

Universitas Negeri Surabaya Faculty of Mathematics and Natural Sciences Data Science Undergraduate Study Program

Document Code

SEMESTER LEARNING PLAN

Courses	CODE			Cou	rse Fa	amily	'	Cre	dit We	ight		SE	MEST	ER	Con Dat	npilatio e	
Multivariate	49202030	63			pulso			T=3	P=0	ECTS	6=4.77	'	4		July	19, 202	
UTHORIZA	SP Devel	per		- Prog	Jram S			se Clu	ster C	oordin	ator	Stu	ıdy Pr	ogran	1 Coo	rdinato	
		Ulfa Siti N Puspitasa			Dewi	Intan		Dr. At	ik Win	tarti, M	1.Kom		Yu	liani Pi	uji Ast	uti, S.	Si., M.S
earning nodel	Project Base	d Learning	learning														
rogram	PLO study program that is charged to the course																
.earning Dutcomes	PLO-8	Work together a	nd hav	e social s	ensitiv	rity an	d brir	ng cha	nge to	the e	nvironm	nent					
PLO)	PLO-11	Able to impleme	nt data	science t	echno	ology i	n rea	l probl	lems								
	PLO-12	Able to design a databases, data	Able to design and develop algorithms for various purposes such as big data analysis, artificial intelligence, databases, data mining, inferential statistics, algorithm design and analysis, and data warehouse.														
	PLO-17																
	Program Ob	jectives (PO)															
	PO - 1	Students unders statistics	Students understand the concept of multivariate statistics, matrices, vectors, and their operations in multivariate statistics														
	PO - 2	Students are abl	Students are able to understand the multivariate reduction process (PCA and FA) and apply it to real problems														
	PO - 3	Students are able to understand the concept of multivariate clustering and apply it to real problems															
	PO - 4	Students are able to understand multivariate modeling concepts such as Linear Models, GLM, SEM, and PLS and can apply them to real problems															
	PO - 5	Students can implement multivariate analysis on real problems and realize creative ideas, as well as explain the results of the analysis scientifically															
	PLO-PO Mat	rix															
								<u> </u>			1			٦			
		P.0		PLO-8	5		PLO-	11		PLO-	12		PLO-	17	_		
		PO-1													_		
		PO-2													_		
		PO-3															
		PO-4															
		PO-5															
	PO Matrix at	the end of each le	arnin	g stage ((Sub-l	PO)											
		P.O								Week							
			1	2 3	4	5	6	7	8	9		11	12	13	14	15	16
		PO-1	+-		·	-	_		-	· ·		-+	-	-		-	
		PO-2	+		~	1											
		PO-3	+	$\left \right $	+	<u> </u>						+					
				1 1	1			+									
							1	1			1	1	1	1	1	1	1
		PO-4 PO-5					1	~	~		1	~	1	~	~	1	1

Short Course Descript	tion matrices, vector distribution, hyp Equation Model	s and their operation othesis testing, the ng (SEM), clustering	ons, as well as clas concept of MANOVA method. Apart from t	sification in mi , the concept hat, students a	ich include the basic co ultivariate statistics, the of Principle Component re expected to be able to m solving using techniqu	concept of norm Analysis, Fuzzy design problem s	al multivariate and Structural solutions using
Referen	ces Main :						
	2. Härdle,	Wolfgang Karl, and L		d multivariate st	3 atistical analysis. Springe ar models. CRC press, 20		
	Supporters:						
	1. Barbara	G. Tabachnick, Usin	g Multivariate Statisti	cs, Sixth Edition	, Pearson, 2013		
Support lecturer		ntan Puspitasari, M.K S.Stat., M.Stat.	ćom.				
Week-	Final abilities of each learning stage	Eval	uation	Lear Stude	elp Learning, ning methods, nt Assignments, stimated time]	Learning materials	Assessment Weight (%)
	(Sub-PO)	Indicator	Criteria & Form	Offline(offline)	Online (<i>online</i>)	[References]	incigin (70)
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
1	Students understand the concept and purpose of multivariate analysis	 Know the application of multivariate methods to real problems. Can differentiate between univariate data and multivariate data. Know the grouping of univariate methods and multivariate methods 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: Basic concepts of multivariate analysis Applications and grouping of multivariate methods References: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	1%
2	Students understand the concepts of vectors, matrices, variance, covariance and their operations	 Able to explain vector concepts and their operations Able to explain the concept of matrices and their operations Able to explain the concept of variants and their operations Able to explain the concept of variants and their operations Able to explain the concept of covariance and its operations 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: Vectors, Matrices, Variants, Covariance Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	1%
3	Students understand the concepts of symmetric matrices, positive definite, Eigen Value, Eigen Vector	1.Able to explain the concept of Eigen Value 2.Able to explain the concept of Eigen Vector	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: Eigen Value, Eigen Vector Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%

