## correct-q2

June 21, 2024

```
[11]: import pandas as pd
      import numpy as np
      from scipy.stats import norm, truncnorm, uniform
      # Load data from URL
      url = "https://raw.githubusercontent.com/yadavhimanshu059/CGS698C/main/notes/
      ⇔Data/word-recognition-times.csv"
      dat = pd.read_csv(url)
      dat = dat.iloc[:, 1:] # Remove the first column
      # Convert 'type' to numeric indicator
      dat['type_indicator'] = np.where(dat['type'] == 'word', 0, 1)
      # Extract data
      y = dat['RT'].values
      type_indicator = dat['type_indicator'].values
      # Define log likelihood function
      def log_likelihood(alpha, beta, sigma, RT, type_indicator):
          mu = alpha + beta * type_indicator
          return np.sum(norm.logpdf(RT, loc=mu, scale=sigma))
      # Prior distributions
      def prior_alpha(alpha):
          return norm.logpdf(alpha, loc=400, scale=50)
      def prior_beta(beta):
          return np.log(truncnorm.pdf(beta, a=0, b=np.inf, loc=0, scale=50))
      # MCMC settings
      nsamp = 10000 # Number of samples
      alpha_chain = np.zeros(nsamp)
      beta_chain = np.zeros(nsamp)
      # Initial values
      alpha_chain[0] = np.random.normal(400, 50)
```

```
beta_chain[0] = truncnorm.rvs((0 - beta_chain[0]) / 0.1, (np.inf -
 ⇔beta_chain[0]) / 0.1, loc=beta_chain[0], scale=0.1)
step alpha = 0.1
step_beta = 0.1
i = 0
while i < nsamp - 1:
    # Propose new values
    proposal_alpha = np.random.normal(alpha_chain[i], step_alpha)
    proposal_beta = truncnorm.rvs((0 - beta_chain[i]) / step_beta, (np.inf -u
 deta_chain[i]) / step_beta, loc=beta_chain[i], scale=step_beta)
    # Calculate log posterior
    post_new = (log_likelihood(proposal_alpha, proposal_beta, 30, y,__
  →type_indicator) +
                prior_alpha(proposal_alpha) +
                prior_beta(proposal_beta))
    post_prev = (log_likelihood(alpha_chain[i], beta_chain[i], 30, y,__
  →type_indicator) +
                 prior alpha(alpha chain[i]) +
                 prior_beta(beta_chain[i]))
    # Hastings ratio
    Hastings_ratio = np.exp(post_new - post_prev)
    p_str = min(Hastings_ratio, 1) # probability of acceptance
    if p_str > uniform.rvs():
        alpha_chain[i + 1] = proposal_alpha
        beta_chain[i + 1] = proposal_beta
        i += 1
# Calculate the 95% credible intervals
alpha credible interval = np.percentile(alpha chain, [2.5, 97.5])
beta_credible_interval = np.percentile(beta_chain, [2.5, 97.5])
# Print results
print("95% credible interval for alpha:", alpha_credible_interval)
print("95% credible interval for beta:", beta_credible_interval)
95% credible interval for alpha: [367.53337729 420.71011892]
95% credible interval for beta: [ 8.65546426 53.39317177]
```

[]: