

Bayesian prevalence analysis of autism prevalence in Chile

Adele Tyson

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```
#source("Chile_prev.R", local = knitr::knit_global())

library(nleqslv) # Only needed for robince bayesian prevalence
library(janitor)
library(gridExtra)
library(readxl)
library(viridis)
library(wesanderson)
library(psych)
library(Hmisc)
library(poolr)
library(epitools)
library(corrplot)
library(caret)
library(mltools)
library(ggrepel)
library(rjags)
library(rstan)
library(posterior)
library(tidybayes)
library(bayesplot)
library(tidyverse)
```

Set up

Load data

```
chile_merged_raw <- read.csv("04_Data/Data_Chile_Merge.csv") %>% clean_names()

chile_merged <- chile_merged_raw %>%
  rename(sex_desc = sex,
         year = agno,
         school_code = rbd,
         school_check_code = dgv_rbd,
         school_name = nom_rbd,
         school_region_code = cod_reg_rbd,
         school_region_name_abr = nom_reg_rbd_a,
         school_province_code = cod_pro_rbd,
         school_commune_code = cod_com_rbd,
         school_commune_name = nom_com_rbd,
         school_dept_code = cod_deprov_rbd,
```



```
select(school_commune_name, health_service_name)
```

Try Bayesian analysis of autism prevalence and specificity and sensitivity of school assessment “Bayesian Estimation of Disease Prevalence and the Parameters of Diagnostic Tests in the Absence of a Gold Standard” Lawrence Joseph, Theresa W. Gyorkos, Louis Coupal <https://www.cambridge.org/core/journals/epidemiology-and-psychiatric-sciences/article/bayesian-approach-to-estimating-the-population-prevalence-of-mood-and-anxiety-disorders-using-multiple-measures/DB1D2CA6C27C7E8C85C60B62B969BB72>

Use sensitivity and specificity of Social Attention and Communication Surveillance–Revised (SACS-R) tool “Diagnostic Accuracy of the Social Attention and Communication Surveillance–Revised With Preschool Tool for Early Autism Detection in Very Young Children” Josephine Barbaro, Nancy Sadka, Melissa Gilbert, et al <https://jamanetwork.com/journals/jamanetworkopen/fullarticle/2789926>

```
chile_bayes_aut <- chile_merged %>%
  filter(age_june30 >= 6 & age_june30 <= 18,
         #special_needs_status == 1,
         sex != 0) %>%
  mutate(autism = ifelse(special_needs_code == 105, 1, 0),
         age_cat = ifelse(age_june30 <= 8, 1, ifelse(age_june30 <= 11, 2, ifelse(age_june30 <= 14, 3, 4)),
         age_cat_name = ifelse(age_cat == 1, "6-8", ifelse(age_cat == 2, "9-11", ifelse(age_cat == 3, "12-14", "15-18")),
         # 1 = 6-8, 2 = 9-11, 3 = 12-14, 4 = 15-18
         age_cat_name = factor(age_cat_name, levels = c("6-8", "9-11", "12-14", "15-18")),
         ethnic_2_group = ifelse(ethnic_3_group == "Aymara", "Other ethnic group", ethnic_3_group),
         school_fee_temp = school_fee,
         school_fee = ifelse(school_fee == "", "No information",
                             ifelse(school_fee == "GRATUITO", "Free",
                                     ifelse(school_fee == "$1.000 A $10.000", "$1,000-$10,000",
                                             ifelse(school_fee == "$10.001 A $25.000", "$10,001-$25,000",
                                                     ifelse(school_fee == "$25.001 A $50.000", "$25,001-$50,000",
                                                         ifelse(school_fee == "$50.001 A $100.000", "$50,001-$100,000",
                                                             ifelse(school_fee == "MAS DE $100.000", "More than $100,000", "SIN INFORMACION")))))))
         school_fee = factor(school_fee, levels = c("Free", "$1,000-$10,000", "$10,001-$25,000", "$25,001-$50,000", "$50,001-$100,000", "More than $100,000", "SIN INFORMACION")),
         school_fee_group = ifelse(school_fee == "Free", "Free",
                                   ifelse(school_fee %in% c("$1,000-$10,000", "$10,001-$25,000", "$25,001-$50,000", "$50,001-$100,000", "More than $100,000", "SIN INFORMACION"),
                                           ifelse(school_fee == "$100,001+", "High",
                                                  ifelse(school_fee == "No information", "No information", "Other"))),
                                   "Other"),
         school_fee_group = factor(school_fee_group, levels = c("Free", "Low", "High", "No information")),
  left_join(chile_communes, by = "school_commune_name") %>%

select(school_region_name_abr,
       sex,
       sex_desc,
       age_june30,
       #edad_alu_2, # equal to age_june30
       age_cat,
       age_cat_name,
       school_rurality_code,
       #rural_rbd_2, # not quite equal to school_rurality_code as it has NA's
       pago_matricula,
       pago_mensual,
       school_fee,
       school_fee_group,
       school_fee_temp,
       ethnicity,
```

```

    mapuche,
    nationality,
    ethnic_3_group,
    ethnic_2_group,
    #asd_chile, # equal to autism
    autism,
    school_commune_name,
    health_service_name
  )

# Prevalence of autism in Chile dataset
sum(chile_bayes_aut$autism) / nrow(chile_bayes_aut) # 0.00476 = 0.476%, very low

## [1] 0.004760322

# Is prevalence the same across geographic regions, age, sex?
n_std_pop <- sum(chile_stdpop$std_pop)

```

Define some functions to keep code clean

```

get_grouped_prev <- function(x, stdpop, grouping_vars) {
  # Calculates sample prevalence, age- and sex-standardised prevalence and group weighting for supplied
  # x = chile_bayes_aut, needs columns called autism, count
  # stdpop = standard population with age and sex counts
  # grouping_vars = variables in x to group by

  n_stdpop <- sum(stdpop$std_pop)

  x_grouped <- x %>%
    group_by(across(all_of(grouping_vars))) %>%
    summarise(count = n()) %>%
    pivot_wider(names_from = autism, values_from = count) %>%
    rename("n_noautism" = "0", "n_autism" = "1", "age" = "age_june30") %>%
    mutate(n_autism = ifelse(is.na(n_autism), 0, n_autism), # If there are no cases of autism in the gr
           sample_pop_size = n_noautism + n_autism, # Total sample population is autism cases + not cas
           sample_prevalence = n_autism / sample_pop_size) %>% # Prevalence of autism in the group
    left_join(stdpop, by = c("age", "sex")) %>%
    mutate(aut_prev_std = n_autism / sample_pop_size * pop_prop, # Prevalence of autism in the group, s
           w = std_pop / (sample_pop_size * n_stdpop), # Weight of the group using standard population
           w2 = pop_prop / sample_pop_size,
           #sum_std_pop = sum(std_pop)
           ) %>%
    ungroup()
  return(x_grouped)
}

get_adjusted_prev <- function(x, grouping_vars) {
  # Turns grouped prevalences into age- and sex- adjusted prevalences with Fay and Feuer Gamma confiden
  # x = output from get_grouped_prev
  x_adj <- x %>%
    group_by(across(all_of(grouping_vars))) %>%
    summarise(sum_sample_pop_size = sum(sample_pop_size),
              crude_rate = sum(n_autism) / sum(sample_pop_size),
              crude_count = sum(n_autism),

```

```

    adjusted_rate = sum(n_autism / sample_pop_size * pop_prop),
    adjusted_count = round(adjusted_rate * sum_sample_pop_size, 0), # had to fudge this to get 1
    #adjusted_count = adjusted_rate * sum_sample_pop_size,
    var = sum(pop_prop^2 * n_autism / sample_pop_size^2),
    #se2 = sqrt(sum((std_pop/sum(std_pop))^2 * n_autism/sample_pop_size^2)),
    w_M = max(w),
    ci_lower = ifelse(var == 0, 0, var / (2*adjusted_rate) * qchisq(p = 0.05/2, df = 2*adjusted_rate)),
    ci_upper = (var + w_M^2) / (2*(adjusted_rate + w_M)) * qchisq(p = 1-0.05/2, df = 2*(adjusted_rate)),
    arrange(across(all_of(grouping_vars)))
  }
}

do_jags_rand_model <- function(x, feat, model, theta_mu, theta_sigma, pars, nBurn = 1000, nIter = 1000,
  # x = output from get_adjusted_prev. Needs to have columns sum_sample_pop_size, adjusted_count
  # feat = feature being used as random effect
  # model = JAGS random effects model
  # theta_mu, theta_sigma = mean and sd of beta prior distribution
  # pars = model parameters to report
  # nBurn = number of burn-in samples
  # nIter = number of posterior iterations

  nFeat <- length(unique(x[[feat]]))
  FeatNames <- sort(unique(x[[feat]]))

  # Define beta prior
  theta_a <- theta_mu * (theta_mu * (1-theta_mu) / theta_sigma^2 - 1)
  theta_b <- (1 - theta_mu) * (theta_mu * (1-theta_mu) / theta_sigma^2 - 1)

  # Initial values for model chains
  rand_ini <- list(list(theta = rep(0.001, nFeat)), #, spec = 0.5, sens = 0.5),
    list(theta = rep(0.01, nFeat))) #, spec = 0.9, sens = 0.9))

  # Run JAGS model
  rand_data <- list(theta_a = theta_a,
    theta_b = theta_b,
    nObs = x$sum_sample_pop_size,
    aut_sample = x$adjusted_count,
    nFeat = nFeat)
  rand_jag <- jags.model(textConnection(model),
    data = rand_data,
    inits = rand_ini,
    n.chains = 2,
    quiet = TRUE)
  update(rand_jag, n.iter = nBurn)
  rand_sam <- coda.samples(model = rand_jag,
    variable.names = pars,
    n.iter = nIter)

  # Convergence checks
  if(convergence_checks) {
    print(mcmc_trace(rand_sam, paste0("theta[", 1:nFeat, "]"))) # Convergence looks fine and rhats <= 1
    print(mcmc_trace(rand_sam, paste0("aut_pred[", 1:nFeat, "]"))) # Convergence looks fine and rhats <= 1
    rand_summ <- summary(subset_draws(as_draws(rand_sam), pars),
      ~quantile(.x, probs=c(0.025, 0.5, 0.975)),

```

```

      ~mcse_quantile(.x, probs=c(0.025, 0.5, 0.975)),
      "rhat") %>%
  arrange(desc(rhat))
  print(rand_summ)
}

# Extract posterior density
prev_post <- as_tibble(as_draws_matrix(rand_sam), rownames = "Iteration") %>%
  select(c("Iteration", contains("theta["))) %>%
  pivot_longer(cols = contains("theta["),
               names_to = "Feat",
               values_to = "predicted_prev") %>%
  mutate(Feat_names = factor(Feat, levels = c(paste0("theta[", 1:nFeat, "]")), labels = FeatNames)) %>%
  select(Iteration, Feat_names, predicted_prev)

return(prev_post)
}

plot_post_density <- function(jags_post, sample_data, feat, theta_mu, theta_sigma) {
  # Plots posterior densities and their 95% credible intervals, and sample prevalence confidence intervals
  # jags_post = output from do_jags_rand_model, ie posterior densities
  # sample_data = output from get_adjusted_prev, ie sample prevalences with confidence intervals
  # feat = the same feature used as the random effect in do_jags_rand_model
  # theta_mu, theta_sigma = mean and sd of beta prior distribution used in do_jags_rand_model

  # calculate posterior credible intervals
  post_ci <- jags_post %>%
  group_by(across(all_of(feat))) %>%
  summarise(post_lower = quantile(predicted_prev, 0.025),
            post_upper = quantile(predicted_prev, 0.975))

  print(ggplot() +
        geom_density(data = jags_post, aes(x = predicted_prev)) +
        geom_vline(data = post_ci, aes(xintercept = post_lower), color = "blue", linetype = "dotted") +
        geom_vline(data = post_ci, aes(xintercept = post_upper), color = "blue", linetype = "dotted") +
        geom_vline(data = sample_data, aes(xintercept = ci_lower), color = "red", linetype = "dashed") +
        geom_vline(data = sample_data, aes(xintercept = ci_upper), color = "red", linetype = "dashed") +
        facet_wrap(as.formula(paste0("~", feat))) +
        labs(title = paste0("Prior mean = ", theta_mu, ", prior sd = ", signif(theta_sigma, 3))))
}

```

Bayesian prevalence analysis

Set global parameters

```

nObs <- nrow(chile_bayes_aut)
nIter <- 1000
nBurn <- 1000
pars <- c("theta_a", "theta_b", "theta", "aut_sample", "aut_pred")

theta_mu_prior <- 0.0046
theta_sigma_prior <- (0.0047-0.0045) / (2*1.96)

```

```

theta_mu_extrapolate <- c(0.005, 0.01, 0.015, 0.3) # 0.5%, 1%, 1.5%, 3% prevalence
theta_sigma_extrapolate <- c(rep(0.0001/1.96, 4)) # Same as chosen prior
#theta_mu <- c(theta_mu_prior, theta_mu_sens)
#theta_sigma <- c(theta_sigma_prior, theta_sigma_sens)
#theta_a <- theta_mu * (theta_mu * (1-theta_mu) / theta_sigma^2 - 1)
#theta_b <- (1 - theta_mu) * (theta_mu * (1-theta_mu) / theta_sigma^2 - 1)

rand_model <- "model {
  for(i in 1:nFeat) { # For each category in the feature grouping
    theta[i] ~ dbeta(theta_a, theta_b)
    aut_sample[i] ~ dbin(theta[i], nObs[i])
    aut_pred[i] ~ dbin(theta[i], nObs[i])
  }
}"

```

Common effects model with unadjusted sample prevalence

Random effects analysis

Standardise prevalence by Chile's age and sex based population sizes using https://seer.cancer.gov/seerstat/WebHelp/Rate_Algorithms.htm and <https://wonder.cdc.gov/wonder/help/cancer/fayfeuerconfidenceintervals.pdf>

See https://github.com/Dpananos/bayes_multiple_measures/blob/master/analysis/sensitivity_analysis.R for more sensitivity analysis ideas

Random effect on region

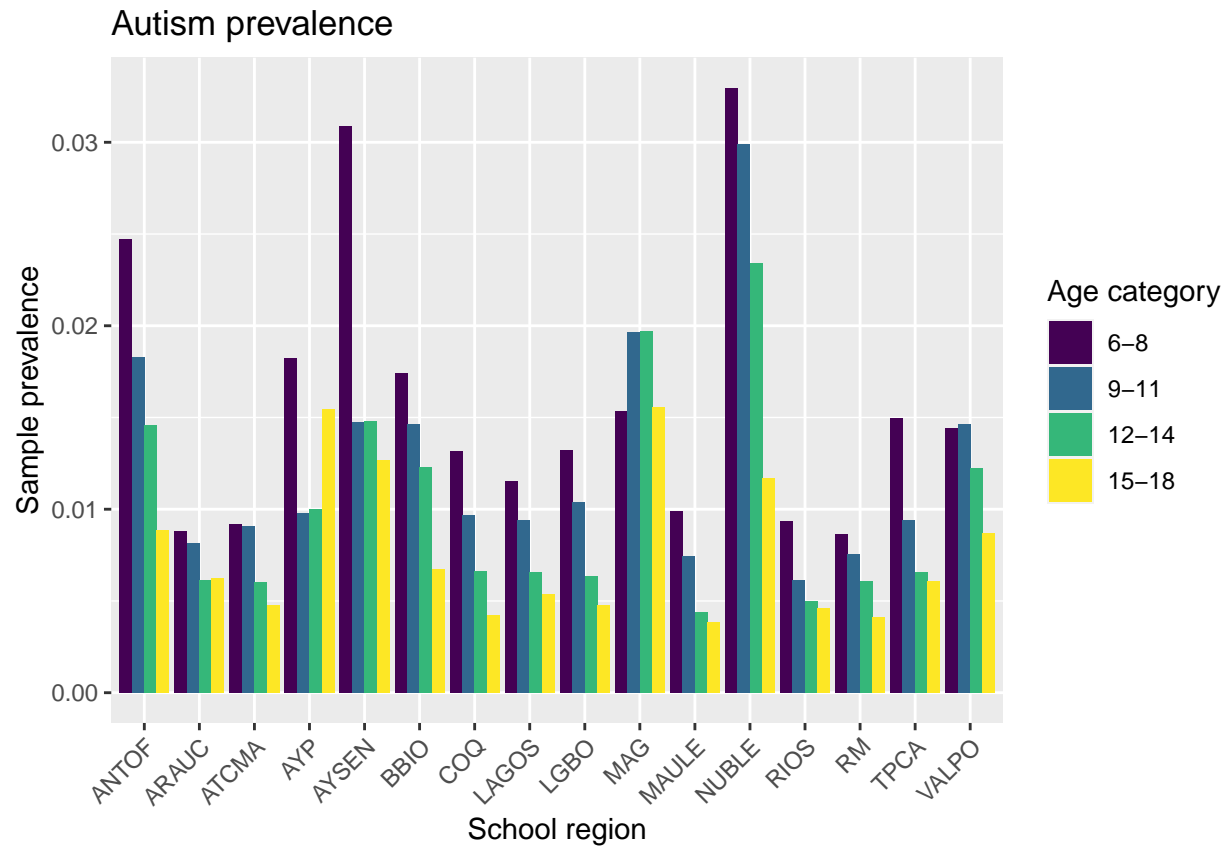
```

aut_prev_region <- get_grouped_prev(x = chile_bayes_aut, stdpop = chile_stdpop,
                                     grouping_vars = c("school_region_name_abr", "age_june30", "age_cat_name"))

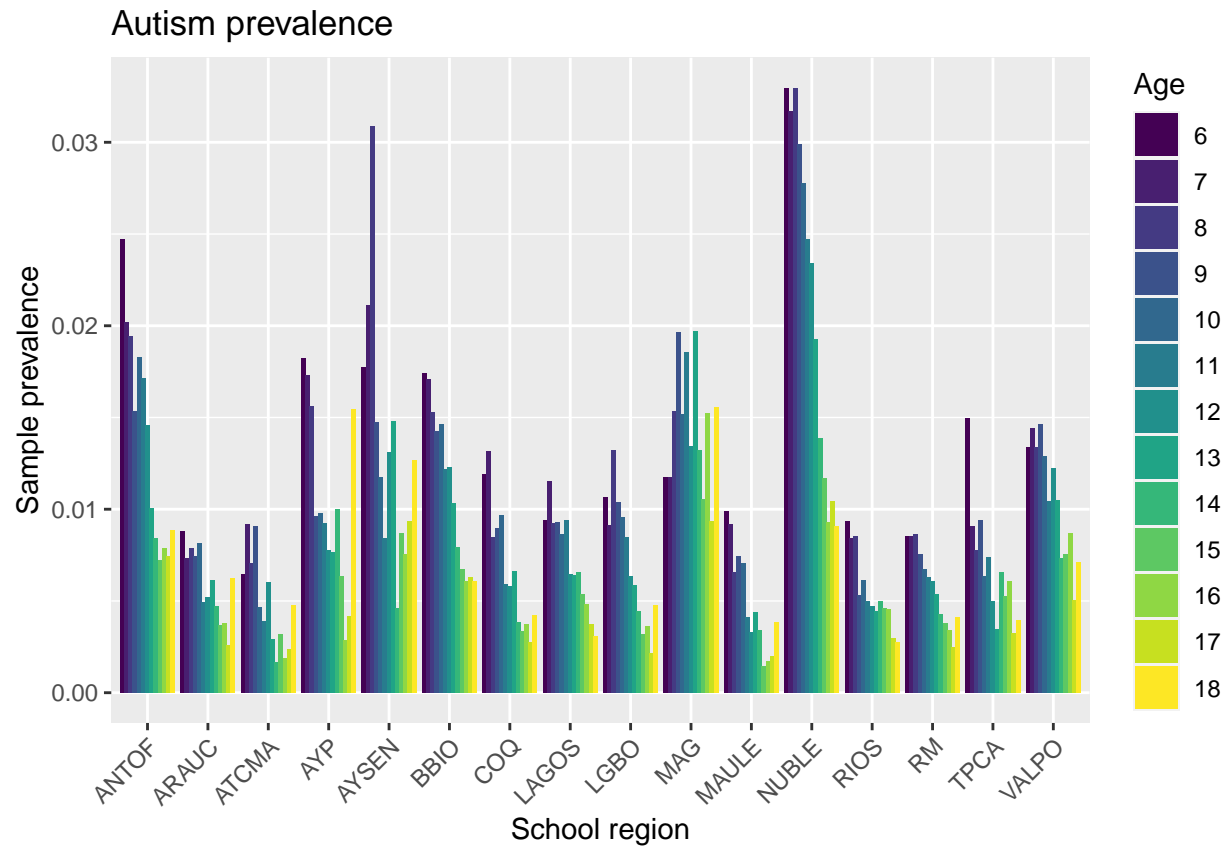
## `summarise()` has grouped output by 'school_region_name_abr', 'age_june30',
## 'age_cat_name', 'sex', 'sex_desc'. You can override using the `.groups`
## argument.

ggplot(data = aut_prev_region) +
  geom_col(aes(x = school_region_name_abr, y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "School region",
       y = "Sample prevalence",
       fill = "Age category")

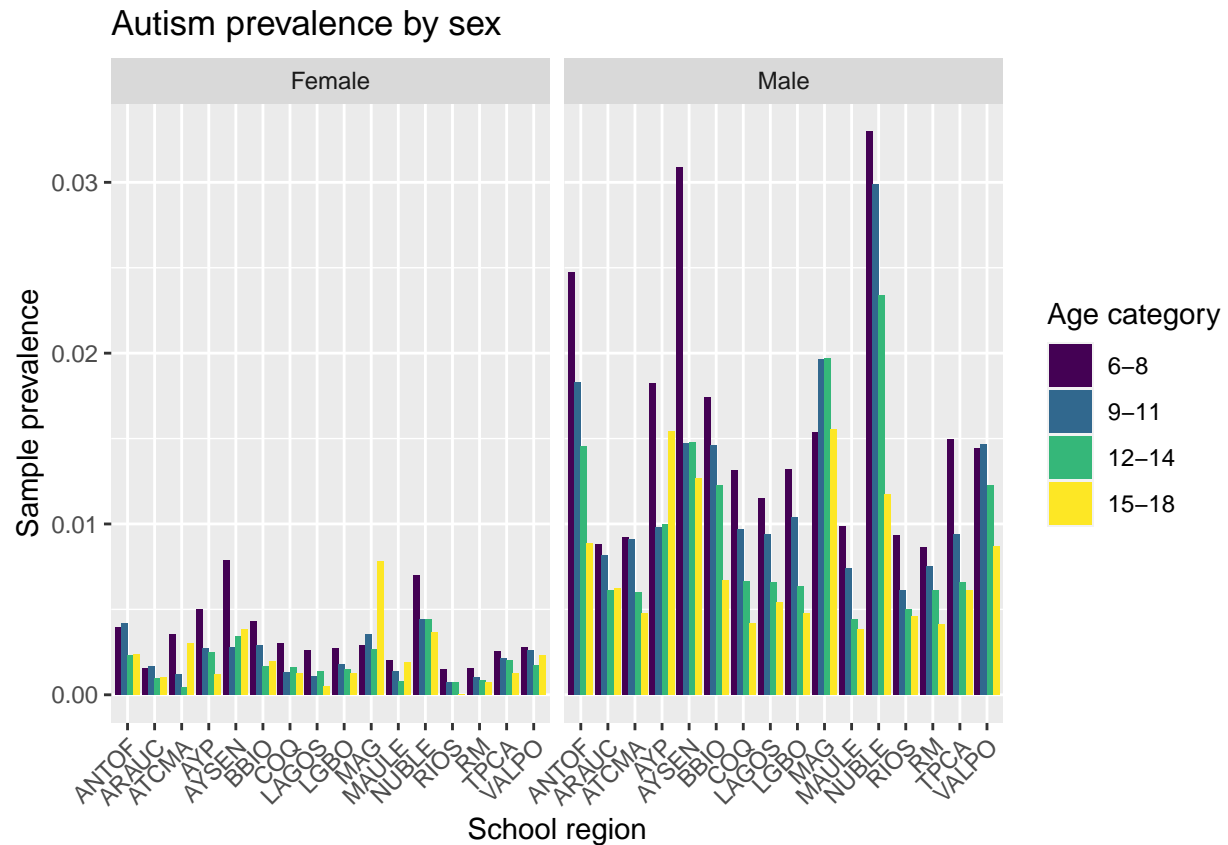
```



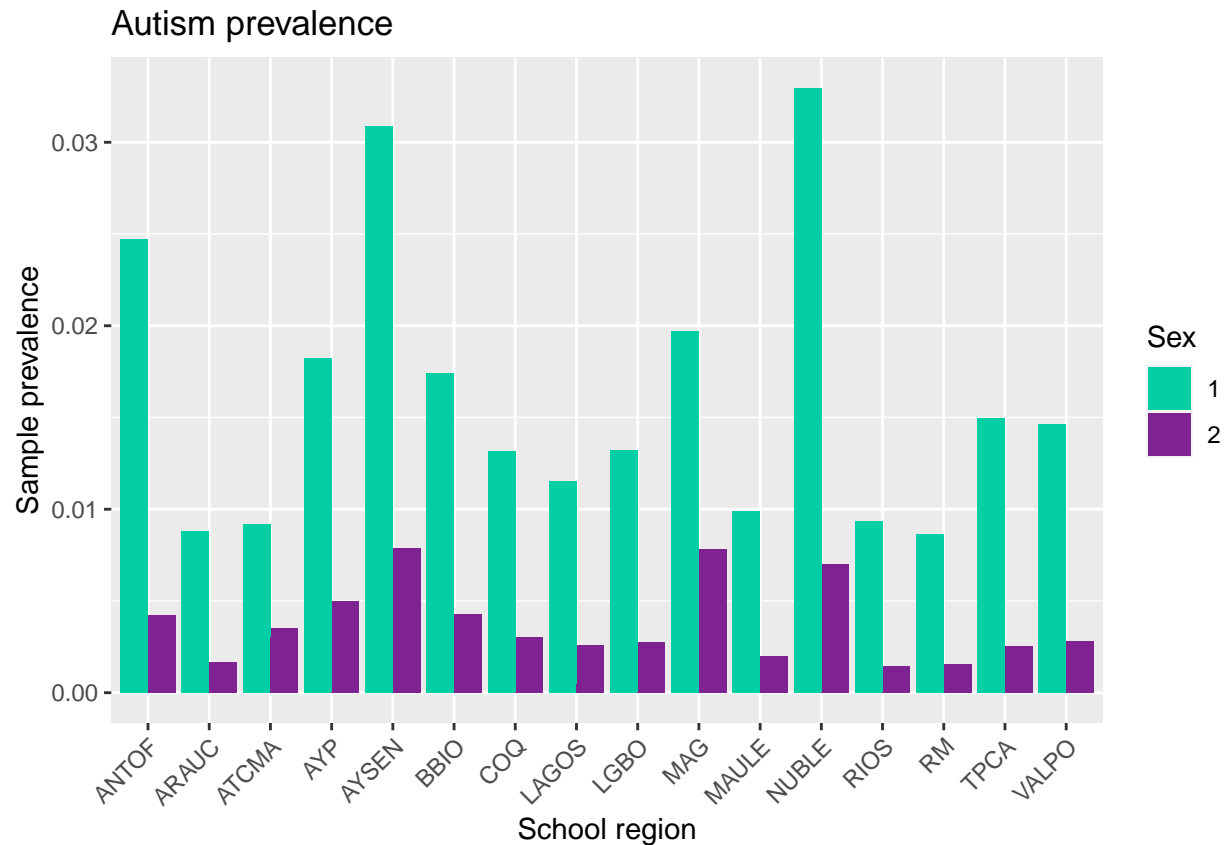
```
ggplot(data = aut_prev_region) +
  geom_col(aes(x = school_region_name_abr, y = sample_prevalence, group = age, fill = as.factor(age)),
    scale_fill_viridis_d() +
    theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
    labs(title = "Autism prevalence",
      x = "School region",
      y = "Sample prevalence",
      fill = "Age")
```

```
ggplot(data = aut_prev_region) +
  geom_col(aes(x = school_region_name_abr, y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
       x = "School region",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_region) +
  geom_col(aes(x = school_region_name_abr, y = sample_prevalence, group = sex, fill = as.factor(sex)),
    # 1 is male, 2 is female
    #scale_fill_manual(values = wes_palette("GrandBudapest1")) +
    scale_fill_manual(values = c("#03CEA4", "#802392")) +
    theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
    labs(title = "Autism prevalence",
      x = "School region",
      y = "Sample prevalence",
      fill = "Sex")
```



```

aut_prev_region_adj <- get_adjusted_prev(aut_prev_region, grouping_vars = "school_region_name_abr")

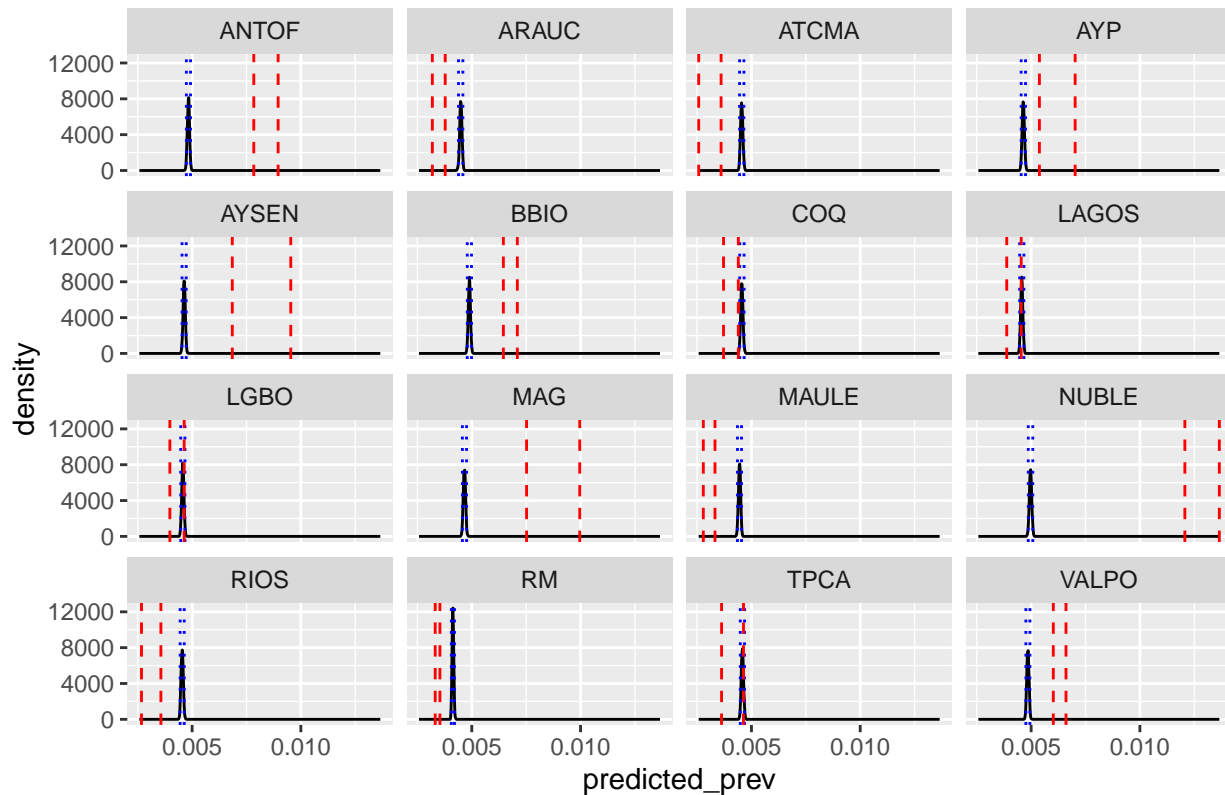
aut_prev_region_post <- do_jags_rand_model(x = aut_prev_region_adj,
  feat = "school_region_name_abr",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_region_name_abr" = "Feat_names")

plot_post_density(aut_prev_region_post, aut_prev_region_adj, feat = "school_region_name_abr", theta_mu = theta_mu_prior, theta_sigma = theta_sigma_prior)

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.

```

Prior mean = 0.0046, prior sd = 5.1e-05



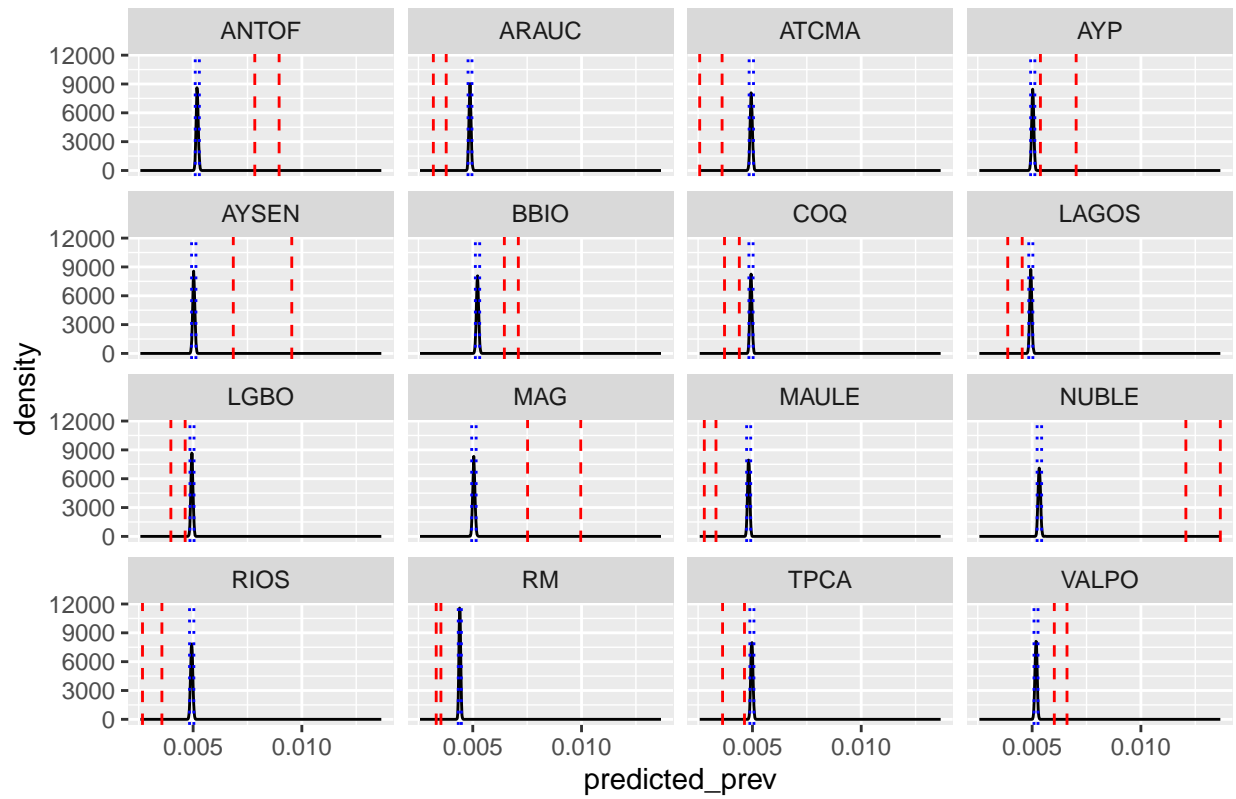
Predictions for higher population prevalence - increase prior mean

