

Project

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Data Cleaning

```
data <- read_sav(here("data_files", "pj_data.sav"))

data<- data %>% #don't need household data
  filter(RECTYPE != 1)

data_filled<- data %>%
  group_by(CASEID) %>%
  mutate(FULLPART = FULLPART[1],
         DAY=DAY[1],
         AGE=AGE[1],
         SEX=SEX[1],
         MARST=MARST[1],
         UHRSWORKT=UHRSWORKT[1],
         EARNWEEK=EARNWEEK[1],
         PAIDHOUR=PAIDHOUR[1],
         HH_NUMOWNKIDS=HH_NUMOWNKIDS[1],
         WB_RESP=WB_RESP[1],
         Leisure_All=Leisure_All[1])%>%
  subset(select = -c(PERNUM, LINENO, WT06, IND2_CPS8, YEAR, AWBWT)) %>%
  filter(RECTYPE!=2) #after pasting values, remove rows with only level 2
head(data_filled)

## # A tibble: 6 x 20
## # Groups:   CASEID [1]
##      RECTYPE CASEID    DAY    AGE    SEX    MARST FULLPART UHRSWORKT EARNWEEK
##      <dbl+lbl> <dbl> <dbl+lbl> <dbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl> <dbl>
## 1 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## 2 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## 3 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## 4 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## 5 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## 6 3 [Activity] 2.02e13 5 [Thu~    60 2 [Fem~ 4 [Div~ 1 [Full~    40    516
## # ... with 11 more variables: PAIDHOUR <dbl+lbl>, HH_NUMOWNKIDS <dbl+lbl>,
## #   WB_RESP <dbl+lbl>, Leisure_All <dbl>, ACTLINE <dbl>, ACTIVITY <dbl+lbl>,
## #   DURATION <dbl>, SCHAPPY <dbl+lbl>, SCSTRESS <dbl+lbl>, MEANING <dbl+lbl>,
## #   WBELIG <dbl+lbl>
```

```
#Recoding Hourly Status Variable (1=Not Paid Hourly, 2 = Paid Hourly)
```

```
data_filled$PAIDHOUR <- as.numeric(data_filled$PAIDHOUR)
```

```
data_filled <- data_filled %>%
```

```
  mutate(PAIDHOUR=recode(PAIDHOUR,
```

```
    '1' = 2,
```

```
    '2' = 1))
```

```
#Recoding Day of Week (1=Weekday, 2=Weekend)
```

```
data_filled$DAY <- as.numeric(data_filled$DAY)
```

```
data_filled <- data_filled %>%
```

```
  mutate(DAY=recode(DAY,
```

```
    '1' = 2,
```

```
    '2' = 1,
```

```
    '3' = 1,
```

```
    '4' = 1,
```

```
    '5' = 1,
```

```
    '6' = 1,
```

```
    '7' = 2))
```

```
#Recoding Marital Status (1=Not Married, 2= Married)
```

```
data_filled$MARST <- as.numeric(data_filled$MARST)
```

```
data_filled <- data_filled %>%
```

```
  mutate(MARST=recode(MARST,
```

```
    '1' = 2,
```

```
    '2' = 2,
```

```
    .default = 1))
```

```
#Recoding Well-Being Measures
```

```
data_filled$MEANING <- as.numeric(data_filled$MEANING)
```

```
data_filled <- data_filled %>%
```

```
  mutate(MEANING=recode(MEANING,
```

```
    '0' = 1,
```

```
    '1' = 2,
```

```
    '2' = 3,
```

```
    '3' = 4,
```

```
    '4' = 5,
```

```
    '5' = 6,
```

```
    '6' = 7))
```

```
data_filled$SCHAPPY <- as.numeric(data_filled$SCHAPPY)
```

```
data_filled <- data_filled %>%
```

```
  mutate(SCHAPPY=recode(SCHAPPY,
```

```
    '0' = 1,
```

```
    '1' = 2,
```

```
    '2' = 3,
```

```
    '3' = 4,
```

```
    '4' = 5,
```

```
    '5' = 6,
```

```
    '6' = 7))
```

```
data_filled$SCSTRESS <- as.numeric(data_filled$SCSTRESS)
```

```
data_filled <- data_filled %>%
```

```
mutate(SCSTRESS=recode(SCSTRESS,
  '0' = 1,
  '1' = 2,
  '2' = 3,
  '3' = 4,
  '4' = 5,
  '5' = 6,
  '6' = 7))
```

