Subject Description Form

Subject Code	LSGI1001
Subject Title	Big Earth Data and Geospatial AI
Credit Value	2
Level	1
Pre-requisite/ Co-requisite/ Exclusion	Nil
Objectives and Development Opportunities	Artificial Intelligence and Data Analytics (AIDA) are crucial tools in the world today, revolutionizing the paradigm of various fields, such as business, public health, environment, science, technology, social science, etc. They provide beneficial methodologies to automatically process the huge volume and variety of data, access the essential information, and facilitate decision-making in different aspects of life. Big Earth Data, big data collected from space for Earth observations,
	provides new opportunities for Earth and environmental sciences, revolutionizing the methods and techniques for understanding of the Earth's systems and dynamics, human-environment interactions, and sustainability of nature and human systems. Geospatial analytics uses data from all kinds of technology — satellite imagery, GPS, location sensors, social media, mobile devices — to analyse and visualize the geo- referenced data for understanding phenomena and finding patterns and trends in complex relationships between people and places. Geospatial analytics adds location and timing to traditional types of data, and this additional context allows for a more complete picture of phenomena and events, a more accurate prediction of patterns and trends, and real-time visualization.
	This course provides a general education on various issues in Earth Observations and Geospatial Artificial Intelligence (EO-GeoAI). Identification and evaluation of Earth features from images/data acquired from space are explored. Aerial photos, Earth resources satellite images, weather satellite images, LiDAR, drone images, social media, mobile devices etc. are used to give insight into important physical, economic, and cultural features on the Earth surface. Rudimentary knowledge on machine learning in remote sensing, geospatial AI and geospatial big data analytics, will further be introduced.
Intended Learning Outcomes	 Upon completion of the subject, students will be able to: (i) Demonstrate an understanding of the foundational concepts of Artificial Intelligence and Data Analytics (AIDA); (ii) Acquire basic skills in using AIDA technologies and applications; (iii) Articulate examples of how the adoption AIDA could enhance their chosen disciplines; (iv) Demonstrate an awareness of global contemporary ethical issues and impact from AIDA applications in daily life; (v) Appreciate different types of Earth observation images/data; (vi) Acquire basic principles of Earth observations; (vii) Show basic understanding of machine learning for remote sensing, geospatial AI and big data analytics

Subject Synopsis/	Topics:							
Indicative Syllabus	1. Earth C The def sensors analytic of using	Dbservations, Enition of rer , big Earth da es and geospa g big Earth da	s, Big Earth Data, and Geospatial Analytics emote sensing, Earth observations, remote data, and digital image processing; geospatial patial imagery analytics; benefits and examples data in analytics.					
	2. Types a Aerial j ground (e.g., G informa bands c	<i>Types and Characteristics of Big Earth Data</i> Aerial photographs; satellite images; drone images; space shuttles; ground-based sensing; geo-referenced information over the web (e.g., Google Earth and Virtual Globe); volunteered geographic information; characteristics of big Earth data (image; pixel; image bands or channels; spatial resolution; and temporal resolution).						
	3. Introdu to mach Applica Geospa image s and sce	ction to Machine learning ation example tial AI metho emantic segn ne recognitio	<i>hine Learning in Remote Sensing</i> Introduction for remote sensing image analysis. es of machine learning to Big Earth Data. ods in machine learning (e.g., deep learning), nentation, object detection, change monitoring, on.					
	4. Geospa Introdu architec perform Data.	<i>tial Artificial</i> ction to fram ctures, and ap nance/cloud c	<i>Intelligence</i> eworks for deep learning, classic network plications in image classification, and high- computing to extract knowledge from Big Earth					
	5. <i>Geospa</i> Basics data mi analysis	<i>tial Data And</i> of spatial data ning, pattern s, rule learnin	alytics a science, social sensing, people sensing, spatial recognition & matching, time series image ag for spatial data mining.					
Teaching/Learning	Teaching	Intended	Remarks					
Methodology	and Learning	Subject Learning						
	Methods	Outcome						
	Lectures	(i), (ii),	AIDA big Earth data, and geospatial					
		(111), (1V), (V), (V)	analytics concepts, principles, and skills will be given through lectures. There will be in-					
		(v), (vi), (vii)	class activities (e.g., discussions, debates,					
			videos) to better engage students					
	Lab	(ii) (v)	in active learning.					
	exercises	(ii), (v), (vii)	student skills through hands-on practices in					
			solving real-world problems with					
	geospatial analytics tasks.		geospatial analytics tasks. Ouizzes will be given to belp students better					
	Zuillos	(iii), (iv),	understand the points discussed in lectures					
		(v), (vi),	and labs, and to gain more related knowledge					
		(V11)	via after-class reading and reviewing and learn how to employ the knowledge to					
			solve geospatial big data problems in real					
			world.					

