selective_play

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```
\# Import\ packages
\#Import\ packages
setwd("~/Desktop/political play in WoT/")
library(brms)
## Loading required package: Rcpp
## Loading 'brms' package (version 2.16.1). Useful instructions
## can be found by typing help('brms'). A more detailed introduction
## to the package is available through vignette('brms_overview').
##
## Attaching package: 'brms'
## The following object is masked from 'package:stats':
##
##
       ar
library(broom.mixed)
library(car)
## Loading required package: carData
library(corrplot)
## corrplot 0.90 loaded
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:car':
##
##
       recode
```

```
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(emmeans)
library(ggplot2)
library(glmmTMB)
## Warning in checkDepPackageVersion(dep_pkg = "TMB"): Package version inconsistency detected.
## glmmTMB was built with TMB version 1.7.21
## Current TMB version is 1.7.22
## Please re-install glmmTMB from source or restore original 'TMB' package (see '?reinstalling' for mor
library(interactions)
library(lme4)
## Loading required package: Matrix
##
## Attaching package: 'lme4'
## The following object is masked from 'package:brms':
##
##
       ngrps
library(lmerTest)
##
## Attaching package: 'lmerTest'
## The following object is masked from 'package:lme4':
##
##
       lmer
## The following object is masked from 'package:stats':
##
##
       step
library(modelsummary)
library(MuMIn)
##
## Attaching package: 'MuMIn'
## The following object is masked from 'package:brms':
##
##
       100
```

```
library(optimx)
library(plyr)
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
## -----
##
## Attaching package: 'plyr'
## The following objects are masked from 'package:dplyr':
##
##
      arrange, count, desc, failwith, id, mutate, rename, summarise,
##
      summarize
library(psych)
##
## Attaching package: 'psych'
## The following object is masked from 'package:modelsummary':
##
##
      SD
## The following objects are masked from 'package:ggplot2':
##
##
      %+%, alpha
## The following object is masked from 'package:car':
##
##
      logit
## The following object is masked from 'package:brms':
##
##
      cs
library(r2mlm)
## Loading required package: nlme
## Attaching package: 'nlme'
## The following object is masked from 'package:optimx':
##
##
      coef<-
```

```
## The following object is masked from 'package:lme4':
##
       lmList
##
## The following object is masked from 'package:dplyr':
##
       collapse
## Registered S3 methods overwritten by 'parameters':
##
     method
                                      from
##
     as.double.parameters_kurtosis
                                      datawizard
##
     as.double.parameters_skewness
                                      datawizard
     as.double.parameters_smoothness datawizard
##
##
     as.numeric.parameters_kurtosis datawizard
##
     as.numeric.parameters_skewness datawizard
##
     as.numeric.parameters smoothness datawizard
##
     print.parameters_distribution datawizard
##
    print.parameters_kurtosis
                                     datawizard
##
    print.parameters_skewness
                                     datawizard
##
     summary.parameters_kurtosis
                                    datawizard
     summary.parameters_skewness
##
                                     datawizard
library(readr)
library(reshape2)
library(sjPlot)
## Learn more about sjPlot with 'browseVignettes("sjPlot")'.
library(splines)
library(stringi)
library(tidyr)
##
## Attaching package: 'tidyr'
## The following object is masked from 'package:reshape2':
##
##
       smiths
## The following objects are masked from 'package:Matrix':
##
##
       expand, pack, unpack
library(tidyverse)
## -- Attaching packages -----
                                                    ----- tidyverse 1.3.1 --
## v tibble 3.1.4
                     v stringr 1.4.0
## v purrr
                      v forcats 0.5.1
           0.3.4
```

```
masks ggplot2::%+%()
## x psych::%+%()
                       masks ggplot2::alpha()
## x psych::alpha()
## x plyr::arrange()
                       masks dplyr::arrange()
## x nlme::collapse() masks dplyr::collapse()
## x purrr::compact() masks plyr::compact()
## x plyr::count()
                       masks dplyr::count()
## x tidyr::expand()
                       masks Matrix::expand()
## x plyr::failwith() masks dplyr::failwith()
## x dplyr::filter() masks stats::filter()
## x plyr::id()
                      masks dplyr::id()
## x dplyr::lag()
                      masks stats::lag()
## x plyr::mutate() masks dplyr::mutate()
## x tidyr::pack() masks Matrix::pack()
## x dplyr::recode() masks car::recode()
## x plyr::rename()
                       masks dplyr::rename()
## x purrr::some()
                       masks car::some()
## x plyr::summarise() masks dplyr::summarise()
## x plyr::summarize() masks dplyr::summarize()
## x tidyr::unpack()
                       masks Matrix::unpack()
theme_set(theme_bw())
#increase the max print, and turn off scientific notation
options(scipen=999)
options(max.print=99999999)
data <- read.csv("data_ica.csv")</pre>
data_long <- read.csv("data_long.csv")</pre>
m01 <- lmer(us_pct ~ military_y + liberal + conser + (1 | id), data = data_long)
m01.1 <- lmer(us_pct ~ age.z + inc.z + edu.z</pre>
              + rating_t1.z + unlock.z + battle.z + week
              + military_y + liberal + conser
              + (1 | id), data = data_long)
m02 <- lmer(cn_pct ~ military_y + liberal + conser + (1 | id), data = data_long)
m02.1 <- lmer(cn_pct ~ age.z + inc.z + edu.z
              + rating t1.z + unlock.z + battle.z + week
              + military_y + liberal + conser
              + (1 | id), data = data_long)
m03 <- lmer(ussr_pct ~ military_y + liberal + conser + (1 | id), data = data_long)
m03.1 <- lmer(ussr_pct ~ age.z + inc.z + edu.z</pre>
              + rating_t1.z + unlock.z + battle.z + week
              + military_y + liberal + conser
              + (1 | id), data = data_long)
msummary(list(
  "U.S. vehicles (Baseline Model 1)" = m01,
 "U.S.S.R. vehicles (Baseline Model 2)" = m03,
```

----- tidyverse_conflicts() --

-- Conflicts -----

Table 1: Multilevel Models Predicting Vehicle Selection

	U.S. vehicles (Baseline Model 1)	U.S.S.R. vehicles (Baseline Model 2)	Chinese vehicles (Baseline M
(Intercept)	0.254***	0.186***	0.030***
	(0.009)	(0.006)	(0.002)
$military_y$	0.043***	-0.020**	-0.006*
	(0.010)	(0.007)	(0.002)
liberal	-0.037**	0.000	0.008*
	(0.014)	(0.010)	(0.003)
conser	-0.006	0.012	0.000
	(0.011)	(0.008)	(0.003)
sd (Intercept)	0.211	0.141	0.048
sd _Observation	0.158	0.143	0.065
age.z			
inc.z			
edu.z			
rating_t1.z			
unlock.z			
battle.z			
week			
AIC	-15782.8	-22376.4	-64479.0
BIC	-15733.7	-22327.3	-64429.9
Log.Lik.	7897.390	11 194.178	32 245.478
REMLcrit	-15794.781	-22388.357	-64490.955

⁺ p < 0.1, * p < 0.05, ** p < 0.01, *** p < 0.001

```
"Chinese vehicles (Baseline Model 3)" = m02,
"U.S. vehicles (Full Model 4)" = m01.1,
"U.S.S.R. vehicles (Full Model 5)" = m03.1,
"Chinese vehicles (Full Model 6)" = m02.1),
title = 'Multilevel Models Predicting Vehicle Selection',
stars = TRUE,
statistic = c('std.error')
)
```

Warning: In version 0.8.0 of the 'modelsummary' package, the default significance markers produced by ## This warning is displayed once per session.

