

Meeting 09-08-23

2023-08-07

1.1 Is there still as sizeable proportion of infections occuring within household?

Using the logit model which includes all the data. Recall from the previous Rmd:

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

```
# Load CmdStanR fits
fit_norm <- readRDS(here::here("data", "logit_pairs_draws_ordered.rds"))
#fit_beta <- readRDS(here::here("data", "beta_pairs_draws_ordered.rds"))

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

eta_prop <- eta_draws / eta_sums

# 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.
## 0.2850  0.3458  0.3593  0.3591  0.3723  0.4398
```

```
print(quantile(eta_HH_prop, c(0.025, 0.975))) # 95% CI
```

```
##      2.5%      97.5%
## 0.3209107 0.3979944
```

Is there a specific time which we should start the modelling from (i.e a date/round for which we start counting the intervention as 'significant'?)

1.2 Does HH proportion change with community (fishing vs inland?)

First we observe a quirk with the data:

```
# Load in pairs data
filename <- here::here("data", "pairs_tsi.csv")
pairs_tsi <- read.csv(filename)
setDT(pairs_tsi)

# Check whether same household pairs are the same community
```

```

same_hh_pairs <- pairs_tsi[same_hh == 1]
source_comm <- same_hh_pairs[,.(COMM.SOURCE)]
recip_comm <- same_hh_pairs[,.(COMM.RECIPIENT)]

t(source_comm == recip_comm) # Observation 81 is false

```

```

##           [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12]
## COMM.SOURCE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,13] [,14] [,15] [,16] [,17] [,18] [,19] [,20] [,21] [,22] [,23]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,24] [,25] [,26] [,27] [,28] [,29] [,30] [,31] [,32] [,33] [,34]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,35] [,36] [,37] [,38] [,39] [,40] [,41] [,42] [,43] [,44] [,45]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,46] [,47] [,48] [,49] [,50] [,51] [,52] [,53] [,54] [,55] [,56]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,57] [,58] [,59] [,60] [,61] [,62] [,63] [,64] [,65] [,66] [,67]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,68] [,69] [,70] [,71] [,72] [,73] [,74] [,75] [,76] [,77] [,78]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,79] [,80] [,81] [,82] [,83] [,84] [,85] [,86] [,87] [,88] [,89]
## COMM.SOURCE  TRUE TRUE FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,90] [,91] [,92] [,93] [,94] [,95] [,96] [,97] [,98] [,99] [,100]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,101] [,102] [,103] [,104] [,105] [,106] [,107] [,108] [,109]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,110] [,111] [,112] [,113] [,114] [,115] [,116] [,117] [,118]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,119] [,120] [,121] [,122] [,123] [,124] [,125] [,126] [,127]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,128] [,129] [,130] [,131] [,132] [,133] [,134] [,135] [,136]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,137] [,138] [,139] [,140] [,141] [,142] [,143] [,144] [,145]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,146] [,147] [,148] [,149] [,150] [,151] [,152] [,153] [,154]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,155] [,156] [,157] [,158] [,159] [,160] [,161] [,162] [,163]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,164] [,165] [,166] [,167] [,168] [,169] [,170] [,171] [,172]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,173] [,174] [,175] [,176] [,177] [,178] [,179] [,180] [,181]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,182] [,183] [,184] [,185] [,186] [,187] [,188] [,189] [,190]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,191] [,192] [,193] [,194] [,195] [,196] [,197] [,198] [,199]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,200] [,201] [,202] [,203] [,204] [,205] [,206] [,207] [,208]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,209] [,210] [,211] [,212] [,213] [,214] [,215] [,216] [,217]
## COMM.SOURCE  TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##           [,218] [,219]
## COMM.SOURCE  TRUE TRUE

```

The 81st entry of the same `_hh` data set has a source and recipient from the same household but different community! Do we ignore this data point? (Yes in this model).

