Some prior predictive checks

```
library(tidyverse)
source("../../init.R")
Navigating to the US directory to load use the US data loading wrappers.
find_and_set_directory("US/exploration")
source("load_and_preprocess.R")
##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##
      smiths
## Attaching package: 'jsonlite'
## The following object is masked from 'package:purrr':
##
##
      flatten
## Rows: 19311 Columns: 139
## -- Column specification -------
## Delimiter: ","
## chr (14): TRACT, NAME, ST, STATE_FIPS, CNTY_FIPS, STCOFIPS, STATE_ABBR, STA...
## dbl (125): total_ppl_acs20E, median_annual_incomeE, house_price_medianE, ren...
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
## Rows: 20044 Columns: 124
## -- Column specification -
## Delimiter: ","
## chr
       (2): STATE_ABBR, zip
## dbl (122): total_ppl_acs20E, median_annual_incomeE, house_price_medianE, bui...
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

find_and_set_directory("US/exploration")

source("compute_descriptives.R")

```
##
## Attaching package: 'kableExtra'
##
## The following object is masked from 'package:dplyr':
##
## group_rows
```

In particular, I want to focus on two tract states, Ohio and Massachusetts, as well as two ZIP states, New Jersey and Illinois.

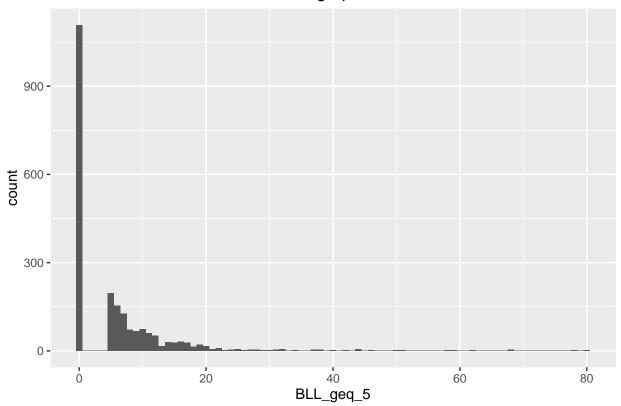
Ohio

