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##### 0 - safe choice A, 1 - risky choice B #####
library(rstan); rstan_options(javascript=FALSE)

## Loading required package: StanHeaders

## Loading required package: ggplot2

## rstan (Version 2.21.8, GitRev: 2e1f913d3ca3)

## For execution on a local, multicore CPU with excess RAM we recommend calling
## options(mc.cores = parallel::detectCores()).
## To avoid recompilation of unchanged Stan programs, we recommend calling
## rstan_options(auto_write = TRUE)

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

options(mc.cores = parallel::detectCores())
rstan_options(auto_write = T)

# Get list of files in 'data_2' folder with the pattern "riskytimed"
files <- dir(path = "data_2", pattern="riskytimed")

# Read all csv files in the list
data_list <- lapply(paste0("data_2/", files), read.table, header = TRUE, skip = 0, fill = TRUE, sep= ",")

# Concatenate rows of all items in the list into a data frame
dat <- do.call("rbind", data_list)

# gamble characteristics
dat$eva = dat$oa1*dat$pa1+dat$oa2*dat$pa2 + dat$oa3*dat$pa3+dat$oa4*dat$pa4
dat$evb = dat$ob1*dat$pb1+dat$ob2*dat$pb2 + dat$ob3*dat$pb3+dat$ob4*dat$pb4
dat$evd = dat$evb - dat$eva
dat$sda = sqrt((dat$oa1-dat$eva)^2*dat$pa1 + (dat$oa2-dat$eva)^2*dat$pa2 + (dat$oa3-dat$eva)^2*dat$pa3)
dat$sdb = sqrt((dat$ob1-dat$evb)^2*dat$pb1 + (dat$ob2-dat$evb)^2*dat$pb2 + (dat$ob3-dat$evb)^2*dat$pb3)
dat$sdd = dat$sdb - dat$sda
dat$evdummy = ifelse(dat$evd>0,1,0)

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# transform to +/- 1; safe - 1, risky +1
dat$cho <- ifelse(dat$choice==0,-1,ifelse(dat$choice==1,1,NA))
ids <- unique(dat$id)
for(j in 1:length(ids)){
  dat$tid[dat$id==ids[j]] <- j
}
tids <- unique(dat$tid)
# only control data
control_dat <- dat[dat$cond=="control",]
# remove fast RTs
rcontrol_dat <- control_dat[control_dat$rt>1,]
# only condition no time pressure
dataList = list(cho = rcontrol_dat$cho, rt = rcontrol_dat$rt, participant = rcontrol_dat$tid, N=nrow(rcontrol_dat))

parameters = c("transf_mu_alpha","transf_mu_threshold","transf_mu_ndt", "transf_mu_theta", 'sd_threshold')

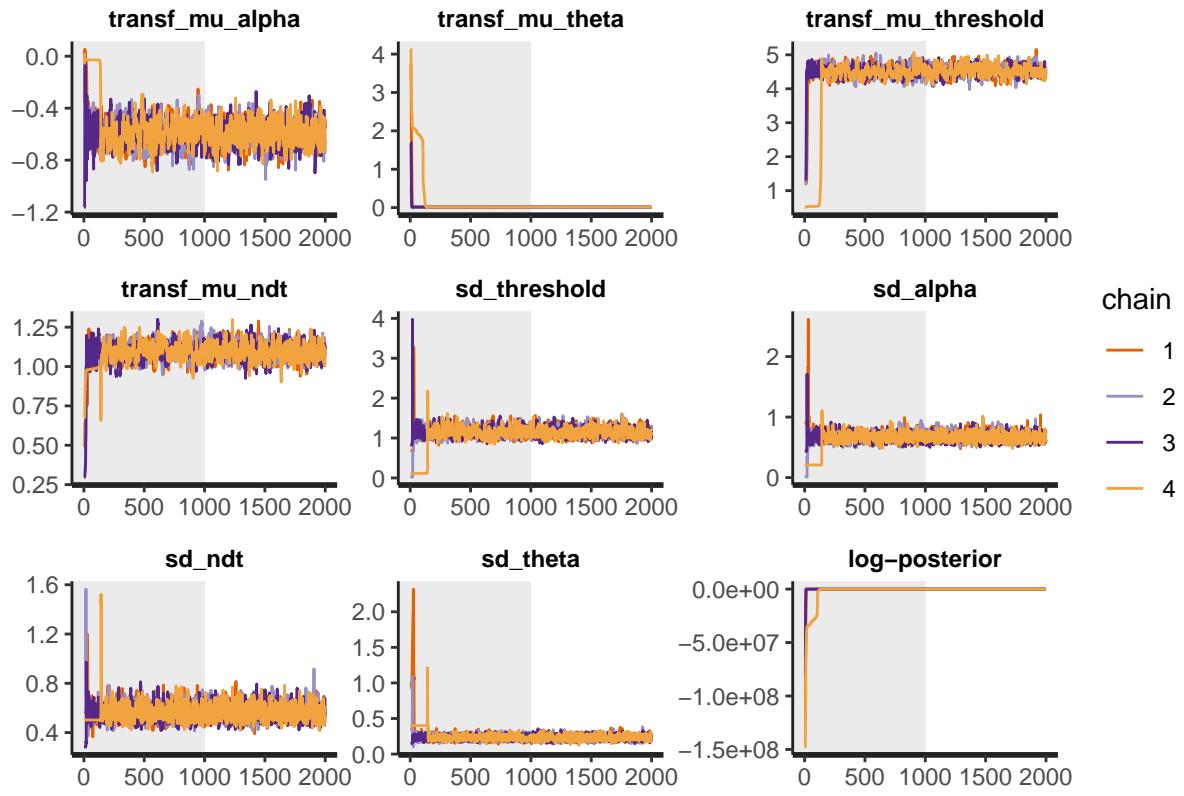
initFunc <-function (i) {
  initList=list()
  for (ll in 1:i){
    initList[[ll]] = list(mu_alpha = runif(1, -5, 5),
                          sd_alpha = runif(1,0,1),
                          mu_threshold = runif(1,-0.5, 2.5),
                          sd_threshold = runif(1, 0, 1),
                          mu_ndt = runif(1, -1.5, 0),
                          sd_ndt = runif(1, 0, 1),
                          mu_theta = runif(1,-4, 6),
                          sd_theta = runif(1,0,1),
                          z_alpha = runif(length(tids),-0.1,0.1),
                          z_theta = runif(length(tids),-0.1,0.1),
                          z_threshold = runif(length(tids),-0.1,0.1),
                          z_ndt = runif(length(tids),-0.1,0.1)
    )
  }
  return(initList)
}

