

# Test 2 Info Sheet

*Stat 341 – Spring 2017*

## Logistics

- In-class test: Friday, April 28
- Take-home test: Due Monday, May 1
- Both portions cover Chapters 1-10 and 12 of *Statistical Rethinking*, emphasizing material since Chapter 5. (Note: We did not cover the portion of Chapter 10 dealing with Poisson models, and out coverage of Chapter 9 was quite light.)

## Topics

This list isn't meant to be exhaustive, but I hope it is useful in preparing for the test.

## Techniques

1. Creating Numerical and Graphical Summaries
  - You have your choice of `lattice`, `ggplot2`, or `ggformula`, but I've been leaning heavily on `ggformula`, since it does what we need most easily. If you use `lattice`, the migration should be pretty easy. If you already know (or are learning) `ggplot2`, it's fine to continue with that. (Anything you can do with `ggformula` can be done with `ggplot2`, since `ggformula` just translates a new interface into `ggplot2` code.)
2. Fitting models with `map()` and `map2stan()`
3. Posterior sampling
  - a. `extract.samples()`
  - b. `link()`
  - c. `sim()`
  - d. creating counterfactual data (for use with `link()` and `sim()`, for example)
4. Creating an ensemble model with `ensemble()`
5. Examining models
  - a. `precis(model)`, `plot(precis(model))`
  - b. `coeftab(model)`, `plot(coeftab(model))`
  - c. `compare(...)`, `plot(compare(...))`
  - d. `plot(model)` to look at chains of a Stan model
  - e. `pairs(model)`
  - f. `WAIC(model)`, `DIC(model)`, `logLik(model)`
  - g. `show(model)`
6. A non-exhaustive list of R commands you should know
  - Plotting commands (`lattice`, `ggplot2`, or `ggformula` versions) for the types of plots we have been making.
  - `apply()`
  - working with data frames: `data_frame()`, `expand.grid()`, `mutate()`, `filter()`, `group_by()`, `summarise()`

- distributions: `dnorm()`, `dbinom()`, `dunif()`, `rnorm()`, `rbinom()`, `runif()`
- miscellaneous: `zscore()`, `str()`, `head()`, `sample()`, `coerce_index()`

## Concepts

1. posterior  $\propto$  prior  $\cdot$  likelihood
2. Overfitting/Underfitting
3. Information Criteria (AIC, DIC, WAIC)
  - a. how to calculate information and entropy
  - b. effective number of parameters
  - c. Know the connections among WAIC and pWAIC from the `compare()` output, and `logLik(model)`
4. What posterior sampling is and what can be done with posterior samples
5. Potential problems with quadratic approximation (`map()`) and why Stan often works better; potential problems with Stan and how to detect them.
6. Metropolis Algorithm
7. Creating models
  - selecting priors that are reasonable
  - creating a model relationship (equation with variables and parameters)
  - linear models, including interaction effects
  - generalized linear models (link functions, etc.)
  - handling categorical variables
  - multilevel models (aka hierarchical models, aka random effects models)
8. Interpreting models
  - interpreting model coefficients (and their uncertainty).
  - interpreting model predictions (and their uncertainty).
  - plots that help visualize what a model “thinks”