We should fit the model to four groups:

- Same household, fishing (1)
- Same household, inland (2)
- Different household, fishing* (4)
- Different household, inland* (5)

But how do we decide which community if the source and recipient are not from the same community? Just from source/recipient community or add a separate group from an inter-community transmission (3). (The model picked uses the last option, so we now have five groups)

```
# Load CmdStanR fit
fit_1_2 <- readRDS(here::here("data", "logit_pairs_draws_1-2.rds"))
print(fit_1_2, max_rows = 1000)
```

##	variable	mean	median	sd	mad	q5	q95	rhat
##	lp_	601.31	601.49	11.73	11.71	581.66	620.25	1.02
##	v[1,1]	0.64	0.88	0.38	0.18	0.06	1.00	1.06
##	v[2,1]	0.69	0.93	0.39	0.11	0.01	1.00	1.01
##	v[3,1]	0.72	0.96	0.36	0.06	0.03	1.00	1.01
##	v[4,1]	0.22	0.06	0.33	0.07	0.00	0.97	1.06
##	v[5,1]	0.80	0.98	0.33	0.03	0.02	1.00	1.04
##	v[1,2]	0.72	0.87	0.31	0.18	0.11	1.00	1.01
##	v[2,2]	0.68	0.81	0.33	0.27	0.03	1.00	1.00
##	v[3,2]	0.70	0.82	0.30	0.24	0.10	1.00	1.00
##	v[4,2]	0.56	0.58	0.37	0.58	0.02	1.00	1.01
##	v[5,2]	0.70	0.82	0.30	0.25	0.11	1.00	1.01
##	alpha[1]	0.80	0.76	0.29	0.27	0.41	1.35	1.01
##	alpha[2]	0.82	0.78	0.31	0.28	0.42	1.40	1.00
##	alpha[3]	0.81	0.77	0.28	0.26	0.43	1.33	1.00
##	alpha[4]	0.99	0.94	0.35	0.34	0.49	1.64	1.01
##	alpha[5]	0.77	0.73	0.27	0.25	0.40	1.27	1.01
##	eta[1]	116.96	116.66	10.90	10.81	99.43	135.48	1.00
##	eta[2]	102.94	102.59	10.20	10.17	86.76	120.25	1.00
##	eta[3]	72.91	72.55	8.55	8.39	59.44	87.67	1.00
##	eta[4]	178.03	177.70	13.37	13.22	156.65	200.70	1.00
##	eta[5]	143.02	142.74	11.87	11.86	123.98	162.78	1.00
##	mus_1[1,1]	-0.92	-0.73	0.70	0.23	-1.49	-0.51	1.05
##	mus_1[2,1]	-1.36	-0.58	2.26	0.19	-5.32	-0.37	1.00
##	mus_1[3,1]	-1.26	-0.87	1.24	0.19	-3.45	-0.64	1.01
##	mus_1[4,1]	-2.75	-2.16	2.12	1.80	-6.46	-0.69	1.06
##	mus_1[5,1]	-0.67	-0.38	1.17	0.11	-2.09	-0.23	1.04
##	mus_1[1,2]	1.35	-0.30	3.30	0.68	-0.90	8.69	1.05
##	mus_1[2,2]	0.99	-0.17	2.99	0.58	-1.08	7.77	1.00
##	mus_1[3,2]	1.46	-0.02	3.27	1.22	-0.94	8.63	1.01
##	mus_1[4,2]	-0.93	-0.80	0.92	0.30	-2.48	-0.09	1.04
##	mus_1[5,2]	1.70	0.35	3.21	0.93	-0.53	8.90	1.04
##	mus_1[1,3]	6.77	5.29	6.84	7.24	-0.46	19.85	1.03
##	mus_1[2,3]	6.48	4.96	6.72	7.18	-0.44	19.39	1.00
##	mus_1[3,3]	6.93	5.29	6.61	6.43	-0.62	19.58	1.01
##	mus_1[4,3]	2.12	-0.22	5.01	0.67	-0.73	13.69	1.02

