

Meeting 04/08/23

2023-08-03

Infections by transmission type

We produce plots showing the posterior distribution of the proportion of infections from each group. The groups are:

1 : Female to male, out-of-household
2 : Female to male, same household
3 : Male to female, out-of-household
4 : Male to female, same household

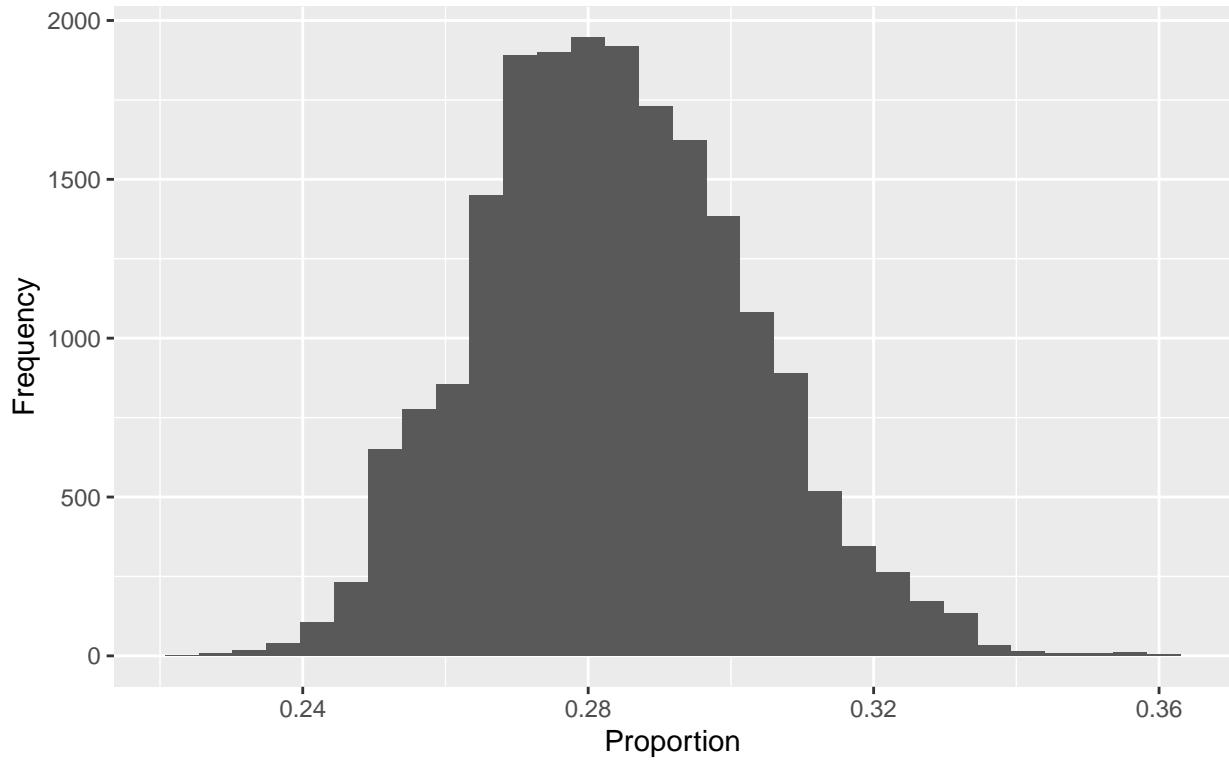
```
# Extract rates (eta)
eta_draws <- as_draws_matrix(fit$draws("eta"))
eta_sums <- rowSums(eta_draws)

eta_prop <- eta_draws / eta_sums

# Plot histograms of proportions
for(group in 1:4){
  p <- ggplot() + geom_histogram(aes(x = eta_prop[, group])) +
    labs(title = paste("Histogram of proportions - group", group),
        x = "Proportion", y = "Frequency",
        caption = "For four chains with 5,000 iterations each")
  print(p)
}

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

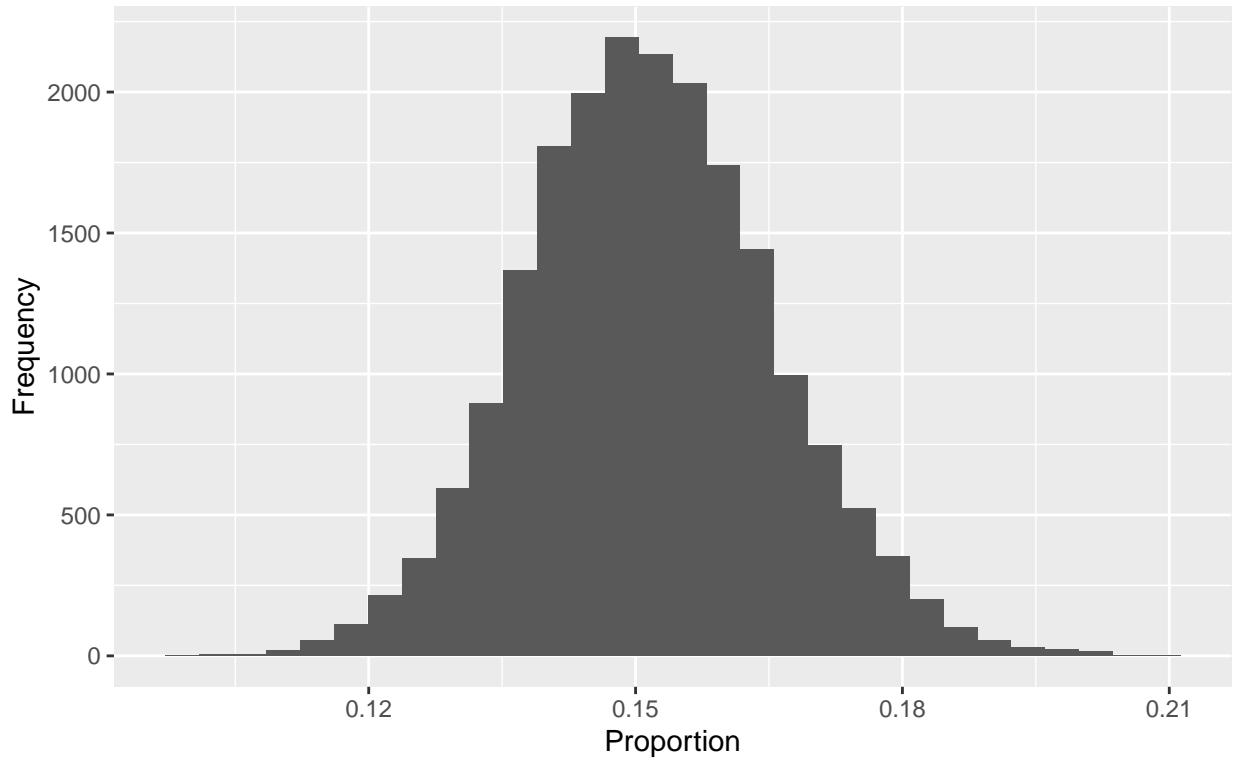
Histogram of proportions – group 1



For four chains with 5,000 iterations each

