

PROGRAMME STRUCTURE Master of Science in Statistics MSc (Applied Statistics) Semester: III

Programme Outcome (PO) - For MSc Applied Statistics Programme	 Master of Science program provides extended theoretical and practical knowledge of different science subjects. Master of Science programme at Sardar Patel University is designed keeping the overall back ground preparation in mind for the student to either seek a job or to become an entrepreneur. The students, after completion of Bachelor of Science can select the master's programme in the subject they have had at the final year or in a related discipline (depending upon eligibility criteria prescribed by university). Programme outcomes: At the end of the program, the students will be able to Have a deep understanding of both the theoretical and practical concepts in the respective subject. Understand laboratory processes and use scientific equipments and work independently. Develop research temperament as a consequence of their theory and practical learning. Communicate scientific information in oral and written form. Understand the issues related to nature and environmental contexts and think rationally for sustainable development.
Programme Specific Outcome	1.
(PSO) - For MSc Applied	2.
Statistics Semester - III	

To Pass	At least 40% Marks in the University Examination in each paper and 40% Marks in the aggregate of University and Internal examination in each
	course of Theory, Practical & 40% Marks in Viva-voce.





			Theory/	Credit	Exam	Component of Marks		
Course Type	Course Code	Name Of Course	Practical		Duration	Internal	External	Total
					in hrs	Total	Total	Total
	PS03CAST51	Knowledge Discovery and Data Mining	Т	4	3	30	70	100
	PS03CAST52	Operations Research - II	Т	4	3	30	70	100
Core Course	PS03CAST53	Practicals (Based on PS03CAST51 and Based on	D	4	3	30	70	100
		PS03CAST51)	Γ			30	70	100
	PS03CAST54	Practicals	Р	4	3	30	70	100
	PS03CAST55	Comprehensive Viva-Voce	Р	4	3	30	70	100
Elective	PS03ESTA51	Planning and Analysis of Industrial Experiment	Т	4	3	30	70	100
Course	PS03ESTA52	Generallized Linear Models	Т	4	3	30	70	100
(Any Two)	PS03ESTA53	Survival Analysis	Т	4	3	30	70	100





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03CAST51	Title of the Course	KNOWLEDGE DISCOVERY AND DATA MINING		
Total Credits of the Course	4	Hours per Week	4		
Course Dijectives: The objective of this course is to provide introduction to the principles and design of data mining and statistical machine learning algorithms. The course is aimed at providing foundations for conceptual aspects of data					
	mining and machine learning algorithms along with their applications to solve real-world problems.				

Course Content				
Unit	Description	Weightage* (%)		
1.	Review of classification methods from multivariate analysis, classification and decision trees. Clustering methods from both statistical and data mining viewpoints.	25		
2.	Unsupervised learning from univariate and multivariate data, dimension reduction and feature selection	25		
3.	Supervised learning from moderate to high dimensional input spaces, artificial neural networks and extensions of regression models, regression trees	25		
4.	Introduction to databases, including simple relational databases, data warehouses and introduction to online analytical data processing. (Revision) Association rules and prediction, data attributes, applications to electronic commerce	25		

Teaching-	Interactive Class Lectures, ICT tools, Live Demonstrations & Algorithm
Learning	Building.
Methodology	

Evaluation Pattern				
Sr. No.	Details of the Evaluation	Weightage		
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%		





2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Cou	Course Outcomes: Having completed this course, the learner will be able to			
1.	understand data mining principles and techniques: Introduce DM as a cutting edge business intelligence method and acquaint the students with the DM techniques for building competitive advantage through proactive analysis, predictive modelling, and identifying new trends and behaviours.			
2.	have a good understanding of the fundamental issues and challenges of machine learning: data, model selection, model complexity, etc.			
3.	have an understanding of the strengths and weaknesses of many popular machine learning approaches.			
4.	appreciate the underlying mathematical relationships within and across Machine Learning algorithms and the paradigms of supervised and un-supervised learning.			
5.	design and implement various machine learning algorithms in a range of real-world applications.			
6.	construct basic feed-forward artificial neural networks with backpropagation. This develops a good foundation for studying deep learning.			

Suggested References:			
Sr. No.	References		
1.	Berson, A. and Smith, S.J. (1997). <i>Data Warehousing, Data Mining, and OLAP</i> . McGraw Hill.		
2.	Breiman, L., Friedman, J.H., Olshen, R.A. and Stone, C.J. (1984). <i>Classification and Regression Trees</i> . Wadsworth and Brooks/Cole.		
3.	Han, J. and Kamber. M. (2000). <i>Data Mining; Concepts and Techniques</i> . Morgan Kaufmann.		
4.	Mitchell, T.M. (1997). Machine Learning. McGraw Hill		
5.	Ripley, B.D. (1996). <i>Pattern Recognition and Neural Networks</i> . Cambridge University Press.		





