Subject Description Form

Subject Code	COMP4434			
Subject Title	Big Data Analytics			
Credit Value	3			
Level	4			
Pre-requisite / Co-requisite / Exclusion	Pre-requisites : AMA1104 or COMP1004, COMP1011 or COMP1012 or ENG2002, COMP2011 or COMP2013, COMP2411 or equivalent introductory database subject			
Objectives	The objectives of this subject are to:			
	1. introduce students the concept and challenge of big data (3 V's: volume, velocity, and variety); and			
	2. teach students in applying skills and tools to manage and analyse the big data.			
Intended	Upon completion of the subject, students will be able to:			
Learning Outcomes	 (a) understand the concept and challenge of big data and why existing technology is inadequate to analyse the big data; 			
	(b) understand how to collect, manage, store, and query various form of big data;			
	(c) understand how to analyse big data using various quantitative methods;			
	 (d) gain hands-on experience on large-scale analytics tools to solve some open big data problems; and 			
	(e) be able to conduct thorough analysis on the impact of big data for business decisions and strategy in real-world applications.			
Subject Synopsis/	Торіс			
Indicative	1. Introduction to Big Data			
Syllabus	Different V's, their challenges and application domains.			
	2. Collection of Big Data			
	Eventual Consistency and NoSQL systems (MongoDB, BigTable, etc.)			
	3. Large-Scale Data Analytics Systems			
	Hadoop, MapReduce, Hive, etc.			
	4. Basic Statistical Analysis			
	5. Machine Learning Systems for Big Data			
	6. Graph Analytics			
	Graph structures, PageRank, Centrality, etc.			

	7. Data Analysis Application: Recommender System						
	8. Data Visualisation						
Teaching/ Learning Methodology	A mix of lectures and lab sessions is used to deliver the various topics in this subjec Lectures are conducted to initiate students with the concepts and techniques of bi data. Students are given the opportunity to gain hands-on experience on both oper source and commercial big data analytics software during the laboratory sessions.						ues of big both open-
Assessment Methods in Alignment	Specific assessment methods/tasks	% weighting	Intended subject learning outcomes to be assessed				s to be
with Intended Learning			а	b	с	d	e
Outcomes	Continuous Assessment						
	1. Lab Exercises / Assignments	60%	\checkmark	~	\checkmark	~	~
	2. Project		\checkmark	~	\checkmark	\checkmark	~
	3. Quiz		\checkmark	~	\checkmark		
	Examination	40%	\checkmark	\checkmark	\checkmark		\checkmark
	Total	100 %					
	Explanation of the ap intended learning outco Continuous assessment which are designed to exercise is designed to relevant knowledge, pr software tools. The pr understanding and usin real problem through te Examination will evalue	omes: ts consist of a facilitate stud o encourage actice in orde oject is desig ng different kn eam. Quizzes a	a project, a lents to ac students r to enrich gned to en nowledge, are to ensu	assignmen chieve inte to acquire n their hand nhance stu , principle are the stud	ts, lab exe inded lear e deep ui ds-on exp idents' al s, techniq dents unde	ercises, an ming outco nderstandi perience wi bility to a pues, tools erstand the	d quizzes, omes. Lab ng of the ith various cquire the to solve a concepts.
Student Study	Class contact:						
Effort Expected	 Lectures 						26 Hrs.
	Tutorials/Laboratory 13 Hrs.						
	Other student study eff	ort:					
	• Review the lecture	;					28 Hrs.
	• Review the lab						14 Hrs.
	• Work on the project	ct					15 Hrs.

	 Prepare the quizzes 	9 Hrs.				
	Prepare the examination	11 Hrs.				
	Total student study effort	116 Hrs.				
Reading List	Reference Books:					
and References	1. Segaran, Toby, and Jeff Hammerbacher, <i>Beautiful data: the stories behind elegant data solution</i> , O'Reilly Media, Inc., 2009					
	2. Dean, Jeffrey and Ghemawat, Sanjay, "MapReduce: so on large clusters", <i>Communications of the ACM</i> , Janua					
	3. Stonebraker, M., Abadi, D., DeWitt, David J., Madde A. and Rasin, A., "MapReduce and Parallel DBM <i>Communications of the ACM</i> , January 2010.					
	4. Dean, Jeffrey and Ghemawat, Sanjay, "MapReduce: A Tool", <i>Communications of the ACM</i> , January 2010.	Flexible Data Processing				
	5. K. Shvachko, H. Kuang, S. Radia and R. Chansler, File System", <i>IEEE Symposium on Mass Storage Sy</i> 2010	-				
	6. White, Tom, <i>Hadoop: The definitive guide</i> , O'Reilly N	Media, Inc., 2012.				
	 Cattell, Rick, "Scalable SQL and NoSQL Data Stores" Volume 39, Issue 4, December 2010. 	", ACM SIGMOD Record,				
	 Page, Lawrence and Brin, Sergey and Motwani, Raje "The PageRank Citation Ranking: Bringing Order Report, Stanford InfoLab, 1999. 	•				
	9. Toby Segaran, <i>Programming Collective Intelligence</i> ,	O'Reilly Media, Inc., 2007				
	10. Han, Jiawei, Micheline Kamber, and Jian Pei, <i>Dat Techniques</i> , 3rd Edition, Morgan Kauffman, 2011.	ta Mining: Concepts and				
	11. Tan, Pang-Ning, Michael Steinbach, and Vipin Kur <i>mining</i> , Pearson Education India, 2016.	nar, Introduction to data				
	12. Murphy, Kevin P., <i>Machine Learning: A Probabilisti</i> 2012.	ic Perspective, MIT press,				
	13. Theobald, Oliver, <i>Machine Learning for Absolute Be</i> <i>Introduction</i> , 2nd Edition, Scatterplot press, 2017.	eginners: A Plain English				
	14. The NumPy community, NumPy: The Absolute Basic	es for Beginners.				
	15. The Pandas community, 10 minutes to Pandas					
	 Géron, A., Hands-on machine learning with S TensorFlow: Concepts, tools, and techniques to b O'Reilly Media, 2019 					

17. Turnbull, J. The Docker Book: Containerization is the new virtualization, Turnbull, 2014.
