García, M., Rodríguez, D. & Chuvieco, E. 2018. A data mining approach for global burned area mapping. *International Journal of Applied Earth Observation and Geoinformation*, 73, 39-51.
- Rani, W. N. M. W. M., Nifa, F. a. A., Ismail, M. N. & Khalid, K. N. Planning for post disaster recovery: Lesson learnt from flood events in Kelantan Malaysia. AIP Conference Proceedings, 2017.
- Razi, M., Ariffin, J., Tahir, W. & Arish, N. 2010. Flood estimation studies using hydrologic modeling system (HEC-HMS) for Johor River, Malaysia. *Journal of Applied Sciences*, 10, 930-939.
- Reba, M. N. M., Roslan, N., Syafiuddin, A. & Hashim, M. Evaluation of empirical radar rainfall model during the massive flood in Malaysia. International Geoscience and Remote Sensing Symposium (IGARSS), 2016. 4406-4409.
- Ren, P., Li, J., Feng, P., Guo, Y. & Ma, Q. 2018. Evaluation of Multiple Satellite Precipitation Products and Their Use in Hydrological Modelling over the Luanhe River Basin, China. *Water*, 10, 677.
- Ricci, P., Verderame, G. M., Manfredi, G., Pollino, M., Borfecchia, F., De Cecco, L., Martini, S., Pascale, C., Ristoratore, E. & James, V. Seismic Vulnerability Assessment Using Field Survey and Remote Sensing Techniques. International Conference on Computational Science and Its Applications, 2011 Berlin, Heidelberg. Springer Berlin Heidelberg, 109-124.
- Ritchie, H. & Roser, M. 2018. Urbanization. *Our World in Data*.
- Rizeei, H. M., Pradhan, B. & Saharkhiz, M. A. 2018. Urban object extraction using Dempster Shafer feature-based image analysis from worldview-3 satellite imagery. *International Journal of Remote Sensing*, 40, 1092-1119.
- Rodriguez-Galiano, V. F., Chica-Olmo, M. & Chica-Rivas, M. 2014. Predictive modelling of gold potential with the integration of multisource information based on random forest: a case study on the Rodalquilar area, Southern Spain. *International Journal of Geographical Information Science*, 28, 1336-1354.
- Romali, N. S. 2018. *Flood Damage and Risk Assessment for Urban Area in Segamat*. Phd Thesis, Universiti Teknologi Malaysia.

- Romali, N. S., Yusop, Z. & Ismail, A. Z. Hydrological Modelling using HEC-HMS for Flood Risk Assessment of Segamat Town, Malaysia. IOP Conference Series: Materials Science and Engineering, 2018a.
- Romali, N. S., Yusop, Z. & Ismail, Z. 2015. Flood damage assessment: A review of flood stage–damage function curve. *Proceedings of the International Symposium on Flood Research and Management*. Springer.
- Romali, N. S., Yusop, Z., Sulaiman, M. & Ismail, Z. 2018b. FLOOD RISK ASSESSMENT: A REVIEW OF FLOOD DAMAGE ESTIMATION MODEL FOR MALAYSIA. *Jurnal Teknologi*, 80, 145-153.
- Romanski, P., Kotthoff, L. & Kotthoff, M. L. 2013. Package ‘FSelector’. URL <http://cran.r-project.org/web/packages/FSelector/index.html>.
- Roslee, R., Tongkul, F., Mariappan, S. & Simon, N. 2018. Flood Hazard Analysis (FHAn) using Multi-Criteria Evaluation (MCE) in Penampang Area, Sabah, Malaysia. *ASM Science Journal*, 11, 104-122.
- Röthlisberger, V., Zischg, A. P. & Keiler, M. 2017. Identifying spatial clusters of flood exposure to support decision making in risk management. *Science of The Total Environment*, 598, 593-603.
- Ruiz Estrada, M. A., Koutronas, E., Tahir, M. & Mansor, N. 2017. Hydrological hazard assessment: THE 2014–15 Malaysia floods. *International Journal of Disaster Risk Reduction*, 24, 264-270.
- Ryherd, S. & Woodcock, C. 1996. Combining spectral and texture data in the segmentation of remotely sensed images. *Photogrammetric engineering and remote sensing*, 62, 181-194.
- S V, S. S., Roy, P. S., V, C. & G, S. R. 2018. Flood risk assessment using multi-criteria analysis: a case study from Kopili River Basin, Assam, India. *Geomatics, Natural Hazards and Risk*, 9, 79-93.
- Saadatkah, N., Tehrani, M. H., Mansor, S., Khuzaimah, Z., Kassim, A. & Saadatkah, R. 2016. Impact assessment of land cover changes on the runoff changes on the extreme flood events in the Kelantan River basin. *Arabian Journal of Geosciences*, 9.
- Saber, M. & Yilmaz, K. K. 2018. Evaluation and Bias Correction of Satellite-Based Rainfall Estimates for Modelling Flash Floods over the Mediterranean region: Application to Karpuz River Basin, Turkey. *Water*, 10, 657.
- Sadeghi, M., Nguyen, P., Naeini, M. R., Hsu, K., Braithwaite, D. & Sorooshian, S. 2021. PERSIANN-CCS-CDR, a 3-hourly 0.04° global precipitation climate data record for heavy precipitation studies. *Scientific Data*, 8, 157.
- Saini, S. & Arora, K. 2014. A study analysis on the different image segmentation techniques. *International Journal of Information & Computation Technology*, 4, 1445-1452.
- Saini, S. S. & Kaushik, S. 2012. Risk and vulnerability assessment of flood hazard in part of Ghaggar Basin: A case study of Guhla block, Kaithal, Haryana, India. *International Journal of Geomatics and Geosciences*, 3, 42-54.
- Salarpour, M., Rahman, N. A. & Yusop, Z. 2011. Simulation of flood extent mapping by infoworks RS-case study for tropical catchment. *Journal of Software Engineering*, 5, 127-135.
- Salarpour, M., Yusop, Z., Yusof, F., Shahid, S. & Jajarmizadeh, M. 2013. Flood frequency analysis based on t-copula for Johor river, Malaysia. *Journal of Applied Sciences*, 13, 1021-1028.



- Sanlang, S., Cao, S., Du, M., Mo, Y., Chen, Q. & He, W. 2021. Integrating Aerial LiDAR and Very-High-Resolution Images for Urban Functional Zone Mapping. *Remote Sensing*, 13, 2573.
- Satgé, F., Xavier, A., Pillco Zolá, R., Hussain, Y., Timouk, F., Garnier, J. & Bonnet, M.-P. 2017. Comparative Assessments of the Latest GPM Mission's Spatially Enhanced Satellite Rainfall Products over the Main Bolivian Watersheds. *Remote Sensing*, 9, 369.
