

```
##### 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: 2elf913d3ca3)
```

```
## 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
```

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':  
##  
## filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
## intersect, setdiff, setequal, union
```

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

dat <- read.csv('final_data.csv')
```

```
dat = dat %>%
  mutate(cho = 0,
         cho = ifelse(response == "f", 1*risk_index, cho),
         cho = ifelse(response == "j", -1*risk_index, cho))

dat <- dat %>%
  filter(skew != 'control')
```

```
ids <- unique(dat$subject)
for(j in 1:length(ids)){
  dat$tid[dat$subject==ids[j]] <- j
}
tids <- unique(dat$tid)

dat <- dat %>%
  filter(test_part == 'cc' | test_part == 'ss')
dat <- dat %>%
  mutate(con = ifelse(test_part == "cc", 1, -1))
dat$rt <- dat$rt/1000
```

```
# only condition no time pressure
dataList = list(cho = dat$cho, rt = dat$rt, participant = dat$tid, N=nrow(dat),
L = length(tids), starting_point=0.5, evd = dat$evd, sdd = dat$sdd, con = dat$con)
```

```

parameters = c("transf_mu_alpha","transf_mu_threshold","transf_mu_ndt", "transf_mu_theta",
"transf_mu_delta_theta", 'sd_threshold',"sd_alpha","sd_ndt", 'sd_theta', 'sd_delta_theta',
"alpha_sbj","threshold_sbj","ndt_sbj",'theta_sbj','delta_theta_sbj', "log_lik")

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, 5),
                          sd_threshold = runif(1, 0, 1),
                          mu_ndt = runif(1, -1.5, 0),
                          sd_ndt = runif(1, 0, 1),
                          mu_theta = runif(1,-20, 1),
                          sd_theta = runif(1,0,1),
                          mu_delta_theta = runif(1, -1, 1),
                          sd_delta_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),
                          z_delta_theta = runif(length(tids),-0.1,0.1)

    )
  }

  return(initList)
}

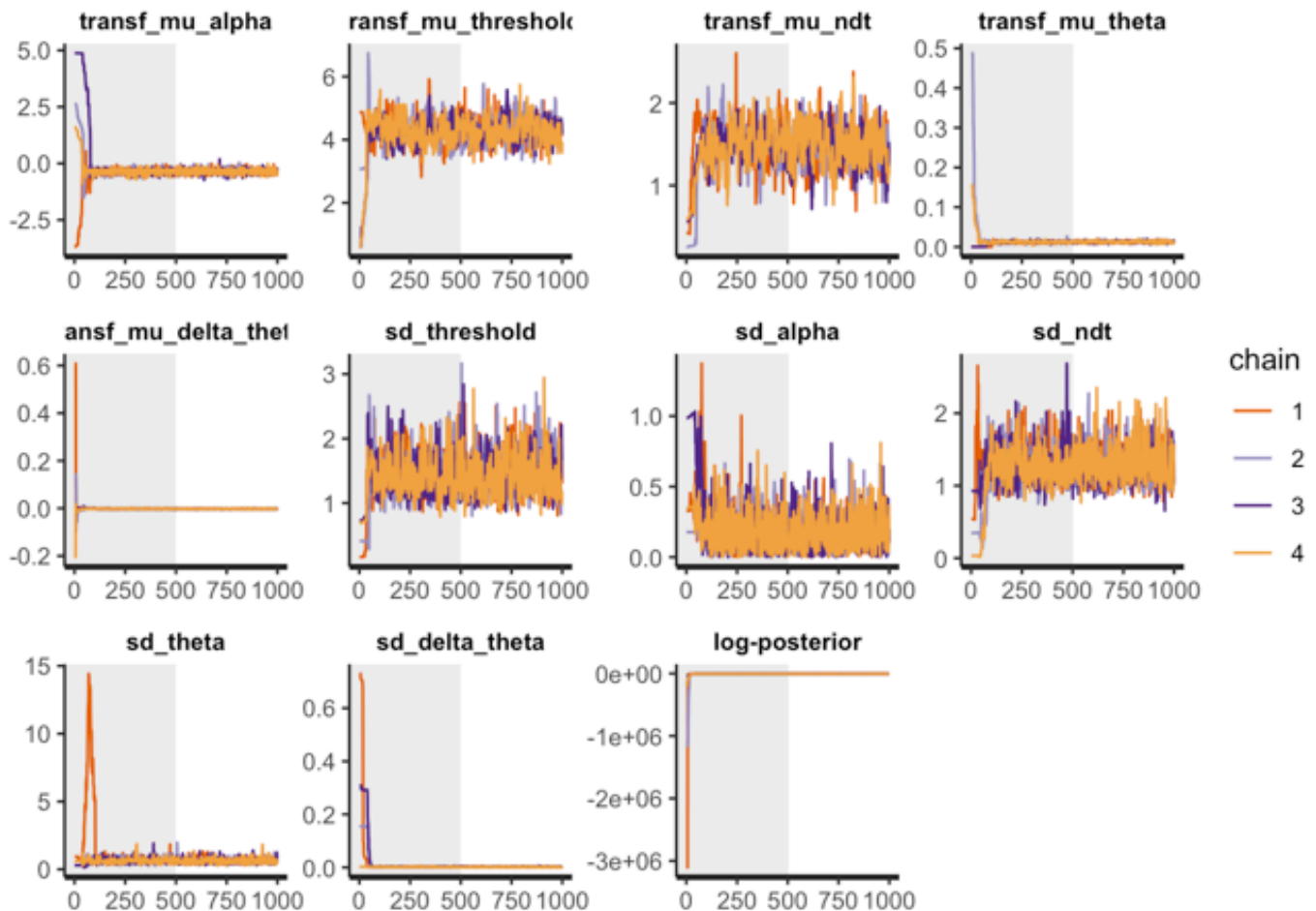
```

```

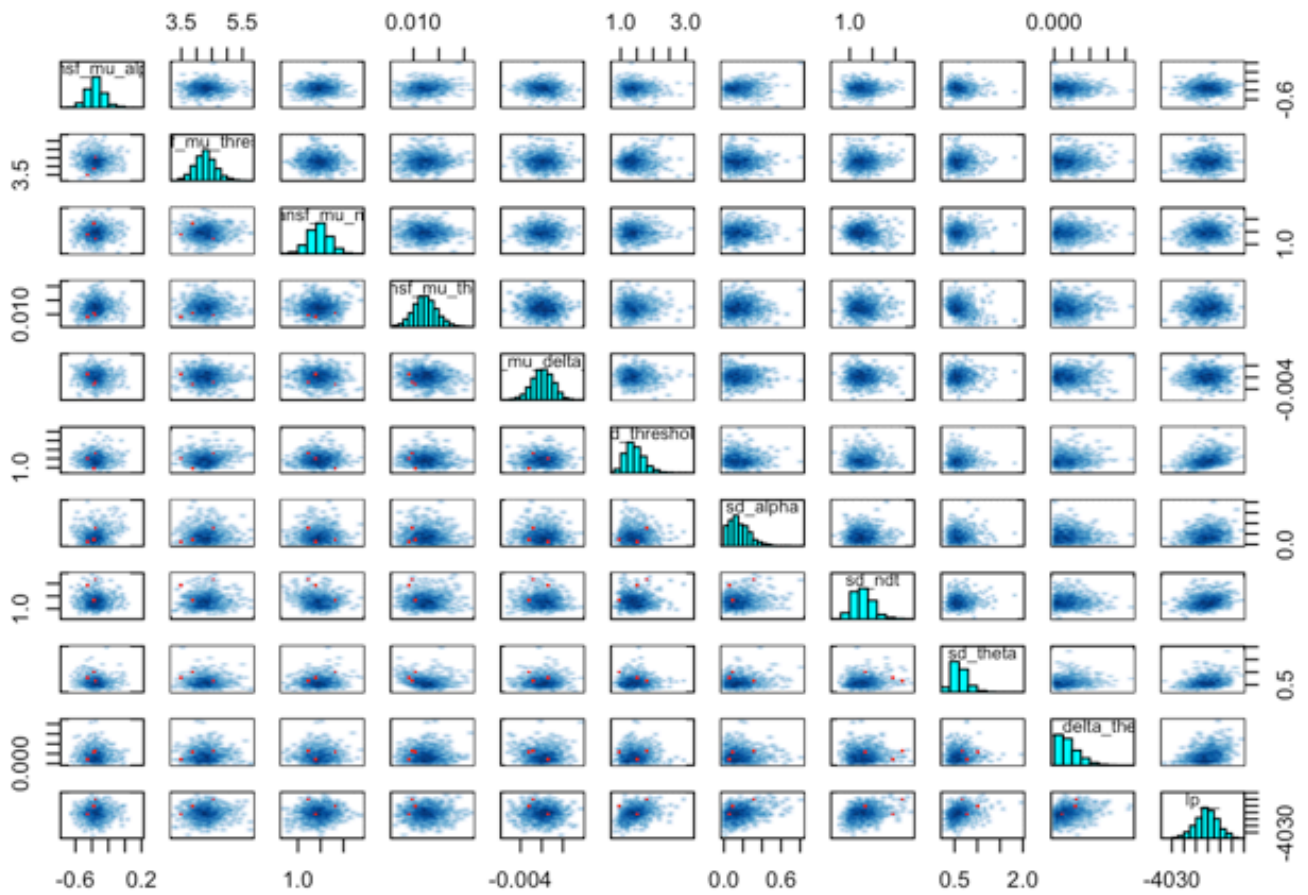
m <- stan_model("MV_Baseline.stan")
dsamples <- sampling(m,
  data=dataList,
  pars=parameters,
  iter=1000,
  chains=4,#If not specified, gives random inits
  init = initFunc(4),
  warmup = 500, # Stands for burn-in; Default = iter/2
  seed = 12, # Setting seed; Default is random seed
  refresh = 0
)

```

```
#parameters = c("transf_mu_alpha", "transf_mu_threshold", "transf_mu_ndt", "transf_mu_theta", "transf_mu_delta_theta", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta_theta', "alpha_sbj", "threshold_sbj", "ndt_sbj", 'theta_sbj', 'delta_theta_sbj', "log_lik")
rstan::traceplot(dsamples, pars=c("transf_mu_alpha", "transf_mu_threshold", "transf_mu_ndt", "transf_mu_theta", "transf_mu_delta_theta", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta_theta', "lp__"), inc_warmup = TRUE, nrow = 3)
```



```
pairs(dsamples, pars = c("transf_mu_alpha", "transf_mu_threshold", "transf_mu_ndt", "transf_mu_theta", "transf_mu_delta_theta", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta_theta', "lp__"))
```



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