m <- stan_model("MV_Baseline.stan")
dsamples <- sampling(m,
                      data=dataList,
                      pars=parameters,
                      iter=2000,
                      chains=4,#If not specified, gives random inits
                      init = initFunc(4),
                      warmup = 1000, # Stands for burn-in; Default = iter/2
                      seed = 12, # Setting seed; Default is random seed
                      refresh = 0
)

#parameters = c("transf_mu_alpha","transf_mu_threshold","transf_mu_ndt", "transf_mu_theta", 'sd_threshold')

rstan::traceplot(dsamples, pars=c("transf_mu_alpha","transf_mu_theta", "transf_mu_threshold","transf_mu_ndt"))

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pairs(dsamples, pars = c("transf_mu_alpha", "transf_mu_theta", "transf_mu_threshold", "transf_mu_ndt", 'sd_
alpha', 'sd_ndt', 'sd_theta', 'log-posterior"))

## Warning in par(usr): argument 1 does not name a graphical parameter

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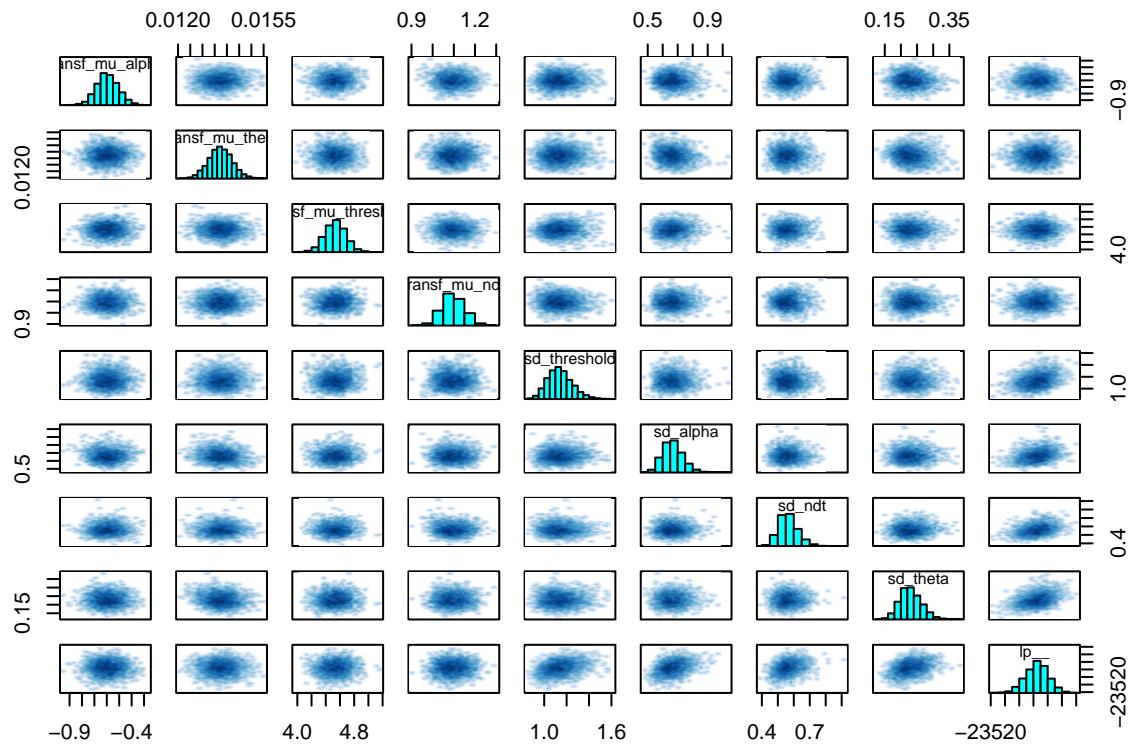
## Warning in par(usr): argument 1 does not name a graphical parameter