```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

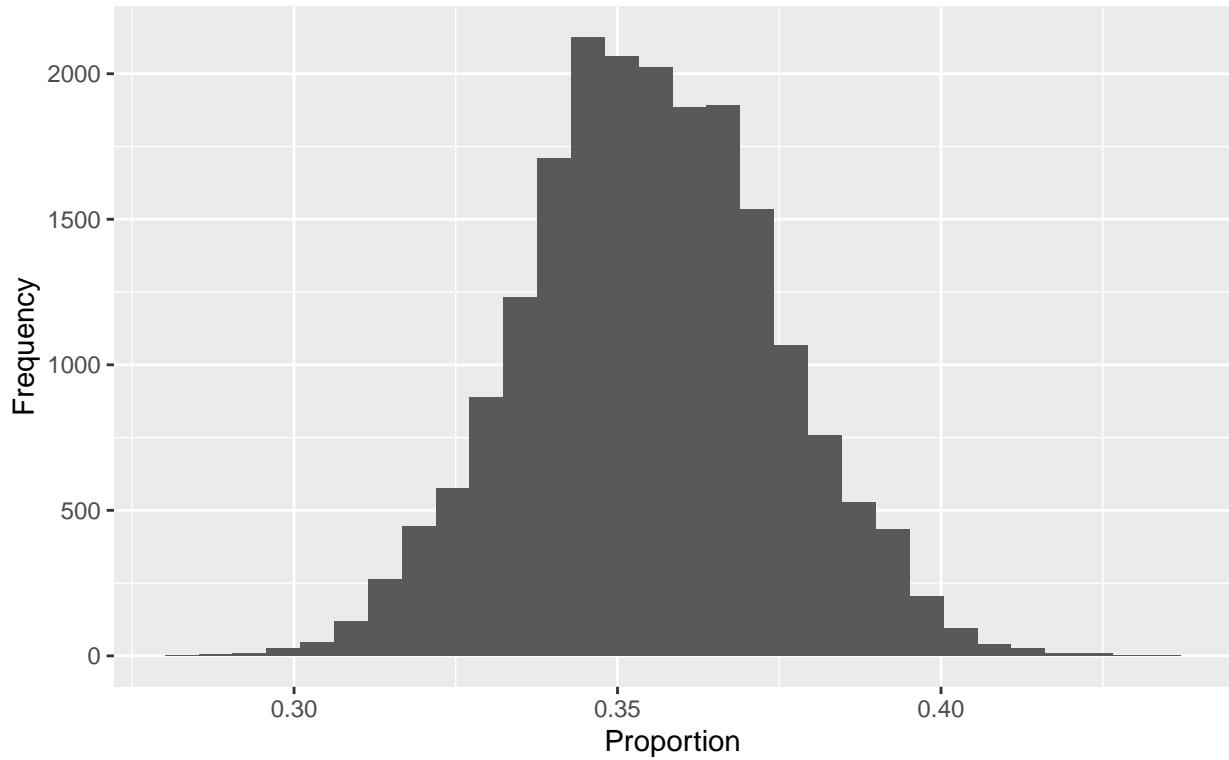
Histogram of proportions – group 2



For four chains with 5,000 iterations each

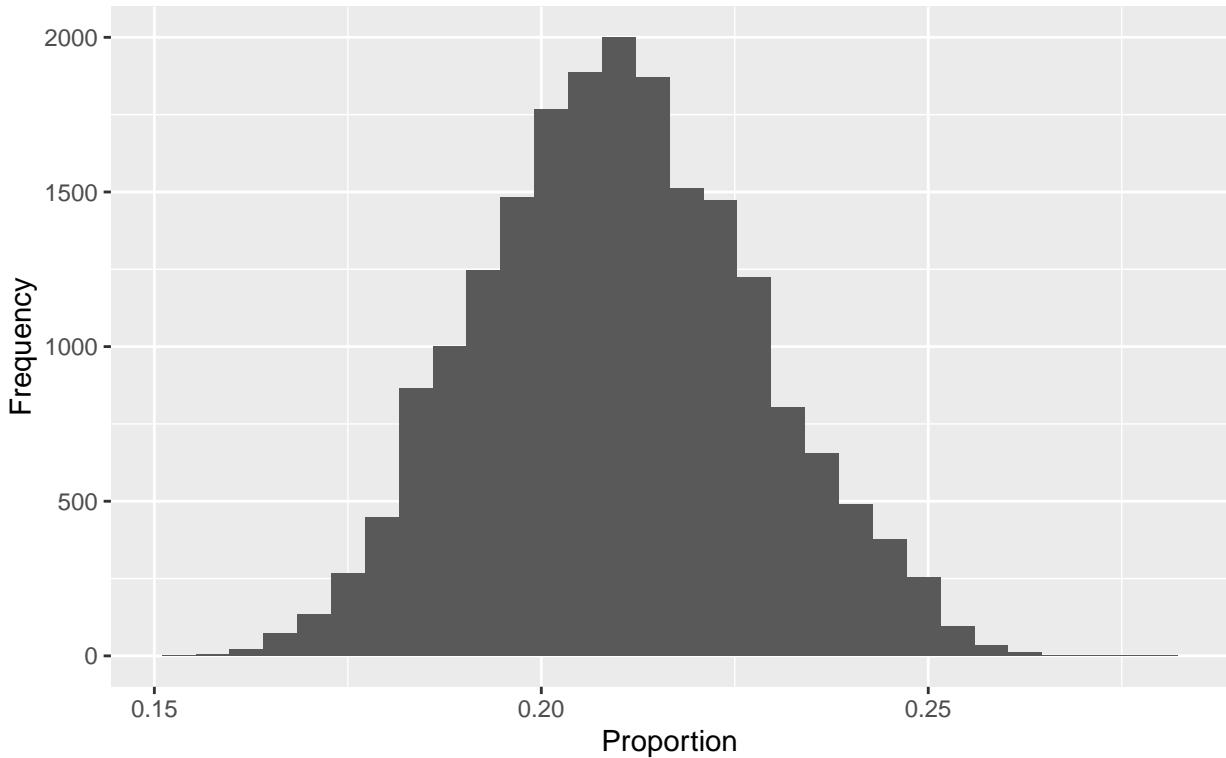
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

Histogram of proportions – group 3



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

Histogram of proportions – group 4



```
# Print out summary statistics
summary(eta_prop)
```

```
## # A tibble: 4 x 10
##   variable  mean  median    sd    mad    q5    q95  rhat ess_bulk ess_tail
##   <chr>     <num>   <num>  <num>  <num>  <num>  <num> <num>    <num>    <num>
## 1 eta[1]    0.284   0.283 0.0187 0.0189 0.254  0.315  1.03    84.0    417.
## 2 eta[2]    0.151   0.151 0.0139 0.0139 0.129  0.175  1.00   1829.   6972.
## 3 eta[3]    0.355   0.354 0.0194 0.0194 0.323  0.388  1.00    889.    440.
## 4 eta[4]    0.210   0.210 0.0177 0.0182 0.182  0.241  1.05    54.1    104.
```

We can see that the plurality of infections are M \rightarrow F and out-of-household. In fact, a majority of cases are indeed out-of-household, though there seems to be a considerable amount of within household transmission too.

As of right now, some of the parameters have a concerning Rhat value (but this should hopefully change with ordered model, fingers crossed!)

```
# Check proportion of HH vs OOH
eta_HH_prop <- rowSums(eta_prop[, c(2,4)])
summary(eta_HH_prop)
```

```
##      Min. 1st Qu. Median      Mean 3rd Qu.      Max.
## 0.2925 0.3473 0.3610 0.3615 0.3751 0.4383
```

```

print(quantile(eta_HH_prop, c(0.025, 0.975))) # 90% CI

##      2.5%    97.5%
## 0.3220805 0.4028591