```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_region_post <- do_jags_rand_model(x = aut_prev_region_adj,
                                             feat = "school_region_name_abr",
                                             model = rand_model,
                                             theta_mu = theta_mu_extrapolate[j],
                                             theta_sigma = theta_sigma_extrapolate[j],
                                             pars = pars,
                                             convergence_checks = FALSE) %>%
    rename("school_region_name_abr" = "Feat_names")

  plot_post_density(aut_prev_region_post,
                    aut_prev_region_adj,
                    feat = "school_region_name_abr",
                    theta_mu = theta_mu_extrapolate[j],
                    theta_sigma = theta_sigma_extrapolate[j])
}
```

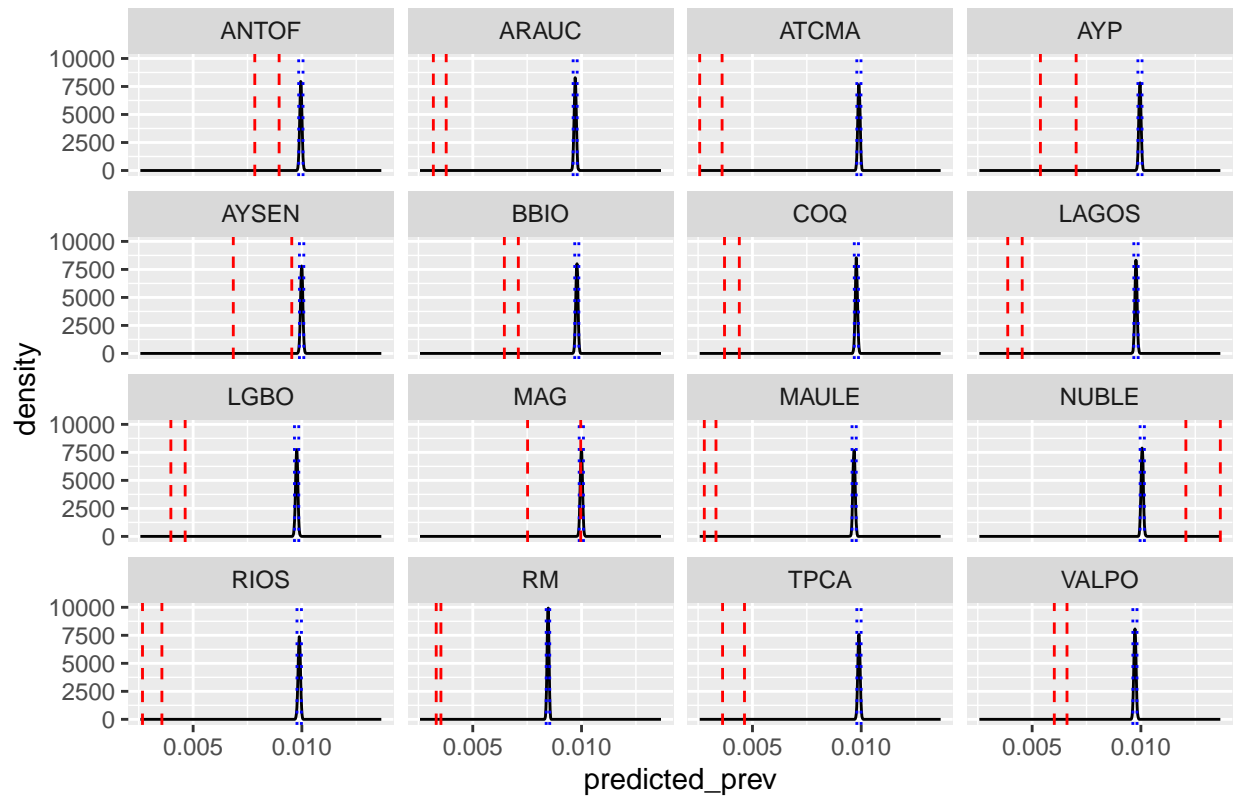
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



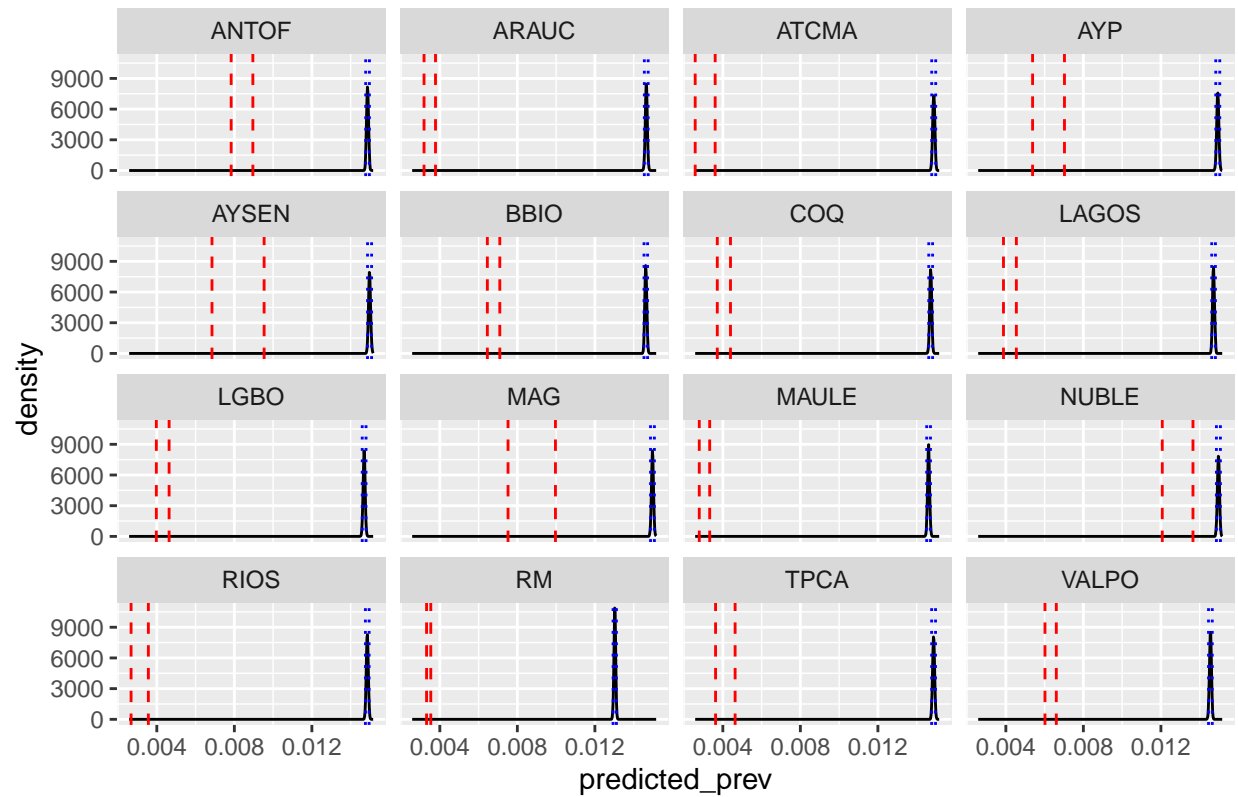
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



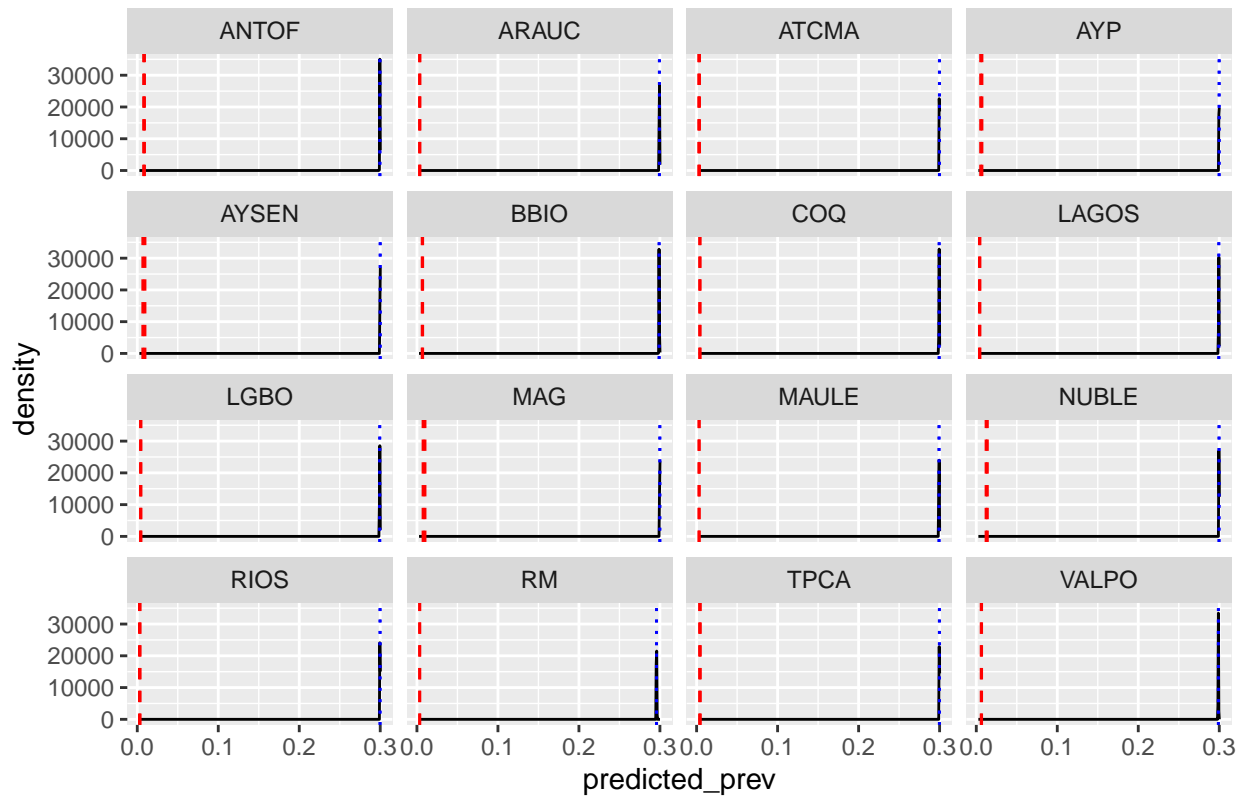
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

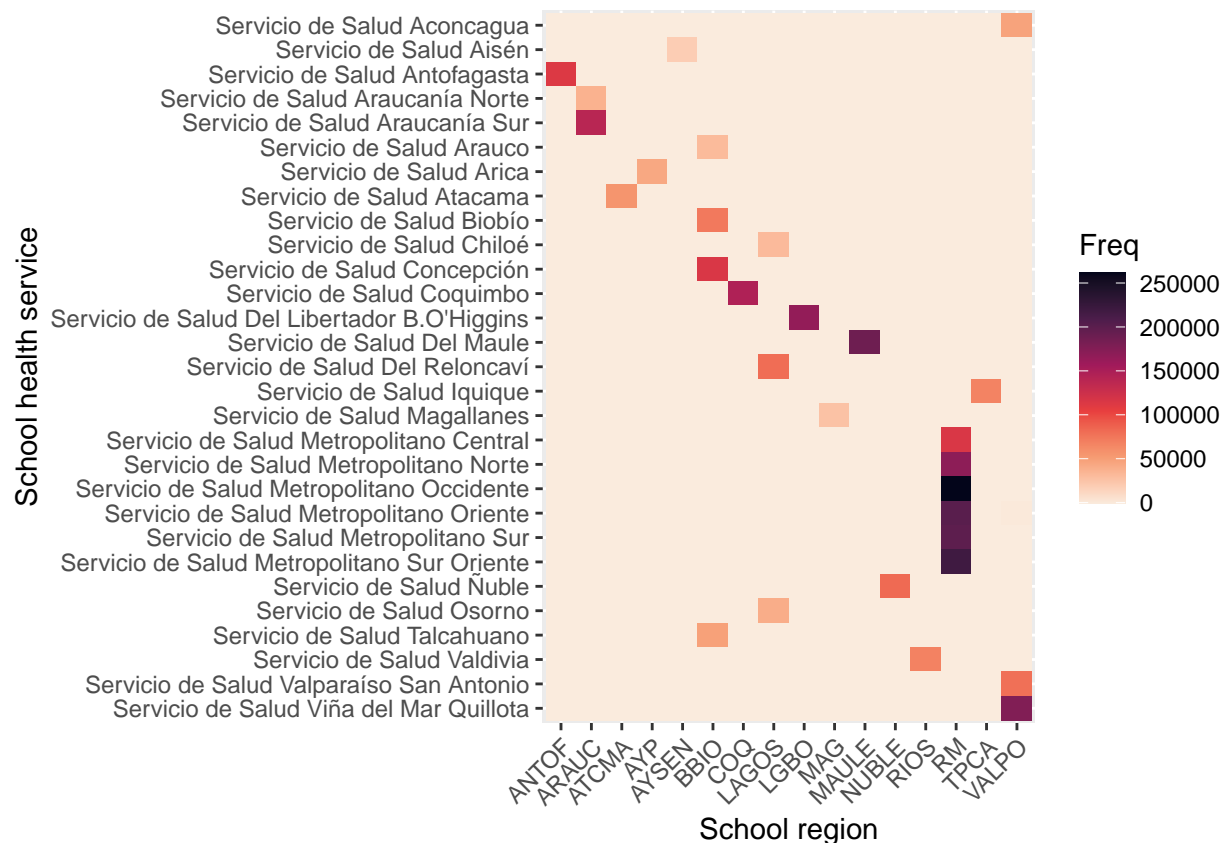
Prior mean = 0.3, prior sd = 5.1e-05



Random effect on health service

Show which regions each health service operates in

```
map_tab_df <- as.data.frame(table(chile_bayes_aut$school_region_name_abr, chile_bayes_aut$health_service))
ggplot(map_tab_df, aes(x = Var1, y = Var2, fill = Freq)) +
  geom_tile() +
  #scale_fill_gradient(low = "white", high = "blue") +
  scale_fill_viridis_c(option = "rocket", direction = -1) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  scale_y_discrete(limits = rev(levels(map_tab_df$Var2))) +
  labs(x = "School region", y = "School health service", fill = "Freq")
```

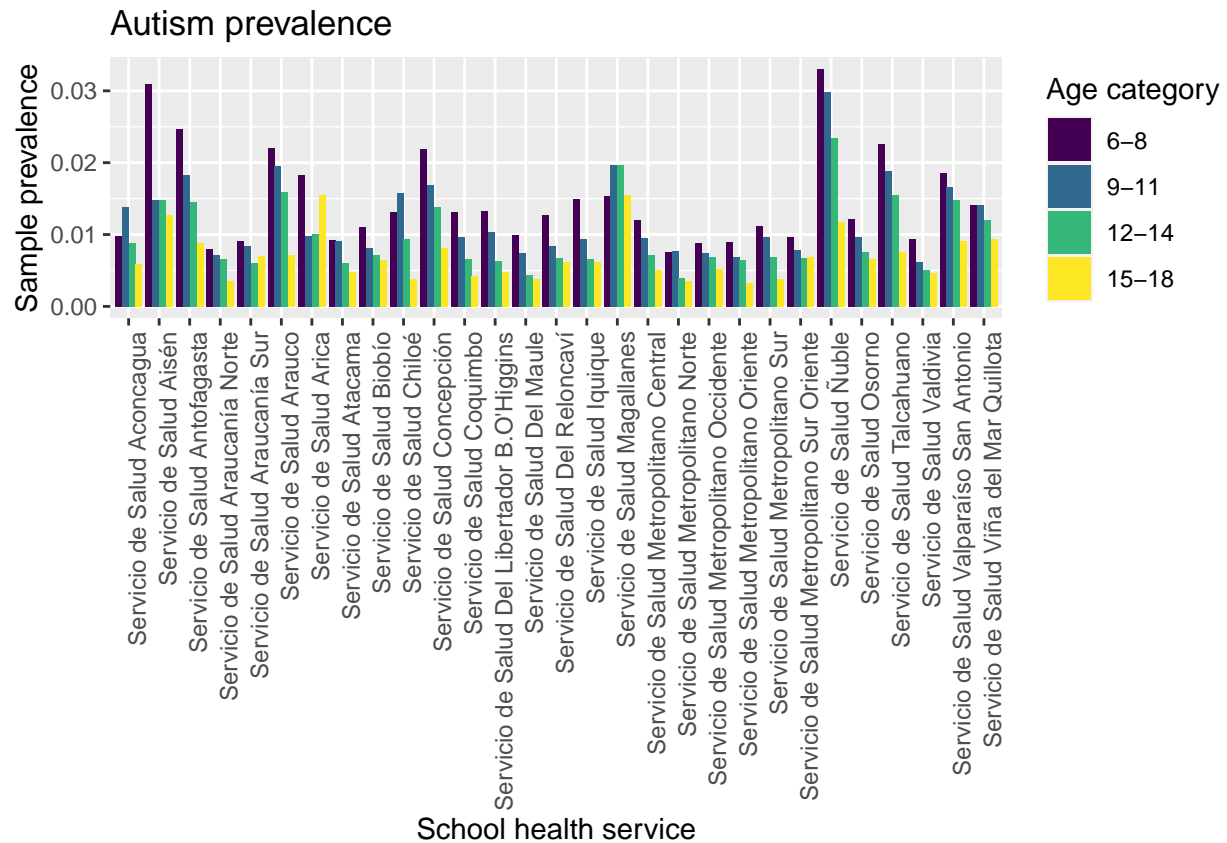



Each health service operates in only one region, some regions (RM, VALPO, LAGOS, BBIO, ARAUC) have multiple health services.

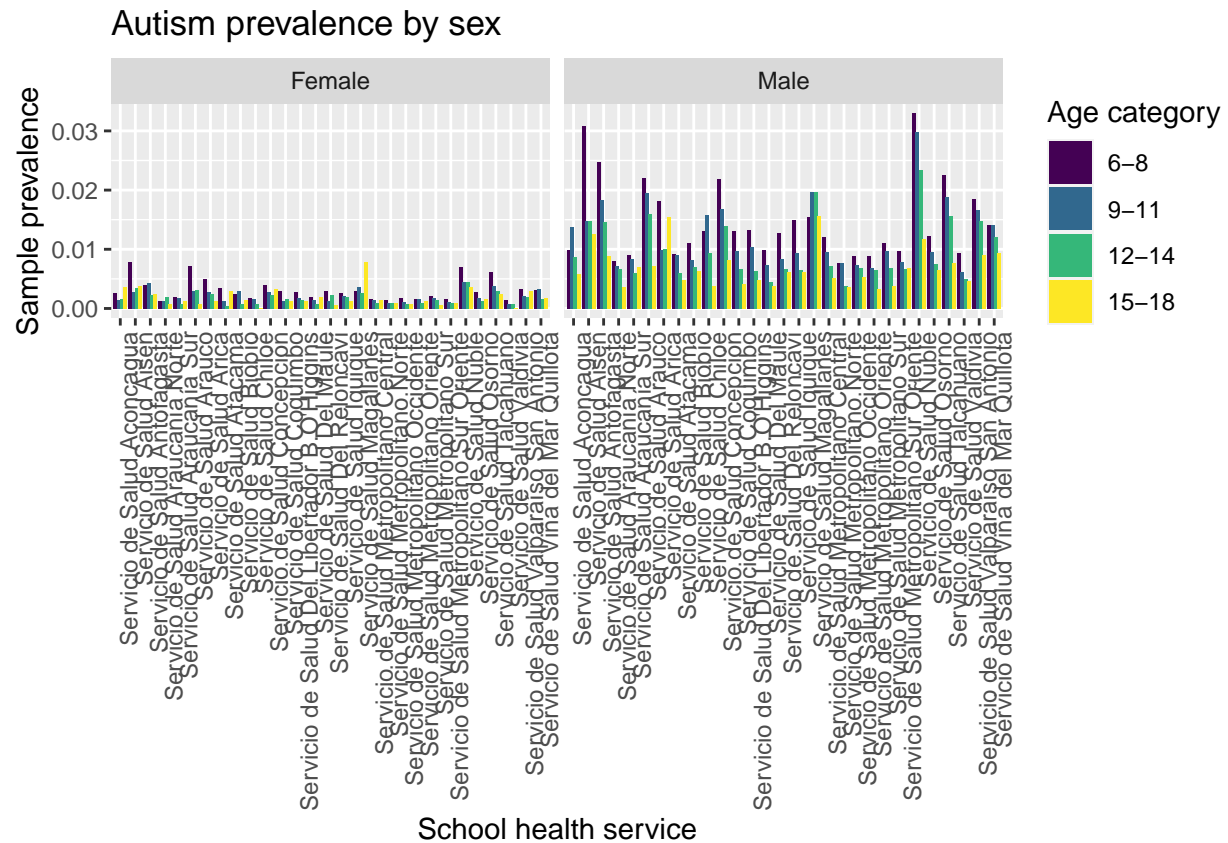
```
aut_prev_health <- get_grouped_prev(x = chile_bayes_aut, stdpop = chile_stdpop,
                                     grouping_vars = c("health_service_name", "age_june30", "age_cat_name"))

## `summarise()` has grouped output by 'health_service_name', 'age_june30',
## 'age_cat_name', 'sex', 'sex_desc'. You can override using the `.groups`
## argument.

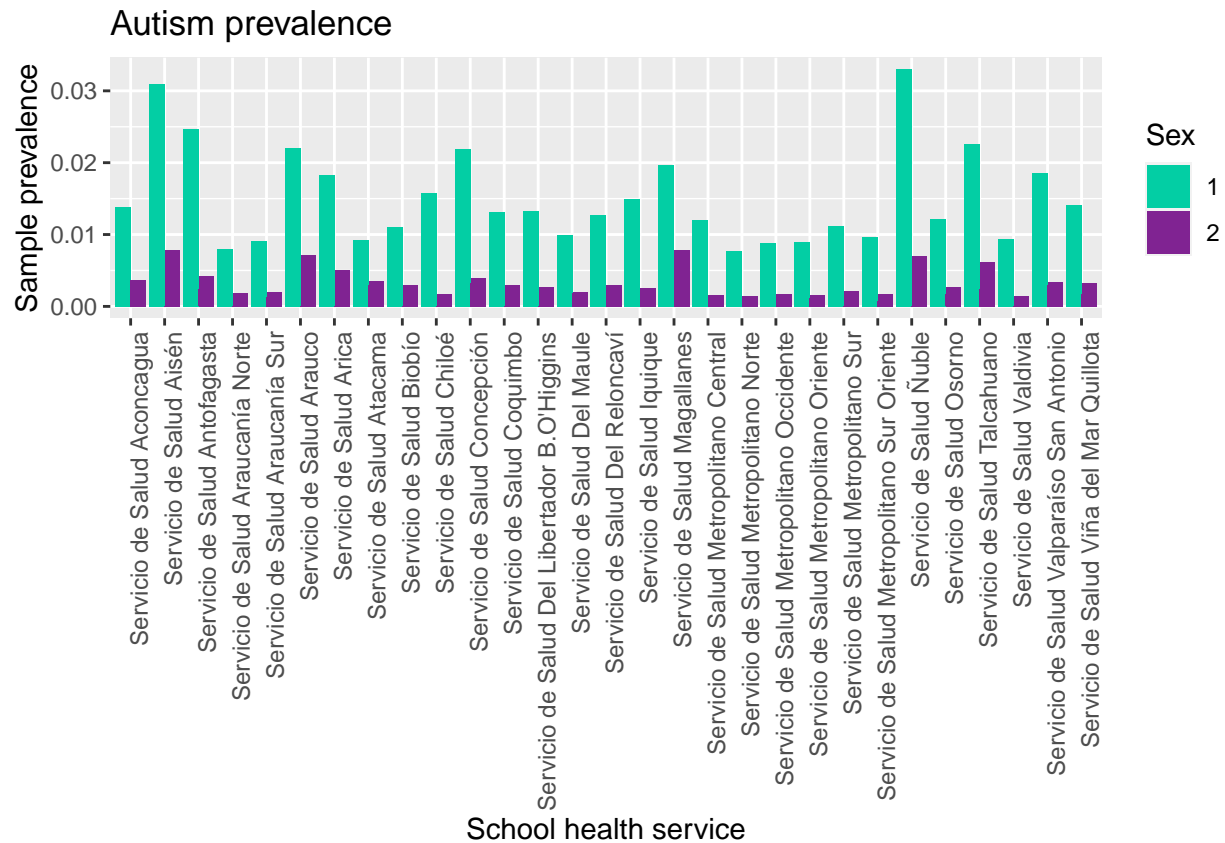
ggplot(data = aut_prev_health) +
  geom_col(aes(x = health_service_name, y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "School health service",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_health) +
  geom_col(aes(x = health_service_name, y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
       x = "School health service",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_health) +
  geom_col(aes(x = health_service_name, y = sample_prevalence, group = sex, fill = as.factor(sex)), position = "dodge") +
  # 1 is male, 2 is female
  scale_fill_manual(values = c("#03CEA4", "#802392")) +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "School health service",
       y = "Sample prevalence",
       fill = "Sex")
```



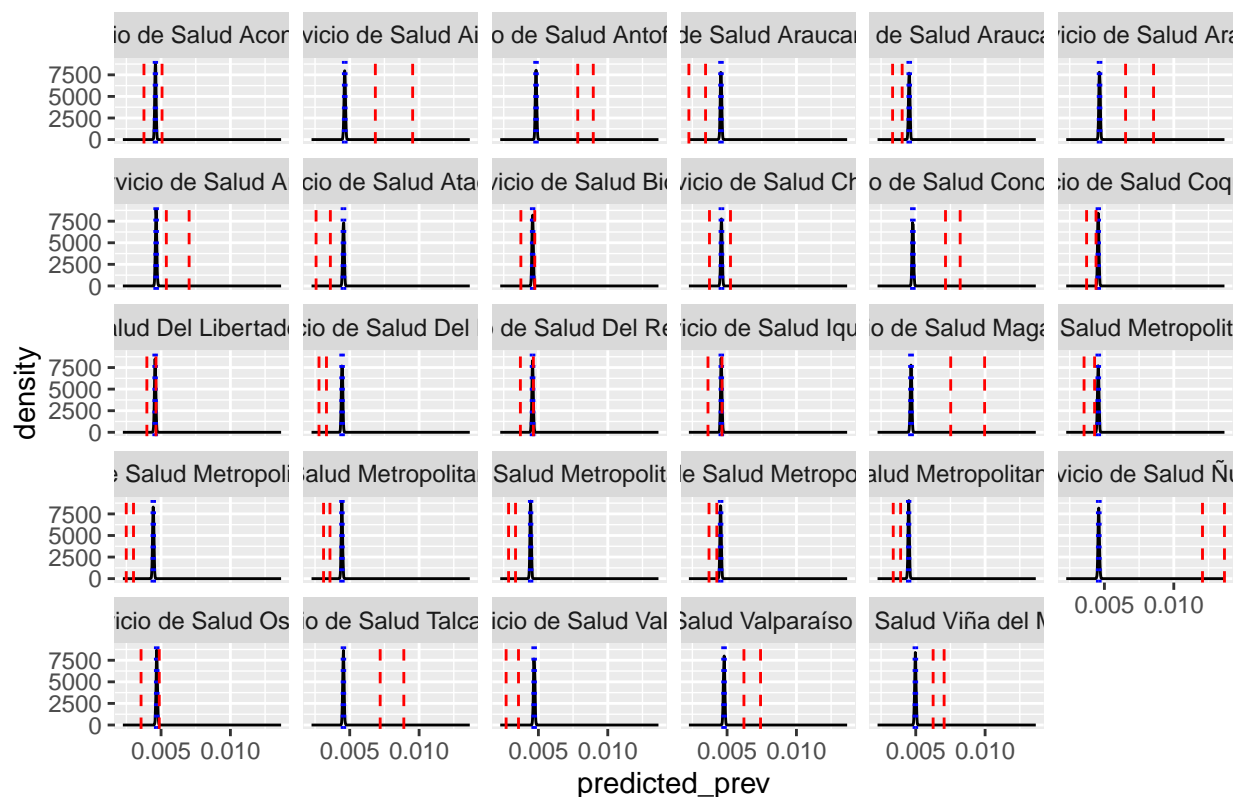
```
aut_prev_health_adj <- get_adjusted_prev(aut_prev_health, grouping_vars = "health_service_name")

aut_prev_health_post <- do_jags_rand_model(x = aut_prev_health_adj,
  feat = "health_service_name",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("health_service_name" = "Feat_names")

plot_post_density(aut_prev_health_post, aut_prev_health_adj, feat = "health_service_name", theta_mu = t
```

Don't know how to automatically pick scale for object of type
<draws_matrix/draws/matrix>. Defaulting to continuous.

Prior mean = 0.0046, prior sd = 5.1e-05



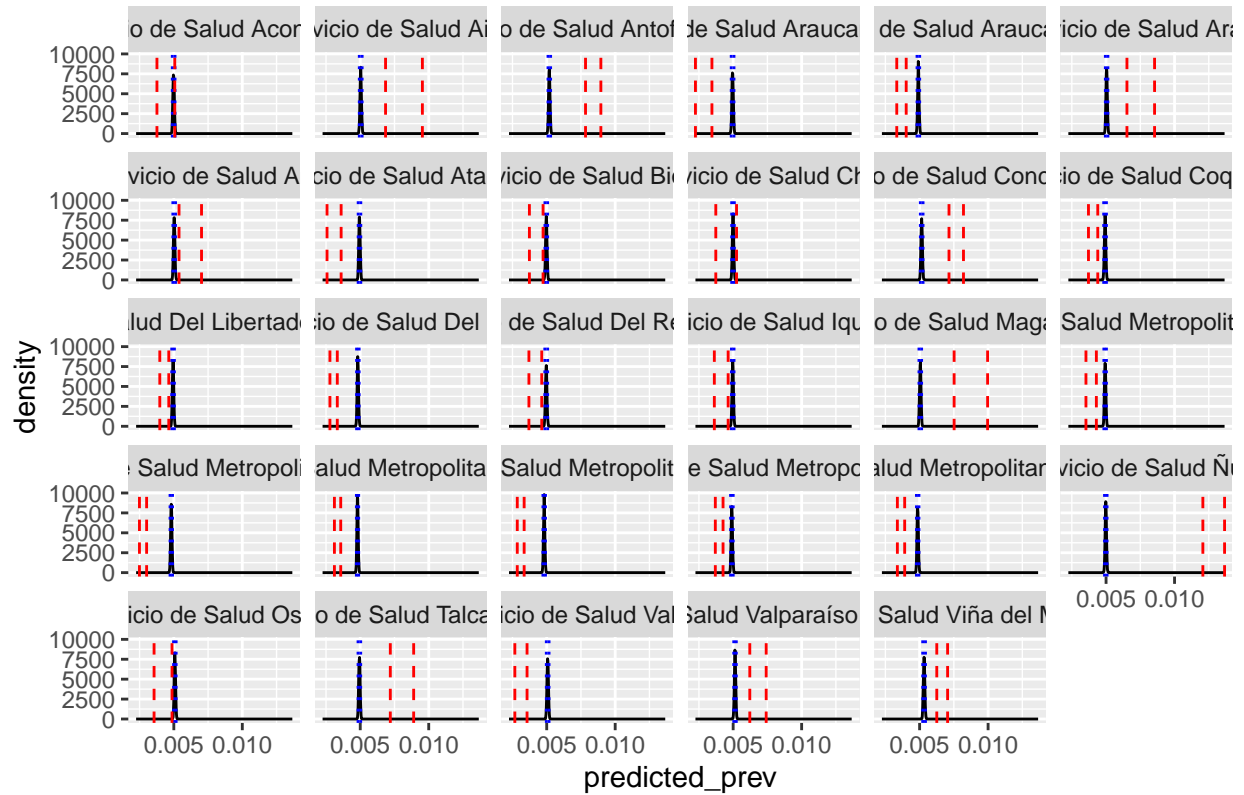
Predictions for higher population prevalence - increase prior mean

```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_health_post <- do_jags_rand_model(x = aut_prev_health_adj,
    feat = "health_service_name",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%
    rename("health_service_name" = "Feat_names")

  plot_post_density(aut_prev_health_post,
    aut_prev_health_adj,
    feat = "health_service_name",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

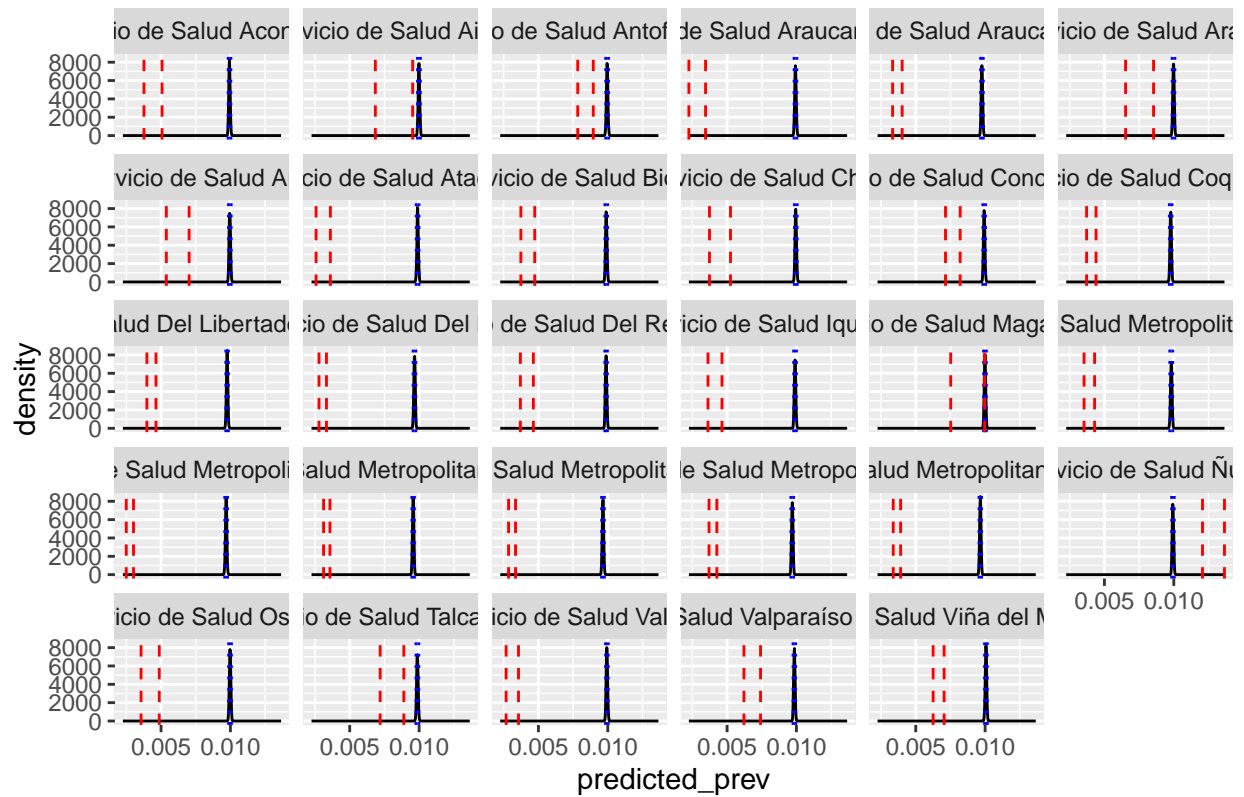
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



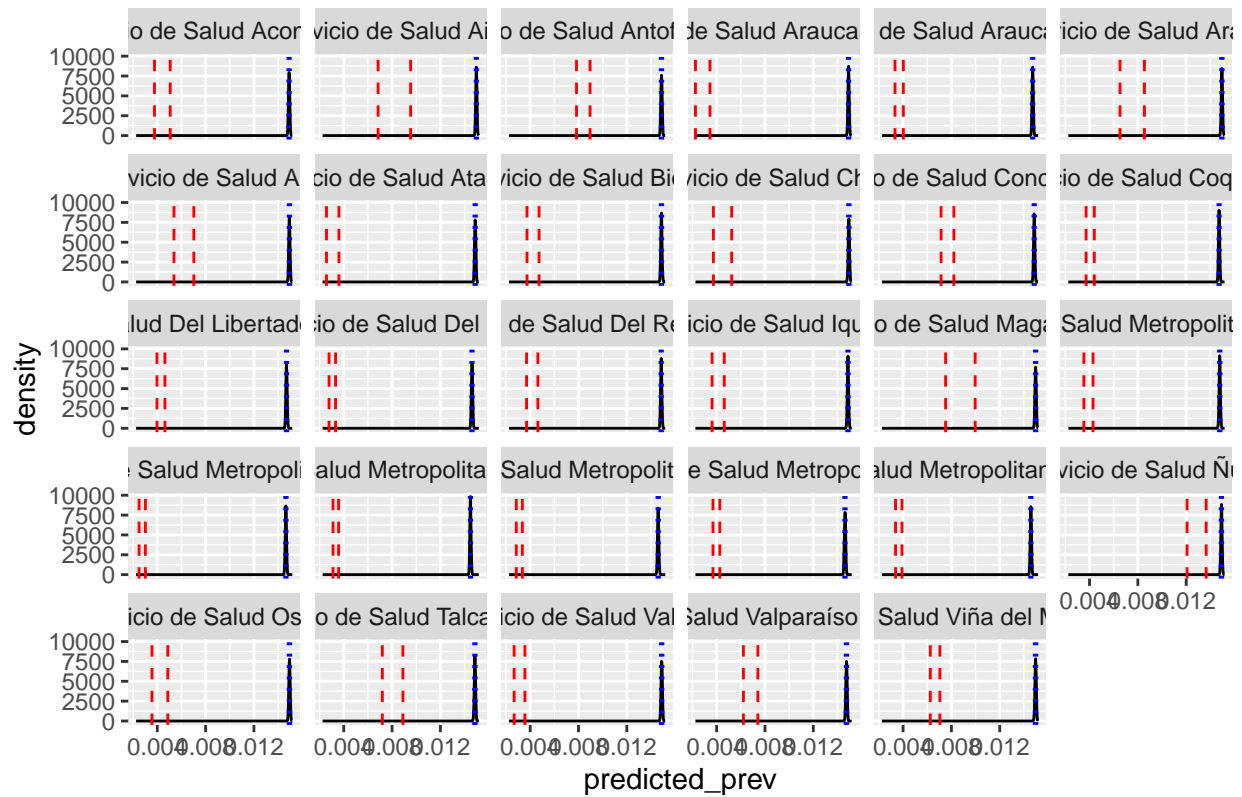
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



