```
library(rstan)
library(dplyr)
options(mc.cores = parallel::detectCores())
rstan_options(auto_write = TRUE)
library(RWiener)
#original parameter values
th = 4.56
ndt = 1.32
beta = .5
theta = .01
alpha = -0.36
delta_theta = 0.002
dat <- read.csv('final data.csv')</pre>
dat <- dat %>%
  filter(test_part == 'cc' | test_part == 'ss',
         skew %in% c("rl", "lr", 'ns'),
         subject == '4xf5xlk2') %>%
  select(P_A1, O_A1, P_A2, O_A2, P_B1, O_B1, P_B2, O_B2, eva, evb, evd, sda, sdb,
sdd, test_part)
dat$con = ifelse(dat$test part == 'cc', 1, -1)
columns to modify <- c('P_A1', 'P_A2', 'P_B1', 'P_B2')</pre>
for(column in columns to modify) {
  dat[[column]] <- dat[[column]] / 100</pre>
}
dat$evd cat <- cut(dat$evd,
                   breaks = c(-Inf, -15, -5, 5, 15, Inf),
                   labels = c("-20", "-10", "0", "10", "20"),
                   right = FALSE)
dat$sdd_cat <- cut(dat$sdd,
                   breaks = c(-Inf, 6, 11, 16),
                   labels = c('5', "10", "15"),
                  right = FALSE)
stim <- dat[order(dat$evd_cat, dat$sdd_cat, dat$con), ]</pre>
stim$drift <- (theta + delta theta * stim$con) * (stim$evd + alpha * stim$sdd)</pre>
```

```
stim2 = stim
stim3 = stim

for(n in 1:nrow(stim2)){
    stim2$simchosum[n] = 0
}

stim4 = stim
for(n in 1:nrow(stim4)){
    stim4$simchosum[n] = 0
}
```

```
sim ddm <- "
data {
    int<lower=1> N;
                                                    // number of data items
                                                    // number of participants
    int<lower=1> L;
    // int<lower=1, upper=L> participant[N];
                                                        // level (participant)
                                               // accuracy (-1, 1)
    int<lower=-1,upper=1> cho[N];
                                                    // rt
    real<lower=0> rt[N];
   real evd[N];
    real sdd[N];
    real<lower=0, upper=1> starting point;
                                                   // starting point diffusion mo
del not to estimate
    int<lower=-1,upper=1> con[N];
}
parameters {
   real alpha_sbj;
   real theta_v;
   real threshold v;
   real ndt_v;
    real delta_theta;
}
transformed parameters {
    real drift ll[N];
                                                    // trial-by-trial drift rate f
or likelihood (incorporates accuracy)
    real drift_t[N];
                                                    // trial-by-trial drift rate f
or predictions
                                                    // trial-by-trial threshold
    real<lower=0> threshold t[N];
    real<lower=0> ndt_t[N];
                                                    // trial-by-trial ndt
    real<lower=0> theta sbj;
    real<lower=0> threshold sbj;
    real<lower=0> ndt_sbj;
```

```
theta_sbj = log(1 + exp(theta_v));
    threshold_sbj = log(1 + exp(threshold_v));
    ndt_sbj = log(1 + exp(ndt_v));
    for (n in 1:N) {
        drift_t[n] = (theta_sbj + delta_theta*con[n])* (evd[n] + alpha_sbj * sdd[
n]);
        drift_ll[n] = drift_t[n]*cho[n];
        threshold_t[n] = threshold_sbj;
        ndt t[n] = ndt sbj;
    }
}
model {
  alpha_sbj ~ normal(0, 5);
    theta_v ~ normal(1,5);
    threshold_v ~ normal(1,3);
    ndt v ~ normal(0,1);
    delta_theta ~ normal(0, 5);
    rt ~ wiener(threshold_t, ndt_t, starting_point, drift_ll);
}
generated quantities {
