

STA2201_Lab7

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15/03/2023

Lip cancer

Here is the lip cancer data given to you in terribly unreproducible and error-prone format.

- `aff.i` is proportion of male population working outside in each region
- `observe.i` is observed deaths in each region
- `expect.i` is expected deaths, based on region-specific age distribution and national-level age-specific mortality rates.

```
observe.i <- c(
  5,13,18,5,10,18,29,10,15,22,4,11,10,22,13,14,17,21,25,6,11,21,13,5,19,18,14,17,3,10,
  7,3,12,11,6,16,13,6,9,10,4,9,11,12,23,18,12,7,13,12,12,13,6,14,7,18,13,9,6,8,7,6,16,4,6,12,5,5,
  17,5,7,2,9,7,6,12,13,17,5,5,6,12,10,16,10,16,15,18,6,12,6,8,33,15,14,18,25,14,2,73,13,14,6,20,8,
  12,10,3,11,3,11,13,11,13,10,5,18,10,23,5,9,2,11,9,11,6,11,5,19,15,4,8,9,6,4,4,2,12,12,11,9,7,7,
  8,12,11,23,7,16,46,9,18,12,13,14,14,3,9,15,6,13,13,12,8,11,5,9,8,22,9,2,10,6,10,12,9,11,32,5,11,
  9,11,11,0,9,3,11,11,11,5,4,8,9,30,110)
expect.i <- c(
  6.17,8.44,7.23,5.62,4.18,29.35,11.79,12.35,7.28,9.40,3.77,3.41,8.70,9.57,8.18,4.35,
  4.91,10.66,16.99,2.94,3.07,5.50,6.47,4.85,9.85,6.95,5.74,5.70,2.22,3.46,4.40,4.05,5.74,6.36,5.13,
  16.99,6.19,5.56,11.69,4.69,6.25,10.84,8.40,13.19,9.25,16.98,8.39,2.86,9.70,12.12,12.94,9.77,
  10.34,5.09,3.29,17.19,5.42,11.39,8.33,4.97,7.14,6.74,17.01,5.80,4.84,12.00,4.50,4.39,16.35,6.02,
  6.42,5.26,4.59,11.86,4.05,5.48,13.13,8.72,2.87,2.13,4.48,5.85,6.67,6.11,5.78,12.31,10.56,10.23,
  2.52,6.22,14.29,5.71,37.93,7.81,9.86,11.61,18.52,12.28,5.41,61.96,8.55,12.07,4.29,19.42,8.25,
  12.90,4.76,5.56,11.11,4.76,10.48,13.13,12.94,14.61,9.26,6.94,16.82,33.49,20.91,5.32,6.77,8.70,
  12.94,16.07,8.87,7.79,14.60,5.10,24.42,17.78,4.04,7.84,9.89,8.45,5.06,4.49,6.25,9.16,12.37,8.40,
  9.57,5.83,9.21,9.64,9.09,12.94,17.42,10.29,7.14,92.50,14.29,15.61,6.00,8.55,15.22,18.42,5.77,
  18.37,13.16,7.69,14.61,15.85,12.77,7.41,14.86,6.94,5.66,9.88,102.16,7.63,5.13,7.58,8.00,12.82,
  18.75,12.33,5.88,64.64,8.62,12.09,11.11,14.10,10.48,7.00,10.23,6.82,15.71,9.65,8.59,8.33,6.06,
  12.31,8.91,50.10,288.00)
aff.i <- c(0.2415,0.2309,0.3999,0.2977,0.3264,0.3346,0.4150,0.4202,0.1023,0.1752,
  0.2548,0.3248,0.2287,0.2520,0.2058,0.2785,0.2528,0.1847,0.3736,0.2411,
  0.3700,0.2997,0.2883,0.2427,0.3782,0.1865,0.2633,0.2978,0.3541,0.4176,
  0.2910,0.3431,0.1168,0.2195,0.2911,0.4297,0.2119,0.2698,0.0874,0.3204,
  0.1839,0.1796,0.2471,0.2016,0.1560,0.3162,0.0732,0.1490,0.2283,0.1187,
  0.3500,0.2915,0.1339,0.0995,0.2355,0.2392,0.0877,0.3571,0.1014,0.0363,
  0.1665,0.1226,0.2186,0.1279,0.0842,0.0733,0.0377,0.2216,0.3062,0.0310,
  0.0755,0.0583,0.2546,0.2933,0.1682,0.2518,0.1971,0.1473,0.2311,0.2471,
  0.3063,0.1526,0.1487,0.3537,0.2753,0.0849,0.1013,0.1622,0.1267,0.2376,
  0.0737,0.2755,0.0152,0.1415,0.1344,0.1058,0.0545,0.1047,0.1335,0.3134,
  0.1326,0.1222,0.1992,0.0620,0.1313,0.0848,0.2687,0.1396,0.1234,0.0997,
```

```
0.0694,0.1022,0.0779,0.0253,0.1012,0.0999,0.0828,0.2950,0.0778,0.1388,
0.2449,0.0978,0.1144,0.1038,0.1613,0.1921,0.2714,0.1467,0.1783,0.1790,
0.1482,0.1383,0.0805,0.0619,0.1934,0.1315,0.1050,0.0702,0.1002,0.1445,
0.0353,0.0400,0.1385,0.0491,0.0520,0.0640,0.1017,0.0837,0.1462,0.0958,
0.0745,0.2942,0.2278,0.1347,0.0907,0.1238,0.1773,0.0623,0.0742,0.1003,
0.0590,0.0719,0.0652,0.1687,0.1199,0.1768,0.1638,0.1360,0.0832,0.2174,
0.1662,0.2023,0.1319,0.0526,0.0287,0.0405,0.1616,0.0730,0.1005,0.0743,
0.0577,0.0481,0.1002,0.0433,0.0838,0.1124,0.2265,0.0436,0.1402,0.0313,
0.0359,0.0696,0.0618,0.0932,0.0097)
```

Question 1

Explain a bit more what the `expect.i` variable is. For example, if a particular area has an expected deaths of 6, what does this mean?