4	Students are able to understand the concepts of Principle Component Analysis and Factor Analysis, as well as implement them in real problems	 Able to explain the concept of Principle Component Analysis Able to explain the concept of Factor Analysis (FA) 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: PCA and FA Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
5	Students are able to understand the concept of the multivariate normal distribution Review	 Able to explain the concept of univariate normal density Able to explain the concept of Bivariate normal density Able to explain the concept of Multivariate normal density and its properties Able to explain the concept of sampling distribution Able to explain the concept of sampling distribution Able to explain the concept of Assessing the assumption of normality Able to explain the concept of how to handle normality 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - univariate normal density - Bivariate normal density - Multivariate normal density and its properties - Sampling distribution - Assessing the assumption of normality - handling normality References: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%

6	Students are able to explain the concept of mean vector hypothesis testing, ANOVA and ANCOVA, MANOVA, MANCOVA, and can apply it to real problems	 Able to determine, calculate and interpret hypothesis tests for the average vector of one and two multivariate normal populations Can carry out mean vector hypothesis testing and its interpretation Can carry out assumption tests Able to understand the concepts of ANOVA and ANCOVA Able to carry out ANOVA and ANCOVA 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Hypothesis testing for one population - Hypothesis testing for two populations - ANOVA - MANOVA - MANOVA - MANCOVA - Interpretation of analysis results References: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
7	Students are able to explain the concept of hierarchical clustering and can apply it to real problems	 Able to carry out cluster analysis with several cluster analysis methods Can evaluate the results of cluster analysis Can check assumptions in cluster analysis 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: Clustering Concept and Hierarchical Approach Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
8	Students are able to explain the concept of non- hierarchical clustering and can apply it to real problems	 Able to carry out cluster analysis with several cluster analysis methods Can evaluate the results of cluster analysis Can check assumptions in cluster analysis 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: Clustering Concept and Hierarchical Approach Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
9	Midterm exam	Test	Criteria: Test Form of Assessment : Test	UTS 150	UTS 150	Material: Clustering Concept and Hierarchical Approach Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	15%

10	Students are able to understand distributional and numerical based parameter estimation	 Able to estimate parameters with MLE Able to estimate parameters using EM 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Least Square - Maximum Likelihood Estimation - Expectation Maximization (EM) Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
11	Students are able to understand the concept of Generalized Linear Models	 Able to understand the concept of Exponential Family Able to understand the concept of Generalized Linear Models Able to apply GLM to real data 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Exponential Family - Generalized Linear Models - Logistic Regression - Poisson Regression Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
12	Students understand the concept and implementation of Structural Equation Modeling (SEM) and Partial Least Square (PLS)	 Able to explain the concept of Structural Equation Modeling (SEM) Able to explain the concept of Partial Least Square (SEM) 	Criteria: Non Test Form of Assessment : Participatory Activities	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Exponential Family - Generalized Linear Models - Logistic Regression - Poisson Regression Reference: Joseph F. Hair, Multivariate Data Analysis, 8th Ed. Cengage, 2018	2%
13	Students can implement multivariate analysis on real problems and realize creative ideas, as well as explain the results of the analysis scientifically	 Able to formulate research questions from real world problems that will be solved using multivariate statistics Able to conduct literature studies related to research questions Able to design problem solving methods using multivariate statistics 	Criteria: Non Test Form of Assessment : Project Results Assessment / Product Assessment	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Method Implementation - Results Analysis - Conclusion Literature:	10%

