

The New School for Social Research  
Advanced Econometrics 1  
Spring 2017  
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### Assignment 4

Due Oct 23 (Mon) 6:00 pm

1. Intro to Rstan: Read the attached paper *Stan: A probabilistic programming language for Bayesian inference and optimization* and answer the following questions
  - (a) Generate fake data ( $N=20$ ) from a normal distribution with mean  $\mu$  of 2.5 and standard deviation  $\sigma$  of 2. Use `set.seed(102)`.
  - (b) Fit a normal model to the simulated data using Stan. Try with 1) a non-informative prior (a uniform prior), 2) a normal priors for  $\mu$  with mean of 2.5 and standard deviation of 15 (keep the uniform prior for  $\sigma$ ), and 3) a normal prior for  $\mu$  with mean of 15 and standard deviation of 2. 1000 iterations (500 warm-ups) and 3 chains are preferred.  
\*A normal model and a normal prior can be conveniently coded by using a built-in sampling distribution `normal()` in Stan.
  - (c) Compare the three different log-posterior densities by reporting the mean, sd and a few quantiles of estimated parameters  $(\mu, \sigma)$ . You might want to use `print()` function.
  - (d) Report  $\hat{R}$  and check if chains are mixed properly. Visualize the result by using `plot()` function.
  - (e) Explain in a few sentences in which way different priors affect the posterior density of parameters in this example.
2. Programming Metropolis-Hasting algorithm
  - (a) Generate fake data ( $N=50$ ) from a normal distribution with mean  $\mu$  of 2 and standard deviation  $\sigma$  of 1. Use `set.seed(102)`. Fit a normal model to the simulated data using Metropolis-Hasting algorithm. Report the posterior distribution of  $\mu$  and  $\sigma$ , and the convergence plot of the parameters.