On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03CAST52	Title of the Course	OPERATIONS RESEARCH – II
Total Credits of the Course	4	Hours per Week	4
Course Objectives:	The main objectiv 1. Ability to unde so that they are ab machines) more ef 2. Knowledge of analysis of manag 3. Skills in the us tools in solving read. 4. Mathematical to Research.	es of the course rstand and analy ole to use resource ffectively. formulating ma erial problems in e of Operations al problems in ir nodels for analy	are as follows: /ze managerial problems in industry ces (capitals, materials, staffing, and athematical models for quantitative in industry. Research approaches and computer industry. ysis of real problems in Operations

Course Content				
Unit	Description	Weightage* (%)		
1.	Transportation Problem (TP): Introduction, Area of application, mathematical model of TP- maximization and minimization problems, Terminologies used in LPP. Degeneracy, Balanced and unbalanced TP. North-West Corner Method (NWCM), Least Cost Method (LCM), Vogel's Approximation Method (VAM), Modified Distribution Method (MODI)	25		
2.	Assignment Problem (AP): Introduction, Area of application, mathematical model of AP – maximization and minimization problems, Hungarian Method, Multiple Optimal Solutions.	25		
3.	Network Analysis: Introduction, Minimal spanning tree problem, Maximal flow problem. PERT and CPM- Terminologies used, Similarity and Differences, steps in PERT and CPM.	25		
4.	Inventory Control-Deterministic and Probabilistic models, Non-linear Programming Problem- Kuhn-Tucker Conditions, Introduction to Simulation Techniques and Sequencing Problems.	25		

Teaching- Learning MethodologyInteractive Class Lectures, ICT tools used
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Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Course Outcomes: Having completed this course, the learner will be able to		
1.	formulate a real-world problem as a mathematical programming model	
2.	solve network models like the shortest path, minimum spanning tree, and maximum flow problems	
3.	understand the applications of, basic methods for, and challenges in non-linear programming	
4.	identify, mathematically express and solve transportation problems.	
5.	learn optimality conditions for single- and multiple-variable unconstrained and constrained non-linear optimization problems, and corresponding solution methodologies	
6.	be able to design and solve simple models of CPM and queuing to improve decision making and develop critical thinking and objective analysis of decision problems.	
7.	be able to solve simple problems of replacement and implement practical cases of decision making under different business environments.	
8.	identify, mathematically express and solve assignment problems.	

Sugges	Suggested References:	
Sr. No.	References	
1.	Kambo, N.S.(1991) Mathematical Programming Techniques Affiliated East-West Press Pvt. Ltd.	
2.	Operations Research, Paneerselvan, PHI.	





3.	Taha, H.A.(1992) Operations Research 5th Ed., Macmillan.
4.	Kanti Swarup, Gupta P. K. and Man Mohan Singh (1977) Operations Research, Sultan Chand & Sons.
5.	N. D. Vohra (2011) Quantitative Techniques in Management, 4th Ed., Mc Graw Hill.
6.	V. K. Kapoor(1998) Problems & Solutions in Operations Research, 2nd Ed., Sultan Chand & Sons.
7.	S. D. Sharma (2001) Operations Research, 13th Ed., Kedar Nath Ram Nath & Co.
8.	J. K. Sharma(2009) Quantitative Techniques For Managerial Decisions, 1st Ed., Macmillan

On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03CAST53	Title of the Course	PRACTICALS
Total Credits of the Course	04	Hours per Week	04
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Course	1.
Objectives:	2.

Course Content		
Unit	Description	Weightage* (%)
1.		25
2.		25
3.		25
4.		25

Teaching- Learning		
Methodology		

Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%





Cou	Course Outcomes: Having completed this course, the learner will be able to		
1.			
2.			
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Suggested References:		
Sr. No.	References	
1.		
2.		

On-line resources to be used if available as reference material
On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03CASTA54	Title of the Course	PRACTICALS
Total Credits of the Course	04	Hours per Week	04

Course	1.
Objectives:	2.

Course Content			
Unit	Description	Weightage* (%)	
1.		25	
2.		25	
3.		25	
4.		25	

Teaching- Learning		
Methodology		

Evaluation Pattern			
Sr. No.	Details of the Evaluation	Weightage	
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%	
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%	
3.	University Examination	70%	





Course Outcomes: Having completed this course, the learner will be able to			
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Sugges	Suggested References:				
Sr. No.	References				
1.					
2.					

On-line resources to be used if available as reference material
On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03CAST55	Title of the Course	COMPREHENSIVE VIVA-VOCE
Total Credits of the Course	01	Hours per Week	04

Course	1.
Objectives:	2.