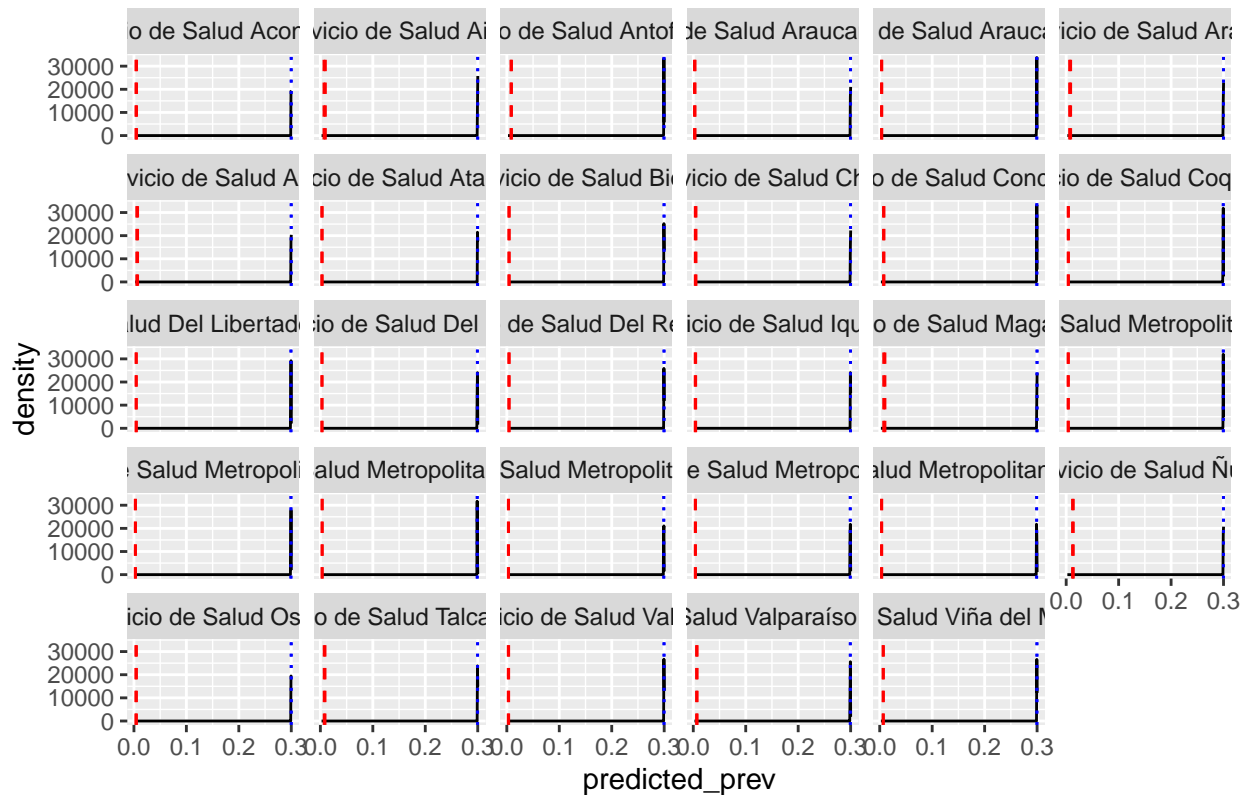
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05

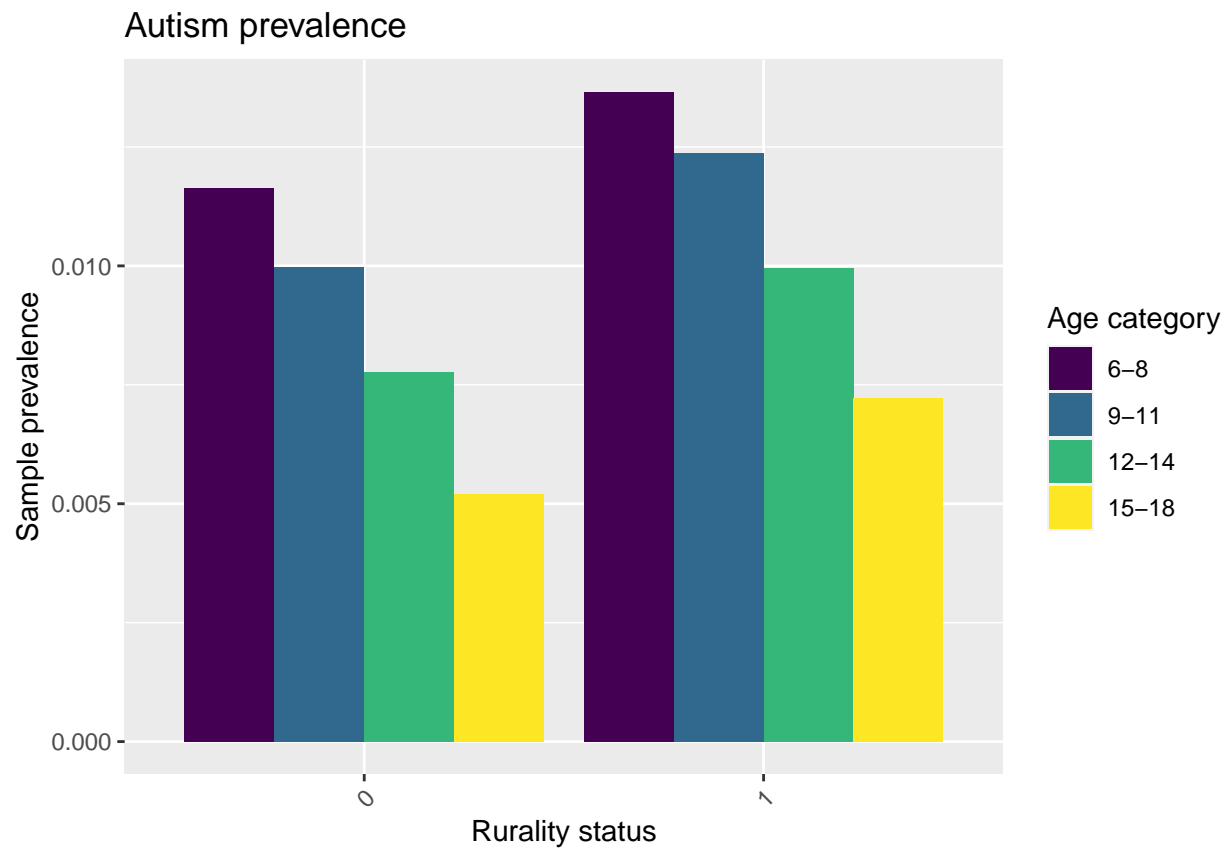


Random effect on rurality

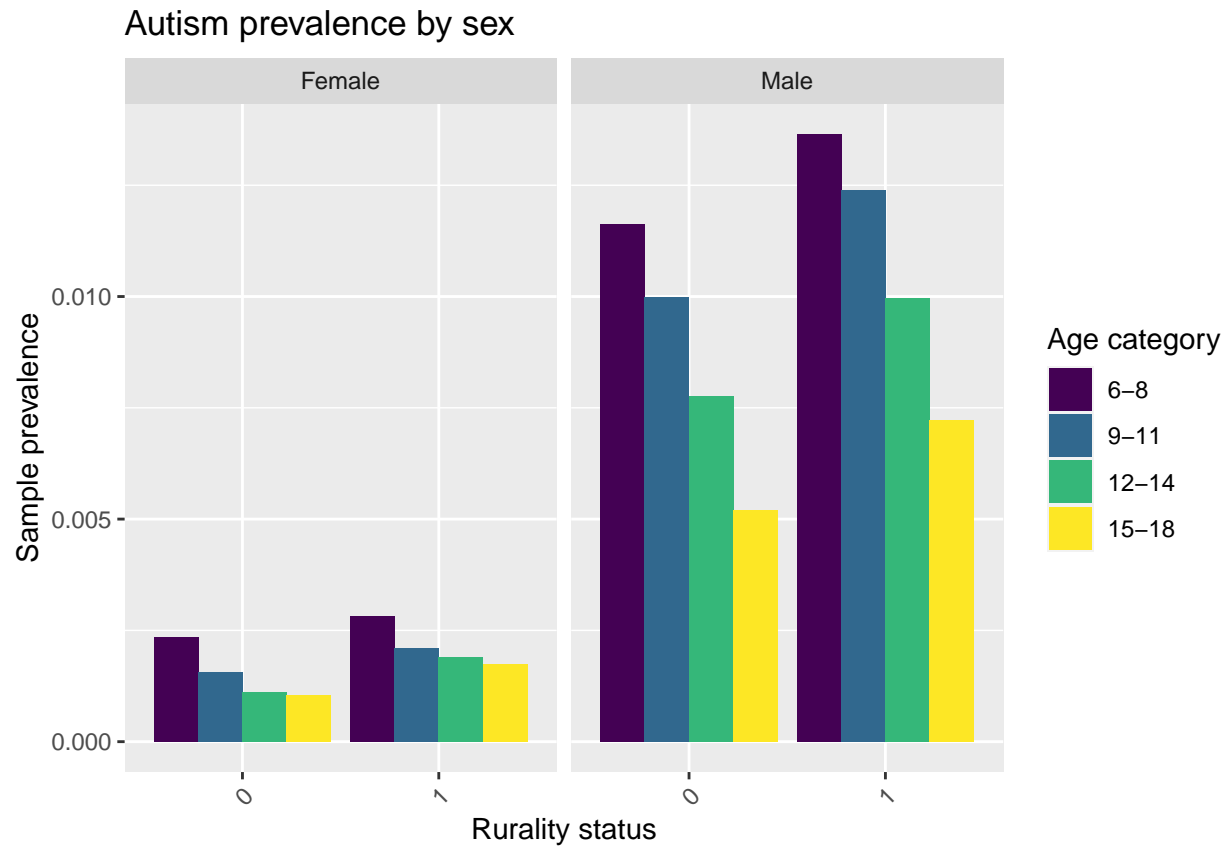
```
aut_prev_rural <- get_grouped_prev(x = chile_bayes_aut, stdpop = chile_stdpop,
                                   grouping_vars = c("school_rurality_code", "age_june30", "age_cat_name"))

## `summarise()` has grouped output by 'school_rurality_code', 'age_june30',
## 'age_cat_name', 'sex', 'sex_desc'. You can override using the `.groups`
## argument.

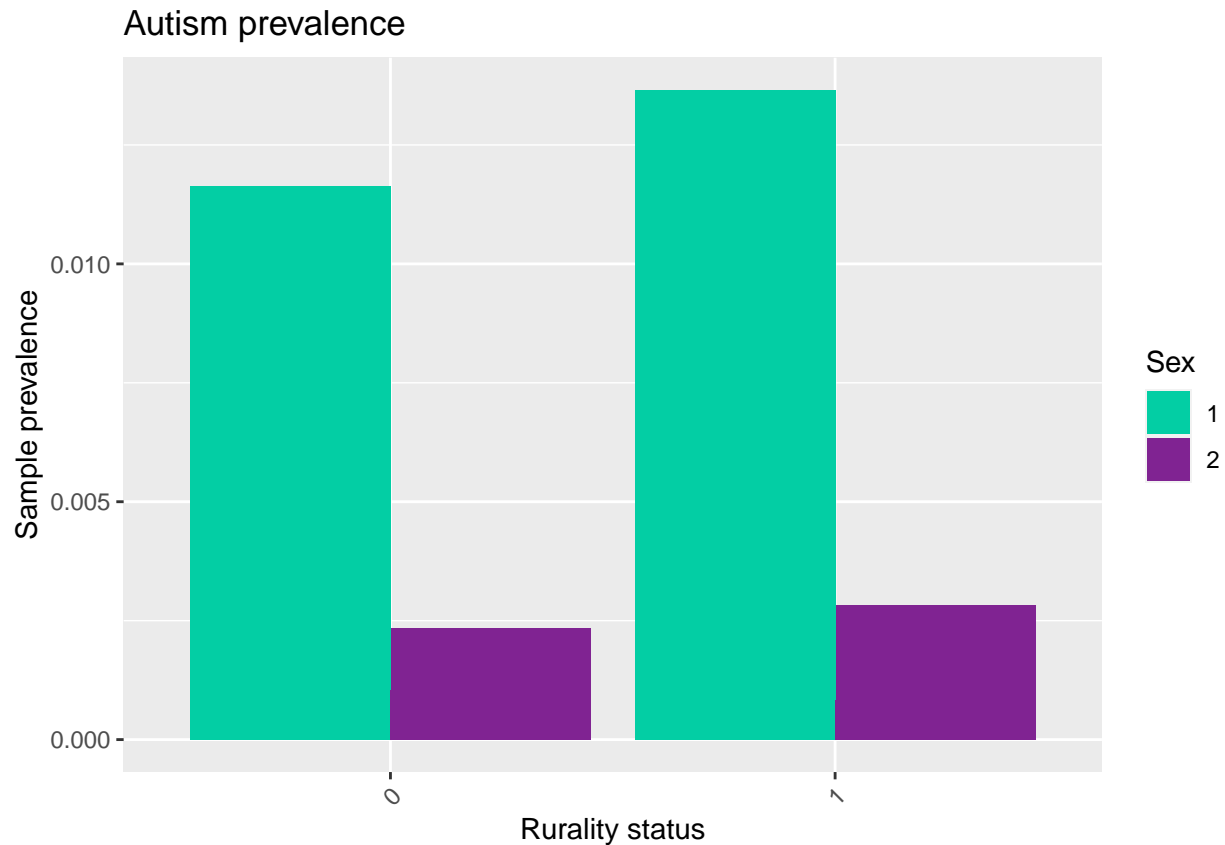
ggplot(data = aut_prev_rural) +
  geom_col(aes(x = as.factor(school_rurality_code), y = sample_prevalence, group = age_cat_name, fill =
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Rurality status",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_rural) +
  geom_col(aes(x = as.factor(school_rurality_code), y = sample_prevalence, group = age_cat_name, fill =
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
        x = "Rurality status",
        y = "Sample prevalence",
        fill = "Age category")
```



```
ggplot(data = aut_prev_rural) +
  geom_col(aes(x = as.factor(school_rurality_code), y = sample_prevalence, group = sex, fill = as.factor(school_rurality_code))) +
  scale_fill_manual(values = c("#03CEA4", "#802392")) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Rurality status",
       y = "Sample prevalence",
       fill = "Sex")
```



```
aut_prev_rural_adj <- get_adjusted_prev(aut_prev_rural, grouping_vars = "school_rurality_code")

aut_prev_rural_post <- do_jags_rand_model(x = aut_prev_rural_adj,
  feat = "school_rurality_code",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_rurality_code" = "Feat_names")

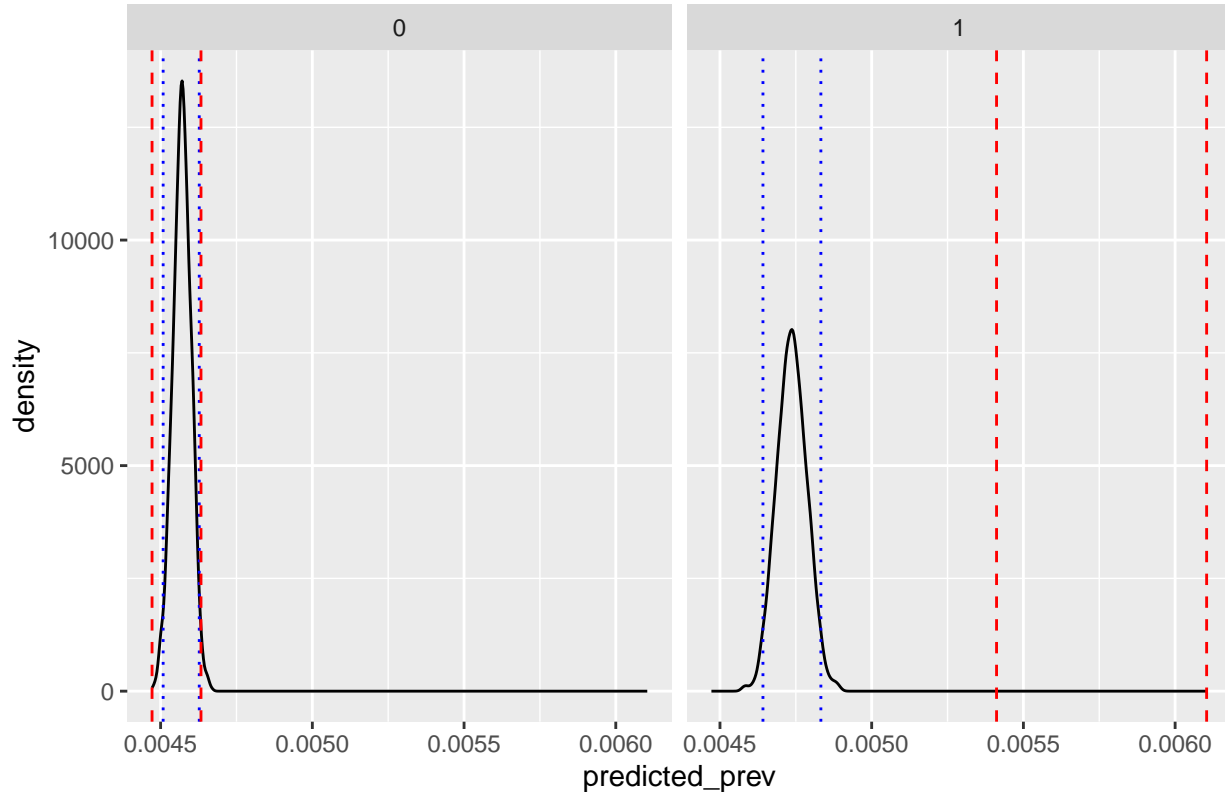
plot_post_density(aut_prev_rural_post, aut_prev_rural_adj, feat = "school_rurality_code", theta_mu = th
```

```
## Warning: Combining variables of class <factor> and <integer> was deprecated in ggplot2
## 3.4.0.
## i Please ensure your variables are compatible before plotting (location:
##   `combine_vars()`)
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

## Warning: Combining variables of class <integer> and <factor> was deprecated in ggplot2
## 3.4.0.
## i Please ensure your variables are compatible before plotting (location:
##   `combine_vars()`)
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
```

```
## generated.
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



Assuming 0 = city, 1 = rural. Narrower sample CI for city because sample size is bigger

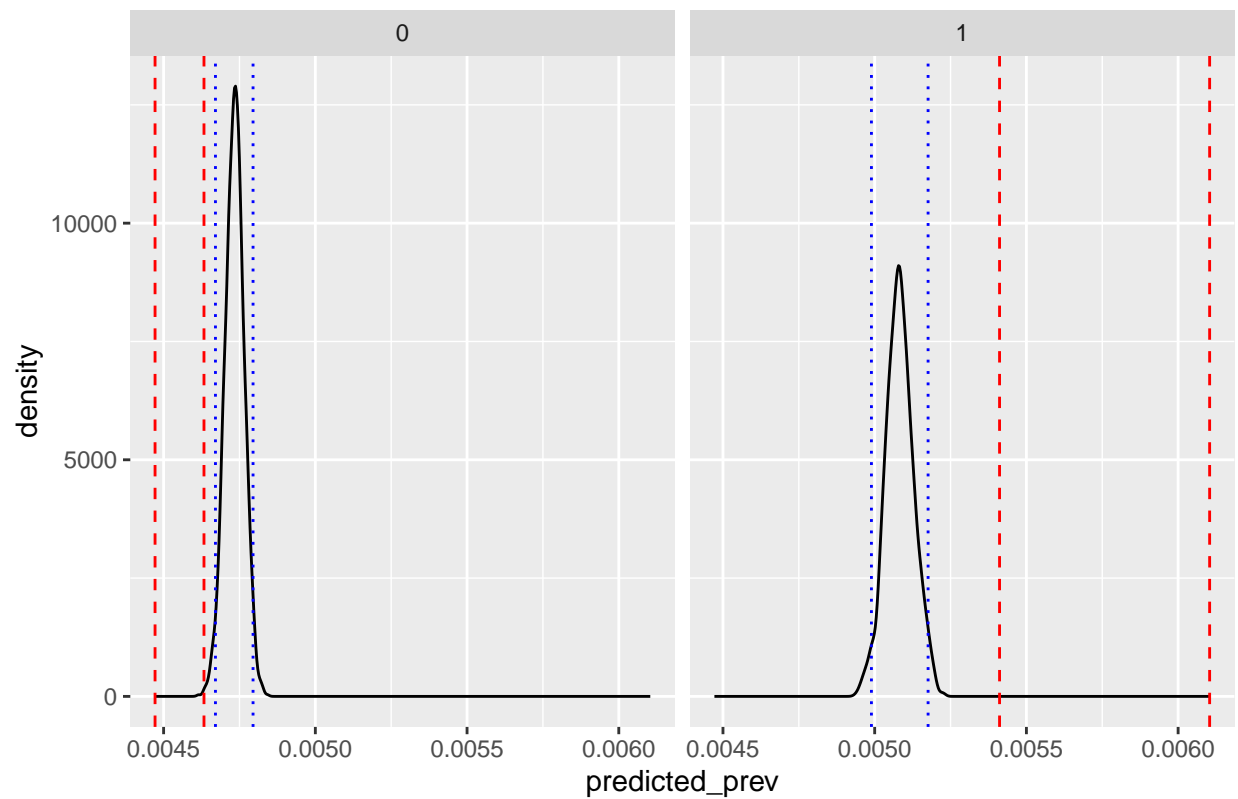
Predictions for higher population prevalence - increase prior mean

```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_rural_post <- do_jags_rand_model(x = aut_prev_rural_adj,
    feat = "school_rurality_code",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%
  rename("school_rurality_code" = "Feat_names")

  plot_post_density(aut_prev_rural_post,
    aut_prev_rural_adj,
    feat = "school_rurality_code",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

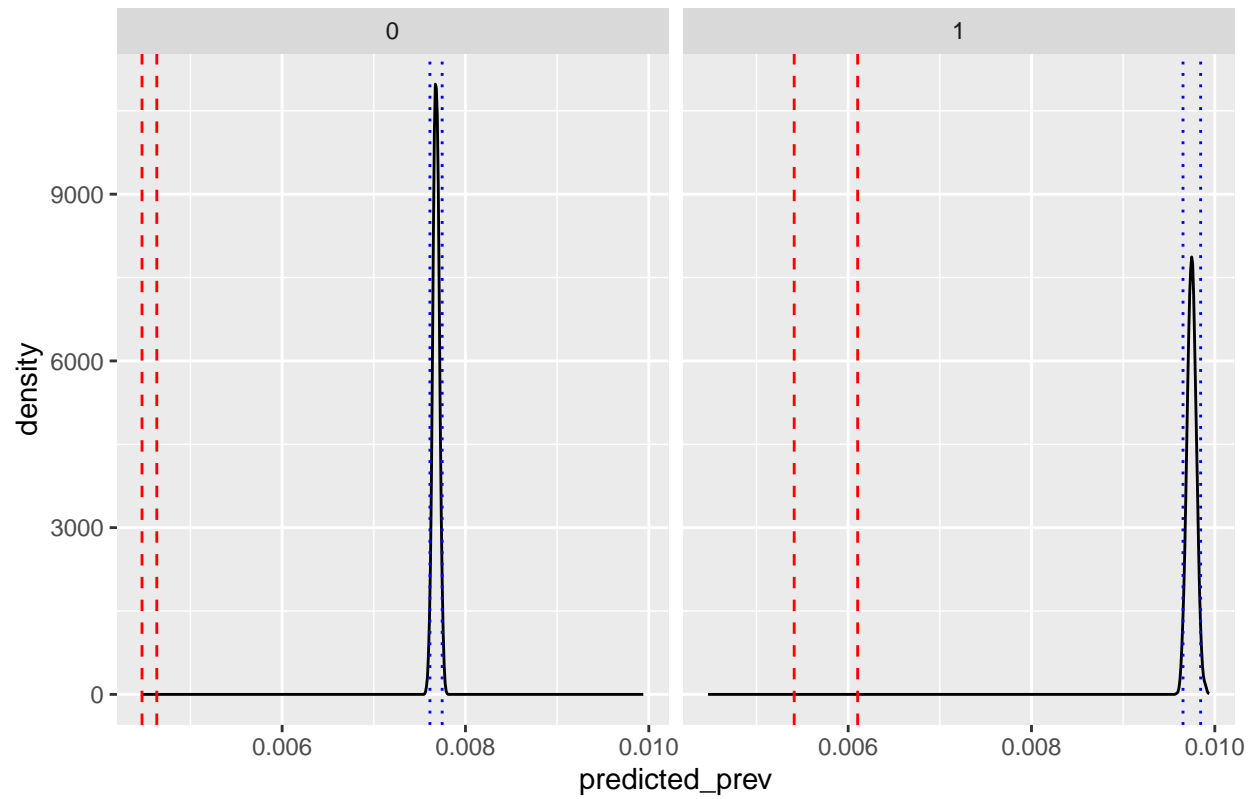
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



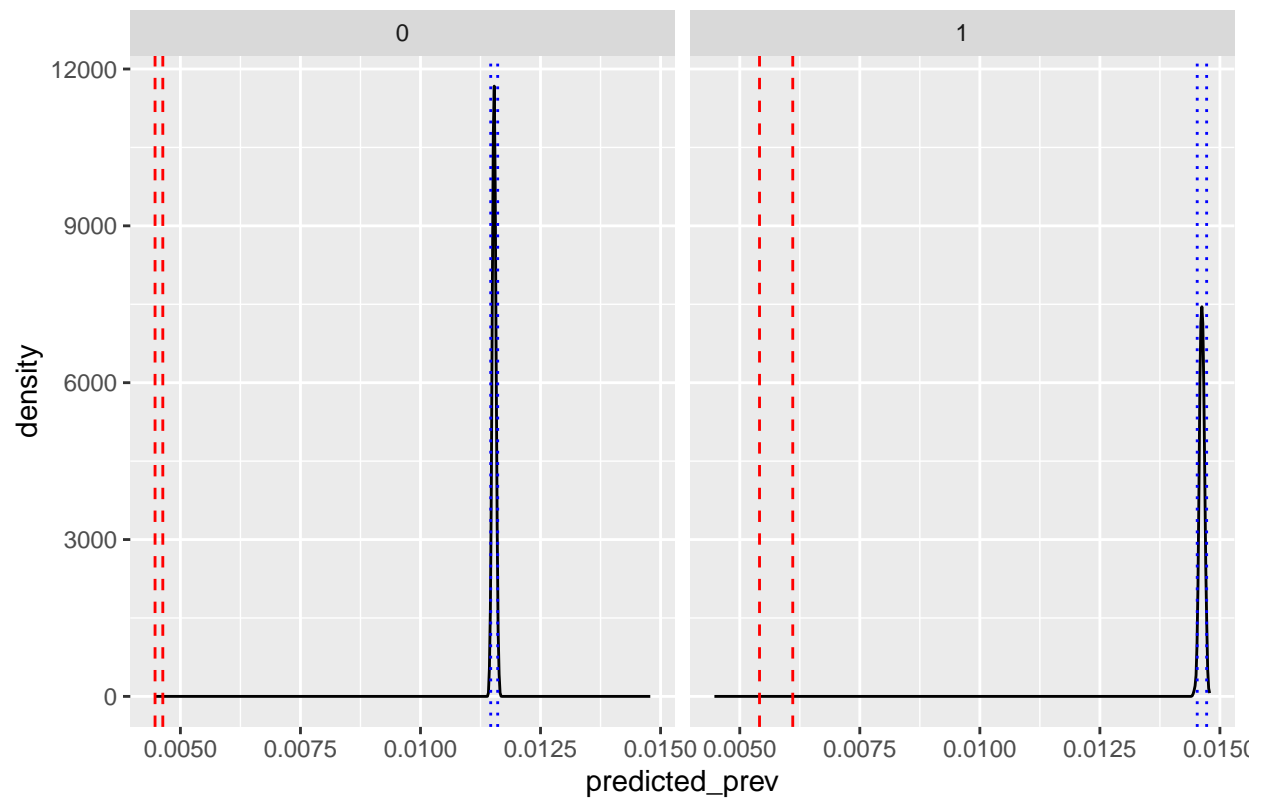
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05

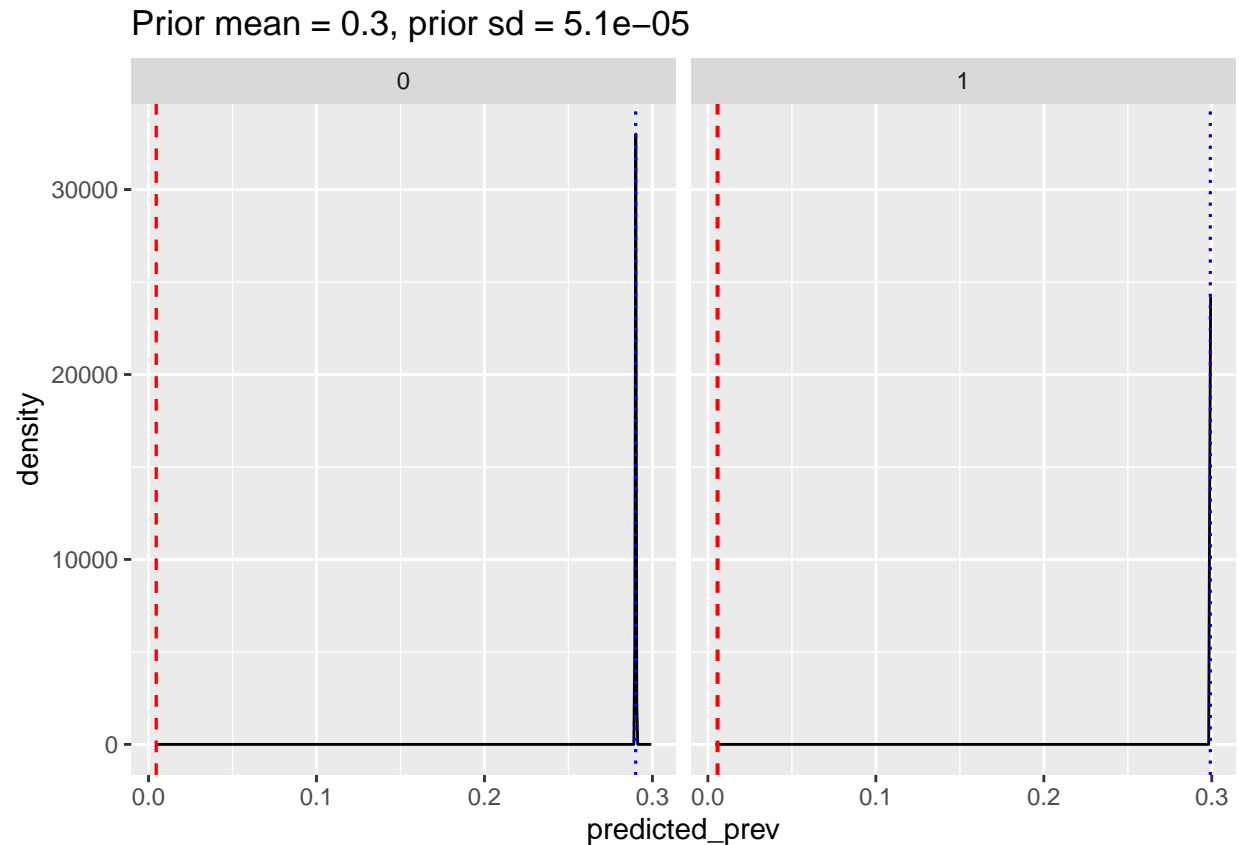


```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

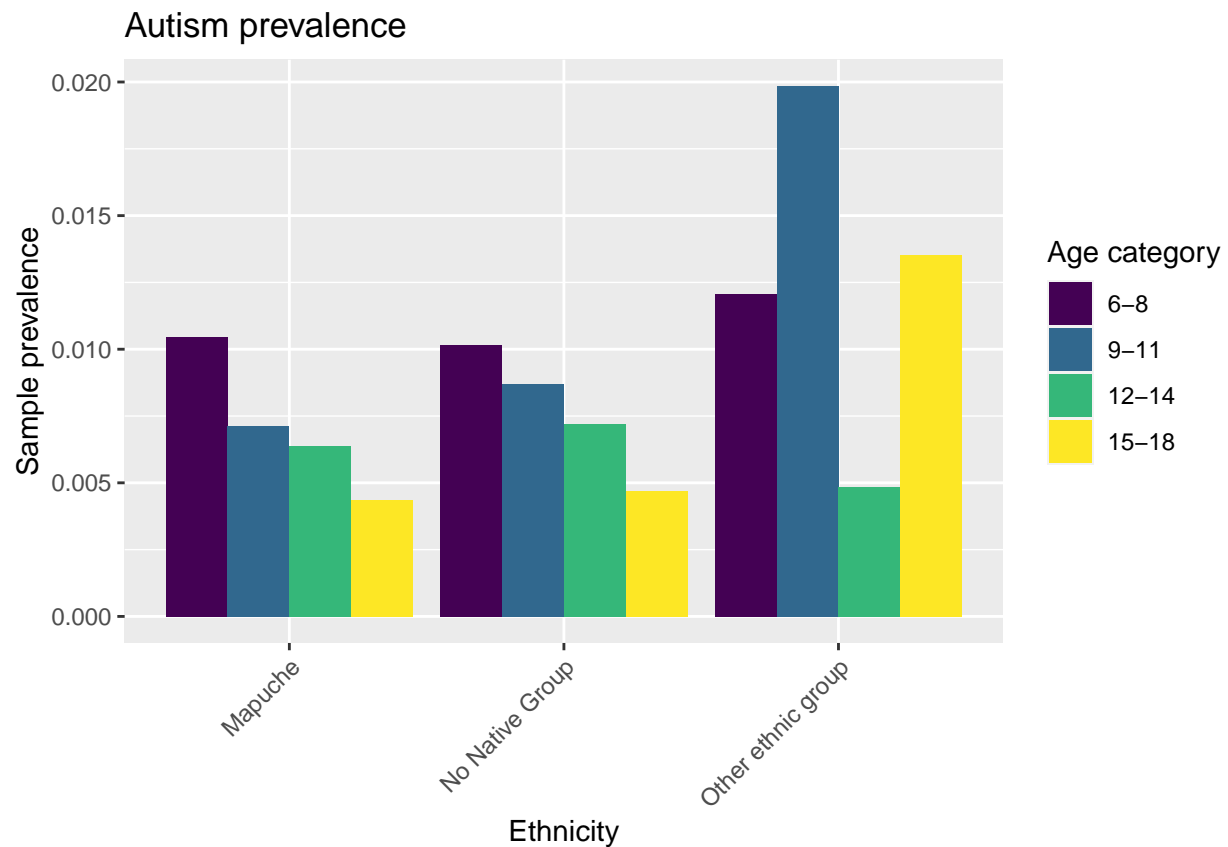
Random effect on ethnicity

Only use regions with large Mapuche populations

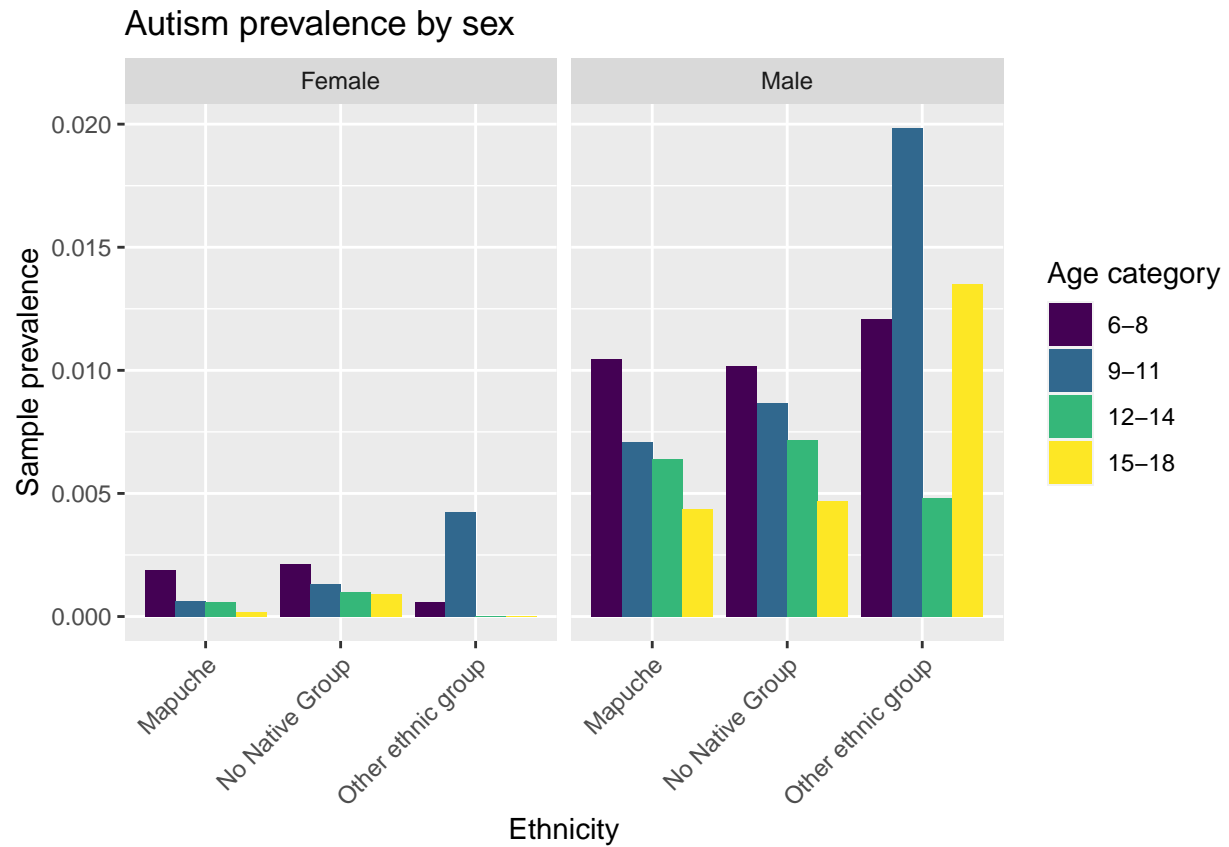
```
aut_prev_ethnic <- chile_bayes_aut %>%
  filter(school_region_name_abr %in% c("ARAUC", "BBIO", "LAGOS", "RIOS", "RM")) %>%
  get_grouped_prev(stdpop = chile_stdpop,
    grouping_vars = c("ethnic_2_group", "age_june30", "age_cat_name", "sex", "sex_desc",

## `summarise()` has grouped output by 'ethnic_2_group', 'age_june30',
## 'age_cat_name', 'sex', 'sex_desc'. You can override using the `.groups`
## argument.