```

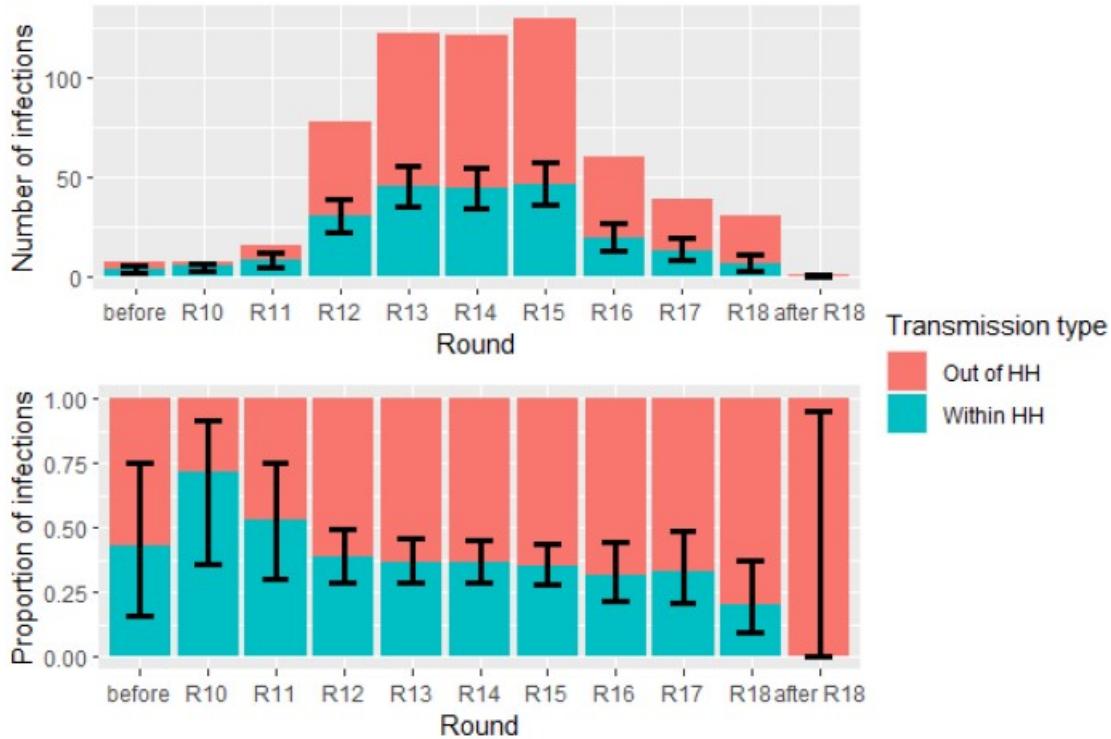


Figure 1: Caption

Resulting densities

We now plot the underlying distributions for each transmission type for one sample from the posterior distribution.

(What is the best way to aggregate the samples? Perhaps generate densities for every 50 iterations and take the mean at each grid point for plotting? May be computationally expensive to take mean of all)

```

source(here::here("helper-functions", "plot_normal.R"))

## Warning: package 'ggpubr' was built under R version 4.2.3

set.seed(635261)

# Randomly choose draw and chain number
draw_no <- sample(5000, 1)
chain_no <- sample(4, 1)

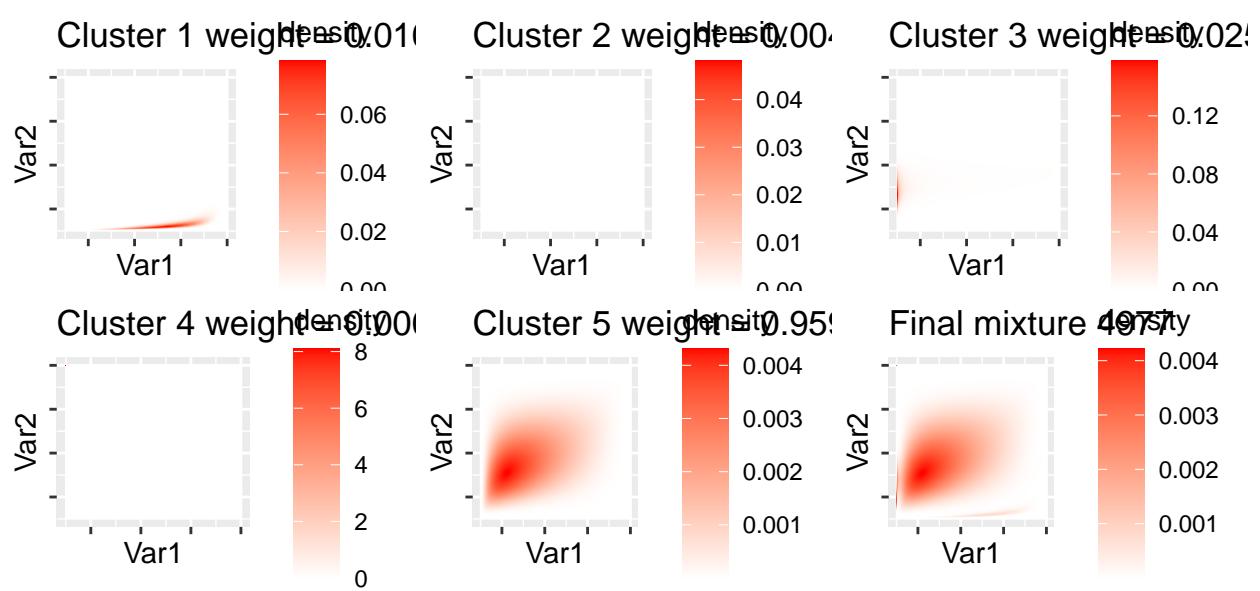
```

```

# Plot densities for all groups
for (group in 1:4){
  cat("Plotting", group, "\n") # progress bar!
  plot_normal(fit, chain_no, draw_no, group, "not_in_use")
}

## Plotting 1

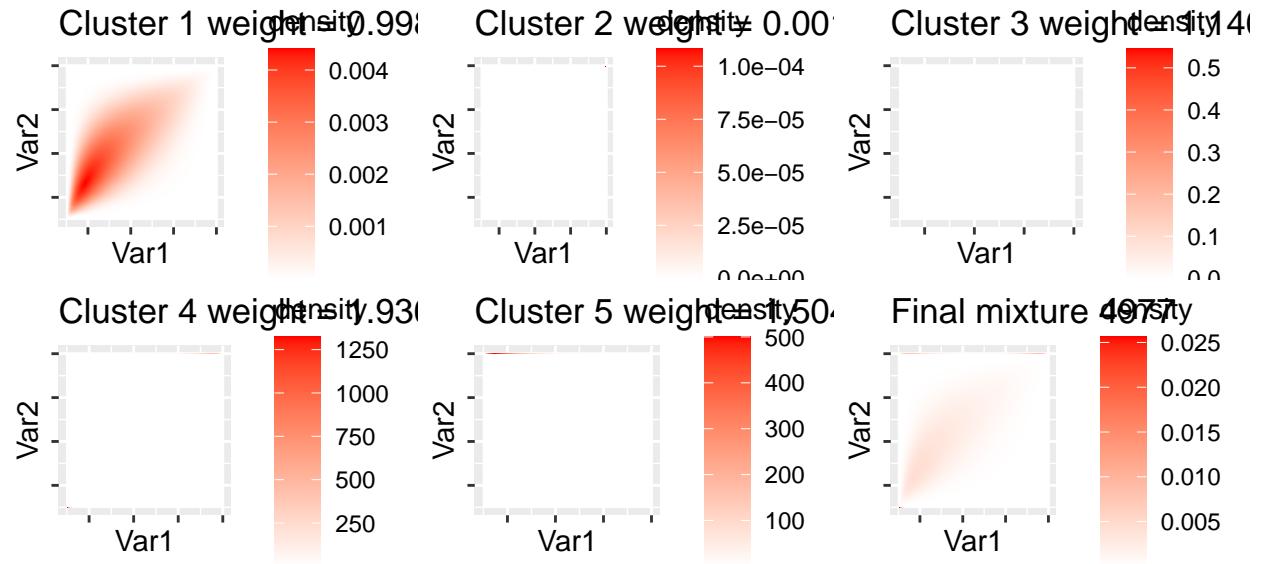
```