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print(dsamples, pars = c("transf_mu_alpha", "transf_mu_theta", "transf_mu_threshold", "transf_mu_ndt", "sd_threshold", "sd_alpha", "sd_ndt", "sd_theta", "lp_"))

## Inference for Stan model: MV_Baseline.
## 4 chains, each with iter=2000; warmup=1000; thin=1;
## post-warmup draws per chain=1000, total post-warmup draws=4000.
##
##                               mean se_mean    sd   2.5%   25%   50%
## transf_mu_alpha      -0.59    0.00  0.09  -0.76 -0.65 -0.60
## transf_mu_theta       0.01    0.00  0.00   0.01  0.01  0.01
## transf_mu_threshold  4.53    0.01  0.15  4.24  4.43  4.53
## transf_mu_ndt        1.10    0.00  0.05  1.00  1.06  1.09
## sd_threshold         1.14    0.00  0.11  0.95  1.06  1.13
## sd_alpha              0.67    0.00  0.07  0.55  0.62  0.66
## sd_ndt               0.56    0.00  0.06  0.46  0.52  0.56
## sd_theta              0.23    0.00  0.03  0.17  0.21  0.23
## lp_                  -23455.11 0.59 16.05 -23487.16 -23465.53 -23454.88
##                               75%    97.5% n_eff Rhat
## transf_mu_alpha      -0.54   -0.42   523  1.00
## transf_mu_theta       0.01    0.01  2385  1.00
## transf_mu_threshold  4.63    4.82  448  1.00
## transf_mu_ndt        1.13   1.20  472  1.02
## sd_threshold         1.21   1.38  759  1.00
## sd_alpha              0.71   0.81 1138  1.00
## sd_ndt               0.60   0.69 1160  1.00
## sd_theta              0.25   0.30 1392  1.00

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## lp__          -23444.13 -23424.77    732 1.00
##
## Samples were drawn using NUTS(diag_e) at Mon Oct  9 22:55:28 2023.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).

ratios_cp <- neff_ratio(dsamples, pars = c("transf_mu_alpha", "transf_mu_theta", "transf_mu_threshold", "sd_alpha", "sd_ndt", "sd_theta"))
df_ratios_cp <- as.data.frame(ratios_cp)
print(df_ratios_cp)

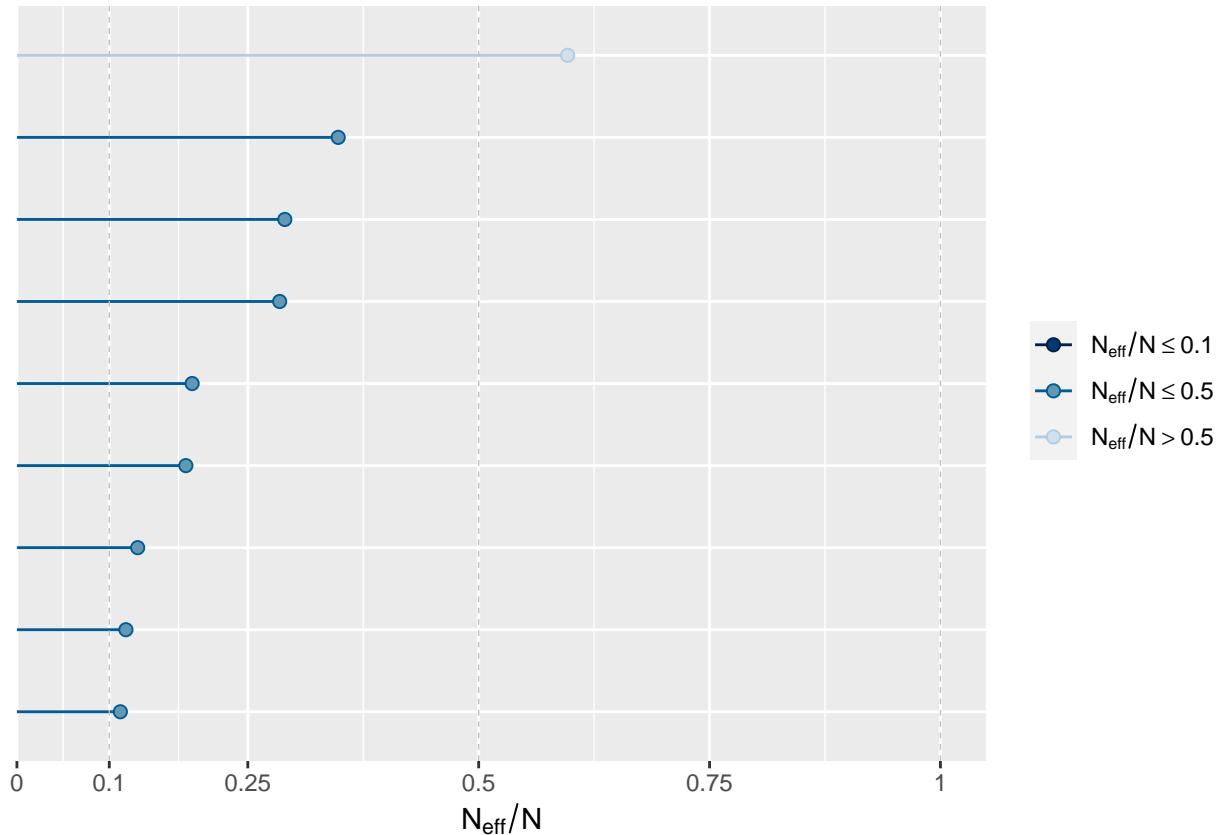
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##           ratios_cp
## transf_mu_alpha      0.1307550
## transf_mu_theta       0.5963391
## transf_mu_threshold  0.1120171
## transf_mu_ndt        0.1180074
## sd_threshold          0.1897305
## sd_alpha              0.2845123
## sd_ndt                0.2900805
## sd_theta               0.3479072
## lp__
## lp__                  0.1829276