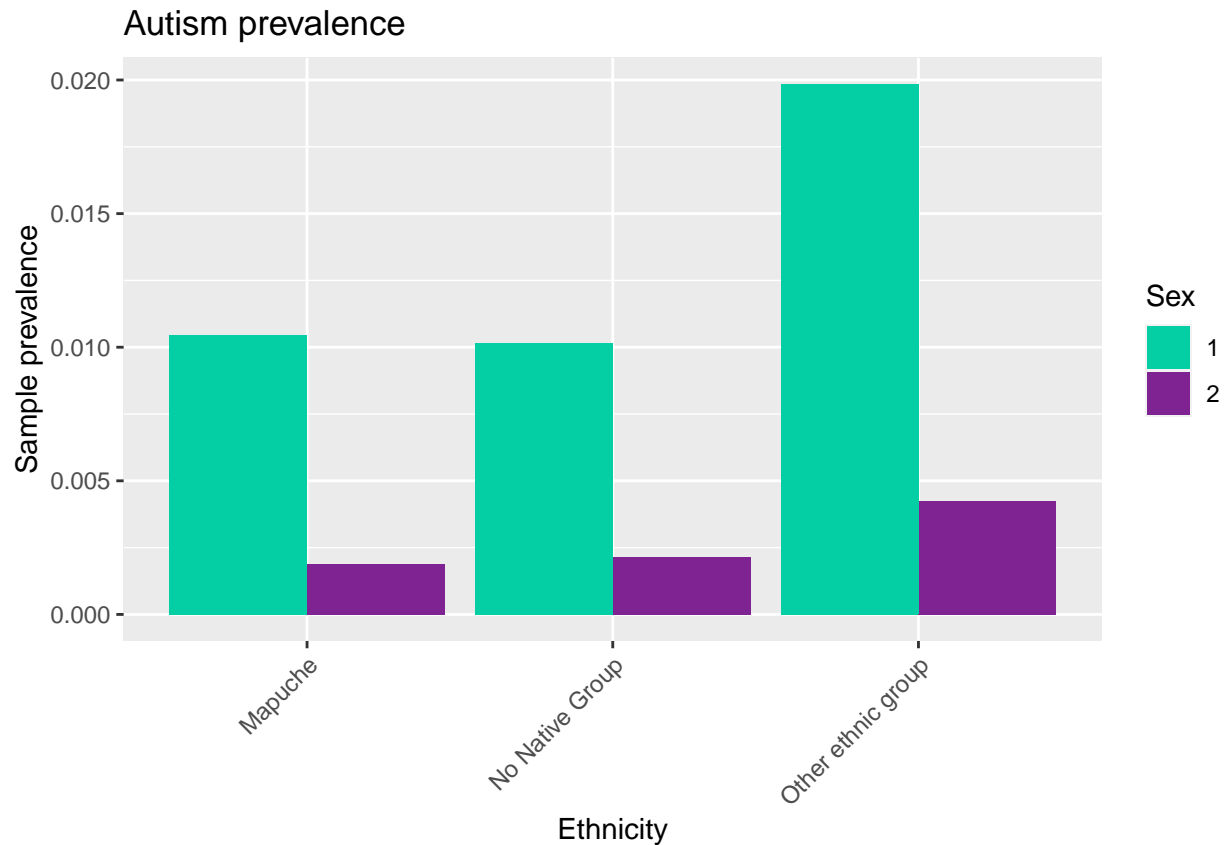
ggplot(data = aut_prev_ethnic) +
  geom_col(aes(x = as.factor(ethnic_2_group), y = sample_prevalence, group = age_cat_name, fill = age_c
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
    x = "Ethnicity",
    y = "Sample prevalence",
    fill = "Age category")
```



```
ggplot(data = aut_prev_ethnic) +
  geom_col(aes(x = as.factor(ethnic_2_group), y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
       x = "Ethnicity",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_ethnic) +
  geom_col(aes(x = as.factor(ethnic_2_group), y = sample_prevalence, group = sex, fill = as.factor(sex))) +
  scale_fill_manual(values = c("#03CEA4", "#802392")) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Ethnicity",
       y = "Sample prevalence",
       fill = "Sex")
```



```

aut_prev_ethnic_adj <- get_adjusted_prev(aut_prev_ethnic, grouping_vars = "ethnic_2_group")

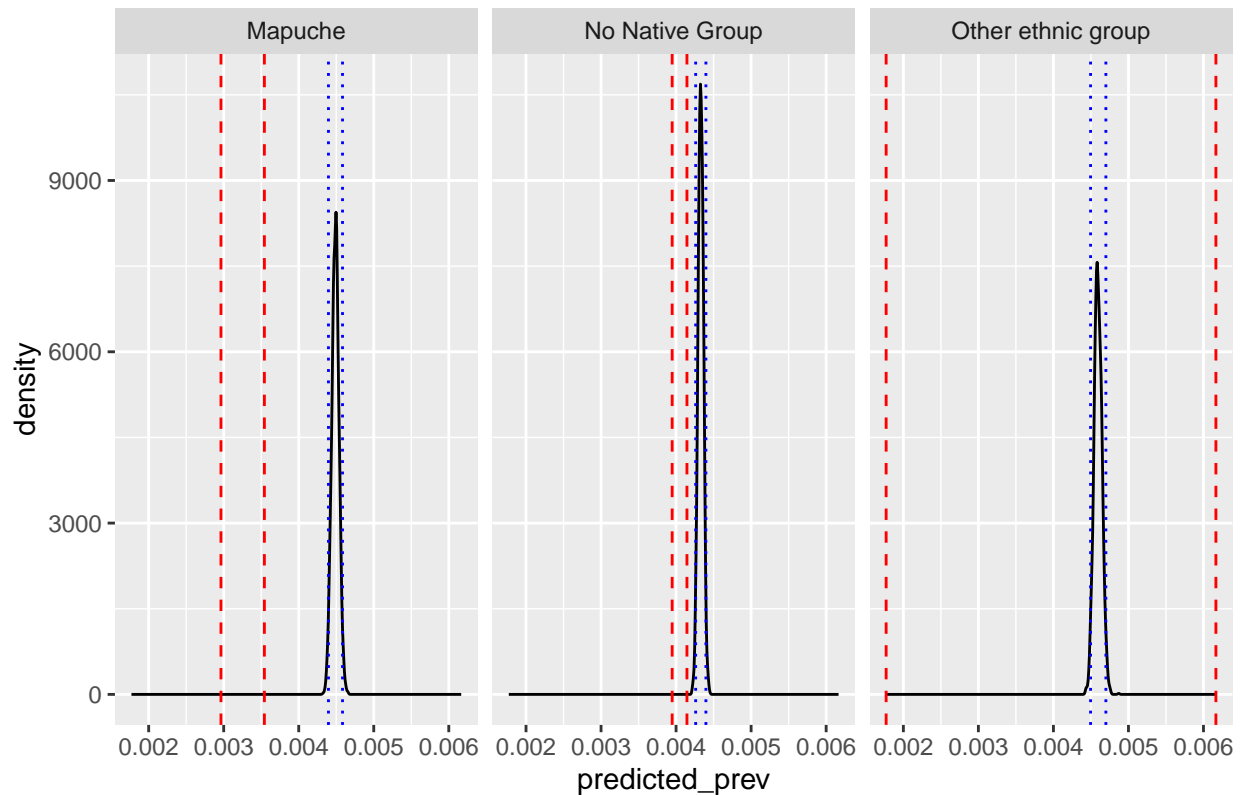
aut_prev_ethnic_post <- do_jags_rand_model(x = aut_prev_ethnic_adj,
  feat = "ethnic_2_group",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("ethnic_2_group" = "Feat_names")

plot_post_density(aut_prev_ethnic_post, aut_prev_ethnic_adj, feat = "ethnic_2_group", theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior, pars = pars, convergence_checks = FALSE)

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.

```

Prior mean = 0.0046, prior sd = 5.1e-05



Predictions for higher population prevalence - increase prior mean

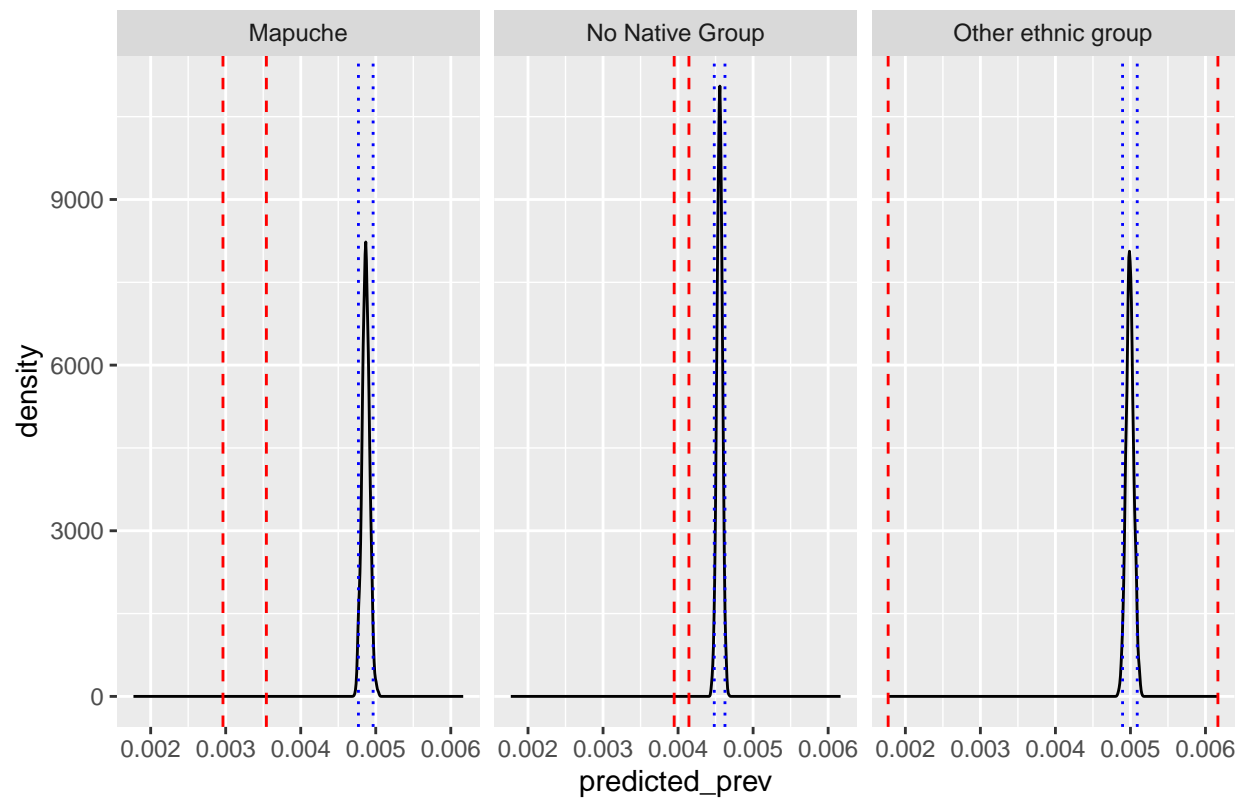
```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_ethnic_post <- do_jags_rand_model(x = aut_prev_ethnic_adj,
    feat = "ethnic_2_group",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("ethnic_2_group" = "Feat_names")

  plot_post_density(aut_prev_ethnic_post,
    aut_prev_ethnic_adj,
    feat = "ethnic_2_group",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

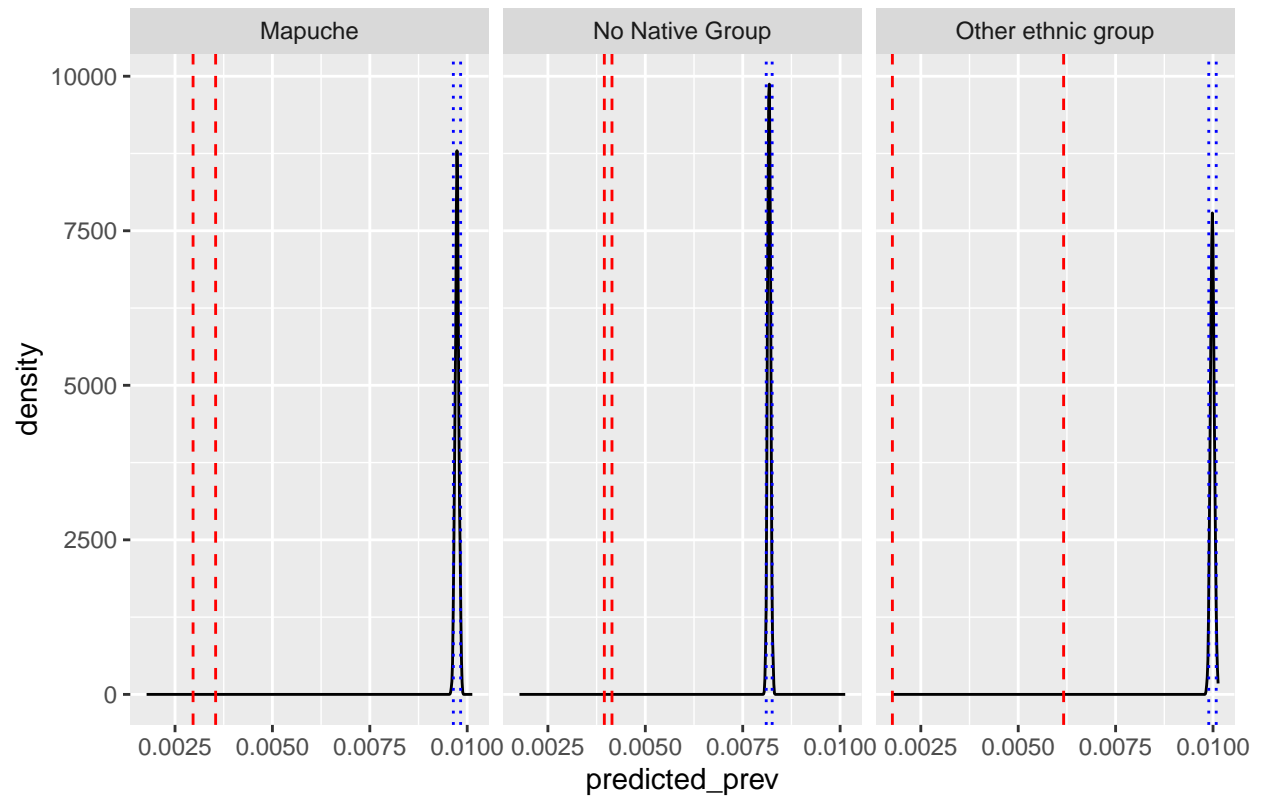
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



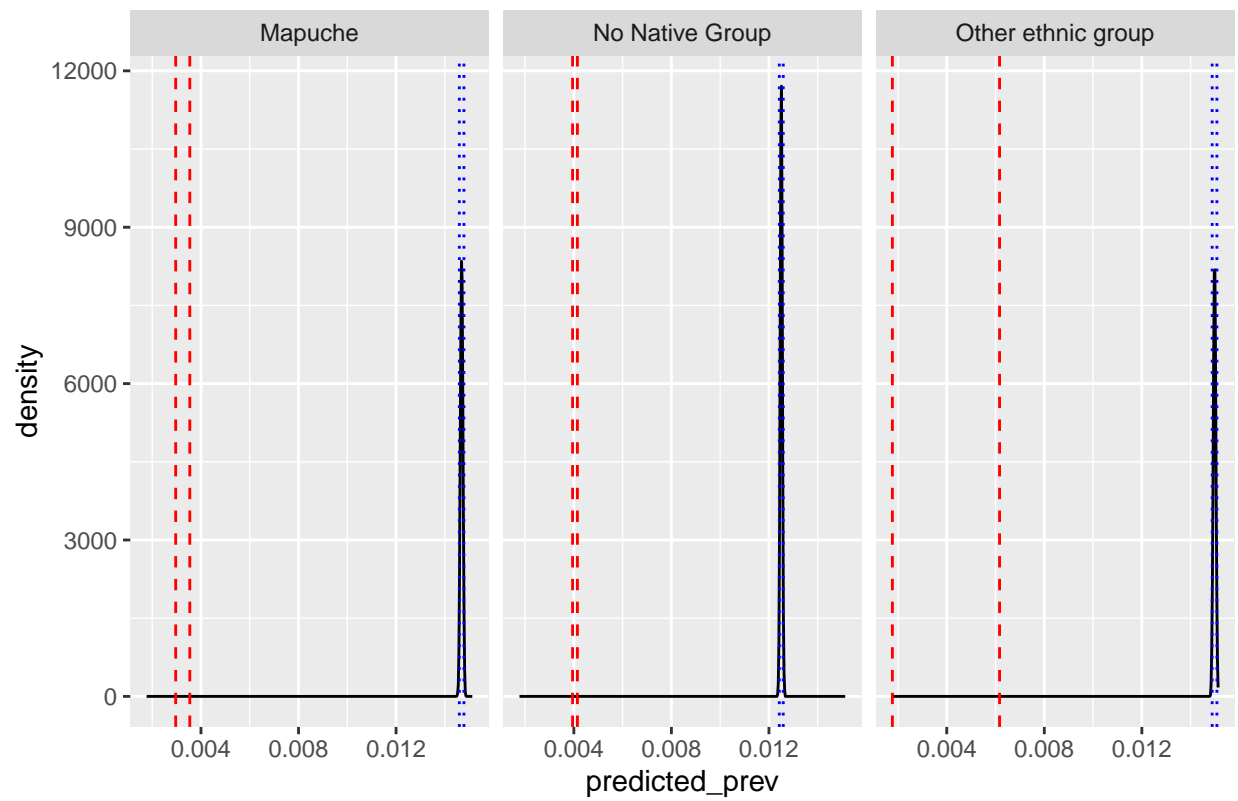
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



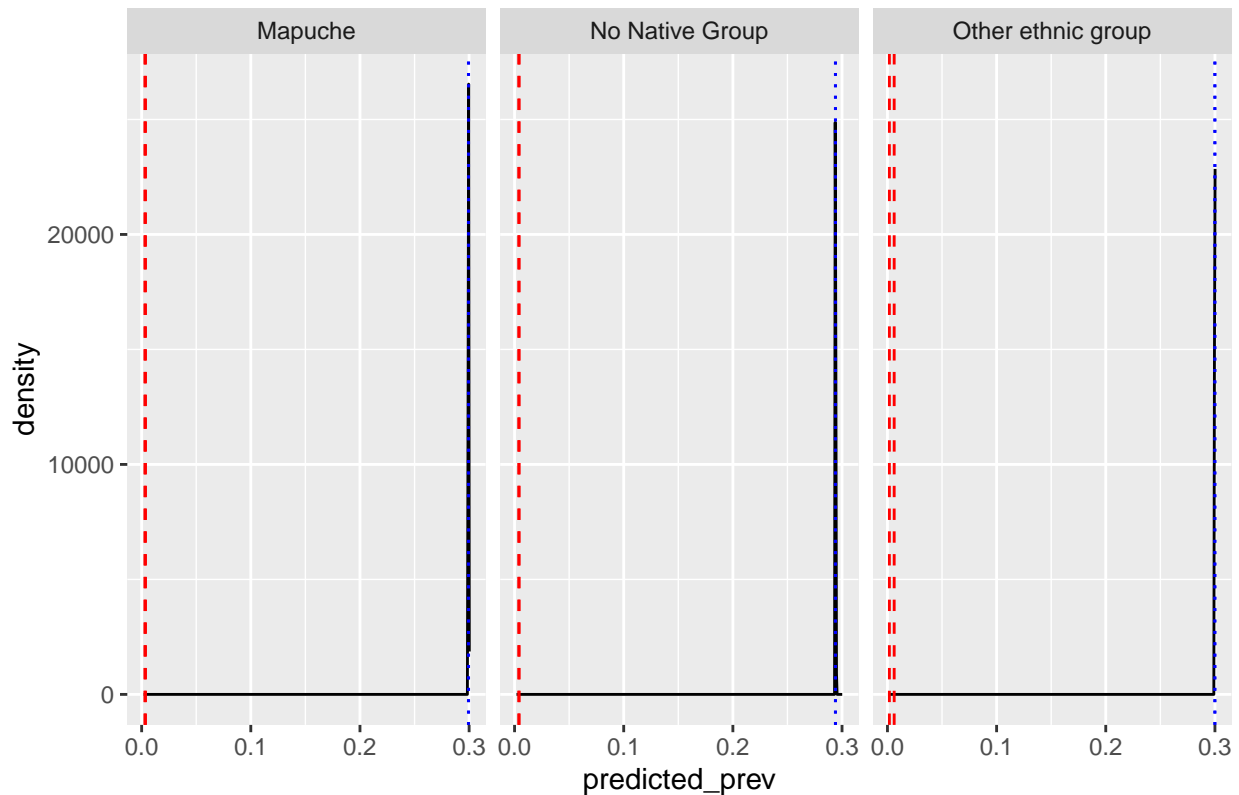
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05

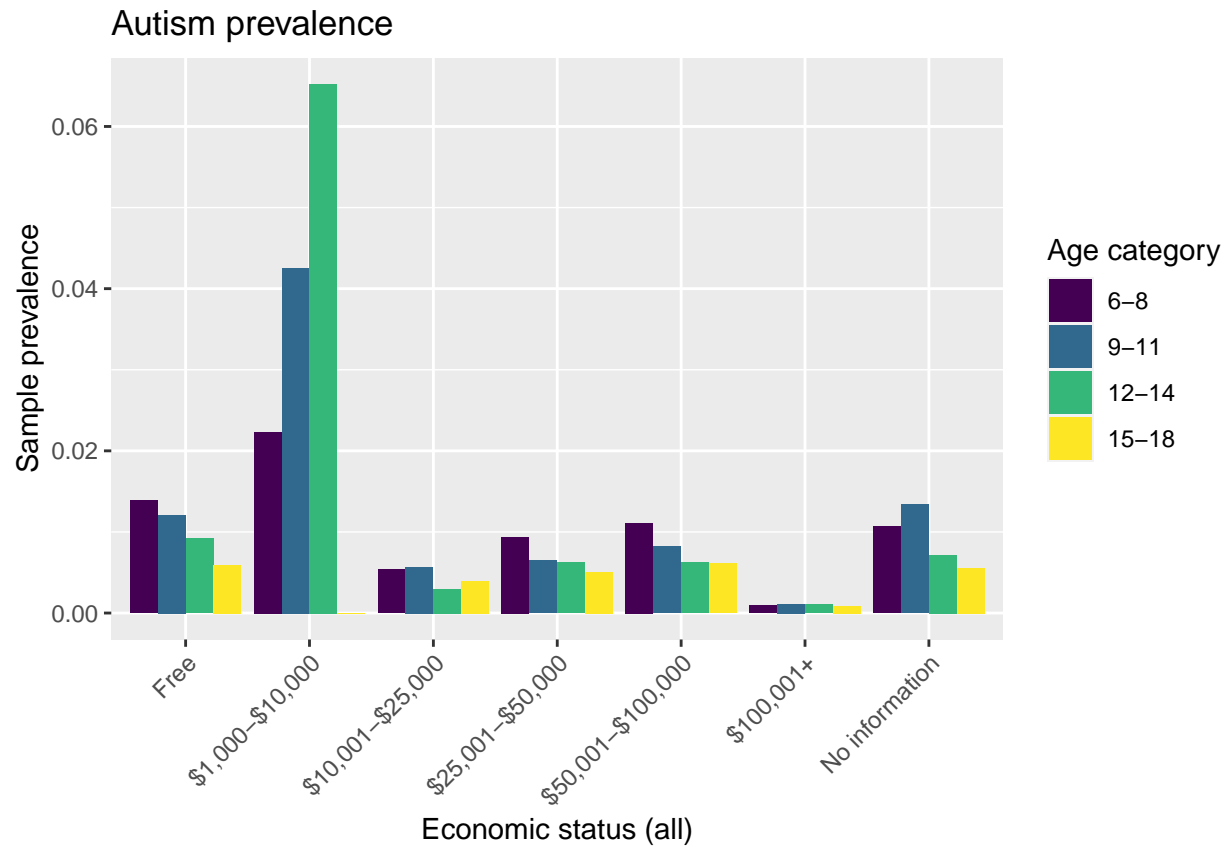


Random effect on economic status - all economic groups

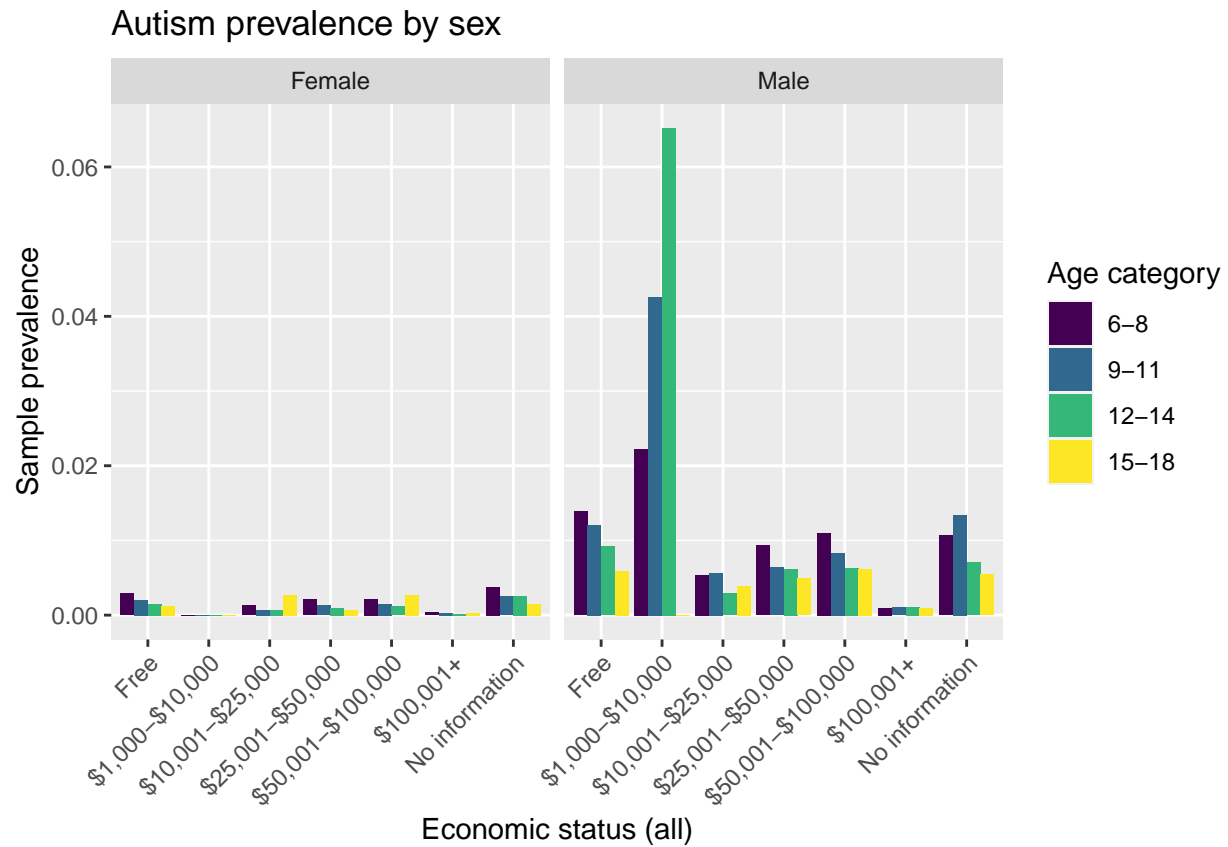
```
aut_prev_econ <- get_grouped_prev(x = chile_bayes_aut, stdpop = chile_stdpop,
                                   grouping_vars = c("school_fee", "school_fee_group", "age_june30", "age_cat_name"))

## `summarise()` has grouped output by 'school_fee', 'school_fee_group',
## 'age_june30', 'age_cat_name', 'sex', 'sex_desc'. You can override using the
## `.groups` argument.

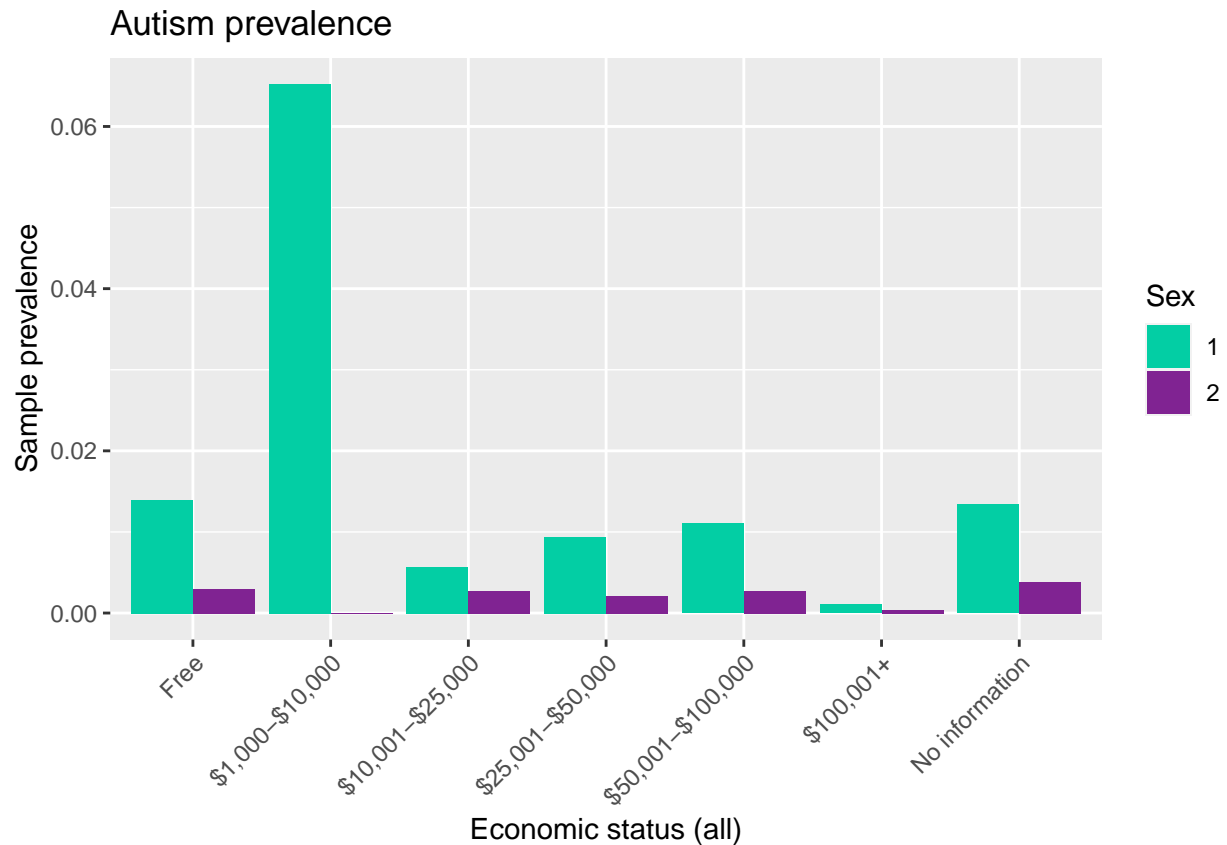
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee), y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Economic status (all)",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee), y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
       x = "Economic status (all)",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee), y = sample_prevalence, group = sex, fill = as.factor(sex)), p
  scale_fill_manual(values = c("#03CEA4", "#802392")) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Economic status (all)",
       y = "Sample prevalence",
       fill = "Sex")
```



```
#aut_prev_econ <- get_grouped_prev(x = chile_bayes_aut, stdpop = chile_stdpop,
#                                grouping_vars = c("school_fee", "age_june30", "sex", "autism"))

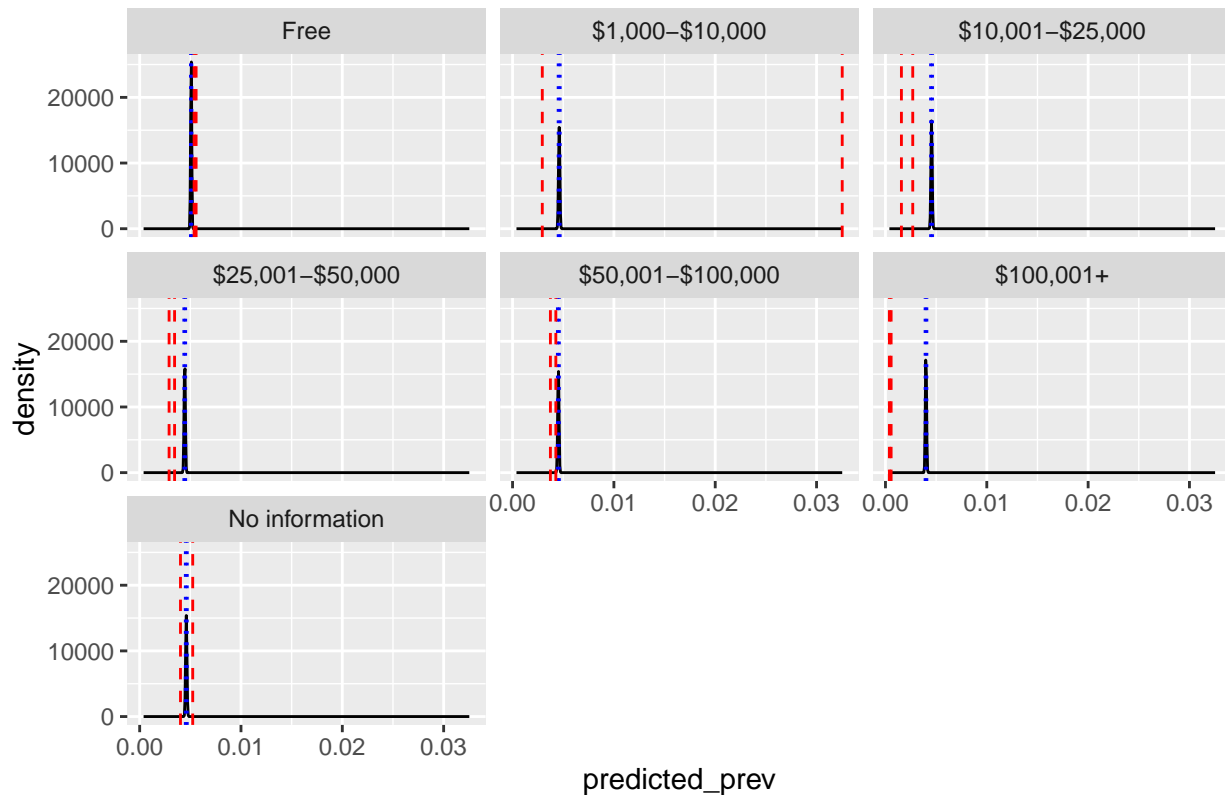
aut_prev_econ_adj <- get_adjusted_prev(aut_prev_econ, grouping_vars = "school_fee")

aut_prev_econ_post <- do_jags_rand_model(x = aut_prev_econ_adj,
                                       feat = "school_fee",
                                       model = rand_model,
                                       theta_mu = theta_mu_prior,
                                       theta_sigma = theta_sigma_prior,
                                       pars = pars,
                                       convergence_checks = FALSE) %>%
  rename("school_fee" = "Feat_names")

plot_post_density(aut_prev_econ_post, aut_prev_econ_adj, feat = "school_fee", theta_mu = theta_mu_prior)

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



Predictions for higher population prevalence - increase prior mean

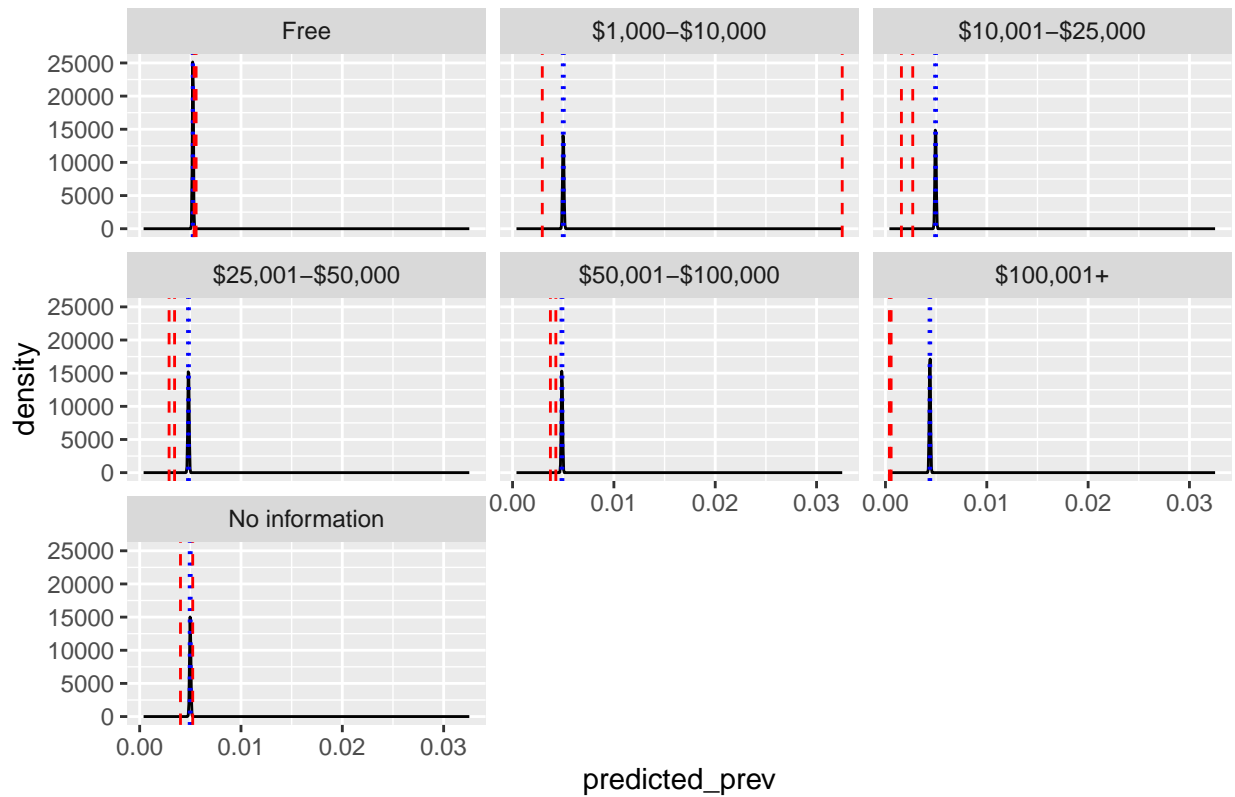
```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_econ_post <- do_jags_rand_model(x = aut_prev_econ_adj,
    feat = "school_fee",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("school_fee" = "Feat_names")

  plot_post_density(aut_prev_econ_post,
    aut_prev_econ_adj,
    feat = "school_fee",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