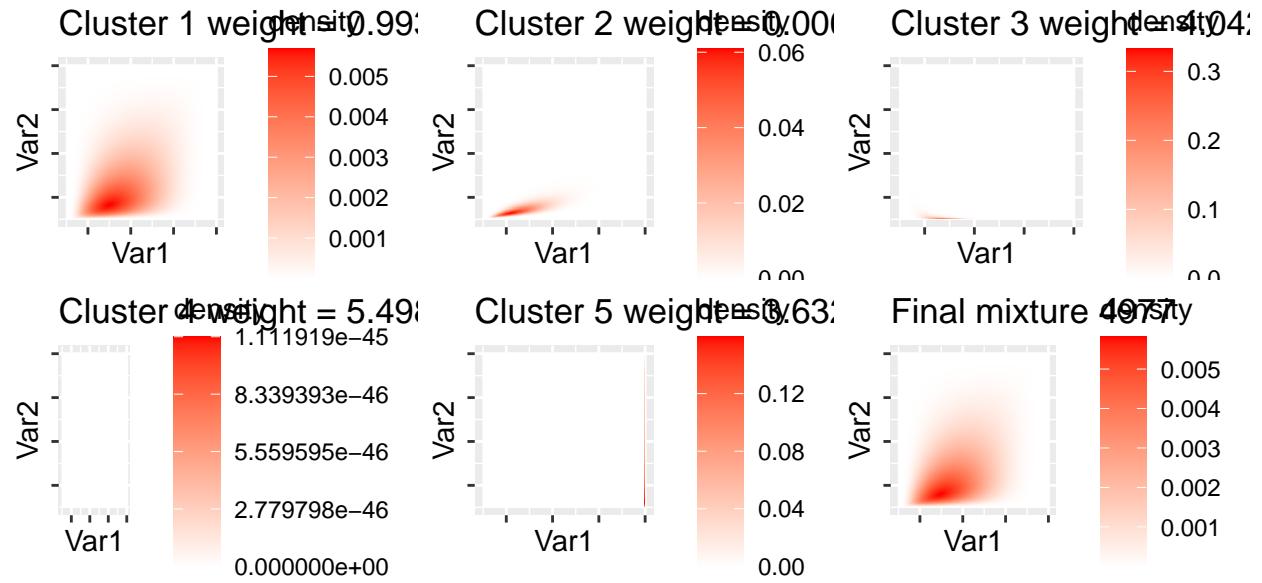
```

## Plotting 2

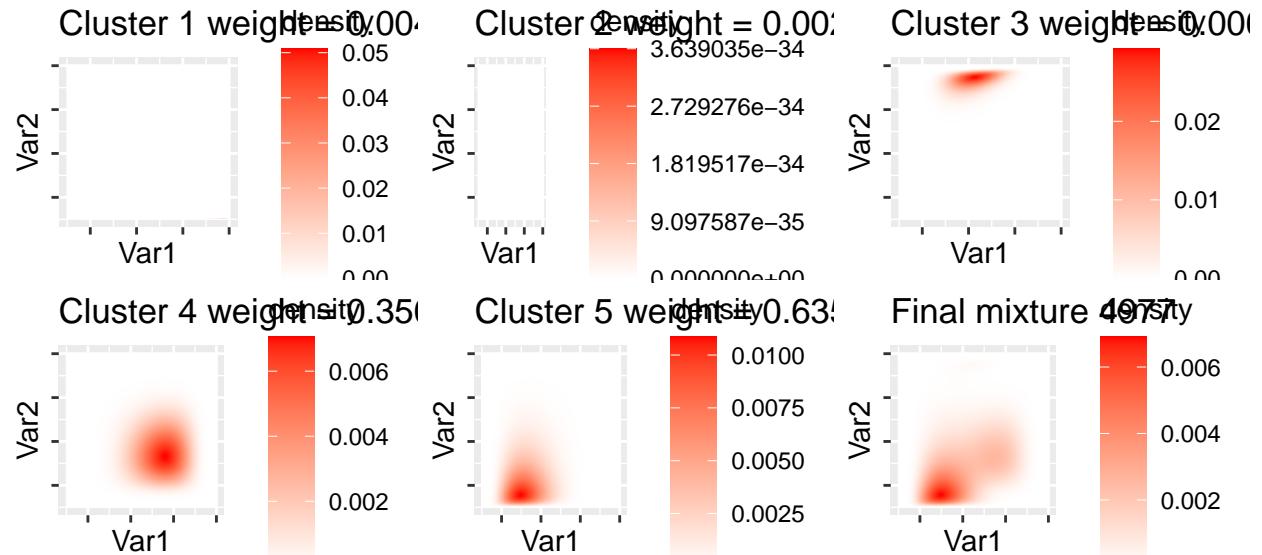
```



```
## Plotting 3
```



```
## Plotting 4
```



Compare to the results from the kde analysis:

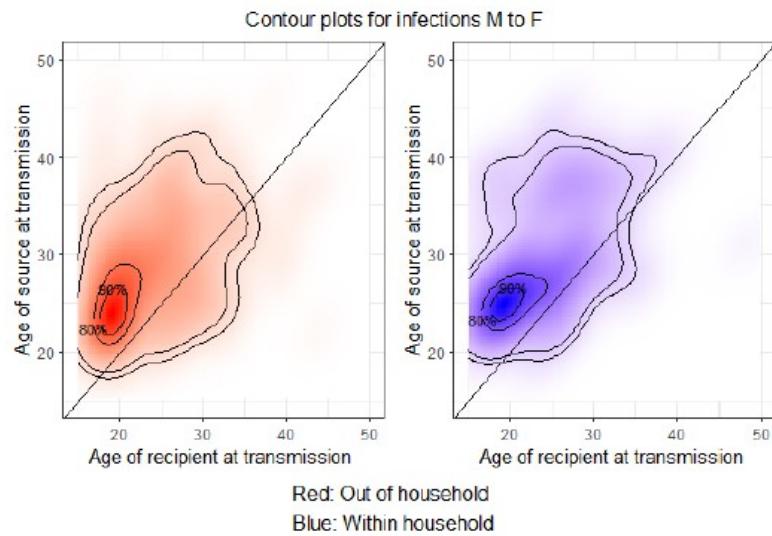


Figure 6: Contour plots for sources' and recipients' ages for infections male to female

Figure 2: Caption

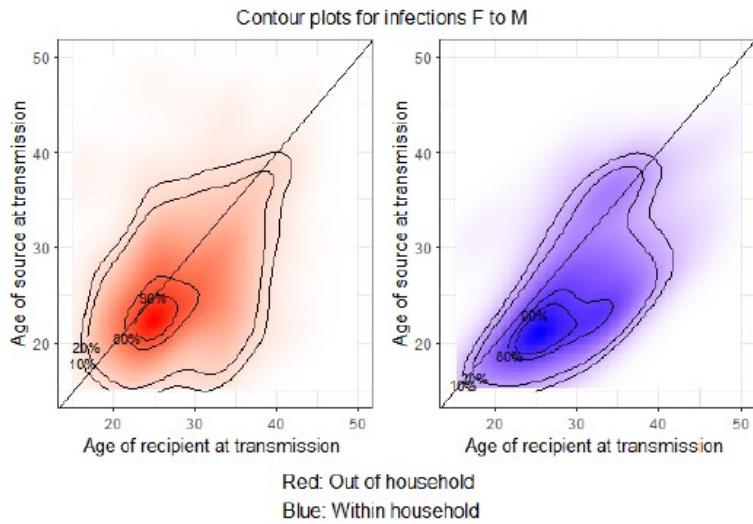


Figure 7: Contour plots for sources' and recipients' ages for infections female to male

Figure 3: Caption

Diagnosis on ordered model

```
fit_ordered <- readRDS(here::here("data", "logit_pairs_draws_ordered.rds"))
print(fit_ordered, max_rows = 200)
```

	variable	mean	median	sd	mad	q5	q95	rhat
##	lp_	833.96	834.38	11.25	11.18	814.76	851.70	1.00
##	v[1,1]	0.49	0.53	0.26	0.35	0.10	0.87	1.60
##	v[2,1]	0.98	0.99	0.04	0.01	0.95	1.00	1.00
##	v[3,1]	0.98	0.99	0.05	0.01	0.92	1.00	1.00
##	v[4,1]	0.81	0.97	0.26	0.05	0.24	1.00	1.02
##	v[1,2]	0.93	0.96	0.12	0.04	0.78	0.99	1.07
##	v[2,2]	0.63	0.70	0.30	0.35	0.08	1.00	1.00
##	v[3,2]	0.69	0.79	0.29	0.28	0.11	1.00	1.00
##	v[4,2]	0.74	0.85	0.28	0.21	0.17	1.00	1.00
##	v[1,3]	0.76	0.85	0.24	0.19	0.24	1.00	1.00
##	v[2,3]	0.60	0.64	0.30	0.38	0.07	0.99	1.00
##	v[3,3]	0.62	0.67	0.30	0.38	0.08	0.99	1.00
##	v[4,3]	0.66	0.73	0.30	0.33	0.10	0.99	1.00
##	v[1,4]	0.60	0.64	0.30	0.38	0.07	0.99	1.00
##	v[2,4]	0.59	0.62	0.31	0.40	0.07	0.99	1.00
##	v[3,4]	0.60	0.65	0.30	0.39	0.07	0.99	1.00
##	v[4,4]	0.59	0.63	0.31	0.40	0.07	0.99	1.00
##	alpha[1]	0.79	0.77	0.24	0.23	0.44	1.22	1.00
##	alpha[2]	0.71	0.69	0.23	0.23	0.38	1.12	1.00
##	alpha[3]	0.69	0.66	0.23	0.21	0.38	1.10	1.00
##	alpha[4]	0.76	0.73	0.24	0.23	0.41	1.19	1.00
##	eta[1]	175.48	174.97	13.80	13.45	153.78	198.37	1.01