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mcmc_neff(ratios_cp, size = 2)
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library(ggplot2)

# Assuming 'fit' is your fitted model
samples_matrix <- as.matrix(dsamples)
means <- colMeans(samples_matrix)
hpd_interval <- t(apply(samples_matrix, 2, function(x) quantile(x, probs=c(0.025, 0.975))))

parameters <- c("transf_mu_alpha", "transf_mu_theta", "transf_mu_threshold",
               "transf_mu_ndt", "sd_threshold", "sd_alpha", "sd_ndt", "sd_theta")

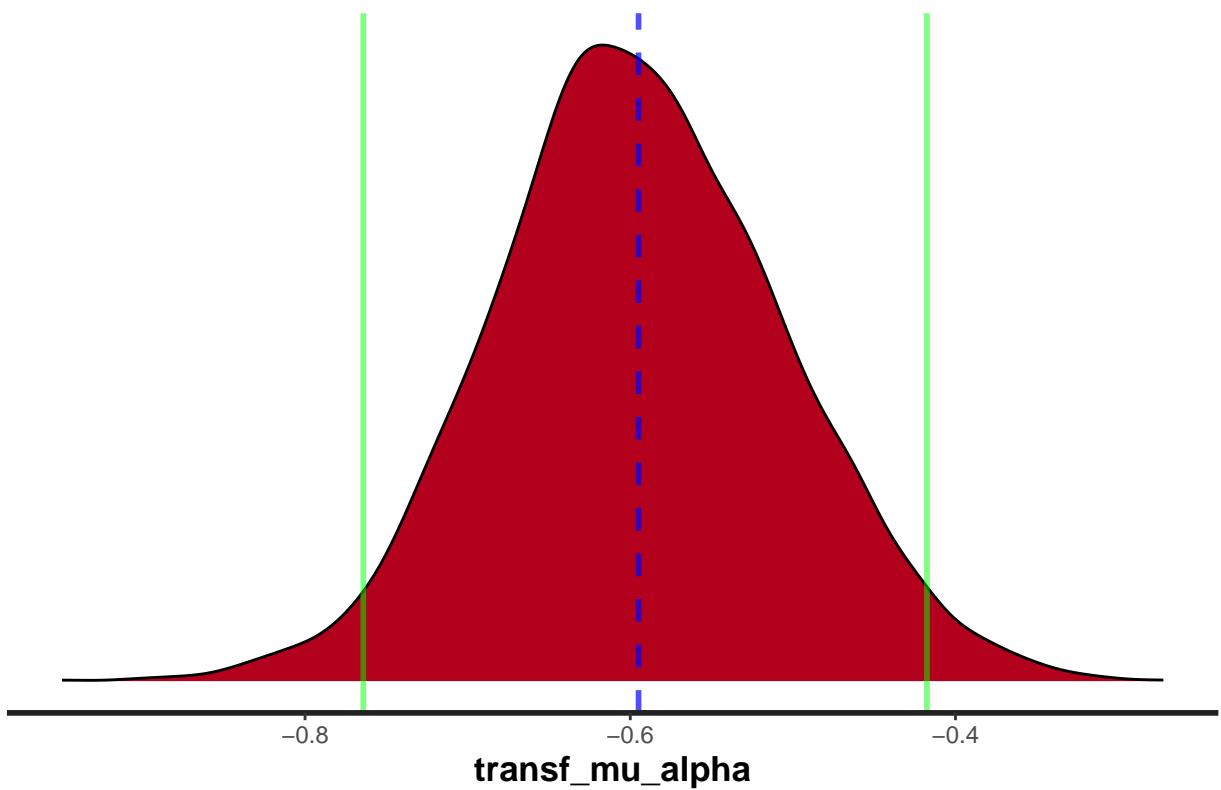
# Loop through each parameter and plot
for (param in parameters) {
  p <- stan_dens(dsamples, pars = param)
  p <- p +
    geom_vline(aes(xintercept=means[param]), color="blue", linetype="dashed", size=1, alpha=0.7, show.limits=FALSE)
    geom_vline(aes(xintercept=hpd_interval[param, 1]), color="green", linetype="solid", size=1, alpha=0.7, show.limits=FALSE)
    geom_vline(aes(xintercept=hpd_interval[param, 2]), color="green", linetype="solid", size=1, alpha=0.7, show.limits=FALSE)
  labs(title=paste("Posterior distribution of", param))