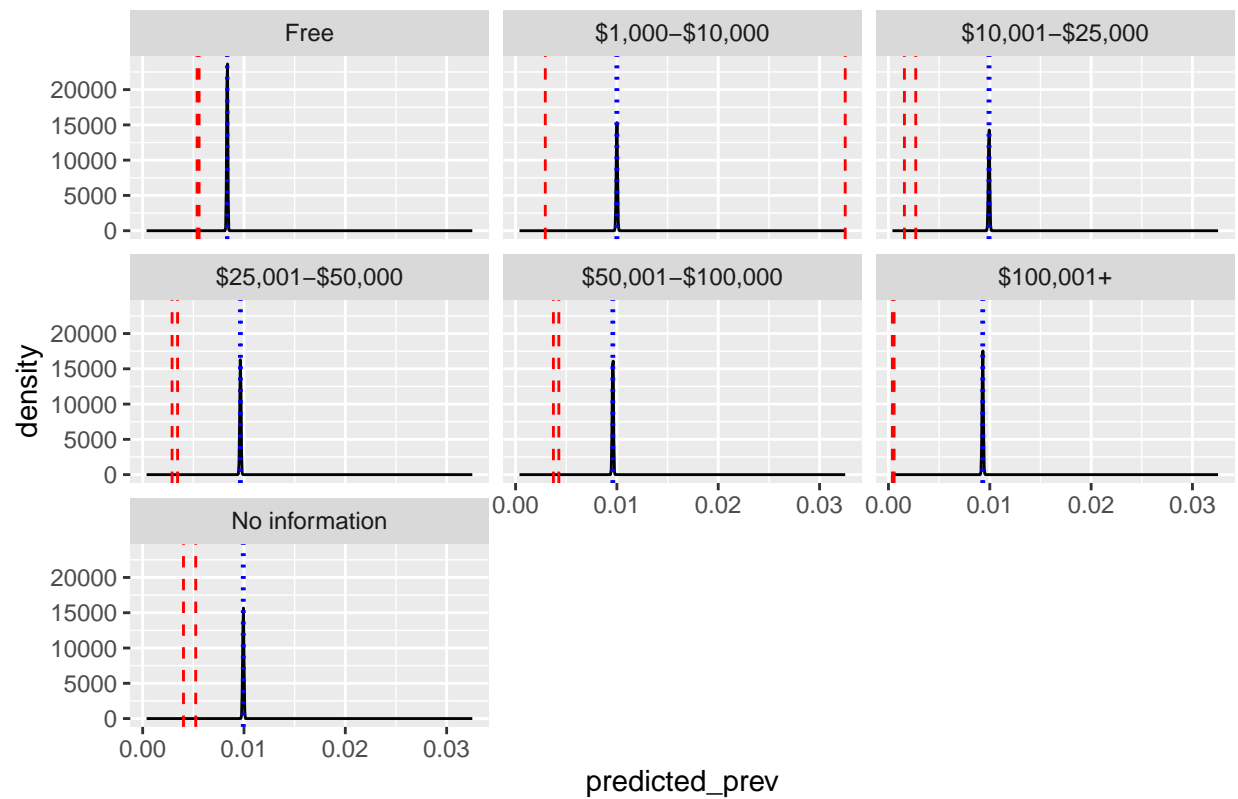
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



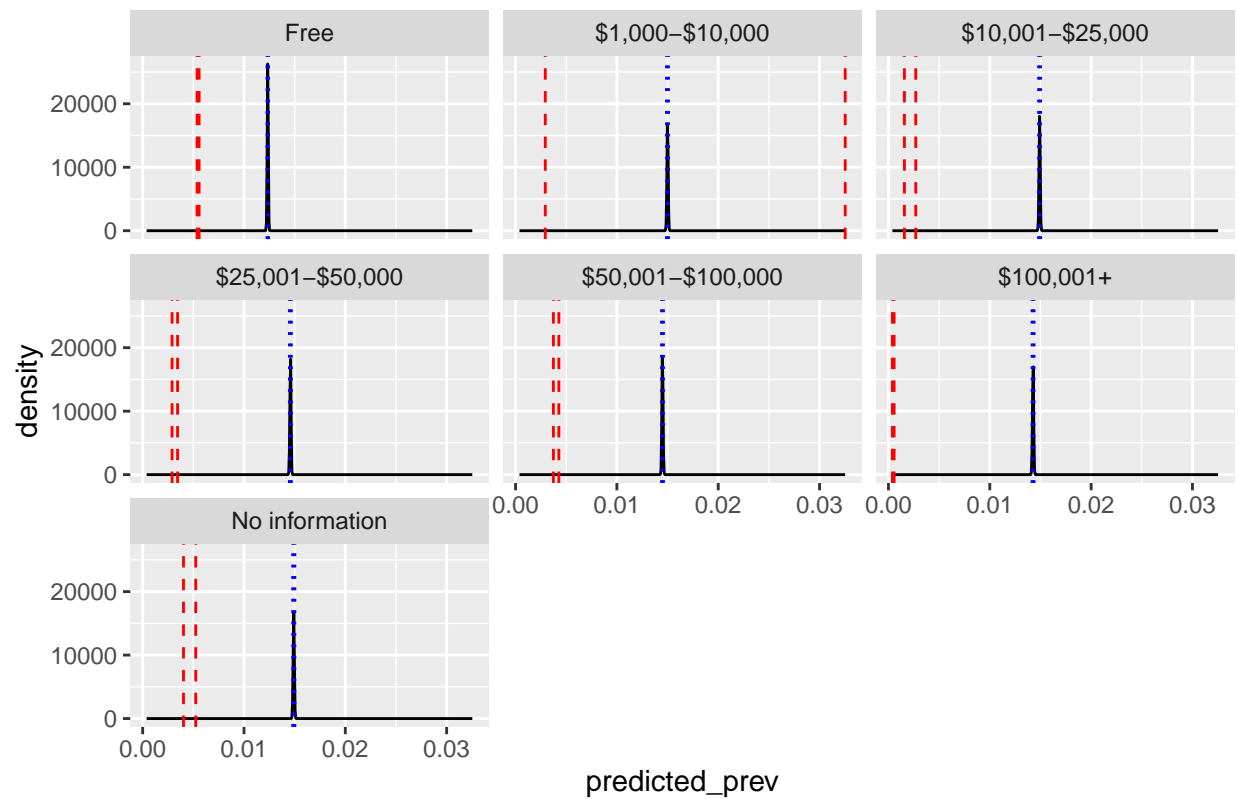
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



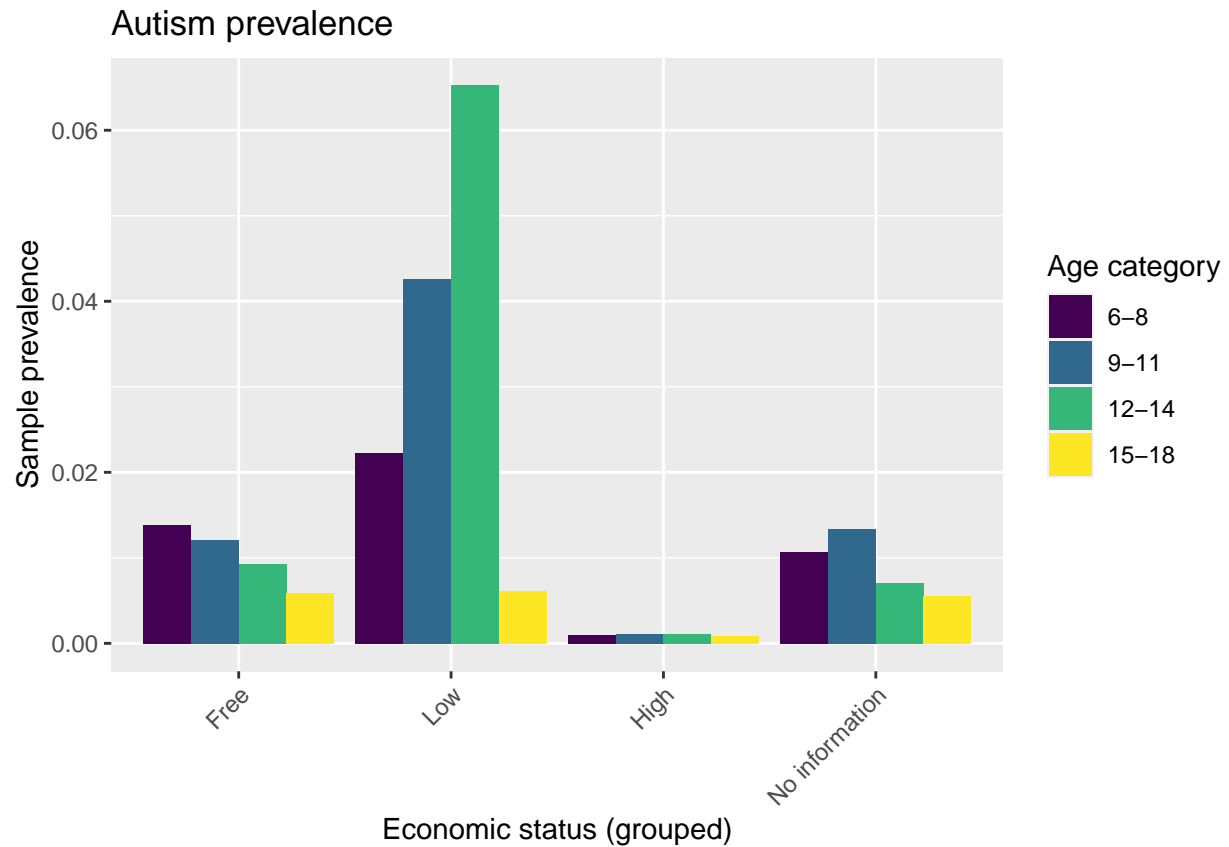
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05

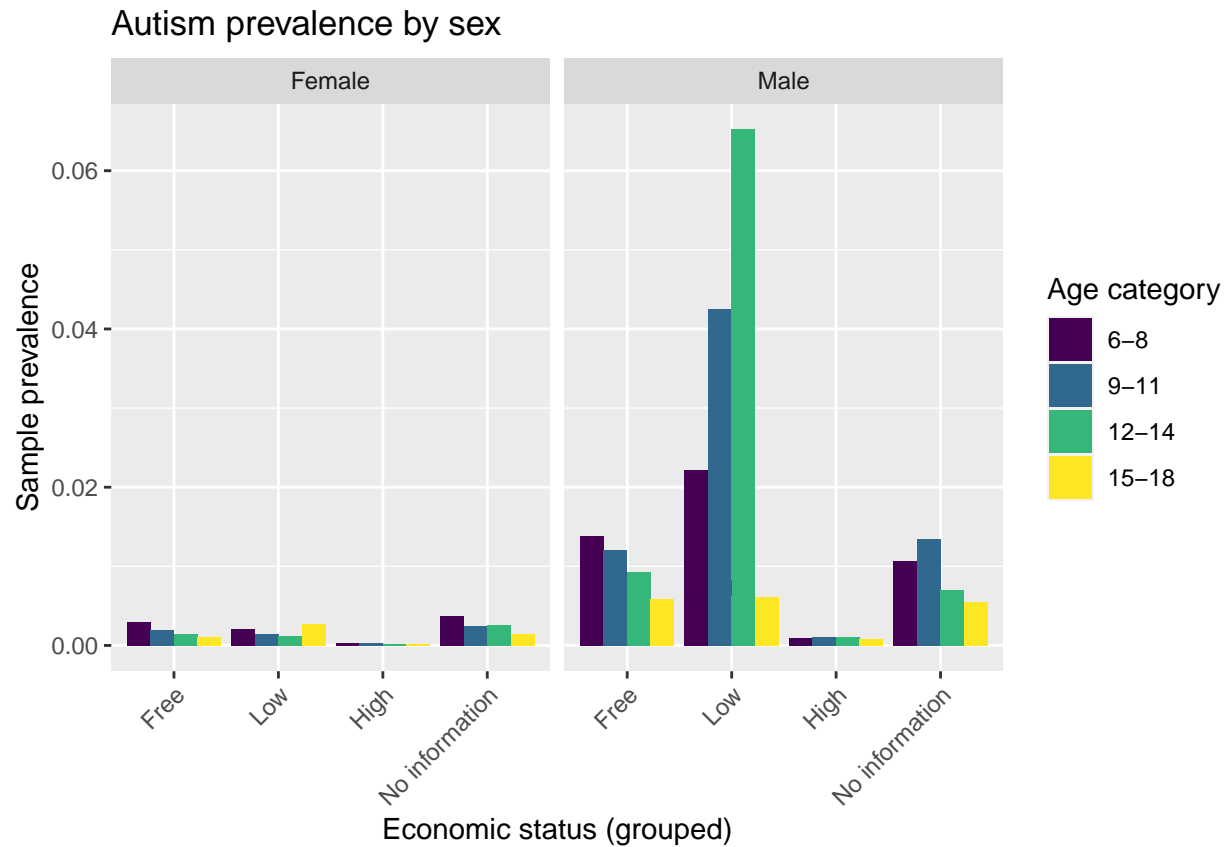


Random effect on economic status - free education, low status, high status and no information

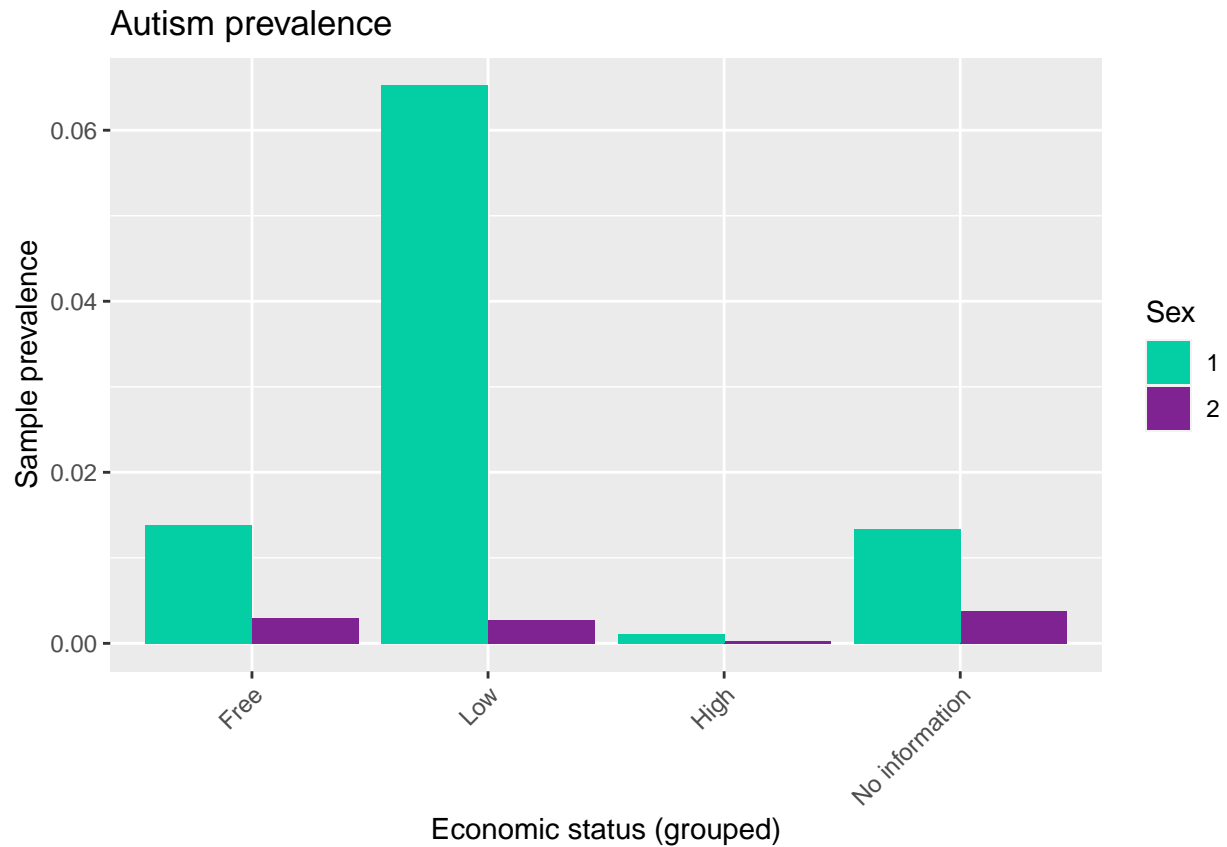
```
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee_group), y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
       x = "Economic status (grouped)",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee_group), y = sample_prevalence, group = age_cat_name, fill = age_cat_name)) +
  scale_fill_viridis_d() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  facet_wrap(~sex_desc) +
  labs(title = "Autism prevalence by sex",
       x = "Economic status (grouped)",
       y = "Sample prevalence",
       fill = "Age category")
```



```
ggplot(data = aut_prev_econ) +
  geom_col(aes(x = as.factor(school_fee_group), y = sample_prevalence, group = sex, fill = as.factor(se
  scale_fill_manual(values = c("#03CEA4", "#802392")) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  labs(title = "Autism prevalence",
        x = "Economic status (grouped)",
        y = "Sample prevalence",
        fill = "Sex")
```



```

aut_prev_econG_adj <- get_adjusted_prev(aut_prev_econ, grouping_vars = "school_fee_group")

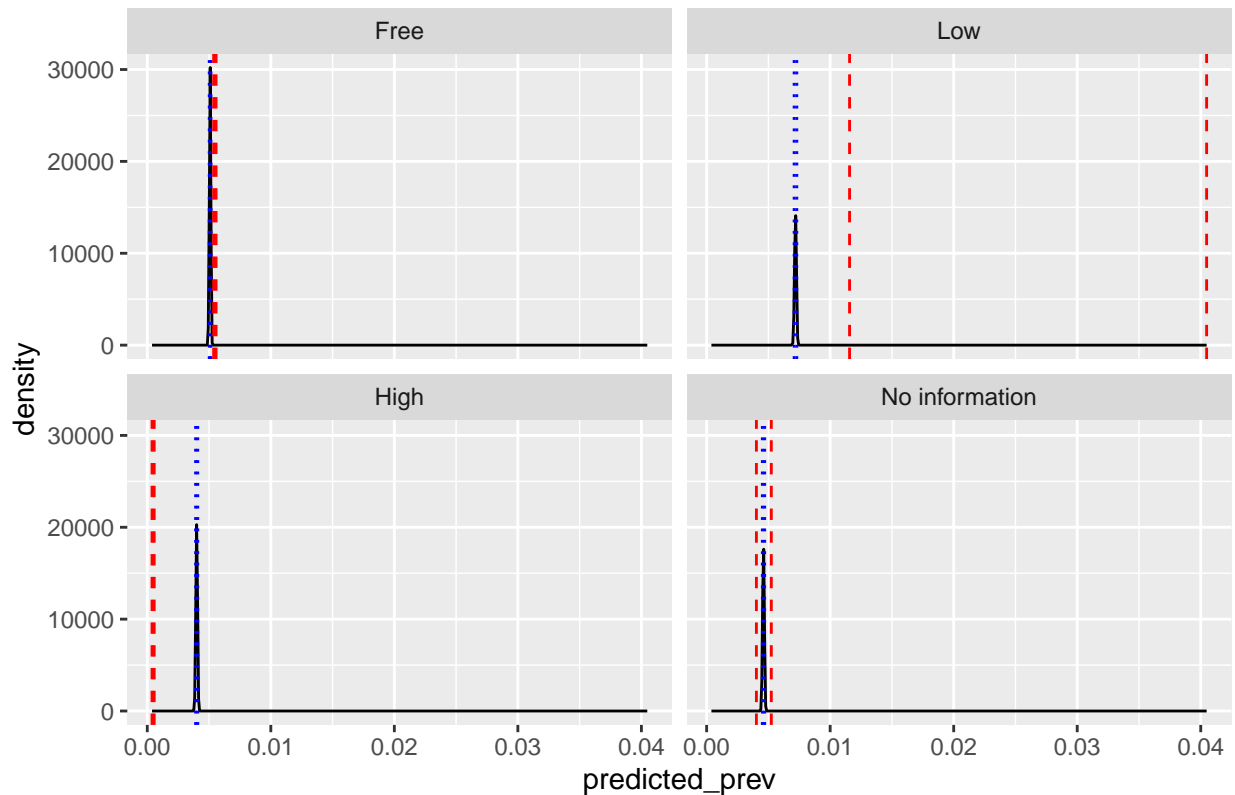
aut_prev_econG_post <- do_jags_rand_model(x = aut_prev_econG_adj,
  feat = "school_fee_group",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_fee_group" = "Feat_names")

plot_post_density(aut_prev_econG_post, aut_prev_econG_adj, feat = "school_fee_group", theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior, pars = pars, convergence_checks = FALSE)

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.

```

Prior mean = 0.0046, prior sd = 5.1e-05



Predictions for higher population prevalence - increase prior mean

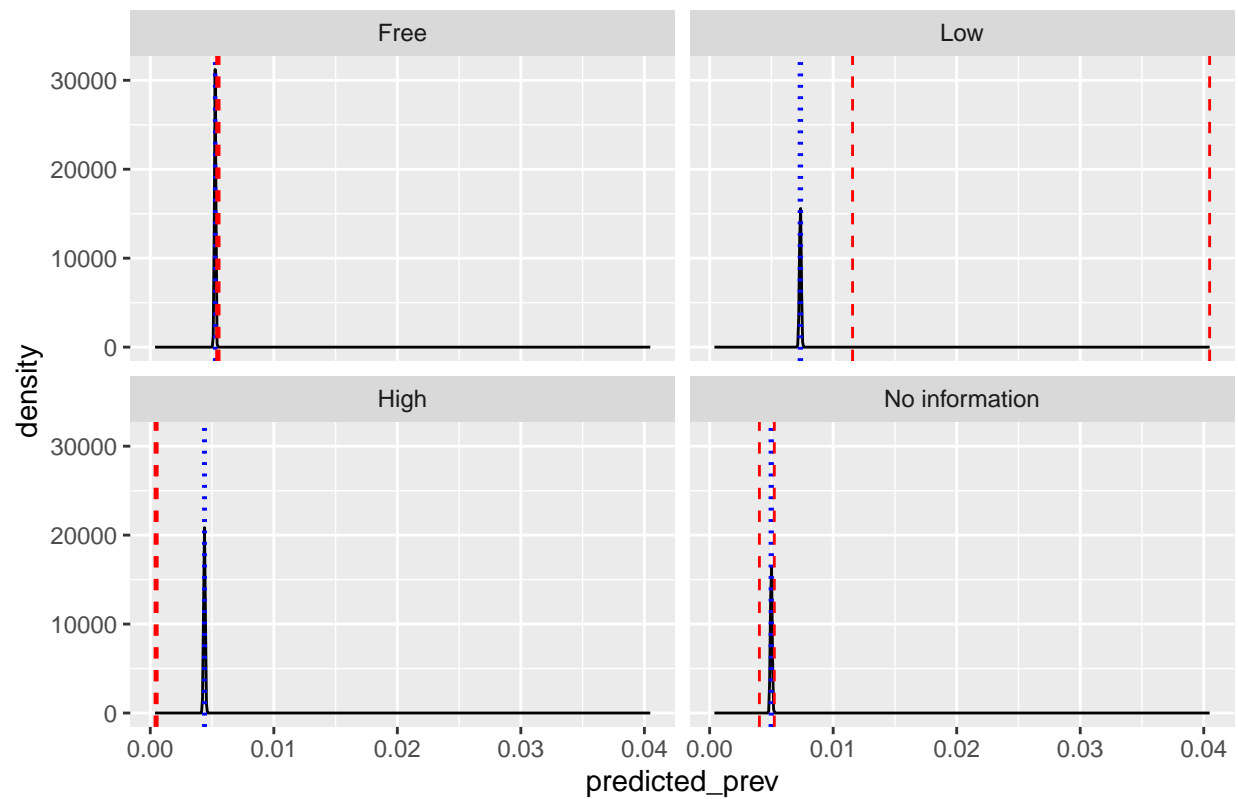
```
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_econG_post <- do_jags_rand_model(x = aut_prev_econG_adj,
    feat = "school_fee_group",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("school_fee_group" = "Feat_names")

  plot_post_density(aut_prev_econG_post,
    aut_prev_econG_adj,
    feat = "school_fee_group",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

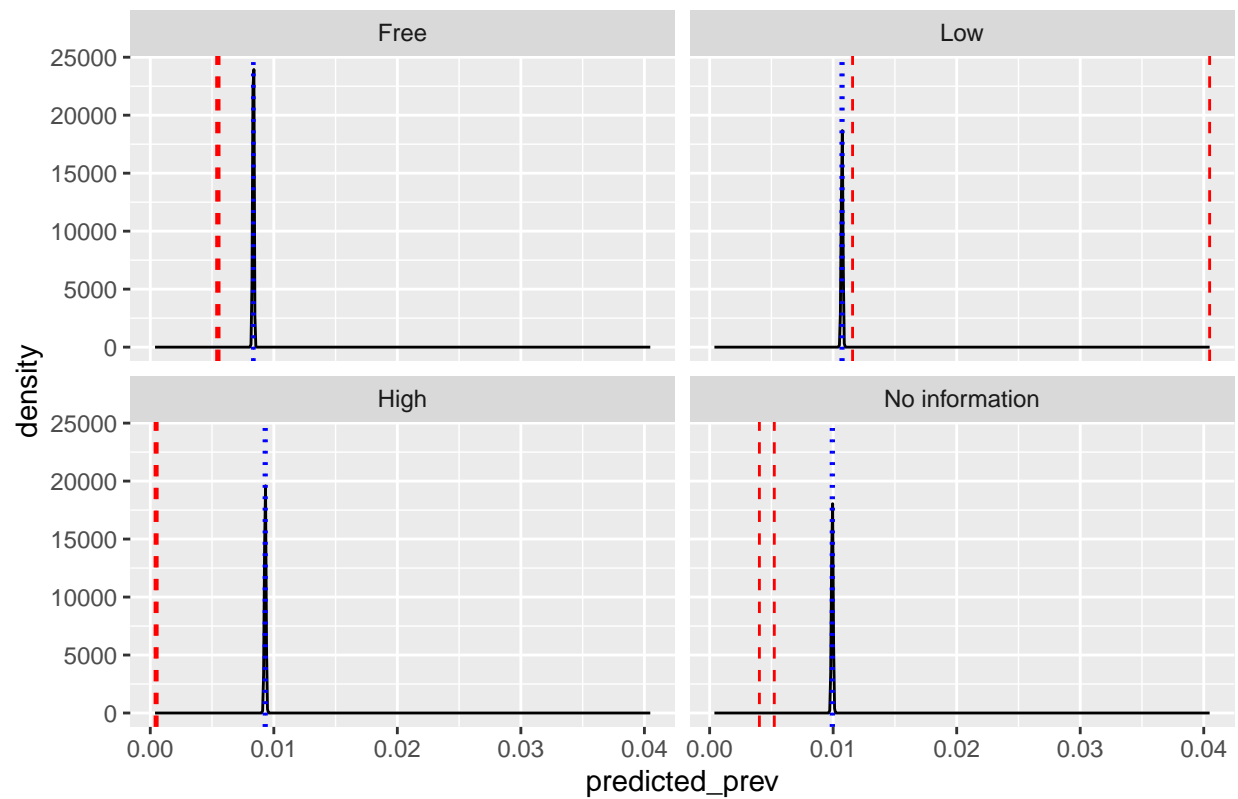
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



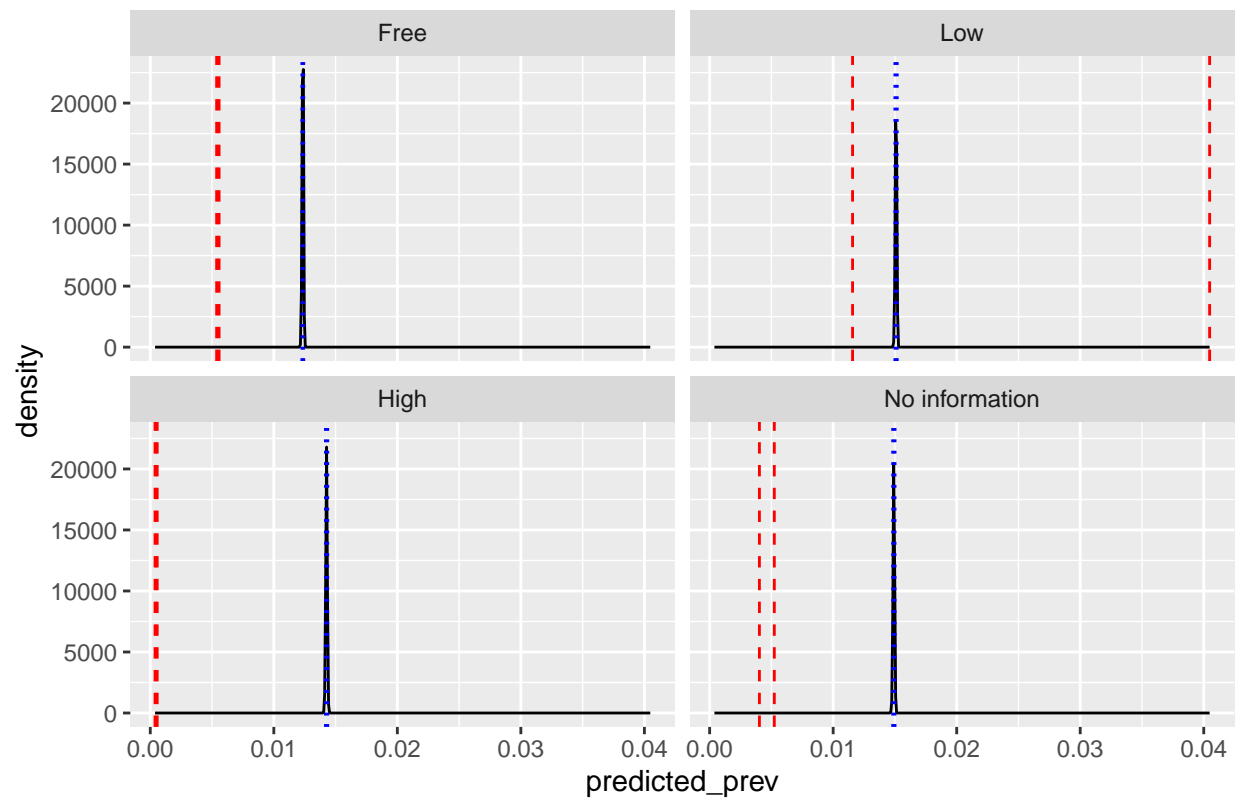
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



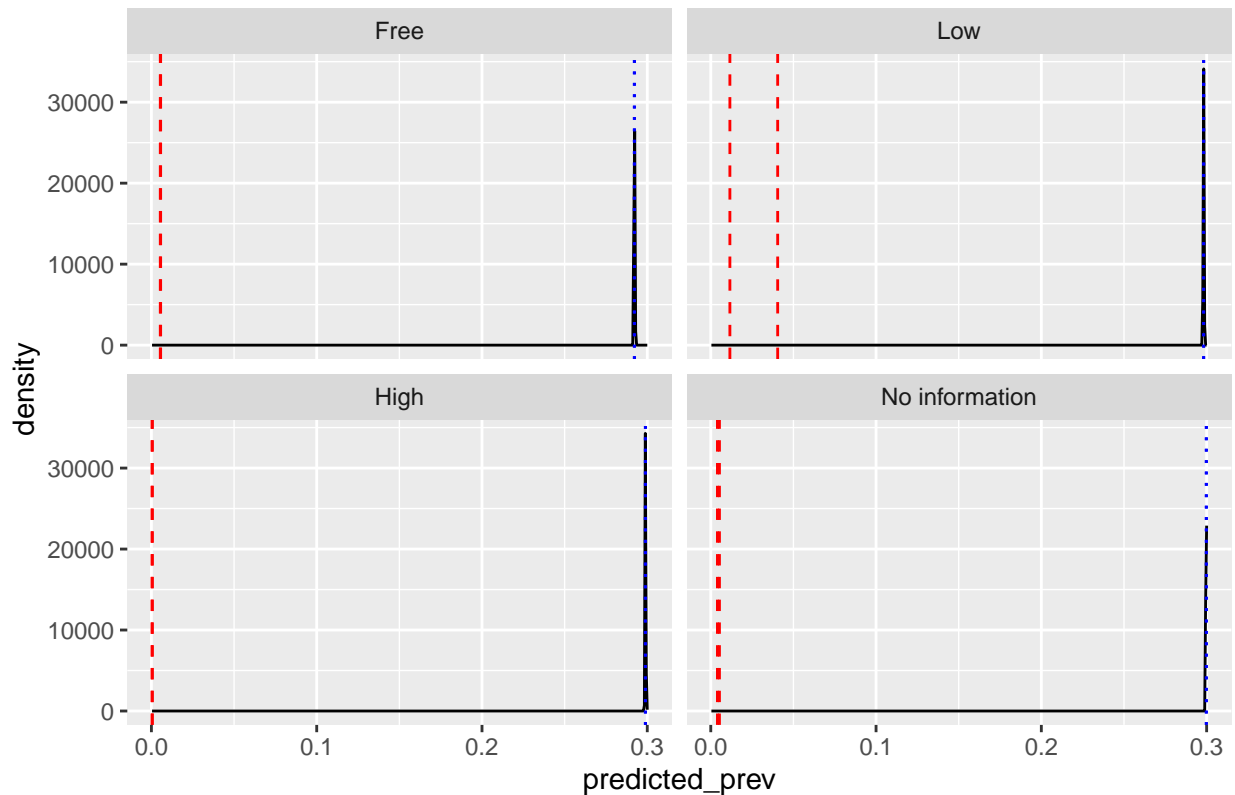
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05



Random effect on sex