Solution

Expected deaths is the implied number of lip cancer deaths for a particular region given that region's age structure and the national level age-specific mortality rates for lip cancer. For example, an expected number of deaths being 6 would mean that for that particular region, we would expect 6 lip cancer deaths if this region were to experience the same age specific mortality rates as at the national level.

Question 2

Run three different models in Stan with three different set-up's for estimating θ_i , that is the relative risk of lip cancer in each region:

1. Intercept α_i is same in each region $= \alpha$
2. α_i is different in each region and modeled separately (with covariate)
3. α_i is different in each region and the intercept is modeled hierarchically (with covariate)

Solution

$$y_i | \theta_i \sim \text{Poisson}(\theta_i \cdot e_i)$$

Look at 3 models for $\log(\theta_i)$

$$\log(\theta_i) = \alpha + \beta x_i$$

Model 1

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr   0.3.5
## v tibble  3.1.8      v dplyr  1.1.0
## v tidyr   1.2.1      v stringr 1.4.1
```

```
## v readr 2.1.3 v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
```

```
library(rstan)
```

```
## Loading required package: StanHeaders
##
## rstan version 2.26.13 (Stan version 2.26.1)
##
## For execution on a local, multicore CPU with excess RAM we recommend calling
## options(mc.cores = parallel::detectCores()).
## To avoid recompilation of unchanged Stan programs, we recommend calling
## rstan_options(auto_write = TRUE)
## For within-chain threading using 'reduce_sum()' or 'map_rect()' Stan functions,
## change 'threads_per_chain' option:
## rstan_options(threads_per_chain = 1)
##
## Do not specify '-march=native' in 'LOCAL_CPPFLAGS' or a Makevars file
##
## Attaching package: 'rstan'
##
## The following object is masked from 'package:tidyr':
##
##     extract
```

```
library(tidybayes)
```

```
stan_data <- list(y = observe.i, log_e = log(expect.i), N=length(observe.i), x=aff.i-mean(aff.i))
```

```
mod1 <- stan(data=stan_data, file = "lab9_1.stan")
```

```
## Warning in readLines(file, warn = TRUE): incomplete
## final line found on 'C:\Users\Alice\Desktop\GRAD
## SCHOOL\STA2201\STA2201AppliedStatAliceHuang\lab9_1.stan'

##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 0.000129 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 1.29 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 1: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 1: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 1: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 1: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 1: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 1: Iteration:  1001 / 2000 [ 50%] (Sampling)
```

```

## Chain 1: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 1: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 1: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 1: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 0.148 seconds (Warm-up)
## Chain 1: 0.127 seconds (Sampling)
## Chain 1: 0.275 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 3.1e-05 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.31 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.135 seconds (Warm-up)
## Chain 2: 0.12 seconds (Sampling)
## Chain 2: 0.255 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 2.6e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.26 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)

```

```

## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.133 seconds (Warm-up)
## Chain 3: 0.169 seconds (Sampling)
## Chain 3: 0.302 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 4.2e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.42 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 0.173 seconds (Warm-up)
## Chain 4: 0.126 seconds (Sampling)
## Chain 4: 0.299 seconds (Total)
## Chain 4:

```

