Day-by-day Objectives for Math 300R

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Lessons 1-18

As done in Fall 2020. Possible revisions to those lessons is not a topic of this proposal.

Lesson 19: Decisions with data (nti)

- 1. Distinguish between the two settings for decision-making:
 - a. **Prediction**: predict an outcome for an individual
 - b. **Relationship**: characterize a relationship with an eye toward intervention or a better understanding of how a mechanism works.
- 2. Given a research question, identify whether it corresponds to a prediction setting or a relationship setting.

Lesson 20: Reality versus gaming (nti)

- 1. Understand that gaming is a way of improving our skills and identifying potential opportunities and problems.
- 2. Enumerate the four stages of the games we will use and identify which ones correspond to non-gaming, real-world work with data.

The Four Stages (to be moved to NTI)

- i. building the deck: Instructors provide a simulation of a mechanism that generates rows of a data frame.
- ii. the deal: Some of these rows will be dealt to you, constituting the data you have to work with. [real-world]
- iii. the play: Build models and extract results. [real-world]
- iv. the reveal: Compare your results from (iii) either to the mechanism given in (i) or to more data generated by the simulation.
- 1. Distinguish between a sample, a row, and a sample of samples.

Lesson 21: DAGs, noise, and simulation (nti)

- 1. Determine whether a proposed graph is directed and acyclic.
- 2. Read notation to identify response variable, explanatory variable, covariates, and effect sizes.
- 3. Characterize the magnitude of random noise.
- 4. Generate data from simulations and summarize variables individually.

Lesson 22: Sampling variation (nti)

- 1. Implement on the computer a procedure to generate a sample, calculate a regression model, and produce a summary.
- 2. Iterate the procedure and collect the summaries across iterations.
- 3. Graphically display the distribution of summaries and generate a compact numerical description ("confidence interval") of the sampling distribution.

Lesson 23: Estimate sampling variation from a single sample (nti)

- 1. Use bootstrapping to estimate sampling variation.
- 2. Infer sampling variation from a regression table.

Lesson 24: Effect size (nti)

- 1. Estimate an effect size from a regression model of the two variables.
- 2. Construct a confidence interval on the effect size.
- 3. Evaluate whether confidence interval indicates that estimated effect size is consistent with simulation.
- 4. Understand and use scaling of confidence interval length as a function of n.

Lesson 25: Mechanics of prediction (nti)

- 1. Given a sample from a DAG simulation, construct a predictor function for a specified response variable.
- 2. Use the predictor function to estimate prediction error on a given DAG sample and summarize with root mean square (RMS) error.
- 3. Distinguish between in-sample and out-of-sample prediction estimates of prediction error.

Lesson 26: Constructing a prediction interval

1. Identify the two components that make up a prediction error, one that scales with n and the other that doesn't.

Lesson 27: Covariates

- 1. Show that including covariates in a prediction model always reduces in-sample mean square residual, but may not reduce residuals out-of-sample.
- 2. Given regression coefficients, calculate model degrees of freedom and residual degrees of freedom.
- 3. Calculate amount of in-sample mean square error reduction to be expected with a useless (random) covariate. (Residual sum of squares divided by residual degrees of freedom.)

Lesson 28: Covariates eat variance

- 1. Construct F statistic as ratio of incremental increase in model mean square due to model term(s) divided by residual mean square.
- 2. Use software to construct ANOVA report and correctly interpret F statistics for prediction model term selection.

Lesson 29: Confounding

- 1. Identify confounding in a DAG
- 2. Choose whether to include covariate depending on form of DAG

Lesson 30: Non-causal correlation

- 1. Distinguish "common cause" and "collider" forms of DAG.
- 2. Construct appropriate DAG to match a narrative hypothesis.

Lesson 31: Experiment and random assignment

- 1. Properly use nomenclature of experiment.
- 2. Correctly re-draw DAG for an ideal experimental intervention.
- 3. Use blocking to set assignment to treatment or control.

Lesson 32: Measuring and accumulating risk

- 1. Distinguish between absolute and relative risk and identify when a change in risk is being presented as absolute or relative.
- 2. Calculate and correctly interpret other presentations of differences in risk: population attributable fraction, NTT, odds ratio.
- 3. Interpret effect size as stated in log odds.

Lesson 33: Constructing a classifier

- 1. Build a classifier from case-control data.
- 2. Cross-tabulate classifier results versus true state. Evaluate false-positive rate, false-negative rate, accuracy.
- 3. Calculate different forms of conditional probability: p(A|B) versus p(B|A) and identify which form of conditional probability is useful for prediction of an individual's outcome.

Lesson 34: Accounting for prevalence

- 1. Explain why case-control data may not give an proper measure of "prevalence."
- 2. Convert

Lesson 35: Hypothesis testing

- 1. Understand and use properly hypothesis testing nomenclature: test statistic, sampling distribution under the null, Type-1 and Type-2 error, rejection threshold, p-value
- 2. Contrast hypothesis testing versus Bayesian framework.

Lesson 36: Calculating a p-value

- 1. The permutation test
- 2. Interpret correctly from regression/ANOVA reports
- 3. Traditional names for hypothesis tests in different "textbook" settings.
- 4. Distinguish between p-value and effect size, that is, "significance" and "substance."

Lesson 37: False discovery with hypothesis testing

- 1. Identify signs of false discovery in a research paper.
- 2. Estimate how overall p-value should change when study is replicated.

Alternative 1

Theme: Classifiers: ROC and loss function

Alternative 2

Theme: Accumulating risk: Logistic regression