- Saudi, A. S. M., Juahir, H., Azid, A. & Azaman, F. 2015. Flood risk index assessment in Johor river basin. *Malaysian Journal of Analytical Sciences*, 19, 991-1000.
- Sayama, T. 2017. Rainfall-Runoff-Inundation Model User's Manual. Disaster Prevention Research Institute (DPRI), Kyoto University.
- Sayama, T., Fukami, K., Tanaka, S. & Takeuchi, K. Rainfall-runoff-inundation analysis for flood risk assessment at the regional scale. Proceeding of the Fifth Conference of Asia Pacific Association of Hydrology and Water Resources (APHW), 2010. 568-576.
- Sayama, T., Ozawa, G., Kawakami, T., Nabesaka, S. & Fukami, K. 2012. Rainfall-runoff-inundation analysis of the 2010 Pakistan flood in the Kabul River basin. *Hydrological Sciences Journal*, 57, 298-312.
- Sayama, T., Tatebe, Y., Iwami, Y. & Tanaka, S. 2015. Hydrologic sensitivity of flood runoff and inundation: 2011 Thailand floods in the Chao Phraya River basin. *Natural Hazards and Earth System Sciences*, 15, 1617-1630.
- Sayama, T., Tatebe, Y. & Tanaka, S. 2017. An emergency response-type rainfall-runoff-inundation simulation for 2011 Thailand floods. *Journal of Flood Risk Management*, 10, 65-78.
- Schanze, J. 2006. Flood Risk Management – A Basic Framework. In: SCHANZE, J., ZEMAN, E. & MARSALEK, J. (eds.) *Flood Risk Management: Hazards, Vulnerability and Mitigation Measures*. Springer Netherlands.
- Schiewe, J. 2002. Segmentation of high-resolution remotely sensed data-concepts, applications and problems. *International Archives of Photogrammetry Remote Sensing and Spatial Information Sciences*, 34, 380-385.
- Schlosser, A. D., Szabó, G., Bertalan, L., Varga, Z., Enyedi, P. & Szabó, S. 2020. Building Extraction Using Orthophotos and Dense Point Cloud Derived from Visual Band Aerial Imagery Based on Machine Learning and Segmentation. *Remote Sensing*, 12, 2397.
- Schumann, G. J., Bates, P. D., Apel, H. & Aronica, G. T. 2018. *Global flood hazard: Applications in modeling, mapping, and forecasting*, John Wiley & Sons.
- Schuol, J., Abbaspour, K. C., Srinivasan, R. & Yang, H. 2008. Estimation of freshwater availability in the West African sub-continent using the SWAT hydrologic model. *Journal of Hydrology*, 352, 30-49.
- Semire, F. A., Mohd-Mokhtar, R., Ismail, W., Mohamad, N. & Mandeep, J. S. 2012. Ground validation of space-borne satellite rainfall products in Malaysia. *Advances in Space Research*, 50, 1241-1249.
- Serpico, S., D'inca, M., Melgani, F. & Moser, G. 2003. *A comparison of feature reduction techniques for classification of hyperspectral remote-sensing data*, SPIE.
- Sertel, E., Topaloğlu, R. H., Şallı, B., Yay Algan, I. & Aksu, G. A. 2018. Comparison of Landscape Metrics for Three Different Level Land Cover/Land Use Maps. *ISPRS International Journal of Geo-Information*, 7, 408.

- Shaari, M. S. M., Abd Karim, M. Z. & Hasan-Basri, B. 2017. Does flood disaster lessen GDP growth? Evidence from Malaysia's manufacturing and agricultural sectors. *Malaysian Journal of Economic Studies*, 54, 61-81.
- Shabri, A. & Jemain, A. A. 2013. Regional flood frequency analysis for Southwest Peninsular Malaysia by LQ-moments. *Journal of Flood Risk Management*, 6, 360-371.
- Shafiai, S. & Khalid, M. S. 2016. Examining of issues on flood disaster management in Malaysia. *International Review of Management and Marketing*, 6, 51-56.
- Shafie, A. 2009. *Extreme Flood Event: A Case Study on Floods of 2006 and 2007 in Johor, Malaysia*. Msc., Colorado State University.
- Shah, S. M. H., Mustaffa, Z. & Yusof, K. W. 2017. Disasters worldwide and floods in the Malaysian region: a brief review. *Indian Journal of Science and Technology*, 10.
- Shaluf, I. M. & Ahmadun, F. L.-R. 2006. Disaster types in Malaysia: an overview. *Disaster Prevention and Management*, 15, 286-298.
- Sharif, M., Burn, D. H. & Wey, K. M. Daily and hourly weather data generation using a k-nearest neighbour approach. Canadian Hydrotechnical Conference, 2007. CHC Winnipeg, 1-10.
- Sharma, R., Ghosh, A. & Joshi, P. K. 2013. Decision tree approach for classification of remotely sensed satellite data using open source support. *Journal of Earth System Science*, 122, 1237-1247.
- Sharma, S. K., Kwak, Y.-J., Kumar, R. & Sarma, B. 2018. Analysis of Hydrological Sensitivity for Flood Risk Assessment. *ISPRS International Journal of Geo-Information*, 7, 51.
- Shen, W., Wu, G., Sun, Z., Xiong, W., Fu, Z. & Xiao, R. Study on classification methods of remote sensing image based on decision tree technology. 2011 International Conference on Computer Science and Service System (CSSS), 2011. IEEE, 4058-4061.
- Shi, D. & Yang, X. 2017. A Relative Evaluation of Random Forests for Land Cover Mapping in an Urban Area. *Photogrammetric Engineering & Remote Sensing*, 83, 541-552.
- Shi, L., Wan, Y., Gao, X. & Wang, M. 2018. Feature Selection for Object-Based Classification of High-Resolution Remote Sensing Images Based on the Combination of a Genetic Algorithm and Tabu Search. *Computational Intelligence and Neuroscience*, 2018, 6595792.
- Shirahata, L. M., Iizuka, K., Yusupujang, A., Rinawan, F. R., Bhattarai, R. & Dong, X. 2017. Production of Global Land Cover Data–GLCNMO2013. *Journal of Geography and Geology*, 9.
- Shivhare, P. & Gupta, V. 2015. Review of Image Segmentation Techniques Including Pre & Post Processing Operations. *International Journal of Engineering and Advanced Technology*, 4, 153-157.
- Shrestha, B. B., Sawano, H., Ohara, M., Yamazaki, Y. & Tokunaga, Y. 2018. Methodology for agricultural flood damage assessment. *Recent Advances in Flood Risk Management*. IntechOpen.
- Sidek, L. M., Rostam, N. E., Hidayah, B., Roseli, Z. A., Majid, W. H. a. W. A., Zahari, N. Z., Salleh, S. H. M., Ahmad, R. D. R. & Ahmad, M. N. Hydrology Analysis and Modelling for Klang River Basin Flood Hazard Map. IOP Conference Series: Earth and Environmental Science, 2016.