Course Content				
Unit	Description	Weightage* (%)		
1.				
2.				
3.				
4.				

Teaching- Learning		
Methodology		

Evaluation Pattern			
Sr. No.	Details of the Evaluation	Weightage	
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%	
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%	
3.	University Examination	70%	





Course Outcomes: Having completed this course, the learner will be able to		
1.		
2.		
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Sugges	Suggested References:		
Sr. No.	References		
1.			
2.			

On-line resources to be used if available as reference material		
On-line Resources		





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03EAST51	Title of the	Planning and Analysis of Industrial		
Course Code	1505LA5151	Course	Experiments		
Total Credits	04	Hours per	04		
of the Course	04	Week	04		
Course	1. To learn how to plan, design and conduct experiments efficiently				
Objectives:	and effectively.				
	2. To analyze the resulting data to obtain objective conclusions.				
	3. To equip the student with modern statistical designs, and their				
	applications in the industry.				

Course	Course Content				
Unit	Description	Weightage* (%)			
1.	A review of basic concepts of design of experiment. Factorial Experiments: Concepts of main effects, interaction, Analysis of full 2n and 32 factorial designs, Confounding: Total and partial confounding. Analysis of 2n and 3n confounded design	25			
2.	2 (n-p) Fractional Factorial Designs: Basic Idea, Generating the Design, The Concept of Design Resolution, Plackett-Burman Designs for Screening, Enhancing Design Resolution via Foldover, Aliases of Interactions: Design Generators, Blocking, Replicating the Design, Adding Center Points, Analyzing the Results of a 2(n-p) Experiment.	25			
3.	3 (n-p) Fractional Factorial Designs: Overview, Designing 3(n-p) Experiments, Box-Behnken Designs, Analyzing the 3(n-p) Design, ANOVA, Parameter Estimates.	25			
4.	Central Composite and Non-Factorial Response Surface Designs: Overview, Design Considerations, Alpha for Rotatability and Orthogonality, Available Standard Designs, Analyzing Central Composite Designs, The Fitted Response Surface, Categorized Response Surfaces. Taguchi Methods: Robust Design Experiments: Overview, Quality and Loss Functions, Signal-to-Noise (S/N) Ratios, Orthogonal Arrays, Analyzing Designs, Accumulation Analysis	25			

Teaching-			
Learning			
Methodology			





Evalu	Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage	
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%	
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%	
3.	University Examination	70%	

Cou	Course Outcomes: Having completed this course, the learner will be able to		
1.	Appreciate the advantages and disadvantages of a design for a particular experiment.		
2.	Understand the potential practical problems in its implementation.		
3.	Construct optimal or good designs for a range of practical experiments.		
4.	Describe how the analysis of the data from the experiment should be carried out		

Sugges	Suggested References:		
Sr. No.	References		
1.	Kshirsagar A.M. (1983) Linear Models (Marcel Dekker)		
2.	John P.W.M.(1971) Linear Models (John Wiley Ltd.)		
3.	Jeff Wu C. F., Hamada M. (2000): Experiments: Planning, Analysis and parameter design optimization, John Wiley & Sons.		
4.	Montgomery D.C. (2001): Design and Analysis of Experiments, 5th edition, Wiley New York.		
5.	Angela Dean and Daneil Voss (1999): Design and Analysis of Experiments, Wiley.		
6.	Phadke, M.S. (1989): Quality Engineering using Robust Design, Prentice-Hall.		





On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03EAST52	Title of the Course	GENERALIZED LINEAR MODELS
Total Credits of the Course	04	Hours per Week	04

Course	To introduce the theory of generalized linear models (GLIM), estimation			
Objectives:	and testing procedure and explain procedure for fitting of Genralized			
	Linear Models for real data.			

Course Content		
Unit	Description	Weightage* (%)
1.	Review of Linear Statistical Models, Discrete Response Data, Introduction of Generalized Linear Models (GLMs), Components: Linear Predictor, Link Function, Natural Parameters, Scale Parameters; Exponential Family of Distributions (EFD): Members of EFD: Normal, Lognormal, Exponential, Gamma, Binomial, Poisson, Negative Binomial; Steps for Model Fitting, Mean and Variance of EFD; Frequent Inference: Estimation of Parameters through Iteratively Reweighted Least Square (IRLS) and Algorithms, Form of Adjacent Dependent Variable and Weights, Analysis of Deviance, Nested Model and Non-Nested Model; Goodness of Fit Criteria: RSquare, Akaike Information Criterion (AIC), Schwarz Information Criterion (SIC), Bayesian Information Criterion (BIC); Step Wise Selection; Testing of Parameters through Wald Test; Confidence Intervals; GLMs Residuals: Residual Analysis, Pearson Residual, Anscombe Residual, Deviance Residual. Model Checking: Hat Matrix, Outlier, Leverage, Influence	25
2.	 Binary Data: Models for Binary Data: Group Data and Ungroup Data, Linear Predictor, Link Function: Logit, Probit and Complementary Log Log; Prospective Study and Retrospective Study, Likelihood Function, Estimation of Parameters through IRLS Method; Deviance; Probit Model, Residual Analysis Polytomous Data: Introduction of Multinomial Logistic Regression and Ordinal Regression, Examples and their inference. Ordinal Regression 	25





