Group Reflection: Adaptive Rejection Sampling STAT 243 Final Project

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This short paper serves as an interpretory document to supplement the attached R code–file names go here–that implement the Adaptive Rejection Sampling (ARS) proposed by Gilks and Wild (1992) and then test the aforementioned ARS code.

1 Initial Approach

We initally began this project by ensuring that each of us understood all portions of the Gilks and Wild ARS algorithm. Even though we would not directly be coding everything, we felt it was important for every group member to be versed in the entire project, particularly because we were already operating as a 3-person group and Michelle was going to be attending a conference and generally unavailable to meet and do extensive amounts of work the week before the project due date. Any one of us needed to be able to jump in and do any step of the project. Following discussions of the overall ARS algorithm, we began the coding process, trying to keep modularity in mind as we went ahead.

2 The Coding Process

2.1 ARS Code

We first decided to modularly encode in functions all of the various equations defined in the paper. Matt started off the group by writing functions to calculate the log of the user-specified density, h and h'. Michelle and Ying then used those functions to write code to calculate functions 1–4 as defined in Gilks & Wild based on starting abcissae defined by the user.

Then came the tricky part – determining how to sample from the piecwise upper hull and coding the selfupdating feature of the ARS algorithm. We initially began the process by attempting to generate extremely long vectors of values along each of the piecewise functions and then use the sample() function to sample one of those points; however, we realized that the costs of using such an simulation to sample would be quite large, particularly for unbounded densities (for how can you generate a vector of values from - ∞ to a specific abcissa?).

After some work, Ying discovered an analytical approach to sampling relying on the CDF, and this approach is what is encoded into the ARS code. First determine the area under each of the piecewise upper hulls (the intersections between each piece denoted by z_*), and normalize such that the total area is equivalent to 1. Then generate $q \sim U(0,1)$. Use the relative normalized areas and this q to determine from which of the upper hull pieces to sample, and then subtract the areas of any of the pieces before this piece from q to get some area A. A is the area under the chosen piece from its leftmost z_* value to the desired x^* sampled from the upper bound, or equivalently:

$$A = \int_{z_*}^{x^*} \frac{e^{ax+b}}{c} dx$$

where c is the total area under all of the piecewise upper hulls before normalization and ax + b is the equation of the chosen piece of the upper hull. Subbing in the upper hull calculation from Gilks and Wild and solving the integral gives:

$$A = \int_{z_*}^{x^*} \frac{e^{ax+b}}{c} dx$$

$$= \int_{z_*}^{x^*} \frac{e^{a(x-x_j)+b}}{c} dx$$

$$= \frac{e^{b-ax_j}}{c} \int_{z_*}^{x^*} e^{ax} dx$$

$$= \frac{e^{b-ax_j}}{c} \frac{1}{a} e^{ax} \Big|_{z_*}^{x^*}$$

$$= \frac{e^{b-ax_j}}{c} \frac{1}{a} \left(e^{ax^*} - e^{az_*} \right)$$

$$x^* = \frac{1}{a} \log \left(A \frac{ca}{e^{b-ax_j}} + e^{az_*} \right)$$

After analytically solving for this sampling method, Ying implemented it into the code. Then Michelle and Matt looped the code through, adding the specifics of sampling and updating steps defined by Gilks and Wild and having the code terminate once the desired sample size has been accepted by the algorithm. Finally, Michelle wrote a nice wrapper function around all the code, thus creating our version of the ARS algorithm.

Our ARS function first requests the user to input a density function, two initial boundary points, and the number of points required to accept for a complete simulation of the density function. The user then much run the ARS wrapper function and the ARStesting function first to initialize these functions, and then can run the code to produce the final samples and plots.

2.2 Testing Code

We approached the testing section of our project similarly to how we approached encoding the ARS algorithm. Michelle began the testing section of the code, which we then broke out into its own function, thinking that we should keep the ARS algorithm and the code testing separate for modularity concerns. We first produce a series of plots comparing an ARS sample from the standard normal to the *rnorm* function built into R. The function prompts the user to make a decision as to whether the plots look similar enough to pass the test; the code proceeds according to the user input.

If the user decides to pass the first test, the testing algorithm compares the empirical CDFs of samples generated from the ARS test to a theoretical distribution using the Kolmogorov-Smirnov (K-S) test. The Kolmogorov-Smirnov (K-S) tests for equality of two continuous probability distributions, with the null hypothesis, in our case, being that the two CDFs are from the same distribution. The function passes or fails this portion of the testing algorithm based on a p-value cutoff, and produces plots as well as status messages to the user. Bsed on the user input, and the final p-value, similarly between distributions can be confirmed. Matt came up with the idea of using the K-S test, and Michelle initially coded the section.

Non-log concave test: Include text here after we test a non-log concave function.

Log concave test with the gamma function: Include test here with a different log-concave function.

3 Reflection and Group Dynamics

We broke up the work in such a way that a single person started a step of the project, pushed his or her work to the GitHub repository, and solicited input, adjustments, and edits from the other team members. In the above section (2), we mentioned individuals who started certain portions of the code; in addition, Matt took the lead as primary, initial author of this reflection paper. However, it should be noted that everyone was involved in every piece of this project, and nothing can be uniquely attributed to a single group member. Our team collaborated using GitHub with a naming convention for all files and folders in the archive. Our code also follows standard naming convention practices. We also communicated daily by email to update each other on our progress when the next stage of work was ready for the next person. The final pieces that we are turning in truly represent a full group effort.

R Code submitted on bSpace by: XXXXXXX