Appendix Difference-in-Difference Analysis: Female Labor Participation

```
#Downloading the libraries, setting the working directory and importing the data set
library(tidyverse)
library(stargazer)
library(dagitty)
library(gridExtra)
library(tinytex)
library(ggplot2)
library(tidyr)
library(dplyr)
library(plyr)
library(reshape2)
library(sandwich)
dir <- "/Users/valeriemaasdamme/Documents/BAM_ASP_A2"</pre>
dirProg <- pasteO(dir, "/programs/")</pre>
dirData <- pasteO(dir, "/Data/")</pre>
dfDiD <- read.csv(file=paste0(dirData, "DiD_dataset.csv"))</pre>
```

Preparing and analyzing the dataset

```
# no need to transform the dataset, already in the long format
str(dfDiD) # all variables are numeric or integer, no need to transform

dfDiD$dPeriod = ifelse(dfDiD$year >= 1993, 1, 0) # dummy variable for period
dfDiD$cChildren = ifelse(dfDiD$children >= 1, 1, 0) # dummy for different groups

dfDiD.sub <- subset(dfDiD, work=="1") #creating a subset of employed women</pre>
```

1 Plotting the dependent variables

```
Earn.plot <- qplot(Year, Earn, data=earn.agg, geom=c("point","line"),</pre>
  colour = Group,
 xlab="Year", ylab="Annual earnings") +
 geom vline(xintercept = 1993) +
 theme bw()
ggsave(file="Earn.pdf", width=7, height=4)
#Finc
finc.agg = aggregate(dfDiD.sub$finc, list(dfDiD.sub$year, dfDiD.sub$cChildren == 1),
                     FUN = mean, na.rm = TRUE)
names(finc.agg) = c("Year", "Children", "Finc")
finc.agg$Group[1:6] = "Women without children"
finc.agg$Group[7:12] = "Women with children"
Finc.plot <- qplot(Year, Finc, data=finc.agg, geom=c("point","line"),</pre>
  colour = Group,
 xlab="Year", ylab="Annual Family Income") +
 geom_vline(xintercept = 1993) +
 theme_bw()
ggsave(file="Finc.pdf", width=7, height=4)
#Work
work.agg = aggregate(dfDiD$work, list(dfDiD$year, dfDiD$cChildren == 1),
                     FUN = mean, na.rm = TRUE)
names(work.agg) = c("Year", "Children", "Work")
work.agg$Group[1:6] = "Women without children"
work.agg$Group[7:12] = "Women with children"
Work.plot <- qplot(Year, Work, data=work.agg, geom=c("point","line"),</pre>
  colour = Group,
 xlab="Year", ylab="Work")+
 geom_vline(xintercept = 1993) +
 theme bw()
ggsave(file="Work.pdf", width=7, height=4)
```

2 Summary statistics of the dataset

```
stargazer(dfDiD, type = "text")
stargazer(dfDiD[, c("children", "finc", "earn", "age", "work", "unearn")], type = "text")
```