	Models with Qualitative or/and Quantitative Covariates; Parallel Line Regression	
3.	Count Data: Introduction Poisson Regression, Likelihood, Estimation and Testing of Parameters: Log Linear Model for Contingency Table and their Analysis Generalized Linear Models with distribution having Constant Coefficient of Variation; Gamma Family; Canonical Link Function; Inference and Residual Analysis of GLMs with Gamma Distribution; Comparison between Response having Gamma distribution and lognormal distribution; Examples and Applications	25
4.	Models for Survival Data: Estimation with Censored Survival Data and Survival Distribution: Exponential Distribution; Weibull Distribution and their Examples. Under and Over Dispersion Problem of Data, Quasi Likelihood for Estimation of Parameters, Properties of Quasi Likelihood, Analysis of Deviance; Quasi Likelihood: Binomial, Poisson, Normal, Gamma, Lognormal, Exponential; Comparison of Quasi Likelihood with Likelihood; Concept of Marginal Likelihood, Conditional Likelihood; Models with Nonlinear Parameters in Covariates; Model Checking: Checking Link Function, Checking Covariance Scale, Checking the Variance Function, Score Test for Extra Parameters, Checking Form of Covariates, Detection of Influential Observations	25

Teaching-	Discussion and question answers based learning
Learning	Black board/Multimedia projector using ICT Tools
Methodology	Learning through Problem solving approach
	Assignments and seminars are given for development of confidence
	among students
	Fitting of various models to data using software (demonstration of
	software R language for data handling)

Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage





1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Cou	Course Outcomes: Having completed this course, the learner will be able to		
1.	Understand the differences between Linear model and generalized model and their underlying assumptions.		
2.	Various estimation procedures for estimation of parameters of different generalized models.		
3.	know Logit, Probit for binary data and their applications		
4.			

Suggested References:			
Sr. No.	References		
1.	Agresti, A. (2002). Categorical Data Analysis, ED.II, Wiley InterScience		
2.	Fahrmier,L .and Tutz,G.(2001). Multivariate Statistical Modeling Based on Generalizes Linear Models, Springer		
3.	Gill, J.(2001).Generalized Linear Models: A Unified Approach, Sage Publication		
4.	Lindsey, J.K. (1997). Applying Generalized Linear Models , Springer		
5.	Maindonald, J. And Braun, J. (2007). Data Analysis and Graphics using R: An example based approach Ed.II, Cambridge University Press		
6.	McCullagh, P. And Nelder, J.A. (1983). Generalized Linear Models- Monographs on Statistics and Applied Probability, Chapman and Hall		
7.	Myers, R.H, Montegomery, D.C., Vinning, G.G and Robinson, T.J.(2010). Generalized Linear Models with Applications in Engineering and the Sciences,		





	Ed.II, Wiley Series in Probability and Statistics, A John Wiley & Sons.
8.	Agresti, A. (2002). Categorical Data Analysis, ED.II, Wiley InterScience
9.	Fahrmier,L .and Tutz,G.(2001). Multivariate Statistical Modeling Based on Generalizes Linear Models, Springer
10.	Gill, J.(2001).Generalized Linear Models: A Unified Approach, Sage Publication

On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – III

Course Code	PS03EAST53	Title of the Course	SURVIVAL ANALYSIS	
Total Credits of the Course	04	Hours per Week	04	

Course Objectives:	1. Survival analysis involves the analysis of time-to-event data and is wide used in health and medicine. The objective of this course is introduci	
	the fundamental concepts of survival analysis and basic principles such as censoring, survival and hazard functions, Cox proportional hazards	
	regression, model diagnostics and model extensions to incorporate recurrent events and competing risks. 2. To introduce various frailty models.	