## mus_1[5,3]	7.52	6.34	6.68	7.57	-0.14	20.04	1.03
## mus_2[1,1]	-1.61	-0.88	2.30	0.36	-2.48	-0.56	1.06
## mus_2[2,1]	-0.78	-0.78	2.57	0.18	-2.96	0.52	1.01
## mus_2[3,1]	-0.99	-0.89	2.09	0.18	-1.56	-0.10	1.01
## mus_2[4,1]	-0.98	-0.98	2.05	0.53	-2.14	0.33	1.03
## mus_2[5,1]	-0.70	-0.76	2.23	0.13	-2.04	-0.36	1.04
## mus_2[1,2]	-0.62	-0.53	6.43	2.42	-11.73	11.44	1.04
## mus_2[2,2]	-1.49	-1.07	5.94	2.47	-9.77	9.74	1.00
## mus_2[3,2]	-0.41	-0.89	6.41	1.65	-11.38	11.77	1.00
## mus_2[4,2]	-1.20	-1.04	1.92	0.46	-2.53	0.12	1.01
## mus_2[5,2]	-1.36	-2.00	6.52	2.67	-11.16	11.38	1.01
## mus_2[1,3]	-0.25	-0.51	8.42	5.51	-14.54	14.87	1.02
## mus_2[2,3]	-0.23	-0.69	8.49	6.12	-14.71	14.75	1.00
## mus_2[3,3]	-0.15	-0.71	8.37	5.53	-14.71	14.54	1.00
## mus_2[4,3]	-0.64	-0.75	5.51	0.61	-9.95	9.40	1.01
## mus_2[5,3]	-0.39	-0.69	8.89	6.94	-15.24	15.39	1.02
## Sigmas[1,1,1,1]	0.82	0.83	0.80	0.19	0.14	1.25	1.03
## Sigmas[2,1,1,1]	1.42	0.96	8.58	0.21	0.41	2.22	1.00
## Sigmas[3,1,1,1]	1.47	0.96	3.83	0.32	0.39	3.41	1.00
## Sigmas[4,1,1,1]	2.60	1.69	5.20	1.63	0.26	7.20	1.02
## Sigmas[5,1,1,1]	0.94	0.83	2.45	0.13	0.40	1.14	1.03
## Sigmas[1,2,1,1]	1.78	0.77	16.09	0.41	0.19	3.66	1.01
## Sigmas[2,2,1,1]	2.18	0.71	17.92	0.49	0.21	6.16	1.00
## Sigmas[3,2,1,1]	2.70	0.66	15.69	0.61	0.17	7.98	1.00
## Sigmas[4,2,1,1]	2.63	0.64	8.83	0.47	0.18	9.50	1.03
## Sigmas[5,2,1,1]	1.93	0.69	10.83	0.51	0.18	4.96	1.01
## Sigmas[1,3,1,1]	2.18	0.65	14.21	0.53	0.16	6.23	1.01
## Sigmas[2,3,1,1]	4.89	0.73	96.15	0.57	0.18	9.15	1.00
## Sigmas[3,3,1,1]	8.42	0.70	268.51	0.65	0.17	8.78	1.00
## Sigmas[4,3,1,1]	3.38	0.64	31.90	0.26	0.23	7.68	1.02
## Sigmas[5,3,1,1]	3.72	0.69	40.94	0.59	0.17	8.24	1.00
## Sigmas[1,1,2,1]	0.11	0.19	0.49	0.23	-0.36	0.41	1.03
## Sigmas[2,1,2,1]	0.14	0.28	2.35	0.18	-0.72	0.64	1.00
## Sigmas[3,1,2,1]	0.26	0.09	2.56	0.16	-0.25	0.97	1.00
## Sigmas[4,1,2,1]	-0.20	-0.02	2.20	0.51	-1.81	0.95	1.01
## Sigmas[5,1,2,1]	0.03	0.09	0.74	0.10	-0.46	0.26	1.03
## Sigmas[1,2,2,1]	0.09	0.09	4.37	0.29	-0.99	0.97	1.01
## Sigmas[2,2,2,1]	0.28	0.25	5.96	0.52	-1.90	2.20	1.00
## Sigmas[3,2,2,1]	0.05	0.05	5.51	0.31	-1.68	2.16	1.00
## Sigmas[4,2,2,1]	0.01	0.08	3.05	0.30	-1.34	0.94	1.01
## Sigmas[5,2,2,1]	-0.11	0.02	7.97	0.33	-1.55	1.64	1.01
## Sigmas[1,3,2,1]	-0.03	0.01	5.87	0.36	-1.84	1.85	1.00
## Sigmas[2,3,2,1]	0.34	0.09	21.53	0.48	-2.43	2.70	1.00
## Sigmas[3,3,2,1]	-3.31	0.00	86.96	0.42	-2.44	2.50	1.00
## Sigmas[4,3,2,1]	-0.30	0.24	10.77	0.22	-1.39	1.02	1.01
## Sigmas[5,3,2,1]	0.01	0.02	17.26	0.39	-2.40	2.29	1.00
## Sigmas[1,1,1,2]	0.11	0.19	0.49	0.23	-0.36	0.41	1.03
## Sigmas[2,1,1,2]	0.14	0.28	2.35	0.18	-0.72	0.64	1.00
## Sigmas[3,1,1,2]	0.26	0.09	2.56	0.16	-0.25	0.97	1.00
## Sigmas[4,1,1,2]	-0.20	-0.02	2.20	0.51	-1.81	0.95	1.01
## Sigmas[5,1,1,2]	0.03	0.09	0.74	0.10	-0.46	0.26	1.03
## Sigmas[1,2,1,2]	0.09	0.09	4.37	0.29	-0.99	0.97	1.01
## Sigmas[2,2,1,2]	0.28	0.25	5.96	0.52	-1.90	2.20	1.00
## Sigmas[3,2,1,2]	0.05	0.05	5.51	0.31	-1.68	2.16	1.00