## eta[2]	92.97	92.63	9.67	9.68	77.51	109.32	1.00
## eta[3]	218.11	217.64	14.91	14.79	194.43	243.29	1.00
## eta[4]	127.54	127.04	11.28	11.32	109.82	147.04	1.00
## mus_1[1,1]	-1.13	-0.95	0.53	0.18	-2.08	-0.76	1.37
## mus_1[2,1]	-0.92	-0.92	0.11	0.11	-1.11	-0.74	1.00
## mus_1[3,1]	-0.43	-0.43	0.07	0.06	-0.54	-0.33	1.00
## mus_1[4,1]	-0.46	-0.37	0.23	0.14	-0.91	-0.22	1.02
## mus_1[1,2]	-0.72	-0.75	0.21	0.12	-0.92	-0.41	1.09
## mus_1[2,2]	1.39	0.68	2.17	1.61	-0.78	5.91	1.00
## mus_1[3,2]	1.51	0.95	1.88	1.26	-0.29	5.49	1.00
## mus_1[4,2]	0.48	0.11	1.27	0.46	-0.33	2.94	1.01
## mus_1[1,3]	0.62	0.41	1.31	0.69	-0.61	2.48	1.00
## mus_1[2,3]	4.43	3.84	3.26	3.19	0.21	10.64	1.00
## mus_1[3,3]	4.45	3.77	3.22	3.07	0.52	10.67	1.00
## mus_1[4,3]	2.70	1.59	2.92	1.86	0.01	8.78	1.01
## mus_1[1,4]	4.63	3.62	3.87	3.39	0.33	12.34	1.00
## mus_1[2,4]	8.19	7.75	4.22	4.26	2.08	15.77	1.00
## mus_1[3,4]	8.20	7.75	4.22	4.33	2.14	15.84	1.00
## mus_1[4,4]	6.53	5.93	4.19	4.36	0.92	14.34	1.01
## mus_1[1,5]	11.05	10.34	6.01	6.11	2.43	21.90	1.00
## mus_1[2,5]	13.91	13.33	5.85	5.64	5.33	24.59	1.00
## mus_1[3,5]	13.78	13.19	5.76	5.65	5.35	24.20	1.00
## mus_1[4,5]	12.70	12.20	5.99	6.02	4.03	23.34	1.00
## mus_2[1,1]	-0.47	-0.52	0.36	0.22	-0.83	-0.03	1.18
## mus_2[2,1]	-0.29	-0.29	0.10	0.10	-0.45	-0.13	1.00
## mus_2[3,1]	-1.15	-1.15	0.08	0.07	-1.28	-1.03	1.00
## mus_2[4,1]	-1.31	-1.23	0.28	0.15	-1.88	-1.02	1.01
## mus_2[1,2]	-0.46	-0.46	0.49	0.27	-0.88	0.05	1.16
## mus_2[2,2]	-0.06	-0.40	8.94	7.32	-15.32	15.22	1.00
## mus_2[3,2]	-0.98	-1.83	7.45	3.75	-13.00	13.31	1.00
## mus_2[4,2]	-1.33	-0.98	4.83	2.77	-7.62	5.30	1.01
## mus_2[1,3]	-2.27	-2.82	3.82	0.94	-4.83	2.99	1.01
## mus_2[2,3]	-0.14	-0.21	9.88	9.53	-16.48	16.33	1.00
## mus_2[3,3]	-0.22	-0.79	9.37	8.76	-15.58	15.31	1.00
## mus_2[4,3]	-0.30	-0.41	8.14	6.35	-14.21	13.85	1.01
## mus_2[1,4]	-0.45	-0.90	9.20	8.51	-15.67	15.24	1.00
## mus_2[2,4]	0.23	0.23	9.92	9.78	-15.91	16.88	1.00
## mus_2[3,4]	0.21	0.08	10.12	9.86	-16.46	17.21	1.00
## mus_2[4,4]	0.21	0.30	9.76	9.32	-16.01	16.44	1.00
## mus_2[1,5]	-0.06	-0.22	9.80	9.73	-16.34	15.90	1.00
## mus_2[2,5]	-0.04	-0.23	9.97	10.22	-16.01	16.39	1.00
## mus_2[3,5]	0.19	0.13	9.85	9.87	-16.06	16.39	1.00
## mus_2[4,5]	0.09	-0.05	9.91	9.98	-15.93	16.36	1.00
## Sigmas[1,1,1,1]	1.70	0.79	1.55	0.64	0.37	4.37	1.54
## Sigmas[2,1,1,1]	1.09	1.08	0.17	0.16	0.85	1.41	1.00
## Sigmas[3,1,1,1]	0.61	0.61	0.07	0.07	0.50	0.73	1.00
## Sigmas[4,1,1,1]	0.53	0.58	0.37	0.14	0.16	0.77	1.02
## Sigmas[1,2,1,1]	2.13	1.78	2.74	1.83	0.42	5.30	1.51
## Sigmas[2,2,1,1]	2.77	0.68	26.89	0.62	0.17	7.28	1.00
## Sigmas[3,2,1,1]	4.69	0.68	78.33	0.57	0.17	6.64	1.00
## Sigmas[4,2,1,1]	1.75	0.61	25.13	0.40	0.19	4.32	1.01
## Sigmas[1,3,1,1]	1.46	0.60	8.90	0.50	0.16	4.09	1.01
## Sigmas[2,3,1,1]	3.40	0.75	23.11	0.72	0.17	9.11	1.00
## Sigmas[3,3,1,1]	4.74	0.74	54.24	0.70	0.17	10.62	1.00