```
data_summary("OH", year = 2010)
## [1] "Loading OH from processed_data"
## Rows: 94292 Columns: 6
## Delimiter: ","
## chr (4): state, BLL_geq_5, BLL_geq_10, tested
## dbl (2): tract, year
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
## Warning: Using 'across()' in 'filter()' was deprecated in dplyr 1.0.8.
## i Please use 'if_any()' or 'if_all()' instead.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
## [1] "Additional features added: "
## Warning in datasummary_skim_numeric(data, output = output, fmt = fmt, histogram
## = histogram, : The histogram argument is only supported for (a) output types
## "default", "html", or "kableExtra"; (b) writing to file paths with extensions
## ".html", ".jpg", or ".png"; and (c) Rmarkdown or knitr documents compiled to
## PDF or HTML. Use 'histogram=FALSE' to silence this warning.
## [[1]]
## # A tibble: 1 x 18
##
     year n_obs lead_cens_5 lead_nocens_5 lead_censR_5 lead_cens_10 lead_nocens_10
##
    <dbl> <int>
                      <int>
                                   <int>
                                                <dbl>
                                                            <int>
                                                                           <int>
## 1 2010 4722
                       2534
                                    2188
                                                0.537
                                                             1350
                                                                            3372
## # i 11 more variables: lead_censR_10 <dbl>, test_cens <int>, test_censR <dbl>,
      sup_threshold_5 <dbl>, sup_threshold_10 <dbl>, median_lead_5 <dbl>,
      q75_lead_5 <dbl>, max_lead_5 <dbl>, median_lead_10 <dbl>,
      q75_lead_10 <dbl>, max_lead_10 <dbl>
##
## [[2]]
##
```

##											
##	1	Unique (#)	Missing	(%)	Mean		SD	Min	Median	Max	1
##	:		:	:		-: -	:	:	I:	:	
##	BLL_geq_5	44	1	0	5.3	3	6.1	0.0	5.0	80.0	
##	BLL_geq_10	8	1	0	1.5	5	2.3	0.0	0.0	14.0	1
##	tested	163	1	0	42.3	3	32.4	5.0	35.0	372.0	1
##	year	1	1	0	2010.0)	0.0	2010.0	2010.0	2010.0	1
##	tested_ell	1	1	0	5.0)	0.0	5.0	5.0	5.0	1
##	ell_5	1	1	0	4.0)	0.0	4.0	1 4.0	4.0	1
##	ell_10	1	1	0	4.0)	0.0	4.0	1 4.0	4.0	
##	median_annual_incomeE	2271	1	0	0.0)	1.0	-2.5	-0.0	5.5	1
##	house_price_medianE	l 1464	1	0	0.0)	1.0	-1.6	-0.1	6.2	1
##	poor_fam_propE	2308	1	0	-0.0)	1.0	-1.1	-0.4	5.6	1
##	black_ppl_propE	2157	1	0	-0.1	.	0.9	-0.6	l -0.5	3.5	1
##	bp_pre_1959E_prop	2358	1	0	-0.0)	1.0	-1.8	-0.1	2.0	1
##	svi_socioeconomic_pctile	2104	1	0	-0.0)	1.0	-2.2	-0.0	1.6	1
##	under_yo5_pplE	529	1	0	229.5	5	137.7	8.0	204.0	1263.0	
##	ped_per_100k	l 69	1	0	87.0)	67.0	0.0	68.1	243.7	1
##											

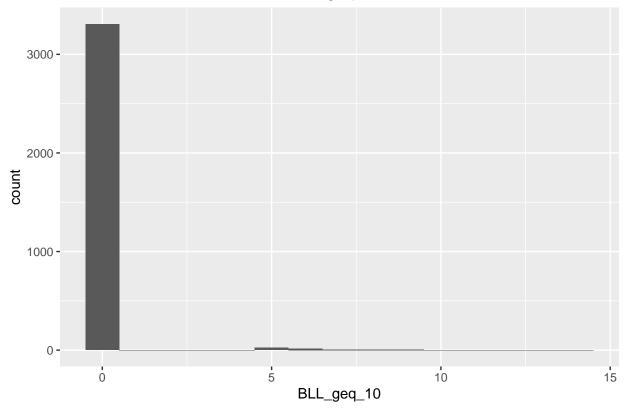
[[3]] ## [[3]][[1]]

Distribution of non-censored BLL_geq_5 values

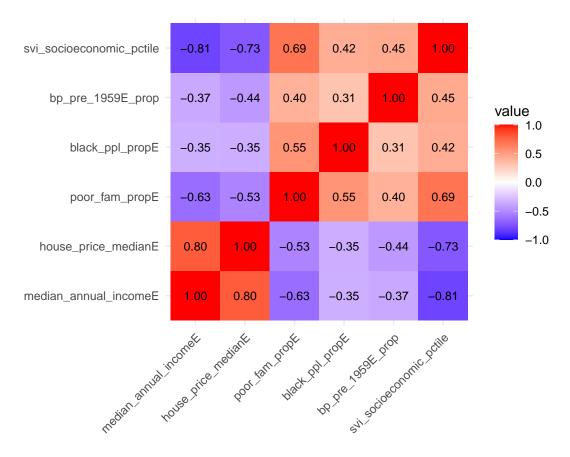


[[3]][[2]]

Distribution of non-censored BLL_geq_10 values



[[4]]



I take one additional preprocessing steps on the predictors passed to the logit side (income, building period, and SVI), I also standardise the pediatrician rate.

```
# standardise pediatrician rate
oh_merged <- oh_merged |>
    mutate(ped_per_100k = (ped_per_100k - mean(ped_per_100k)) / sd(ped_per_100k))

