PH 712 Probability and Statistical Inference

Part O: Syllabus Review, calculus basics & R Basics

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Syllabus

Contact

- Instructor: Zhiyang (Gee) Zhou, PhD, Asst. Prof. (Biostatistics)
 - Email: zhou67@uwm.edu
 - -Homepage: zhiyanggeezhou.github.io
- Lectures
 - Mon/Wed 11:30–12:45 at Zilber 110
- Office Hours
 - By appointment

Grading

- Assignments (30%; 4 or 5)
 - Digital copies submitted
 - Attaching (if applicable) both outputs and source codes
 - Including necessary interpretation
 - Organized in a clear and readable way
 - Accepting NO late submission
- Midterm (35%; Oct. 23 or 28)
 - 75-minute
 - Open-book
 - In-person
- Final (35%; Dec. 16, tentatively)
 - 2-hour
 - Open-book
 - In-person

Materials

- Reading list (recommended but NOT required)
 - [HMC] R. Hogg, J. McKean, & A. Craig. (2018). Introduction to Mathematical Statistics, 8th Ed. Boston: Pearson.
 - [CB] G. Casella & R. L. Berger. (2002). Statistical Inference, 2nd Ed. Pacific Grove: Thompson Learning..

 D. Salsburg (2001). The Lady Tasting Tea: How Statistics Revolutionized Science in the Twentieth Century. New York: WH Freeman.

- Lecture notes and beyond
 - Posted at Canvas and zhiyanggeezhou.github.io
 - Subject to update without prior notice

"All models are wrong, but some are useful."

- G. E. P. Box. (1976). Journal of the American Statistical Association, 71:791–799.

What is a statistical model?

- Two types of statistical models (Breiman, 2001)
 - Stochastic model vs. machine learning model (PH 812 Statistical Learning & Data Mining)
- Stochastic model: the distribution of random variables (RVs) of interest
 - Recall the linear regression and logit regression (PH 711 Intermediate Biostatistics) \mathbf{P}
 - Parametric vs non-parametric vs semi-parametric

Statistical modelling

Confirmed RVs of interest \rightarrow Data collection and cleaning \rightarrow Specified models \rightarrow Model fitting and inference \rightarrow Interpretation

Statistical inference

• To figure out the underlying true model - E.g., is the RV distributed as $\mathcal{N}(0,1)$?

Topics to be covered

- Basic concepts of univariate random variable;
- Variable transformation;
- Sufficiency;
- Consistency and limiting distributions;
- Point estimation finite/large samples;
- Hypothesis testing finite/large samples;
- Interval estimation with finite/large samples
- and so forth

Introduction to Univariate Differentiation and Integration

Univariate differentiation

- The process of finding the derivative of a univariate function
 - Interpretation for derivatives:
 - * Geometric: the derivative at a point gives the slope of the tangent line to the curve at that point
 - * Practical: measuring the rate of changes, i.e., how a function changes as its input changes
 E.g. the velocity being the derivative of position with respect to time
- Basic rules of differentiation
 - Refer to sections "Rules for basic functions" and "Rules for combined functions" at https: //en.wikipedia.org/wiki/Derivative#Rules_of_computation

Univariate integration

- The process of finding the integral of a univariate function, reversing the differentiation process
 - Interpretation for integrals:
 - * Geometric: representing the area under the curve of a function over a given interval
 - * Practical: accumulating tiny quantities
 - · E.g. the traveled distance being the integral of the velocity function of time
- Indefinite integrals
 - Representing the general form of an antiderivative of a function
 - Including a constant C, as integrating a function reverses the differentiation process, but there could be multiple functions that differentiate to the same result (differing by a constant)
 - * E.g., $x^2 + 1$ and $x^2 + 2$ share the identical derivative 2x and hence $\int 2x dx = x^2 + C$ with indefinite C.
- Definite integrals
 - Calculating the net area under the curve of a function between two specific limits
 - Dropping the indefinite C in the corresponding definite integral and plugging in limits
- * E.g., $\int_a^b 2x dx = x^2 |_{x=a}^{x=b} = b^2 a^2$ Basic rules of indefinite integration
- - Refer to the sections "Rational functions", "Exponential functions" and "Logarithms" at https: //en.wikipedia.org/wiki/Lists of integrals

R basics

- Previously covered in any courses? PH702? PH718?
- Installation
 - download and install BASE R from https://cran.r-project.org
 - download and install Rstudio from https://www.rstudio.com
 - download and install packages via *Rstudio*
- Working directory
 - When you ask R to open a certain file, it will look in the working directory for this file.
 - When you tell R to save a data file or figure, it will save it in the working directory.

```
getwd()
mainDir <- "c:/"</pre>
subDir <- "stat3690"</pre>
dir.create(file.path(mainDir, subDir), showWarnings = FALSE)
setwd(file.path(mainDir, subDir))
```

- Packages
 - installation: install.packages()
 - loading: library()

```
install.packages('nlme')
library(nlme)
```

- Help manual: help(), ?, google, stackoverflow, etc.
- *R* is free but not cheap

- Open-source
- Citing packages
- NO quality control
- Requiring statistical sophistication
- Time-consuming to become a master
- References for the fusion of ${\cal R}$ and statistical methds
 - G. James, D. Witten, T. Hastie and R. Tibshirani (2023) An Introduction to Statistical Learning: with Applications in R, 2nd Ed.
 - M. L. Rizzo (2019) Statistical Computing with R, 2nd Ed.
 - O. Jones, R. Maillardet, A. Robinson (2014) Introduction to Scientific Programming and Simulation Using R, 2nd Ed.