- Simonovic, S. P. 2002. Two New Non-structural Measures for Sustainable Management of Floods. *Water International*, 27, 38-46.

- Singh, V. P. 1995. *Computer models of watershed hydrology*, Water Resources Publications Highlands Ranch, CO.
- Singh, Y. K., Sinha, N. & Singh, S. K. Heart Disease Prediction System Using Random Forest. 2017 Singapore. Springer Singapore, 613-623.
- Sinnakaudan, S. K., Ab Ghani, A., Ahmad, M. S. S. & Zakaria, N. A. 2003. Flood risk mapping for Pari River incorporating sediment transport. *Environmental Modelling & Software*, 18, 119-130.
- Sinnakaudan, S. K., Ab Ghani, A., Kiat, C. C., Ahmad, M. S. S. & Zakaria, N. A. 2002. *Integrated triangular irregular network (ITIN) model for flood risk analysis case study: Pari River, Ipoh, Malaysia*, Singapore, World Scientific Publ Co Pte Ltd.
- Sinnakaudan, S. K. & Bakar, S. H. A. 2005. Tight coupling of SFlood and ArcView GIS 3.2 for flood risk analysis. *Geo-information for Disaster Management*.
- Sitterson, J., Knightes, C., Parmar, R., Wolfe, K., Muche, M. & Avant, B. 2017. An Overview of Rainfall-Runoff Model Types.
- Smith, D. I. 1994. Flood Damage Estimation - a Review of Urban Stage-Damage Curves and Loss Functions. *Water Sa*, 20, 231-238.
- Soetanto, R. & Proverbs, D. G. 2004. Impact of flood characteristics on damage caused to UK domestic properties: the perceptions of building surveyors. *Structural Survey*, 22, 95-104.
- Solín, L. & Skubinčan, P. 2013. Flood risk assessment and management: Review of concepts, definitions and methods. *Geograficky Casopis*, 65, 23-44.
- Soo, E. Z. X., Jaafar, W. Z. W., Lai, S. H., Islam, T. & Srivastava, P. 2018. Evaluation of satellite precipitation products for extreme flood events: case study in Peninsular Malaysia. *Journal of Water and Climate Change*.
- Sorooshian, S., Hsu, K.-L., Gao, X., Gupta, H. V., Imam, B. & Braithwaite, D. 2000. Evaluation of PERSIANN System Satellite-Based Estimates of Tropical Rainfall. *Bulletin of the American Meteorological Society*, 81, 2035-2046.
- Spence, R. J. S., Kelman, I., Calogero, E., Toyos, G., Baxter, P. J. & Komorowski, J. C. 2005. Modelling expected physical impacts and human casualties from explosive volcanic eruptions. *Nat. Hazards Earth Syst. Sci.*, 5, 1003-1015.
- Stephenson, V. & D'ayala, D. 2014. A new approach to flood vulnerability assessment for historic buildings in England. *Natural Hazards and Earth System Sciences*, 14, 1035-1048.
- Stisen, S. & Sandholt, I. 2010. Evaluation of remote-sensing-based rainfall products through predictive capability in hydrological runoff modelling. *Hydrological Processes*, 24, 879-891.
- Strahler, A. H. 1980. The use of prior probabilities in maximum likelihood classification of remotely sensed data. *Remote Sensing of Environment*, 10, 135-163.
- Su, C.-H., Zhang, J., Gruber, A., Parinussa, R., Ryu, D., Crow, W. T. & Wagner, W. 2016. Error decomposition of nine passive and active microwave satellite soil moisture data sets over Australia. *Remote Sensing of Environment*, 182, 128-140.
- Su, T. 2019. Object-based feature selection using class-pair separability for high-resolution image classification. *International Journal of Remote Sensing*, 41, 238-271.
- Su, W., Li, J., Chen, Y., Liu, Z., Zhang, J., Low, T. M., Suppiah, I. & Hashim, S. a. M. 2008. Textural and local spatial statistics for the object - oriented

- classification of urban areas using high resolution imagery. *International Journal of Remote Sensing*, 29, 3105-3117.
- Subramaniam, S. K., Gannapathy, V. R., Anas, S. A., Diah, A. B. M., Suaidi, M. K. & Hamidon, A. H. 2009. *Intellectual and remotely self monitored Flood Observatory System for high frequency flood prone locations*, Athens, World Scientific and Engineering Acad and Soc.
- Subramaniam, S. K., Gannapathy, V. R., Subramonian, S. & Hamidon, A. H. 2010. Flood level indicator and risk warning system for remote location monitoring using flood observatory system. *WSEAS Transactions on Systems and Control*, 5, 153-163.
- Sulaiman, N. A., Mastor, T. A., Mat, M. S. C. & Samad, A. M. Flood hazard zoning and risk assessment for Bandar Segamat sustainability using analytical hierarchy process (AHP). Proceedings - 2015 IEEE 11th International Colloquium on Signal Processing and Its Applications, CSPA 2015, 2015. 72-77.
- Sulaiman, N. H., Kamarudin, M. K. A., Toriman, M. E., Juahir, H., Ata, F. M., Azid, A., Wahab, N. J. A., Umar, R., Khalit, S. I., Makhtar, M., Arfan, A. & Sideng, U. 2017. Relationship of rainfall distribution and water level on major flood 2014 in Pahang River Basin, Malaysia. *EnvironmentAsia*, 10, 1-8.
- Sulaiman, W. N. A., Heshmatpoor, A. & Rosli, M. H. 2010. Identification of flood source areas in Pahang river basin, Peninsular Malaysia. *EnvironmentAsia*, 3, 73-78.
- Sulaiman, Z., Mohamad, N., Ismail, T. a. T., Johari, N. & Hussain, N. H. N. 2016. Infant feeding concerns in times of natural disaster: Lessons learned from the 2014 flood in Kelantan, Malaysia. *Asia Pacific Journal of Clinical Nutrition*, 25, 625-630.
- Sun, Q., Miao, C., Duan, Q., Ashouri, H., Sorooshian, S. & Hsu, K. L. 2018a. A review of global precipitation data sets: Data sources, estimation, and intercomparisons. *Reviews of Geophysics*, 56, 79-107.
- Sun, W., Sun, Y., Li, X., Wang, T., Wang, Y., Qiu, Q. & Deng, Z. 2018b. Evaluation and Correction of GPM IMERG Precipitation Products over the Capital Circle in Northeast China at Multiple Spatiotemporal Scales. *Advances in Meteorology*, 2018, 14.
- Sun, X., Lin, X., Shen, S. & Hu, Z. 2017. High-resolution remote sensing data classification over urban areas using random forest ensemble and fully connected conditional random field. *ISPRS International Journal of Geo-Information*, 6.
