

Course Name and Number: Fundamentals of Computational Math IS 605 Fall 2018

Credits: 3

Prerequisites: None

Course Description:

The course will provide an overview of the commonly used mathematical techniques in Data Analytics & Statistics. A lot of emphasis will be given to computational techniques and implementing math in programs. Students will be expected to do a fair bit of hands-on programming. We'll be primarily using R as our programming environment. All assignments will need to be submitted in as R-Markdown documents. Students can expect to walk away with a good understanding of the kinds of math they'll need to be successful in the exciting area of Data Analytics.

Program Learning Outcomes/Competencies addressed by the course:

1. Business Understanding. Apply frameworks and processes to build data-analytic solutions from an understanding of business goals.
2. Data Programming. Use industry standard statistical tools and simulation packages.
3. Foundational Math and Statistics. Emphasis on probability, statistics, and computational methods.
4. Data Understanding. Collect, describe, model, explore and verify data.
5. Prescriptive Modeling. Selecting prescriptive modeling techniques, generating test designs, building and assessing models.
6. Model Implementation and Deployment. Students will learn to implement mathematical models.
7. Presentation. Students will deliver presentations of their project results.

Learning Objectives

1. Apply linear algebra to solve data science problems.
2. Apply probability and statistics to solve data science problems.
3. Apply univariate and multivariate calculus to solve data science problems.

Assignments and grading:

1. Assignments: During the course, you will be completing a series of individual assignments, participating in forum discussions, and completing a final examination.

2. Examinations: This course has a single examination, a comprehensive final.

3. Grade Evaluation: Grades in this course are determined by the percentage of points obtained.

Assignment	Percentage
Homework	15 x 3% = 45%
Discussion Contribution	15 x 1% = 15%
Examination	1 x 40% = 40%
Total	100%

4. Grading Rubric: The grading rubrics follow.

- a) Each homework will be graded as {zero, check minus, check, and check plus}. The associated grades are {0%, 1%, 2%, 3%}.
- b) Discussions are applied analysis from the texts. You must post a response by Wednesday at midnight (ET) and respond to at least one of your colleagues' contributions by Saturday at midnight (ET), providing meaningful feedback on the analysis.

5. Late Policy for Homework: Late work is penalized 20% per day. All assignment due dates and times are shown in Blackboard.

6. Grades: I assign grades based on the following table.

Quality of Performance	Letter Grade	Range %	GPA/ Quality Pts.
Excellent - work is of exceptional quality	A	93 - 100	4
	A-	90 - 92.9	3.7
Good - work is above average	B+	87 - 89.9	3.3
Satisfactory	B	83 - 86.9	3
Below Average	B-	80 - 82.9	2.7
Poor	C+	77 - 79.9	2.3
	C	70 - 76.9	2
Failure	F	< 70	0

Course Materials (All Open Source)

<https://open.umn.edu/opentextbooks/BookDetail.aspx?bookId=21>

Introduction to Probability, Grinstead, C. Snell, J., 1997

<https://open.umn.edu/opentextbooks/BookDetail.aspx?bookId=5>

A First Course in Linear Algebra, Beezer, R., 2008

<https://open.umn.edu/opentextbooks/BookDetail.aspx?bookId=399>

Linear Regression Using R: An Introduction to Data Modeling, Lilja, D., 2016

<https://open.umn.edu/opentextbooks/BookDetail.aspx?bookId=198>

APEX Calculus, Hartman, G. 2014

Relevant Software and Other Tools:

Students should have R Studio & R installed in their computers. Relevant libraries that are required will be posted in the assignments and course materials. Students are expected to submit R-Markdown files for their assignments.

Course Meeting Time:

See the course website.

The Course schedule follows. (L=Linear Algebra text, P=Probability Text, C=Calculus Text, R=Regression)

Week of	Topics	Reading	Assignment (due Sunday at 11:59 PM ET)
27 Aug	Vectors, Matrices & Systems of Equations	L, Chapters SLE, V, M	HW1
3 Sep	Trace, Determinant, Factorization of Matrices	L, Chapters T, VS, D	HW2
10 Sep	Eigenvalues & Eigenvectors	L, Chapters E	HW3

17 Sep	Linear Transformations, Representations	L, Chapter s LT, R	HW4
24 Sep	Discrete and Continuous Probability Distributions	P, Chapters 1 and 2	HW5
1 Oct	Combinatorics and Conditional Probability	P, Chapters 3 and 4	HW6
8 Oct	Important Distributions / Densities, Expected Value & Variance	P, Chapters 5 and 6	HW7
15 Oct	Sums of Random Variables, Law of Large Numbers	P, Chapters 7 and 8	HW8
22 Oct	Central Limit Theorem & Generating Functions	P, Chapters 9 and 10	HW9
29 Oct	Markov Chains & Random Walks	P, Chapters 11 and 12	HW10
5 Nov	Regression Analysis in R	R, Chapters 1-3	HW11
12 Nov	Regression Analysis in R	R, Chapters 4-6	HW12
19 Nov	Univariate & Multivariate Calculus	C, As Needed	HW13
26 Nov	Taylor Series Approximations	C, Chapter 8	HW14
3 Dec	Functions of Several Variables	C, Chapter 12	HW15
3 Dec	Final Examination	All	Final Exam: Due 5 Dec

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Excellent - work is above and beyond class expectations	A-	90-92.9	3.7
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Failure	F	< 70	0