Courses online

- https://www.pluralsight.com/search?q=R
-
- Data types: let str() or class() tell you
 - numbers (integer, real, or complex)
 - characters ("abc")
 - logical (TRUE or FALSE)
 - date & time
 - factor (commonly encountered in this course)
 - NA (different from Inf, "'', 0, NaN etc.)
- Data structures: let str() or class() tell you
 - vector: an ordered collection of the same data type
 - matrix: two-dimensional collection of the same data type
 - $-\,$ array: more than two dimensional collection of the same data type
 - data frame: collection of vectors of same length but of arbitrary data types
 - list: collection of arbitrary objects
- Data input and output
 - create
 - * vector: c(), seq(), rep()
 - * matrix: matrix(), cbind(), rbind()
 - $\ast\,$ data frame
 - output: write.table(), write.csv(), write.xlsx()
 - import: read.table(), read.csv(), read.xlsx()
 - * header: whether or not assume variable names in first row
 - * stringsAsFactors: whether or not convert character string to factors
 - scan(): a more general way to input data
 - save.image() and load(): save and reload workspace
 - source(): run R script
- Parenthesis in ${\cal R}$
 - paenthesis () to enclose inputs for functions
 - square brackets [], [[]] for indexing
 - braces {} to enclose for loop or statements such as if or ifelse

```
# Create numeric vectors
v1 = c(1,2,3); v1
v2 = seq(4, 6, by=0.5); v2
v3 = c(v1, v2); v3
v4 = rep(pi, 5); v4
v5 = rep(v1,2); v5
v6 = rep(v1, each=2); v6
# Create Character vector
v7 <- c("one", "two", "three"); v7
# Select specific elements
v1[c(1,3)]
v7[2]
# Create matrices
m1 = matrix(-1:4, nrow=2); m1
m2 = matrix(-1:4, nrow=2, byrow=TRUE); m2
m3 = cbind(m1,m2); m3
(m4 = cbind(m1, m2))
# Create a data frame
e <- c(1,2,3,4)
f <- c("red", "white", "black", NA)</pre>
g <- c(TRUE, TRUE, TRUE, FALSE)
mydata <- data.frame(e,f,g)</pre>
names(mydata) <- c("ID", "Color", "Passed") # name variable</pre>
mydata
# Output
write.csv(mydata, file='mydata.csv', row.names=F)
# Import
(simple = read.csv('mydata.csv', header=TRUE, stringsAsFactors=TRUE))
class(simple)
class(simple[[1]])
class(simple[[2]])
class(simple[[3]])
(simple = read.csv('mydata.csv', header=FALSE, stringsAsFactors=FALSE))
class(simple[[3]])
# EXERCISE
# Create a matrix with 2 rows and 6 columns such that it contains the numbers 1,4,7,...,34.
# Make sure the numbers are increasing row-wise; ie, 4 should be in the second column.
# Use the seq() function to generate the numbers. Do NOT type them out by hand!
# ANSWER
matrix(seq(from=1, to=34, by=3), nrow=2)
```

• Elementary arithmetic operators

-+, -, *, /, ^

- $-\log, \exp, \sin, \cos, \tan,$ sqrt
- FALSE and TRUE becoming 0 and 1, respectively
- sum(), mean(), median(), min(), max(), var(), sd(), summary()
- Matrix calculation
 - element-wise multiplication: A * B

- matrix multiplication: A %*% B
- singlar value decomposition: eigen(A)
- Loops: for() and while()
- Probabilities
 - normal distribution: dnorm(), pnorm(), qnorm(), rnorm()
 - uniform distribution: dunif(), punif(), qunif(), runif()
 - multivariate normal distribution: dmvnorm(), rmvnorm()

```
# Generate two datasets
set.seed(100)
x = rnorm(250, mean=0, sd=1)
y = runif(250, -3, 3)
```

- Basic graphics
 - strip chart, histogram, box plot, scatter plot
 - Package ggplot2 (RECOMMENDED)

```
# Strip chart
stripchart(x)
# Histogram
hist(x)
# Box plot
boxplot(x)
# Side-bu-side box plot
xy = data.frame(normal=x, uniform=y)
boxplot(xy)
# Scatter Plot with fitted line
plot(x, y ,xlab="x", ylab = "y", main = "scatter plot between x and y")
abline(lm(y~x))
# EXERCISE
# Play with a data set called "Gasoline" included in the package "nlme".
# 1. How many variables are contained in this data set? What are they?
# 2. Generate a histogram of yield and calculate the five number summary for it.
# What is the shape of the histogram?
# 3. Generate side-by-side boxplots,
  comparing the temperature at which all the gasoline is vaporized (endpoint) to sample.
#
# Does it seem that the temperatures at which all the gasoline is vaporized differ by sample?
# 4. Generate a plot that illustrates the relationship between yield and endpoint.
# Describe the relationship between these two variables.
# 5. What if the plot created in Q4 were separated by sample?
# Generate a plot of yield v.s. endpoint, separated by sample.
# ANSWER
attach(nlme::Gasoline)
```

```
# 1. Six variables: yield, endpoint, sample, API, vapor, ASTM
# 2.
summary(yield)
hist(yield, nclass=50)
# 3.
boxplot(endpoint ~ Sample)
anova(lm(endpoint ~ Sample))
# 4.
plot(x=endpoint, y=yield, xlab="endpoint",ylab = "yield",
      main = "scatter plot between endpoint and yield")
abline(lm(yield~endpoint))
# 5.
par(mfrow=c(2,5))
for (i in 1:10){
 plot(x=endpoint[Sample==i], y=yield[Sample==i], xlab='', ylab='', main=paste('Sample=', i))
 abline(lm(yield[Sample==i]~endpoint[Sample==i]))
}
# Do not forget to detach the dataset after using it.
detach(nlme::Gasoline)
```