```
m_gca1 <- glmmTMB(login_hr.log ~ week + (week | id), #general</pre>
                                                    data = data_long)
m_pw1 <- glmmTMB(login_hr.log ~ phase1 + phase2 + phase3 + (phase1 + phase2 + phase3 | id),
                                              data = data long)
anova(m_gca1, m_pw1)
## Data: data_long
## Models:
## m_gca1: login_hr.log ~ week + (week | id), zi=~0, disp=~1
## m_pw1: login_hr.log ~ phase1 + phase2 + phase3 + (phase1 + phase2 + , zi=~0, disp=~1
                                   phase3 | id), zi=~0, disp=~1
## m_pw1:
                                            AIC
                                                              BIC logLik deviance Chisq Chi Df
                                                                                                                                                                                                  Pr(>Chisq)
                           Df
## m_gca1 6 104616 104667 -52302
                                                                                                    104604
                                                                                                      99678 4926 9 < 0.00000000000000022 ***
## m_pw1 15 99708 99836 -49839
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
msummary(list(
     "Growth Curve Model" = m gca1,
     "Piecewise Growth Model" = m_pw1),
    title = 'Multilevel Growth Curve Models Predicting Gaming Behavors',
     stars = TRUE,
     statistic = c('std.error')
#Model equations for models in RQ2
Level 1:
                                      Outcome_{ti} = \beta_{0i} + \beta_{1i}US\_pmc_{ti} + \beta_{2i}USSR\_pmc_{ti} + \beta_{3i}China\_pmc_{ti} + e_{ti}
Level 2:
                    \beta_{0i} = \gamma_{00} + \gamma_{01} \text{US\_pm}_i + \gamma_{02} \text{USSR\_pm}_i + \gamma_{03} \text{China\_pm}_i + \gamma_{04} \text{Ideology}_i + \gamma_{05} \text{Military}_i + \gamma_{05}
                                                                                                                                              \gamma_{06}Phase1_i + \gamma_{07}Phase2_i + \gamma_{08}Phase3_i +
                                                                                                                                           \gamma_{09}Rating<sub>i</sub> + \gamma_{10}Unlocked<sub>i</sub> + \gamma_{11}Battle<sub>i</sub>+
                                                                                                                                                   \gamma_{12}Age<sub>i</sub> + \gamma_{13}Edu<sub>i</sub> + \gamma_{14}Income<sub>i</sub> + u_{0i}
m06.d <- glmmTMB(login_hr.log ~ age.z + edu.z + inc.z</pre>
                                           + rating_t1.z + unlock.z + battle.z
                                           + phase1 + phase2 + phase3
                                           + us_cmc + cn_cmc + ussr_cmc
                                           + us_cm + cn_cm + ussr_cm
                                           + liberal + conser + military_y
                                           + (us_cmc + cn_cmc + ussr_cmc | id),
                                           data = data_long, REML = TRUE)
data_long$battle_wk.log <- log(data_long$battle_wk+1) #battle_wk is very skewed
m07.d <- glmmTMB(battle_wk.log ~ age.z + edu.z + inc.z
                                                 + rating_t1.z + unlock.z + battle.z
```

Table 2: Multilevel Growth Curve Models Predicting Gaming Behavors

	Growth Curve Model	Piecewise Growth Model
(Intercept)	1.795***	1.656***
` - /	(0.031)	(0.031)
week	0.037***	,
	(0.002)	
sd(Intercept)	1.458	1.435
sdweek	0.098	
$cor_{\underline{\hspace{1cm}}}(Intercept).week$	-0.623	
sdObservation	0.824	0.709
phase1		0.071***
		(0.004)
phase2		0.066***
		(0.005)
phase3		-0.094***
		(0.007)
sd_phase1		0.168
sd_phase2		0.209
sd_phase3		0.270
cor(Intercept).phase1		-0.417
cor(Intercept).phase2		-0.341
cor(Intercept).phase3		-0.014
$cor_phase1.phase2$		-0.095
$cor\phase1.phase3$		-0.298
corphase2.phase3		-0.205
AIC	104615.9	99707.9
BIC	104667.1	99835.9
Log.Lik.	-52301.951	-49838.948

⁺ p < 0.1, * p < 0.05, ** p < 0.01, *** p < 0.001