```
mod1
```

```

## Inference for Stan model: anon_model.
## 4 chains, each with iter=2000; warmup=1000; thin=1;
## post-warmup draws per chain=1000, total post-warmup draws=4000.
##
##               mean se_mean  sd   2.5%   25%   50%   75%   97.5%
## alpha        -0.01    0.00 0.02  -0.05  -0.02  -0.01   0.00   0.03
## beta          2.42    0.00 0.18   2.06   2.30   2.42   2.54   2.76
## log_theta[1]   0.17    0.00 0.02   0.12   0.15   0.17   0.19   0.22
## log_theta[2]   0.15    0.00 0.02   0.10   0.13   0.15   0.16   0.19
## log_theta[3]   0.55    0.00 0.05   0.46   0.52   0.55   0.59   0.64
## log_theta[4]   0.31    0.00 0.03   0.24   0.29   0.31   0.33   0.37
## log_theta[5]   0.38    0.00 0.04   0.31   0.35   0.38   0.40   0.45
## log_theta[6]   0.40    0.00 0.04   0.32   0.37   0.40   0.42   0.47
## log_theta[7]   0.59    0.00 0.05   0.49   0.56   0.59   0.62   0.69
## log_theta[8]   0.60    0.00 0.05   0.50   0.57   0.60   0.64   0.70
## log_theta[9]  -0.17    0.00 0.02  -0.21  -0.18  -0.17  -0.15  -0.12
## log_theta[10]  0.01    0.00 0.02  -0.03   0.00   0.01   0.02   0.05
## log_theta[11]  0.20    0.00 0.03   0.15   0.19   0.20   0.22   0.25
## log_theta[12]  0.37    0.00 0.04   0.30   0.35   0.37   0.40   0.44
## log_theta[13]  0.14    0.00 0.02   0.09   0.12   0.14   0.16   0.19

```

## log_theta[14]	0.20	0.00	0.03	0.14	0.18	0.20	0.21	0.25
## log_theta[15]	0.08	0.00	0.02	0.04	0.07	0.08	0.10	0.13
## log_theta[16]	0.26	0.00	0.03	0.20	0.24	0.26	0.28	0.32
## log_theta[17]	0.20	0.00	0.03	0.15	0.18	0.20	0.22	0.25
## log_theta[18]	0.03	0.00	0.02	-0.01	0.02	0.03	0.05	0.08
## log_theta[19]	0.49	0.00	0.04	0.41	0.46	0.49	0.52	0.57
## log_theta[20]	0.17	0.00	0.02	0.12	0.15	0.17	0.19	0.22
## log_theta[21]	0.48	0.00	0.04	0.40	0.45	0.48	0.51	0.56
## log_theta[22]	0.31	0.00	0.03	0.25	0.29	0.31	0.33	0.37
## log_theta[23]	0.28	0.00	0.03	0.22	0.26	0.28	0.30	0.34
## log_theta[24]	0.17	0.00	0.03	0.12	0.16	0.17	0.19	0.22
## log_theta[25]	0.50	0.00	0.04	0.42	0.47	0.50	0.53	0.59
## log_theta[26]	0.04	0.00	0.02	0.00	0.02	0.04	0.05	0.08
## log_theta[27]	0.22	0.00	0.03	0.17	0.21	0.22	0.24	0.28
## log_theta[28]	0.31	0.00	0.03	0.24	0.29	0.31	0.33	0.37
## log_theta[29]	0.44	0.00	0.04	0.36	0.42	0.44	0.47	0.52
## log_theta[30]	0.60	0.00	0.05	0.50	0.56	0.60	0.63	0.69
## log_theta[31]	0.29	0.00	0.03	0.23	0.27	0.29	0.31	0.35
## log_theta[32]	0.42	0.00	0.04	0.34	0.39	0.42	0.44	0.49
## log_theta[33]	-0.13	0.00	0.02	-0.18	-0.15	-0.13	-0.12	-0.09
## log_theta[34]	0.12	0.00	0.02	0.07	0.10	0.12	0.13	0.16
## log_theta[35]	0.29	0.00	0.03	0.23	0.27	0.29	0.31	0.35
## log_theta[36]	0.63	0.00	0.05	0.52	0.59	0.63	0.66	0.73
## log_theta[37]	0.10	0.00	0.02	0.05	0.08	0.10	0.11	0.14
## log_theta[38]	0.24	0.00	0.03	0.18	0.22	0.24	0.26	0.29
## log_theta[39]	-0.20	0.00	0.03	-0.25	-0.22	-0.20	-0.19	-0.15
## log_theta[40]	0.36	0.00	0.03	0.29	0.34	0.36	0.39	0.43
## log_theta[41]	0.03	0.00	0.02	-0.01	0.02	0.03	0.05	0.07
## log_theta[42]	0.02	0.00	0.02	-0.02	0.01	0.02	0.04	0.06
## log_theta[43]	0.18	0.00	0.03	0.13	0.17	0.18	0.20	0.23
## log_theta[44]	0.07	0.00	0.02	0.03	0.06	0.07	0.09	0.12
## log_theta[45]	-0.04	0.00	0.02	-0.08	-0.05	-0.04	-0.02	0.01
## log_theta[46]	0.35	0.00	0.03	0.28	0.33	0.35	0.38	0.42
## log_theta[47]	-0.24	0.00	0.03	-0.29	-0.25	-0.24	-0.22	-0.18
## log_theta[48]	-0.05	0.00	0.02	-0.09	-0.07	-0.05	-0.04	-0.01
## log_theta[49]	0.14	0.00	0.02	0.09	0.12	0.14	0.15	0.19
## log_theta[50]	-0.13	0.00	0.02	-0.17	-0.14	-0.13	-0.11	-0.08
## log_theta[51]	0.43	0.00	0.04	0.36	0.41	0.43	0.46	0.51
## log_theta[52]	0.29	0.00	0.03	0.23	0.27	0.29	0.31	0.35
## log_theta[53]	-0.09	0.00	0.02	-0.13	-0.10	-0.09	-0.08	-0.05
## log_theta[54]	-0.17	0.00	0.02	-0.22	-0.19	-0.17	-0.16	-0.13
## log_theta[55]	0.16	0.00	