	E- (i learning Modules	ii), (vii)	Each s Webin (ILN) (https: , whice order a spece examp studer analyt daily l	stude nars i with //ww h rec to get cific t ole. T nts ca ics w life an	nt sho n the certif w.iee juires t certi echno Throug n lear vill he nd in t	buld w IEEE icates e.org// them ficates ologica gh the n how lp to s their c	atch a Learni educat to ans Each l topic E- le geosp olve p areer.	t least ing N tion/il swer o n Web c with earnin patial practio	t five etworl n.htm questic pinar c n real ag moo AI and cal issu	(l) pons in overs world dules, d data ues in	
Assessment Methods in Alignment with Intended Learning	Specific assessment methods/task	% weightingIntended subject learning ou to be assessed (Please tick as appropriate)						g outc k as	outcomes as		
Outcomes				(i)	(ii)	(iii)	(iv)	(v)	(vi)	(vii)	
	• Lab exercises	40%			\checkmark			\checkmark	\checkmark	\checkmark	
	• E-learning modules	g 20%				\checkmark				\checkmark	
	• Quiz	15%		\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	
	• Final test	25%	1	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	
	Total	100	%			1	1	1	1		
	For the continua- learning outcom five Webinars a AIDA, especial (i.e., learning of problems. In ad test for learning	ous assess nes (ii) (v) and submit ly the EO- utcome (iii dition, stu g outcomes	ments, 1 , (vi) ar certific GeoAI i) and ($^{\circ}$ dents w s (i) – (v	four 1 nd (vi cates appli vii)) a vill be vii).	ab ex i); the to sho ication and th e asses	ercises en eacl ow the ns, in t eir ski ssed by	s are fr n stude ir unde heir sj lls to s y one o	irst us ent sh erstan pecifi solve quiz a	sed to ould v dings c disc real-w and on	assess vatch of iplines vorld e final	
Student Study	Class contact (time-tabled):										
Enort Expected	• Lectures 2 hours in weeks 1-3, 5, 9, 11-12; and 1 hour in week 7.						,	15 Hrs.			
	 Labs Labs 2 hours x 4 lab exercises in weeks 4, 6, 8, and 10. Quiz and test Quiz: 1 hour in week 7. Final test: 2 hours in week 13. 					8 Hrs.					
						3 Hrs.					
	Other student study effort:										
	• E-Learning modules 2 hours x 5 learning modules in weeks 3, 5, 9, 11 and 12.				12.	10 Hrs.					
• Self-study and review								40 Hrs.			
	Total student study effort				76 Hrs.						

Reading List and	Reference Books:					
References						
Kererences	1. Beecher, K., Computational Thinking, BCS, 2017.					
	2. Teetor, P., <i>R Cookbook</i> , O'Reilly Media, 2011.					
	3. Wickham, H. and Grolemund, G., <i>R for Data Science</i> , O'Reilly Media, 2017.					
	4. Jake VanderPlas, <i>Python Data Science Handbook</i> , O'Reilly Media, Inc., 2016.					
	5. Peter Bruce and Andrew Bruce, <i>Practical Statistics for Data Scientists</i> , O'Reilly Media, 2017.					
	6. Boyd, S. and Vandenberghe, L., <i>Introduction to Applied Linear Algebra</i> , Cambridge University Press, 2018.					
	7. Stewart, J., <i>Calculus: Early Transcendentals</i> , 8th Edition, Cengage Learning, 2015.					
	8. Hastiem, T., Tibshirani, R., and Friedman, J., <i>The Elements of Statistical Learning: Data Mining, Inference, and Prediction</i> , 2nd Edition, Springer, 2009					
	9. Russell, S. and Norvig, P., <i>Artificial Intelligence: A Modern Approach</i> , 3rd Edition, Pearson, 2009.					
	10. Bishop, Christopher M., <i>Pattern Recognition and Machine Learning</i> . Springer, 2016.					
	11. Weng, Q., An Introduction to Contemporary Remote Sensing. McGraw-Hill Professional, 2012.					
	 Avery, T. E. and Berlin, G. L., Fundamentals of Remote Sensing and Airphoto Interpretation (fifth edition). Saddle River, N.J.: Prentice Hall, 1992. 					
	 Stefan Hinz (Editor), Martin Weinmann (Editor), Andreas Braun (Editor), Object and Pattern Recognition in Remote Sensing, Whittles Publishing, 2021. 					
	14. Martin Werner and Yao-Yi Chiang, <i>Handbook of Big Geospatial Data</i> , Springer, 2021.					

• Note: Prof. Weng is the Co-Lead of the IEEE GRSS Educational Webinar Series, who will arrange for students to watch five Webinars for free.

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