- Suparta, W. & Rahman, R. 2016. Spatial interpolation of GPS PWV and meteorological variables over the west coast of Peninsular Malaysia during 2013 Klang Valley Flash Flood. *Atmospheric Research*, 168, 205-219.
- Suparta, W., Rahman, R. & Singh, M. S. J. Monitoring the variability of precipitable water vapor over the Klang Valley, Malaysia during flash flood. IOP Conference Series: Earth and Environmental Science, 2014.
- Supian, M. N. a. A., Razak, F. A. & Bakar, S. A. Twitter communication during 2014 flood in Malaysia: Informational or emotional? AIP Conference Proceedings, 2017.
- Suriya, S. & Mudgal, B. V. 2012. Impact of urbanization on flooding: The Thirusoolam sub watershed – A case study. *Journal of Hydrology*, 412-413, 210-219.

- Syed Hussain, T. P. R. & Ismail, H. 2013. Flood frequency analysis of Kelantan River Basin, Malaysia. *World Applied Sciences Journal*, 28, 1989-1995.
- Tadjudin, S. & Landgrebe, D. A. Covariance estimation for limited training samples. IGARSS '98. Sensing and Managing the Environment. 1998 IEEE International Geoscience and Remote Sensing. Symposium Proceedings. (Cat. No.98CH36174), 6-10 July 1998 1998. 2688-2690 vol.5.
- Taherzadeh, E., Shafri, H. Z. & Shahi, K. 2014. Roof material detection based on object-based approach using worldview-2 satellite imagery. *World Academy of Science, Engineering and Technology, International Journal of Computer, Electrical, Automation, Control and Information Engineering*, 8, 1737-1740.
- Tam, T. H. 2014. *Flood Risk Mapping using Geospatial Technologies and Hydrodynamic Modelling in Kota Tinggi, Johor*. Master Thesis, Universiti Teknologi Malaysia.
- Tam, T. H., Abd Rahman, M. Z., Harun, S. & Kaoje, I. U. 2018. Mapping of Highly Heterogeneous Urban Structure Type for Flood Vulnerability Assessment. *Int. Arch. Photogramm. Remote Sens. Spatial Inf. Sci.*, XLII-4/W9, 229-235.
- Tam, T. H., Ibrahim, A. L., Rahman, M. Z. A. & Mazura, Z. 2014. Flood loss assessment in the Kota Tinggi. *IOP Conference Series: Earth and Environmental Science*, 18, 012120.
- Tan, M., Ibrahim, A., Duan, Z., Cracknell, A. & Chaplot, V. 2015. Evaluation of Six High-Resolution Satellite and Ground-Based Precipitation Products over Malaysia. *Remote Sensing*, 7, 1504.
- Tan, M., Samat, N., Chan, N. & Roy, R. 2018a. Hydro-Meteorological Assessment of Three GPM Satellite Precipitation Products in the Kelantan River Basin, Malaysia. *Remote Sensing*, 10, 1011.
- Tan, M. L., Gassman, P. W. & Cracknell, A. P. 2017. Assessment of Three Long-Term Gridded Climate Products for Hydro-Climatic Simulations in Tropical River Basins. *Water*, 9, 229.
- Tan, M. L. & Santo, H. 2018b. Comparison of GPM IMERG, TMPA 3B42 and PERSIANN-CDR satellite precipitation products over Malaysia. *Atmospheric Research*, 202, 63-76.
- Tang, G., Ma, Y., Long, D., Zhong, L. & Hong, Y. 2016. Evaluation of GPM Day-1 IMERG and TMPA Version-7 legacy products over Mainland China at multiple spatiotemporal scales. *Journal of Hydrology*, 533, 152-167.
- Tang, G., Zeng, Z., Ma, M., Liu, R., Wen, Y. & Hong, Y. 2017. Can Near-Real-Time Satellite Precipitation Products Capture Rainstorms and Guide Flood Warning for the 2016 Summer in South China? *IEEE Geoscience and Remote Sensing Letters*, 14, 1208-1212.
- Tang, J., Alelyani, S. & Liu, H. 2014. Feature selection for classification: A review. *Data classification: Algorithms and applications*, 37.
- Tangang, F. T., Juneng, L., Salimun, E., Vinayachandran, P. N., Seng, Y. K., Reason, C. J. C., Behera, S. K. & Yasunari, T. 2008. On the roles of the northeast cold surge, the Borneo vortex, the Madden-Julian Oscillation, and the Indian Ocean Dipole during the extreme 2006/2007 flood in southern Peninsular Malaysia. *Geophysical Research Letters*, 35.
- Tapsell, S. M., Penning-Rowsell, E. C., Tunstall, S. M. & Wilson, T. L. 2002. Vulnerability to flooding: health and social dimensions. *Philosophical transactions of the royal society of London. Series A: Mathematical, Physical and Engineering Sciences*, 360, 1511-1525.

- Tarabalka, Y., Chanussot, J. & Benediktsson, J. A. 2010. Segmentation and classification of hyperspectral images using watershed transformation. *Pattern Recognition*, 43, 2367-2379.
- Tateishi, R., Hoan, N. T., Kobayashi, T., Alsaadeh, B., Tana, G. & Phong, D. X. 2014. Production of global land cover data-GLCNMO2008. *Journal of Geography and Geology*, 6, 99.
- Tateishi, R., Uriyangqai, B., Al-Bilbisi, H., Ghar, M. A., Tsend-Ayush, J., Kobayashi, T., Kasimu, A., Hoan, N. T., Shalaby, A. & Alsaadeh, B. 2011. Production of global land cover data-GLCNMO. *International Journal of Digital Earth*, 4, 22-49.
- Taubenböck, H., Esch, T., Wurm, M., Roth, A. & Dech, S. 2010. Object-based feature extraction using high spatial resolution satellite data of urban areas. *Journal of Spatial Science*, 55, 117-132.
- Taubenböck, H., Post, J., Roth, A., Zosseder, K., Strunz, G. & Dech, S. 2008. A conceptual vulnerability and risk framework as outline to identify capabilities of remote sensing. *Nat. Hazards Earth Syst. Sci.*, 8, 409-420.
- Taubenböck, H., Roth, A., Dech, S., Mehl, H., München, J., Stempniewski, L. & Zschau, J. 2009. Assessing building vulnerability using synergistically remote sensing and civil engineering. *Urban and regional data management. Taylor & Francis Group, London*, 287-300.
- Taubenböck, H., Wurm, M., Netzband, M., Zwenzner, H., Roth, A., Rahman, A. & Dech, S. 2011. Flood risks in urbanized areas – multi-sensoral approaches using remotely sensed data for risk assessment. *Natural Hazards and Earth System Science*, 11, 431-444.