Course Content		
Unit	Description	Weightage* (%)
1.	Survival data, Concepts of time, order and random and hybrid censoring, Life distributions - exponential, gamma, Lognormal, Pareto, linear failure rate, Ageing classes - IFR, IFRA, NBU, NBUE, HNBUE and their duals, Bathtub failure rate. Parametric inference, point estimation, confidence Intervals, scores, tests based on LR, MLE	25
2.	Life tables, failure rate, mean residual life and their elementary properties. Estimation of survival function - Actuarial estimator, Kaplan - Meier estimator, Estimation under the assumption of IFR/DFR	25
3.	Semi-parametric regression for failure rate - Cox's proportional hazards model, partial likelihood, estimation and inference methods for the Cox models, time-dependent covariates, residuals and model diagnosis, functional forms of the Cox models, goodness-of-fit tests for the Cox models, Competing risk models, Repair models, Probabilistic models, Joint distribution of failure times Unconditional tests for the time truncated case, Tests for exponentiality, two sample nonparametric problem.	25
4.	Nelson-Aalen estimators, counting processes and martingales, modeling counting processes, Regression models for modeling multiple events, Frailty models, Shared frailty models, Identifiability of frailty models, Frailty regression models, Bivariate and correlated frailty models, Additive frailty models.	25





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Teaching-		
Learning		
Methodology		

Evaluation Pattern			
Sr. No.	Details of the Evaluation	Weightage	
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%	
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%	
3.	University Examination	70%	

Cou	Course Outcomes: Having completed this course, the learner will be able to		
1.	decide the type of censoring and truncation that is the basis for given survival data		
2.	estimate survival functions using parametric and non-parametric methods		
3.	construct a life table using the Kaplan-Meier approach		
4.	interpret coefficients in Cox proportional hazards regression analysis		
5.	compare survival functions of two or more populations; use software for survival analysis		
6.	explain theoretical concepts underlying parametric survival models; and		
7.	analyse survival data using basic parametric models and interpret the results		

Sugges	sted References:
Sr. No.	References
1.	Klein, J. P. and Moeschberger, M. L. (1997). Survival Analysis: Techniques for Censored and Truncated Data, Springer, New York





2.	Collett, D. (2003). Modelling Survivaldata in Medical Research, Second Edition, Chapman & Hall/CRC
3.	Therneau, T. M. and Grambsch, P. M. (2000). Modeling Survival Data, Extending the Cox Model, Springer, New York.
4.	Cox, D.R. and Oakes, D. (1984). Analysis of Survival Data, Chapman and Hall.
5.	Deshpande, J.V. and Purohit, S.G. (2005). Life Time Data: Statistical Models and Methods, Word Scientific.
6.	Duchateau, L. and Johnson, P. (2008). The Frailty Model. Springer: New York.
7.	Hanagal, D. D. (2011). Modeling Survival Data Using Frailty Models. CRC Press.
8.	Kalbfleish, JD. and Prentice, RL. (2002). The Statistical Analysis of Failure Time Data. New York: Wiley.
9.	Hougaard, P. (2000). Analysis of Multivariate Survival Data. Springer: New York.
10.	Wienke, A. (2011). Frailty Models in Survival Analysis, CRC Press: New York.

On-line resources to be used if available as reference material
On-line Resources





PROGRAMME STRUCTURE Master of Science in Statistics MSc (Applied Statistics) Semester: IV

Programme Outcome (PO) - For MSc Applied Statistics Programme	 Master of Science program provides extended theoretical and practical knowledge of different science subjects. Master of Science programme at Sardar Patel University is designed keeping the overall back ground preparation in mind for the student to either seek a job or to become an entrepreneur. The students, after completion of Bachelor of Science can select the master's programme in the subject they have had at the final year or in a related discipline (depending upon eligibility criteria prescribed by university). Programme outcomes: At the end of the program, the students will be able to Have a deep understanding of both the theoretical and practical concepts in the respective subject. Understand laboratory processes and use scientific equipments and work independently. Develop research temperament as a consequence of their theory and practical learning. Understand the issues related to nature and environmental contexts and think rationally for sustainable development. The students are able to handle unexpected situations by critically analyzing the problem.
Programme Specific Outcome	1.
(PSO) - For MSc Applied	2.
Statistics Semester - III	

To Pass	At least 40% Marks in the University Examination in each paper and 40% Marks in the aggregate of University and Internal examination in each
	course of Theory, Practical & 40% Marks in Viva-voce.





			Theory/	Credit	Exam	Component of Marks		
Course Type	Course Code	Name Of Course	Theory/ Dractical		Duration	Internal	External	Total
			Practical		in hrs	Total	Total	Total
Core Course	PS04CAST51	Computer Oriented Statistical Methods	Т	4	3	30	70	100
	PS04CAST52	Practicals	Р	4	6	30	70	100
	PS04CAST53	Project Work	Р	12	18	30	70	100
	PS04CAST54	Comprehensive Viva	Р	1	6	30	70	100
Elective	PS04ESTA51	Clinical Trials	Т	4	3	30	70	100
Course	PS04ESTA52	Econometrics and Time Series Analysis	Т	4	3	30	70	100





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04CAST51	Title of the Course	COMPUTER ORIENTED STATISTICAL METHODS
Total Credits of the Course	04	Hours per Week	04

Course Objectives:	1. Develop understanding of generating samples from a specified distribution as input to a simulation model.
	2. Illustrate some widely-used techniques for generating random variates: Inverse-transform technique, Acceptance-rejection technique, Special properties.
	3. To introduce various multivarite techniques for reduction and extraction of multivariate data.
	4. To study interdependency and inter-relationship among the variables.

