

```
##### 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
```

```
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$oa
3-dat$eva)^2*dat$pa3 + (dat$oa4-dat$eva)^2*dat$pa4)
dat$sdb = sqrt((dat$ob1-dat$evb)^2*dat$pb1 + (dat$ob2-dat$evb)^2*dat$pb2 + (dat$ob
3-dat$evb)^2*dat$pb3 + (dat$ob4-dat$evb)^2*dat$pb4)
dat$sdd = dat$sdb - dat$sda
dat$evdummy = ifelse(dat$evd>0,1,0)
```

```
# 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,]

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
```

```
rcontrol_dat <- rcontrol_dat %>%
  rowwise() %>%
  mutate(
    oa_condition = sum(c_across(starts_with("oa"))) == 0),
    ob_condition = sum(c_across(starts_with("ob"))) == 0)
  ) %>%
  filter(
    (oa_condition == 2 & ob_condition == 2) |
    (oa_condition == 0 & ob_condition == 0)
  )
```

rcontrol_dat

```
## # A tibble: 2,995 × 42
## # Rowwise:
##   date      id block order cond  trial gameid leftright  oa1  oa2  oa3  oa
4
##   <chr>  <int> <int> <int> <chr> <int>  <int>      <int> <int> <int> <int> <int>
>
## 1 2019_...    1     1     0 cont...    1    144      1     99    93    90    7
## 2 2019_...    1     1     0 cont...    3    121      1     42    51    47    7
## 3 2019_...    1     1     0 cont...    5    118      1     62    52    65    6
## 4 2019_...    1     1     0 cont...    8    126      0     16    55    51    4
## 5 2019_...    1     1     0 cont...    9    142      1     98    65    64    8
## 6 2019_...    1     1     0 cont...   10    129      0     12    29    81    3
## 7 2019_...    1     1     0 cont...   27    123      1     26    24    51    9
## 8 2019_...    1     1     0 cont...   29    122      1     35    58    25    3
## 9 2019_...    1     1     0 cont...   32    117      1     65    85    92    4
## 10 2019_...   1     1     0 cont...   37    139      0     44    31    16
##
## # i 2,985 more rows
## # i 30 more variables: pa1 <dbl>, pa2 <dbl>, pa3 <dbl>, pa4 <dbl>, ob1 <int>,
## # ob2 <int>, ob3 <int>, ob4 <int>, pb1 <dbl>, pb2 <dbl>, pb3 <dbl>,
## # pb4 <dbl>, choice <int>, rt <dbl>, paytrial <int>, payout <int>,
## # alter <int>, geschlecht <int>, comment <chr>, eva <dbl>, evb <dbl>,
## # evd <dbl>, sda <dbl>, sdb <dbl>, sdd <dbl>, evdummy <dbl>, cho <dbl>,
## # tid <int>, oa_condition <int>, ob_condition <int>
```

```
# only condition no time pressure
dataList = list(cho = rcontrol_dat$cho, rt = rcontrol_dat$rt, participant = rcontrol_dat$tid, N=nrow(rcontrol_dat), L = length(tids), starting_point=0.5, evd = rcontrol_dat$evd, sdd = rcontrol_dat$sdd, con = rep(1, length(rcontrol_dat$trial)))
```

```
parameters = c("transf_mu_alpha","transf_mu_threshold","transf_mu_ndt", "transf_mu_theta","transf_mu_delta", 'sd_threshold',"sd_alpha","sd_ndt", 'sd_theta', 'sd_delta', "alpha_sbj","threshold_sbj","ndt_sbj",'theta_sbj','delta_sbj', "log_lik")
```