Exclusions

```
#include participants who answered the well-being module
data_short <- data_filled %>%
  filter(WBELIG==1) %>%
  subset(select = -c(WBELIG, RECTYPE))

#randomly selected to report well-being during leisurely activities (ATUS activity coding category 12 and 13)
data_short <- data_short %>%
  filter(ACTIVITY < 140000) %>%
  filter(ACTIVITY >=120000) %>%
  filter(SCHAPPY %in% (1:7)) %>%
  filter(SCSTRESS %in% (1:7)) %>%
  filter(MEANING %in% (1:7)) %>%
  filter(PAIDHOUR != 99) %>% #don't have information on hourly payment status
  filter(UHRSWORKT != 9999) %>% #don't have number of hours worked
  filter(UHRSWORKT != 9995) %>% #responded as hours varied
  filter(EARNWEEK != 0)

data_short <- data_short %>%
  group_by(CASEID) %>%
  filter(n()>=2) %>%
  mutate (CASEID=cur_group_id())%>%
  ungroup()
```

Variable Description

- CASEID: unique identifying number for participant
- DAY: whether the day of the week of the diary day was a weekday (DAY = 1) or a weekend (DAY = 2)
- AGE: person's age in years as of his/her last birthday
- SEX: whether the individual is male (SEX = 1) or female (SEX = 2)
- MARST: person's marital status, either not married (MARST = 1) or married (MARST = 2)
- FULLPART: indicates whether the individual usually works full time (FULLPART = 1) or part time (FULLPART = 2).
- UHRSWORKT: total number of hours the respondent usually works per week
- EARNWEEK: income per week
- PAIDHOUR: whether the respondent is paid on an hourly basis (PAIDHOUR = 2) or not (PAIDHOUR = 1)
- HH_NUMOWNKIDS: the number of the respondent's own children under the age of 18 who live in the household

- DURATION: length of the activity in minutes
- SCHAPPY: seven point scale that indicates how much happiness the respondent felt during (1 - Not At all, 7 - Very) the activity.
- SCSTRESS: seven point scale that indicates how much stress the respondent felt during the activity (1 - Not At all, 7 - Very)
- MEANING: seven point scale that indicates how meaningful the activity was to the respondent (1 - Not At all, 7 - Very)

Model Equation

Lv-1:

$$\text{SCHAPPY}_{ij} = \beta_{0j} + \beta_{1j}\text{DURATION}_{ij} + e_{ij}$$

Lv-2:

$$\beta_{0j} = \gamma_{00} + \gamma_{01}\text{PAIDHOUR}_j + \gamma_{02}\log(\text{INCOME})_j + u_{0j}$$