```
chile_stdpop_f <- chile_stdpop %>%
  filter(sex == 2) %>%
  mutate(pop_prop = std_pop / sum(std_pop))
chile_stdpop_m <- chile_stdpop %>%
  filter(sex == 1) %>%
  mutate(pop_prop = std_pop / sum(std_pop))

aut_prev_f <- chile_bayes_aut %>%
  filter(sex == 2) %>%
  get_grouped_prev(stdpop = chile_stdpop_f, grouping_vars = c("age_june30", "sex", "autism"))

## `summarise()` has grouped output by 'age_june30', 'sex'. You can override using
## the `.groups` argument.

aut_prev_adj_f <- get_adjusted_prev(aut_prev_f, grouping_vars = c()) %>% mutate(sex = 2)

aut_prev_m <- chile_bayes_aut %>%
  filter(sex == 1) %>%
  get_grouped_prev(stdpop = chile_stdpop_m, grouping_vars = c("age_june30", "sex", "autism"))

## `summarise()` has grouped output by 'age_june30', 'sex'. You can override using
## the `.groups` argument.

aut_prev_adj_m <- get_adjusted_prev(aut_prev_m, grouping_vars = c()) %>% mutate(sex = 1)
```

```

aut_prev_sex_adj <- rbind(aut_prev_adj_m, aut_prev_adj_f)
# have to put m first because 1 comes before 2 and otherwise will mess up naming in do_jags_rand_model

aut_prev_sex_post <- do_jags_rand_model(x = aut_prev_sex_adj,
  feat = "sex",
  model = rand_model,
  theta_mu = theta_mu_prior,
  theta_sigma = theta_sigma_prior,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("sex" = "Feat_names")

plot_post_density(aut_prev_sex_post, aut_prev_sex_adj, feat = "sex", theta_mu = theta_mu_prior, theta_s

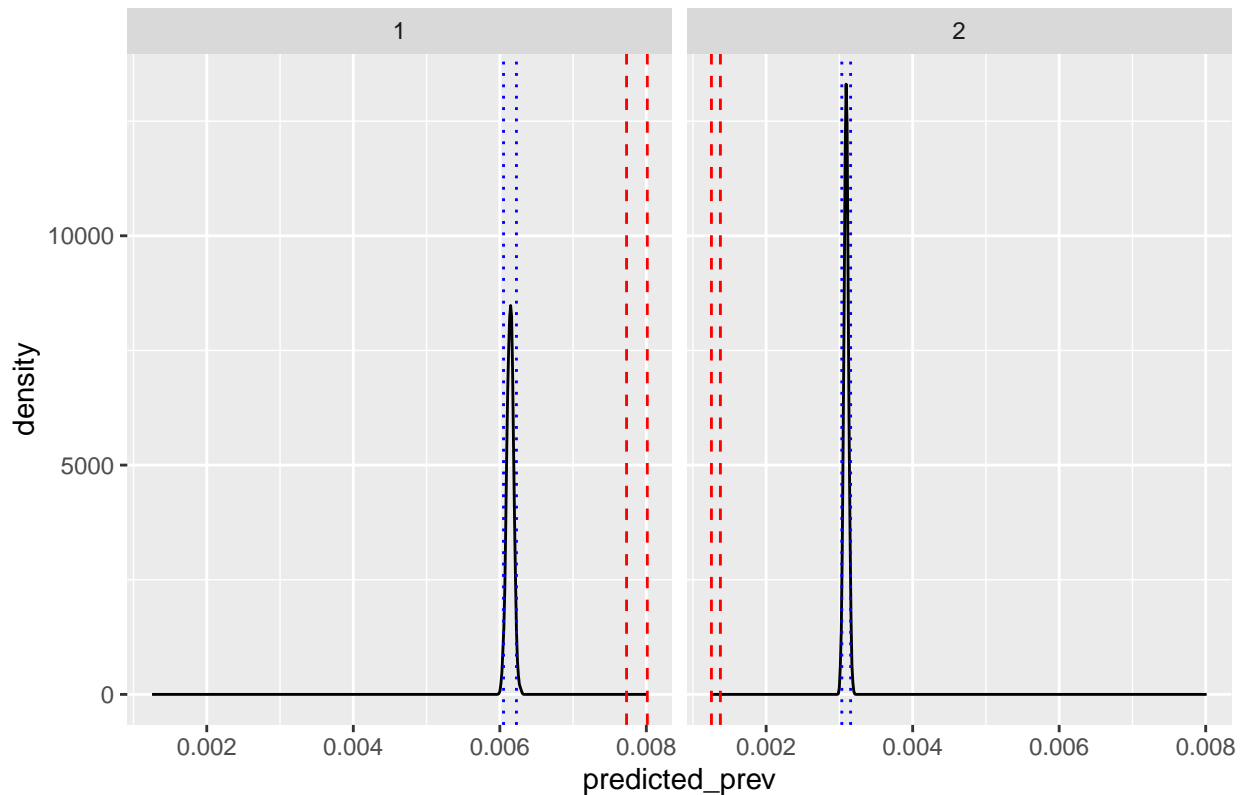
## Warning: Combining variables of class <factor> and <numeric> was deprecated in ggplot2
## 3.4.0.
## i Please ensure your variables are compatible before plotting (location:
##   `combine_vars()`)
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

## Warning: Combining variables of class <numeric> and <factor> was deprecated in ggplot2
## 3.4.0.
## i Please ensure your variables are compatible before plotting (location:
##   `combine_vars()`)
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.

```

Prior mean = 0.0046, prior sd = 5.1e-05



Random effect on region, sexes separate

Need sex specific priors - use age-adjusted prevalence from above

```
theta_mu_prior_f <- aut_prev_adj_f$adjusted_rate
theta_sigma_prior_f <- sqrt(aut_prev_adj_f$var)
```

```
theta_mu_prior_m <- aut_prev_adj_m$adjusted_rate
theta_sigma_prior_m <- sqrt(aut_prev_adj_m$var)
```

Females

```
aut_prev_region_f <- chile_bayes_aut %>%
```

```
  filter(sex == 2) %>%
```

```
  get_grouped_prev(stdpop = chile_stdpop_f, grouping_vars = c("school_region_name_abr", "age_june30", "sex"))
```

```
## `summarise()` has grouped output by 'school_region_name_abr', 'age_june30',
```

```
## 'sex'. You can override using the `.groups` argument.
```

```
aut_prev_region_adj_f <- get_adjusted_prev(aut_prev_region_f, grouping_vars = "school_region_name_abr")
```

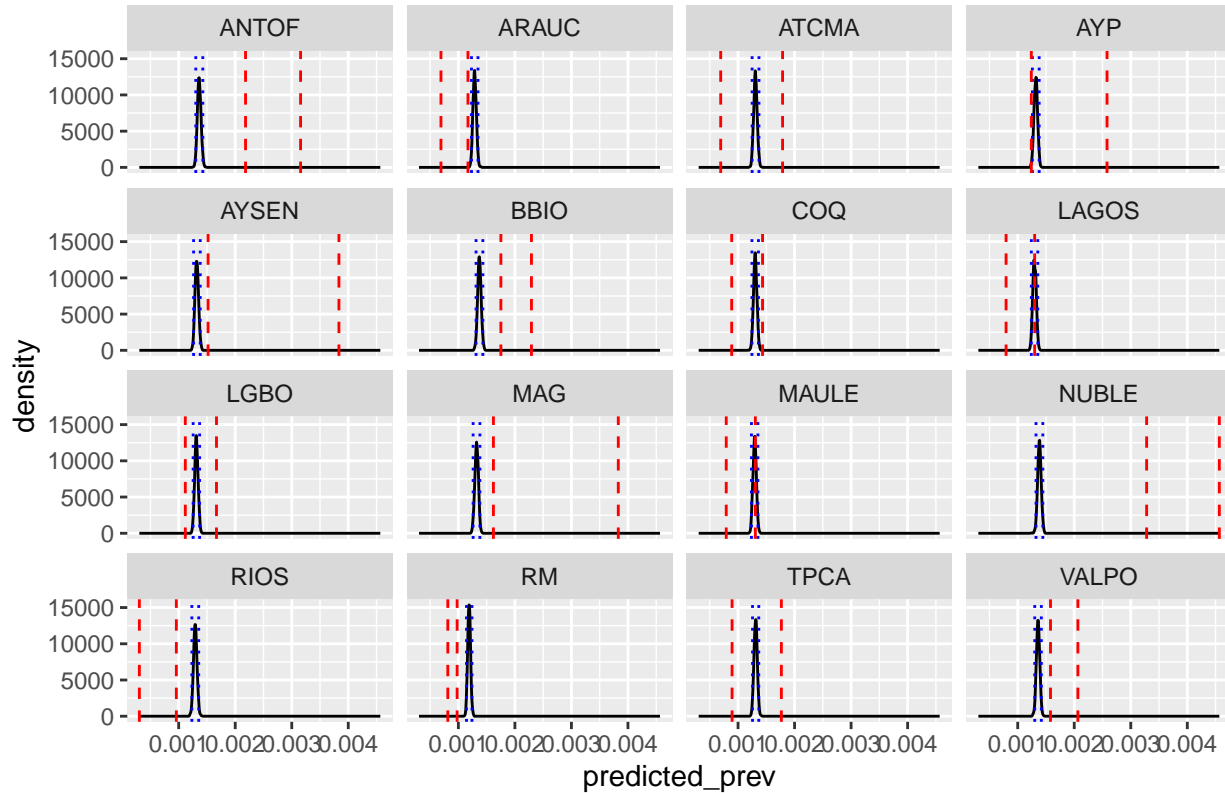
```
aut_prev_region_post_f <- do_jags_rand_model(x = aut_prev_region_adj_f,
  feat = "school_region_name_abr",
  model = rand_model,
  theta_mu = theta_mu_prior_f,
  theta_sigma = theta_sigma_prior_f,
  pars = pars,
  convergence_checks = FALSE) %>%
```

```
rename("school_region_name_abr" = "Feat_names")
```

```
plot_post_density(aut_prev_region_post_f, aut_prev_region_adj_f, feat = "school_region_name_abr", theta,
```

```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



```
# Males
aut_prev_region_m <- chile_bayes_aut %>%
  filter(sex == 1) %>%
  get_grouped_prev(stdpop = chile_stdpop_m, grouping_vars = c("school_region_name_abr", "age_june30", "

## `summarise()` has grouped output by 'school_region_name_abr', 'age_june30',
## 'sex'. You can override using the `.groups` argument.

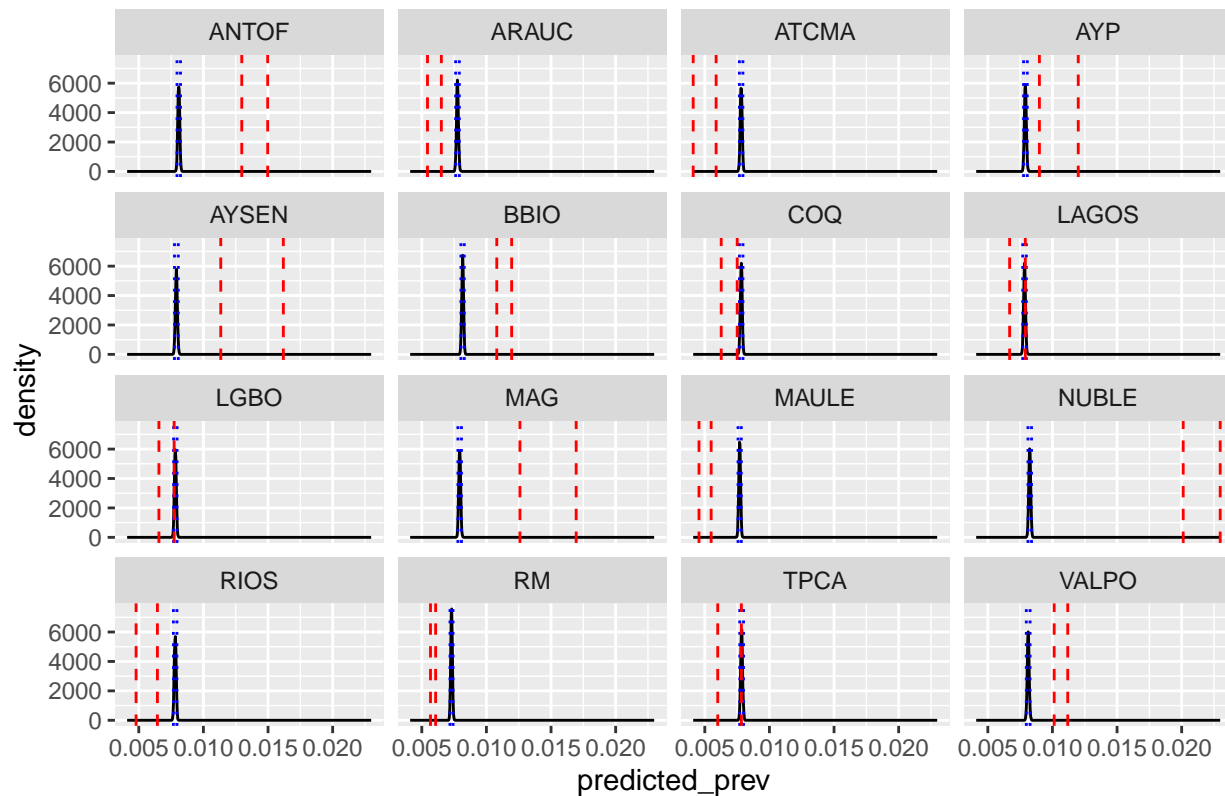
aut_prev_region_adj_m <- get_adjusted_prev(aut_prev_region_m, grouping_vars = "school_region_name_abr")

aut_prev_region_post_m <- do_jags_rand_model(x = aut_prev_region_adj_m,
  feat = "school_region_name_abr",
  model = rand_model,
  theta_mu = theta_mu_prior_m,
  theta_sigma = theta_sigma_prior_m,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_region_name_abr" = "Feat_names")

plot_post_density(aut_prev_region_post_m, aut_prev_region_adj_m, feat = "school_region_name_abr", theta,
```

```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



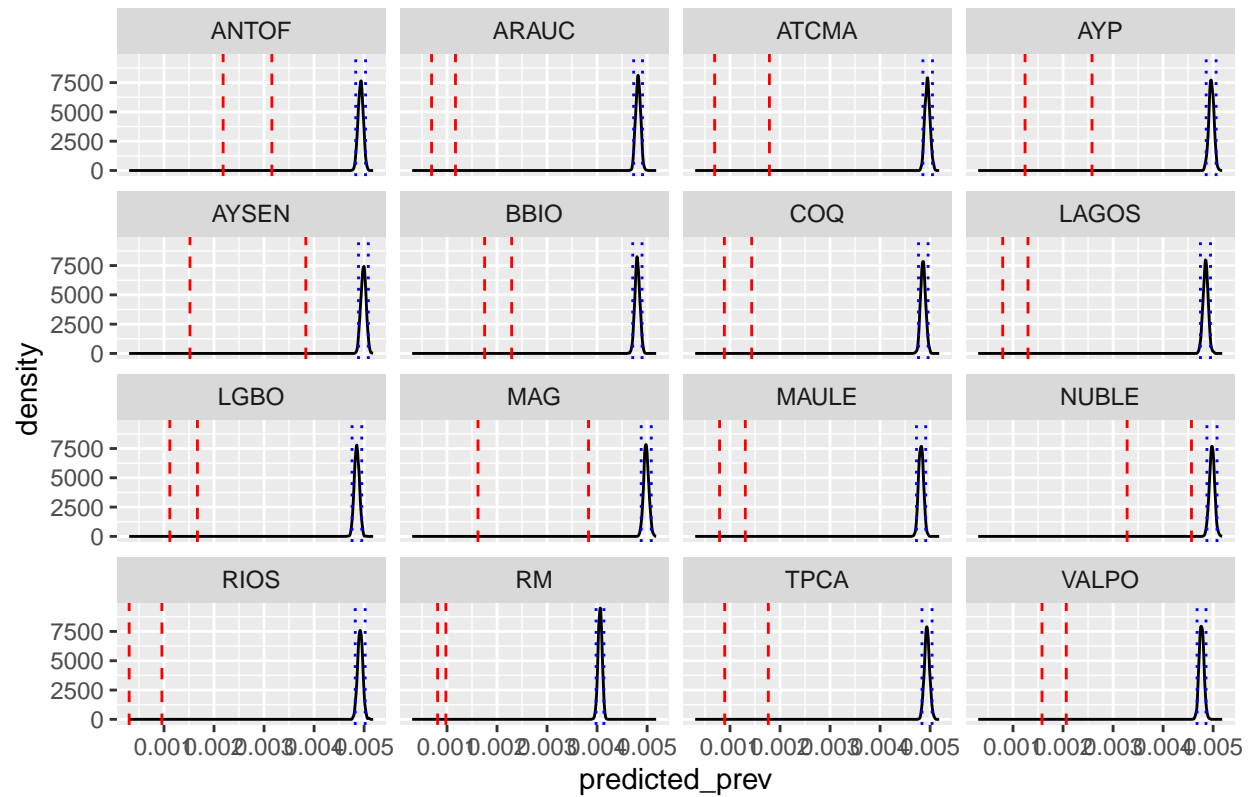
Maybe need different extrapolation priors for females?

```
# Females
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_region_post_f <- do_jags_rand_model(x = aut_prev_region_adj_f,
                                              feat = "school_region_name_abr",
                                              model = rand_model,
                                              theta_mu = theta_mu_extrapolate[j],
                                              theta_sigma = theta_sigma_extrapolate[j],
                                              pars = pars,
                                              convergence_checks = FALSE) %>%
    rename("school_region_name_abr" = "Feat_names")

  plot_post_density(aut_prev_region_post_f,
                    aut_prev_region_adj_f,
                    feat = "school_region_name_abr",
                    theta_mu = theta_mu_extrapolate[j],
                    theta_sigma = theta_sigma_extrapolate[j])
}
```

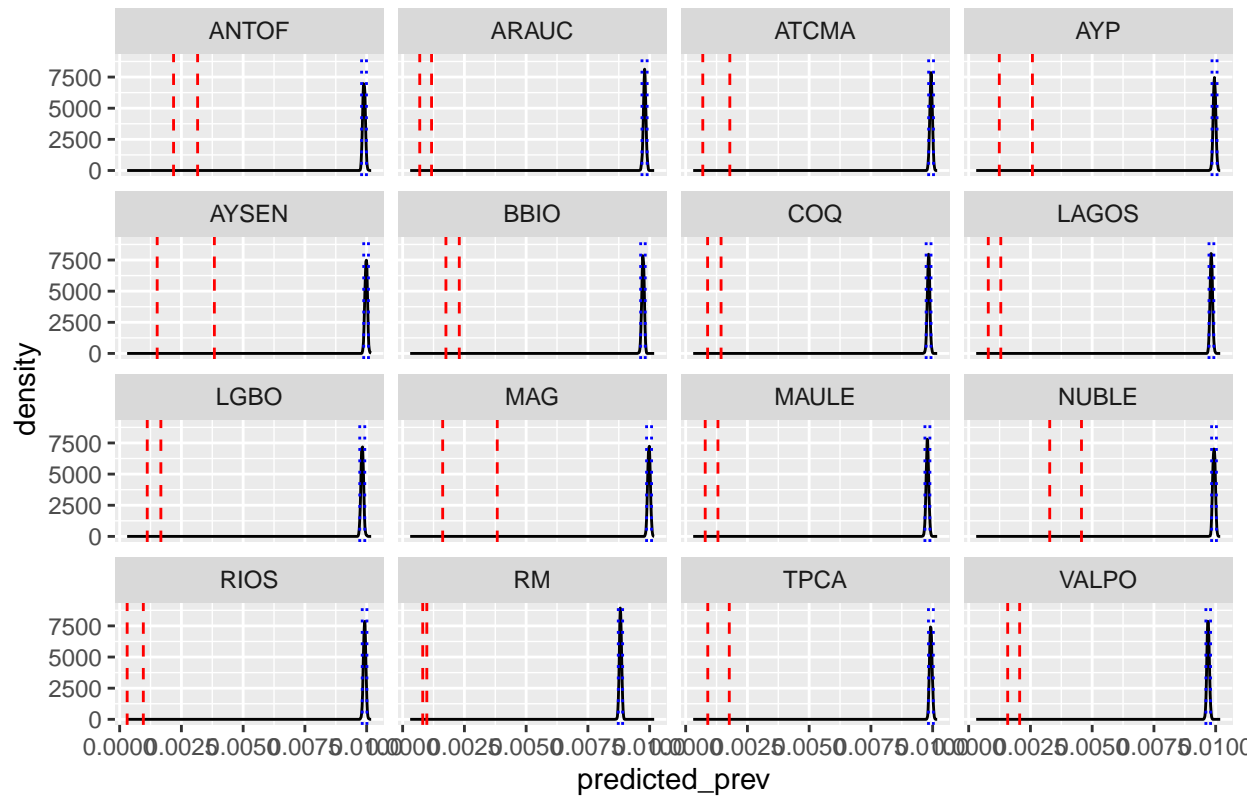
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



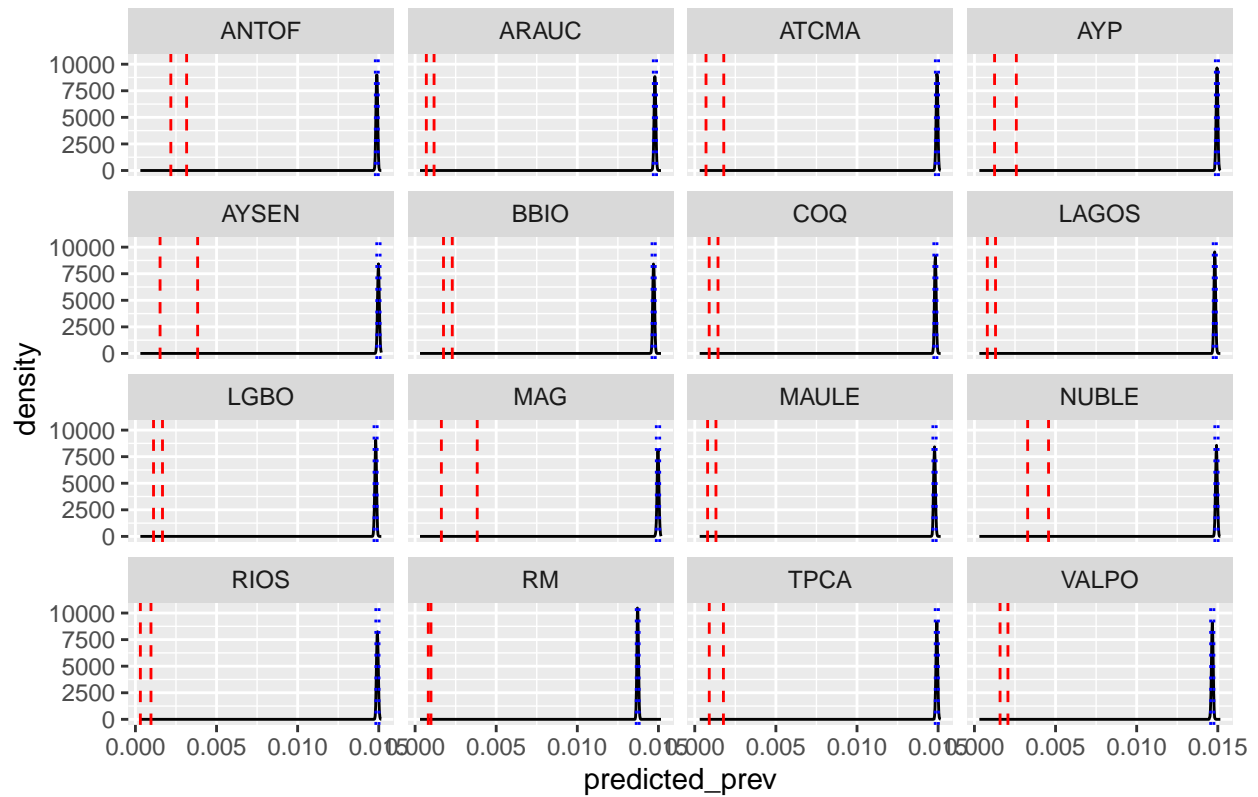
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



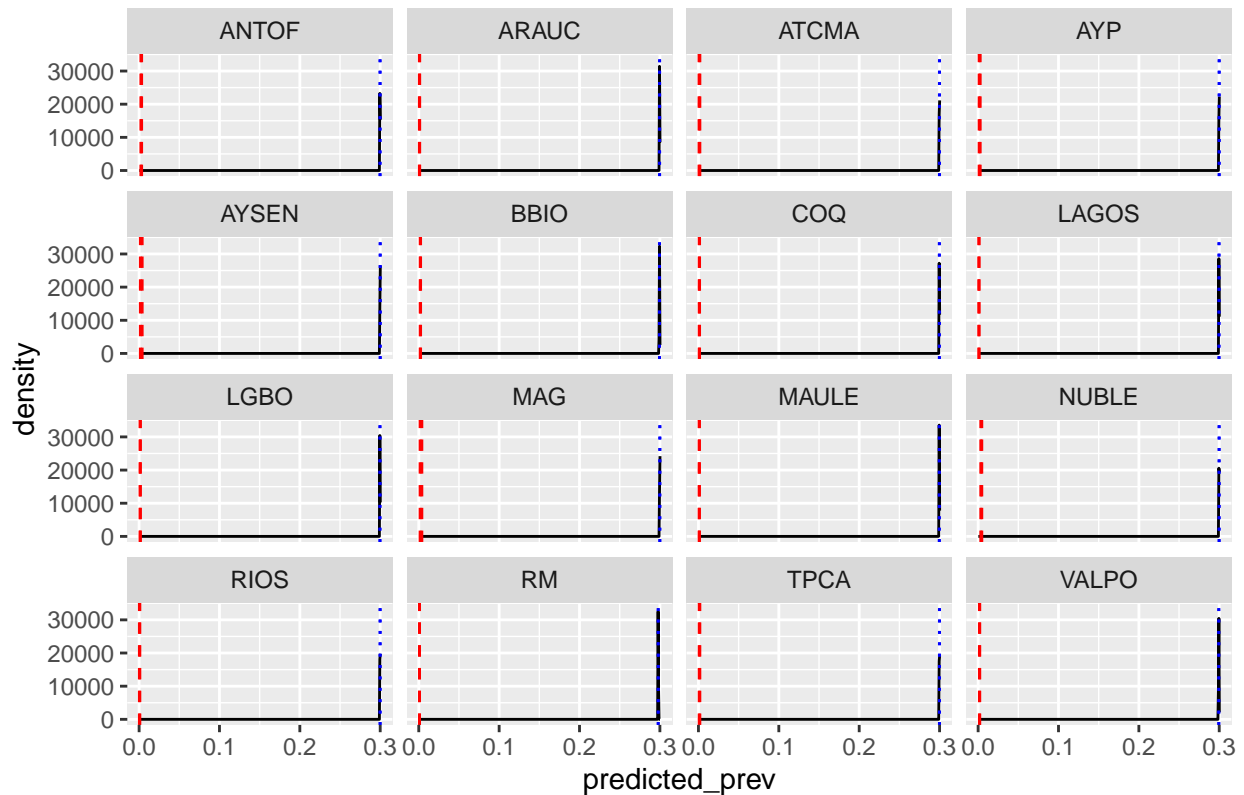
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05

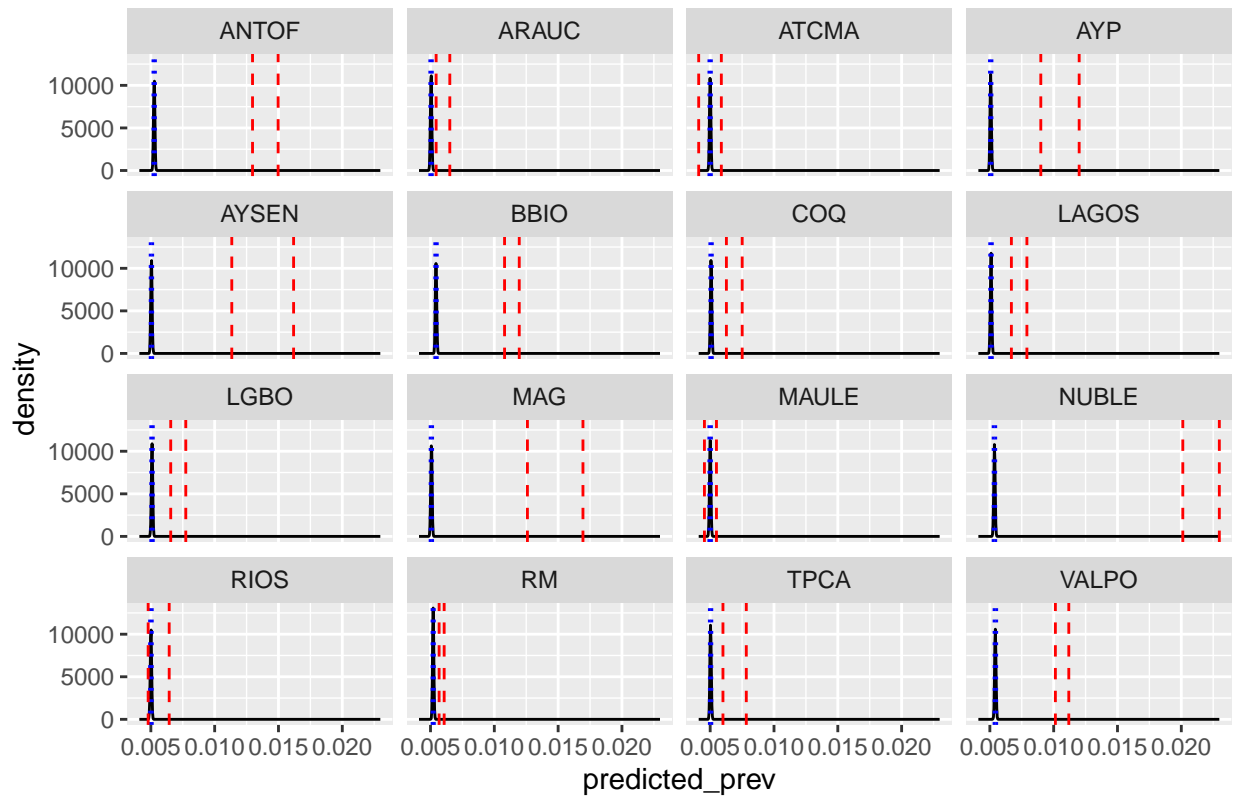


```
# Males
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_region_post_m <- do_jags_rand_model(x = aut_prev_region_adj_m,
                                              feat = "school_region_name_abr",
                                              model = rand_model,
                                              theta_mu = theta_mu_extrapolate[j],
                                              theta_sigma = theta_sigma_extrapolate[j],
                                              pars = pars,
                                              convergence_checks = FALSE) %>%
    rename("school_region_name_abr" = "Feat_names")

  plot_post_density(aut_prev_region_post_m,
                    aut_prev_region_adj_m,
                    feat = "school_region_name_abr",
                    theta_mu = theta_mu_extrapolate[j],
                    theta_sigma = theta_sigma_extrapolate[j])
}
```

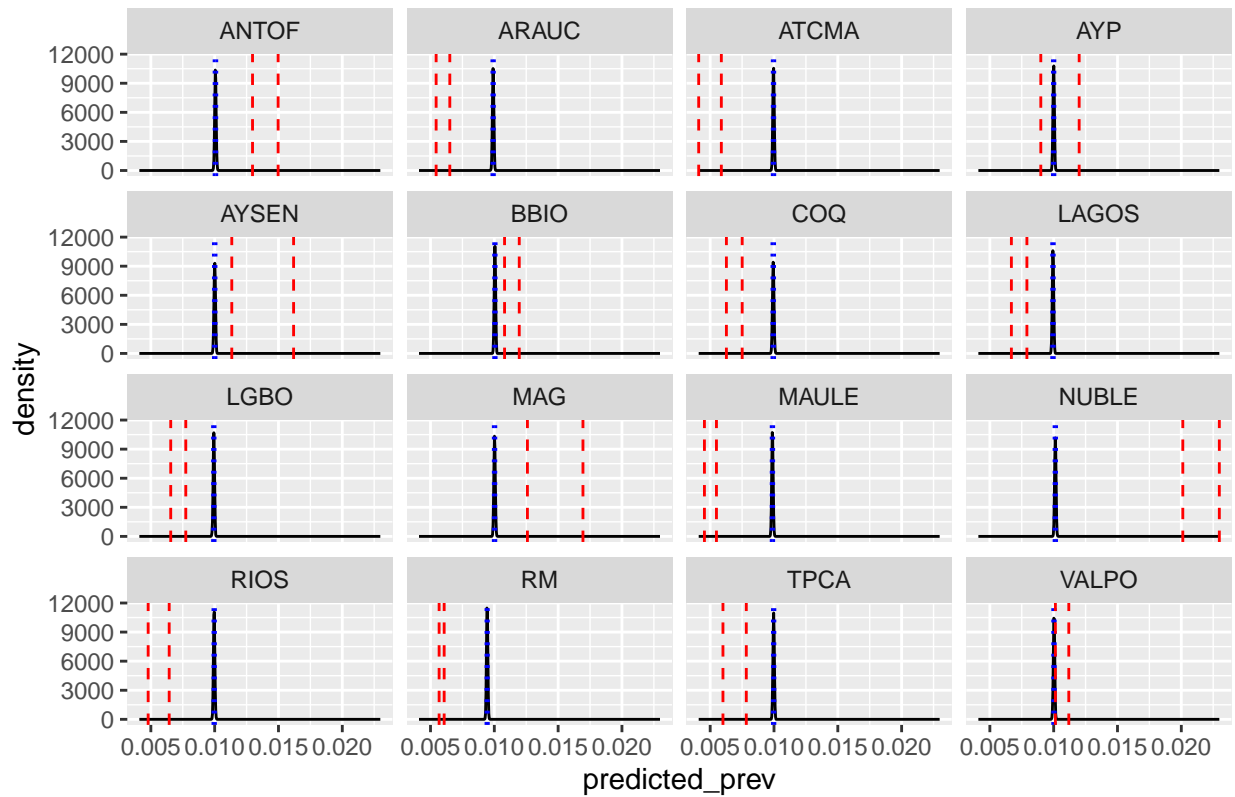
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



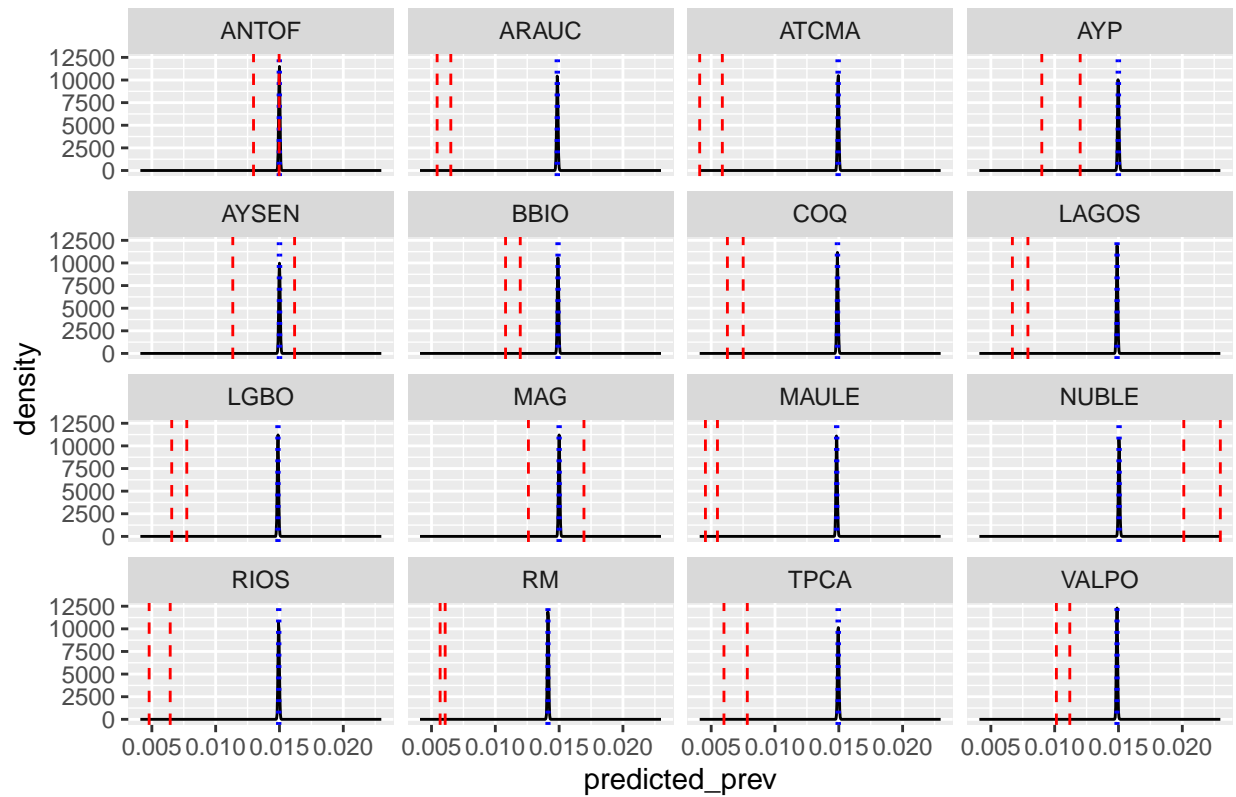
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



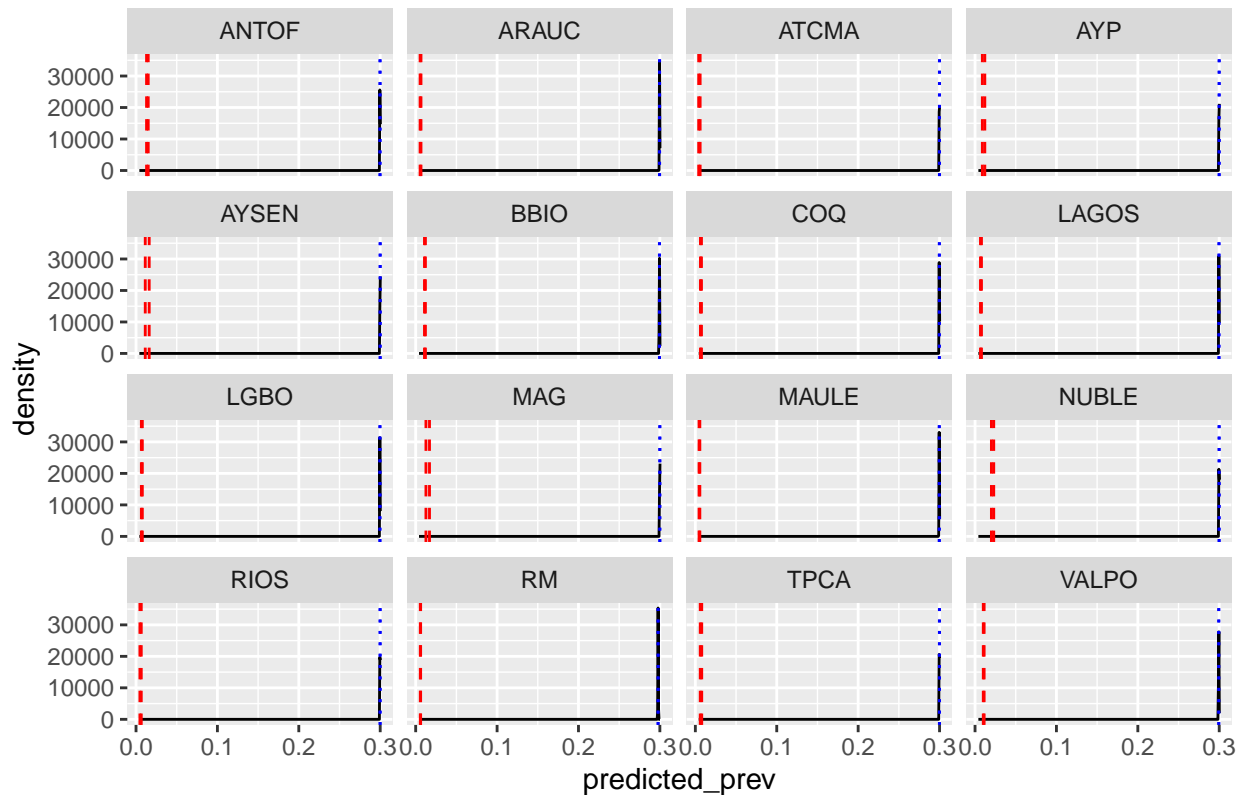
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.3, prior sd = 5.1e-05



Random effect on ethnicity, sexes separate

Need sex-specific priors

```
# Females
aut_prev_ethnic_f <- chile_bayes_aut %>%
  filter(sex == 2) %>%
  get_grouped_prev(stdpop = chile_stdpop_f, grouping_vars = c("ethnic_2_group", "age_june30", "sex", "a

## `summarise()` has grouped output by 'ethnic_2_group', 'age_june30', 'sex'. You
## can override using the `.groups` argument.

