STA302H1S: Methods of Data Analysis I LEC5101 – Winter 2025

Course Syllabus

Instructor Contact Information & Course Details	
Instructor:	Ismaïla Ba (email: ismaila.ba@utoronto.ca)
Class Schedule:	Wednesdays, 6:00PM – 9:00PM in MC 102.
Course Website:	https://q.utoronto.ca/courses/380340
Course Description:	The course provides a solid introduction to data analysis with a focus on the theory and application of linear regression. Topics to be covered include: initial examination of data, correlation, simple and multiple regression models using least squares, inference for regression parameters for normally distributed errors, confidence and prediction intervals, model diagnostics and remedial measures when the model assumptions are violated, interactions and dummy variables, ANOVA, and model selection and validation. Statistical software will be used throughout and will be required for the completion of various assessments during the term. The development of strong written communication skills will be emphasized.
Learning Outcomes:	 By the end of the course, all students should be able to: 1. Recognize the importance of assumptions and limitations of linear regression models to gauge when linear models are appropriate to use and to be critical of their results. 2. Interpret the results of an analysis involving linear models for technical and nontechnical audiences. 3. Apply methods of linear models and data cleaning to new datasets correctly using statistical software in a reproducible way. 4. Explain statistical concepts and theory of linear models to various audiences as would be required in the job market or collaborative environment. 5. Outline the correct use of linear models in a coherent and reproducible analysis plan. 6. Distinguish the common and unique characteristics between linear regression model and any other given machine learning models, which is important for learning or designing a new model.
Prerequisite:	Pre-requisites are strictly enforced by the department , not the instructor . If you do not have the equivalent pre-requisites, you will be un-enrolled from the course. Students should have a second year statistics course (STA238H1/STA248H1/STA255H1/STA261H1/ECO227Y1/STAB57H3/STA258H5/STA260H5/ECO227Y5), a computer science course (CSC108H1/CSC110Y1/CSC120H1/CSC148H1/CSCA08H3/CSCA20H3/CSCA48H3/CSC108H5/CSC148H5) and a mathematics course (MAT223H1/MAT224H1/MAT240H1/MATA22H3/MATA23H3/MAT223H5/MAT240H5/MATB24H3/MAT224H5), or equivalent preparation as determined by the department.

Course Materials

Course Content:	We have a Quercus course page for this course. All lecture slides and materials will be posted on this Quercus course page. In addition, any important announcements will also be posted in Quercus. Please make sure to check it regularly.
Textbook:	This course does not strictly follow any particular textbook, but rather merges materials from a number of sources. All of the below recommended textbooks are freely available as an electronic copy through the University of Toronto Library. Our primary reference text will be
	• A Modern Approach to Regression with R, by Simon J. Sheather (Springer).
	Other helpful references from which practice problems may be assigned are:
	• Applied Regression Modeling, 2nd edition, by Iain Pardoe (Wiley),
	• Methods and Applications of Linear Models: Regression and the Analysis of Variance, 2nd edition, by Ronald R. Hocking (Wiley),
	• Applied Linear Regression, 3rd edition, by Sanford Weisberg (Wiley), and
	• Linear Models in Statistics, 2nd edition, by Alvin C. Rencher and G. Bruce Schaalje (Wiley).
	These are all useful books, but may present the material in a different order or in a different way. They are still good for additional explanation and practice problems. Other useful resources will be posted on the Quercus course page.
Statistical Software:	We will be using the R Statistical Software for performing statistical analyses in this course. R is a free software that can either be downloaded onto your personal computer or used in a cloud environment. We encourage all students to use RStudio through the JupyterHub for University of Toronto. This will allow you to login with your official UofT credentials and use RStudio without the need for a local installation and can be run on any device that has access to an internet connection. More information about using RStudio in JupyterHub will be provided early in the term. R code shown in class will be available on the course page and, along with any additional resources, should be sufficient to complete any assessment involving data analysis.
	Course components
In-person Lectures:	Lectures will be conducted in person. Slides will be available on Quercus. Class time each week will comprise of a combination of lecturing, in-class activities, and code-along sessions. Where possible, you are encouraged to bring a laptop or tablet to follow along with the code.
Office Hours:	Instructor and TAs will hold office hours in a combination of online and in-person for- mats. The office hour schedule and mode of delivery will be posted on Quercus once finalized. It is recommended that you visit office hours whenever you have a question about the material. It is always important to have material clarified as quickly as possi- ble. Don't wait until the last minute to ask your questions!
Quercus Discussion Board:	We will use the Quercus Discussion Board as an online discussion forum, which can be accessed through the Quercus course page. All questions about course material should be posted here or asked during instructor or TAs office hours. The instructor and TAs will monitor the board and will help answer questions but students are encouraged to answer posts and help their fellow classmates.