# load STAN model
library(cmdstanr)

## This is cmdstanr version 0.6.1

## - CmdStanR documentation and vignettes: mc-stan.org/cmdstanr

## - CmdStan path: /Users/lasse/.cmdstan/cmdstan-2.33.0

## - CmdStan version: 2.33.0

## A newer version of CmdStan is available. See ?install_cmdstan() to install it.

## To disable this check set option or environment variable CMDSTANR_NO_VER_CHECK=TRUE.
```

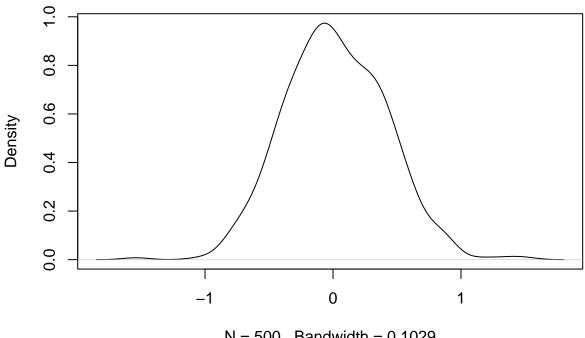
```
find_and_set_directory("models/prior_predictive_checks")
ppc_stan <- cmdstan_model("poisson_thinned_exclusion_ppc.stan")</pre>
## Warning in readLines(stan_file): incomplete final line found on
## 'poisson_thinned_exclusion_ppc.stan'
# create function to prep data for given STAN model
stan_data <- function(state_final){</pre>
  list(
   N_obs = state_final |> filter(!BLL_geq_5_suppressed) |> count() |> pull(n),
   N_cens = state_final |> filter(BLL_geq_5_suppressed) |> count() |> pull(n),
   y_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(BLL_geq_5),
   median_income_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(median_annual_incomeE),
   house_price_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(house_price_medianE),
   poverty_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(poor_fam_propE),
   black_prop_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(black_ppl_propE),
   building_period_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(bp_pre_1959E_prop),
   svi obs = state final |> filter(!BLL geq 5 suppressed) |> pull(svi socioeconomic pctile),
    # censored
   median_income_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(median_annual_incomeE),
   house price cens = state final |> filter(BLL geq 5 suppressed) |> pull(house price medianE),
   poverty_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(poor_fam_propE),
   black_prop_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(black_ppl_propE),
   building_period_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(bp_pre_1959E_prop),
    svi_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(svi_socioeconomic_pctile),
    # pediatricians & kids
   z_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(ped_per_100k),
   z_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(ped_per_100k),
   kids_obs = state_final |> filter(!BLL_geq_5_suppressed) |> pull(under_yo5_pplE),
   kids_cens = state_final |> filter(BLL_geq_5_suppressed) |> pull(under_yo5_pplE)
}
```

Setting the priors as follows:

In the logit: - intercept of N(0, 1.2) for "uniform" log-odds - slopes of N(0, 0.5) for all 4 predictors (income, building period, SVI, and pediatrician rate) In the poisson: - intercept of N(-1.58.2) based on the NHANES national average of 2.6% - slopes of N(0,0.5) for all intercepts - for higher variances some observations go beyond the rate tolerance - except income and the building period, which I center -0.1 and 0.1 respectively.

```
# plot N(0, 0.6)
plot(density(rnorm(500,0,0.4)))
```

density.default(x = rnorm(500, 0, 0.4))



N = 500 Bandwidth = 0.1029

```
# sample from prior predictive distribution
oh_stan <- stan_data(oh_merged)</pre>
ppc_samples <- ppc_stan$sample(</pre>
  data = oh_stan,
  iter_sampling = 500,
  refresh = NULL,
  fixed_param = TRUE) # required for RNG?
```