- Tehrany, M. S., Pradhan, B. & Jebur, M. N. 2013. Spatial prediction of flood susceptible areas using rule based decision tree (DT) and a novel ensemble bivariate and multivariate statistical models in GIS. *Journal of Hydrology*, 504, 69-79.
- Tehrany, M. S., Pradhan, B. & Jebur, M. N. 2014. Flood susceptibility mapping using a novel ensemble weights-of-evidence and support vector machine models in GIS. *Journal of Hydrology*, 512, 332-343.
- Tehrany, M. S., Pradhan, B. & Jebur, M. N. 2015a. Flood susceptibility analysis and its verification using a novel ensemble support vector machine and frequency ratio method. *Stochastic Environmental Research and Risk Assessment*, 29, 1149-1165.
- Tehrany, M. S., Pradhan, B., Mansor, S. & Ahmad, N. 2015b. Flood susceptibility assessment using GIS-based support vector machine model with different kernel types. *CATENA*, 125, 91-101.
- Tekeli, A. E. & Fouli, H. 2016. Evaluation of TRMM satellite-based precipitation indexes for flood forecasting over Riyadh City, Saudi Arabia. *Journal of Hydrology*, 541, 471-479.
- Teng, J., Jakeman, A. J., Vaze, J., Croke, B. F. W., Dutta, D. & Kim, S. 2017. Flood inundation modelling: A review of methods, recent advances and uncertainty analysis. *Environmental Modelling & Software*, 90, 201-216.
- Tesfagiorgis, K., Mahani, S. E., Krakauer, N. Y. & Khanbilvardi, R. 2011. Bias correction of satellite rainfall estimates using a radar-gauge product – a case study in Oklahoma (USA). *Hydrol. Earth Syst. Sci.*, 15, 2631-2647.
- Thakur, P., Maiti, S., Kingma, N., Hari Prasad, V., Aggarwal, S. P. & Bhardwaj, A. 2012. Estimation of structural vulnerability for flooding using geospatial tools in the rural area of Orissa, India. *Natural Hazards*, 61, 501-520.

- Thanh Noi, P. & Kappas, M. 2018. Comparison of Random Forest, k-Nearest Neighbor, and Support Vector Machine Classifiers for Land Cover Classification Using Sentinel-2 Imagery. *Sensors*, 18, 18.
- Thieken, A., Olschewski, A., Kreibich, H., Kobsch, S. & Merz, B. 2008. Development and evaluation of FLEMOps—a new Flood Loss Estimation Model for the private sector. *WIT Transactions on Ecology and the Environment*, 118, 315-324.
- Thieken, A. H., Müller, M., Kreibich, H. & Merz, B. 2005. Flood damage and influencing factors: New insights from the August 2002 flood in Germany. *Water Resources Research*, 41.
- Thomlinson, J. R., Bolstad, P. V. & Cohen, W. B. 1999. Coordinating Methodologies for Scaling Landcover Classifications from Site-Specific to Global: Steps toward Validating Global Map Products. *Remote Sensing of Environment*, 70, 16-28.
- Thouret, J.-C., Ettinger, S., Guitton, M., Santoni, O., Magill, C., Martelli, K., Zuccaro, G., Revilla, V., Charca, J. A. & Arguedas, A. 2014. Assessing physical vulnerability in large cities exposed to flash floods and debris flows: the case of Arequipa (Peru). *Natural Hazards*, 73, 1771-1815.
- Tingsanchali, T. & Karim, M. F. 2005. Flood hazard and risk analysis in the southwest region of Bangladesh. *Hydrological Processes*, 19, 2055-2069.
- Tobler, W. R. 1970. A computer movie simulating urban growth in the Detroit region. *Economic geography*, 46, 234-240.
- Tokarczyk, P. 2013. Beyond hand-crafted features in remote sensing.
- Tokarczyk, P., Wegner, J. D., Walk, S. & Schindler, K. 2015. Features, Color Spaces, and Boosting: New Insights on Semantic Classification of Remote Sensing Images. *IEEE Transactions on Geoscience and Remote Sensing*, 53, 280-295.
- Tonbul, H. & Kavzaoğlu, T. Application of Taguchi Optimization and ANOVA Statistics in Optimal Parameter Setting of Multi-Resolution Segmentation. 2019 9th International Conference on Recent Advances in Space Technologies (RAST), 11-14 June 2019. 387-391.
- Tong, X., Xie, H. & Weng, Q. 2014. Urban Land Cover Classification With Airborne Hyperspectral Data: What Features to Use? *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 7, 3998-4009.
- Toriman, M. E., Jaafar, O., Maulud, K. N. A., Sharifah Mastura, S. A., Aziz, N. a. A., Gasim, M. B., Er, A. C., Jali, M. F. M., Jamil, N. R. & Abdullah, M. P. 2011. Modeling flood inundation in river catchment using hydraulic and geographical information system (GIS) simulation approach. *Journal of Engineering and Applied Sciences*, 6, 428-432.
- Tralli, D. M., Blom, R. G., Zlotnicki, V., Donnellan, A. & Evans, D. L. 2005. Satellite remote sensing of earthquake, volcano, flood, landslide and coastal inundation hazards. *ISPRS Journal of Photogrammetry and Remote Sensing*, 59, 185-198.
- Tran, T. V., Julian, J. P. & De Beurs, K. M. 2014. Land Cover Heterogeneity Effects on Sub-Pixel and Per-Pixel Classifications. *ISPRS International Journal of Geo-Information*, 3, 540-553.
- Try, S., Lee, G., Yu, W., Oeurng, C. & Jang, C. 2018. Large-Scale Flood-Inundation Modeling in the Mekong River Basin. *Journal of Hydrologic Engineering*, 23, 05018011.
- Udin, W. S., Ismail, N. a. B., Bahar, A. M. A. & Khan, M. M. A. 2018a. GIS-based River Flood Hazard Mapping in Rural Area: A Case Study in Dabong,

- Kelantan, Peninsular Malaysia. *Asian Journal of Water, Environment and Pollution*, 15, 47-55.
- Udin, W. S. & Malek, N. A. Flood risk analysis in Sg. Sam, Kuala Krai, Kelantan using remote sensing and GIS technique. IOP Conference Series: Earth and Environmental Science, 2018b.
- Uhu 2004. Two billion vulnerable to floods by 2050; number expected to double or more in two generations.
- Undrr. 2015. *Terminology* [Online]. Geneva: United Nations International Strategy for Disaster Reduction (UNISDR) Available: <https://www.unisdr.org/we/inform/terminology> [Accessed].
- Unesco-Ihe. 2019. *Flood Vulnerability Indices* [Online]. Available: <http://unihefvi.free.fr/vulnerability.php> [Accessed 21 April 2019].
- Unisdr 2009a. 2009 UNISDR Terminology on Disaster Risk Reduction. Geneva, Switzerland: United Nations International Strategy for Disaster Reduction (UNISDR).