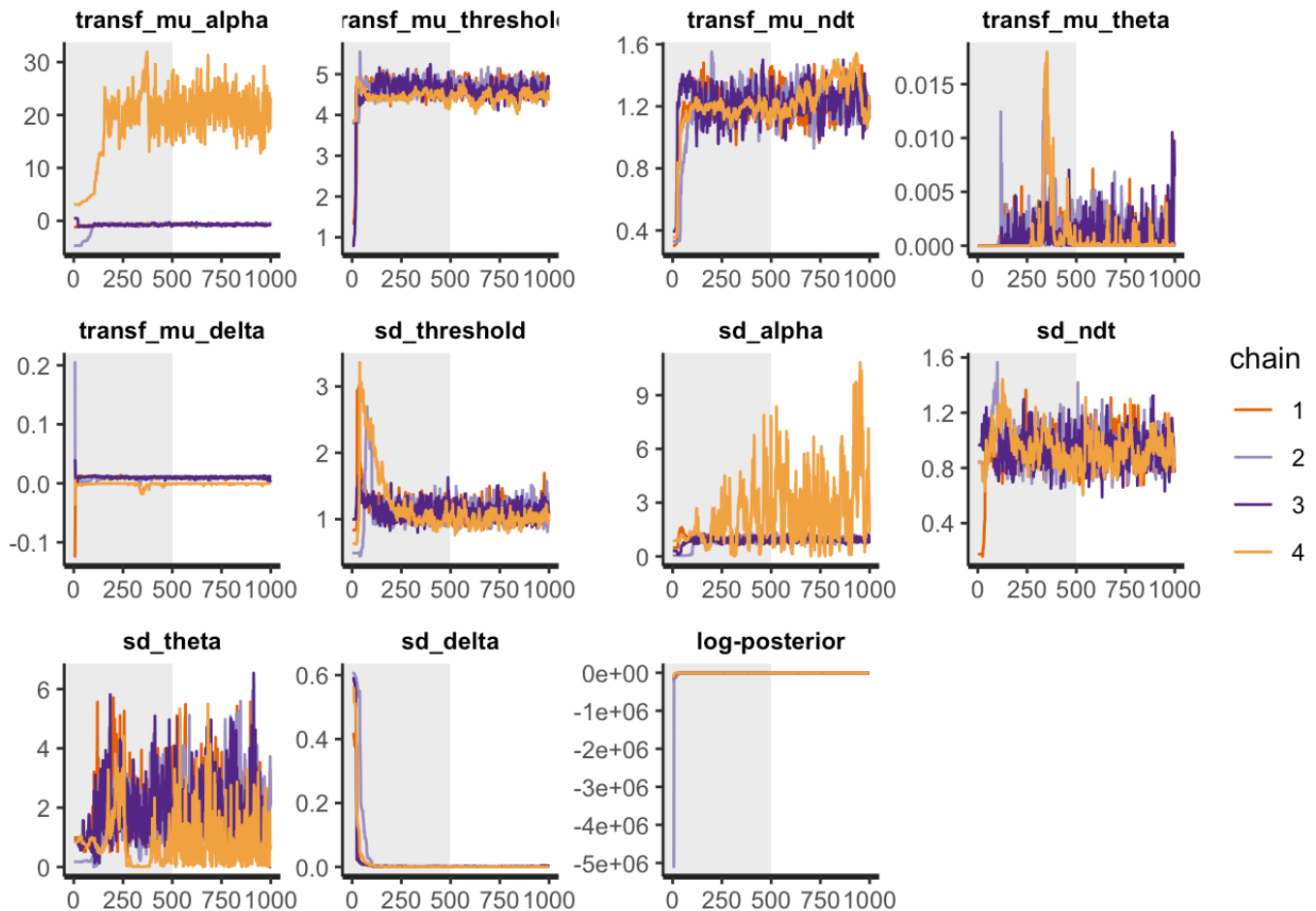
```
initFunc <-function (i) {
  initList=list()
  for (ll in 1:i){
    initList[[ll]] = list(mu_alpha = runif(1, -5, 5),
                          sd_alpha = runif(1,0,1),
                          mu_threshold = runif(1,-0.5, 5),
                          sd_threshold = runif(1, 0, 1),
                          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 = runif(1, -1, 1),
                          sd_delta = 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 = runif(length(tids),-0.1,0.1)

    )
  }

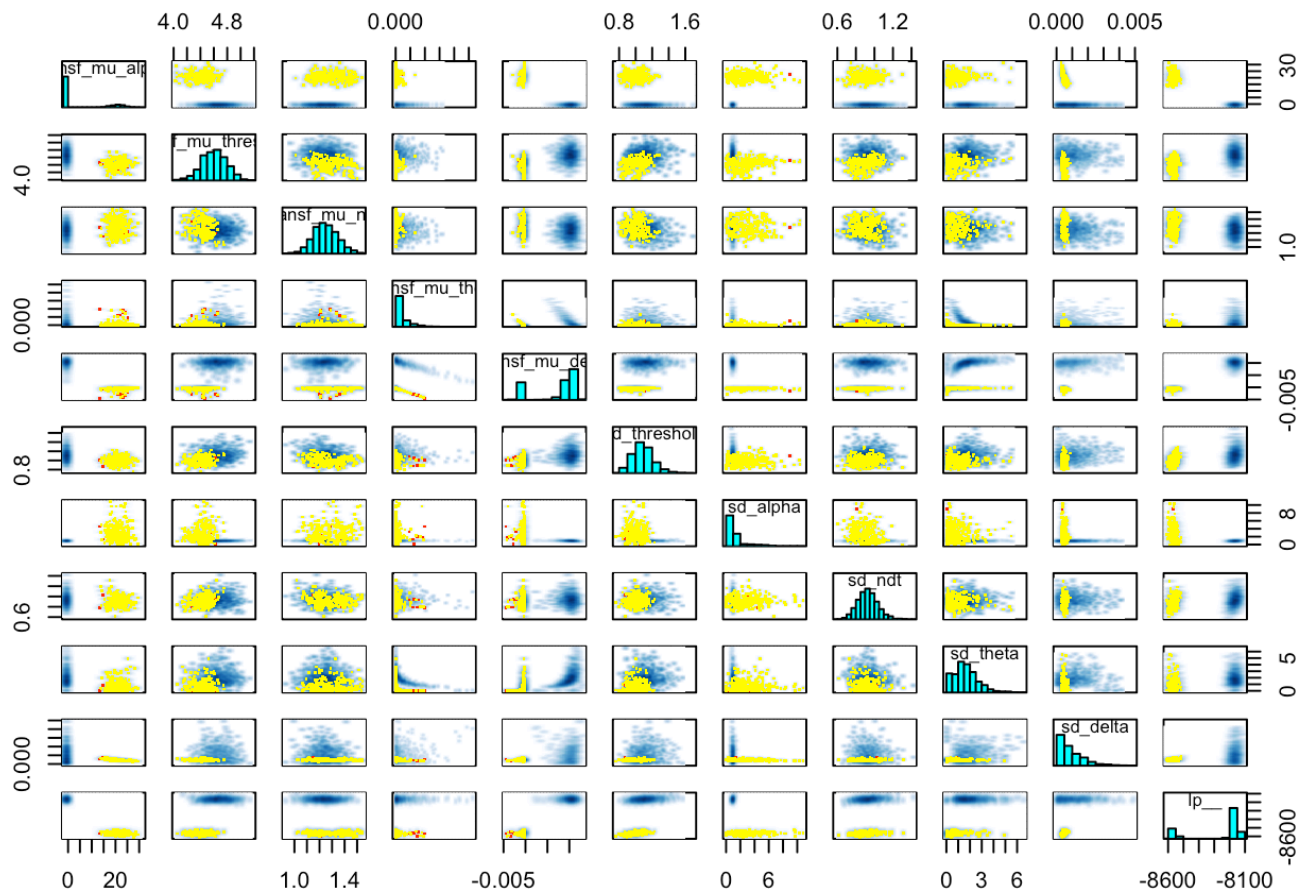
  return(initList)
}
```

```
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", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta', "alpha_sbj", "threshold_sbj", "ndt_sbj", 'theta_sbj', 'delta_sbj', "log_lik")
rstan::traceplot(dsamples, pars=c("transf_mu_alpha", "transf_mu_threshold", "transf_mu_ndt", "transf_mu_theta", "transf_mu_delta", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta', "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", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta', "lp__"))
```



```
print(dsamples, pars = c("transf_mu_alpha", "transf_mu_threshold", "transf_mu_ndt",
  "transf_mu_theta", "transf_mu_delta", 'sd_threshold', "sd_alpha", "sd_ndt", 'sd_theta', 'sd_delta', "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	4.71	6.62	9.47	-0.91	-0.74	-0.64	2.99
