Predictive Analytics in Healthcare HI 743

Department of Health Informatics and Administration Zilber College of Public Health University of Wisconsin - Milwaukee

January 23, 2025

Overview

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- 3.1 Syllabus Review
- 3.2 Weekly Itinerary

HI 743 - Predictive Analytics in Healthcare

- Instructor: Ryan Gallagher
- Time & Place: Thursdays 2:30PM 5:00PM in EH 103
- **Textbooks**: Fundamentals of Machine Learning for Predictive Data Analytics: Algorithms, Worked Examples, and Case Studies by John D. Kelleher, Brian Mac Namee and Aoife D' Arey
 - Introduction to Statistical Learning with Applications in R (Second Edition) by Gareth James, Daniela Witten, Trevor Hastie, Robert Tibshirani (Freely Distributed at statlearning.com)
- Required Materials: Textbook, A Computer (Laptop preferably), R (software), & RStudio (IDE for R)

About Me



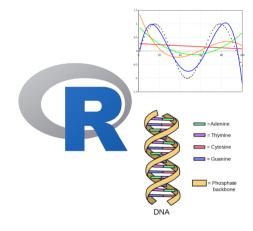




Biostatistician MCW/CW Advanced Genomics Lab

M.A. in Biostatistics & Data Science (MCW) B.S. in Statistics & Applied Physics, double major (UWEC)

- Biostatistics
 - Hypothesis Testing
 - Model Fitting
 - Differential Expression Analysis
- Bioinformatics
 - Pipeline Development
 - Genomic Data Analysis
- Programming
 - R for Statistics
 - Python for Data Processing Pipeline
 - Bash for Linux Cluster



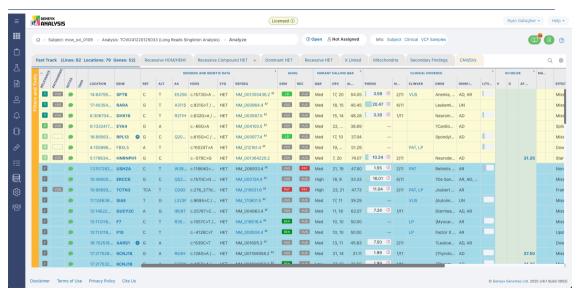
I work on my lab's **Structural Variance Initiative**, which:

- 1. Recruits & Consents Pediatrics Patients
- 2. Sequences Entire Genome from Biological Sample
 - 2.1 Oxford Nanopore Long-Read Sequencing
 - 2.2 Bionano Optical Genome Mapping
- 3. Processes Raw Sequencing Reads (Python + Bash)
 - 3.1 Sequencing Data Management ($\sim 1.5 \text{TB}$)
 - 3.2 Raw Sequence Alignment + QC
 - 3.3 Identify SNPs + Small INDELs, CNVs, SVs
- 4. Annotate Identified Variants
- 5. Identify Potential Disease Causing Variants









About You

Syllabus

Lets look at our syllabus

Week 2 - Course Motivations

Lecture:

- What is Predictive Data Analytics?
- What is Machine Learning?
- How does Machine Learning work?
- How is Machine Learning relevant to my field of interest?

Week 3 - Intro to R Language

Lecture:

- What is R?
- Why use R?
- Brief History of R

- Hands-on Tutorial in R
- Data Manipulation with *Tidyverse* Package
- R Figures

Week 4 - Data Exploration

Lecture:

- Data Quality
- Data Preparation
- Missing Values / Outliers

- Hands-On Data Cleaning
- Data Visualization

Week 5 - Error Based Regression

Lecture:

- Simple Linear Regression
- Multiple Linear Regression & Interpretation
- Nonlinear Regression & Interpretation

- Multiple & Nonlinear Regression
- Interpreting R Output
- Introduction to RMarkdown for Reporting

Week 6 - Classification / Logistic Regression

Lecture:

- Logistic Regression
- Generalized Linear Models
- Multinomial Logistic Regression

- Logistic Regression in R
- glm() function

Week 7 - Case Study (or Classification Continued)

This week will either be a continuation of the Classification methods.

Or

We will look at a case study which applies Classification methods to real-world data.

Project Idea is Due Following Week.

Week 8 - Information Based Learning

Lecture:

- Basics of Decision Trees
- Random Forest, Bagging, Boosting, etc..
- Tree "Pruning"

- Fitting Classification Trees
- Fitting Regression Trees
- Bagging Algorithms

Week 9 - SPRING BREAK!

No Class!

Week 10 - Unsupervised / Similarity-based Learning

Lecture:

- Principal Component Analysis
- Clustering Methods
- Challenges of Unsupervised Learning

- K-Means Clustering
- PCA Example

Week 11 - Probability / Bayesian Methods in ML

Lecture:

- Bayes' Thoerem
- Naive Bayes Model
- Probability Density Functions

R Lab:

Naive Bayes Worked Examples

Week 12 - Neural Networks and Deep Learning

Lecture:

- What is Deep Learning?
- Single Layer Neural Networks
- Multilayer Neural Networks
- Convolutional Neural Networks

R Lab:

• Single, Multi, and Convolutional Neural Networks Applied

Week 13 - Model Evaluation

Lecture:

- Model Performance Measures
- Resampling Methods
- Cross Validation

- Validation Set Approach
- Cross-Validation
- Bootstrapping

Week 14 - Practical ML Application

Lecture:

- When to use Machine Learning vs. Typical Methods
- Choosing a Machine Learning Approach
- Matching Approaches to Data

Lab portion will be dedicated to working on Final Project.

Week 15 - Work on Final Project

Classtime will be dedicated to working on Final Project.

Project Due May 9th.