0.02	0.11	0.14	0.16	0.17	0.20
## log_theta[56]	0.17	0.00	0.02	0.12	0.15	0.17	0.18	0.21
## log_theta[57]	-0.20	0.00	0.03	-0.25	-0.22	-0.20	-0.19	-0.15
## log_theta[58]	0.45	0.00	0.04	0.37	0.42	0.45	0.48	0.53
## log_theta[59]	-0.17	0.00	0.02	-0.21	-0.18	-0.17	-0.15	-0.12
## log_theta[60]	-0.33	0.00	0.03	-0.39	-0.35	-0.33	-0.31	-0.26
## log_theta[61]	-0.01	0.00	0.02	-0.05	-0.02	-0.01	0.00	0.03
## log_theta[62]	-0.12	0.00	0.02	-0.16	-0.13	-0.12	-0.10	-0.07
## log_theta[63]	0.12	0.00	0.02	0.07	0.10	0.12	0.13	0.16
## log_theta[64]	-0.10	0.00	0.02	-0.15	-0.12	-0.10	-0.09	-0.06
## log_theta[65]	-0.21	0.00	0.03	-0.26	-0.23	-0.21	-0.19	-0.16
## log_theta[66]	-0.24	0.00	0.03	-0.29	-0.25	-0.24	-0.22	-0.18
## log_theta[67]	-0.32	0.00	0.03	-0.38	-0.34	-0.32	-0.30	-0.26

## log_theta[68]	0.12	0.00	0.02	0.08	0.11	0.12	0.14	0.17
## log_theta[69]	0.33	0.00	0.03	0.26	0.31	0.33	0.35	0.39
## log_theta[70]	-0.34	0.00	0.03	-0.40	-0.36	-0.34	-0.32	-0.28
## log_theta[71]	-0.23	0.00	0.03	-0.28	-0.25	-0.23	-0.21	-0.18
## log_theta[72]	-0.27	0.00	0.03	-0.33	-0.29	-0.27	-0.25	-0.22
## log_theta[73]	0.20	0.00	0.03	0.15	0.19	0.20	0.22	0.25
## log_theta[74]	0.30	0.00	0.03	0.23	0.28	0.30	0.32	0.36
## log_theta[75]	-0.01	0.00	0.02	-0.05	-0.02	-0.01	0.01	0.03
## log_theta[76]	0.20	0.00	0.03	0.14	0.18	0.20	0.21	0.25
## log_theta[77]	0.06	0.00	0.02	0.02	0.05	0.06	0.08	0.11
## log_theta[78]	-0.06	0.00	0.02	-0.10	-0.07	-0.06	-0.04	-0.02
## log_theta[79]	0.15	0.00	0.02	0.10	0.13	0.15	0.16	0.19
## log_theta[80]	0.18	0.00	0.03	0.13	0.17	0.18	0.20	0.23
## log_theta[81]	0.33	0.00	0.03	0.26	0.31	0.33	0.35	0.39
## log_theta[82]	-0.04	0.00	0.02	-0.09	-0.06	-0.04	-0.03	0.00
## log_theta[83]	-0.05	0.00	0.02	-0.10	-0.07	-0.05	-0.04	-0.01
## log_theta[84]	0.44	0.00	0.04	0.36	0.42	0.44	0.47	0.52
## log_theta[85]	0.25	0.00	0.03	0.20	0.23	0.25	0.27	0.31
## log_theta[86]	-0.21	0.00	0.03	-0.26	-0.22	-0.21	-0.19	-0.16
## log_theta[87]	-0.17	0.00	0.02	-0.21	-0.18	-0.17	-0.15	-0.12
## log_theta[88]	-0.02	0.00	0.02	-0.06	-0.04	-0.02	-0.01	0.02
## log_theta[89]	-0.11	0.00	0.02	-0.15	-0.12	-0.11	-0.09	-0.06
## log_theta[90]	0.16	0.00	0.02	0.11	0.15	0.16	0.18	0.21
## log_theta[91]	-0.24	0.00	0.03	-0.29	-0.25	-0.24	-0.22	-0.18
## log_theta[92]	0.25	0.00	0.03	0.20	0.23	0.25	0.27	0.31
## log_theta[93]	-0.38	0.00	0.03	-0.44	-0.40	-0.38	-0.36	-0.31
## log_theta[94]	-0.07	0.00	0.02	-0.11	-0.09	-0.07	-0.06	-0.03
## log_theta[95]	-0.09	0.00	0.02	-0.13	-0.10	-0.09	-0.07	-0.05
## log_theta[96]	-0.16	0.00	0.02	-0.20	-0.17	-0.16	-0.14	-0.11
## log_theta[97]	-0.28	0.00	0.03	-0.34	-0.30	-0.28	-0.26	-0.23
## log_theta[98]	-0.16	0.00	0.02	-0.21	-0.18	-0.16	-0.14	-0.11
## log_theta[99]	-0.09	0.00	0.02	-0.13	-0.10	-0.09	-0.08	-0.05
## log_theta[100]	0.35	0.00	0.03	0.28	0.32	0.35	0.37	0.41
## log_theta[101]	-0.09	0.00	0.02	-0.14	-0.11	-0.09	-0.08	-0.05
## log_theta[102]	-0.12	0.00	0.02	-0.16	-0.13	-0.12	-0.10	-0.07
## log_theta[103]	0.07	0.00	0.02	0.03	0.05	0.07	0.08	0.11
## log_theta[104]	-0.26	0.00	0.03	-0.32	-0.28	-0.26	-0.25	-0.21
## log_theta[105]	-0.10	0.00	0.02	-0.14	-0.11	-0.10	-0.08	-0.05
## log_theta[106]	-0.21	0.00	0.03	-0.26	-0.23	-0.21	-0.19	-0.16
## log_theta[107]	0.24	0.00	0.03	0.18	0.22	0.24	0.26	0.29
## log_theta[108]	-0.08	0.00	0.02	-0.12	-0.09	-0.08	-0.06	-0.03
## log_theta[109]	-0.11	0.00	0.02	-0.16	-0.13	-0.12	-0.10	-0.07
## log_theta[110]	-0.17	0.00	0.02	-0.22	-0.19	-0.17	-0.16	-0.12
## log_theta[111]	-0.25	0.00	0.03	-0.30	-0.26	-0.25	-0.23	-0.19
## log_theta[112]	-0.17	0.00	0.02	-0.21	-0.18	-0.17	-0.15	-0.12
## log_theta[113]	-0.23	0.00	0.03	-0.28	-0.24	-0.23	-0.21	-0.17
## log_theta[114]	-0.35	0.00	0.03	-0.42	-0.37	-0.35	-0.33	-0.29
## log_theta[115]	-0.17	0.00	0.02	-0.21	-0.18	-0.17	-0.15	-0.12
## log_theta[116]	-0.17	0.00	