## Sigmas[4,3,1,1]	4.42	0.69	68.89	0.64	0.16	9.08	1.00
## Sigmas[1,4,1,1]	3.02	0.66	27.20	0.61	0.16	8.19	1.00
## Sigmas[2,4,1,1]	5.59	0.71	63.62	0.67	0.16	11.74	1.00
## Sigmas[3,4,1,1]	3.45	0.76	27.20	0.72	0.17	8.93	1.00
## Sigmas[4,4,1,1]	3.45	0.72	37.14	0.67	0.17	9.53	1.00
## Sigmas[1,5,1,1]	3.67	0.75	24.52	0.70	0.16	10.12	1.00
## Sigmas[2,5,1,1]	3.29	0.71	15.87	0.66	0.17	10.03	1.00
## Sigmas[3,5,1,1]	3.11	0.72	34.70	0.66	0.17	7.53	1.00
## Sigmas[4,5,1,1]	55.24	0.70	3959.89	0.64	0.17	10.54	1.00
## Sigmas[1,1,2,1]	0.31	0.28	0.45	0.13	-0.03	0.79	1.17
## Sigmas[2,1,2,1]	0.57	0.56	0.12	0.11	0.40	0.79	1.00
## Sigmas[3,1,2,1]	0.25	0.25	0.06	0.06	0.15	0.35	1.00
## Sigmas[4,1,2,1]	0.29	0.30	0.15	0.11	0.06	0.49	1.01
## Sigmas[1,2,2,1]	0.27	0.31	0.87	0.14	-0.15	0.79	1.17
## Sigmas[2,2,2,1]	0.30	0.00	9.41	0.46	-2.08	2.37	1.00
## Sigmas[3,2,2,1]	-0.64	0.02	32.15	0.44	-2.06	1.92	1.00
## Sigmas[4,2,2,1]	0.25	0.13	7.28	0.37	-1.28	2.41	1.00
## Sigmas[1,3,2,1]	-0.16	-0.07	2.85	0.40	-1.67	1.21	1.00
## Sigmas[2,3,2,1]	-0.32	0.00	10.25	0.49	-3.61	2.88	1.01
## Sigmas[3,3,2,1]	-0.56	0.00	18.48	0.49	-3.08	2.78	1.00
## Sigmas[4,3,2,1]	0.62	0.02	18.75	0.46	-2.75	2.86	1.00
## Sigmas[1,4,2,1]	-0.21	-0.01	13.32	0.44	-2.53	2.35	1.00
## Sigmas[2,4,2,1]	1.43	0.00	39.25	0.48	-3.41	3.17	1.00
## Sigmas[3,4,2,1]	0.06	0.00	8.78	0.47	-2.69	2.72	1.00
## Sigmas[4,4,2,1]	-0.12	0.00	6.26	0.46	-2.67	2.48	1.00
## Sigmas[1,5,2,1]	-0.14	-0.02	12.52	0.47	-3.07	2.73	1.00
## Sigmas[2,5,2,1]	0.19	0.01	7.43	0.45	-2.50	3.45	1.00
## Sigmas[3,5,2,1]	0.28	0.01	10.05	0.45	-2.18	2.57	1.00
## Sigmas[4,5,2,1]	-17.07	0.00	677.87	0.48	-2.73	3.04	1.00
## Sigmas[1,1,1,2]	0.31	0.28	0.45	0.13	-0.03	0.79	1.17
## Sigmas[2,1,1,2]	0.57	0.56	0.12	0.11	0.40	0.79	1.00
## Sigmas[3,1,1,2]	0.25	0.25	0.06	0.06	0.15	0.35	1.00
## Sigmas[4,1,1,2]	0.29	0.30	0.15	0.11	0.06	0.49	1.01
## Sigmas[1,2,1,2]	0.27	0.31	0.87	0.14	-0.15	0.79	1.17
## Sigmas[2,2,1,2]	0.30	0.00	9.41	0.46	-2.08	2.37	1.00
## Sigmas[3,2,1,2]	-0.64	0.02	32.15	0.44	-2.06	1.92	1.00
## Sigmas[4,2,1,2]	0.25	0.13	7.28	0.37	-1.28	2.41	1.00
## Sigmas[1,3,1,2]	-0.16	-0.07	2.85	0.40	-1.67	1.21	1.00
## Sigmas[2,3,1,2]	-0.32	0.00	10.25	0.49	-3.61	2.88	1.01
## Sigmas[3,3,1,2]	-0.56	0.00	18.48	0.49	-3.08	2.78	1.00
## Sigmas[4,3,1,2]	0.62	0.02	18.75	0.46	-2.75	2.86	1.00
## Sigmas[1,4,1,2]	-0.21	-0.01	13.32	0.44	-2.53	2.35	1.00
## Sigmas[2,4,1,2]	1.43	0.00	39.25	0.48	-3.41	3.17	1.00
## Sigmas[3,4,1,2]	0.06	0.00	8.78	0.47	-2.69	2.72	1.00
## Sigmas[4,4,1,2]	-0.12	0.00	6.26	0.46	-2.67	2.48	1.00
## Sigmas[1,5,1,2]	-0.14	-0.02	12.52	0.47	-3.07	2.73	1.00
## Sigmas[2,5,1,2]	0.19	0.01	7.43	0.45	-2.50	3.45	1.00
## Sigmas[3,5,1,2]	0.28	0.01	10.05	0.45	-2.18	2.57	1.00
## Sigmas[4,5,1,2]	-17.07	0.00	677.87	0.48	-2.73	3.04	1.00
## Sigmas[1,1,2,2]	0.66	0.60	0.58	0.14	0.37	0.97	1.09
## Sigmas[2,1,2,2]	0.80	0.79	0.12	0.12	0.62	1.02	1.00
## Sigmas[3,1,2,2]	1.03	1.03	0.11	0.10	0.86	1.22	1.00
## Sigmas[4,1,2,2]	1.20	1.07	0.75	0.25	0.69	2.06	1.01
## Sigmas[1,2,2,2]	0.68	0.62	0.41	0.15	0.40	1.12	1.09

```

##  Sigmas[2,2,2,2]      2.77  0.72  20.28  0.66  0.17  7.10 1.00
##  Sigmas[3,2,2,2]      4.32  0.77  33.85  0.70  0.17  7.40 1.00
##  Sigmas[4,2,2,2]      3.57  0.86  15.97  0.80  0.20  11.98 1.00
##  Sigmas[1,3,2,2]      1.40  0.68  3.26  0.57  0.17  4.32 1.00
##  Sigmas[2,3,2,2]      5.37  0.75  33.53  0.71  0.17  11.97 1.01
##  Sigmas[3,3,2,2]      3.58  0.73  16.69  0.69  0.17  10.45 1.00
##  Sigmas[4,3,2,2]      3.58  0.75  19.88  0.72  0.17  10.10 1.00
##  Sigmas[1,4,2,2]      2.41  0.70  11.88  0.63  0.17  8.06 1.00
##  Sigmas[2,4,2,2]      5.23  0.72  51.65  0.66  0.17  12.73 1.00
##  Sigmas[3,4,2,2]      2.75  0.70  11.75  0.64  0.16  8.68 1.00
##  Sigmas[4,4,2,2]      2.33  0.70  7.72  0.64  0.17  8.11 1.00
##  Sigmas[1,5,2,2]      3.40  0.71  20.33  0.68  0.16  9.19 1.00
##  Sigmas[2,5,2,2]      2.61  0.71  9.39  0.66  0.17  8.95 1.00
##  Sigmas[3,5,2,2]      2.40  0.69  10.17  0.64  0.17  7.71 1.00
##  Sigmas[4,5,2,2]      28.58  0.73  406.26  0.68  0.17  11.40 1.00
##  weights[1,1]          0.49  0.53  0.26  0.35  0.10  0.87 1.60
##  weights[2,1]          0.98  0.99  0.04  0.01  0.95  1.00 1.00
##  weights[3,1]          0.98  0.99  0.05  0.01  0.92  1.00 1.00
##  weights[4,1]          0.81  0.97  0.26  0.05  0.24  1.00 1.02
##  weights[1,2]          0.48  0.44  0.26  0.34  0.09  0.86 1.56
##  weights[2,2]          0.01  0.00  0.04  0.00  0.00  0.04 1.00
##  weights[3,2]          0.02  0.00  0.05  0.00  0.00  0.08 1.00
##  weights[4,2]          0.16  0.02  0.24  0.03  0.00  0.71 1.02
##  weights[1,3]          0.03  0.01  0.06  0.01  0.00  0.06 1.01
##  weights[2,3]          0.00  0.00  0.01  0.00  0.00  0.01 1.00
##  weights[3,3]          0.00  0.00  0.01  0.00  0.00  0.01 1.00
##  weights[4,3]          0.02  0.00  0.07  0.00  0.00  0.15 1.02
##  weights[1,4]          0.00  0.00  0.01  0.00  0.00  0.01 1.01
##  weights[2,4]          0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  weights[3,4]          0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  weights[4,4]          0.00  0.00  0.01  0.00  0.00  0.01 1.00
##  weights[1,5]          0.00  0.00  0.00  0.00  0.00  0.01 1.00
##  weights[2,5]          0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  weights[3,5]          0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  weights[4,5]          0.00  0.00  0.00  0.00  0.00  0.01 1.01
##  cumprod_one_minus_v[1,1] 0.51  0.47  0.26  0.35  0.13  0.90 1.60
##  cumprod_one_minus_v[2,1] 0.02  0.01  0.04  0.01  0.00  0.05 1.00
##  cumprod_one_minus_v[3,1] 0.02  0.01  0.05  0.01  0.00  0.08 1.00
##  cumprod_one_minus_v[4,1] 0.19  0.03  0.26  0.05  0.00  0.76 1.02
##  cumprod_one_minus_v[1,2] 0.03  0.02  0.06  0.01  0.00  0.07 1.01
##  cumprod_one_minus_v[2,2] 0.00  0.00  0.01  0.00  0.00  0.02 1.00
##  cumprod_one_minus_v[3,2] 0.00  0.00  0.01  0.00  0.00  0.01 1.00
##  cumprod_one_minus_v[4,2] 0.03  0.00  0.07  0.01  0.00  0.17 1.02
##  cumprod_one_minus_v[1,3] 0.01  0.00  0.01  0.00  0.00  0.02 1.01
##  cumprod_one_minus_v[2,3] 0.00  0.00  0.00  0.00  0.00  0.01 1.00
##  cumprod_one_minus_v[3,3] 0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  cumprod_one_minus_v[4,3] 0.00  0.00  0.01  0.00  0.00  0.01 1.01
##  cumprod_one_minus_v[1,4] 0.00  0.00  0.00  0.00  0.00  0.01 1.00
##  cumprod_one_minus_v[2,4] 0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  cumprod_one_minus_v[3,4] 0.00  0.00  0.00  0.00  0.00  0.00 1.00
##  cumprod_one_minus_v[4,4] 0.00  0.00  0.00  0.00  0.00  0.01 1.01
##  ess_bulk ess_tail
##      937     3162
##      6       80

