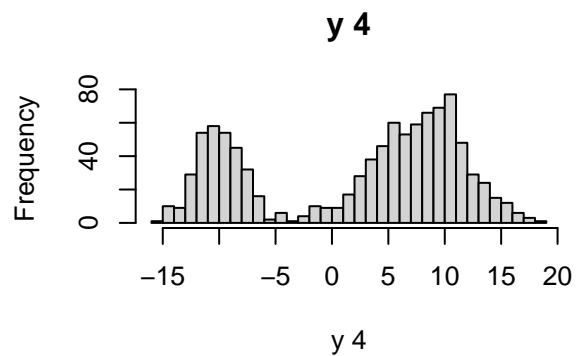
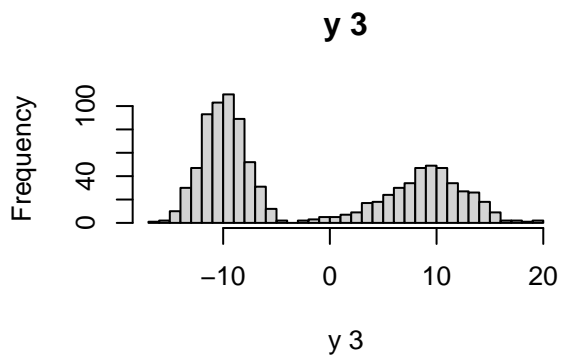
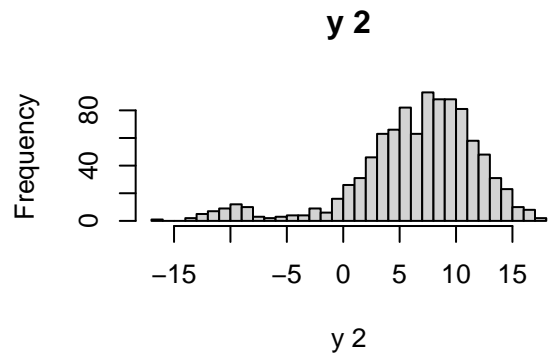
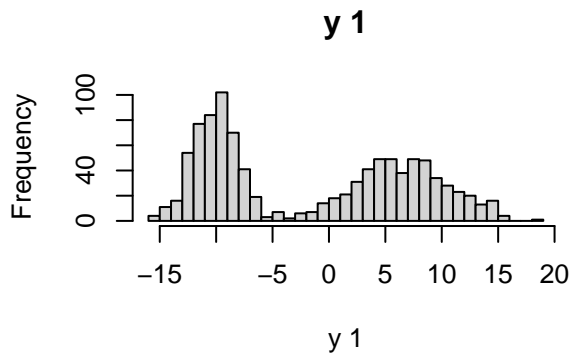


Extra Credit

Yuki Joyama (yj2803)

```
# import data
data <- read.table("hw4_data_prob3.txt", header = TRUE)

# check histograms
par(mfrow = c(2, 2))
for (i in 1:4) {
  hist(data[[i]], main = paste("y", i), xlab = paste("y", i), breaks = 30)
}
```



```
par(mfrow = c(1, 1))

# data cleaning for stan
stan_data <- list(
```

```

N = nrow(data),
D = ncol(data),
K = 3,
y = as.matrix(data)
)

# compile the stan model
stan_model <- stan_model("mixture_model.stan")

## Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
## using C compiler: 'Apple clang version 15.0.0 (clang-1500.1.0.2.5)'
## using SDK: 'MacOSX14.2.sdk'
## clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/StanHeaders/include"
## In file included from <built-in>:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/StanHeaders/include:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/RcppEigen/include:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/Matrix.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/MatrixBase.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/MatrixBase.h:1:
## #include <cmath>
## ~~~~~
## 1 error generated.
## make: *** [foo.o] Error 1

# fit the model
fit <- sampling(
  stan_model,
  data = stan_data,
  iter = 1000,
  chains = 4,
  control = list(adapt_delta = 0.99, max_treedepth = 20)
)

##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 0.001745 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 17.45 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration: 1 / 1000 [ 0%] (Warmup)
## Chain 1: Iteration: 100 / 1000 [ 10%] (Warmup)
## Chain 1: Iteration: 200 / 1000 [ 20%] (Warmup)
## Chain 1: Iteration: 300 / 1000 [ 30%] (Warmup)
## Chain 1: Iteration: 400 / 1000 [ 40%] (Warmup)