##	Sigmas[4,2,1,2]	0.01	0.08	3.05	0.30	-1.34	0.94	1.01
##	Sigmas[5,2,1,2]	-0.11	0.02	7.97	0.33	-1.55	1.64	1.01
##	Sigmas[1,3,1,2]	-0.03	0.01	5.87	0.36	-1.84	1.85	1.00
##	Sigmas[2,3,1,2]	0.34	0.09	21.53	0.48	-2.43	2.70	1.00
##	Sigmas[3,3,1,2]	-3.31	0.00	86.96	0.42	-2.44	2.50	1.00
##	Sigmas[4,3,1,2]	-0.30	0.24	10.77	0.22	-1.39	1.02	1.01
##	Sigmas[5,3,1,2]	0.01	0.02	17.26	0.39	-2.40	2.29	1.00
##	Sigmas[1,1,2,2]	0.76	0.81	0.67	0.35	0.19	1.19	1.03
##	Sigmas[2,1,2,2]	1.59	1.28	2.94	0.57	0.33	3.36	1.00
##	Sigmas[3,1,2,2]	1.17	0.96	2.47	0.22	0.47	1.70	1.01
##	Sigmas[4,1,2,2]	1.02	0.71	2.87	0.41	0.22	2.07	1.01
##	Sigmas[5,1,2,2]	1.01	0.97	0.74	0.20	0.46	1.37	1.02
##	Sigmas[1,2,2,2]	1.26	0.52	5.17	0.31	0.16	3.64	1.02
##	Sigmas[2,2,2,2]	3.10	1.37	10.10	1.45	0.21	8.95	1.00
##	Sigmas[3,2,2,2]	2.01	0.83	10.51	0.58	0.19	5.43	1.00
##	Sigmas[4,2,2,2]	0.95	0.76	2.32	0.32	0.22	1.60	1.00
##	Sigmas[5,2,2,2]	2.13	0.74	10.80	0.59	0.19	5.90	1.00
##	Sigmas[1,3,2,2]	1.93	0.63	7.24	0.53	0.17	6.45	1.00
##	Sigmas[2,3,2,2]	3.98	0.98	22.85	0.95	0.18	9.68	1.00
##	Sigmas[3,3,2,2]	7.72	0.77	97.00	0.69	0.17	8.58	1.00
##	Sigmas[4,3,2,2]	2.80	0.83	14.33	0.38	0.22	4.48	1.01
##	Sigmas[5,3,2,2]	2.97	0.73	13.53	0.65	0.18	8.49	1.00
##	weights[1,1]	0.64	0.88	0.38	0.18	0.06	1.00	1.06
##	weights[2,1]	0.69	0.93	0.39	0.11	0.01	1.00	1.01
##	weights[3,1]	0.72	0.96	0.36	0.06	0.03	1.00	1.01
##	weights[4,1]	0.22	0.06	0.33	0.07	0.00	0.97	1.06
##	weights[5,1]	0.80	0.98	0.33	0.03	0.02	1.00	1.04
##	weights[1,2]	0.27	0.08	0.33	0.12	0.00	0.83	1.04
##	weights[2,2]	0.19	0.03	0.28	0.04	0.00	0.89	1.01
##	weights[3,2]	0.21	0.03	0.30	0.04	0.00	0.88	1.00
##	weights[4,2]	0.41	0.26	0.36	0.34	0.01	0.97	1.02
##	weights[5,2]	0.14	0.01	0.25	0.02	0.00	0.80	1.03
##	weights[1,3]	0.08	0.00	0.20	0.01	0.00	0.67	1.04
##	weights[2,3]	0.12	0.01	0.27	0.01	0.00	0.92	1.00
##	weights[3,3]	0.07	0.01	0.18	0.01	0.00	0.52	1.01
##	weights[4,3]	0.37	0.27	0.37	0.39	0.00	0.96	1.02
##	weights[5,3]	0.06	0.00	0.18	0.00	0.00	0.61	1.03
##	cumprod_one_minus_v[1,1]	0.36	0.12	0.38	0.18	0.00	0.94	1.06
##	cumprod_one_minus_v[2,1]	0.31	0.07	0.39	0.11	0.00	0.99	1.01
##	cumprod_one_minus_v[3,1]	0.28	0.04	0.36	0.06	0.00	0.97	1.01
##	cumprod_one_minus_v[4,1]	0.78	0.94	0.33	0.07	0.03	1.00	1.06
##	cumprod_one_minus_v[5,1]	0.20	0.02	0.33	0.03	0.00	0.98	1.04
##	cumprod_one_minus_v[1,2]	0.08	0.00	0.20	0.01	0.00	0.67	1.04
##	cumprod_one_minus_v[2,2]	0.12	0.01	0.27	0.01	0.00	0.92	1.00
##	cumprod_one_minus_v[3,2]	0.07	0.01	0.18	0.01	0.00	0.52	1.01
##	cumprod_one_minus_v[4,2]	0.37	0.27	0.37	0.39	0.00	0.96	1.02
##	cumprod_one_minus_v[5,2]	0.06	0.00	0.18	0.00	0.00	0.61	1.03
##	ess_bulk	ess_tail						
##	411	1814						
##	76	37						
##	286	510						
##	259	337						
##	68	267						
##	139	97						

