Subject Code	MM5433			
Subject Title	Decision Analytics by Machine Learning			
Credit Value	3			
Level	5			
Normal Duration	1-semester			
Pre-requisite/ Co-requisite/ Exclusion	None			
Objectives	<ol> <li>Introduce basic programming concepts in R-language.</li> <li>Introduce students to machine learning in decision-making context.</li> <li>Justify the use of machine learning in the workplace.</li> <li>Demonstrate pitfalls of machine learning.</li> </ol>			
Intended Learning Outcomes	Upon completion of the subject, students will be able to: a. Able to read/write/evaluate code written in R-language b. Demonstrate practical skills on simple predictive analytics c. Scrutinize insights based on predictive analytics d. Use machine learning as a routine tool for effective decisions			
Subject Synopsis/ Indicative Syllabus	This subject offers students a journey from basic data analytics to advanced machine learning concepts, using R and XGBoost. Each week, through a representative business example study, we uncover how data shape effective management and decision making. The subject gradually builds on R-programming and machine learning knowledge, giving students hands-on experience with R- assignments linked to the weekly topics. Basic understanding of statistics is required. Part I: Beginning programming in R - Art of writing programs - Algorithms vs code - R-syntax			
	<ul> <li>Part II: Fundamentals of data analytics</li> <li>Importance of data</li> <li>Big data</li> <li>The process of data collection</li> <li>The process of data cleaning</li> </ul>			
	Part III: Human behavior - Non-linear relationships - Missing responses - Biases - Choices and value estimates			
	<ul> <li>Part IV: Machine learning hiccups</li> <li>overfitting and underfitting</li> <li>Model explanations</li> <li>Text analysis</li> </ul>			

Teaching/Learning Methodology	<ul><li>39 hours of class activities including lectures on the main concepts and models, together with applicational case studies, tutorials, class/group problem discussions, and presenting pre-class analysis of their work. Weekly representative simple case-based problems connect programming exercises to workplace problems.</li><li>Weekly R programming assignments slowly build up expertise in predictive analytics. Students should be able to work on the assignments on their regular laptop from home.</li></ul>						
Assessment Methods in Alignment with Intended Learning Outcomes	Specific assessment methods/tasks	% weighting	Intended subject learning outcomes to be assessed (Please tick as appropriate)				
			а	b	c	d	
	Continuous Assessment*	100%					
	1. Class Participation	10%	~	~	~	~	
	2. Weekly assignments	60%	~	~	~		
	3. Final decision making writeup	30%	~	~	~	~	
	Total	100 %					
	*Weighting of assessment meth different, subject to each subje		ntinuous assessment maybe				
	To pass this subject, studen above in the overall subject	ss this subject, students are required to obtain Grade D o in the overall subject grade.					
	<ul> <li>Explanation of the appropriateness of the assessment met in assessing the intended learning outcomes:</li> <li>Class Participation: Students are required to attend class engage in discussions surrounding organizational issues and do on applicational case studies.</li> <li>Weekly Assignment: After-class assessment of the continuderstanding of the concepts, issues, models and application machine learning techniques by providing answers to g questions.</li> </ul>						
	Final decision making write-up: The writeup is a potential to the organizational problem using machine learning. Th is expected to be a senior executive of the firm and hence be presented in a simple form with charts for those execu understand and critique.					reader should	

Student Study Effort Expected	Class contact:			
	Seminars	39 Hrs.		
	Other student study effort:			
	<ul> <li>Preparation for lectures</li> </ul>	39 Hrs.		
	<ul> <li>Preparation for individual assignment / group project / class quiz</li> </ul>	44 Hrs.		
	Total student study effort	122 Hrs.		
Reading List and References	<ul> <li>Matloff, N. (2011). The Art of R Programmin Statistical Software Design. No Starch Press. F <u>https://www.amazon.com/Art-Programming-Statist</u> Design/dp/1593273843</li> <li>Kuhn, M., &amp; Johnson, K. (2013). Applied Predic Springer. Retrieved from https://www.amazor Predictive-Modeling-Max-Kuhn/dp/1461468485</li> <li>Molnar, C. (2022). Interpretable Machine Learning Making Black Box Models Explainable. Independe Retrieved from https://www.amazon.com/Interpret Learning-Making-Explainable/dp/B09TMWHVB4 available for free https://christophm.github.io/in book/)</li> <li>Lantz, B. (2023). Machine Learning with R: Learn building and improving machine learning mode preparation to model tuning, evaluation, and working 4th Edition. Packt Publishing. Retri https://www.amazon.com/Machine-Learning-cleans tidyverse/dp/1801071322</li> </ul>	Retrieved from stical-Software- ictive Modeling. on.com/Applied- ng: A Guide For lently published. retable-Machine- 4 (web book interpretable-ml- m techniques for dels, from data ng with big data, trieved from		

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