14	Students can implement multivariate analysis on real problems and realize creative ideas, as well as explain the results of the analysis scientifically	 Able to formulate research questions from real world problems that will be solved using multivariate statistics Able to conduct literature studies related to research questions Able to design problem solving methods using multivariate statistics 	Criteria: Non Test Form of Assessment : Project Results Assessment / Product Assessment	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Method Implementation - Results Analysis - Conclusion Literature:	10%
15	Students can implement multivariate analysis on real problems and realize creative ideas, as well as explain the results of the analysis scientifically	 Able to formulate research questions from real world problems that will be solved using multivariate statistics Able to conduct literature studies related to research questions Able to design problem solving methods using multivariate statistics 	Criteria: Non Test Form of Assessment : Project Results Assessment / Product Assessment	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Method Implementation - Results Analysis - Conclusion Literature:	15%
16	Students can implement multivariate analysis on real problems and realize creative ideas, as well as explain the results of the analysis scientifically	 Able to formulate research questions from real world problems that will be solved using multivariate statistics Able to conduct literature studies related to research questions Able to design problem solving methods using multivariate statistics 	Criteria: Non Test Form of Assessment : Project Results Assessment / Product Assessment, Portfolio Assessment	Collaborative Learning (Lectures, discussions and questions and answers) 150	Collaborative Learning (Lectures, discussions and questions and answers) Independent Learning 150	Material: - Method Implementation - Results Analysis - Conclusion Literature:	30%

Evaluation Percentage Recap: Project Based Learning

No	Evaluation	Percentage
1.	Participatory Activities	20%

2.	Project Results Assessment / Product Assessment	50%
3.	Portfolio Assessment	15%
4.	Test	15%
		100%

Notes

- Learning Outcomes of Study Program Graduates (PLO Study Program) are the abilities possessed by each Study
 Program graduate which are the internalization of attitudes, mastery of knowledge and skills according to the level of their
 study program obtained through the learning process.
- The PLO imposed on courses are several learning outcomes of study program graduates (CPL-Study Program) which are used for the formation/development of a course consisting of aspects of attitude, general skills, special skills and knowledge.
- 3. **Program Objectives (PO)** are abilities that are specifically described from the PLO assigned to a course, and are specific to the study material or learning materials for that course.
- 4. **Subject Sub-PO (Sub-PO)** is a capability that is specifically described from the PO that can be measured or observed and is the final ability that is planned at each learning stage, and is specific to the learning material of the course.
- 5. Indicators for assessing abilities in the process and student learning outcomes are specific and measurable statements that identify the abilities or performance of student learning outcomes accompanied by evidence.
- Assessment Criteria are benchmarks used as a measure or measure of learning achievement in assessments based on predetermined indicators. Assessment criteria are guidelines for assessors so that assessments are consistent and unbiased. Criteria can be quantitative or qualitative.
- 7. Forms of assessment: test and non-test.
- 8. Forms of learning: Lecture, Response, Tutorial, Seminar or equivalent, Practicum, Studio Practice, Workshop Practice, Field Practice, Research, Community Service and/or other equivalent forms of learning.
- Learning Methods: Small Group Discussion, Role-Play & Simulation, Discovery Learning, Self-Directed Learning, Cooperative Learning, Collaborative Learning, Contextual Learning, Project Based Learning, and other equivalent methods.
- 10. Learning materials are details or descriptions of study materials which can be presented in the form of several main points and sub-topics.
- 11. The assessment weight is the percentage of assessment of each sub-PO achievement whose size is proportional to the level of difficulty of achieving that sub-PO, and the total is 100%.
- 12. TM=Face to face, PT=Structured assignments, BM=Independent study.