```
## Running MCMC with 4 sequential chains...
##
## Chain 1 Iteration:
                                         (Sampling)
                        1 / 500 [ 0%]
## Chain 1 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
## Chain 1 Iteration: 200 / 500 [ 40%]
                                         (Sampling)
## Chain 1 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 1 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 1 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 1 finished in 1.8 seconds.
## Chain 2 Iteration:
                        1 / 500 [ 0%]
                                         (Sampling)
## Chain 2 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
## Chain 2 Iteration: 200 / 500 [ 40%]
                                         (Sampling)
## Chain 2 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 2 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 2 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 2 finished in 1.8 seconds.
## Chain 3 Iteration:
                        1 / 500 [ 0%]
                                         (Sampling)
```

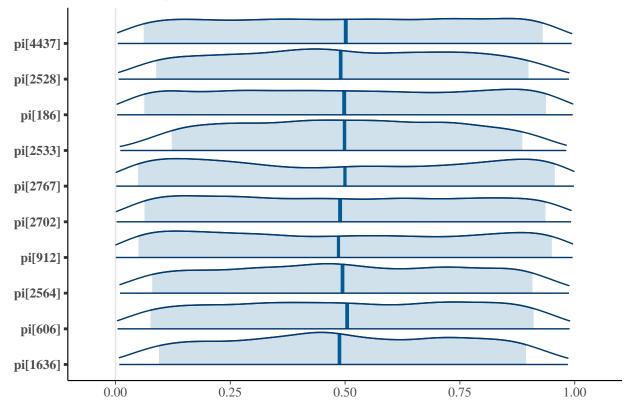
```
## Chain 3 Iteration: 100 / 500 [ 20%]
                                        (Sampling)
## Chain 3 Iteration: 200 / 500 [ 40%]
                                        (Sampling)
## Chain 3 Iteration: 300 / 500 [ 60%]
                                        (Sampling)
## Chain 3 Iteration: 400 / 500 [ 80%]
                                        (Sampling)
## Chain 3 Iteration: 500 / 500 [100%]
                                        (Sampling)
## Chain 3 finished in 1.8 seconds.
## Chain 4 Iteration:
                       1 / 500 [ 0%]
                                        (Sampling)
## Chain 4 Iteration: 100 / 500 [ 20%]
                                        (Sampling)
## Chain 4 Iteration: 200 / 500 [ 40%]
                                        (Sampling)
## Chain 4 Iteration: 300 / 500 [ 60%]
                                        (Sampling)
## Chain 4 Iteration: 400 / 500 [ 80%]
                                        (Sampling)
## Chain 4 Iteration: 500 / 500 [100%]
                                        (Sampling)
## Chain 4 finished in 1.8 seconds.
##
## All 4 chains finished successfully.
## Mean chain execution time: 1.8 seconds.
## Total execution time: 7.7 seconds.
ppc_samples$summary(variables = c("y_thinned"))
## # A tibble: 4,722 x 10
                                               q5
##
      variable
                    mean median
                                                    q95 rhat ess_bulk ess_tail
                                   sd
                                        mad
##
      <chr>
                   <num>
                                                                          <num>
## 1 y_thinned[1] 1.80
                              0 25.0
                                          0
                                                0
                                                      4 1.00
                                                                 1907.
                                                                          1933.
## 2 y_thinned[2]
                    1.23
                              0 12.3
                                          0
                                                0
                                                      4 1.00
                                                                 1827.
                                                                          1816.
## 3 y_thinned[3]
                    1.64
                              0 17.4
                                          0
                                                0
                                                      4 1.00
                                                                 2009.
                                                                          1763.
## 4 y_thinned[4]
                    1.01
                                                      4 1.00
                                                                 1953.
                              0 8.13
                                          0
                                                0
                                                                          1898.
## 5 y_thinned[5]
                    1.00
                              0 8.30
                                          0
                                                0
                                                      4 1.00
                                                                 2156.
                                                                          2018.
## 6 y_thinned[6]
                    1.99
                              0 20.7
                                          0
                                                0
                                                      6 1.00
                                                                 1994.
                                                                          1821.
## 7 y_thinned[7]
                    1.04
                              0 8.88
                                          0
                                                      4 1.00
                                                                 2044.
                                                0
                                                                          1989.
## 8 y_thinned[8]
                    1.11
                              0 10.9
                                          0
                                                0
                                                      4 1.00
                                                                 1700.
                                                                          1758.
## 9 y_thinned[9]
                    1.02
                              0 8.95
                                          0
                                                0
                                                      4 1.00
                                                                 2113.
                                                                          1972.
## 10 y_thinned[10] 1.29
                              0 7.16
                                          0
                                                0
                                                      5 1.00
                                                                 2032.
                                                                          2000.
## # i 4,712 more rows
# add y_thinned from $summary to oh_merged
oh_merged_ppc <- oh_merged |>
  bind cols(ppc samples$summary(variables = c("y thinned")) |>
              filter(str_detect(variable, "y_thinned")) |>
              select(c(mean, median, q5, q95)))
Plot the thinning rates
library(bayesplot)
## This is bayesplot version 1.10.0
## - Online documentation and vignettes at mc-stan.org/bayesplot
## - bayesplot theme set to bayesplot::theme_default()
      * Does _not_ affect other ggplot2 plots
##
```

* See ?bayesplot_theme_set for details on theme setting