- Unisdr 2009b. UNISDR Terminology on Disaster Risk Reduction. United Nations International Strategy for Disaster Reduction Geneva.
- Unisdr 2011. Global assessment report on disaster risk reduction: Revealing risk, redefining development. Geneva: United Nations Geneva.
- Unisdr 2015. The human cost of weather-related disasters, 1995–2015. *United Nations, Geneva*.
- Unisdr 2017. Words into Action guidelines: National disaster risk assessment.
- Ushio, T., Sasashige, K., Kubota, T., Shige, S., Okamoto, K. I., Aonashi, K., Inoue, T., Takahashi, N., Iguchi, T., Kachi, M., Oki, R., Morimoto, T. & Kawasaki, Z.-I. 2009. A Kalman Filter Approach to the Global Satellite Mapping of Precipitation (GSMaP) from Combined Passive Microwave and Infrared Radiometric Data. *Journal of the Meteorological Society of Japan. Ser. II*, 87A, 137-151.
- Usman Kaoje, I. 2021. *Geospatial Physical Flood Vulnerability Assessment Using Indicator-Based Approach*. PhD Thesis, Universiti Teknologi Malaysia.
- Uwakwe, A. C. 2015. *Assesment of Physical Vulnerability to Flood in Saint Lucia: Case Studies: Castries Old Central Business District and Dennery Village*, University of Twente Faculty of Geo-Information and Earth Observation (ITC).
- Valdés-Pineda, R., Demaría, E. M. C., Valdés, J. B., Wi, S. & Serrat-Capdevilla, A. 2016. Bias correction of daily satellite-based rainfall estimates for hydrologic forecasting in the Upper Zambezi, Africa. *Hydrol. Earth Syst. Sci. Discuss.*, 2016, 1-28.
- Van Der Sande, C., De Jong, S. & De Roo, A. 2003. A segmentation and classification approach of IKONOS-2 imagery for land cover mapping to assist flood risk and flood damage assessment. *International Journal of Applied Earth Observation and Geoinformation*, 4, 217-229.
- Vapnik, V. 1995. *The Nature of Statistical Learning Theory*, New York, Springer-Verlag.
- Varikoden, H., Samah, A. A. & Babu, C. A. 2010. Spatial and temporal characteristics of rain intensity in the peninsular Malaysia using TRMM rain rate. *Journal of Hydrology*, 387, 312-319.
- Vetter, T., Huang, S., Aich, V., Yang, T., Wang, X., Krysanova, V. & Hattermann, F. 2015. Multi-model climate impact assessment and intercomparison for three large-scale river basins on three continents. *Earth Syst. Dynam.*, 6, 17-43.



- Vijouyeh, H. G. & Taşkın, G. A comprehensive evaluation of feature selection algorithms in hyperspectral image classification. 2016 IEEE International Geoscience and Remote Sensing Symposium (IGARSS), 10-15 July 2016 2016. 489-492.
- Vinet, F., Saidi, M. E. M., Douvinet, J., Fehri, N., Nasrallah, W., Menad, W. & Mellas, S. 2016. Urbanization and land use as a driver of flood risk. *In: (FRANCE), A. N. D. R. P. L. E. (ed.) The Mediterranean Region under Climate Change: A Scientific Update*. Marseille.
- Voigt, S., Giulio-Tonolo, F., Lyons, J., Kučera, J., Jones, B., Schneiderhan, T., Platzeck, G., Kaku, K., Hazarika, M. K., Czarán, L., Li, S., Pedersen, W., James, G. K., Proy, C., Muthike, D. M., Bequignon, J. & Guha-Sapir, D. 2016. Global trends in satellite-based emergency mapping. *Science*, 353, 247-252.
- Vojinovic, Z., Hammond, M., Golub, D., Hirunsalee, S., Weesakul, S., Meesuk, V., Medina, N., Sanchez, A., Kumara, S. & Abbott, M. 2016. Holistic approach to flood risk assessment in areas with cultural heritage: a practical application in Ayutthaya, Thailand. *Natural Hazards*, 81, 589-616.
- Volpi, M., Tuia, D., Bovolo, F., Kanevski, M. & Bruzzone, L. 2013. Supervised change detection in VHR images using contextual information and support vector machines. *International Journal of Applied Earth Observation and Geoinformation*, 20, 77-85.
- Voltersen, M., Berger, C., Hese, S. & Schmillius, C. 2014. Object-based land cover mapping and comprehensive feature calculation for an automated derivation of urban structure types at block level. *Remote Sensing of Environment*, 154, 192-201.
- Vu, T. T. & Ranzi, R. 2017. Flood risk assessment and coping capacity of floods in central Vietnam. *Journal of Hydro-Environment Research*, 14, 44-60.
- Wahab, A. M. & Muhamad Ludin, A. N. Flood vulnerability assessment using artificial neural networks in Muar Region, Johor Malaysia. IOP Conference Series: Earth and Environmental Science, 2018.
- Wan, K. M. & Billa, L. 2018. Post-flood land use damage estimation using improved Normalized Difference Flood Index (NDFI3) on Landsat 8 datasets: December 2014 floods, Kelantan, Malaysia. *Arabian Journal of Geosciences*, 11.
- Wang, J., Xu, Y., Yang, L., Wang, Q., Yuan, J. & Wang, Y. 2020. Data Assimilation of High-Resolution Satellite Rainfall Product Improves Rainfall Simulation Associated with Landfalling Tropical Cyclones in the Yangtze River Delta. *Remote Sensing*, 12, 276.
- Wang, M. & Li, R. 2014. Segmentation of High Spatial Resolution Remote Sensing Imagery Based on Hard-Boundary Constraint and Two-Stage Merging. *IEEE Transactions on Geoscience and Remote Sensing*, 52, 5712-5725.
- Wang, M., Yuan, S., Pan, J., Fang, L., Zhou, Q. & Yang, G. 2016. Seamline Determination for High Resolution Orthoimage Mosaicking Using Watershed Segmentation. *Photogrammetric Engineering & Remote Sensing*, 82, 121-133.
- Ward, P. J., Van Pelt, S. C., De Keizer, O., Aerts, J., Beersma, J. J., Van Den Hurk, B. & Te Linde, A. H. 2014. Including climate change projections in probabilistic flood risk assessment. *Journal of Flood Risk Management*, 7, 141-151.
- Wardah, T., Abu Bakar, S. H., Bardossy, A. & Maznorizan, M. 2008. Use of geostationary meteorological satellite images in convective rain estimation for flash-flood forecasting. *Journal of Hydrology*, 356, 283-298.
- Wardah, T., Suzana, R., Huda, S. Y. S. N. & Kamil, A. A. Multi-sensor data inputs rainfall estimation for flood simulation and forecasting. CHUSER 2012 - 2012

- IEEE Colloquium on Humanities, Science and Engineering Research, 2012. 374-379.
