

# STAT 4360: Introduction to Statistical Learning

Fall 2023

**Instructor:** Jiayi Wang (jiayi.wang2@utdallas.edu)

**Time:** Monday, Wednesday 1:00pm-2:15pm

**Location:** GR 3.302

**Office Hours:** Monday & Wednesday 2:15pm - 3:15pm or by appointment (Please call me in Teams)

**Office Location:** SCI 1.172

**TA:** Nisansala Wickramasinghe (wickramasinghe.wickramasinghe@utdallas.edu)

**TA office hours:** Tuesday & Thursday 3:00pm - 4:00pm or by appointment

**TA office Location:** FN 3.118 E

## Course Overview

### Description

Statistical and machine learning methods will be taught at an introductory level. We will cover supervised learning methods including linear regression; classification; cross-validation; bootstrap; model selection and regularization; tree-based methods; and support vector machines. We will also learn some unsupervised learning methods including principal components analysis and clustering. Computer packages such as R or Python will be used for implementation of methods and data analysis.

### Prerequisites

STAT 3355 or equivalent.

### Learning Objectives

1. Understand basic statistical and machine learning methods.
2. Analyze data using basic statistical and machine learning methods.
3. Use software packages (R or Python) for data analysis.

## Textbook and Resources

- **Textbook:**

- Required: *An Introduction to Statistical Learning with Applications in R* by James, Witten, Hastie, and Tibshirani. Visit the book's website [https://hastie.su.domains/ISLR2/ISLRv2\\_website.pdf](https://hastie.su.domains/ISLR2/ISLRv2_website.pdf) to download a PDF of the book and other related materials.

- Others:

1. *The Elements of Statistical Learning* by Hastie, Tibshirani, and Friedman. Visit <https://web.stanford.edu/~hastie/ElemStatLearn> for a PDF of the book and other related materials.
2. *Machine Learning: A Probabilistic Perspective* by Murphy, MIT Press, 2012.
3. *An Introduction to R* (<https://cran.r-project.org/doc/manuals/r-release/R-intro.pdf>)

4. *Advanced R* (<http://adv-r.had.co.nz/>).

5. *Intro to Machine Learning with Python* (<https://python-course.eu/machine-learning/>).

- **Course Website:** Use eLearning ([elearning.utdallas.edu](http://elearning.utdallas.edu)) to access lecture notes, homework assignments, exams, and to check your performance. The lecture recordings are available on Teams (STAT 4360.0W1 Fall 23). You can find lecture recordings in the general channel.
  - The course materials are available to all students registered for this class as they are intended to supplement the classroom experience. These materials are for registered students' use only. Classroom materials may not be reproduced or shared with those not in class, or uploaded to other online environments except to implement an approved Office of Student Access Ability accommodation. Failure to comply with these University requirements is a violation of the Student Code of Conduct (<https://policy.utdallas.edu/utdsp5003>).

## Grading Policy

Course grade will be assigned as  $A^+$ (96-100),  $A$ (93-95),  $A^-$ (90-92),  $B^+$ (87-89),  $B$ (84-86),  $B^-$ (80-83),  $C^+$ (75-79),  $C$ (70-74),  $C^-$ (65-69),  $D^+$ (60-64),  $D$ (55-59),  $D^-$ (50-54) or  $F$ (0-49) with components and weights as follows.

\* These cutoffs may be slightly adjusted at the end of the semester.

### 1. Mini Projects (45%)

- Six mini projects will be assigned during the semester.
- The projects will need to be submitted through eLearning. They may not have equal weight.
- Some bonus questions may be included in the homework, which give extra credits.
- No late work will be accepted. Students who miss a homework due to an emergency or illness must provide appropriate documentation in a timely manner. Work submitted by a student as makeup work for an excused absence is not considered late work and is exempted from the late work policy.

### 2. Exams: Exam 1 (25%); Exam 2 (30%)

- The tests will be based on a specified range of course content to be announced in the class. They will involve multiple-choice problems. There may be some problems which require mathematical derivation. The use of a calculator is required.
- The tests are closed-book, closed-notes. A cheat sheet may be provided if deemed necessary.
- Some bonus questions may be included in the exams, which give extra credits.
- No make-up exam will be given unless there is a serious medical emergency and appropriate documentation is provided in a timely manner. It may be possible to take it early for reasons such as travel.

### 3. Homework (0%)

- Students are supposed to go through exercises in the textbook (*An Introduction to Statistical Learning with Applications in R*) according to the process of lectures by themselves. The answers of questions in the textbook are available online. Exam problems will have similar essence as the problems in the homework.

Incomplete grade may be possible only in the case of a documented serious medical emergency near the end of the semester, with at least 70% of work completed at an on-going passing grade.

## Tentative Course Schedule

Week	Topic	Textbook Chapters
1-3	Statistical Learning & Linear Regression	1-3
4-6	Classification & Resampling	4-5
7-9	Regularization & Non-linear regression	6-7
10-12	Tree-based methods & Support Vector machines	8-9
13-15	Unsupervised learning (and Deep learning if time permits)	12, 10

## Tentative Exam and Project Schedule

Due date	Item
Sep 6	Mini Project 1
Sep 20	Mini Project 2
Oct 11	Exam 1
Oct 18	Mini Project 3
Nov 1	Mini Project 4
Nov 15	Mini Project 5
Nov 29	Mini Project 6
Dec 6	Exam 2

## Other Policies

### Class Attendance and Participation

Regular and punctual class attendance is expected. Students who fail to attend class regularly are inviting scholastic difficulty. Regular class participation is expected. Students who fail to participate in class regularly are inviting scholastic difficulty. A portion of the grade for this course is directly tied to your participation in this class. It also includes engaging in group or other activities during class that solicit your feedback on homework assignments, readings, or materials covered in the lectures (and/or labs). Successful participation is defined as consistently adhering to University requirements, as presented in this syllabus. Failure to comply with these University requirements is a violation of the Student Code of Conduct.

### Academic Integrity

This creed was voted on by the UT Dallas student body in 2014. It is a standard that Comets choose to live by and encourage others to do the same:

“As a Comet, I pledge honesty, integrity, and service in all that I do.”

You can learn more about academic integrity, and your rights and responsibilities at <https://conduct.utdallas.edu/integrity/>.

### Academic Support Resources

The information contained in the following link lists the University’s academic support resources for all students. Please go to <http://go.utdallas.edu/academic-support-resources>.

### UT Dallas Syllabus Policies and Procedures

The information contained in the following link constitutes the University’s policies and procedures segment of the course syllabus. Please review the catalog sections regarding the credit/no credit or pass/fail grading option and withdrawal from class.

Please go to <http://go.utdallas.edu/syllabus-policies> for these policies.