```
ppc_samples$draws(format = "draws_df") |>
    # select columns that contain "pi"
    select(starts_with("pi")) |>
    # randomly select 100 columns of those (for speed)
    select(sample(1:nrow(oh_merged_ppc), 10)) |>
    mcmc_areas(prob = 0.9) +
    ggtitle("Prior thinning rates for Ohio")
```

Warning: Dropping 'draws_df' class as required metadata was removed.

Prior thinning rates for Ohio



```
library(bayesplot)

# extract and plot draws

ppc_samples$draws(format = "draws_df", variables = c("y_thinned")) |>
    # select columns that contain "y_thinned"

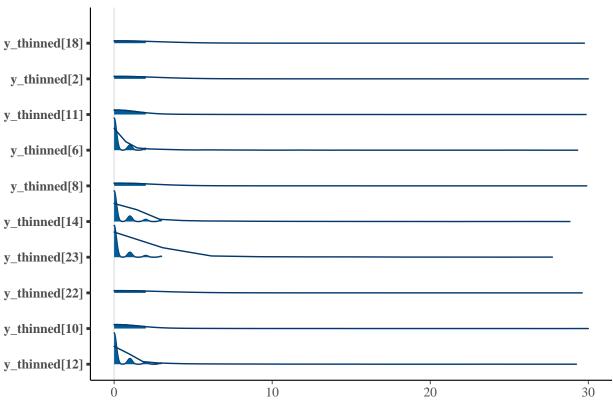
    select(starts_with("y_thinned")) |>
    # randomly select 100 columns of those (for speed)

    select(sample(1:length(oh_merged), 10)) |>
    mcmc_areas(prob = 0.8, alpha = 0.1) +
    ggtitle("Prior predictive checks for Ohio") +
    xlim(0,30)
```

Warning: The following arguments were unrecognized and ignored: alpha

```
## Warning: Dropping 'draws_df' class as required metadata was removed.
## Scale for x is already present.
## Adding another scale for x, which will replace the existing scale.
## Warning: Removed 10 rows containing missing values ('geom_segment()').
```

Prior predictive checks for Ohio



Massachusetts

```
data_summary("MA", year = 2010)

