

Coursework Project Part 1

ST2195-Programming for Data Science

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1. Part 1a)

For part 1(a), we'll implement the Metropolis-Hastings algorithm to generate samples from the given probability density function $f(x) = \frac{1}{2}\exp(-|x|)$.

Steps that I will be taking to implement:

N=10000: Total number of samples.

s=1: Standard deviation for the proposal distribution.

We will define the target function $f(x)$ and its logarithm for numerical stability and run the Metropolis-Hastings algorithm with the given proposal distribution. Then we will create a histogram and kernel density estimate (KDE) of the generated samples. Finally, we will overlay the graph of $f(x)$ on the histogram and KDE plot.

Calculate and print the sample mean and standard deviation of the generated samples. I imported libraries such as NumPy, matplotlib and seaborn to help me with calculations and plotting of the histogram.

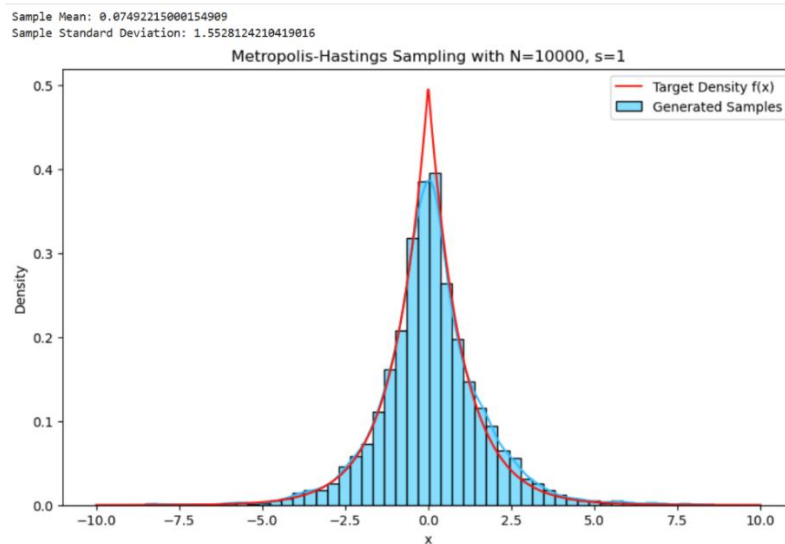


Figure 1: The result of the code(python) part a

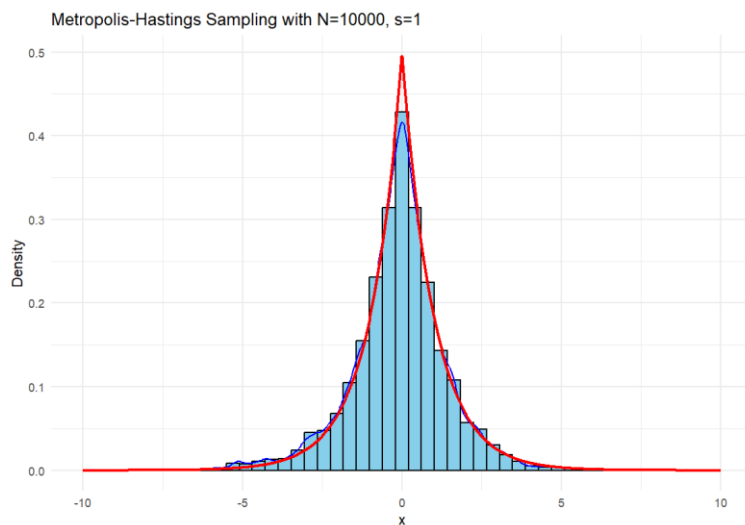


Figure 2: The result of the code (R) part a

2. Part 1b)

2.1. Computing the Convergence Diagnostic \hat{R}

For part (b), we'll calculate the \hat{R} diagnostic to verify convergence.

Steps that we will take includes generating multiple chains, $J=4$ chains, each with $N=2000$ samples and different initial values. Next, calculate within-chain means M_j and variances V_j . Then we will calculate the overall mean M and between-chain variance B . We will then compute \hat{R} using the provided formula and plot \hat{R} over a range of s values. For the coding, I used R and used libraries such as ggplot2 for the histogram

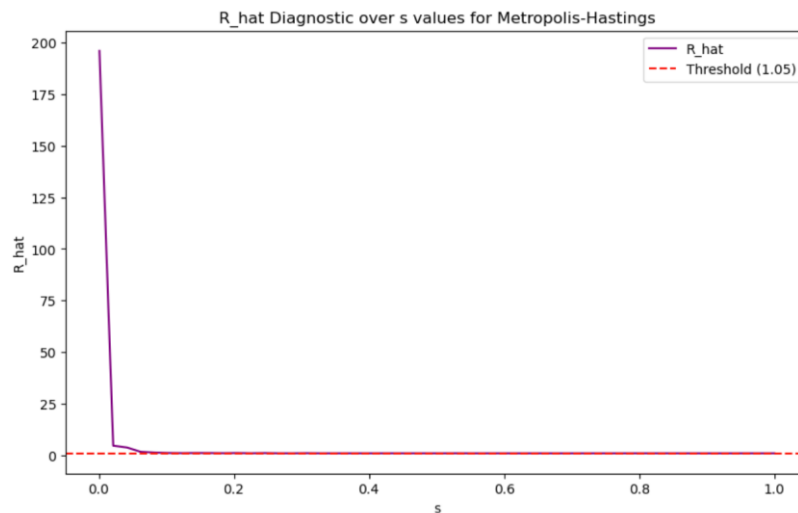


Figure 3: The result of the code(python) part b

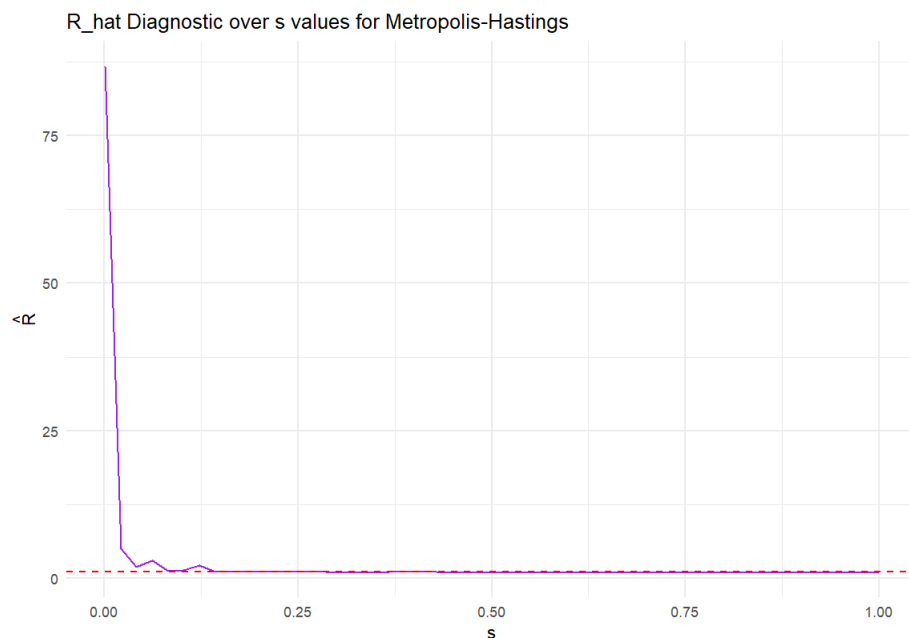


Figure 4: The result of the code(R) part b