3 Difference-in-Difference

```
avgWork <- ddply (dfDiD, .(dPeriod, cChildren), summarise,</pre>
                 avgWork = mean(work, na.rm=TRUE))
#Remodel the avg table from long to wide, add row for the difference in averages
avgtable.Earn <- dcast (avgEarn, dPeriod ~ cChildren, value.var = "avgEarn")
avgtable.Earn <- rbind(avgtable.Earn, avgtable.Earn[2,]-avgtable.Earn[1,])
rownames(avgtable.Earn) <- c("Before", "After", "Difference") # renaming the rows
colnames(avgtable.Earn) <- c("dPeriod", "Women without children (0)",</pre>
                              "Women with children (1)") # renaming the columns
avgtable.Earn[3, "dPeriod"] <- NA
avgtable.Finc <- dcast (avgFinc, dPeriod ~ cChildren, value.var = "avgFinc")
avgtable.Finc <- rbind(avgtable.Finc, avgtable.Finc[2,]-avgtable.Finc[1,])</pre>
rownames(avgtable.Finc) <- c("Before", "After", "Difference")</pre>
colnames(avgtable.Finc) <- c("dPeriod", "Women without children (0)",</pre>
                              "Women with children (1)")
avgtable.Finc[3, "dPeriod"] <- NA
avgtable.Work <- dcast (avgWork, dPeriod ~ cChildren, value.var = "avgWork")
avgtable.Work <- rbind(avgtable.Work, avgtable.Work[2,]-avgtable.Work[1,])
rownames(avgtable.Work) <- c("Before", "After", "Difference")</pre>
colnames(avgtable.Work) <- c("dPeriod", "Women without children (0)",</pre>
                              "Women with children (1)")
avgtable.Work[3, "dPeriod"] <- NA</pre>
stargazer(avgtable.Earn, summary=FALSE, align = TRUE, type="text",
          title = "Average Annual Earnings")
stargazer(avgtable.Finc, summary=FALSE, align = TRUE, type="text",
          title = "Average Indicator Annual Family Income")
stargazer(avgtable.Work, summary=FALSE, align = TRUE, type="text",
          title = "Average Indicator Work Status")
```

4 Regression analysis

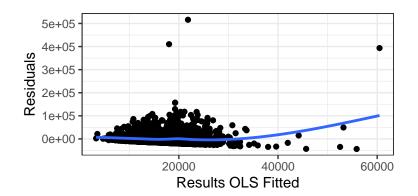
Control variables

```
# adding urate, unearn and children as control variables
mdl.control.earn <- earn ~ cChildren + dPeriod + cChildren:dPeriod +
  urate + unearn + children
rsltOLS.control.earn <- lm(mdl.control.earn, data=dfDiD.sub)
# Finc
mdl.control.finc <- finc ~ cChildren + dPeriod + cChildren:dPeriod +
  urate + unearn + children
rsltOLS.control.finc <- lm(mdl.control.finc, data=dfDiD.sub)
# Work
mdl.control.work <- work ~ cChildren + dPeriod + cChildren:dPeriod +</pre>
  urate + unearn + children
rsltOLS.control.work <- lm(mdl.control.work, data=dfDiD)
stargazer(rsltOLS.control.earn, rsltOLS.control.finc,
          rsltOLS.control.work,
          intercept.bottom = FALSE,
          align = TRUE,
          no.space=TRUE, type="text")
```

Robust standard errors

```
#Test for heteroskedasticity
rsltOLS.control.earn2 <- lm(mdl.control.earn, data=dfDiD.sub)
rsltOLS.control.finc2 <- lm(mdl.control.finc, data=dfDiD.sub)
rsltOLS.control.work2 <- lm(mdl.control.work, data=dfDiD)

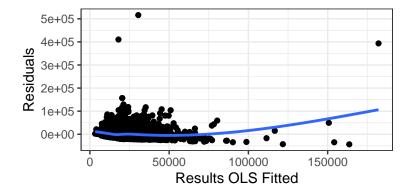
# EARN
ggplot(data = data.frame(fit = fitted(rsltOLS.control.earn2),
    rsid = residuals(rsltOLS.control.earn2)),
    aes(fit, rsid)) +
    geom_point() +
    stat_smooth(se = F) +
    theme_bw() +
    labs(x = "Results OLS Fitted") +
    labs(y = "Residuals")</pre>
```



```
lmtest::bptest(rsltOLS.control.earn2)
# p < 0.01, heteroskedastiscity is detected.

#FINC

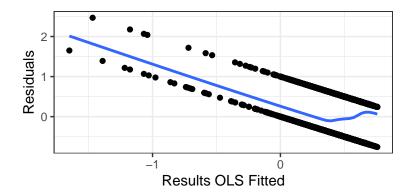
ggplot(data = data.frame(fit = fitted(rsltOLS.control.finc2),
    rsid = residuals(rsltOLS.control.finc2)),
    aes(fit, rsid)) +
    geom_point() +
    stat_smooth(se = F) +
    theme_bw() +
    labs(x = "Results OLS Fitted") +
    labs(y = "Residuals")</pre>
```



```
lmtest::bptest(rsltOLS.control.finc2)
# p < 0.01, heteroskedastiscity is detected.

#WORK

ggplot(data = data.frame(fit = fitted(rsltOLS.control.work2),
    rsid = residuals(rsltOLS.control.work2)),
    aes(fit, rsid)) +
    geom_point() +
    stat_smooth(se = F) +
    theme_bw() +
    labs(x = "Results OLS Fitted") +
    labs(y = "Residuals")</pre>
```