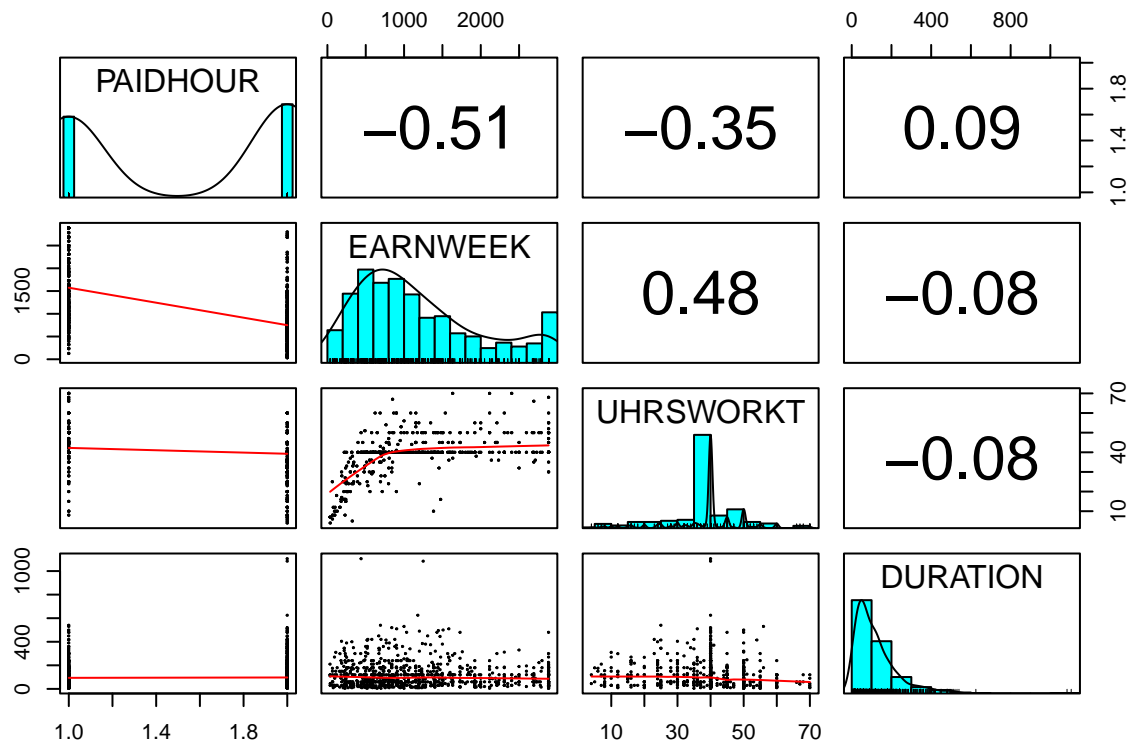
$$\beta_{1j} = \gamma_{10} + u_{1j}$$

Descriptive Statistics

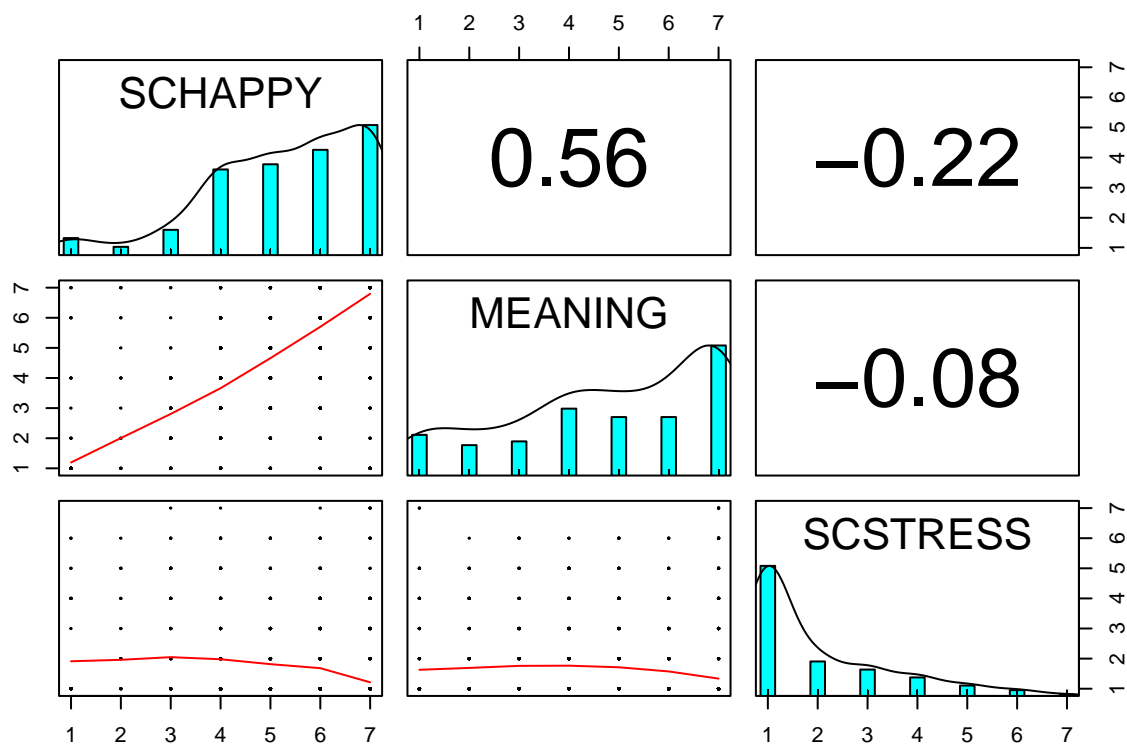
```
summary(data_short)
```

##	CASEID	DAY	AGE	SEX	
##	Min. : 1.0	Min. :1.000	Min. :15.00	Min. :1.000	
##	1st Qu.:109.0	1st Qu.:1.000	1st Qu.:33.00	1st Qu.:1.000	
##	Median :214.5	Median :2.000	Median :43.00	Median :1.000	
##	Mean :215.4	Mean :1.616	Mean :44.14	Mean :1.435	
##	3rd Qu.:322.2	3rd Qu.:2.000	3rd Qu.:55.00	3rd Qu.:2.000	
##	Max. :429.0	Max. :2.000	Max. :85.00	Max. :2.000	
##	MARST	FULLPART	UHRSWORKT	EARNWEEK	
##	Min. :1.000	Min. :1.000	Min. : 4.00	Min. : 35.0	
##	1st Qu.:1.000	1st Qu.:1.000	1st Qu.:40.00	1st Qu.: 593.2	
##	Median :1.000	Median :1.000	Median :40.00	Median : 990.4	
##	Mean :1.456	Mean :1.184	Mean :39.24	Mean :1190.7	
##	3rd Qu.:2.000	3rd Qu.:1.000	3rd Qu.:42.25	3rd Qu.:1634.2	
##	Max. :2.000	Max. :2.000	Max. :70.00	Max. :2884.6	
##	PAIDHOUR	HH_NUMOWNKIDS	WB_RESP	Leisure_All	ACTLINE
##	Min. :1.000	Min. :0.0000	Min. :1	Min. : 0.0	Min. : 1.000
##	1st Qu.:1.000	1st Qu.:0.0000	1st Qu.:1	1st Qu.: 198.0	1st Qu.: 5.000
##	Median :2.000	Median :0.0000	Median :1	Median : 322.0	Median : 9.000
##	Mean :1.536	Mean :0.4766	Mean :1	Mean : 359.6	Mean : 9.901
##	3rd