```
+ phase1 + phase2 + phase3
                 + us_cmc + cn_cmc + ussr_cmc
                 + us_cm + cn_cm + ussr_cm
                 + liberal + conser + military_y
                 + (us_cmc + cn_cmc + ussr_cmc | id),
                 data = data_long, REML = TRUE)
m08.d <- glmmTMB(win_pct ~ age.z + edu.z + inc.z</pre>
                 + rating_t1.z + unlock.z + battle.z
                 + phase1 + phase2 + phase3
                 + us_cmc + cn_cmc + ussr_cmc
                 + us_cm + cn_cm + ussr_cm
                 + liberal + conser + military_y
                 + (us_cmc + cn_cmc + ussr_cmc | id),
                 data = data_long, REML = TRUE)
data_long$team_wk.log <- log(data_long$team_wk+1) #team_wk is very skewed
m09.d <- glmmTMB(team_wk.log ~ age.z + edu.z + inc.z</pre>
                 + rating_t1.z + unlock.z + battle.z
                 + phase1 + phase2 + phase3
                 + us_cmc + cn_cmc + ussr_cmc
                 + us_cm + cn_cm + ussr_cm
                 + liberal + conser + military_y
                 + (us_cmc + cn_cmc + ussr_cmc | id),
                 data = data_long, REML = TRUE)
msummary(list(
  "Gaming time (Model 7)" = m06.d,
  "Battle count (Model 8)" = m07.d,
 "Team battle (Model 9)" = m09.d,
 "Win rate (Model 10)" = m08.d),
 title = 'Multilevel Growth Curve Models Predicting Gaming Behavors and Performance',
 stars = TRUE,
  statistic = c('std.error')
)
ran_int <- lmer(cn_pct ~ 1 + (1 | id), data = data_long)</pre>
variance_components <- as.data.frame(VarCorr(ran_int))</pre>
between_var <- variance_components$vcov[1]</pre>
within_var <- variance_components$vcov[2]</pre>
(icc <- between_var / (between_var + within_var))</pre>
## [1] 0.3505328
#For plot please refer to my prosepctus (or paper)
```

Table 3: Multilevel Growth Curve Models Predicting Gaming Behavors and Performance

	Gaming time (Model 7)	Battle count (Model 8)	Team battle (Model 9)	Win rate (Mo
(Intercept)	2.342***	3.556***	0.038***	0.470***
	(0.066)	(0.068)	(0.006)	(0.004)
age.z	0.001	-0.050+	-0.005*	-0.006**
	(0.028)	(0.028)	(0.002)	(0.002)
edu.z	-0.012	-0.007	0.000	0.002
	(0.024)	(0.025)	(0.002)	(0.001)
inc.z	0.113***	0.053*	-0.002	-0.001
	(0.025)	(0.026)	(0.002)	(0.001)
$rating_t1.z$	-0.029	-0.025	0.002	0.027***
	(0.026)	(0.027)	(0.002)	(0.002)
unlock.z	-0.044	-0.053+	0.006*	-0.004*
	(0.030)	(0.030)	(0.003)	(0.002)
battle.z	0.365***	0.368***	0.011***	0.001
1 1	(0.031)	(0.032)	(0.003)	(0.002)
phase1	0.048***	0.055***	-0.003***	0.003***
1 0	(0.003)	(0.003)	(0.000)	(0.001)
phase2	0.024***	0.019***	0.001	0.002**
1 0	(0.004)	(0.004)	(0.000)	(0.001)
phase3	-0.089***	-0.091***	-0.004***	-0.001
	(0.005)	(0.005)	(0.001)	(0.001)
us_cmc	-0.081	-0.059	-0.008	0.003
	$(0.051) \\ 0.498***$	$(0.058) \\ 0.913***$	(0.005)	(0.009)
cn_cmc			0.002	0.022
	(0.127) $0.234***$	(0.152) $0.415***$	(0.011) $0.025****$	(0.017)
ussr_cmc	(0.056)	(0.066)		0.001
us em	(0.050) -0.202+	(0.000) -0.236*	$(0.006) \\ -0.023*$	(0.009) -0.001
us_cm	-0.202+ (0.108)	-0.250 (0.112)	(0.010)	(0.006)
cn cm	0.151	0.083	-0.017	-0.038
cn_cm	(0.393)	(0.412)	(0.034)	(0.023)
ussr cm	0.434**	0.423**	0.058***	-0.020°
ussiciii	(0.147)	(0.154)	(0.013)	(0.009)
liberal	-0.007	0.032	0.002	0.008*
in order	(0.063)	(0.064)	(0.006)	(0.004)
conser	-0.048	-0.049	0.003	0.002
	(0.050)	(0.051)	(0.004)	(0.003)
military_y	-0.080+	-0.084+	-0.002	0.000
· —	(0.045)	(0.046)	(0.004)	(0.003)
sd(Intercept)	0.839	0.863	0.074	0.036
sdus_cmc	1.212	1.530	0.106	0.223
sdcn_cmc	1.670	2.206	0.122	0.212
sdussr_cmc	1.298	1.701	0.163	0.195
cor(Intercept).us_cmc	-0.072	-0.061	-0.176	0.024
cor(Intercept).cn_cmc	-0.036	-0.044	-0.027	-0.114
cor(Intercept).ussr_cmc	-0.006	-0.077	0.199	-0.083
corus_cmc.cn_cmc	0.255	0.189	0.621	0.553
corus_cmc.ussr_cmc	0.257	0.337	0.304	0.377
$cor_cn_cmc.ussr_cmc$	0.409	0.427	0.411	0.594
sd _Observation	0.664	0.686	0.074	0.112
AIC	51994.4	54058.2	-46060.4	-30719.
BIC	52234.5	54298.3	-45820.3	-30479.
Log.Lik.	-25967.200	-26999.117	23060.213	15389.67