- Waske, B., Fauvel, M., Benediktsson, J. A. & Chanussot, J. 2009. Machine learning techniques in remote sensing data analysis. *Kernel methods for remote sensing data analysis*, 3-24.
- Waske, B., Van Der Linden, S., Benediktsson, J. A., Rabe, A. & Hostert, P. 2010. Sensitivity of Support Vector Machines to Random Feature Selection in Classification of Hyperspectral Data. *IEEE Transactions on Geoscience and Remote Sensing*, 48, 2880-2889.
- Watanachaturaporn, P., Arora, M. K. & Varshney, P. K. Sub-pixel land cover classification using support vector machines. *Proc. Annu. Conf. Amer. Soc. Photogramm. Remote Sens.*, 2006. 1575-1584.
- Wei, C.-P., Piramuthu, S. & Shaw, M. J. 2003. Knowledge discovery and data mining. *Handbook on Knowledge Management*. Springer.
- White, P., Pelling, M., Sen, K., Seddon, D., Russell, S. & Few, R. 2005. Disaster risk reduction: a development concern. *London: DfID*.
- Wieland, M., Pittore, M., Parolai, S., Zschau, J., Moldobekov, B. & Begaliev, U. 2012. Estimating building inventory for rapid seismic vulnerability assessment: Towards an integrated approach based on multi-source imaging. *Soil Dynamics and Earthquake Engineering*, 36, 70-83.
- Wong, S. N. & Sarker, M. L. R. 2014. Land use/land cover mapping using multi-scale texture processing of high resolution data. *IOP Conference Series: Earth and Environmental Science*, 18, 012185.
- Woodcock, C. E., Strahler, A. H. & Jupp, D. L. B. 1988. The use of variograms in remote sensing: I. Scene models and simulated images. *Remote Sensing of Environment*, 25, 323-348.
- Woodley, A., Chappell, T., Geva, S. & Nayak, R. Efficient Feature Selection and Nearest Neighbour Search for Hyperspectral Image Classification. 2016 International Conference on Digital Image Computing: Techniques and Applications (DICTA), 30 Nov.-2 Dec. 2016 2016. 1-8.
- Woolhiser, D., Smith, R. & Goodrich, D. 1990. A kinematic runoff and erosion model: documentation and user manual, ARS 77. *US Department of Agriculture*.
- Wu, H., Cheng, Z., Shi, W., Miao, Z. & Xu, C. 2014. An object-based image analysis for building seismic vulnerability assessment using high-resolution remote sensing imagery. *Natural Hazards*, 71, 151-174.
- Wu, S.-S., Xu, B. & Wang, L. 2006. Urban land-use classification using variogram-based analysis with an aerial photograph. *Photogrammetric Engineering & Remote Sensing*, 72, 813-822.
- Wu, S. S., Qiu, X., Usery, E. L. & Wang, L. 2009. Using geometrical, textural, and contextual information of land parcels for classification of detailed urban land use. *Annals of the Association of American Geographers*, 99, 76-98.
- Wu, X., Peng, J., Shan, J. & Cui, W. 2015. Evaluation of semivariogram features for object-based image classification. *Geo-spatial Information Science*, 18, 159-170.
- Wu, Y., Ke, Y., Gong, H., Chen, B. & Zhu, L. Comparison of object-based and pixel-based methods for urban land-use classification from WorldView-2 imagery. 2014 Third International Workshop on Earth Observation and Remote Sensing Applications (EORSA), 11-14 June 2014 2014. 284-288.

- Xie, P. & Arkin, P. A. 1997. Global Precipitation: A 17-Year Monthly Analysis Based on Gauge Observations, Satellite Estimates, and Numerical Model Outputs. *Bulletin of the American Meteorological Society*, 78, 2539-2558.
- Xie, P., Janowiak, J. E., Arkin, P. A., Adler, R., Gruber, A., Ferraro, R., Huffman, G. J. & Curtis, S. 2003. GPCP pentad precipitation analyses: An experimental dataset based on gauge observations and satellite estimates. *Journal of Climate*, 16, 2197-2214.
- Xu, D., Chen, B., Shen, B., Wang, X., Yan, Y., Xu, L. & Xin, X. 2019. The Classification of Grassland Types Based on Object-Based Image Analysis with Multisource Data. *Rangeland Ecology & Management*, 72, 318-326.
- Xu, R., Tian, F., Yang, L., Hu, H., Lu, H. & Hou, A. 2017. Ground validation of GPM IMERG and TRMM 3B42V7 rainfall products over southern Tibetan Plateau based on a high-density rain gauge network. *Journal of Geophysical Research: Atmospheres*, 122, 910-924.
- Xu, X., Xu, S., Jin, L. & Song, E. 2011. Characteristic analysis of Otsu threshold and its applications. *Pattern Recognition Letters*, 32, 956-961.
- Ya'acob, N., Ismail, N. S., Mustafa, N. & Yusof, A. L. Investigation of flood pattern using ANOVA statistic and remote sensing in Malaysia. IOP Conference Series: Earth and Environmental Science, 2014.
- Yahaya, N. S., Lim, C.-S., Jamaluddin, U. A. & Pereira, J. J. 2015. The December 2014 Flood in Kelantan: A Post-Event Perspective. *Warta Geologi*, 41, 54-57.
- Yahya, N., Thiruchelvam, S., Ghazali, A., Muda, R. S., Mat Isa, A. A., Jin, N. Y., Norkhairi, F. F., Hakimie, H., Kadir, A. K., Mohamed Sahari, K. S., Hasini, H., Mustapha, K. N., Muda, Z. C. & Mamat, A. F. 2018. Building human resilience: The role of community based training and awareness programme (CBTAP) for dam related flood risk management. *ASM Science Journal*, 11, 201-211.
- Yan, G., Mas, J. F., Maathuis, B. H. P., Xiangmin, Z. & Van Dijk, P. M. 2006. Comparison of pixel - based and object - oriented image classification approaches—a case study in a coal fire area, Wuda, Inner Mongolia, China. *International Journal of Remote Sensing*, 27, 4039-4055.
- Yang, X. 2011. Parameterizing Support Vector Machines for Land Cover Classification. *Photogrammetric Engineering & Remote Sensing*, 77, 27-37.
- Yeganeh, N. & Sabri, S. 2014. Flood vulnerability assessment in Iskandar Malaysia using multi-criteria evaluation and fuzzy logic. *Research Journal of Applied Sciences, Engineering and Technology*, 8, 1794-1806.
- Yi, L., Zhang, W. & Li, X. 2018. Assessing Hydrological Modelling Driven by Different Precipitation Datasets via the SMAP Soil Moisture Product and Gauged Streamflow Data. *Remote Sensing*, 10, 1872.