  print(p)
}

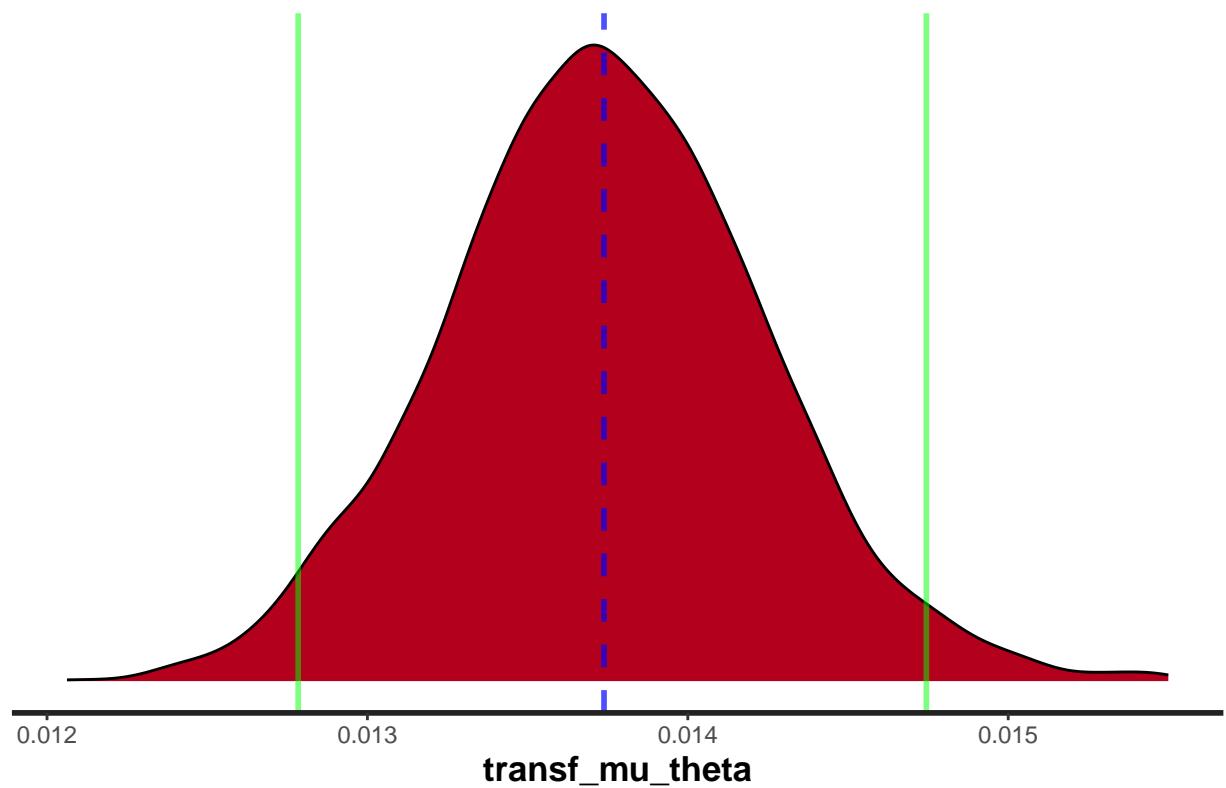
## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

