

Semester	<b>FOUR</b>
Paper Number	<b>13</b>
Paper Code	<b>MDTS 4411</b>
Paper Title	<b>Bayesian Data Analysis and Data Governance &amp; Compliance</b>
No. of Credits	<b>6</b>
Course description	<p>CORE</p> <p>Composite Paper</p> <p>Module 1: 2 classes/week</p> <p>Module 2: 2classes/week</p> <p>No. of classes assigned Theory: 4 classes per week</p> <p>Practical: 3 classes per week</p>
Course Objective	<p>At the end of the course, a student is expected to understand the</p> <ol style="list-style-type: none"> <li>1. At the end of the course, a student is expected to understand the</li> <li>2. Bayesian estimation and credible regions.</li> <li>3. Hierarchical and Empirical Bayesian analysis.</li> <li>4. Complex posterior computation using Acceptance-rejection, importance sampling and simulation methods such as MCMC.</li> <li>5. Bayesian Linear and logistic regression.</li> <li>6. How privacy, security and transparency standards in using data are met.</li> <li>7. How misuse of data may impact the society at large.</li> <li>8. Have an insight into the consequences of violation of property rights.</li> </ol>
Syllabus	<p><b>Module I</b></p> <p><b>Bayesian Data Analysis</b></p> <p><b>Unit1:</b> Subjective definition of probability, Conditional Probability, Marginal Probability, Bayes theorem, Applications of Bayes theorem in Spam filter, Bayesian search, etc. [3]</p> <p><b>Unit2:</b> Prior and posterior distributions, Posterior estimates, credible intervals, highest posterior density regions; Hierarchical models, Hierarchical Bayes and Empirical Bayes. [7]</p> <p><b>Unit3:</b> Acceptance-rejection sampling, importance sampling; Markov chain basics, Markov chain Monte Carlo (MCMC), Gibbs sampling, Metropolis-Hastings MCMC, MCMC diagnostics. [10]</p> <p><b>Unit4:</b> Bayesian Linear regression; Bayesian logistic regression. [6]</p>

	<p><b>Module-II</b></p> <p><b>Data Governance &amp; Compliance</b></p> <p><b>Data Sources:</b> Different sources of data and their interrelations. Database, Data Lake, Data Mesh (6L)</p> <p><b>Data Quality :</b> Different aspects of data quality measures and their statistical relevance (6L)</p> <p><b>Data and Information:</b> Theory of Information. Data Governance, Ownership and consent. Use/Overuse/Misuse of Data. Equilibrium. The FAT Flow framework of Data Science Ethics. (5L)</p> <p><b>Data Privacy:</b> Privacy and Confidentiality. Aspects and Challenges. PII information. Methods of maintaining data privacy. Privacy Models. Trade-off between protecting privacy and loss of information. (6L)</p> <p><b>Data Privacy &amp; Security:</b> Importance of data security and its solutions. Governance and compliance. Data Security Audits. (3L)</p> <p>In all the above topics, case studies can be discussed to elaborate on the ideas.</p>	
List of Practical	Based on Bayesian Data Analysis.	
Reading/Reference Lists	<ol style="list-style-type: none"> <li>1. An Introduction to Bayesian analysis: theory and Methods; Jayanta Kumar Ghosh, Mohan Delampady, Tapas Samanta.</li> <li>2. Bayes and Empirical Bayes Methods for Data Analysis; Bradley P. Carlin, Thomas A Louis</li> <li>3. Bayesian Data Analysis; Andrew Gelman, John B. Carlin, Hel S. Stern, David B. Dunson, Aki Vehtari and Donald B. Rubin.</li> <li>4. Theory of Statistics; Mark J. Schervish</li> <li>5. Monte Carlo Statistical Methods; Christian Robert and George Casella.</li> <li>6. Markov Chain Monte Carlo in Practice; W. R. Gilks, S. Richardson, D. J. Spiegelhalter.</li> <li>7. Handbook of Markov Chain Monte Carlo; Steve Brooks, Andrew Gelman, Galin L. Jones, Xiao-Li Meng.</li> <li>8. Guide to Data Privacy Models, Technologies, Solutions; Vicenç Torra.</li> <li>9. Data Science Ethics- Concepts, Techniques and Cautionary Tales: David Martens, Oxford University Press, 2022.</li> <li>10. Data Ethics and Challenges: Shukla S. et al, Springer Series in Applied Sciences and Technologies, 2022.</li> <li>11. Ethics of Data and Analytics- Concepts and Cases: Kristen Martin, CRC Press, 2022.</li> <li>12. Data Governance and Data Management: Contextualizing Data Governance Drivers, Technologies, and Tools; Rupa Mahanti.</li> </ol>	
Evaluation	<p>Theory</p> <p>CIA: 10</p> <p>End Sem Exam: 50 (30+20)</p> <p>Total : 60</p>	<p>Practical(Based on Bayesian Inference)</p> <p>Continuous Assessment: 30</p> <p>End Sem Viva: 10</p> <p>Total: 40</p>

Paper Structure for End Semester Theory	Short questions: 5 marks each	Long questions: 10 marks each
Module I	2 out of 4	2 out of 3
Module II	2 out of 4	1 out of 2