    vector[N] log_lik;
    {for (n in 1:N) {
        log_lik[n] = wiener_lpdf(rt[n] | threshold_t[n], ndt_t[n], starting_point,
drift_ll[n]);
    }
}
}
```

```
# Set the number of iterations
n iter <- 20
`%+=%` = function(e1,e2) eval.parent(substitute(e1 <- e1 + e2))
# Create empty vectors to store the outcome parameters for each iteration
th recover <- numeric(n iter)
theta_recover <- numeric(n_iter)</pre>
ndt_recover <- numeric(n_iter)</pre>
alpha recover <- numeric(n iter)</pre>
delta_theta_recover <- numeric(n_iter)</pre>
th_bias <- numeric(n_iter)</pre>
theta bias <- numeric(n iter)</pre>
ndt bias <- numeric(n iter)</pre>
alpha_bias <- numeric(n_iter)</pre>
delta_theta_bias <- numeric(n_iter)</pre>
th dev <- numeric(n iter)
theta_dev <- numeric(n_iter)</pre>
ndt_dev <- numeric(n_iter)</pre>
alpha_dev <- numeric(n_iter)</pre>
delta_theta_dev <- numeric(n_iter)</pre>
# Run the model for n_iter iterations
for (i in 1:n iter) {
  for(n in 1:nrow(stim)){
```

```
cres <- rwiener(1,th, ndt, beta, (theta + delta theta * stim$con[n]) * (stim$e</pre>
vd[n] + alpha * stim$sdd[n]))
    stim$simrt[n] <- as.numeric(cres[1])</pre>
    stim$simcho[n] <- ifelse(cres[2]=="upper",1,-1)</pre>
  }
  for(n in 1:nrow(stim2)){
    stim2$simchosum[n] %+=% ifelse(stim$simcho[n]==1,1,0)
    }
  parameters = c("alpha_sbj","threshold_sbj","ndt_sbj",'theta_sbj', 'delta_theta')
  dataList = list(cho = stim$simcho,rt = stim$simrt, N=90, L = 1, starting point
=0.5, evd = stim$evd, sdd = stim$sdd, con = stim$con)
  # Run the diffusion model for the current iteration
  dsamples <- stan(model code = sim ddm,
                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
                refresh = 0
  samples <- rstan::extract(dsamples, pars = c('alpha sbj', 'theta sbj', 'threshol</pre>
d_sbj', 'ndt_sbj', 'delta_theta'))
  # Store the outcome parameters for the current iteration
  th_recover[i] <- mean(samples$threshold_sbj)</pre>
  theta_recover[i] <- mean(samples$theta_sbj)</pre>
  ndt_recover[i] <- mean(samples$ndt_sbj)</pre>
  alpha recover[i] <- mean(samples$alpha sbj)</pre>
  delta_theta_recover[i] <- mean(samples$delta_theta)</pre>
  th bias[i] <- (mean(samples$threshold sbj)-th)/th</pre>
  theta bias[i] <- (mean(samples$theta sbj)-theta)/theta
  ndt_bias[i] <- (mean(samples$ndt_sbj)-ndt)/ndt</pre>
```

```
alpha_bias[i] <- (mean(samples$alpha_sbj)-alpha)/alpha
delta_theta_bias[i] <- (mean(samples$delta_theta)-delta_theta)/delta_theta

th_dev[i] <- abs(mean(samples$threshold_sbj)-th)/th
theta_dev[i] <- abs(mean(samples$theta_sbj)-theta)/theta
ndt_dev[i] <- abs(mean(samples$ndt_sbj)-ndt)/ndt
alpha_dev[i] <- abs(mean(samples$alpha_sbj)-alpha)/alpha
delta_theta_dev[i] <- abs(mean(samples$delta_theta)-delta_theta)/delta_theta</pre>
}
```

Trying to compile a simple C file

```
## Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
## clang -mmacosx-version-min=10.13 -I"/Library/Frameworks/R.framework/Resources/i
nclude" -DNDEBUG
                   -I"/Library/Frameworks/R.framework/Versions/4.2/Resources/libra
ry/Rcpp/include/" -I"/Library/Frameworks/R.framework/Versions/4.2/Resources/libra
ry/RcppEigen/include/" -I"/Library/Frameworks/R.framework/Versions/4.2/Resources/
library/RcppEigen/include/unsupported" -I"/Library/Frameworks/R.framework/Version
s/4.2/Resources/library/BH/include" -I"/Library/Frameworks/R.framework/Versions/4.