Course Content		
Unit	Description	Weightage* (%)
1.	Generation of Random Numbers From Uniform, Binomoial, Poisson, Exponential, Weibull, Normal, Gamma, T, F, Multivariate Normal Distribution and Different Stochastic Processes Using Pseudo Random Number Generation Algorithms Like Linear Congruential Method (Lcg), Inverse Method, Rejection Method, etc.	25
2.	Simulation Principles: Rejection Method; Variance Reduction; Importance Sampling. Simulation of Probability Distribution of Different Statistics Using Monte Carlo and Similar Techniques. Estimation of Bias, MSE and other Statistics using Bootstrap and Similar Techniques. MCMC Algorithms: Metropolis-Hastings Algorithm; Gibbs Sampling.	25
3.	Logistic Regression Models: Introduction; The Multiple Logistic Regression Model; Fitting The Logistic Regression Model; Testing for The Significance of The Model. Application of Logistic Regression in study of Matched Case Control Data. Cox's Regression Model:	25





	Proportional Hazard Model. Estimation and tests of parameters of the proportional Hazard Model. Use of this in comparison of two or more life distributions. Discriminant analysis.	
4.	Multivariate Techniques: (I) Principal Component Analysis (II) Factor Analysis (III) Canonical Correlation (IV) Cluster Analysis.	25

Teaching-	Discussion and question answers based learning	
Learning	Black board/Multimedia projector using ICT Tools	
Methodology	Learning through Problem solving approach	
	Assignments and seminars are given for development of confidence	
	among students	

Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Cou	Course Outcomes: Having completed this course, the learner will be able to		
1.	generate random variables or samples using various methods of random variable generation		
2.	know various methods for simulation		
3.	use appropriate multivariate technique for data analysis depending of objectives		

Suggested References:		
Sr. No.	References	





1.	Anderson, T.W. (2003). An Introduction to Multivariate Statistical Analysis, Ed. IV, Wiley
2.	Bhuyan, K.C. (2004). Multivariate Analysis and its Applications, New Central Book Agency
3.	Efron, B. and Tibshirani. R.J. (1993); An Introduction to the Bootstrap, Chapman and Hall.
4.	Gemerman, D. and Lopes, H.F. (2006). Markov Chain Monte Carlo: Stochastic Simulation for Bayesian Inference, Ed. II, Chapman and Hall
5.	Hair, J. F., Black, W.C., Babin, B. J., Anderson, R.E. and Tatham, R. L. (2006). Multivariate Data Analysis, Ed, VI, Pearson Education
6.	Hardle, W. and Simar, L. (2007) Applied Multivariate Statistical Analysis, Springer
7.	Johnson, R. A. and Wichern, D. W. (2007). Applied Multivariate Statistical Analysis, Prentice-Hall International
8.	Kroese, D. P., Taimre, T. and Botev, Z. I. (2011). Handbook of Monte Carlo Method, Wiley
9.	Manly, B.F. and Navarro Alberto, J.A. (2017). Multivariate Statistical Methods, Ed. IV, CRC Press
10.	McLachlan, G.J. and Krishnan, T. (1997) The EM Algorithms and Extensions.(Wiley.)

On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04CAST52	Title of the Course	PRACTICALS
Total Credits of the Course	4	Hours per Week	6

Course Objectives:	1.
	2.
	3.
	4.

Course Content		
Unit	Description	Weightage* (%)
1.		25
2.		25
3.		25
4.		25

Teaching- Learning Methodology	Discussion and question answers based learning Black board/Multimedia projector using ICT Tools Learning through Problem solving approach Assignments and seminars are given for development of confidence
	Assignments and seminars are given for development of confidence among students

Evalu	ation Pattern	
Sr. No.	Details of the Evaluation	Weightage





1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Course Outcomes: Having completed this course, the learner will be able to		
1.		
2.		
3.		

Suggested References:		
Sr. No.	References	
1.		
2.		
3.		
4.		
5.		
6.		
7.		
8.		
9.		
10.		





On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04CAST53	Title of the Course	PROJECT
Total Credits of the Course	12	Hours per Week	18

Course Objectives:	1.
	2.
	3.
	4.

Course Content		
Unit	Description	Weightage* (%)
1.		25
2.		25
3.		25
4.		25

Teaching- LearningDiscussion and question answers based learningBlack board/Multimedia projector using ICT ToolsMethodologyLearning through Problem solving approachAssignments and seminars are given for development of confidence	
Assignments and seminars are given for development of confide	
	among students

Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage





1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Course Outcomes: Having completed this course, the learner will be able to		
1.		
2.		
3.		