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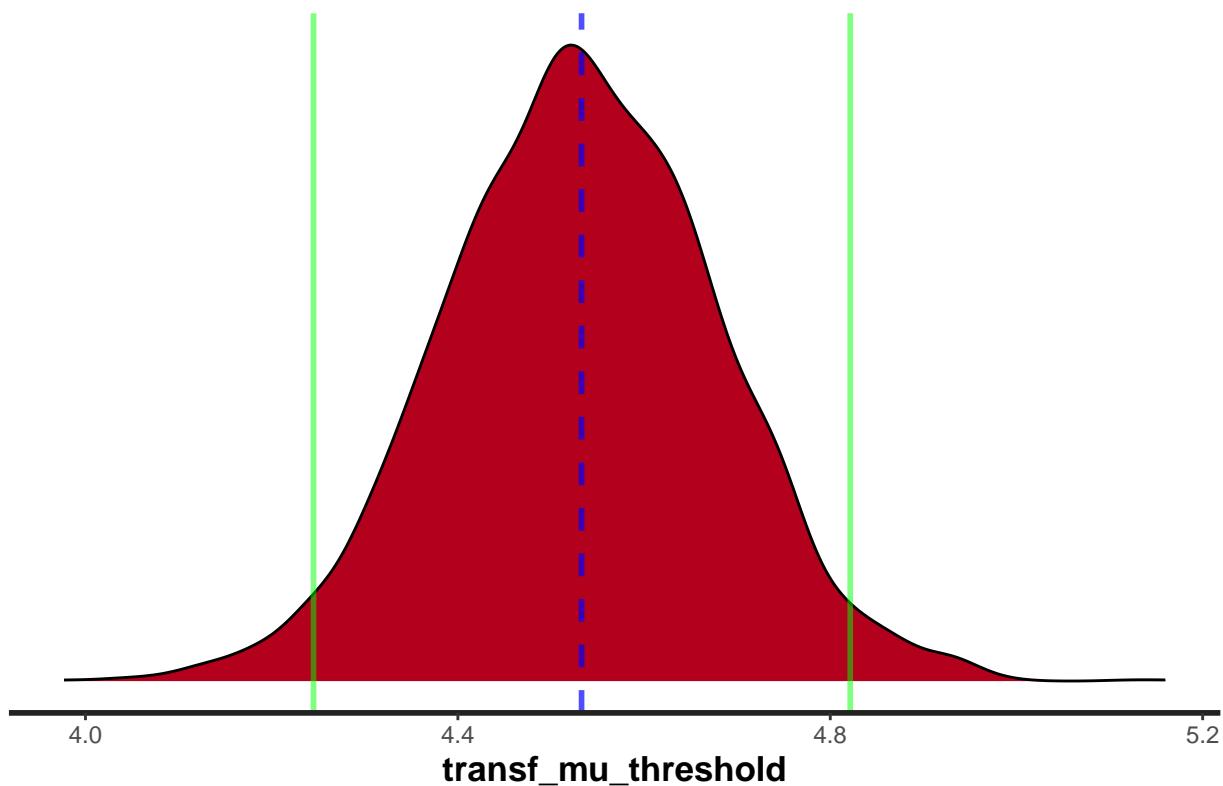
Posterior distribution of `transf_mu_alpha`



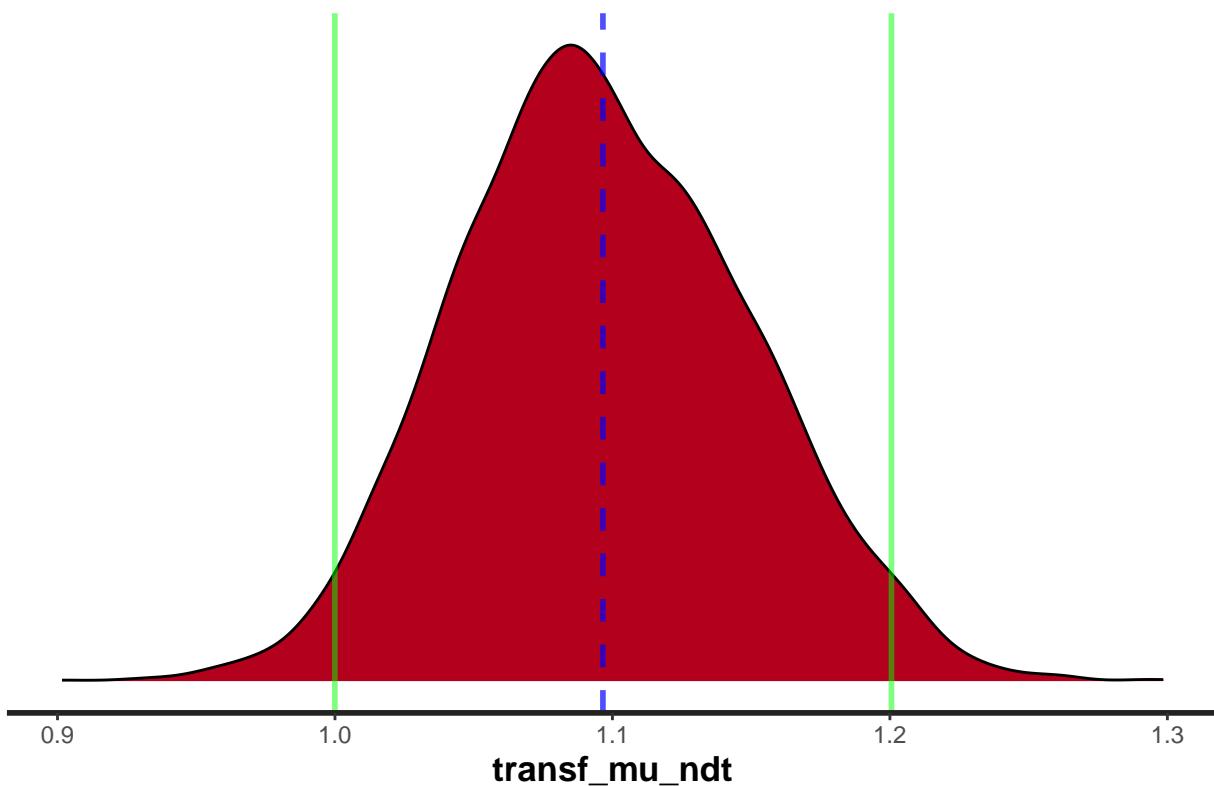
Posterior distribution of transf_mu_theta