lmtest::bptest(rslt0LS.control.work2)
p < 0.01, heteroskedastiscity is detected</pre>

	Dej	Dependent variable:			
	(1)	earn (2)	(3)		
cChildren	•	-4,734.562*** (835.140)	•		
dPeriod		-567.581 (799.615)			
urate		429.112*** (146.978)			
unearn		339.835* (192.934)			
children	·	-1,505.236*** (264.835)	•		
cChildren:dPeriod	•	1,490.156 (916.133)	•		
Constant	· ·	16,363.220*** (1,383.430)	-		
Observations	7,052	7,052	7,052		

	Dependent variable:			
	(1)	finc (2)	(3)	
cChildren	-4,734.562***	-4,734.562***	-4,734.562***	
	(945.280)	(835.140)	(945.280)	
dPeriod	-567.581	-567.581	-567.581	
	(696.500)	(799.615)	(696.500)	
urate	429.112***	429.112***	429.112***	
	(161.063)	(146.978)	(161.063)	
unearn	1,339.835***	1,339.835***	1,339.835***	
	(37.979)	(192.934)	(37.979)	
children	-1,505.236*** (301.752)	-1,505.236*** (264.835)	•	
cChildren:dPeriod	1,490.156	1,490.156	1,490.156	
	(945.477)	(916.133)	(945.477)	
Constant	16,363.220*** (1,285.565)	16,363.220*** (1,383.430)	•	
Observations R2 Adjusted R2 Residual Std. Error (df = 7045) F Statistic (df = 6; 7045)	7,052	7,052	7,052	
	0.172	0.172	0.172	
	0.171	0.171	0.171	
	19,061.410	19,061.410	19,061.410	
	244.155***	244.155***	244.155***	
Note:		*p<0.1; **p<0	.05; ***p<0.01	

Dependent variable:		
(1)	work (2)	(3)
-0.017 (0.017)	-0.017 (0.016)	-0.017 (0.017)
		-0.021*** (0.003)
-0.017*** (0.001)	-0.017*** (0.001)	-0.017*** (0.001)
0.035** (0.017)	0.035** (0.017)	0.035** (0.017)
0.817*** (0.024)	0.817*** (0.024)	0.817*** (0.024)
13,746 0.090 0.089 0.477 225.866***		
	(1) -0.017 (0.017) -0.031** (0.013) -0.021*** (0.003) -0.017*** (0.001) -0.052*** (0.004) 0.035** (0.017) 0.817*** (0.024) 13,746 0.090 0.089 0.477	work (1) (2) -0.017 -0.017 (0.017) (0.016) -0.031** -0.031** (0.013) (0.013) -0.021*** -0.021*** (0.003) (0.003) -0.017*** -0.017*** (0.001) (0.001) -0.052*** -0.052*** (0.004) (0.004) 0.035** 0.035** (0.017) (0.017) 0.817*** 0.817*** (0.024) (0.024) 13,746 0.090 0.090 0.089 0.089

#No impact on the significance of the DiD effect, all three significant (p<0.05). $\#Standard\ error\ for\ all\ three\ remains\ the\ same$