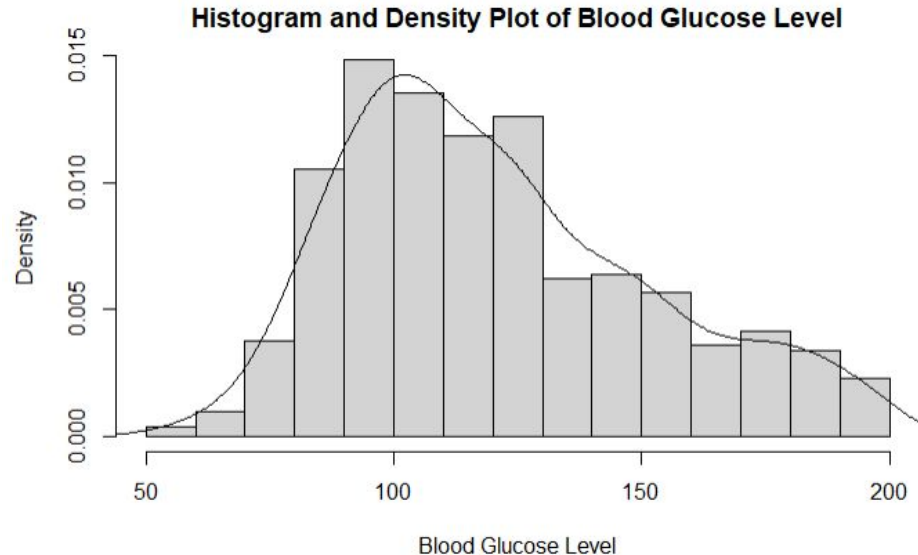
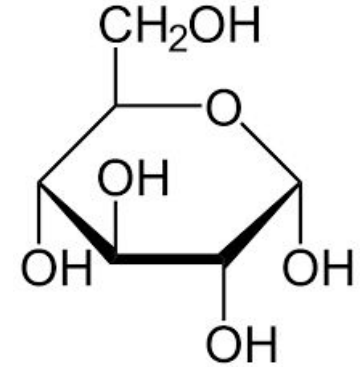


Bayesian Modeling of Blood Glucose

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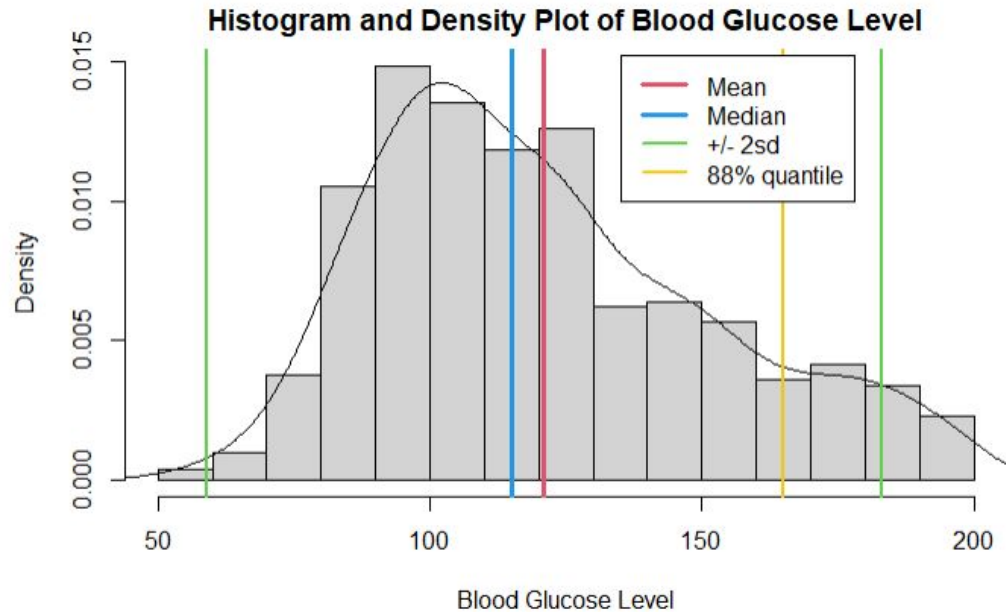
Introduction

- A population of 532 women living near Phoenix, Arizona were tested for diabetes.
- Many factors were recorded, including blood glucose (sugar) level.
- 200 mg/dL or higher after two hours (after eating) suggests diabetes [MC].
- 126 mg/dL or higher on two separate tests (fasted) is diagnosed as diabetes [MC].
- Average blood glucose is 99 mg/dL or lower (fasted) [CDC].



Exploratory Data Analysis

- Data mean \approx 121 mg/dL.
- Data Median = 115 mg/dL.
- Data sd \approx 31 mg/dL.
- 200 of the 523 (38%) observations have glucose of 126 mg/dL or higher.
- About 12% of adult women in the U.S. have diabetes [CDC].
- Using a cutoff of 165 mg/dL, we get 63 of the 532 women (about 12%).