##	435	3656
##	855	1089
##	1581	3827
##	469	636
##	847	2946
##	460	1022
##	841	1139
##	1292	2671
##	979	2017
##	622	728
##	23253	12722
##	26085	14460
##	23326	13601
##	23599	13336
##	20206	12445
##	85	44
##	295	338
##	321	352
##	58	222
##	152	110
##	113	121
##	343	640
##	331	524
##	137	267
##	171	123
##	156	116
##	404	807
##	409	510
##	381	2415
##	218	157
##	87	43
##	728	279
##	965	401
##	1603	776
##	419	133
##	3782	2801
##	6045	4948
##	4791	3930
##	876	796
##	3148	4685
##	15073	8453
##	15805	10685
##	12784	9510
##	4574	2251
##	14315	10732
##	578	590
##	3253	1025
##	1244	1109
##	244	2282
##	2217	511
##	5918	2861
##	6158	5775
##	2316	6727
##	148	394

##	5502	4204
##	2527	5780
##	5918	3753
##	2795	2619
##	995	466
##	5520	4101
##	206	1547
##	1111	521
##	1540	723
##	1003	5810
##	737	493
##	960	2584
##	3494	5650
##	5855	4193
##	1730	580
##	3486	3482
##	6047	4654
##	2572	3690
##	3562	2190
##	748	201
##	4957	3510
##	206	1547
##	1111	521
##	1540	723
##	1003	5810
##	737	493
##	960	2584
##	3494	5650
##	5855	4193
##	1730	580
##	3486	3482
##	6047	4654
##	2572	3690
##	3562	2190
##	748	201
##	4957	3510
##	238	877
##	1410	750
##	3601	932
##	1872	2228
##	2558	450
##	2946	2711
##	2656	6846
##	4362	3749
##	2222	1442
##	3626	3383
##	4217	4036
##	3153	3104
##	2630	1709
##	734	203
##	4126	2761
##	76	37
##	286	510
##	259	337

```
##      68      267
##     139      97
##     110     1253
##     486     642
##     327     844
##     215     963
##     210     649
##     137      79
##     390     581
##     375     436
##     296     897
##     201     104
##      76      37
##     286     510
##     259     337
##      68     267
##     139      97
##     137      79
##     390     581
##     375     436
##     296     897
##     201     104
```

Now we calculate confidence intervals for the proportion of HH infections for both fishing and inland communities:

```
# Extract rates (eta)
eta_draws <- as_draws_matrix(fit_1_2$draws("eta"))
eta_sums <- rowSums(eta_draws)

eta_prop <- eta_draws / eta_sums

# Check proportion of HH vs OOH for both fishing and inland
eta_HH_fishing <- eta_prop[, 1] / rowSums(eta_prop[, c(1, 4)])
eta_HH_inland <- eta_prop[, 2] / rowSums(eta_prop[, c(2, 5)])

print(quantile(eta_HH_fishing, c(0.025, 0.975))) # 95% CI
```

```
##      2.5%      97.5%
## 0.3409573 0.4538992
```

```
print(quantile(eta_HH_inland, c(0.025, 0.975))) # 95% CI
```

```
##      2.5%      97.5%
## 0.3565298 0.4799180
```

The confidence intervals have a good amount of overlap which suggests there might not be any change in proportion of infections which are within HH between the two communities.