Suggested References:		
Sr. No.	References	
1.		
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On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04CAST54	Title of the Course	COMPREHENSIVE VIVA
Total Credits of the Course	1	Hours per Week	4

Course Objectives:	1.
	2.
	3.
	4.

Course Content				
Unit	Description	Weightage* (%)		
1.		25		
2.		25		
3.		25		
4.		25		

Teaching- Learning Methodology	Discussion and question answers based learning Black board/Multimedia projector using ICT Tools Learning through Problem solving approach Assignments and seminars are given for development of confidence
	Assignments and seminars are given for development of confidence among students

Evaluation Pattern			
Sr. No.	Details of the Evaluation	Weightage	





1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Cou	Course Outcomes: Having completed this course, the learner will be able to				
1.					
2.					
3.					

Suggested References:				
Sr. No.	References			
1.				
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On-line resources to be used if available as reference material

On-line Resources





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04EAST51	Title of the Course	CLINICAL TRIALS
Total Credits of the Course	04	Hours per Week	04

Course Objectives:	 To gain knowledge of bio-statistics techniques used in design and analysis of Clinical trials
	2. To train in analysis of commonly conducted pharmaceutical clinical trials
	3. To learn some novel contemporary statistical designs, statistical tests and statistical analysis techniques used in clinical trials

Course	Course Content				
Unit	Description	Weightage* (%)			
1.	Introduction to clinical trials, the need, ethics, protocol of clinical trials, Overview of phase $1 - IV$ and DF, SE, CTE trials, data management and case studies. Bias and random error in clinical studies, Endpoints of clinical trials and sample size estimation in SE and CTE trials	25			
2.	Design of clinical trials parallel vs. cross over designs, cross sectional vs. longitudinal designs, review of factorial designs. Randomization techniques for group allocation.	25			
3.	Analysis of outcomes from Phase I- III trials, analysis of survival data from clinical trials, techniques for Interim analysis, intent to treat analysis.	25			
4.	Application areas Meta analysis, Multi-center trials, Bioequivalence trials	25			

Teaching-	Interactive	Class	Lectures,	ICT	Tools,	Problem	solving	and	Group
Learning Methodology	Seminar.								





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Evaluation Pattern				
Sr. No.	Details of the Evaluation	Weightage		
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%		
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%		
3.	University Examination	70%		

Course Outcomes: Having completed this course, the learnerwill be able to		
1.	Writing and understanding CT protocol. Understand, differentiate and identify among four Phases of a complete Clinical Trial	
2.	Understand study design of published clinical trial. Make choice and carry out randomized allocation of two treatments as specified. Execute clinical trial as per the experimental design.	
3.	Given the clinical trial objective and response type, understand the sample size estimation routine or formula and apply to calculate it.	
4.	Perform analysis of survival clinical trials, meta-analysis for systematic review of published clinical trials, interim analysis of trials employing group sequential testing.	

Suggested References:		
Sr. No.	References	
1.	Shein-Chung Chow and Jen-Pei Liu (2014). Design and Analysis of Clinical Trials, Concepts and Methodologies, 3 rd ed., John Wiley	
2.	Millard, S. P. and Krause, A. (2010). Applied Statistics in the Pharmaceutical Industry with Case Studies using S-plus, Springer Verlag New York	
3.	Senn, S (2002). Cross – Over Trials in Clinical Research, 2 nd ed., Statistics in Practice, John Wiley	
4.	Jones, B. and Kenward, M. G. (2014). Design and Analysis of Cross-Over Trials, 3 rd ed. CRC press	





5.	Mike WL. Cheung (2015). Meta – Analysis, A Structural Equation Modeling Approach, John Wiley
6.	Piantadosi, S. (2005). Clinical Trials –A Methodological Perspective 3 rd ed. Wiley.
7.	Mallinckrodt, C. and Lipkovich, I. (2017). Analyzing Longitudinal Clinical Trial Data, A practical guide, CRC Press, T&F G
8.	Molenberghs, G. and Kenward, M. G. (2007). Missing Data in Clinical Studies, Statistics in Practice, John Wiley
9.	Peace, K. E. (2009). Design and Analysis of Clinical Trials with Time-to-Event Endpoints (Edited), CRC Press, T&F G
10.	Pong, A. and Shein—Chung Chow (2011). Handbook of Adaptive Designs in Pharmaceutical and Clinical Development (Edited), CRC Press T&F G
11.	Atkinson, A. C. and Biswas, A. (2014). Randomised Response-Adaptive Designs in Clinical Trials, Monograph on Statistics and Applied Probability, CRC Press, T&F G