- Yoshimoto, S. & Amarnath, G. 2017. Applications of Satellite-Based Rainfall Estimates in Flood Inundation Modeling—A Case Study in Mundeni Aru River Basin, Sri Lanka. *Remote Sensing*, 9, 998.
- Yu, Q., Gong, P., Clinton, N., Biging, G., Kelly, M. & Schirokauer, D. 2006. Object-based Detailed Vegetation Classification with Airborne High Spatial Resolution Remote Sensing Imagery. *Photogrammetric Engineering & Remote Sensing*, 72, 799-811.
- Yuan, F., Zhang, L., Win, K. W. W., Ren, L., Zhao, C., Zhu, Y., Jiang, S. & Liu, Y. 2017. Assessment of GPM and TRMM Multi-Satellite Precipitation Products in Streamflow Simulations in a Data-Sparse Mountainous Watershed in Myanmar. *Remote Sensing*, 9, 302.

- Yue, A., Zhang, C., Yang, J., Su, W., Yun, W. & Zhu, D. 2013. Texture extraction for object-oriented classification of high spatial resolution remotely sensed images using a semivariogram. *International Journal of Remote Sensing*, 34, 3736-3759.
- Yusoff, A., Din, N. M., Yusoff, S. & Khan, S. U. Big data analytics for Flood Information Management in Kelantan, Malaysia. 2015 IEEE Student Conference on Research and Development, SCORED 2015, 2015. 311-316.
- Yusoff, I. M., Ramli, A., Alkasirah, N. a. M. & Nasir, N. M. 2018. Exploring the managing of flood disaster: A Malaysian perspective. *Geografica: Malaysian Journal of Society and Space*, 14.
- Zahari, N. Z. & Hashim, A. M. Adequacy of Flood Relief Shelters: A Case Study in Perak, Malaysia. E3S Web of Conferences, 2018.
- Zaitoun, N. M. & Aqel, M. J. 2015. Survey on Image Segmentation Techniques. *Procedia Computer Science*, 65, 797-806.
- Zakaria, S. F., Zin, R. M., Mohamad, I., Balubaid, S., Mydin, S. H. & Mdr, E. R. The development of flood map in Malaysia. AIP Conference Proceedings, 2017. AIP Publishing, 110006.
- Zeng, J., Li, Z., Chen, Q., Bi, H., Qiu, J. & Zou, P. 2015. Evaluation of remotely sensed and reanalysis soil moisture products over the Tibetan Plateau using in-situ observations. *Remote Sensing of Environment*, 163, 91-110.
- Zhan, W., Pan, M., Wanders, N. & Wood, E. F. 2015. Correction of real-time satellite precipitation with satellite soil moisture observations. *Hydrol. Earth Syst. Sci.*, 19, 4275-4291.
- Zhang, D.-D., Zhang, L., Zaborovsky, V., Xie, F., Wu, Y.-W. & Lu, T.-T. 2019a. Research on the pixel-based and object-oriented methods of urban feature extraction with GF-2 remote-sensing images. *arXiv preprint arXiv:1903.03412*.
- Zhang, L., Zhang, Q., Du, B., Huang, X., Tang, Y. Y. & Tao, D. 2016. Simultaneous Spectral-Spatial Feature Selection and Extraction for Hyperspectral Images. *IEEE Transactions on Cybernetics*, 48, 16-28.
- Zhang, S. L. & Chang, T. C. 2015. A Study of Image Classification of Remote Sensing Based on Back-Propagation Neural Network with Extended Delta Bar Delta. *Mathematical Problems in Engineering*, 2015, 178598.
- Zhang, Z., Wang, D., Wang, G., Qiu, J. & Liao, W. 2019b. Use of SMAP Soil Moisture and Fitting Methods in Improving GPM Estimation in Near Real Time. *Remote Sensing*, 11, 368.
- Zhao, W., Du, S. & Emery, W.J. 2017. Object-Based Convolutional Neural Network for High-Resolution Imagery Classification. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 10, 3386-3396.
- Zheng, W., Zhu, X., Wen, G., Zhu, Y., Yu, H. & Gan, J. 2018. Unsupervised feature selection by self-paced learning regularization. *Pattern Recognition Letters*.
- Zheng, X., Wang, Y., Gan, M., Zhang, J., Teng, L., Wang, K., Shen, Z. & Zhang, L. 2016. Discrimination of Settlement and Industrial Area Using Landscape Metrics in Rural Region. *Remote Sensing*, 8, 845.
- Zheng, X., Wu, B., Weston, M. V., Zhang, J., Gan, M., Zhu, J., Deng, J., Wang, K. & Teng, L. 2017. Rural Settlement Subdivision by Using Landscape Metrics as Spatial Contextual Information. *Remote Sensing*, 9, 486.
- Zhou, C., Gao, W., Hu, J., Du, L. & Du, L. 2021. Capability of IMERG V6 Early, Late, and Final Precipitation Products for Monitoring Extreme Precipitation Events. *Remote Sensing*, 13, 689.

- Zhou, L., Rasmy, M., Takeuchi, K., Koike, T., Selvarajah, H. & Ao, T. 2021. Adequacy of Near Real-Time Satellite Precipitation Products in Driving Flood Discharge Simulation in the Fuji River Basin, Japan. *Applied Sciences*, 11, 1087.
- Zhou, W., Huang, G., Troy, A. & Cadenasso, M. L. 2009. Object-based land cover classification of shaded areas in high spatial resolution imagery of urban areas: A comparison study. *Remote Sensing of Environment*, 113, 1769-1777.
- Zhou, Y., Zhang, R., Wang, S. & Wang, F. 2018. Feature Selection Method Based on High-Resolution Remote Sensing Images and the Effect of Sensitive Features on Classification Accuracy. *Sensors*, 18, 2013.
- Zhu, H., Cai, L., Liu, H. & Huang, W. 2016. Information Extraction of High Resolution Remote Sensing Images Based on the Calculation of Optimal Segmentation Parameters. *PLOS ONE*, 11, e0158585.
- Zhu, J. & Tang, C. Urban flood damage estimation using GIS and remote sensing. *Advanced Computer Theory and Engineering (ICACTE)*, 2010 3rd International Conference on, 20-22 Aug. 2010. V2-232-V2-237.
- Ziaei, Z., Pradhan, B. & Mansor, S. B. 2014. A rule-based parameter aided with object-based classification approach for extraction of building and roads from WorldView-2 images. *Geocarto International*, 29, 554-569.
- Zin, W. W., Kawasaki, A., Takeuchi, W., San, Z. M. L. T., Htun, K. Z., Aye, T. H. & Win, S. 2018. Flood Hazard Assessment of Bago River Basin, Myanmar. *Journal of Disaster Research*, 13, 14-21.