

Appendix: R codes for the analyses

```
#Import data file and check data types
```

```
library(dplyr)
```

```
write <- read.csv("write.csv")
```

```
str(write)
```

```
#Change the type of all the covariates from integer to factor
```

```
write <- mutate_if(write, is.integer, as.factor)
```

```
#-----  
#Stage 1. Matching through propensity score methods.  
#-----
```

```
install.packages("MatchIt")
```

```
install.packages("optmatch")
```

```
library(MatchIt)
```

```
library(optmatch)
```

```
# Step 1.1: Selecting covariates
```

```
#Build propensity score model using all the covariates
```

```
Edu <- as.formula(paste("EducationCategory~",paste(names(write)[c(2:25)],collapse="+")))
```

```
# Step 1.2: Estimating propensity score and matching.
```

```
#Optimal pair matching (one to one)
```

```
pair.m <- matchit(Edu, data = write, method = "optimal", distance = "logit", ratio = 1)
```

```
#Optimal full matching
```

```
##(one to multiple and multiple to one; in this example we set the ratio to be 1 to 8)
```

```
full.m <- matchit(Edu, data = write, method = "full", distance = "logit", min.controls = 1/8,  
  max.controls = 8)
```

#Step 1.3: Evaluating matching results.

```
summary(pair.m)
```

```
plot(pair.m, type="jitter", interactive = FALSE)
```

```
plot(pair.m, type="hist", interactive = FALSE)
```

```
summary(full.m)
```

```
plot(full.m, type="jitter", interactive = FALSE)
```

```
plot(full.m, type="hist")
```

```
#save the matched data
```

```
full <- match.data(full.m)
```

```
#-----  
#Stage 2. Running linear mixed effects regression DIF analyses.  
#-----
```

```
install.packages("lme4")
```

```
library(lme4)
```

```
install.packages("sjstats")
```

```
library(sjstats) #for R-squared approximation
```

```
str(full)
```

```
full$subclass <- as.factor(full$subclass)
```

```
#Compute the ability proxy variable
```

```
full $L_zscore <- scale(full $L_score, center = TRUE, scale = TRUE)
```

```
full $R_zscore <- scale(full $R_score, center = TRUE, scale = TRUE)
```

```
full $S_zscore <- scale(full $S_score, center = TRUE, scale = TRUE)
```

```
full $W2_zscore <- scale(full $W2, center = TRUE, scale = TRUE)
```

```
full $ztot<- full $L_zscore + full $R_zscore + full $S_zscore + full $W2_zscore
```

#Center the ability proxy variable, ztot, within subclass

```
cmc.ztot <- full %>%
```

```
  select(RegID,ztot,subclass) %>%
```

```
  group_by(subclass) %>%
```

```
  mutate(c.ztot=ztot-mean(ztot))
```

```
full <- merge(full, subset(cmc.ztot, select=c("RegID", "c.ztot")),by="RegID")
```

#Step 2.1: Establishing a null model (Model 1).

```
M1.null <- lmer(W1 ~ 1 + (1 |subclass), data= full, REML=FALSE)
```

```
summary(M1.null)
```

```
r2(M1.null)
```

#Step 2.2: Running a baseline model with ability approximation variable(s) (Model 2).

```
M2.base <- lmer(W1 ~ 1 + ztot.c + (1 |subclass), data= full, REML=FALSE)
```

```
summary(M2.base)
```

```
r2(M2.base)
```

#Step 2.3: Detecting uniform DIF (Model 3).

```
M3.uniform <- lmer(W1 ~ 1 + ztot.c + EducationCategory + (1 |subclass), data= full,  
  REML=FALSE)
```

```
summary(M3.uniform)
```

```
r2(M3.uniform)
```

#Step 2.4: Detecting non-uniform DIF (Model 4).

```
M4.non.uni <- lmer(W1 ~ 1 + ztot.c * EducationCategory + (1 |subclass), data= full,  
  REML=FALSE)
```

```
summary(M4.non.uni)
```

```
r2(M4.non.uni)
```

```
#Model comparison
```

```
anova(M2.base, M3.uniform)
```

```
anova(M3.uniform, M4.non.uni)
```

```
#Step 2.5: Conducting sensitivity analysis.
```

```
install.packages("sensitivityfull")
```

```
library(sensitivityfull)
```

```
library(tidyr)
```

```
#Restructure data from long to wide format
```

```
full.format<-full[with(full, order(subclass, -(as.numeric(as.character(EducationCategory))))), ]
```

```
full.format<-full.format %>% group_by(subclass) %>% mutate(seq.id = row_number())
```

```
#Add a column counting number of treatment unit (coded 1) within each subclass
```

```
full.format<-full.format %>% group_by(subclass) %>% mutate(count.t =  
  sum(EducationCategory == 1))
```

```
#Add a column with True/False to identify if a subclass only has 1 treatment unit
```

```
full.format$treated1 <- ifelse(full.format$count.t == 1, TRUE, FALSE)
```

```
#Remove columns that will not be used in sensitivity analysis
```

```
full.format1<-select(full.format, subclass, seq.id, treated1, W1)
```

```
full.wide<-spread(full.format1, seq.id, W1)
```

```
#Remove extra columns
```

```
full.matrix<-full.wide[-c(1:2)]
```

```
full.matrix<-as.matrix(full.matrix)
```

```
#Sensitivity analysis. Increase the value of gamma from 1 to 2 with an increment of 0.1.
```

```
senfm(full.matrix, full.wide$treated1, gamma = 1, inner = 0, trim = 3, lambda = 1/2,  
      tau = 0, alternative="less")
```