MASTER OF SCIENCE IN APPLIED STATISTICS M.Sc. Applied Statistics, Semester – IV

Course Code	PS04EAST52	Title of the Course	ECONOMETRICS AND TIME SERIES ANALYSIS
Total Credits of the Course	04	Hours per Week	04

Course Objectives:	1. Students have concise knowledge of basic regression analysis so that they are able to understand its applications in different fields in economics.
	2. To provide students with some useful tools of econometrics which help in analysis of economic data
	 To equip students with various forecasting techniques and knowledge on modern statistical methods for analyzing time series data

Course Content		
Unit	Description	Weightage* (%)
1.	Econometrics: Definition, Methodology, Examples, Nature and Source of Data; Classical Linear Regression Model (CLRM): Assumptions, estimation of parameters through Maximum Likelihood Method and Ordinary Least Square Method, Properties of Estimator; Testing of Hypothesis and confidence intervals, Testing of Subset of Regressors, Point Predictor, Model Selection Criterion; R^2 , Adj R^2 , AIC, BIC Mallow's C_p Statistic; Significance Test and Confidence Interval; Dummy Variable: Nature, introduction, examples, Chow Test, Seasonal Adjustment	25
2.	Hetroscedasticity: Reason of Hetroscedasticity; Consequences of using OLS in presence of Hetroscedasticity; Detection: Informal Method, Formal Method; Park Test, Goldfield-Quant Test; Remedial Measures, Method of Generalized Least Squares (GLS), Autocorrelation: Nature of the Problem, Consequences of Autocorrelation, Detection: Graphical Method; Durbin-Watson d Test, A General Test of Autocorrelation, The Breusch-Goldfrey (BG) Test; GLS when correlation coefficient is known as well as unknown;	25





3.	Multicolinearity Problem, Its implications and tools for handling the problem; Detection of Multicolinearity; Remedial Measures; Ridge Regression; Use of Principle Component Analysis; Introduction to Simultaneous Equation Models; The identification Problem	25
4.	Introduction to Time Series Analysis; Some Basic Concepts: white noise, stationary, non stationary time series, ACF and PACF plot; Unit Root Test (Augmented Dickey-Fuller Test); Forecasting: Exponential Smoothing Methods, AR Process, MA Process, ARMA Process, ARIMA Process; The Box-Jenkins (BJ) Methodology; VAR Model; Auto Regressive Conditional Hetroscedasticity (ARCH) and Generalized Autoregressive Conditional Hetroscedasticity (GARH) Model;	25

Teaching-	Discussion and question answers based learning
Learning	Black board/Multimedia projector using ICT Tools
Methodology Learning through Problem solving approach	
	Assignments and seminars are given for development of confidence
	among students

Evaluation Pattern		
Sr. No.	Details of the Evaluation	Weightage
1.	Internal Written / Practical Examination (As per CBCS R.6.8.3)	15%
2.	Internal Continuous Assessment in the form of Practical, Viva-voce, Quizzes, Seminars, Assignments, Attendance (As per CBCS R.6.8.3)	15%
3.	University Examination	70%

Course Outcomes: Having completed this course, the learner will be able to		
1.	to specify assumptions, formulate and estimate appropriate models, interpret the results and test their statistical significance.	





2.	explain core concepts and techniques in econometrics with special focus on the classical linear regression model.
3.	understand the assumptions upon which different econometric methods are based and their implications.
4.	distinguish regression analysis model and time series model.
5.	understand various components of time series and various time series models.

Suggested References:		
Sr. No.	References	
1.	Cameron, A.C. and Trivedi, P.K. (2005). Microeconometrics Methods and Applications, Cambridge University Press	
2.	Cooray, T.M.J.A. (2008). Applied Time Series Analysis and Forecasting, Narosa Publishing House, New Delhi	
3.	Green, W. H. (1993). Econometric Analysis, Ed. II, MACMILLAN Publishing	
4.	Greene, W.H. (2003) Econometric Analysis. Ed. V, Pearson Education	
5.	Gourieroux, C and Jasiak, J. (2007). Financial Econometrics: Problems, Models and Methods, New Age International	
6.	Gujarathi, D.N., Porter, D.C. and Sangeetha (2012). Basic Econometrics, Ed. V , Tata MacGraw Hill, New Delhi	
7.	Johnston, J. and Dinardo, J (1997). Econometric methods, Ed. IV, McGraw Hill	
8.	Ruppert, D.(2004). Statistics and Finance: An Introduction, Springer (India) Pvt. Ltd.	
9.	Shumway, R.H. (1988). Applied Statistical time Series Analysis, Prentice Hall , New Jersey	
10.	Theil, H. (1982) : Introduction to the theory and practice of Econometrics, John Wiley.	

On-line resources to be used if available as reference material

On-line Resources