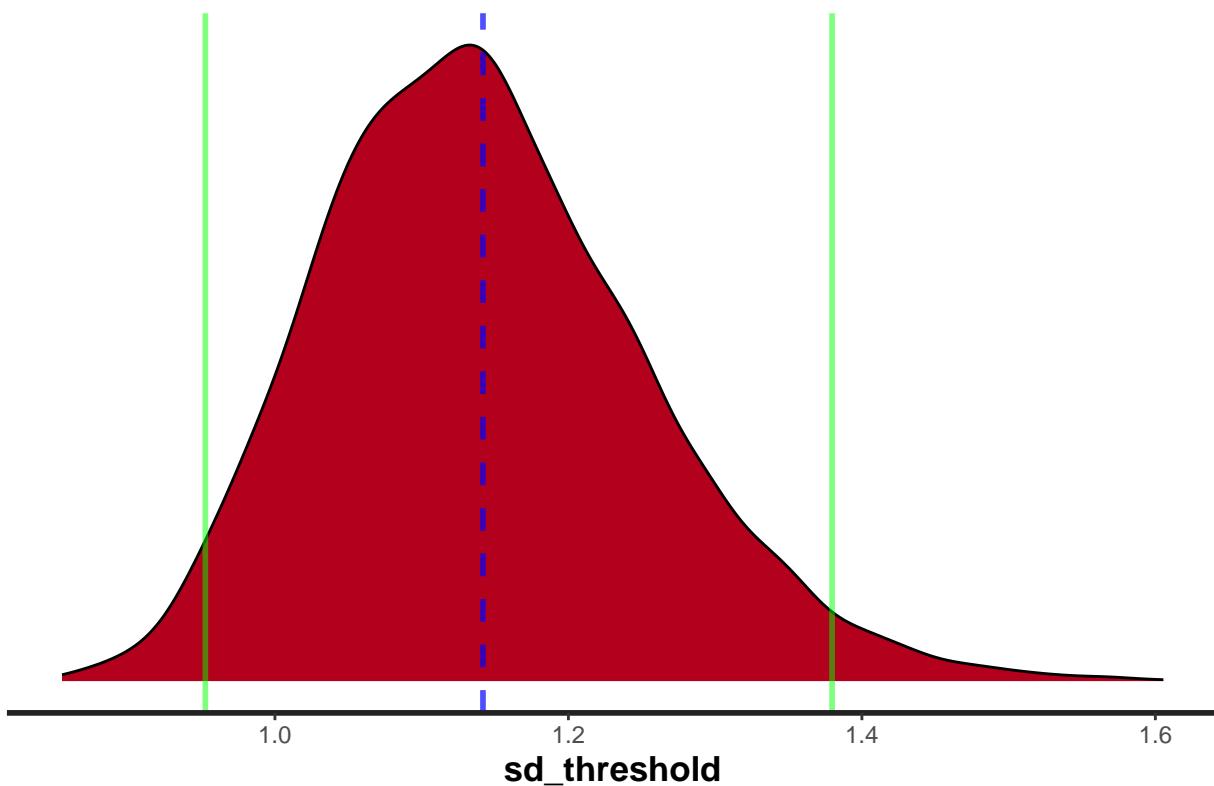
Posterior distribution of transf_mu_threshold