```

```

## Chain 1: Iteration: 500 / 1000 [ 50%] (Warmup)
## Chain 1: Iteration: 501 / 1000 [ 50%] (Sampling)
## Chain 1: Iteration: 600 / 1000 [ 60%] (Sampling)
## Chain 1: Iteration: 700 / 1000 [ 70%] (Sampling)
## Chain 1: Iteration: 800 / 1000 [ 80%] (Sampling)
## Chain 1: Iteration: 900 / 1000 [ 90%] (Sampling)
## Chain 1: Iteration: 1000 / 1000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 145.602 seconds (Warm-up)
## Chain 1: 248.547 seconds (Sampling)
## Chain 1: 394.149 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 0.001135 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 11.35 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration: 1 / 1000 [ 0%] (Warmup)
## Chain 2: Iteration: 100 / 1000 [ 10%] (Warmup)
## Chain 2: Iteration: 200 / 1000 [ 20%] (Warmup)
## Chain 2: Iteration: 300 / 1000 [ 30%] (Warmup)
## Chain 2: Iteration: 400 / 1000 [ 40%] (Warmup)
## Chain 2: Iteration: 500 / 1000 [ 50%] (Warmup)
## Chain 2: Iteration: 501 / 1000 [ 50%] (Sampling)
## Chain 2: Iteration: 600 / 1000 [ 60%] (Sampling)
## Chain 2: Iteration: 700 / 1000 [ 70%] (Sampling)
## Chain 2: Iteration: 800 / 1000 [ 80%] (Sampling)
## Chain 2: Iteration: 900 / 1000 [ 90%] (Sampling)
## Chain 2: Iteration: 1000 / 1000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 71.991 seconds (Warm-up)
## Chain 2: 43.594 seconds (Sampling)
## Chain 2: 115.585 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 0.001161 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 11.61 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:

```

```

## Chain 3: Iteration: 1 / 1000 [ 0%] (Warmup)
## Chain 3: Iteration: 100 / 1000 [ 10%] (Warmup)
## Chain 3: Iteration: 200 / 1000 [ 20%] (Warmup)
## Chain 3: Iteration: 300 / 1000 [ 30%] (Warmup)
## Chain 3: Iteration: 400 / 1000 [ 40%] (Warmup)
## Chain 3: Iteration: 500 / 1000 [ 50%] (Warmup)
## Chain 3: Iteration: 501 / 1000 [ 50%] (Sampling)
## Chain 3: Iteration: 600 / 1000 [ 60%] (Sampling)
## Chain 3: Iteration: 700 / 1000 [ 70%] (Sampling)
## Chain 3: Iteration: 800 / 1000 [ 80%] (Sampling)
## Chain 3: Iteration: 900 / 1000 [ 90%] (Sampling)
## Chain 3: Iteration: 1000 / 1000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 134.107 seconds (Warm-up)
## Chain 3: 125.603 seconds (Sampling)
## Chain 3: 259.71 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 0.001161 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 11.61 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 1000 [ 0%] (Warmup)
## Chain 4: Iteration: 100 / 1000 [ 10%] (Warmup)
## Chain 4: Iteration: 200 / 1000 [ 20%] (Warmup)
## Chain 4: Iteration: 300 / 1000 [ 30%] (Warmup)
## Chain 4: Iteration: 400 / 1000 [ 40%] (Warmup)
## Chain 4: Iteration: 500 / 1000 [ 50%] (Warmup)
## Chain 4: Iteration: 501 / 1000 [ 50%] (Sampling)
## Chain 4: Iteration: 600 / 1000 [ 60%] (Sampling)
## Chain 4: Iteration: 700 / 1000 [ 70%] (Sampling)
## Chain 4: Iteration: 800 / 1000 [ 80%] (Sampling)
## Chain 4: Iteration: 900 / 1000 [ 90%] (Sampling)
## Chain 4: Iteration: 1000 / 1000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 50.035 seconds (Warm-up)
## Chain 4: 34.737 seconds (Sampling)
## Chain 4: 84.772 seconds (Total)
## Chain 4:

```

```

# summary of posterior distribution of parameters:
print(fit)

```

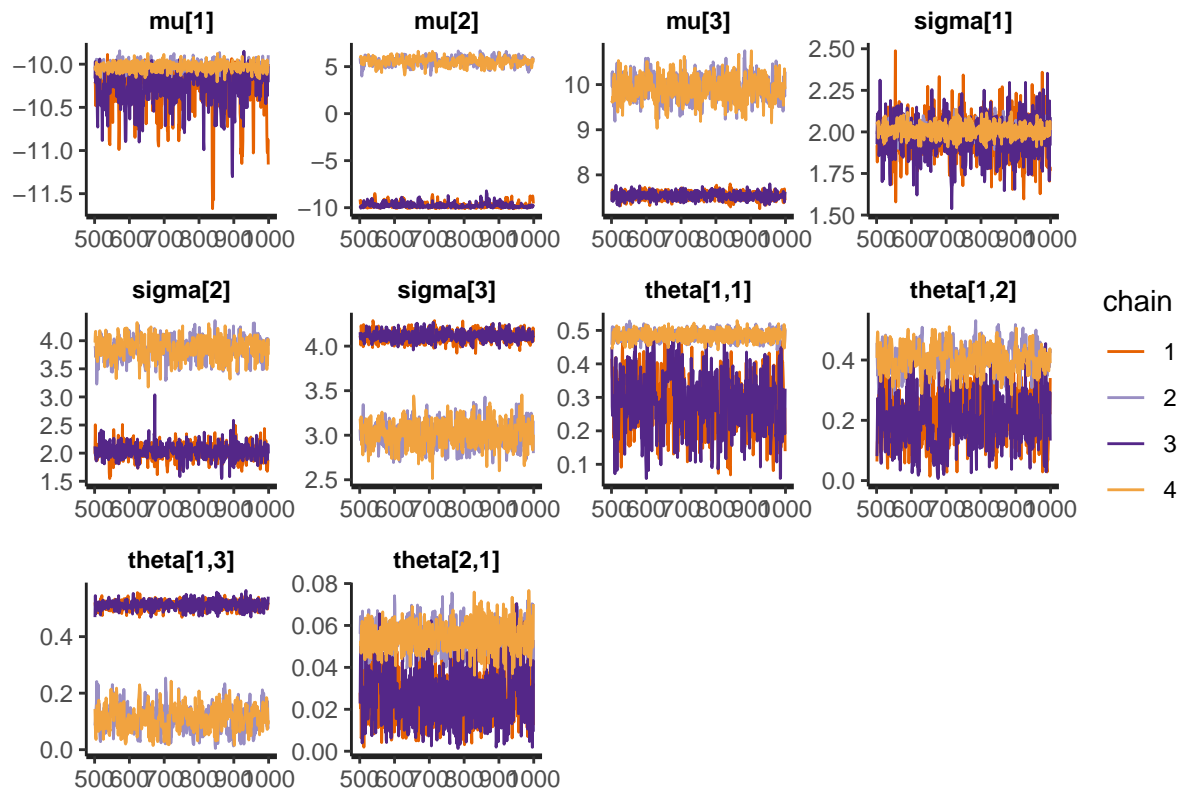
```

## Inference for Stan model: anon_model.
## 4 chains, each with iter=1000; warmup=500; thin=1;
## post-warmup draws per chain=500, total post-warmup draws=2000.
##
##              mean se_mean    sd      2.5%      25%      50%      75%
## mu[1]        -10.15    0.08  0.21    -10.73    -10.20    -10.08    -10.02
## mu[2]         -2.08    5.41  7.66    -10.03     -9.81     -2.11      5.59
## mu[3]          8.73    0.85  1.21      7.40      7.54      8.42      9.92
## sigma[1]       1.99    0.01  0.09      1.76      1.95      2.00      2.04
## sigma[2]       2.94    0.65  0.93      1.79      2.03      3.11      3.88
## sigma[3]       3.56    0.39  0.56      2.78      3.02      3.69      4.11
## theta[1,1]     0.39    0.07  0.12      0.13      0.29      0.45      0.49
## theta[1,2]     0.30    0.07  0.12      0.06      0.20      0.34      0.41
## theta[1,3]     0.31    0.14  0.20      0.04      0.11      0.36      0.51
## theta[2,1]     0.04    0.01  0.02      0.01      0.03      0.05      0.06
## theta[2,2]     0.30    0.19  0.27      0.01      0.03      0.19      0.57
## theta[2,3]     0.66    0.20  0.29      0.26      0.38      0.78      0.94
## theta[3,1]     0.43    0.11  0.17      0.09      0.29      0.53      0.58
## theta[3,2]     0.21    0.07  0.12      0.07      0.11      0.15      0.29
## theta[3,3]     0.36    0.04  0.06      0.26      0.31      0.38      0.42
## theta[4,1]     0.24    0.05  0.09      0.06      0.16      0.28      0.31
## theta[4,2]     0.25    0.07  0.11      0.05      0.15      0.26      0.34
## theta[4,3]     0.52    0.12  0.17      0.25      0.35      0.57      0.68
## lp__          -12505.45  23.52 33.38 -12543.90 -12538.52 -12509.65 -12471.88
##              97.5% n_eff  Rhat
## mu[1]         -9.94     7  1.22
## mu[2]          6.27     2 21.33
## mu[3]         10.37     2  6.40
## sigma[1]       2.16    108  1.04
## sigma[2]       4.15     2  5.86
## sigma[3]       4.22     2  5.53
## theta[1,1]     0.51     3  1.96
## theta[1,2]     0.48     3  1.84
## theta[1,3]     0.54     2  6.42
## theta[2,1]     0.07     3  1.80
## theta[2,2]     0.68     2  5.56
## theta[2,3]     0.95     2  5.90
## theta[3,1]     0.61     2  2.25
## theta[3,2]     0.49     3  1.59
## theta[3,3]     0.44     2  2.74
## theta[4,1]     0.34     3  2.02
## theta[4,2]     0.43     3  1.99
## theta[4,3]     0.71     2  4.68
## lp__          -12468.40     2 12.51
##