Qu.:2.000	3rd Qu.:1.0000	3rd Qu.:1	3rd Qu.: 515.0	3rd Qu.:13.000
##	Max. :2.000	Max. :4.0000	Max. :1	Max. :1145.0	Max. :35.000
##	ACTIVITY	DURATION	SCHAPPY	SCSTRESS	MEANING
##	Min. :120101	Min. : 5	Min. :1.000	Min. :1.00	Min. :1.000
##	1st Qu.:120303	1st Qu.: 45	1st Qu.:4.000	1st Qu.:1.00	1st Qu.:3.000
##	Median :120303	Median : 90	Median :6.000	Median :1.00	Median :5.000
##	Mean :121337	Mean : 120	Mean :5.295	Mean :1.98	Mean :4.831
##	3rd Qu.:120307	3rd Qu.: 160	3rd Qu.:7.000	3rd Qu.:3.00	3rd Qu.:7.000
##	Max. :130299	Max. :1105	Max. :7.000	Max. :7.00	Max. :7.000

```
data_short %>%
  select(PAIDHOUR, EARNWEEK, UHRSWORKT, DURATION) %>%
  psych::pairs.panels(ellipses = FALSE, cex = 0.2, cex.cor = 1)
```



```
data_short %>%
  select(SCHAPPY, MEANING, SCSTRESS) %>%
  psych::pairs.panels(ellipses = FALSE, cex = 0.2, cex.cor = 1)
```



ICC

```
m0_hap <- lmer(SCHAPPY ~ (1 | CASEID), data = data_short)
performance::icc(m0_hap)
```

```
## # Intraclass Correlation Coefficient
##
##     Adjusted ICC: 0.621
##     Unadjusted ICC: 0.621
```

Multilevel Analysis

```
m1 <- lmer(SCHAPPY ~ PAIDHOUR * DURATION * log(EARNWEEK) + (1 | CASEID),
            data = data_short)
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