## [1] "Loading MA from processed_data"

## Rows: 16192 Columns: 6

## -- Column specification ------

## Delimiter: ","

## chr (3): state, town, BLL_geq_5

## dbl (3): year, tested, tract

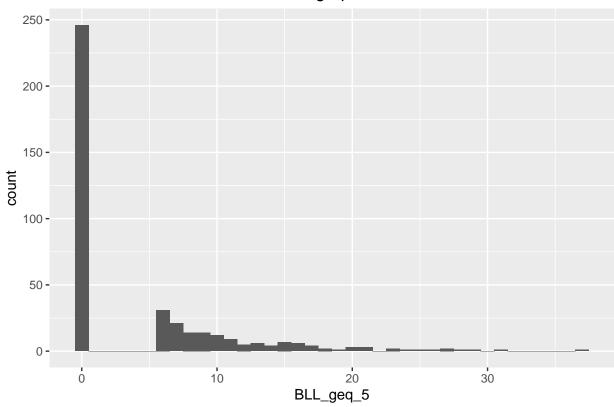
##

## i Use 'spec()' to retrieve the full column specification for this data.

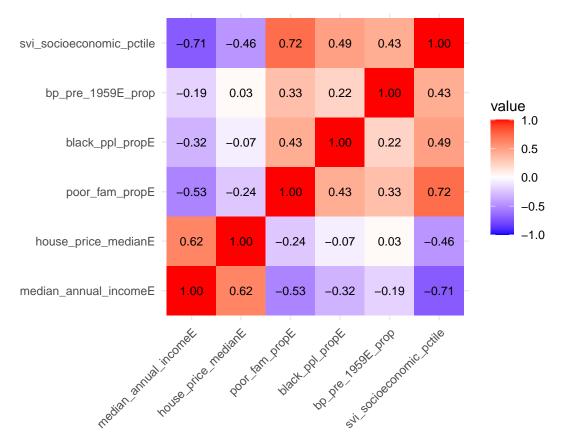
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
## Warning: Using 'across()' in 'filter()' was deprecated in dplyr 1.0.8.
## i Please use 'if_any()' or 'if_all()' instead.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
## [1] "Additional features added: "
## Warning in datasummary_skim_numeric(data, output = output, fmt = fmt, histogram
## = histogram, : The histogram argument is only supported for (a) output types
## "default", "html", or "kableExtra"; (b) writing to file paths with extensions
## ".html", ".jpg", or ".png"; and (c) Rmarkdown or knitr documents compiled to
## PDF or HTML. Use 'histogram=FALSE' to silence this warning.
## [[1]]
## # A tibble: 1 x 11
     year n_obs lead_cens_5 lead_nocens_5 lead_censR_5 test_cens test_censR
    <dbl> <int> <int> <int> <int> <int>
## 1 2010
         966
                     567
                                 399
                                          0.587
## # i 4 more variables: sup threshold 5 <dbl>, median lead 5 <dbl>,
## # q75_lead_5 <dbl>, max_lead_5 <dbl>
## [[2]]
##
##
                        | Unique (#) | Missing (%) | Mean |
                                                             SD | Min | Median | Max |
## |:----:|----:|----:|----:|----:|
## | year
                                   1 |
                                                0 | 2010.0 | 0.0 | 2010.0 | 2010.0 | 2010.0 |
                                  27 |
                                                                     0.0 |
                                                                             5.0 |
## | BLL_geq_5
                                                0 |
                                                      4.7 |
                                                             4.3 |
## | tested
                                  276
                                                0 | 141.4 | 70.4 |
                                                                     3.0 | 133.0 | 411.0 |
## | tested_ell
                                  1 |
                                                0 |
                                                     1.0
                                                             0.0
                                                                     1.0 |
                                                                            1.0
                                                                                     1.0
## | ell_5
                                   1 |
                                               0 |
                                                      4.0
                                                             0.0
                                                                    4.0 |
                                                                             4.0
                                                                                     4.0 |
## | median_annual_incomeE
                                  953 |
                                               0 |
                                                    0.1 |
                                                             1.0 |
                                                                   -2.3 |
                                                                           -0.0
                                                                                     5.0 l
                                  889 |
                                               0 |
                                                     0.1 |
                                                             1.0 |
                                                                   -1.5 |
                                                                           -0.2 |
## | house_price_medianE
                                                                                     6.6
                                               0 |
                                                                           -0.4 l
## | poor fam propE
                                  958 |
                                                    -0.1 |
                                                             0.9 |
                                                                    -0.9 l
                                                                                     5.7
## | black_ppl_propE
                                  908 I
                                               0 | -0.0 |
                                                             1.0 | -0.6 |
                                                                           -0.4 |
                                                                                     5.8 l
## | bp_pre_1959E_prop
                                  964 |
                                               0 | -0.1 |
                                                             1.0 | -2.3 | -0.1 |
                                                                                     2.0 |
## | svi_socioeconomic_pctile |
                               913 |
                                               0 | -0.1 |
                                                             0.9 | -1.3 | -0.4 |
                                                                                     2.1 l
                                               0 | 263.0 | 134.5 |
                                                                   14.0 | 243.5 | 1057.0 |
419 |
                                              0 | 139.7 | 93.0 | 0.0 | 101.1 | 386.2 |
## | ped_per_100k
                                 14 |
##
## [[3]]
## [[3]][[1]]
```

Distribution of non-censored BLL_geq_5 values



[[4]]



```
# standardise pediatrician rate
ma_merged <- ma_merged |>
  mutate(ped_per_100k = (ped_per_100k - mean(ped_per_100k)) / sd(ped_per_100k))
# sample from prior predictive distribution
ma stan <- stan data(ma merged)</pre>
ppc_samples <- ppc_stan$sample(</pre>
  data = ma_stan,
  iter_sampling = 500,
  refresh = NULL,
  fixed_param = TRUE)
## Running MCMC with 4 sequential chains...
##
## Chain 1 Iteration:
                         1 / 500 [ 0%]
                                         (Sampling)
## Chain 1 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
## Chain 1 Iteration: 200 / 500 [ 40%]
                                         (Sampling)
## Chain 1 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 1 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 1 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 1 finished in 0.4 seconds.
## Chain 2 Iteration:
                       1 / 500 [ 0%]
                                         (Sampling)
## Chain 2 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
```

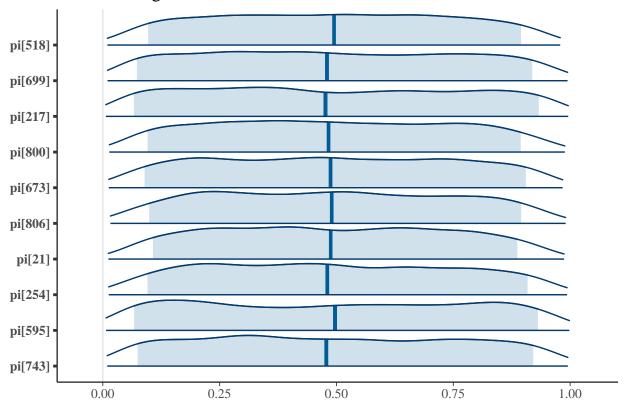
(Sampling)

Chain 2 Iteration: 200 / 500 [40%]

