

#### STRATHMORE INSTITUTE OF MATHEMATICAL SCIENCES

# MASTER OF SCIENCE (DATA SCIENCE AND ANALYTICS)

# **COURSE OUTLINE**

**DSA 8305: LINEAR MODELS** 

#### LECTURER INFORMATION

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**Aim:** To introduce linear models as a broad class of statistical models applying the general principles of likelihood-based inference and further, to enable the student to acquire deep knowledge in the theory and practice of modern regression modelling within the unifying context of generalized linear models. A second aim is to get students acquainted with research questions in this domain.

# **Course Learning Outcomes:**

At the end of this course the student will be able to:

- 1. Formulate linear models including model building, parameter estimation, hypothesis testing, model validation, and model diagnostics.
- 2. Apply various linear models to real life data and interpret the results.
- 3. Critique research papers in the area of linear models and applications in scientific way.
- 4. Use R and other software to perform data analysis and interpret the results.

Contact Hours: Lecture hours: 35 Lecture hours, 30 Lab hour

Prerequisite: None.

#### Course content

	Course content					
Week	Topic/ Sub-topics	Intended Learning Outcome(s)	Class Activity			
1-4	<ul> <li>Theory of general linear models.</li> <li>Model construction.</li> <li>Full and less than full rank models.</li> <li>Estimation and properties of estimates.</li> <li>Hypothesis testing.</li> </ul>	By the end of this topic the student will be able to  - Describe what model construction entails Formulate linear models Learn on methods of estimating parameters and their properties - Formulate and test hypothesis Perform variable and model selection - Perform model prediction	<ul> <li>Lecture illustrations</li> <li>Group discussions.</li> <li>Questions and answers sessions</li> <li>Use numerical methods to solve problems.</li> <li>Practical assessment using R</li> </ul>			

Week	Topic/ Sub-topics	Intended Learning Outcome(s)	Class Activity				
5-8	<ul> <li>Variable and model selection</li> <li>Prediction</li> <li>Non-linear models</li> <li>Nonlinear models via nonlinear least squares.</li> <li>Basic splines and Generalized Additive Models</li> </ul>	<ul> <li>Solve the formulated models using computer software.</li> <li>By the end of this topic the student will be able to</li> <li>Describe nonlinear models and nonlinear least squares.</li> <li>Describe splines and GAM models and be able to analyze them.</li> </ul>	<ul> <li>Lecture illustrations</li> <li>Group discussions.</li> <li>Questions and answers sessions</li> <li>Use numerical methods to solve problems.</li> <li>Practical assessment</li> </ul>				
	using R  CAT 1 (WEEK 8)						
0.11							
9-11	<ul> <li>Ridge Regression and Principal Components Regression for collinear data</li> <li>Models for non-independent errors.</li> </ul>	<ul> <li>By the end of this topic the student will be able to</li> <li>Formulate models relating to ridge regression and solve PCA problems.</li> <li>Solve models with non-independent errors.</li> </ul>	<ul> <li>Lecture illustrations</li> <li>Group discussions.</li> <li>Questions and answers sessions</li> <li>Practical assessment using R</li> </ul>				
	Semi parametric regression models  • Scatter plot smoothing. • Mixed models. • Additive models. • Interaction models and generalized regression	<ul> <li>By the end of this topic the student will be able to</li> <li>Perform smoothing relating to linear models.</li> <li>To understand and be able to analyze mixed models, additive models and generalized models.</li> </ul>	<ul> <li>Lecture illustrations</li> <li>Group discussions.</li> <li>Questions and answers sessions</li> <li>Class presentations</li> <li>Practical assessment using R</li> </ul>				
	CAT 2 (WEEK 15)						

# **Course Delivery Methodology:**

- Lectures
- Case Studies
- Group Work and Presentations

# **Course Materials**

Name	Author	Publisher
Analysis of Categorical Data; 2nd Edition	Agresti A.	Springer Berlin Heidelberg, (2011) ISBN: 978-3-642-04897-5, ISBN: 978-3-642-
		04898-2
An introduction to generalized linear	Dobson A. J., & Barnett A.	Chapman and Hall/CRC, (2018).
models	G.	
An R and S-Plus Companion to Applied	Fox John	Sage Publication, (2010)
Regression; 2nd Edition		ISBN: 141297514X, ISBN: 978-1412975148

# **Academic Assessment**

- Homework assignments and quizzes
- Two sit-in CATS, and the final exam.
- The breakdown for each component is as follows:

Туре	Weighting (%)
Coursework (CATs and Assignments)	40
End of semester Examination (Final Exam)	60
Total	100

#### Communication

- The primary means of communication outside of class will be by email.
- Students must use the university email accounts to receive important University information, including messages related to this class.

# Collaborative work

• Collaboration in course work is certainly encouraged as this promotes team spirit and group synergy as long provided originality is preserved.

# **Computational Software:**

- The main computational software that I use is R.
- R is open source and is free.

# **Document Development Software:**

- The main document development software that I use is TeX.
- This is also a free software. One has to install the basic MikTeX before installing any TeX interface.