```
## Inference for Stan model: MV_Baseline.
## 4 chains, each with iter=1000; warmup=500; thin=1;
## post-warmup draws per chain=500, total post-warmup draws=2000.
##
##               mean se_mean   sd    2.5%    25%    50%    75%
## transf_mu_alpha      -0.36    0.00 0.10   -0.55   -0.42   -0.36   -0.30
## transf_mu_threshold    4.29    0.02 0.36    3.62    4.05    4.28    4.50
## transf_mu_ndt         1.46    0.01 0.23    1.00    1.31    1.46    1.61
## transf_mu_theta        0.01    0.00 0.00    0.01    0.01    0.01    0.01
## transf_mu_delta_theta  0.00    0.00 0.00    0.00    0.00    0.00    0.00
## sd_threshold          1.44    0.01 0.30    0.97    1.22    1.40    1.62
## sd_alpha              0.17    0.00 0.11    0.01    0.08    0.15    0.23
## sd_ndt                1.29    0.01 0.23    0.93    1.13    1.27    1.43
## sd_theta              0.61    0.01 0.18    0.35    0.49    0.59    0.71
## sd_delta_theta        0.00    0.00 0.00    0.00    0.00    0.00    0.00
## lp__                 -4001.05    0.43 9.28 -4020.21 -4006.90 -4000.96 -3994.85
##               97.5% n_eff Rhat
## transf_mu_alpha      -0.14   1206 1.00
## transf_mu_threshold    5.07   293 1.01
## transf_mu_ndt         1.90   269 1.00
## transf_mu_theta        0.02   553 1.01
## transf_mu_delta_theta  0.00  1961 1.00
## sd_threshold          2.10   506 1.01
## sd_alpha              0.43   716 1.01
## sd_ndt                1.82   528 1.01
## sd_theta              1.04   593 1.01
## sd_delta_theta        0.00   719 1.00
## lp__                 -3983.38   458 1.00
##
## Samples were drawn using NUTS(diag_e) at Wed Nov 22 13:04:07 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).
```

```
library(ggplot2)
library(tidyverse) # for the gather function
```

```
## — Attaching core tidyverse packages — tidyverse 2.0.0 —
—
## ✓ forcats    1.0.0      ✓ stringr    1.5.0
## ✓ lubridate  1.9.2      ✓ tibble     3.1.8
## ✓ purrr      1.0.1      ✓ tidyr      1.3.0
## ✓ readr      2.1.4
## — Conflicts — tidyverse_conflicts() —
—
## ✖ tidyr::extract() masks rstan::extract()
## ✖ dplyr::filter()  masks stats::filter()
## ✖ dplyr::lag()     masks stats::lag()
## i Use the `conflicted::conflicted_package()` to force all conflicts to become errors
```

```
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", "transf_mu_delta_theta")

# Reshape data to a long format
df_long <- as.data.frame(samples_matrix) %>%
  gather(key = "parameter", value = "value", parameters)
```

```
## Warning: Using an external vector in selections was deprecated in tidyselect 1.
1.0.
## i Please use `all_of()` or `any_of()` instead.
##   # Was:
##   data %>% select(parameters)
##
##   # Now:
##   data %>% select(all_of(parameters))
##
## See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```

# Convert hpd_interval to a data frame and name the columns
hpd_interval_sub <- hpd_interval[parameters, ]
hpd_df <- as.data.frame(hpd_interval_sub)
colnames(hpd_df) <- c("lower", "upper")
rownames(hpd_df) <- parameters
hpd_df$parameter <- rownames(hpd_df)

# Aesthetic enhancements
theme_set(theme_minimal(base_size = 14)) # Set the default theme

custom_palette <- c("density_fill" = "lightgray",
                    "mean_line" = "blue",
                    "hpd_line" = "darkgreen")

# Add text labels for mean, lower, and upper HPD values
df_long <- df_long %>%
  group_by(parameter) %>%
  mutate(mean = means[parameter])

hpd_df <- hpd_df %>%
  mutate(mid = (lower + upper) / 2)

p <- ggplot(df_long, aes(x = value)) +
  geom_density(aes(fill = "density_fill")) +
  scale_fill_manual(values = custom_palette, guide = FALSE) +
  geom_vline(aes(xintercept = mean, color = "mean_line"), linetype = "dashed", size = 1, alpha = 0.7) +
  geom_text(data = df_long, aes(x = mean, y = 0, label = round(mean, 2)), vjust = -0.5, hjust = 0.5, size = 4, color = custom_palette["mean_line"]) +
  geom_vline(data = hpd_df, aes(xintercept = lower, color = "hpd_line"), linetype = "solid", size = 1, alpha = 0.5) +
  geom_text(data = hpd_df, aes(x = lower, y = 0, label = round(lower, 2)), vjust = -0.5, hjust = -0.5, size = 4, color = custom_palette["hpd_line"]) +
  geom_vline(data = hpd_df, aes(xintercept = upper, color = "hpd_line"), linetype = "solid", size = 1, alpha = 0.5) +
  geom_text(data = hpd_df, aes(x = upper, y = 0, label = round(upper, 2)), vjust = -0.5, hjust = 1.5, size = 4, color = custom_palette["hpd_line"]) +
  facet_wrap(~ parameter, scales = "free", ncol = 2) +
  scale_color_manual(values = custom_palette, guide = FALSE) +
  labs(title = "Posterior distributions")

```

```

## 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.

```



```
print(p)
```

```
## Warning: The `guide` argument in `scale_*()` cannot be `FALSE`. This was deprecated in
## ggplot2 3.3.4.
## i Please use "none" instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

Posterior distributions