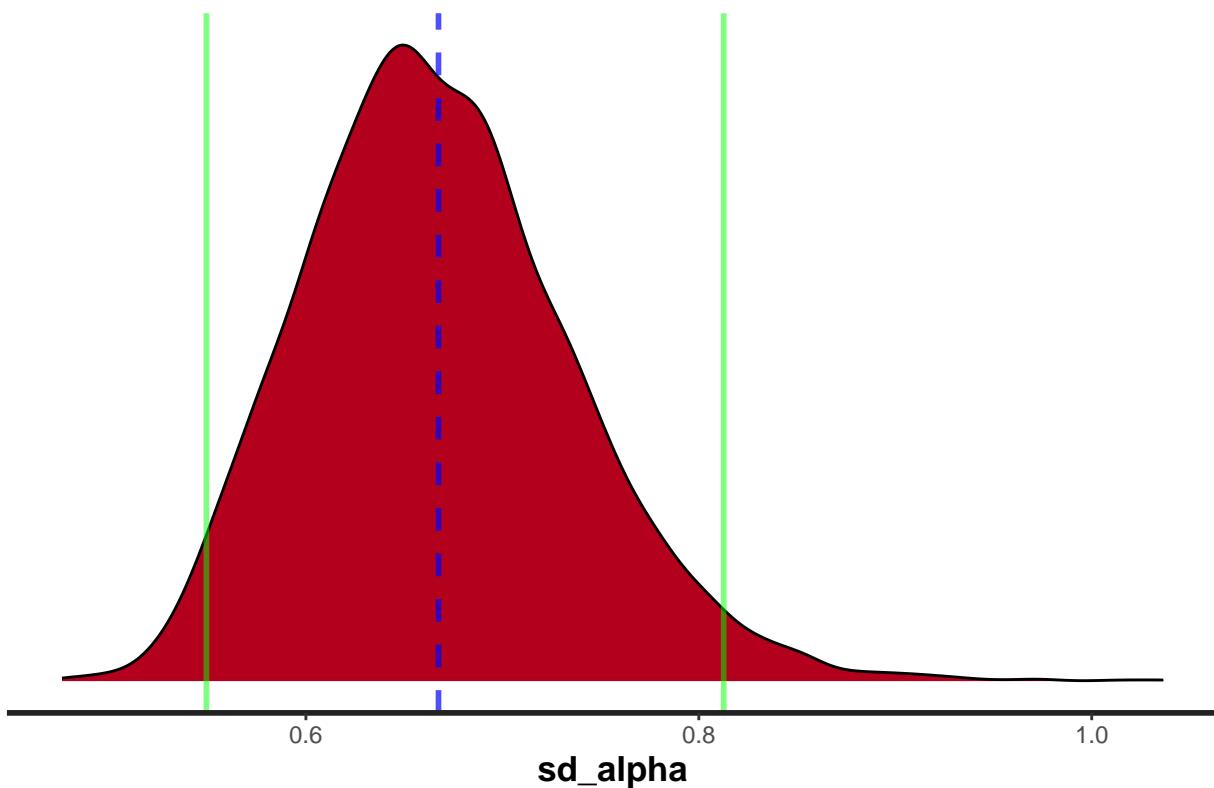
Posterior distribution of transf_mu_ndt



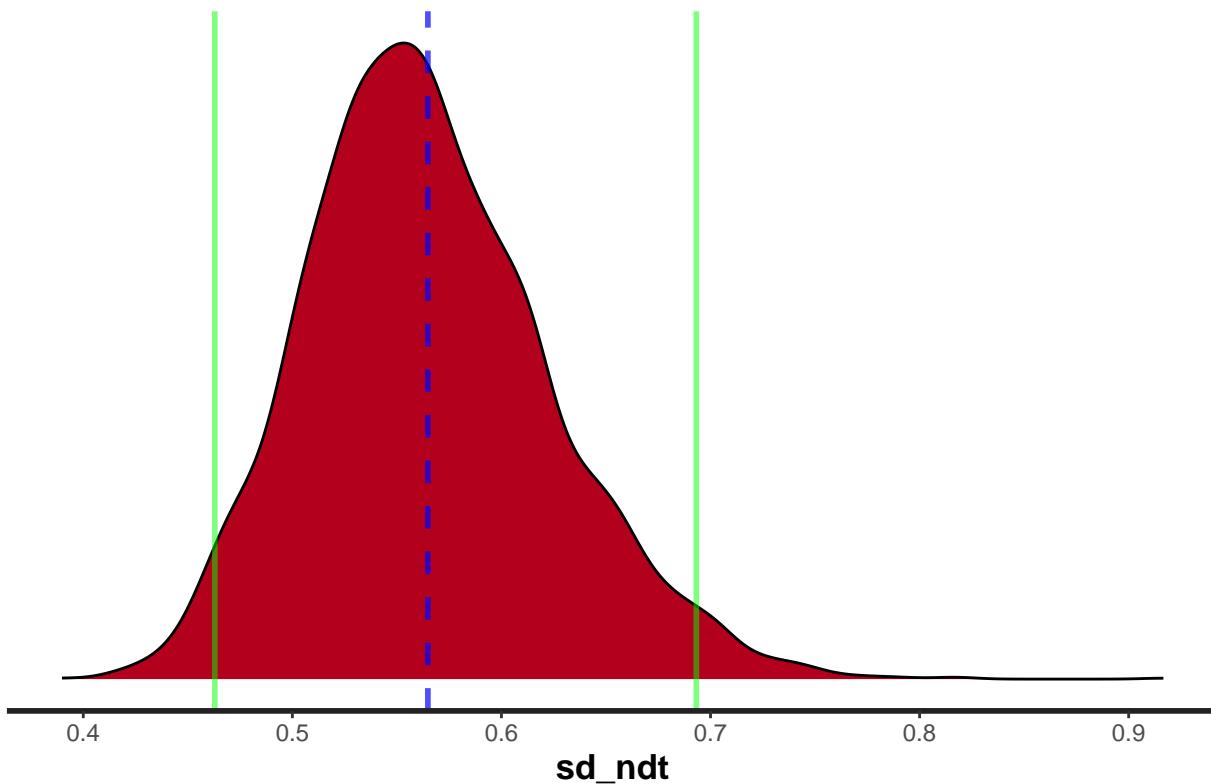
Posterior distribution of sd_threshold



Posterior distribution of sd_{α}



Posterior distribution of sd_ndt



Posterior distribution of sd_theta