```

```
##      2903     2743
##      1914     1253
##      184      304
##      37      558
##     2838     1246
##     3349     4995
##      520     2103
##    1392      709
##    8633     6215
##    3888     3738
##   1275     1984
##   3638     1656
##   2773     2797
##   5567     3564
##   3094     6065
##   2821     1150
##   2730     1765
##   1889     1129
##   2288     2691
##    966      644
##   2648     1976
##   4462     4979
##   3980     8681
##      8      84
##   3140     1942
##   1929     1449
##   221      491
##    26      102
##   5159     4215
##   3702     4776
##    926     2252
##   1385     1019
##   7644     7793
##   2063     2794
##    853     1412
##   1613     1569
## 10093     7812
##   2848     3287
##   1312     1319
##   4149     2761
##   9027     5494
##   7496    10269
##   4589     5722
##      15      82
##   2704     1818
##   2732     1811
##    274      438
##      17     109
##   4499     8334
##   4278     4676
##   2164     3464
##   2548     1560
##   5656     4932
##   4014     4378
```

```
##      4903      6763
##      4223      2838
##      4106      1004
##      5157      3601
##      4611      5657
##      5385      7703
##      4099      4902
##      5460      9152
##      3960      5844
##      7       92
##      2620     1894
##      1925     1363
##      244      466
##      7       143
##      5015     5989
##      1877     1439
##      1891     2648
##      1508     1698
##      2077     4515
##      2850     1626
##      2015     2684
##      2792     4678
##      1230      608
##      3774     3541
##      2199     3662
##      3165     2706
##      1875     1210
##      3802     4245
##      1542      756
##      970       94
##      2956     1511
##      2053     2013
##      343       299
##      474       118
##      3696     3247
##      2226     2166
##      2570     2433
##      2334     2549
##      712       270
##      2171     1576
##      2696     1988
##      4425     2413
##      2919     1928
##      3184     3343
##      3082     2967
##      1924     1703
##      2355      824
##      3288     3120
##      2142     1463
##      970       94
##      2956     1511
##      2053     2013
##      343       299
##      474       118
```

```
##      3696      3247
##      2226      2166
##      2570      2433
##      2334      2549
##      712       270
##     2171      1576
##     2696      1988
##     4425      2413
##     2919      1928
##     3184      3343
##     3082      2967
##     1924      1703
##     2355       824
##     3288      3120
##     2142      1463
##      827       102
##     3878      2777
##     2883      2174
##      510       235
##     1161       110
##     3292      3204
##     2329      1580
##     1452      2589
##     3257      2726
##      727       264
##     2305      1447
##     1958      1676
##     2576      2984
##     2306      1595
##     2070      2803
##     3104      2767
##     1822      1898
##     2543      2311
##     3028      2526
##     1125       646
##      6        80
##     2903      2743
##     1914      1253
##      184       304
##       6        84
##     2722      2638
##     1960      1263
##      192       264
##      943       932
##     2906      4606
##     3865      6177
##      400       228
##     1112       656
##     4060      4679
##     3782      5964
##     1751      1531
##     1340       639
##     3060      2102
##     2821      1853
```

```

##      1196    2945
##        6     80
##      2897   2743
##      1914   1253
##      184    304
##      783    914
##      3194   3149
##      3368   6135
##      384    239
##     1027    636
##      3906   1814
##      3167   4252
##      1342   3562
##      1340    639
##      3060   2102
##      2821   1853
##      1196   2945

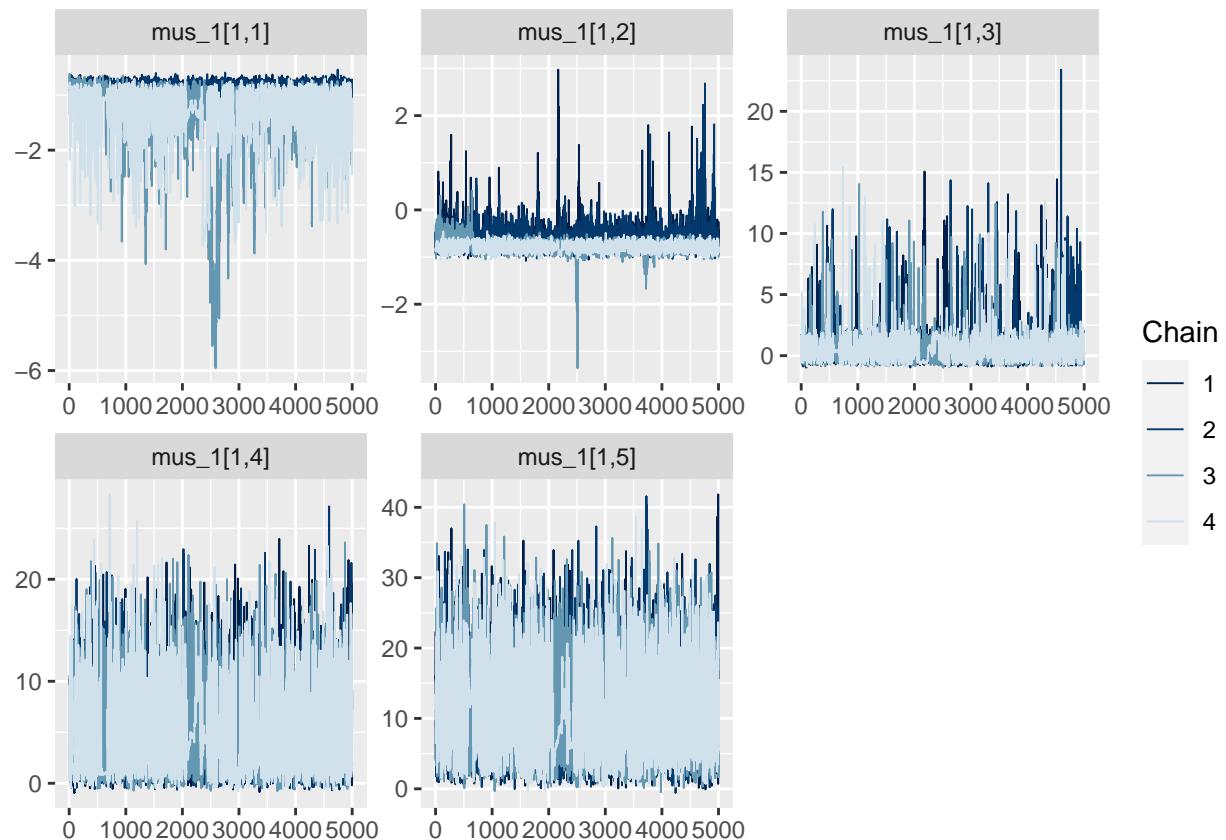
```