	Class communication
How your instructor will communicate with you:	All communication will be made through Quercus announcements or during lectures. Please ensure that you check Quercus regularly so you don't miss anything important.
Where to send content questions:	We will use the Quercus Discussion Board to collect student questions regarding course content, assignments, etc. All questions should be posted here.
When to email the instructor:	The instructor will only respond to emails of a private or sensitive nature. If you email the instructor with content related questions, you will be asked to repost your question on the content board so the answer may benefit all students. Should you need to email the instructor about a sensitive or personal nature, please use your official mail.utoronto.ca email, include your full name and student number. Include your lecture section (L5101) in the subject line so it is received by the correct person. Send all course related emails to sta302@course.utoronto.ca. Please allow up to 48 hours for a reply. Emails will not be monitored on evenings and weekends.
A note on email and discussion board etiquette:	Please make sure that you communicate politely and respectfully with all members of the teaching team and your fellow classmates. Written communications can sometimes take a tone other than what was intended (e.g. can come off as dismissive, rude or insulting), so make sure you re-read or read out loud your email/post before sending it to make sure it has the tone you intended. For more tips on respectful communication, see professional communication tips. The Quercus discussion board is a teaching and learning tool and therefore should only be used as such. Any posts that detract from the learning goal of the board will be removed to keep the board a safe space.
	Course Work, Examinations & Grading
In-class quizzes:	Students will have 3 in-class quizzes via Quercus as scheduled above. Therefore, make sure to bring your electronic device that can assess Quercus. At every quiz, students will be given 10-15 minutes to finish 2 to 3 problems. The problem formats will be individual or multiple choices, true or false, blank filling, matching and others. It usually does not contain heavy computations or derivations, but it is more about the methodologies and conceptual understandings. You will get full marks as long as you answer 1 out of 2, or 2 out of 3 questions correctly.
Midterm Test:	There will be one IN CLASS midterm test scheduled on February 7, 2025 from 5PM-7PM. More details will be communicated closer to the test date. The test will account for 25% of the final grade and will cover material from Lectures 1-5.
Final project:	The final project will consist of a data analysis on a dataset. Students will be required to demonstrate their understanding of the methods taught in the course by developing a reasonable regression model that addresses a valid research question using the techniques from the course. The students will be responsible for choosing the correct methods to apply and providing appropriate justifications defending their choices. The final project is a scaffolded assessment involving three parts:
	Part 1 - Research question: Students will be tasked with defining a research question that can be answered with a dataset using linear regression. This portion of the project will require students to provide their research question, explain why linear regression would be a reasonable method to answer this question, and highlight important characteristics of their dataset.

Part 2 - Analysis Plan Flowchart: Students will be asked to put together a flowchart outlining the steps that they plan to take in their data analysis for the final project on their chosen dataset. This will help in developing a consistent analysis flow and in structuring the methods section of their final report.

Part 3 - Final Project Report: Students will put together a scientific report that outlines the relevance of their proposed research question, the process of their analysis, the results of the performed data analysis, and a discussion of the meaning of the results as well as limitations of the analysis with respect to the statistical tools used/decisions made or the data used.

All parts of the final project must be done **in groups of two to three students**. All group members are expected to contribute to the project equally and provide an outline of their involvement in the project. More detailed instructions for each part will be provided on Quercus at a later date.

Final exam: This will be held during the April exam period and will cover all topics.

