# stan examples

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## 12/8/2021

This file contains examples of 'Stan'.

The first example is for Gaussian model.

The second example is for Beta-binomial model.

#### Set up

#### Ex 1 Gaussian

The mock data is 1,000 i.i.d.  $X \sim \text{Normal}(\mu=5,\sigma=1)$ .

#### Template Result:

```
## Inference for Stan model: my_model_norm.
## 4 chains, each with iter=2000; warmup=1000; thin=1;
## post-warmup draws per chain=1000, total post-warmup draws=4000.
##
##
                                                   50%
                                                                 97.5% n_eff Rhat
            mean se_mean
                           sd
                                  2.5%
                                           25%
                                                           75%
                                                                  5.04 3319
## mu
            4.98
                    0.00 0.03
                                  4.92
                                          4.96
                                                  4.98
                                                          5.00
                                          1.02
## sigma
            1.04
                    0.00 0.02
                                 0.99
                                                  1.04
                                                          1.05
                                                                   1.08
                                                                         3017
                                                                                 1
         -535.51
                    0.03 1.03 -538.40 -535.88 -535.20 -534.78 -534.52
## lp__
##
## Samples were drawn using NUTS(diag_e) at Wed Jan 5 00:19:58 2022.
## 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).
```

#### Pair plots:

#### Try to extract sample results

```
## [1] 4.98610 4.97626 5.00561 4.91539 4.87870 4.90267 5.01703 4.94508 4.97180 ## [10] 4.98501
```

#### Ex\_2 Beta-binomial

#### Parameters and set up:

The number of sampling times is 1000; Prior distribution  $\theta \sim Beta(1,1)$ ,  $Y \sim Bin(100,\theta)$ ;

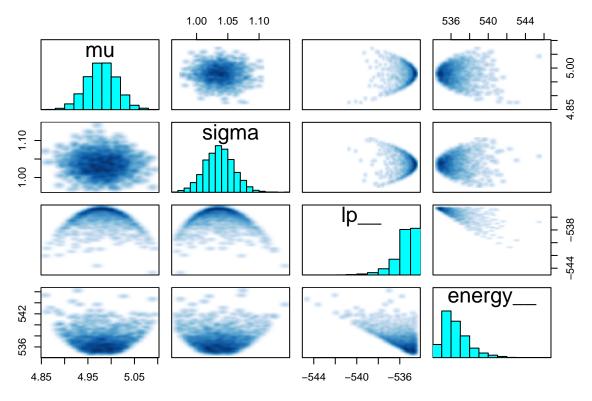


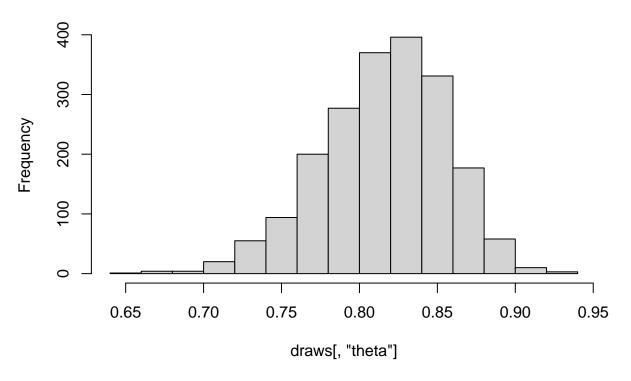
Figure 1: Pair plots for Gaussian model

Observed data is 82 responses in the first cohort of 100 patients;

#### Posterior distribution for $\theta$ :

```
## Inference for the input samples (4 chains: each with iter = 1000; warmup = 0):
##
##
            Q5
                 Q50
                       Q95
                           Mean SD Rhat Bulk_ESS Tail_ESS
           0.7
                 0.8
                       0.9
                             0.8 0.0
                                      1.01
                                                581
                                                         705
## lp__ -51.1 -49.3 -49.0 -49.6 0.8 1.01
                                                805
                                                         725
##
## For each parameter, Bulk_ESS and Tail_ESS are crude measures of
## effective sample size for bulk and tail quantities respectively (an ESS > 100
## per chain is considered good), and Rhat is the potential scale reduction
## factor on rank normalized split chains (at convergence, Rhat <= 1.05).
```

# Histogram of draws[, "theta"]



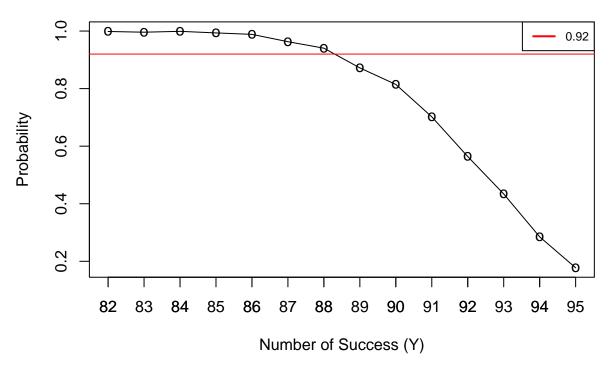
For  $P(\theta \le 0.92 | Y = 82)$ 

## [1] 0.9985

As  $P(\theta \le 0.92|Y=82)=0.9985>0.9$ , we stop the trial at the first interim analysis for observing 82 responses in the first 100 patients.

## [1] 0.1775

## Posterior distribution of theta



## [1] "P|Y=88: 0.9405"

## [1] "P|Y=89: 0.872"