The Model: Sampling Distribution

We choose to use a mixture model for the data. For each of the $n = 532$ study participants, we assign a group membership variable X_i such that

$$X_i = \begin{cases} 1 & \text{with probability } \pi \\ 2 & \text{with probability } 1 - \pi \end{cases}$$

Then the observed data Y_i is given the following density:

$$p(y_i|x_i) = \begin{cases} \text{dnorm}(y_i; \theta_1, \sigma_1^2) & x_i = 1 \\ \text{dnorm}(y_i; \theta_2, \sigma_2^2) & x_i = 2 \end{cases}$$

Note that the X_i are independent and the Y_i are independent given the X_i .

The Model: Prior Distribution

We use the following prior distribution for the model:

$$p(\pi, \theta_1, \theta_2, \sigma_1^2, \sigma_2^2) = p(\pi)p(\theta_1)p(\theta_2)p(\sigma_1^2)p(\sigma_2^2)$$

$$\pi \sim \text{beta}(\alpha, \text{beta})$$

$$\theta_j \sim \text{normal}(\mu_0, \tau_0^2) \text{ for both } j = 1, 2$$

$$\sigma_j^2 \sim \text{inverse - gamma}(\nu_0/2, \sigma_0^2\nu_0/2) \text{ for both } j = 1, 2$$

Full Conditional Distribution

The full conditional distribution can be written as:

$$p(X_i = x_i | \pi, \theta_1, \theta_2, \sigma_1^2, \sigma_2^2, \mathbf{Y}, \mathbf{X}_{-i}) \propto \text{dbinom}(x_i; n, 1, p_2 / (p_1 + p_2)) + 1$$

$$p(\pi | \theta_1, \theta_2, \sigma_1^2, \sigma_2^2, \mathbf{Y}, \mathbf{X}) \propto \text{dbeta}(\pi; \alpha + n_1, \beta + n_2)$$

$$p(\theta_1 | \pi, \theta_2, \sigma_1^2, \sigma_2^2, \mathbf{Y}, \mathbf{X}) \propto \text{dnorm}(\theta_1; \mu_{n,1}, \sigma_{n,1}^2)$$

$$p(\sigma_1^2 | \pi, \theta_1, \theta_2, \sigma_2^2, \mathbf{Y}, \mathbf{X}) \propto \text{dinverse} - \text{gamma}(\sigma_1^2; \nu_{n,1}/2, \tau_{n,1}^2 \nu_{n,1}/2)$$

$$p(\sigma_2^2 | \pi, \theta_1, \theta_2, \sigma_1^2, \mathbf{Y}, \mathbf{X}) \propto \text{dinverse} - \text{gamma}(\sigma_2^2; \nu_{n,2}/2, \tau_{n,2}^2 \nu_{n,2}/2)$$

Where

$$p_1 = \text{dnorm}(y_i; \theta_1, \sigma_1^2)$$

$$p_2 = \text{dnorm}(y_i; \theta_2, \sigma_2^2)$$

$$\mu_{n,1} = \frac{\frac{1}{\tau_0} \mu_0 + \frac{n_1}{\sigma_1^2} \bar{y}_1}{\frac{1}{\tau_0} + \frac{n_1}{\sigma_1^2}}$$

$$\sigma_{n,1}^2 = \frac{1}{\frac{1}{\tau_0} + \frac{n_1}{\sigma_1^2}}$$

$$\mu_{n,2} = \frac{\frac{1}{\tau_0} \mu_0 + \frac{n_2}{\sigma_2^2} \bar{y}_2}{\frac{1}{\tau_0} + \frac{n_2}{\sigma_2^2}}$$

$$\sigma_{n,2}^2 = \frac{1}{\frac{1}{\tau_0} + \frac{n_2}{\sigma_2^2}}$$

$$\nu_{n,1} = \nu_0 + n_1$$

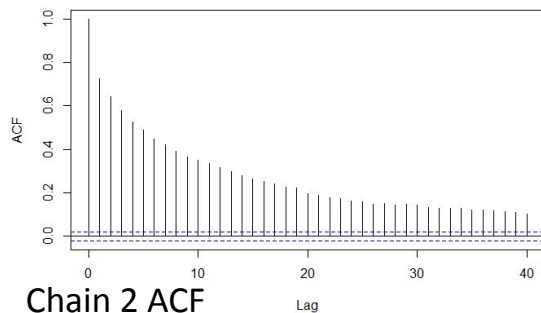
$$\tau_{n,1}^2 = \frac{\nu_0 \sigma_0^2 + \sum_{i, x_i=1} (y_i - \theta)^2}{\nu_{n,1}}$$

$$\nu_{n,2} = \nu_0 + n_2$$

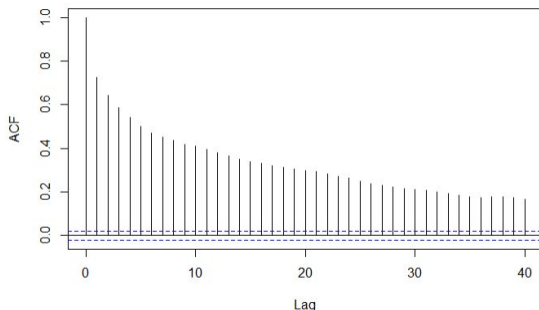
$$\tau_{n,2}^2 = \frac{\nu_0 \sigma_0^2 + \sum_{i, x_i=2} (y_i - \theta)^2}{\nu_{n,2}}$$