```
summary(m1)
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: SCHAPPY ~ PAIDHOUR * DURATION * log(EARNWEEK) + (1 | CASEID)
## Data: data_short
##
## REML criterion at convergence: 3146.5
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.7136 -0.4293  0.1439   0.4167   3.8653
##
## Random effects:
## Groups Name Variance Std.Dev.
## CASEID (Intercept) 1.5191  1.2325
## Residual 0.8997  0.9485
## Number of obs: 896, groups: CASEID, 429
##
## Fixed effects:
##
##              Estimate Std. Error    df t value
## (Intercept)  1.971710   3.148857 723.401164  0.626
## PAIDHOUR      1.612238   1.758803 720.539556  0.917
## DURATION      0.016549   0.017104 642.149268  0.968
## log(EARNWEEK)  0.412585   0.438300 724.487191  0.941
## PAIDHOUR:DURATION -0.005422   0.009415 657.569120 -0.576
## PAIDHOUR:log(EARNWEEK) -0.198693   0.249987 719.618843 -0.795
## DURATION:log(EARNWEEK) -0.002087   0.002409 645.577581 -0.866
## PAIDHOUR:DURATION:log(EARNWEEK) 0.000658   0.001344 662.059919  0.490
##
##              Pr(>|t|)
## (Intercept)  0.531
## PAIDHOUR      0.360
## DURATION      0.334
## log(EARNWEEK)  0.347
## PAIDHOUR:DURATION 0.565
## PAIDHOUR:log(EARNWEEK) 0.427
## DURATION:log(EARNWEEK) 0.387
## PAIDHOUR:DURATION:log(EARNWEEK) 0.625
##
## Correlation of Fixed Effects:
##              (Intr) PAIDHOUR DURATION 1(EARN PAIDHOUR:DURATION PAIDHOUR:(
## PAIDHOUR      -0.969
## DURATION      -0.582  0.562
## 1(EARNWEEK)    -0.996  0.972   0.584
## PAIDHOUR:DURATION 0.570 -0.585 -0.973 -0.576
## PAIDHOUR:(E    0.952 -0.994 -0.556 -0.965  0.584
## DURATION:(E    0.577 -0.561 -0.996 -0.584  0.974      0.559
## PAIDHOUR:DURATION: -0.561  0.581   0.961   0.571 -0.996      -0.586
##
## DURATION:
## PAIDHOUR
## DURATION
## 1(EARNWEEK)
## PAIDHOUR:DURATION
```

```
## PAIDHOUR:(E
## DURATION:(E
## PAIDHOUR:DURATION: -0.969
## fit warnings:
## Some predictor variables are on very different scales: consider rescaling
```

```
#Covariate
```

```
m1_c <- lmer(SCHAPPY ~ DURATION * PAIDHOUR * log(EARNWEEK) + AGE + MARST + DAY + UHRSWORKT + HH_NUMOWNK
            data = data_short)
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

```
summary(m1_c)
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: SCHAPPY ~ DURATION * PAIDHOUR * log(EARNWEEK) + AGE + MARST +
## DAY + UHRSWORKT + HH_NUMOWNKIDS + (1 | CASEID)
## Data: data_short
##
## REML criterion at convergence: 3165.8
##
## Scaled residuals:
## Min      1Q  Median      3Q      Max
## -3.7147 -0.4083  0.1173  0.4246  3.8355
##
## Random effects:
## Groups Name Variance Std.Dev.
## CASEID (Intercept) 1.5199  1.2328
## Residual 0.9002  0.9488
## Number of obs: 896, groups: CASEID, 429
##
## Fixed effects:
##
## Estimate Std. Error df t value
## (Intercept) 1.936e+00 3.198e+00 7.075e+02 0.605
## DURATION 1.739e-02 1.713e-02 6.398e+02 1.015
## PAIDHOUR 1.658e+00 1.779e+00 7.067e+02 0.932
## log(EARNWEEK) 3.501e-01 4.419e-01 7.137e+02 0.792
## AGE 4.091e-03 4.875e-03 4.225e+02 0.839
## MARST 7.145e-02 1.618e-01 4.189e+02 0.442
## DAY 8.577e-03 1.415e-01 4.342e+02 0.061
## UHRSWORKT 2.526e-03 8.622e-03 4.205e+02 0.293
## HH_NUMOWNKIDS 1.318e-01 8.795e-02 4.228e+02 1.498
## DURATION:PAIDHOUR -6.302e-03 9.432e-03 6.543e+02 -0.668
## DURATION:log(EARNWEEK) -2.201e-03 2.412e-03 6.432e+02 -0.912
## PAIDHOUR:log(EARNWEEK) -2.054e-01 2.527e-01 7.061e+02 -0.813
## DURATION:PAIDHOUR:log(EARNWEEK) 7.826e-04 1.347e-03 6.589e+02 0.581
## Pr(>|t|)
## (Intercept) 0.545
```



```
## DURATION                0.310
## PAIDHOUR                0.352
## log(EARNWEEK)           0.428
## AGE                    0.402
## MARST                  0.659
## DAY                    0.952
## UHRSWORKT              0.770
## HH_NUMOWNKIDS          0.135
## DURATION:PAIDHOUR       0.504
## DURATION:log(EARNWEEK)  0.362
## PAIDHOUR:log(EARNWEEK)  0.416
## DURATION:PAIDHOUR:log(EARNWEEK) 0.561

##
## Correlation matrix not shown by default, as p = 13 > 12.
## Use print(x, correlation=TRUE) or
##     vcov(x)         if you need it

## fit warnings:
## Some predictor variables are on very different scales: consider rescaling
```

```
msummary(list(
  "M1" = m1,
  "M1-Covarites" = m1_c
))
```

```
## Warning: 'modelsummary' uses the 'performance' package to extract goodness-of-fit statistics from model objects
## This warning is displayed once per session.
```