The first group seems to have a bit of issues with mixing, so let's plot their traces:

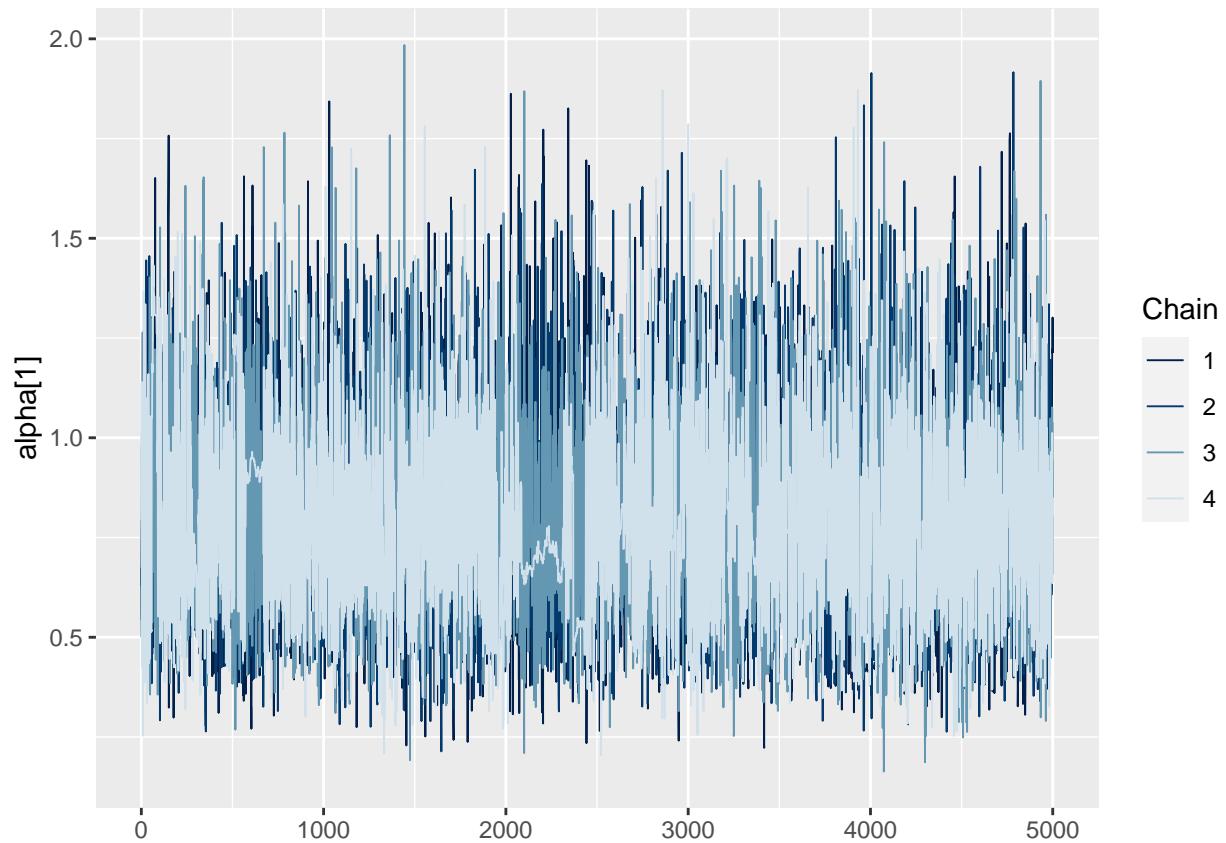
```

# Plot mu_1 for group 1
mcmc_trace(fit_ordered$draws((c("mus_1[1,1]", "mus_1[1,2]", "mus_1[1,3]",
                                "mus_1[1,4]", "mus_1[1,5]"))))

```

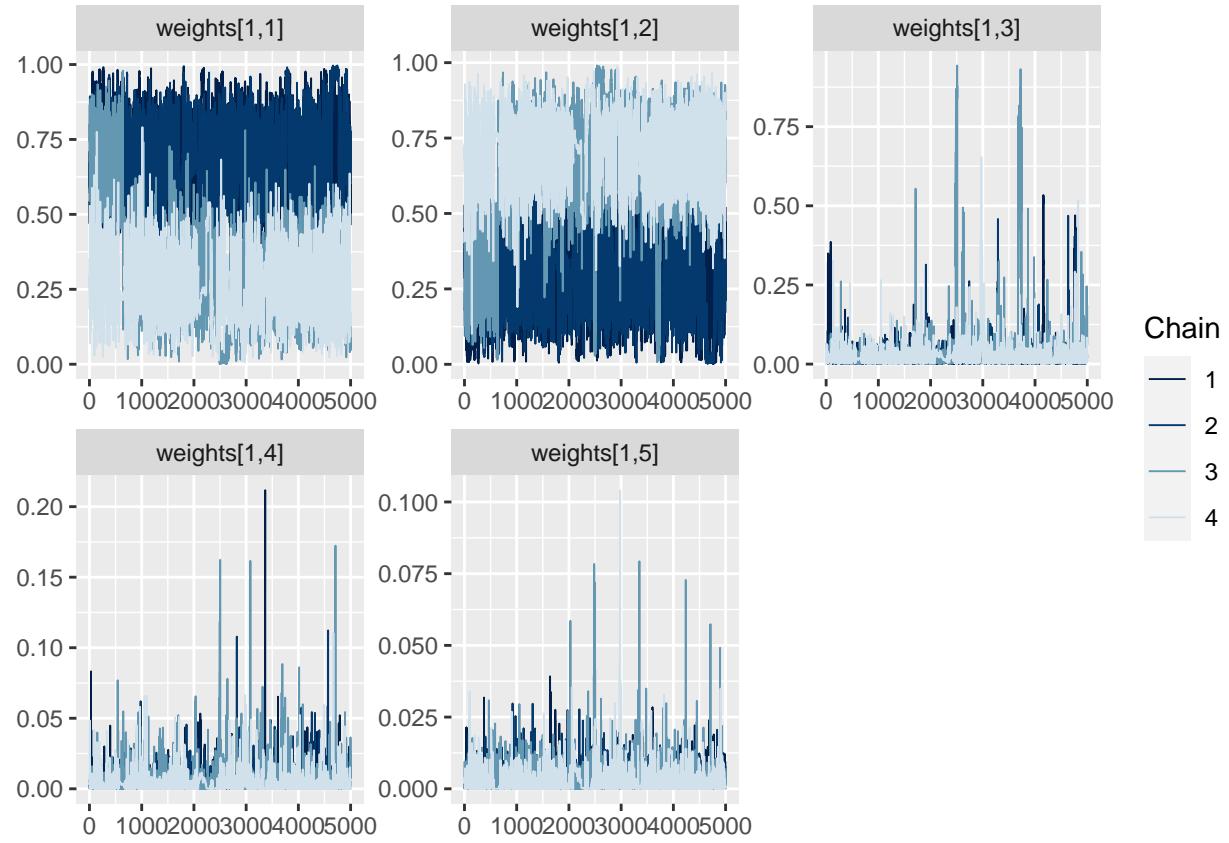


```
# Plot alpha for group 1  
mcmc_trace(fit_ordered$draws("alpha[1]"))
```



Perhaps chain 4 is the issue?

```
# Plot weights for group 1  
mcmc_trace(fit_ordered$draws((c("weights[1,1]", "weights[1,2]", "weights[1,3]",  
"weights[1,4]", "weights[1,5]"))))
```



Probably still multi-modal, two main clusters. All chains identify the last three clusters to be undesirable. No idea why chain 4 suddenly stopped working in the middle.