# STAT/CSSS 564: Assignment 3

May 2nd, 2017

- 1. Fork this repository to your account
- 2. Edit the file solutions.Rmd with your solutions to the problems.
- 3. Submit a pull request to have it graded. Include either or both a HTML and PDF file.

For updates and questions follow the Slack channel: #assignment3.

This assignment will require the following R packages:

library("rstan")

#### Statistical Simulation

Read @KingTomzWittenberg2000a. They propose a statistical simulation approach for interpreting statistical analysis; see section "Simulation-Based Approaches to Interpretation". Compare and contrast this to a full Bayesian approach.

#### Student-t Prior

The robust regression with Student-t error example uses the following prior on the degrees of freedom parameter.

$$\nu \sim Gamma(2, 0.01)$$

The Student-t distribution is used because it has wider tails and thus is less sensitive to outliers than a normal distribution. However, the researcher generally has no information about the value of the degrees of freedom.

- 1. Plot this prior distribution, and the values of the 5th and 95th quantiles. You can use dgamma(x, 2, scale = 0.01) and qgamma(x, 2, scale = 0.01). What is
- 2. Additionally, the prior is truncated at 2. Why? Hint: What moments of the Student-t distribution are not-defined for values between 2.

#### Student-t as a Mixture of Normals

The Student-t distribution is a scale mixture of normals.<sup>1</sup> This means that a Student-t distribution can be represented as normal distributions in which the variances are drawn from different distributions. Suppose X is distributed Student-t with degrees of freedom  $\nu$ , location  $\mu$ , and scale  $\sigma$ ,

$$X \sim t_{\nu}(\mu, \sigma).$$

Samples from Y can be drawn by

$$x_i \sim N(\mu, \lambda_i^2 \gamma^2)$$

If the local variance parameters are distributed inverse-gamma

$$1/\lambda^2 \sim \text{Gamma}(\nu/2, \nu/2).$$

Many distributions used in regression shrinkage: Double Exponential (Laplace), and Hierarchical Shrinkage (Horseshoe), have this representation.

<sup>&</sup>lt;sup>1</sup>mix

You can draw a sample from this:

```
df <- 10
n <- 1000
sigma <- rgamma(n, 0.5 * df, 0.5 * df)
x <- rnorm(n, sd = sqrt(1 / sigma ^ 2))</pre>
```

Plot samples drawn in this way against either samples or theoretical values of the Student-t distribution. Try a few values of the degrees of freedom. Try something small (3) and large (100).

You can draw samples directly from a Student-t with rt. A quantile-quantile plot (geom\_qq) or a density plot with the function (geom\_density and stat\_function).

Note: there isn't a right answer to this. Well, actually, there is, and you know it. They are equivalent, a proof is in the link. So for credit, do a little work, and show it. This pattern appears often, so wrap your head around it.

### Separation

Continue what was covered in class.

#### Transformations of Coefficients

@Rainey2016b notes that unbiased estimators of parameters does not imply that transformations of those parameters are unbiased estimators.

## Poisson and Negative Binomial Example

Fill in