2/Resources/library/StanHeaders/include/src/" -I"/Library/Frameworks/R.framework/
Versions/4.2/Resources/library/StanHeaders/include/" -I"/Library/Frameworks/R.fra
mework/Versions/4.2/Resources/library/RcppParallel/include/" -I"/Library/Framewor
ks/R.framework/Versions/4.2/Resources/library/rstan/include" -DEIGEN NO DEBUG -DB
OOST DISABLE ASSERTS -DBOOST PENDING INTEGER LOG2 HPP -DSTAN THREADS -DBOOST NO
_AUTO_PTR -include '/Library/Frameworks/R.framework/Versions/4.2/Resources/librar
y/StanHeaders/include/stan/math/prim/mat/fun/Eigen.hpp' -D_REENTRANT -DRCPP_PARAL
LEL USE TBB=1
               -I/usr/local/include
                                     -fPIC -Wall -g -O2 -c foo.c -o foo.o
## In file included from <built-in>:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.2/Resources/li
brary/StanHeaders/include/stan/math/prim/mat/fun/Eigen.hpp:13:
## In file included from /Library/Frameworks/R.framework/Versions/4.2/Resources/li
brary/RcppEigen/include/Eigen/Dense:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.2/Resources/li
brary/RcppEigen/include/Eigen/Core:88:
## /Library/Frameworks/R.framework/Versions/4.2/Resources/library/RcppEigen/includ
e/Eigen/src/Core/util/Macros.h:628:1: error: unknown type name 'namespace'
## namespace Eigen {
## ^
## /Library/Frameworks/R.framework/Versions/4.2/Resources/library/RcppEigen/includ
e/Eigen/src/Core/util/Macros.h:628:16: error: expected ';' after top level declara
## namespace Eigen {
##
##
## In file included from <built-in>:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.2/Resources/li
brary/StanHeaders/include/stan/math/prim/mat/fun/Eigen.hpp:13:
## In file included from /Library/Frameworks/R.framework/Versions/4.2/Resources/li
brary/RcppEigen/include/Eigen/Dense:1:
## /Library/Frameworks/R.framework/Versions/4.2/Resources/library/RcppEigen/includ
e/Eigen/Core:96:10: fatal error: 'complex' file not found
## #include <complex>
##
## 3 errors generated.
## make: *** [foo.o] Error 1
```

```
#create a summary df of all parameters
df summary <- data.frame(original th = th,</pre>
                 recovered_th = th_recover,
                 bias_th = th_bias,
                 deviation th = th dev,
                 original theta = theta,
                 recovered_theta = theta_recover,
                 bias theta = theta bias,
                 deviation theta = theta dev,
                 original ndt = ndt,
                 recovered ndt = ndt recover,
                 bias ndt = ndt bias,
                 deviation_ndt = ndt_dev,
                 original alpha = alpha,
                 recovered alpha = alpha recover,
                 bias_alpha = alpha_bias,
                 deviation_alpha = alpha_dev,
                 original delta theta = delta theta,
                 recovered_delta_theta = delta_theta_recover,
                 bias_delta_theta = delta_theta_bias,
                 deviation_delta_theta = delta_theta_dev
                 )
```

```
##
      parameter true_value mean_recovered
                                              mean_bias mean_deviation
## 1
             th
                      4.560
                               4.562158642 0.0004733865
                                                             0.04242890
## 2
           theta
                      0.010
                               0.007345112 - 0.2654887788
                                                             0.40999227
## 3
            ndt
                     1.320
                              1.268747189 -0.0388278874
                                                             0.09423101
                             -0.540735721 0.5020436697
## 4
           alpha
                     -0.360
                                                            -1.20532921
## 5 delta_theta
                      0.002
                               0.001622808 -0.1885957955
                                                             0.83556092
```

```
##
       parameter true value median recovered
## 1
              th
                       4.560
                                   4.622587130
## 2
           theta
                       0.010
                                   0.007467987
## 3
             ndt
                       1.320
                                   1.246827249
           alpha
                      -0.360
                                  -0.557378886
## 4
## 5 delta theta
                       0.002
                                   0.001397025
```

```
#check whether the risky choice proportion can be successfully recovered by the me
an-variance model
#firstly, use recovered parameter values to simulation choice data
for (i in 1:n_iter) {

   for(n in 1:nrow(stim3)){
        cres <- rwiener(1,mean(df_summary$recovered_th), mean(df_summary$recovered_n
dt), beta, (mean(df_summary$recovered_theta) + mean(df_summary$recovered_delta_the
ta)* stim3$con[n] ) * (stim3$evd[n] + mean(df_summary$recovered_alpha) * stim3$sd
d[n]))
        stim3$simrt[n] <- as.numeric(cres[1])
        stim3$simcho[n] <- ifelse(cres[2]=="upper",1,-1)

}
for(n in 1:nrow(stim4)){
        stim4$simchosum[n] %+=% ifelse(stim3$simcho[n]==1,1,0)
      }
}</pre>
```

```
library(ggplot2)
ggplot(subset_data, aes(x = factor(trial), y = value, fill = type, colour = type))
+
geom_bar(stat = "identity", position = "dodge")+
ylim(0,1)
```