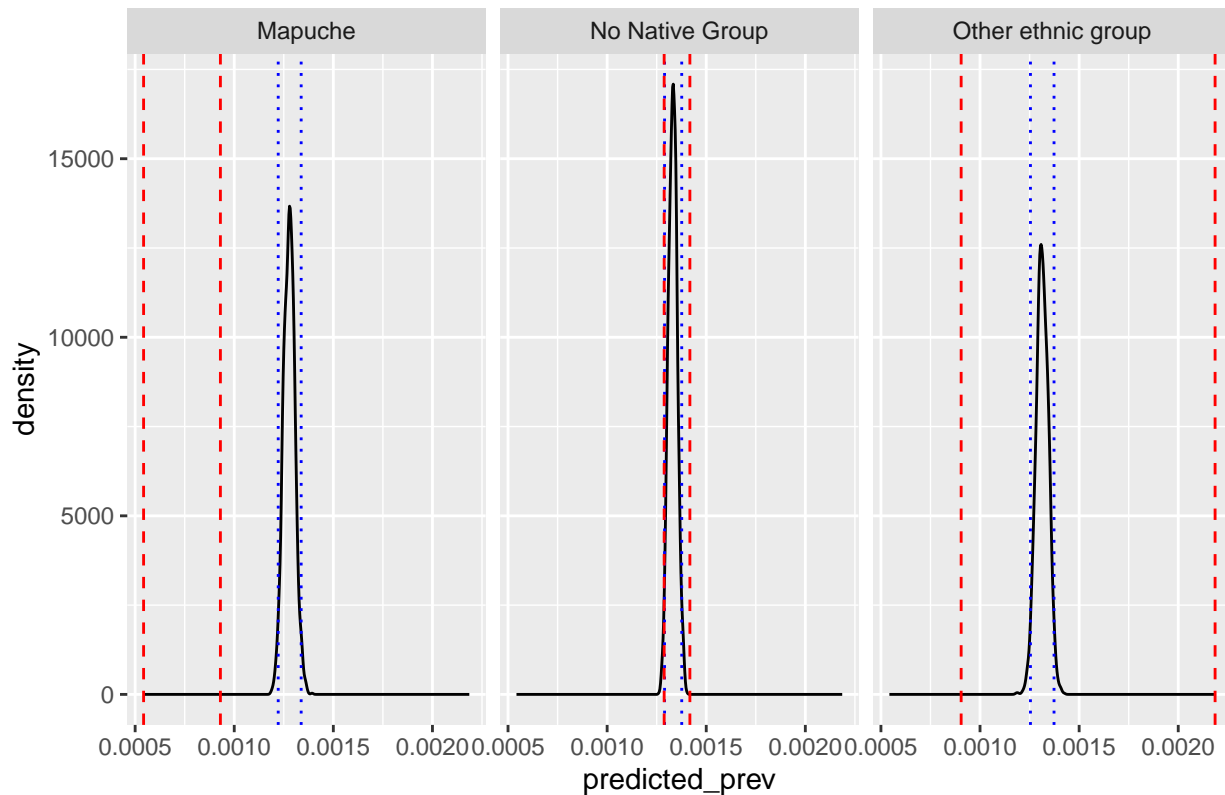
aut_prev_ethnic_adj_f <- get_adjusted_prev(aut_prev_ethnic_f, grouping_vars = "ethnic_2_group")

aut_prev_ethnic_post_f <- do_jags_rand_model(x = aut_prev_ethnic_adj_f,
  feat = "ethnic_2_group",
  model = rand_model,
  theta_mu = theta_mu_prior_f,
  theta_sigma = theta_sigma_prior_f,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("ethnic_2_group" = "Feat_names")

plot_post_density(aut_prev_ethnic_post_f, aut_prev_ethnic_adj_f, feat = "ethnic_2_group", theta_mu = th

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



```
# Males
aut_prev_ethnic_m <- chile_bayes_aut %>%
  filter(sex == 1) %>%
  get_grouped_prev(stdpop = chile_stdpop_m, grouping_vars = c("ethnic_2_group", "age_june30", "sex", "a

## `summarise()` has grouped output by 'ethnic_2_group', 'age_june30', 'sex'. You
## can override using the `.groups` argument.

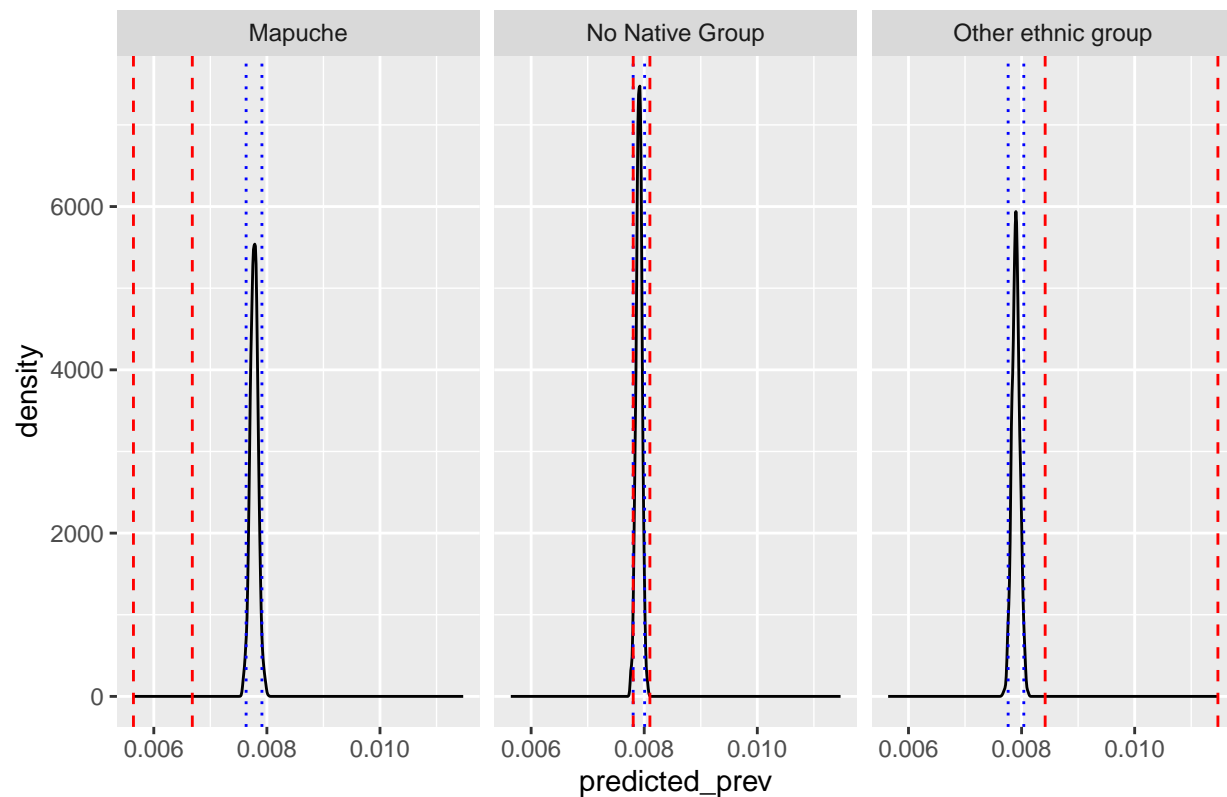
aut_prev_ethnic_adj_m <- get_adjusted_prev(aut_prev_ethnic_m, grouping_vars = "ethnic_2_group")

aut_prev_ethnic_post_m <- do_jags_rand_model(x = aut_prev_ethnic_adj_m,
  feat = "ethnic_2_group",
  model = rand_model,
  theta_mu = theta_mu_prior_m,
  theta_sigma = theta_sigma_prior_m,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("ethnic_2_group" = "Feat_names")

plot_post_density(aut_prev_ethnic_post_m, aut_prev_ethnic_adj_m, feat = "ethnic_2_group", theta_mu = th

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.0046, prior sd = 5.1e-05



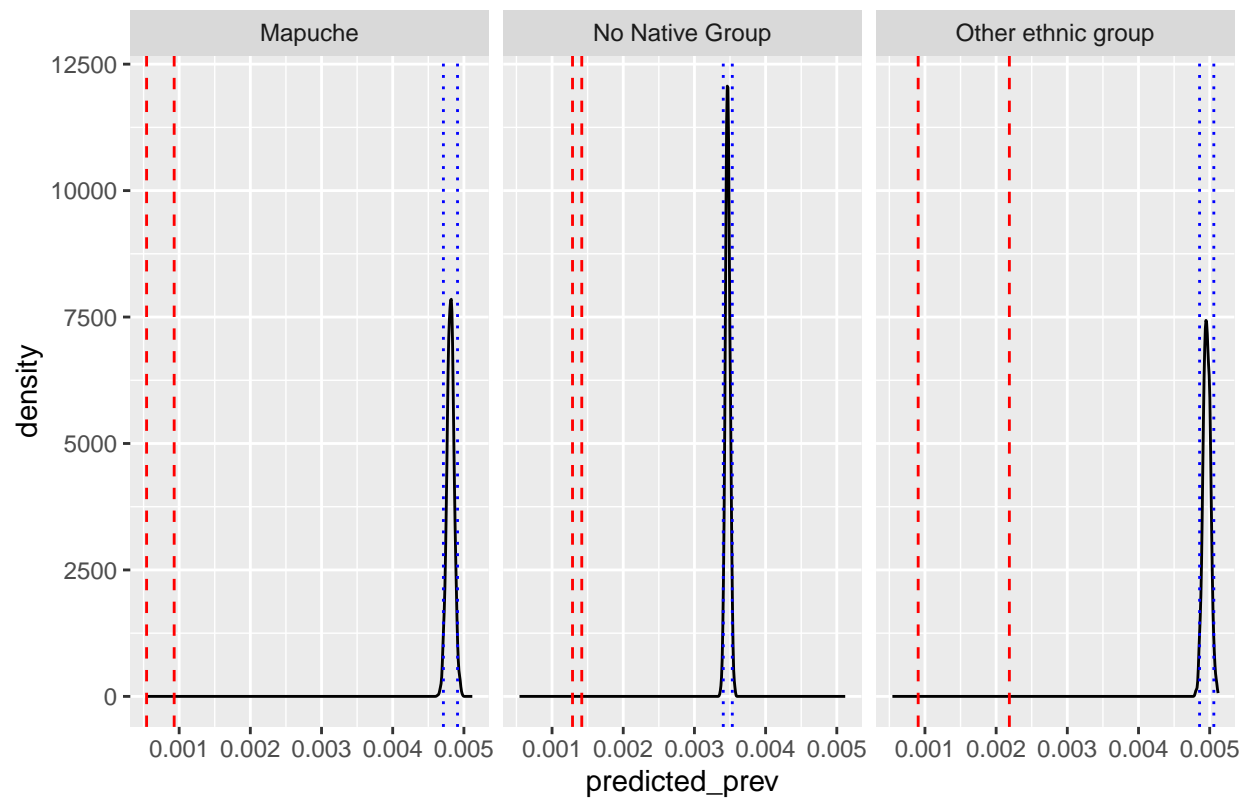
```
# Females
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_ethnic_post_f <- do_jags_rand_model(x = aut_prev_ethnic_adj_f,
    feat = "ethnic_2_group",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("ethnic_2_group" = "Feat_names")

  plot_post_density(aut_prev_ethnic_post_f,
    aut_prev_ethnic_adj_f,
    feat = "ethnic_2_group",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

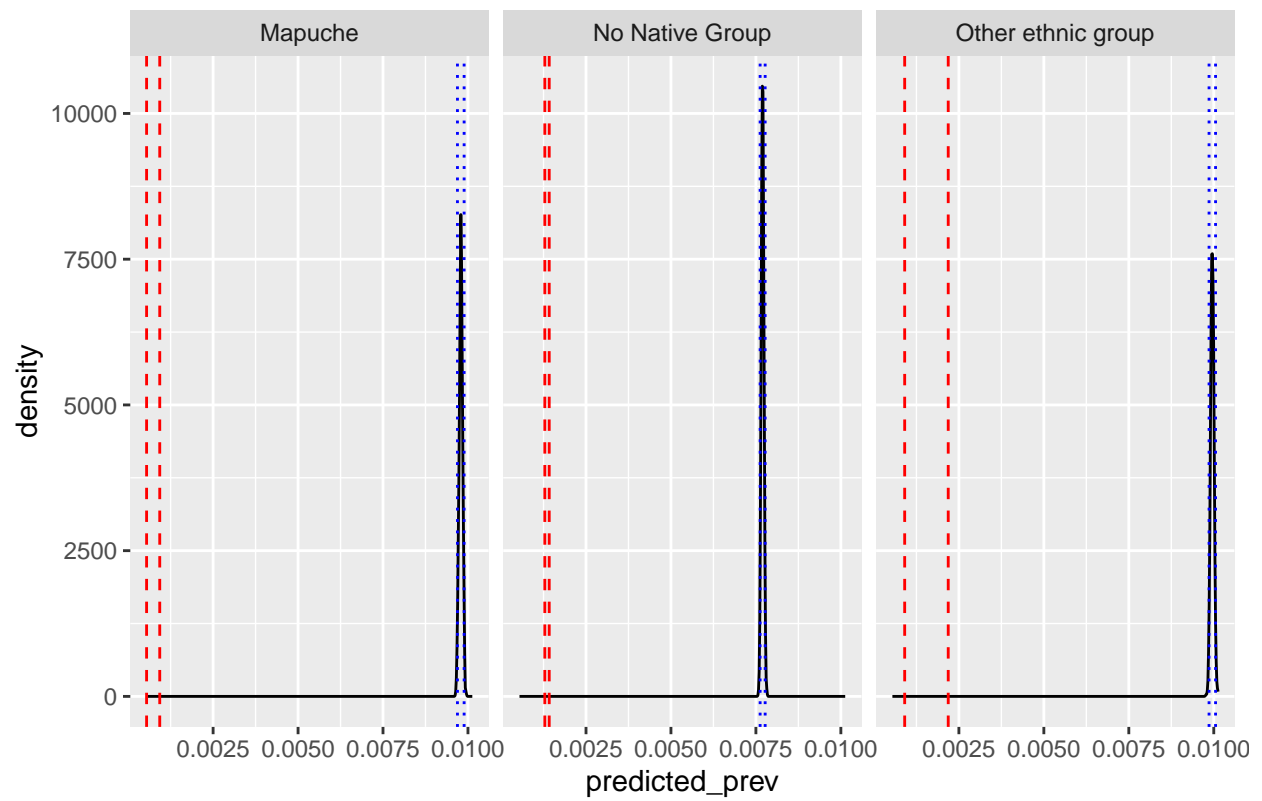
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



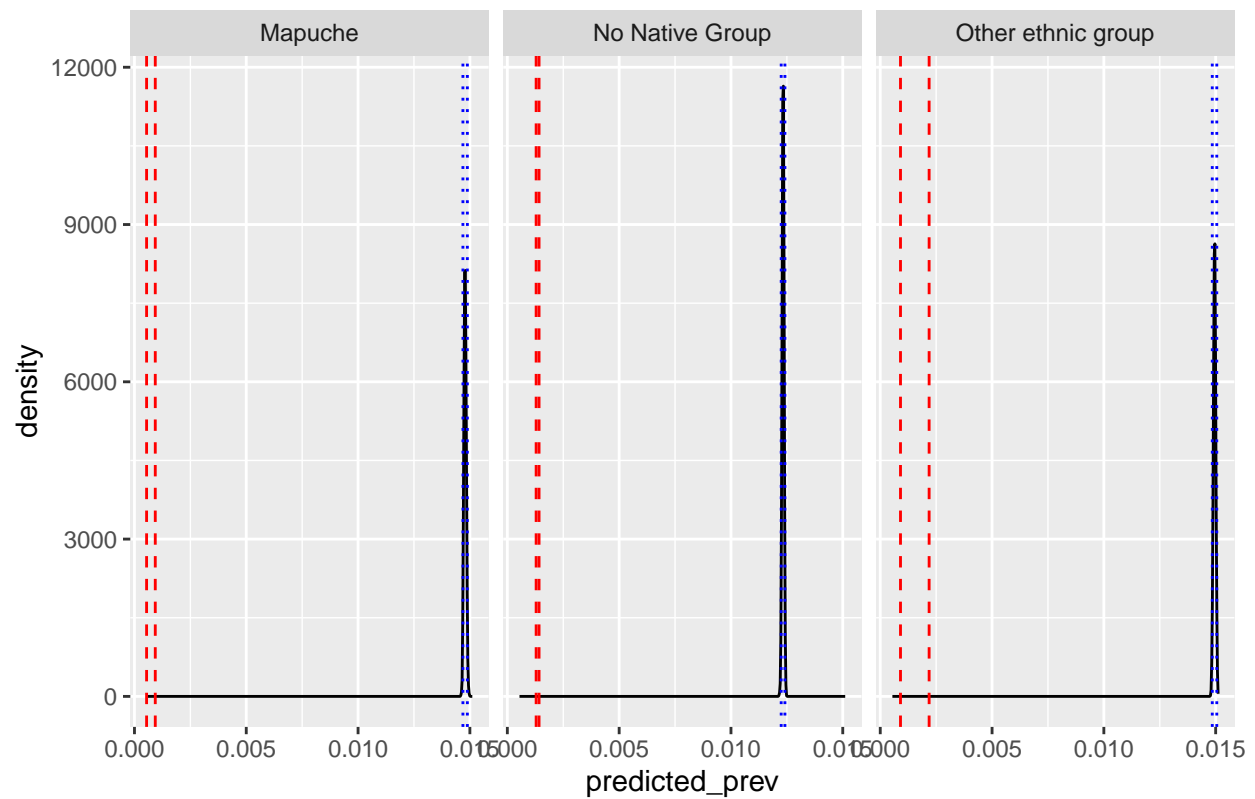
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.01, prior sd = 5.1e-05



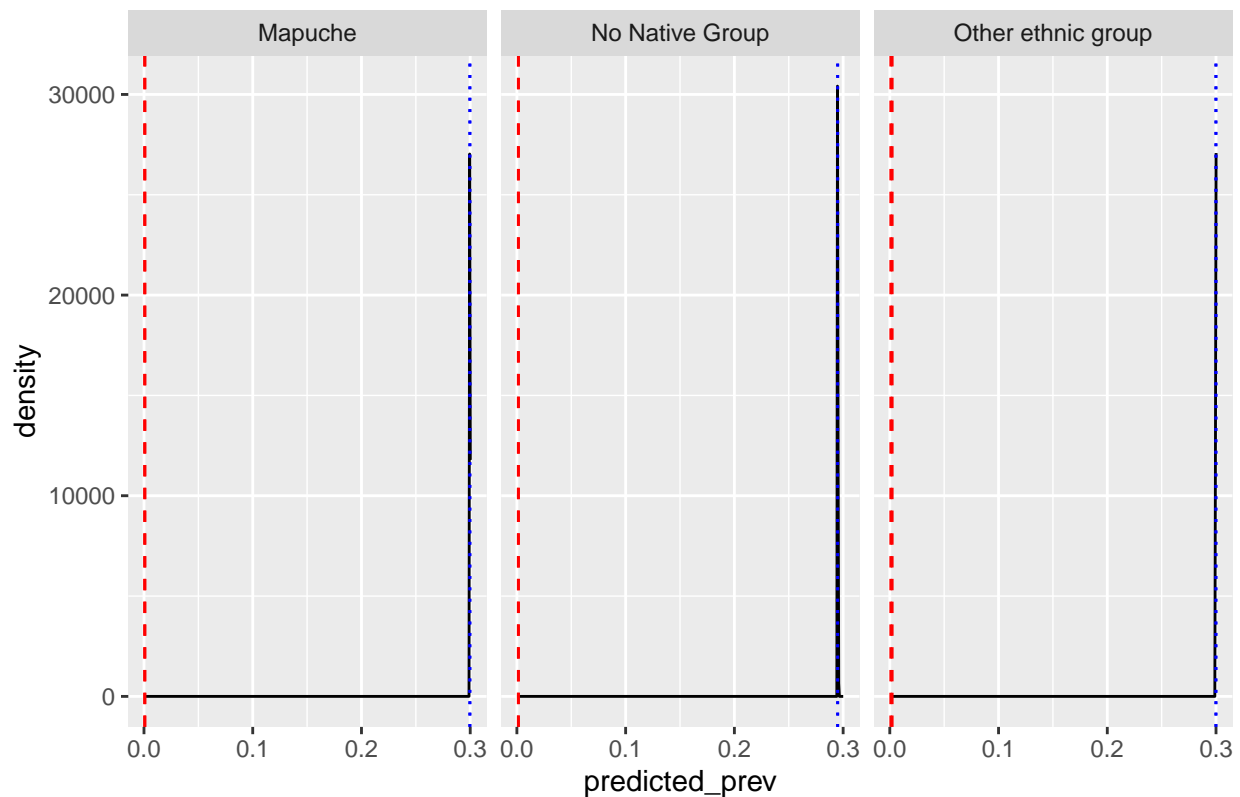
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.3, prior sd = 5.1e-05



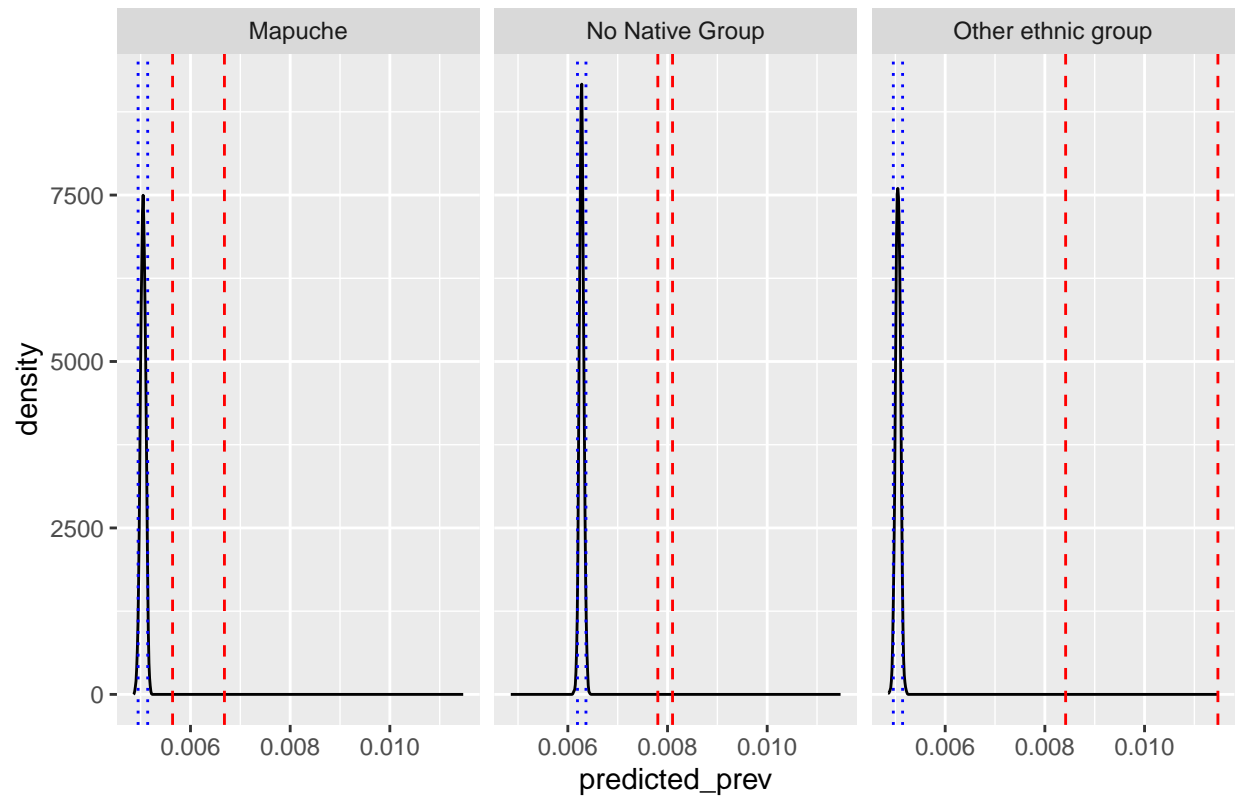
```
# Males
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_ethnic_post_m <- do_jags_rand_model(x = aut_prev_ethnic_adj_m,
                                              feat = "ethnic_2_group",
                                              model = rand_model,
                                              theta_mu = theta_mu_extrapolate[j],
                                              theta_sigma = theta_sigma_extrapolate[j],
                                              pars = pars,
                                              convergence_checks = FALSE) %>%

  rename("ethnic_2_group" = "Feat_names")

  plot_post_density(aut_prev_ethnic_post_m,
                    aut_prev_ethnic_adj_m,
                    feat = "ethnic_2_group",
                    theta_mu = theta_mu_extrapolate[j],
                    theta_sigma = theta_sigma_extrapolate[j])
}
```

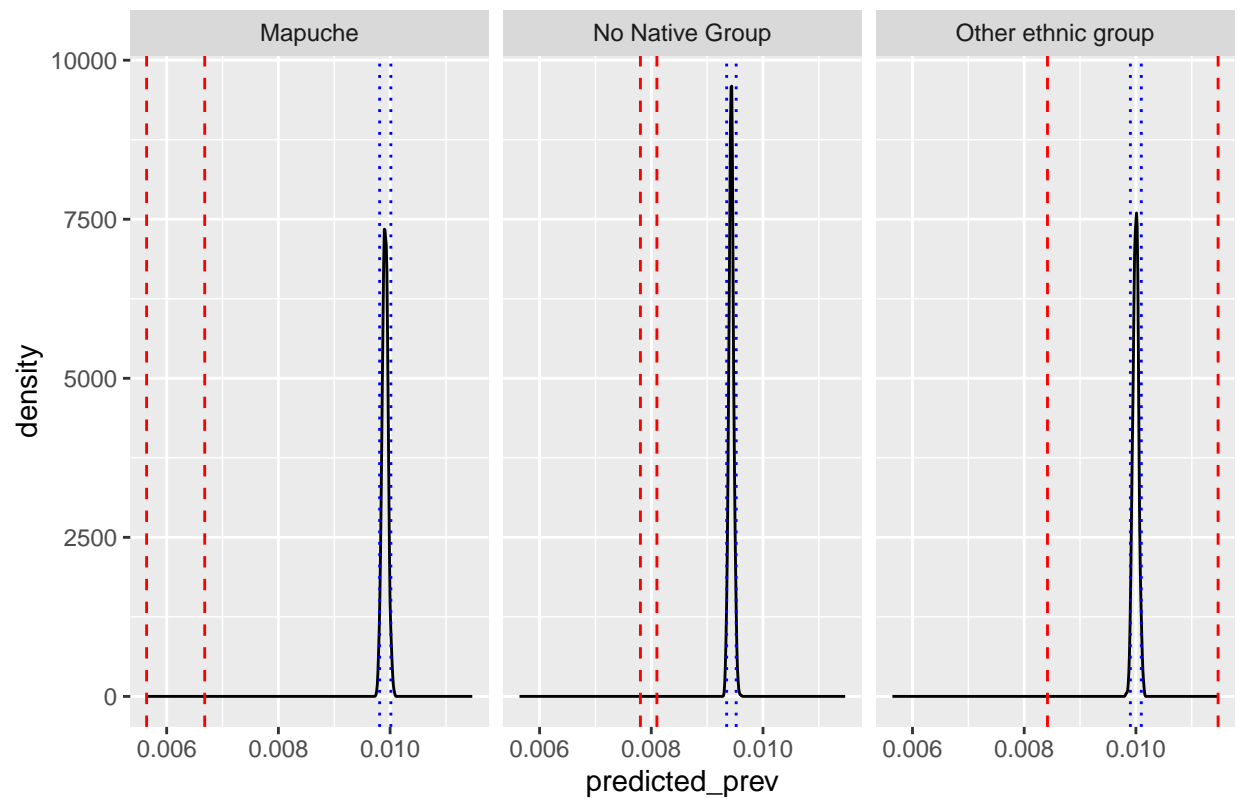
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



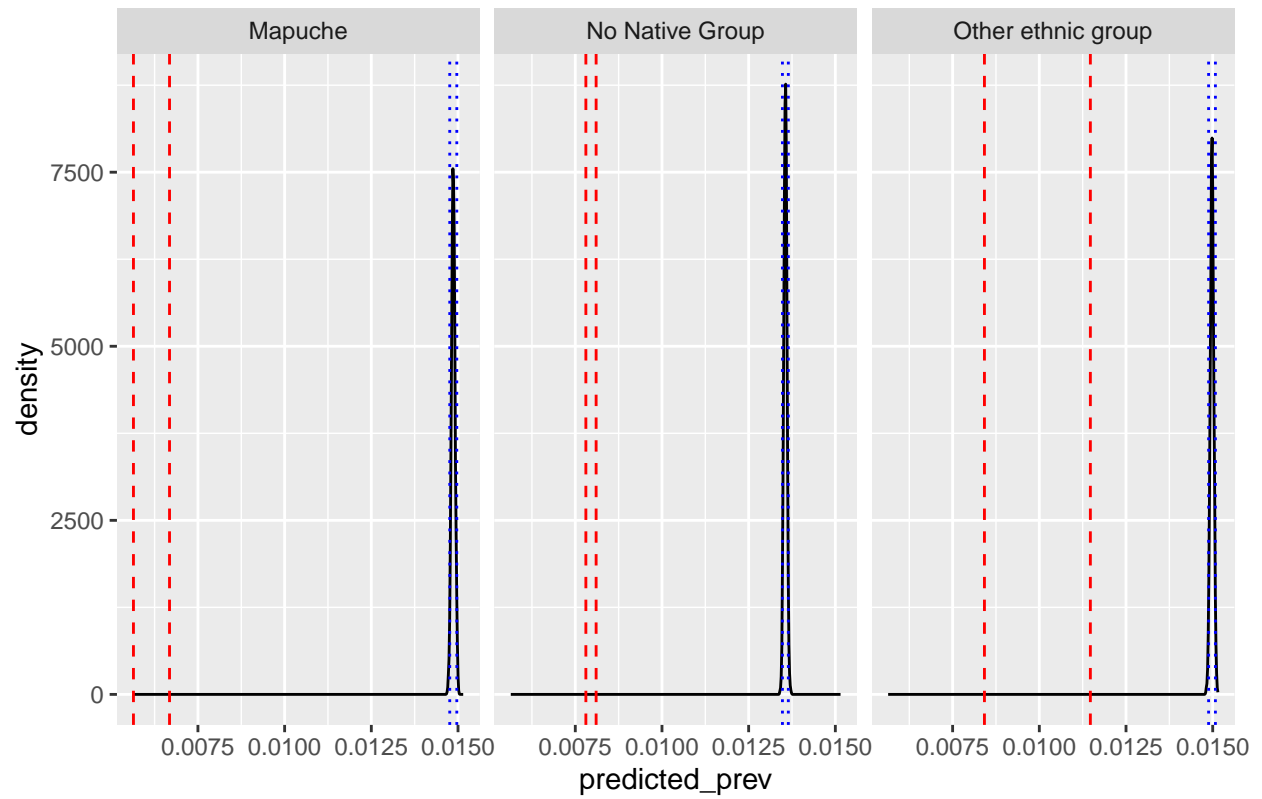
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