0.02	-0.22	-0.19	-0.17	-0.16	-0.12
## log_theta[117]	-0.21	0.00	0.03	-0.26	-0.23	-0.21	-0.20	-0.16
## log_theta[118]	0.30	0.00	0.03	0.24	0.28	0.30	0.32	0.36
## log_theta[119]	-0.23	0.00	0.03	-0.28	-0.24	-0.23	-0.21	-0.17
## log_theta[120]	-0.08	0.00	0.02	-0.12	-0.09	-0.08	-0.06	-0.04
## log_theta[121]	0.18	0.00	0.03	0.13	0.16	0.18	0.20	0.23

## log_theta[122]	-0.18	0.00	0.02	-0.22	-0.19	-0.18	-0.16	-0.13
## log_theta[123]	-0.14	0.00	0.02	-0.18	-0.15	-0.14	-0.12	-0.09
## log_theta[124]	-0.16	0.00	0.02	-0.21	-0.18	-0.16	-0.15	-0.12
## log_theta[125]	-0.02	0.00	0.02	-0.06	-0.04	-0.02	-0.01	0.02
## log_theta[126]	0.05	0.00	0.02	0.01	0.04	0.05	0.07	0.09
## log_theta[127]	0.24	0.00	0.03	0.19	0.22	0.24	0.26	0.30
## log_theta[128]	-0.06	0.00	0.02	-0.10	-0.07	-0.06	-0.04	-0.02
## log_theta[129]	0.02	0.00	0.02	-0.02	0.00	0.02	0.03	0.06
## log_theta[130]	0.02	0.00	0.02	-0.02	0.01	0.02	0.03	0.06
## log_theta[131]	-0.05	0.00	0.02	-0.10	-0.07	-0.05	-0.04	-0.01
## log_theta[132]	-0.08	0.00	0.02	-0.12	-0.09	-0.08	-0.06	-0.04
## log_theta[133]	-0.22	0.00	0.03	-0.27	-0.24	-0.22	-0.20	-0.17
## log_theta[134]	-0.26	0.00	0.03	-0.32	-0.28	-0.26	-0.25	-0.21
## log_theta[135]	0.05	0.00	0.02	0.01	0.04	0.05	0.07	0.10
## log_theta[136]	-0.10	0.00	0.02	-0.14	-0.11	-0.10	-0.08	-0.05
## log_theta[137]	-0.16	0.00	0.02	-0.21	-0.18	-0.16	-0.14	-0.11
## log_theta[138]	-0.24	0.00	0.03	-0.30	-0.26	-0.24	-0.23	-0.19
## log_theta[139]	-0.17	0.00	0.02	-0.22	-0.19	-0.17	-0.16	-0.12
## log_theta[140]	-0.06	0.00	0.02	-0.11	-0.08	-0.06	-0.05	-0.02
## log_theta[141]	-0.33	0.00	0.03	-0.39	-0.35	-0.33	-0.31	-0.27
## log_theta[142]	-0.32	0.00	0.03	-0.38	-0.34	-0.32	-0.30	-0.26
## log_theta[143]	-0.08	0.00	0.02	-0.12	-0.09	-0.08	-0.06	-0.04
## log_theta[144]	-0.29	0.00	0.03	-0.35	-0.31	-0.30	-0.28	-0.24
## log_theta[145]	-0.29	0.00	0.03	-0.34	-0.31	-0.29	-0.27	-0.23
## log_theta[146]	-0.26	0.00	0.03	-0.31	-0.28	-0.26	-0.24	-0.20
## log_theta[147]	-0.17	0.00	0.02	-0.21	-0.18	-0.17	-0.15	-0.12
## log_theta[148]	-0.21	0.00	0.03	-0.26	-0.23	-0.21	-0.19	-0.16
## log_theta[149]	-0.06	0.00	0.02	-0.10	-0.07	-0.06	-0.05	-0.02
## log_theta[150]	-0.18	0.00	0.02	-0.23	-0.20	-0.18	-0.17	-0.13
## log_theta[151]	-0.23	0.00	0.03	-0.28	-0.25	-0.23	-0.22	-0.18
## log_theta[152]	0.30	0.00	0.03	0.24	0.28	0.30	0.32	0.36
## log_theta[153]	0.14	0.00	0.02	0.09	0.12	0.14	0.15	0.18
## log_theta[154]	-0.09	0.00	0.02	-0.13	-0.10	-0.09	-0.07	-0.04
## log_theta[155]	-0.19	0.00	0.02	-0.24	-0.21	-0.19	-0.18	-0.14
## log_theta[156]	-0.11	0.00	0.02	-0.16	-0.13	-0.11	-0.10	-0.07
## log_theta[157]	0.02	0.00	0.02	-0.03	0.00	0.02	0.03	0.06
## log_theta[158]	-0.26	0.00	0.03	-0.32	-0.28	-0.26	-0.25	-0.21
## log_theta[159]	-0.23	0.00	0.03	-0.29	-0.25	-0.23	-0.22	-0.18
## log_theta[160]	-0.17	0.00	0.02	-0.22	-0.19	-0.17	-0.16	-0.12
## log_theta[161]	-0.27	0.00	0.03	-0.33	-0.29	-0.27	-0.25	-0.21
## log_theta[162]	-0.24	0.00	0.03	-0.29	-0.26	-0.24	-0.22	-0.19
## log_theta[163]	-0.26	0.00	0.03	-0.31	-0.27	-0.26	-0.24	-0.20
## log_theta[164]	-0.01	0.00	0.02	-0.05	-0.02	-0.01	0.01	0.04
## log_theta[165]	-0.12	0.00	0.02	-0.17	-0.14	-0.12	-0.11	-0.08
## log_theta[166]	0.01	0.00	0.02	-0.03	0.00	0.01	0.03	0.06
## log_theta[167]	-0.02	0.00	0.02	-0.06	-0.03	-0.02	0.00	0.02
## log_theta[168]	-0.08	0.00	0.02	-0.13	-0.10	-0.08	-0.07	-0.04
## log_theta[169]	-0.21	0.00	0.03	-0.26	-0.23	-0.21	-0.20	-0.16
## log_theta[170]	0.11	0.00	0.02	0.07	0.10	0.11	0.13	0.16
## log_theta[171]	-0.01	0.00	0.02	-0.05	-0.03	-0.01	0.00	0.03
## log_theta[172]	0.08	0.00	0.02	0.03	0.06	0.08	0.09	0.12
## log_theta[173]	-0.09	0.00	0.02	-0.14	-0.11	-0.09	-0.08	-0.05
## log_theta[174]	-0.29	0.00	0.03	-0.34	-0.31	-0.29	-0.27	-0.23
## log_theta[175]	-0.34	0.00	0.03	-0.41	-0.37	-0.34	-0.32	-0.28