Grading Scheme:	Assessment	Date due	Percent
	In-class quizzes		
	Quiz 1	January 22	2%
	Quiz 2	February 12	1%
	Quiz 3	March 12	2%
	Midterm Test	Feb. 7 from 5PM-7PM	25%
	Final Project (3 parts)		
	Part 1: Research Question/Proposal	February 14 by 8PM	5%
	Part 2: Analysis Flowchart	March 7 by 8PM	5%
	Part 3: Written Final Report	April 4 by 8PM	15%
	Final Exam	Scheduled by FAS	45%
	Total		100%

Please note that the last day to drop the course without penalty is March 10, 2025.

Late assessment and extension request policy

Extensions of the final project:	All groups for the final project will have access to extensions of up to 7 days to help manage illness, deadlines or other unexpected situations, but they need to inform the instructor with justification. Groups may use these extensions on the research proposal or final project report, but they need to inform the instructor in advanced or at most 7 days after the required submission deadline. While using these extensions for your final project, all group members must agree to use this extension, so groups should strive to have clear communication throughout the term. Groups who turn in the work by the assigned deadline (i.e. do not use the extension) will receive their graded work and feedback earlier than groups who use extensions. Extensions beyond this will not be granted.
Extreme Situations/Prolonged Illness:	Should a student be experiencing a prolonged illness or other situation that prevents them from turning in their work or contributing to the group project, they should imme- diately contact their instructor and College Registrar to inform them of their situation. They should also submit an Absence Declaration form on ACORN or a Verification of Illness (VOI) form that lists every day during which they were incapacitated and unable to work. These documentations should be sent to sta302@course.utoronto.ca. Accommodations will not be considered without a completed declaration or VOI, and will only be considered for extreme circumstances at the request of the College Registrar.

Accessibility-Related Extension Requests:	Students registered with Accessibility Services should notify the instructor as soon as possible if additional time is needed on assessments that are eligible for such accommodation. Please notify the instructor by email of your situation and cc your accessibility advisor in the process. The instructor will work with the accessibility advisor to determine an appropriate accommodation for your situation. However, note that group work can generally not be granted further extensions beyond those in the above policy.
	Missed assessment policy
	If you experience a prolonged absence due to illness or emergency that prevents you from completing any number of assessments, please contact your College Registrar as soon as possible so that any necessary arrangements can be made.
Missed In-classes quizzes:	There will be no accommodations made for missing the in-classes quizzes.
Missed Term Test:	If you miss the term test, then you must notify the teaching team via email (sta302@course.utoronto.ca) no later than one week after the date of the test. Valid reasons for missing the term test include:
	• Absence declaration via ACORN (see below for important information on eligibility).
	• U of T Verification of Illness or Injury Form (VOI).
	• College Registrar's letter.
	• Letter of Academic Accommodation from Accessibility Service.
	Information about student absences is available here.
	If a student misses the term test for a valid reason then the weight of the term test (25%) will be shifted to the weight of the final exam. In such case, the weight of the final exam will be 70%.
Missed Final Project:	Due to the nature of these assessment, there will be no further extensions on the research proposal or project (see Extensions of the final project) under any circumstances. Late projects will not be accepted and there are no accommodations available for individuals missed contributions to their group projects.

Regrade requests

Regrade requests will be accepted for all assessments. To be considered, you must clearly identify the question you have concerns about, provide a detailed justification for your concern and make specific references to your answer, the feedback you received and to the relevant course material. All regrade requests must be submitted by email no later than one week after the grade for that assessment is released. No regrade requests will be accepted after this deadline. The instructor reserves the right to re-evaluate the entire assessment (i.e., grades may go up, down, or remain unchanged). Please allow a few weeks for the instructor to process regrade requests.

Intellectual property

Course materials provided on Quercus, including lecture slides, assessments, videos, and solutions, are the intellectual property of your instructor and are intended solely for the use of students currently enrolled in this course. Sharing these materials with any person or organization outside of the course constitutes unauthorized use and violates copyright.

ChatGPT and other generative AI are freely available tools that can perform a variety of functions for us. However, it's important to understand how such tools are allowed to be used in this course. Acceptable uses of generative AI in this course include:

- Editing or rephrasing written work that has already been written by the student to improve the syntax, grammar and overall readability of the work.
- Synthesizing or explaining course concepts while learning and studying to contribute to their understanding of the course material.
- Looking up appropriate syntax of individual R functions for use in a data analysis or for understanding errors that may arise when running R code.