```

```
## Samples were drawn using NUTS(diag_e) at Wed Dec 11 19:49:36 2024.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).
```

```
# traceplots
traceplot(fit)
```



Variational Bayes

```
# VB works
vb_fit <-
  vb(
    stan_model,
    data = stan_data,
    iter = 1000,
    elbo_samples = 500,
    algorithm = c("fullrank"),
    output_samples = 1000,
```

```
tol_rel_obj = 0.00001
)
```

```
## Chain 1: -----
## Chain 1: EXPERIMENTAL ALGORITHM:
## Chain 1:   This procedure has not been thoroughly tested and may be unstable
## Chain 1:   or buggy. The interface is subject to change.
## Chain 1: -----
## Chain 1:
## Chain 1:
## Chain 1:
## Chain 1: Gradient evaluation took 0.001454 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 14.54 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Begin eta adaptation.
## Chain 1: Iteration:   1 / 250 [  0%] (Adaptation)
## Chain 1: Iteration:  50 / 250 [ 20%] (Adaptation)
## Chain 1: Iteration: 100 / 250 [ 40%] (Adaptation)
## Chain 1: Iteration: 150 / 250 [ 60%] (Adaptation)
## Chain 1: Iteration: 200 / 250 [ 80%] (Adaptation)
## Chain 1: Iteration: 250 / 250 [100%] (Adaptation)
## Chain 1: Success! Found best value [eta = 0.1].
## Chain 1:
## Chain 1: Begin stochastic gradient ascent.
## Chain 1:   iter          ELBO    delta_ELBO_mean    delta_ELBO_med    notes
## Chain 1:   100        -14857.805          1.000          1.000
## Chain 1:   200        -14508.785          0.512          1.000
## Chain 1:   300        -14409.065          0.015          0.024
## Chain 1:   400        -14346.607          0.006          0.007
## Chain 1:   500        -14282.541          0.004          0.004
## Chain 1:   600        -14244.869          0.004          0.004
## Chain 1:   700        -14186.891          0.003          0.004
## Chain 1:   800        -14146.482          0.003          0.004
## Chain 1:   900        -14141.265          0.002          0.003
## Chain 1:  1000        -14068.186          0.003          0.005
## Chain 1: Informational Message: The maximum number of iterations is reached! The algorithm may not ha
## Chain 1: This variational approximation is not guaranteed to be meaningful.
## Chain 1:
## Chain 1: Drawing a sample of size 1000 from the approximate posterior...
## Chain 1: COMPLETED.
```

```
# vb estimates
```

```
print(vb_fit)
```

```
## Inference for Stan model: anon_model.
```

```
## 1 chains, each with iter=1000; warmup=0; thin=1;
```

```
## post-warmup draws per chain=1000, total post-warmup draws=1000.
```

```
##
```