```

## log_theta[176] -0.32 0.00 0.03 -0.37 -0.34 -0.32 -0.30 -0.26
## log_theta[177] -0.02 0.00 0.02 -0.06 -0.04 -0.02 -0.01 0.02
## log_theta[178] -0.24 0.00 0.03 -0.29 -0.25 -0.24 -0.22 -0.18
## log_theta[179] -0.17 0.00 0.02 -0.22 -0.19 -0.17 -0.15 -0.12
## log_theta[180] -0.23 0.00 0.03 -0.29 -0.25 -0.23 -0.22 -0.18
## log_theta[181] -0.27 0.00 0.03 -0.33 -0.29 -0.27 -0.26 -0.22
## log_theta[182] -0.30 0.00 0.03 -0.35 -0.32 -0.30 -0.28 -0.24
## log_theta[183] -0.17 0.00 0.02 -0.22 -0.19 -0.17 -0.16 -0.12
## log_theta[184] -0.31 0.00 0.03 -0.37 -0.33 -0.31 -0.29 -0.25
## log_theta[185] -0.21 0.00 0.03 -0.26 -0.23 -0.21 -0.19 -0.16
## log_theta[186] -0.14 0.00 0.02 -0.19 -0.16 -0.14 -0.13 -0.10
## log_theta[187] 0.13 0.00 0.02 0.09 0.12 0.13 0.15 0.18
## log_theta[188] -0.31 0.00 0.03 -0.37 -0.33 -0.31 -0.29 -0.25
## log_theta[189] -0.07 0.00 0.02 -0.12 -0.09 -0.07 -0.06 -0.03
## log_theta[190] -0.34 0.00 0.03 -0.40 -0.36 -0.34 -0.32 -0.27
## log_theta[191] -0.33 0.00 0.03 -0.39 -0.35 -0.33 -0.31 -0.27
## log_theta[192] -0.25 0.00 0.03 -0.30 -0.26 -0.25 -0.23 -0.19
## log_theta[193] -0.26 0.00 0.03 -0.32 -0.28 -0.26 -0.25 -0.21
## log_theta[194] -0.19 0.00 0.02 -0.24 -0.20 -0.19 -0.17 -0.14
## log_theta[195] -0.39 0.00 0.03 -0.46 -0.41 -0.39 -0.37 -0.32
## lp__ 3710.75 0.02 1.03 3708.04 3710.32 3711.09 3711.49 3711.75
## n_eff Rhat
## alpha 3809 1
## beta 3838 1
## log_theta[1] 3892 1
## log_theta[2] 3881 1
## log_theta[3] 3884 1
## log_theta[4] 3911 1
## log_theta[5] 3877 1
## log_theta[6] 3868 1
## log_theta[7] 3883 1
## log_theta[8] 3882 1
## log_theta[9] 3709 1
## log_theta[10] 3817 1
## log_theta[11] 3904 1
## log_theta[12] 3879 1
## log_theta[13] 3878 1
## log_theta[14] 3902 1
## log_theta[15] 3851 1
## log_theta[16] 3922 1
## log_theta[17] 3902 1
## log_theta[18] 3827 1
## log_theta[19] 3886 1
## log_theta[20] 3891 1
## log_theta[21] 3887 1
## log_theta[22] 3909 1
## log_theta[23] 3923 1
## log_theta[24] 3893 1
## log_theta[25] 3886 1
## log_theta[26] 3829 1
## log_theta[27] 3911 1
## log_theta[28] 3911 1
## log_theta[29] 3887 1
## log_theta[30] 3882 1

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## log_theta[72]	3753	1
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## log_theta[185]	3729	1
## log_theta[186]	3764	1
## log_theta[187]	3876	1
## log_theta[188]	3767	1
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## log_theta[190]	3767	1
## log_theta[191]	3767	1
## log_theta[192]	3743	1

```
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## log_theta[194]  3719    1
## log_theta[195]  3769    1
## lp__           1873    1
##
## Samples were drawn using NUTS(diag_e) at Wed Mar 15 17:09:25 2023.
## 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).
```

Model 2

$$\log(\theta_i) = \alpha_i + \beta x_i$$

where

$$\alpha_i \sim N(0, 1)$$

```
mod2 <- stan(data=stan_data, file = "lab9_2.stan")
```

```
## Warning in readLines(file, warn = TRUE): incomplete
## final line found on 'C:\Users\Alice\Desktop\GRAD
## SCHOOL\STA2201\STA2201AppliedStatAliceHuang\lab9_2.stan'

##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 6.9e-05 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 0.69 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 1: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 1: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 1: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 1: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 1: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 1: Iteration:  1001 / 2000 [ 50%] (Sampling)
## Chain 1: Iteration:  1200 / 2000 [ 60%] (Sampling)
## Chain 1: Iteration:  1400 / 2000 [ 70%] (Sampling)
## Chain 1: Iteration:  1600 / 2000 [ 80%] (Sampling)
## Chain 1: Iteration:  1800 / 2000 [ 90%] (Sampling)
## Chain 1: Iteration:  2000 / 2000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 0.746 seconds (Warm-up)
## Chain 1:                0.642 seconds (Sampling)
## Chain 1:                1.388 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 3.1e-05 seconds
```

```

## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.31 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 2: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.693 seconds (Warm-up)
## Chain 2:                0.638 seconds (Sampling)
## Chain 2:                1.331 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 5.5e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.55 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 3: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.64 seconds (Warm-up)
## Chain 3:                0.574 seconds (Sampling)
## Chain 3:                1.214 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 4.8e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.48 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:

```

```
## Chain 4: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 4: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration:  1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration:  1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration:  1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration:  1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration:  2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 1.069 seconds (Warm-up)
## Chain 4:                0.848 seconds (Sampling)
## Chain 4:                1.917 seconds (Total)
## Chain 4:
```

Model 3

$$\log \theta_i = \alpha_i + \beta x_i$$

with

$$\alpha_i \sim N(\mu, \sigma^2)$$

$$\mu \sim N(0, 1)$$

$$\sigma \sim N(0, 1)$$

```
mod3 <- stan(data=stan_data, file = "lab9_3.stan")
```

```
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 6.8e-05 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 0.68 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 1: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 1: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 1: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 1: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 1: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 1: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 1: Iteration:  1200 / 2000 [ 60%] (Sampling)
## Chain 1: Iteration:  1400 / 2000 [ 70%] (Sampling)
## Chain 1: Iteration:  1600 / 2000 [ 80%] (Sampling)
## Chain 1: Iteration:  1800 / 2000 [ 90%] (Sampling)
```

```

## Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 0.443 seconds (Warm-up)
## Chain 1: 0.422 seconds (Sampling)
## Chain 1: 0.865 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 4.3e-05 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.43 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.748 seconds (Warm-up)
## Chain 2: 0.531 seconds (Sampling)
## Chain 2: 1.279 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 2.6e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.26 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.536 seconds (Warm-up)
## Chain 3: 0.464 seconds (Sampling)

```



```

## Chain 3:          1 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 3.5e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.35 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 4: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 0.617 seconds (Warm-up)
## Chain 4:          0.453 seconds (Sampling)
## Chain 4:          1.07 seconds (Total)
## Chain 4:

```

Question 3

Make two plots (appropriately labeled and described) that illustrate the differences in estimated θ_i 's across regions and the differences in θ s across models.

```

mod1 %>% gather_draws(log_theta[i]) %>%
  median_qi() %>%
  rename(median_mod1 = .value, lower_mod1 = .lower, upper_mod1 = .upper) %>%
  select(i, median_mod1:upper_mod1) -> res_mod1

mod2 %>% gather_draws(log_theta[i]) %>%
  median_qi() %>%
  rename(median_mod2 = .value, lower_mod2 = .lower, upper_mod2 = .upper) %>%
  select(i, median_mod2:upper_mod2) -> res_mod2

mod3 %>% gather_draws(log_theta[i]) %>%
  median_qi() %>%
  rename(median_mod3 = .value, lower_mod3 = .lower, upper_mod3 = .upper) %>%
  select(i, median_mod3:upper_mod3) -> res_mod3

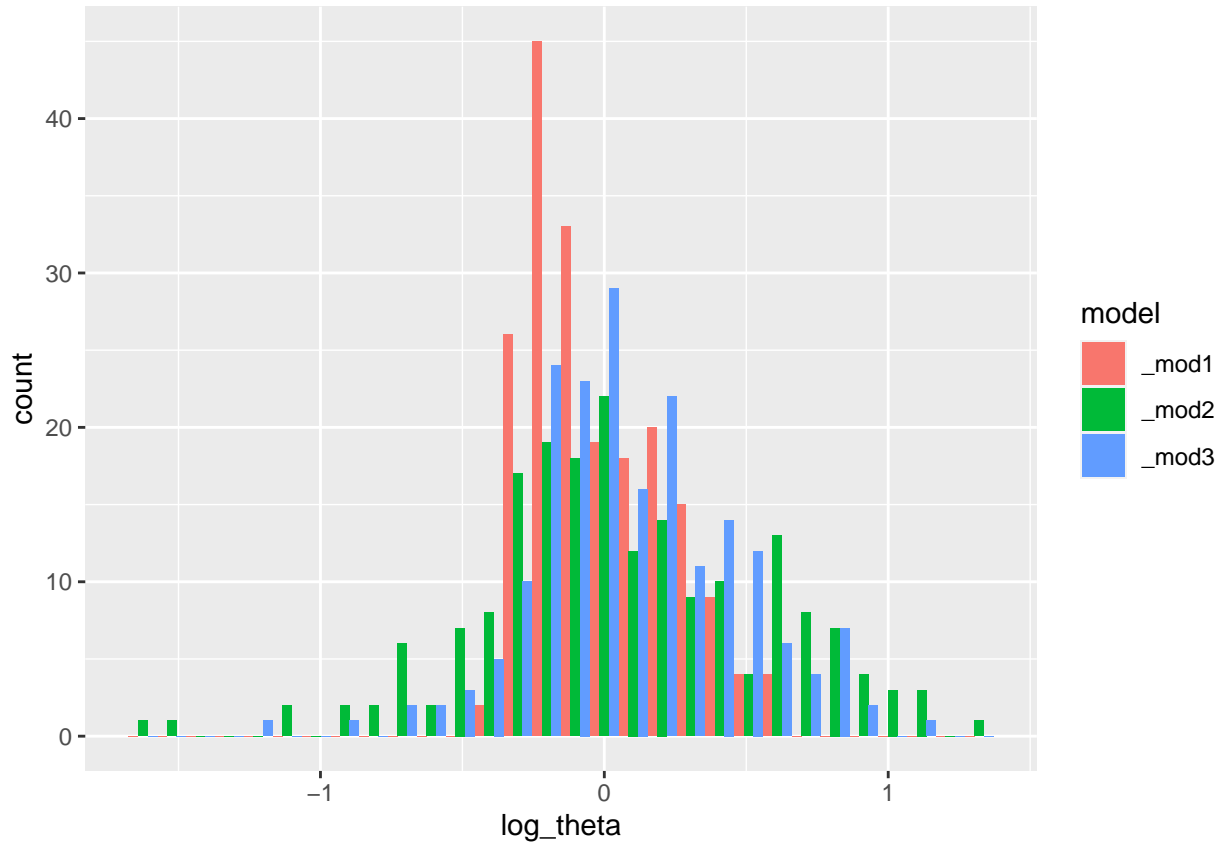
res <- res_mod1 %>% left_join(res_mod2) %>% left_join(res_mod3)

## Joining with 'by = join_by(i)'
## Joining with 'by = join_by(i)'

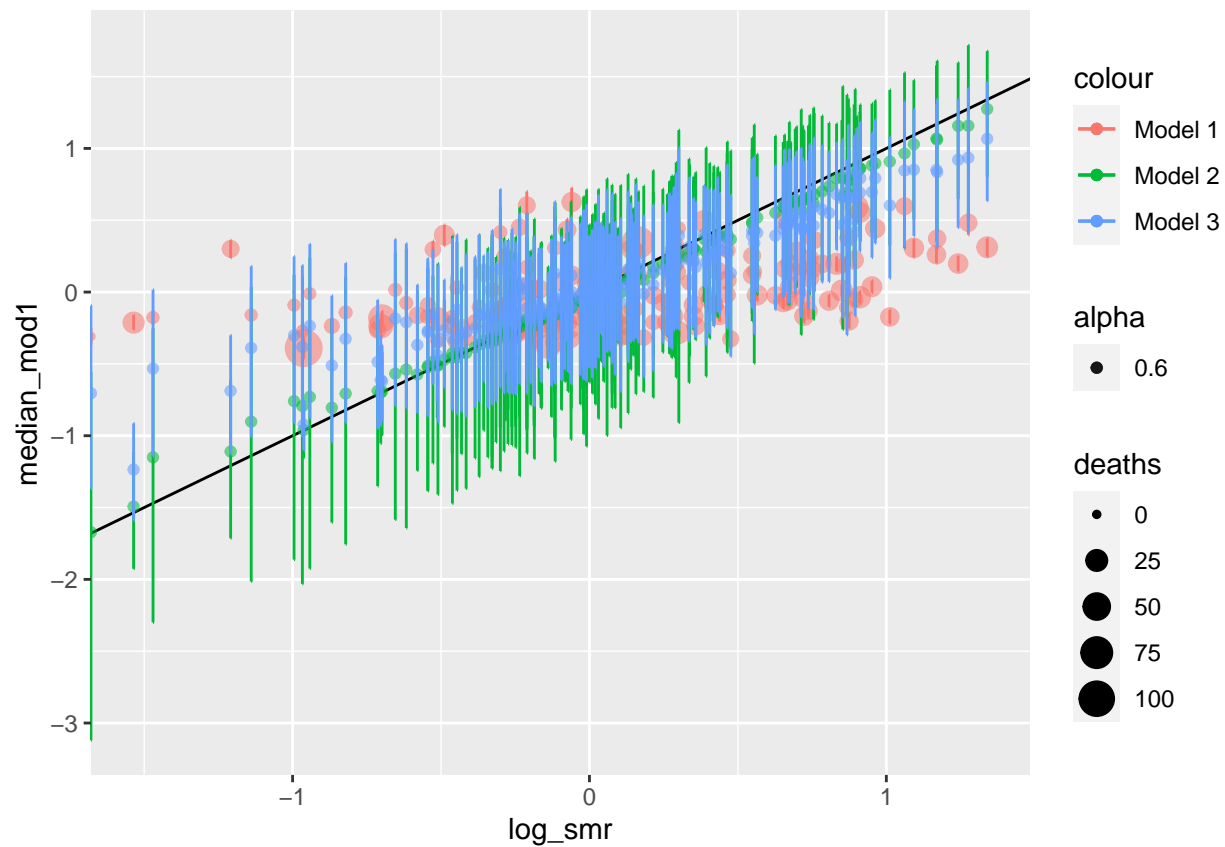
```

```
res %>% select(median_mod1, median_mod2, median_mod3) %>%
  pivot_longer(median_mod1:median_mod3, names_to = "model", values_to = "log_theta") %>%
  mutate(model = str_remove(model, "median")) %>%
  ggplot(aes(log_theta, fill = model)) + geom_histogram(position = "dodge")
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



```
res %>% mutate(deaths = observe.i) %>%
  mutate(log_smr = log(observe.i/expect.i)) -> res
res %>%
  ggplot(aes(log_smr, median_mod1, colour = "Model 1")) +
  geom_point(aes(size = deaths, alpha = 0.6)) +
  geom_errorbar(aes(ymin = lower_mod1, ymax = upper_mod1, color="Model 1")) +
  geom_abline(slope = 1, intercept = 0) +
  geom_point(aes(log_smr, median_mod2, color = "Model 2", alpha = 0.6)) +
  geom_errorbar(aes(ymin = lower_mod2, ymax = upper_mod2, color="Model 2")) +
  geom_point(aes(log_smr, median_mod3, color = "Model 3", alpha = 0.6)) +
  geom_errorbar(aes(ymin = lower_mod3, ymax = upper_mod3, color="Model 3"))
```



Model 2 and Model 3 look similar, but there is more uncertainty around Model 2 (bias variance tradeoff).