However, the work turned in by students must ultimately be their own and students will therefore be accountable for the work they turn in. Unacceptable uses of generative AI in this course include:

- Copying from any generative artificial intelligence applications, including ChatGPT and other AI writing and coding assistants, for the purpose of completing assignments in this course.
- Producing an entire data analysis, written report, or any other piece of work meant for grades.

In summary, generative AI like ChatGPT can be really helpful in your learning process and to improve skills valued in the workplace. However, it cannot be used as a substitute for learning and material produced from these tools should not be passed off as your own. This would be considered academic misconduct (see below). The instructor therefore reserves the right to ask students to explain their work and their process for creating their assignment.

Academic Integrity

The University treats cases of plagiarism and cheating very seriously. It is the students' responsibility for knowing the content of the University of Toronto's Code of Behaviour on Academic Matters. All suspected cases of academic dishonesty will be investigated following procedures outlined in the above document. If you have questions or concerns about what constitutes appropriate academic behaviour or appropriate research and citation methods, you are expected to seek out additional information on academic integrity from your instructor or from other institutional resources (see http://academicintegrity.utoronto.ca/). Here are a few guidelines regarding academic integrity:

- Using ChatGPT and other generative AI for any purpose not outlined above.
- Being dishonest when reporting an illness or personal emergency to get an extension or accommodation is an academic offence.
- You may consult class notes/lecture slides during take-home assessments, however sharing or discussing questions or answers with other students is an academic offence.
- Students must complete all assessments individually. Working together is not allowed unless otherwise specified.
- Paying anyone else to complete your assessments for you is academic misconduct.
- Completing assessments for another student is academic misconduct.
- Sharing your answers/work/code with others is academic misconduct.
- All work that you submit must be your own! You must not copy mathematical derivations, computer output and input, or written answers, etc. from anyone or anywhere else. Unacknowledged copying or unauthorised collaboration will lead to severe disciplinary action, beginning with an automatic grade of zero for all involved and escalating from there. Please read the UofT Policy on Cheating and Plagiarism, and don't plagiarise.

The University of Toronto offers academic accommodations for students with disabilities. If you require accommodations, or have any accessibility concerns about the course, the classroom, or course materials, please contact Accessibility. Services as soon as possible: accessibility.services@utoronto.ca or http://accessibility.utoronto.ca.

Tentative Schedule of Topics

Below is a tentative schedule of topics to be covered in class. The schedule is subject to change and modification.

Week (Dates)	Topic
1 (Jan. 6-10)	Simple Linear Regression Basics: relationships, notation, estimation, inter-
	pretation.
2 (Jan. 13-17)	Multiple linear regression basics: relationships, notation, estimation, inter-
	pretation.
3 (Jan. 20-24)	Assumptions of linear regression: assumptions, residual plots and violation
	detection, ethics workshop 1 (to be confirmed).
4 (Jan. 27-31)	Correcting assumptions: transformations, sampling distributions, and the
	first slight of polynomial regression and generalized linear models.
5 (Feb. 3-7)	MIDTERM TEST
6 (Feb. 10-14)	Inference in Linear Regression: hypothesis tests, confidence intervals, and
	prediction intervals.
(Feb. 17-21)	READING WEEK.
7 (Feb. 24-28)	Decomposition of variance I: sum of squares decomposition, ANOVA F test,
	partial F test.
8 (Mar. 3-7)	Decomposition of variance II: coefficients of determination, multicolinearity.
9 (Mar. 10-14)	Problematic observations: outliers, leverage points, influential points, detec-
	tion and impact, writing workshop 1 (to be confirmed).
10 (Mar. 17-21)	Model Building and Variable Selection: context, likelihood criteria, auto-
	mated methods, hypothesis tests.
11 (Mar. 24-28)	Model Validation: how to validate your models, data analysis process
	overview, writing workshop 2 (to be confirmed).
12 (Mar. 31 - Apr.	Review/Office Hour Class: final exam review and/or Q&A
4)	
Apr. 9-30	Final assessment period