Results

```
msummary(m1,
  estimate = c("{estimate} [{conf.low}, {conf.high}]"),
  statistic = NULL, # suppress the extra rows for SEs
  shape = effect + term ~ model,
  title = "Table 1: Model coefficients")
```

If hourly workers have a greater propensity to make economic evaluation of time, they should display greater sensitivity to the opportunity cost of time and as a consequence, derive less happiness from leisurely activities. To explore people's subjective experience of leisurely activities, I used the happiness measure from the ATUS well-being module. To worker's hourly work status was dummy coded "1" for non-hourly workers and "2" for hourly workers. To account for the worker's opportunity cost of time, we used the duration spent on leisurely activities and their income as an indicator of the foregone earning from spending (wasting) time on leisure. To control for potential differences between hourly and non hourly workers, we controlled for various variables: age, marital status, number of hours worked, number of children under 18 and the day of diary. To test the hypothesis that happiness derived from engaging in leisurely activities is influenced by both hourly payment status and the opportunity cost of time, I fitted a multilevel model with hourly status, duration, and logged income on happiness to explore the three-way interaction among the three variables.

	M1	M1-Covarites
(Intercept)	1.972 (3.149)	1.936 (3.198)
PAIDHOUR	1.612 (1.759)	1.658 (1.779)
DURATION	0.017 (0.017)	0.017 (0.017)
log(EARNWEEK)	0.413 (0.438)	0.350 (0.442)
PAIDHOUR \times DURATION	-0.005 (0.009)	
PAIDHOUR \times log(EARNWEEK)	-0.199 (0.250)	-0.205 (0.253)
DURATION \times log(EARNWEEK)	-0.002 (0.002)	-0.002 (0.002)
PAIDHOUR \times DURATION \times log(EARNWEEK)	0.001 (0.001)	
SD (Intercept)	1.233	1.233
SD (Observations)	0.949	0.949
AGE		0.004 (0.005)
MARST		0.071 (0.162)
DAY		0.009 (0.141)
UHRSWORKT		0.003 (0.009)
HH_NUMOWNKIDS		0.132 (0.088)
DURATION \times PAIDHOUR		-0.006 (0.009)
DURATION \times PAIDHOUR \times log(EARNWEEK)		0.001 (0.001)
Num.Obs.	896	896
RMSE	0.75	0.75

Table 1: Table 1: Model coefficients

	Model 1
(Intercept)	1.972 [-4.208, 8.152]
PAIDHOUR	1.612 [-1.840, 5.064]
DURATION	0.017 [-0.017, 0.050]
log(EARNWEEK)	0.413 [-0.448, 1.273]
PAIDHOUR \times DURATION	-0.005 [-0.024, 0.013]
PAIDHOUR \times log(EARNWEEK)	-0.199 [-0.689, 0.292]
DURATION \times log(EARNWEEK)	-0.002 [-0.007, 0.003]
PAIDHOUR \times DURATION \times log(EARNWEEK)	0.001 [-0.002, 0.003]
SD (Intercept)	1.233
SD (Observations)	0.949
Num.Obs.	896
RMSE	0.75