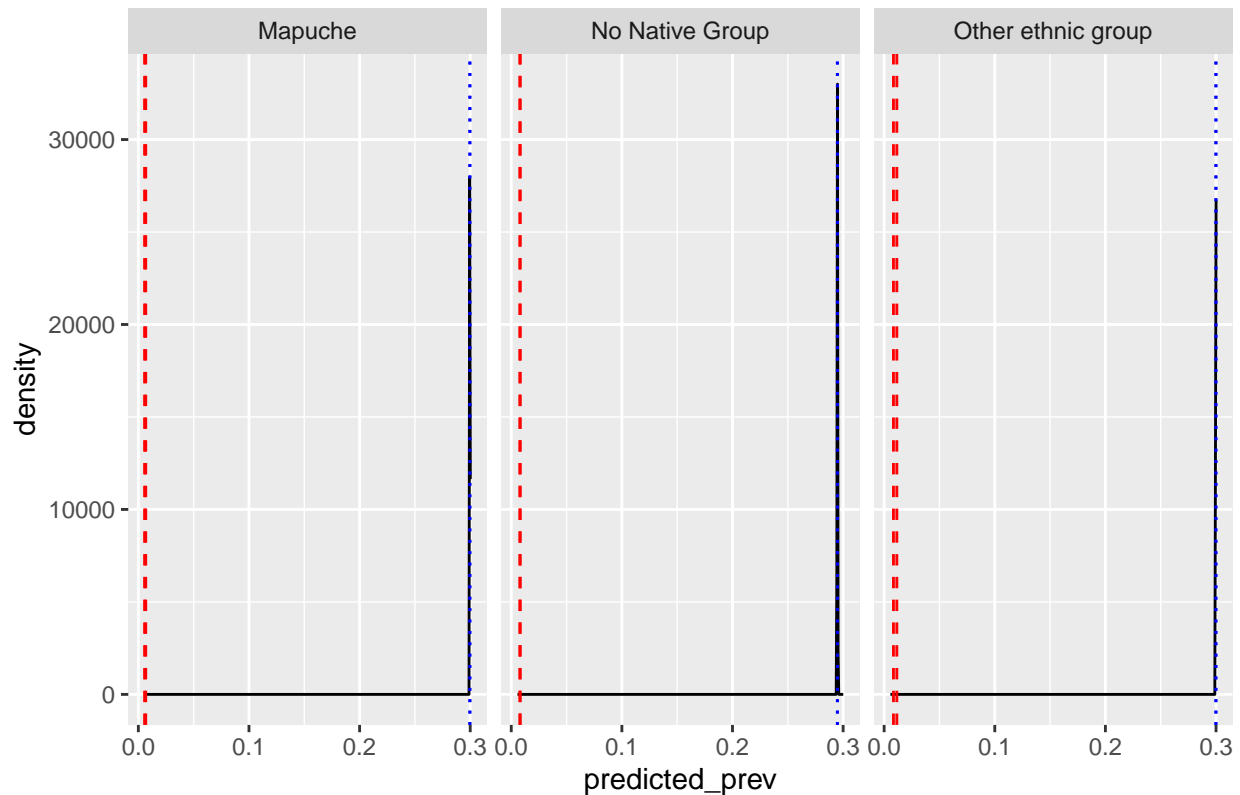
```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.3, prior sd = 5.1e-05



Random effect on economic status, sexes separate

```
# Females
aut_prev_econ_f <- chile_bayes_aut %>%
  filter(sex == 2) %>%
  get_grouped_prev(stdpop = chile_stdpop_f, grouping_vars = c("school_fee", "age_june30", "sex", "autism"))

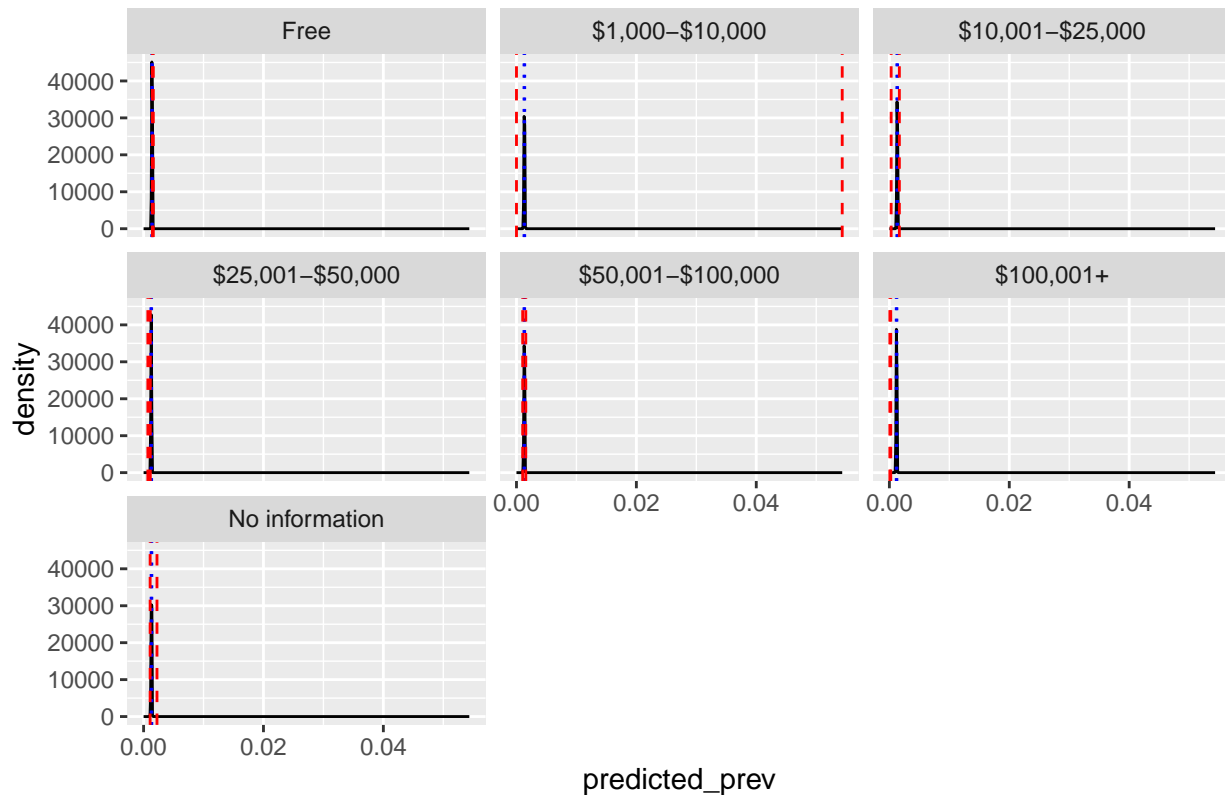
## `summarise()` has grouped output by 'school_fee', 'age_june30', 'sex'. You can
## override using the `.groups` argument.
aut_prev_econ_adj_f <- get_adjusted_prev(aut_prev_econ_f, grouping_vars = "school_fee")

aut_prev_econ_post_f <- do_jags_rand_model(x = aut_prev_econ_adj_f,
  feat = "school_fee",
  model = rand_model,
  theta_mu = theta_mu_prior_f,
  theta_sigma = theta_sigma_prior_f,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_fee" = "Feat_names")

plot_post_density(aut_prev_econ_post_f, aut_prev_econ_adj_f, feat = "school_fee", theta_mu = theta_mu_p
```

Don't know how to automatically pick scale for object of type
<draws_matrix/draws/matrix>. Defaulting to continuous.

Prior mean = 0.0046, prior sd = 5.1e-05



```
# Males
aut_prev_econ_m <- chile_bayes_aut %>%
  filter(sex == 1) %>%
  get_grouped_prev(stdpop = chile_stdpop_m, grouping_vars = c("school_fee", "age_june30", "sex", "autism"))

## `summarise()` has grouped output by 'school_fee', 'age_june30', 'sex'. You can
## override using the `.groups` argument.

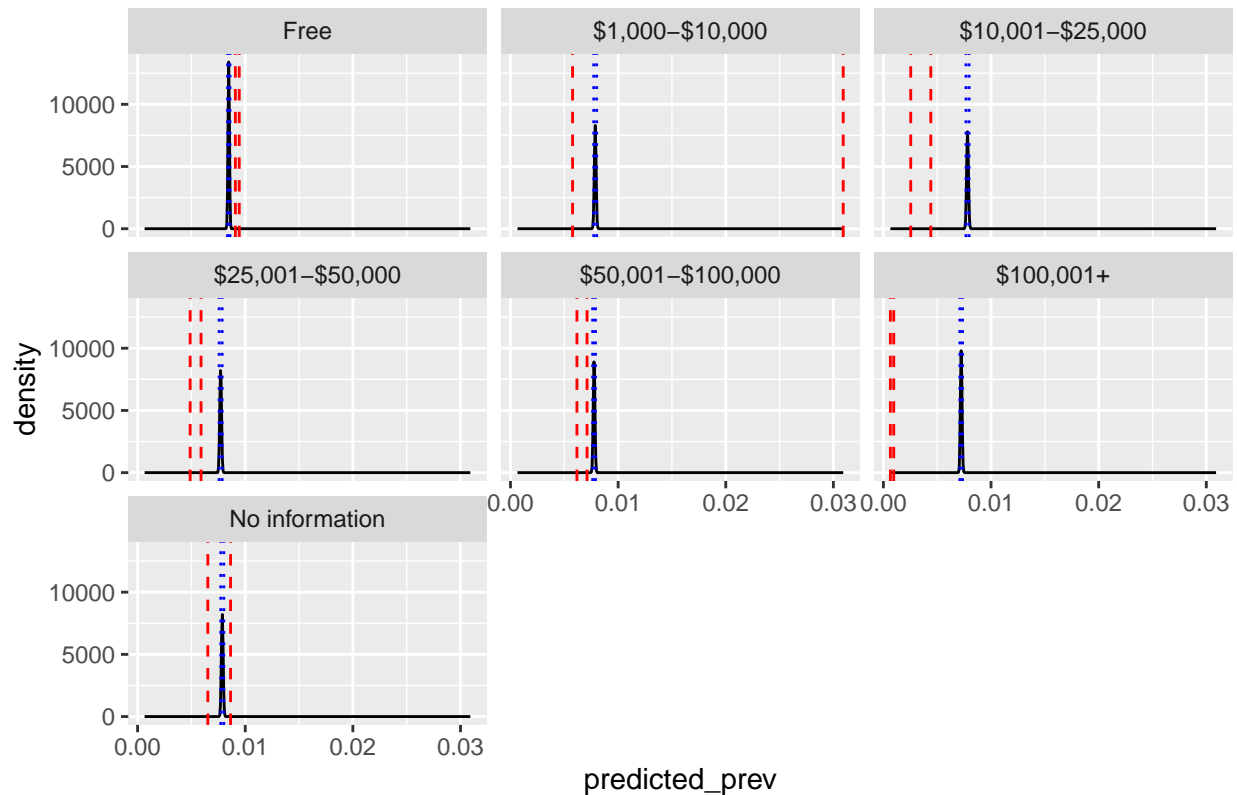
aut_prev_econ_adj_m <- get_adjusted_prev(aut_prev_econ_m, grouping_vars = "school_fee")

aut_prev_econ_post_m <- do_jags_rand_model(x = aut_prev_econ_adj_m,
  feat = "school_fee",
  model = rand_model,
  theta_mu = theta_mu_prior_m,
  theta_sigma = theta_sigma_prior_m,
  pars = pars,
  convergence_checks = FALSE) %>%
  rename("school_fee" = "Feat_names")

plot_post_density(aut_prev_econ_post_m, aut_prev_econ_adj_m, feat = "school_fee", theta_mu = theta_mu_p)

## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.0046, prior sd = 5.1e-05



Maybe need to redefine gamma upper CI for female 1,000-10,000 because there are 0 cases.

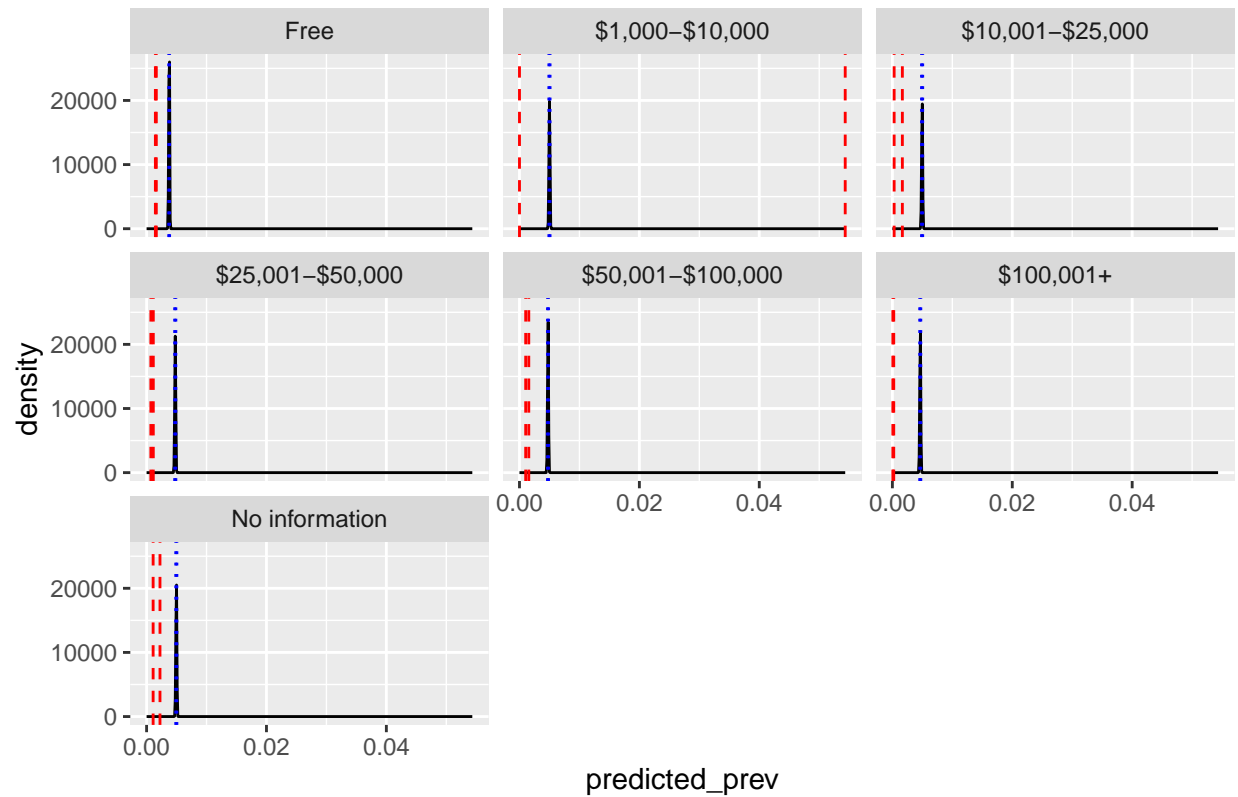
```
# Females
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_econ_post_f <- do_jags_rand_model(x = aut_prev_econ_adj_f,
    feat = "school_fee",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("school_fee" = "Feat_names")

  plot_post_density(aut_prev_econ_post_f,
    aut_prev_econ_adj_f,
    feat = "school_fee",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}

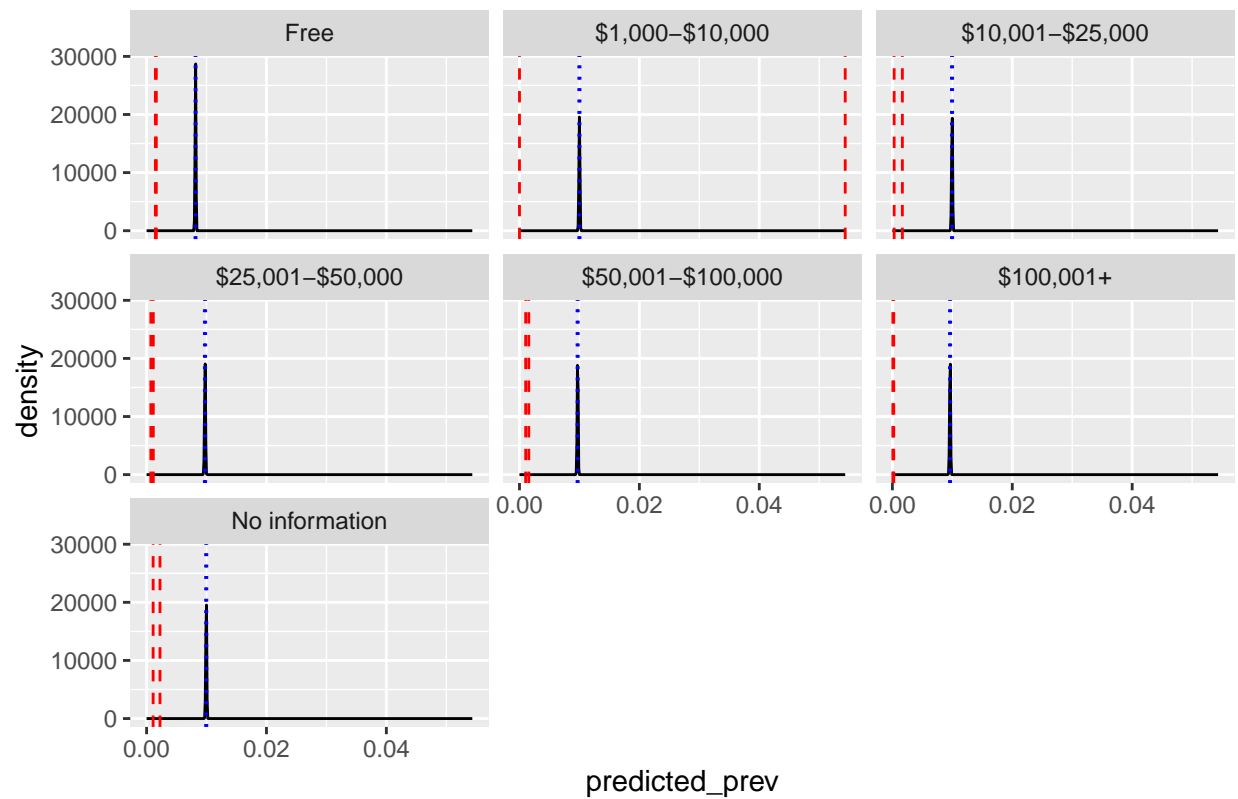
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



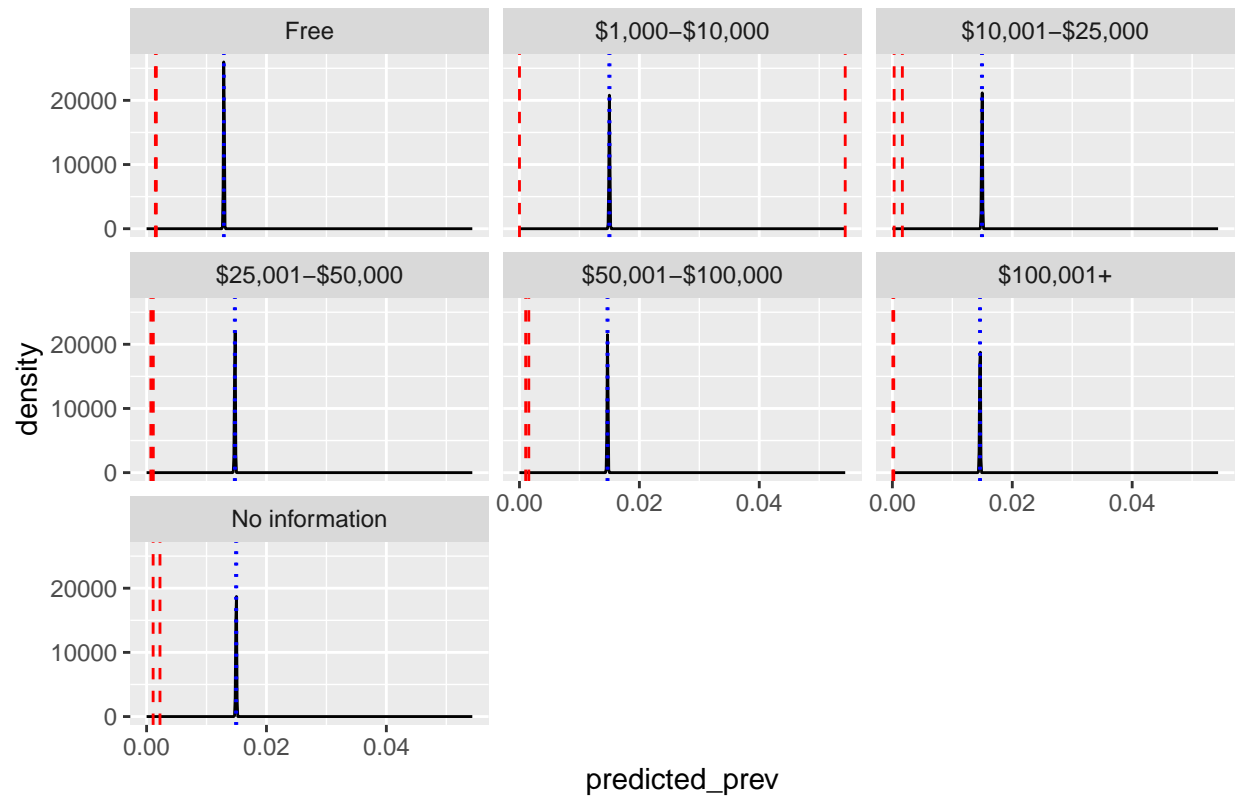
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



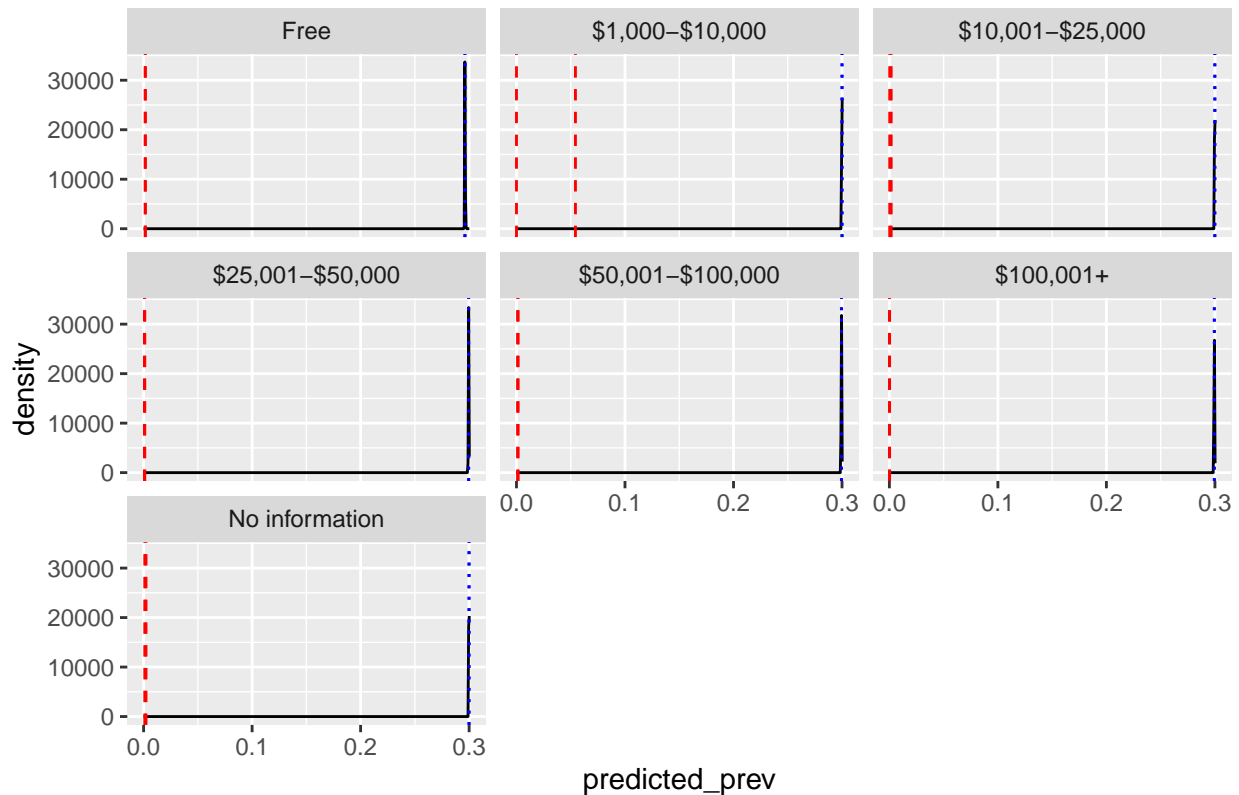
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type  
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.3, prior sd = 5.1e-05



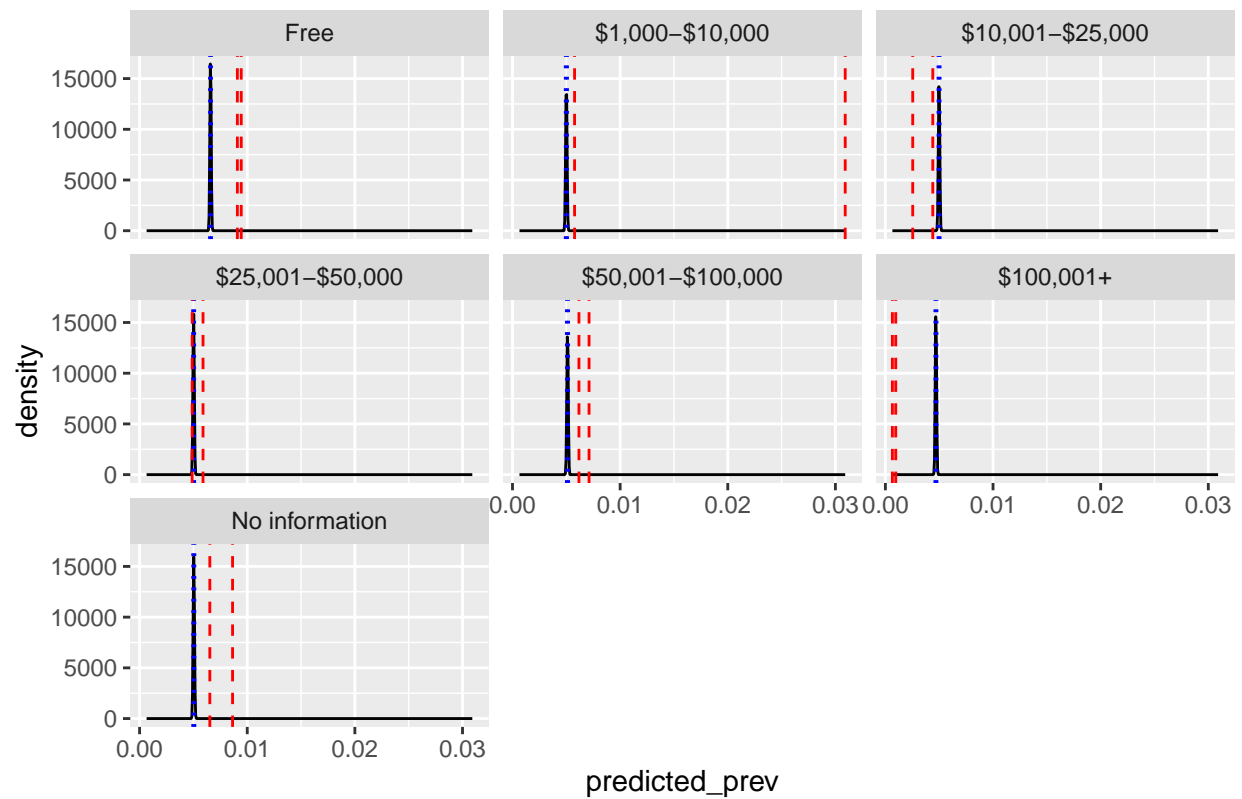
```
# Males
for(j in 1:length(theta_mu_extrapolate)) {
  aut_prev_econ_post_m <- do_jags_rand_model(x = aut_prev_econ_adj_m,
    feat = "school_fee",
    model = rand_model,
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j],
    pars = pars,
    convergence_checks = FALSE) %>%

  rename("school_fee" = "Feat_names")

  plot_post_density(aut_prev_econ_post_m,
    aut_prev_econ_adj_m,
    feat = "school_fee",
    theta_mu = theta_mu_extrapolate[j],
    theta_sigma = theta_sigma_extrapolate[j])
}
```

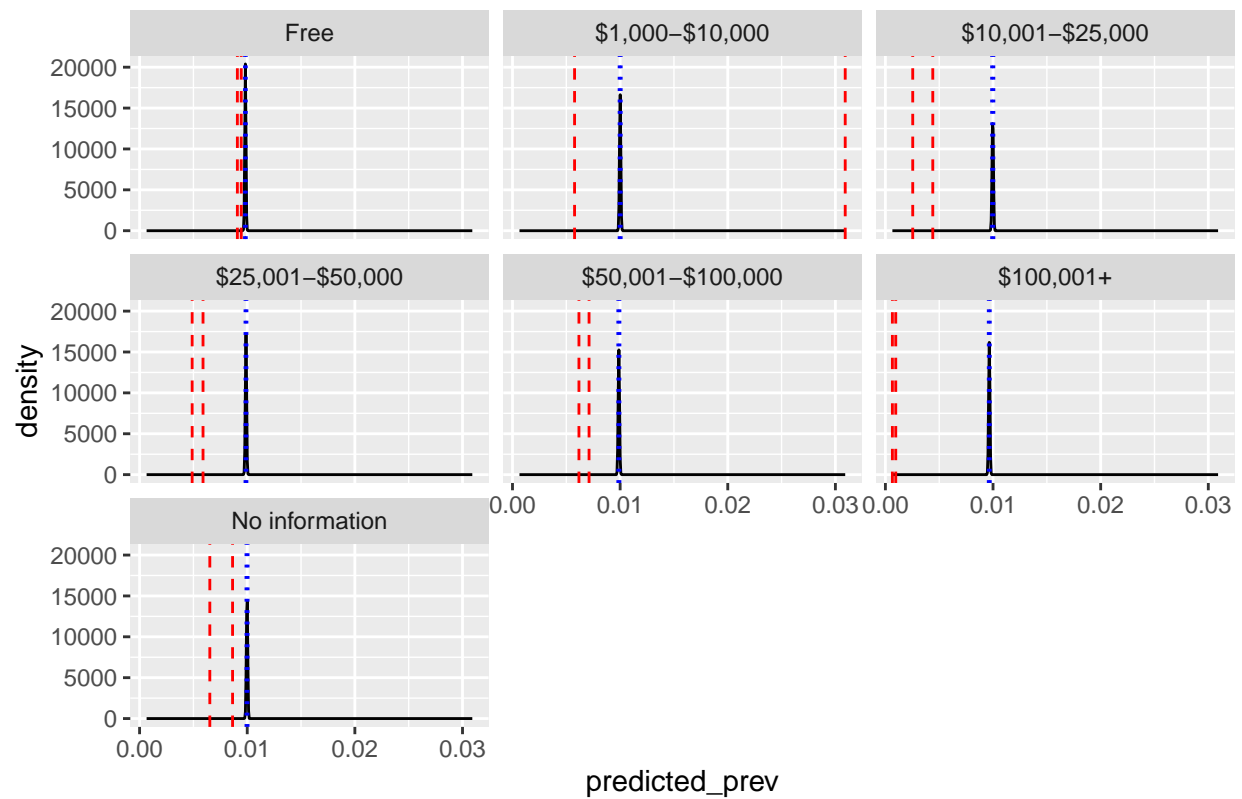
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.005, prior sd = 5.1e-05



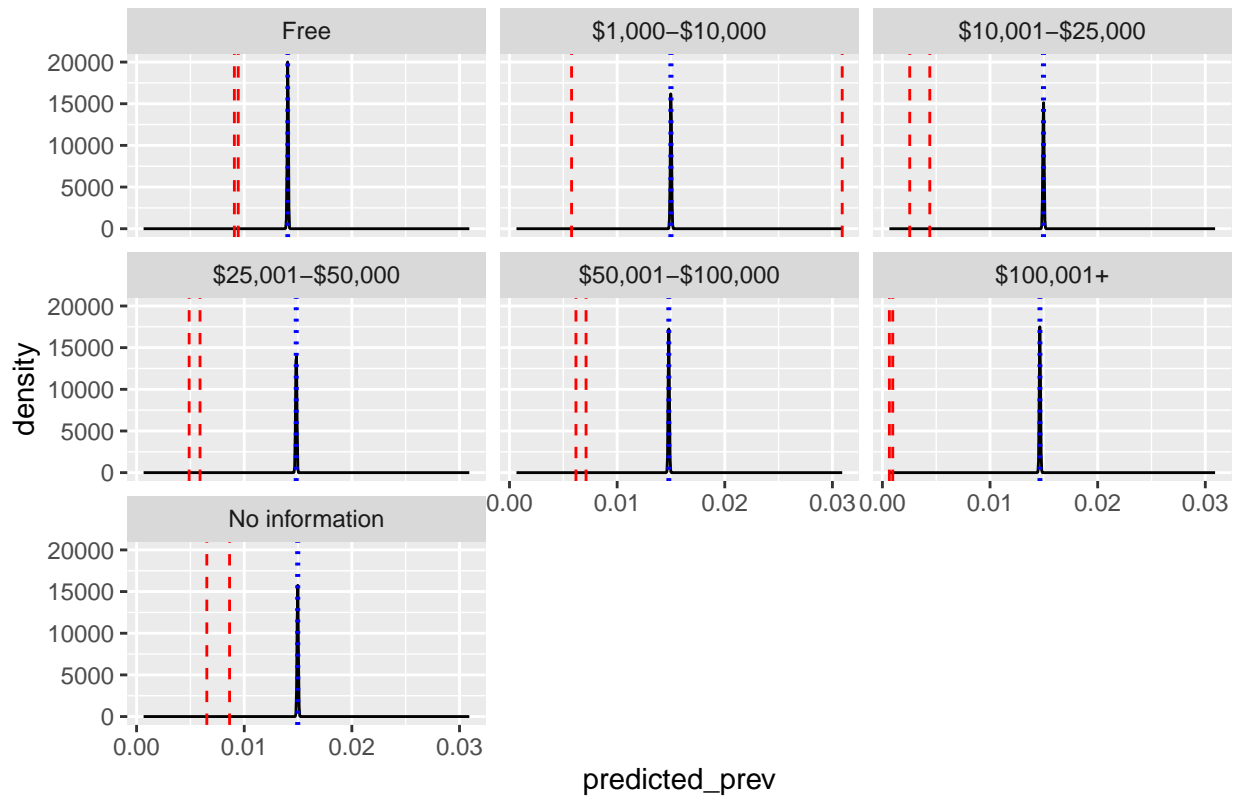
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.01, prior sd = 5.1e-05



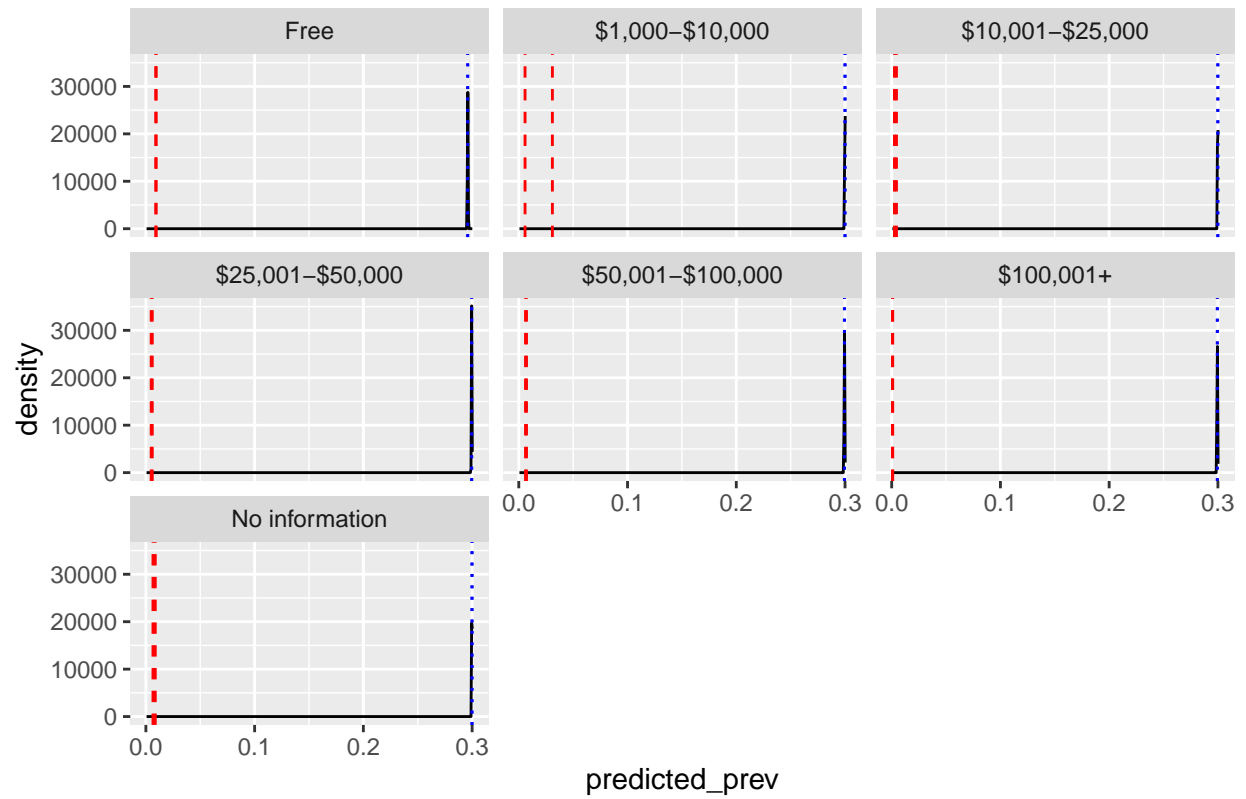
```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```

Prior mean = 0.015, prior sd = 5.1e-05



```
## Don't know how to automatically pick scale for object of type
## <draws_matrix/draws/matrix>. Defaulting to continuous.
```


Prior mean = 0.3, prior sd = 5.1e-05



Could do random effect on sex and region together