transf_mu_threshold	4.60	0.08	0.18	4.22	4.48	4.61	4.73
transf_mu_ndt	1.25	0.01	0.10	1.07	1.18	1.24	1.31
transf_mu_theta	0.00	0.00	0.00	0.00	0.00	0.00	0.00
transf_mu_delta	0.01	0.00	0.00	0.00	0.00	0.01	0.01
sd_threshold	1.09	0.03	0.12	0.88	1.00	1.08	1.16
sd_alpha	1.44	0.58	1.36	0.63	0.86	0.95	1.11
sd_ndt	0.93	0.01	0.11	0.73	0.86	0.92	1.00
sd_theta	1.66	0.31	1.05	0.04	0.95	1.56	2.23
sd_delta	0.00	0.00	0.00	0.00	0.00	0.00	0.00
lp__	-8265.09	121.82	173.05	-8586.26	-8296.69	-8172.84	-8157.84

```
##
##
```

	97.5%	n_eff	Rhat
transf_mu_alpha	24.68	2	6.67
transf_mu_threshold	4.95	5	1.31
transf_mu_ndt	1.45	45	1.11
transf_mu_theta	0.00	40	1.07
transf_mu_delta	0.01	2	4.69
sd_threshold	1.36	16	1.11
sd_alpha	5.59	5	1.36
sd_ndt	1.16	319	1.02
sd_theta	4.08	11	1.15
sd_delta	0.00	21	1.12
lp__	-8136.54	2	11.04

```
##
## Samples were drawn using NUTS(diag_e) at Thu Nov 16 04:09:39 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() —
—
## ✗ tidyrr::extract() masks rstan::extract()
## ✗ dplyr::filter()   masks stats::filter()
## ✗ dplyr::lag()      masks stats::lag()
## i Use the http://conflicted.r-lib.org/ 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")
```

```
# 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