	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	khat
## mu[1]	0.61	NaN	0.21	0.21	0.46	0.61	0.75	1.02	NaN	9.53
## mu[2]	5.71	NaN	2.37	2.64	4.07	5.17	6.75	11.79	NaN	9.48
## mu[3]	7.66	NaN	1.97	5.26	6.40	7.17	8.37	12.70	NaN	9.54
## sigma[1]	9.47	NaN	0.19	9.10	9.35	9.47	9.60	9.84	NaN	9.58
## sigma[2]	2.14	NaN	1.49	0.47	1.15	1.76	2.63	5.85	NaN	9.48
## sigma[3]	3.81	NaN	0.29	3.29	3.62	3.80	4.00	4.40	NaN	9.58
## theta[1,1]	0.93	NaN	0.03	0.87	0.91	0.93	0.95	0.97	NaN	9.57
## theta[1,2]	0.02	NaN	0.02	0.00	0.01	0.02	0.03	0.07	NaN	9.56
## theta[1,3]	0.05	NaN	0.02	0.01	0.03	0.04	0.06	0.10	NaN	9.56
## theta[2,1]	0.18	NaN	0.04	0.11	0.15	0.17	0.20	0.26	NaN	9.57
## theta[2,2]	0.06	NaN	0.04	0.01	0.03	0.05	0.08	0.16	NaN	9.56
## theta[2,3]	0.77	NaN	0.06	0.62	0.74	0.78	0.81	0.86	NaN	9.54
## theta[3,1]	0.96	NaN	0.02	0.90	0.95	0.96	0.97	0.99	NaN	9.58
## theta[3,2]	0.01	NaN	0.01	0.00	0.00	0.00	0.01	0.03	NaN	9.56
## theta[3,3]	0.03	NaN	0.02	0.01	0.02	0.03	0.05	0.09	NaN	9.56
## theta[4,1]	0.66	NaN	0.05	0.55	0.63	0.66	0.70	0.76	NaN	9.57
## theta[4,2]	0.03	NaN	0.03	0.00	0.01	0.02	0.04	0.13	NaN	9.56
## theta[4,3]	0.31	NaN	0.06	0.20	0.27	0.31	0.34	0.42	NaN	9.57
## lp__	0.00	NaN	0.00	0.00	0.00	0.00	0.00	0.00	NaN	9.56

```
##
```

```
## Approximate samples were drawn using VB(fullrank) at Wed Dec 11 19:49:43 2024.
```

```
# get estimated and generating values for wanted parameters
```

```
pars <- vb_fit %>% names %>% `[1:10] %>% sort()
```

```
sim_summary <- as.data.frame(summary(vb_fit)[[1]])
```

```
estimated_values <- sim_summary[pars, c("mean", "2.5%", "97.5%")]
```

```
rstan::traceplot(vb_fit)
```