```
## Chain 2 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 2 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 2 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 2 finished in 0.4 seconds.
## Chain 3 Iteration:
                      1 / 500 [ 0%]
                                         (Sampling)
## Chain 3 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
## Chain 3 Iteration: 200 / 500 [ 40%]
                                         (Sampling)
## Chain 3 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 3 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 3 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 3 finished in 0.4 seconds.
## Chain 4 Iteration: 1 / 500 [ 0%]
                                         (Sampling)
## Chain 4 Iteration: 100 / 500 [ 20%]
                                         (Sampling)
## Chain 4 Iteration: 200 / 500 [ 40%]
                                         (Sampling)
## Chain 4 Iteration: 300 / 500 [ 60%]
                                         (Sampling)
## Chain 4 Iteration: 400 / 500 [ 80%]
                                         (Sampling)
## Chain 4 Iteration: 500 / 500 [100%]
                                         (Sampling)
## Chain 4 finished in 0.4 seconds.
## All 4 chains finished successfully.
## Mean chain execution time: 0.4 seconds.
## Total execution time: 1.9 seconds.
# add y_thinned from $summary to ma_merged
ma_merged_ppc <- ma_merged |>
  bind_cols(ppc_samples$summary(variables = c("y_thinned")) |>
              filter(str_detect(variable, "y_thinned")) |>
              select(c(mean, median, q5, q95)))
# extract prior thinning rates
ppc_samples$draws(format = "draws_df") |>
  # select columns that contain "pi"
  select(starts_with("pi")) |>
  # randomly select 100 columns of those (for speed)
  select(sample(1:nrow(ma_merged), 10)) |>
  mcmc_areas(prob = 0.9) +
  ggtitle("Prior thinning rates for Massachusetts")
```

Warning: Dropping 'draws_df' class as required metadata was removed.

Prior thinning rates for Massachusetts

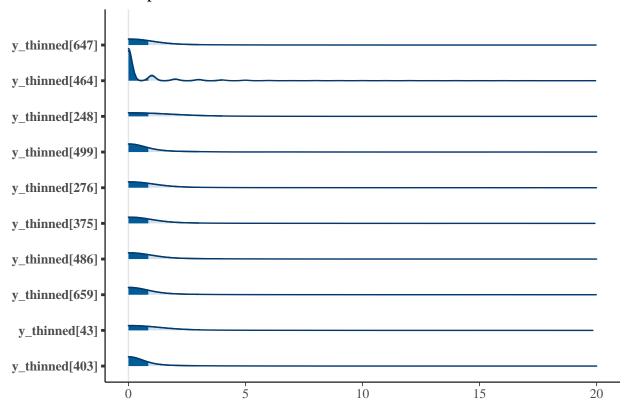


```
# extract and plot draws
ppc_samples$draws(format = "draws_df", variables = c("y_thinned")) |>
    # select columns that contain "y_thinned"
    select(starts_with("y_thinned")) |>
    # randomly select 100 columns of those (for speed)
    select(sample(1:nrow(ma_merged_ppc), 10)) |>
    mcmc_areas(prob = 0.9, alpha = 0.1) +
    ggtitle("Prior predictive checks for Massachusetts") +
    xlim(0,20)
```

```
## Warning: The following arguments were unrecognized and ignored: alpha
```

- ## Warning: Dropping 'draws_df' class as required metadata was removed.
- ## Scale for x is already present.
- ## Adding another scale for x, which will replace the existing scale.
- ## Warning: Removed 10 rows containing missing values ('geom_segment()').

Prior predictive checks for Massachusetts



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