1.3 Is the proportion of HH transmissions larger MF or FM?

We use the same model and draws from Q1.1:


```

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

eta_prop <- eta_draws / eta_sums

# Check proportion of HH vs OOH for both MF and FM
eta_HH_MF <- eta_prop[, 4] / rowSums(eta_prop[, c(3, 4)])
eta_HH_FM <- eta_prop[, 2] / rowSums(eta_prop[, c(1, 2)])

print(quantile(eta_HH_MF, c(0.025, 0.975))) # 95% CI

```

```

##      2.5%      97.5%
## 0.3183093 0.4202151

```

```

print(quantile(eta_HH_FM, c(0.025, 0.975))) # 95% CI

```

```

##      2.5%      97.5%
## 0.2879586 0.4062843

```

Again, there is a significant overlap it seems so potentially not a huge difference.

3

We use the same fit as above. We treat “young” as being aged 15-24.

```

#library(parallel)
#n_cores <- detectCores() - 1
source(here::here("helper-functions", "plot_normal.R"))

```

```

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

```

```

ages <- seq(15.5, 49.5, by = 1)
draws <- seq(100, 5000, by = 100)
grid <- expand.grid(ages, ages)

final_intensities <- list() # one entry per group
eta_draws <- as_draws_matrix(fit_norm$draws("eta"))

for (group in 1:4){
  # Calculate densities for 50 iterations
  intensities <- lapply(draws, plot_normal, fit = fit_norm, chain_no = 1,
                        group_no = group, ages = ages, plot = FALSE)
  etas <- as.numeric(eta_draws[draws, group])

  for (index in 1:length(etas)){
    intensities[[index]] <- etas[index] * intensities[[index]]
  }

  # Calculate mean density

```

```

intensity_matrix <- matrix(unlist(intensities), nrow = 50, byrow = TRUE)
final_intensity <- apply(intensity_matrix, 2, mean)
final_intensities[[group]] <- final_intensity
}

```

3.1 Are young men predominantly affected by older women through non-HH?

We want the F -> M groups for HH and non-HH respectively (so group 1 for OOH and group 2 for HH)

```

# Extract relevant intensities and combine into one data.table
intensity_FM_OOH <- final_intensities[[1]]
intensity_FM_HH <- final_intensities[[2]]

recip_15_24_indices <- which(grid$Var2 < 24) # X: Source, Y: Recipient

yng_grid <- grid[recip_15_24_indices, ]
yng_grid$FM_OOH <- intensity_FM_OOH[recip_15_24_indices]
yng_grid$FM_HH <- intensity_FM_HH[recip_15_24_indices]

setDT(yng_grid)

# Calculate intensities among young people by age of source
yng_grid[, FM_OOH_sum := sum(FM_OOH), by = Var1]
yng_grid[, FM_HH_sum := sum(FM_HH), by = Var1]

# Drop unnecessary columns
yng_grid[, c("FM_OOH", "FM_HH", "Var2") := NULL]

# Plot intensities

p <- ggplot(yng_grid, aes(x = Var1)) +
  geom_line(aes(y = FM_OOH_sum, color = "Out-of-household")) +
  geom_line(aes(y = FM_HH_sum, color = "Within household")) +
  labs(x = "Age of source (Female)", y = "Relative intensity",
       title = "FM transmission HH vs non-HH intensity",
       caption = "Among young (16-24) recipients",
       color = NULL)
p

```



Now repeat the analysis for M -> F transmissions

```
# Extract relevant intensities and add to previous data.table
intensity_MF_OOH <- final_intensities[[3]]
intensity_MF_HH <- final_intensities[[4]]

yng_grid$MF_OOH <- intensity_MF_OOH[recip_15_24_indices]
yng_grid$MF_HH <- intensity_MF_HH[recip_15_24_indices]

# Calculate intensities among young people by age of source
yng_grid[, MF_OOH_sum := sum(MF_OOH), by = Var1]
yng_grid[, MF_HH_sum := sum(MF_HH), by = Var1]

# Drop unnecessary columns
#yng_grid[, c("MF_OOH", "MF_HH", "Var2") := NULL]

# Plot intensities

p <- ggplot(yng_grid, aes(x = Var1)) +
  geom_line(aes(y = MF_OOH_sum, color = "Out-of-household")) +
  geom_line(aes(y = MF_HH_sum, color = "Within household")) +
  labs(x = "Age of source (Male)", y = "Relative intensity",
       title = "MF transmission HH vs non-HH intensity",
       caption = "Among young (16–24) recipients",
       color = NULL)
```

p