Model Diagnostics

Chain 1 ACF



Chain 2 ACF

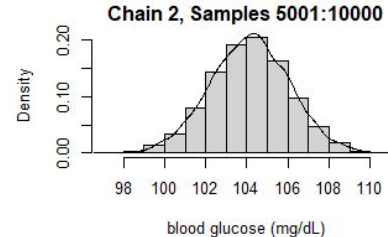
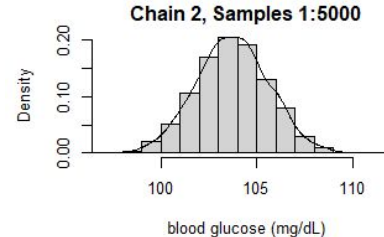
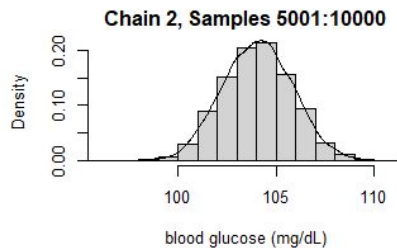
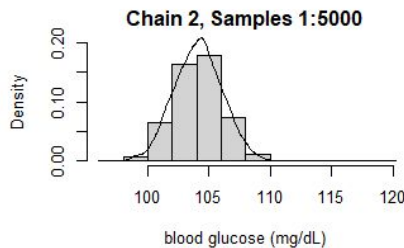
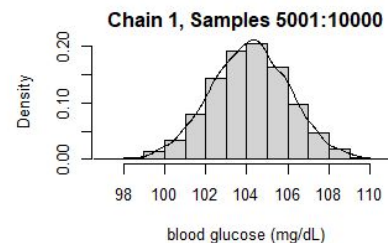
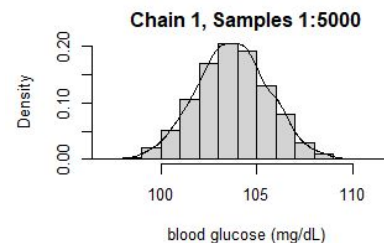
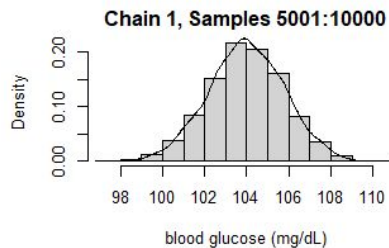
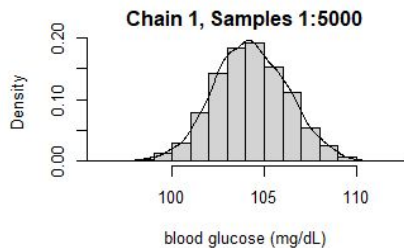


Gibbs Sampler Effective Sizes

parameter	chain1 effective size	chain 2 effective size
theta1	464	225
theta2	227	425
pi	233	216

Model Diagnostics - JAGS Comparison

Gibbs Sampler low blood glucose group mean sample splits JAGS low blood glucose group mean sample splits



Posterior Statistics

Posterior Statistics for Chain 1

variable <chr>	mean <dbl>	2.5% quantile <dbl>	97.5% quantile <dbl>
group 1 blood glucose (mg/dL)	104.13	100.36	107.94
group 2 blood glucose (mg/dL)	149.25	137.93	160.93
proportion group 1	0.62	0.48	0.75
min of groups	104.13	100.36	107.94

Posterior Statistics for Chain 2

variable <chr>	mean <dbl>	2.5% quantile <dbl>	97.5% quantile <dbl>
group 1 blood glucose (mg/dL)	148.80	137.47	160.67
group 2 blood glucose (mg/dL)	104.10	100.46	107.85
proportion group 1	0.38	0.25	0.53
min of groups	104.10	100.46	107.85

Conclusion

- The posterior means for group 1 and 2 blood glucose indicate the model is able to separate groups into diabetics and nondiabetics (“low glucose” and “high glucose”).
- The average blood glucose for a non-diabetic is very close to the posterior mean for the “low glucose” group.
- The ADA recommends diabetics to keep their blood sugar below 130 when fasted, or below 180 two hours after eating. 150 is a happy medium.
- The posterior percent of participants in the “high glucose” group (40%) is much higher than percent of women in the U.S. with diabetes (12%).

References

[CDC] CDC Diabetes tests:

<https://www.cdc.gov/diabetes/basics/getting-tested.html#:~:text=A%20fasting%20blood%20sugar%20level,higher%20indicates%20you%20have%20diabetes.>

[MC] Mayo Clinic Diabetes:

<https://www.mayoclinic.org/diseases-conditions/diabetes/diagnosis-treatment/drc-20371451>

[ADA] Americans with Diabetes Associates (Diabetes.org): <https://diabetes.org/diabetes/a1c/diagnosis>