

# Hobbies and Emotional Regulation (DERS)

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## Introduction

Research Question: Is there a correlation between the Hobbies and the responses to the Difficulties in Emotional Regulation Scale (DERS) Questionnaire?

Data:

- Hobbies importance  
Hobbies\_Imp\_1 to Hobbies\_Imp\_8, also together as Imp\_overall  
(Likert: 1 = Not at all important, ..., 5 = Extremely important)
- Hobbies time  
Hobbies\_Time\_1 to Hobbies\_Time\_8, also together as Time\_overall  
(Likert: 1 = < 1 hour/week, ..., 7 = more than 20 hours/week)
- Mental health metric DERS\_1 to DERS\_16, summarized as DERS\_mean.  
("indicate how often difficulties in emotional regulation": Almost never (0-10%) (1), ..., Almost always (91-100%) (5))

To answer the research question it is needed to:

- Compare all 16 hobby items individually along with Imp\_overall and Time\_overall, to the DERS
- Then summary, anova, and graph the most significant items (these are Hobbies\_Time\_1 and Hobbies\_Imp\_4).

---

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.0      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.1
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(mosaic)
```

```
## Registered S3 method overwritten by 'mosaic':
##   method                from
##   fortify.SpatialPolygonsDataFrame ggplot2
```

```
##
## The 'mosaic' package masks several functions from core packages in order to add
## additional features. The original behavior of these functions should not be affected by this.
##
## Attaching package: 'mosaic'
##
## The following object is masked from 'package:Matrix':
##
##     mean
##
## The following objects are masked from 'package:dplyr':
##
##     count, do, tally
##
## The following object is masked from 'package:purrr':
##
##     cross
##
## The following object is masked from 'package:ggplot2':
##
##     stat
##
## The following objects are masked from 'package:stats':
##
##     binom.test, cor, cor.test, cov, fivenum, IQR, median, prop.test,
##     quantile, sd, t.test, var
##
## The following objects are masked from 'package:base':
##
##     max, mean, min, prod, range, sample, sum
library(knitr)
```

## Data Preparation

```
Data <- read.csv("Data/student_mental_health.csv")[1:1193, ]

ders_items <- paste0("DERS_", 1:16)
imp_items <- paste0("Hobbies_Imp_", 1:8)
time_items <- paste0("Hobbies_Time_", 1:8)

Data_filtered <- Data %>%
  filter(Catch_question != "NA") %>%
  mutate(
    DERS_mean = rowMeans(select(., all_of(ders_items)), na.rm = TRUE),
    Imp_overall = rowMeans(select(., all_of(imp_items)), na.rm = TRUE),
    Time_overall = rowMeans(select(., all_of(time_items)), na.rm = TRUE)
  )

Data_analysis <- Data_filtered %>% filter(!is.na(DERS_mean))

dim(Data_analysis)
```

```
## [1] 748 151
```

## Table of Hobby Means

```
all_hobby_vars <- c(imp_items, time_items, "Imp_overall", "Time_overall")

hobby_table <- tibble(
  Variable = all_hobby_vars,
  Mean     = sapply(Data_analysis[all_hobby_vars], mean, na.rm = TRUE),
  SD       = sapply(Data_analysis[all_hobby_vars], sd,   na.rm = TRUE),
  n        = sapply(Data_analysis[all_hobby_vars], function(x) sum(!is.na(x)))
)

kable(hobby_table, digits = 3,
      caption = "Means and Standard Deviations for All Hobby Variables")
```

Table 1: Means and Standard Deviations for All Hobby Variables

| Variable       | Mean  | SD    | n   |
|----------------|-------|-------|-----|
| Hobbies_Imp_1  | 2.267 | 1.307 | 748 |
| Hobbies_Imp_2  | 2.116 | 1.091 | 748 |
| Hobbies_Imp_3  | 2.586 | 1.111 | 748 |
| Hobbies_Imp_4  | 3.229 | 1.017 | 748 |
| Hobbies_Imp_5  | 2.961 | 1.115 | 748 |
| Hobbies_Imp_6  | 4.167 | 0.798 | 748 |
| Hobbies_Imp_7  | 2.418 | 0.989 | 748 |
| Hobbies_Imp_8  | 3.270 | 1.063 | 748 |
| Hobbies_Time_1 | 1.717 | 1.131 | 748 |
| Hobbies_Time_2 | 1.509 | 0.925 | 748 |
| Hobbies_Time_3 | 2.028 | 1.293 | 748 |
| Hobbies_Time_4 | 4.162 | 1.388 | 748 |
| Hobbies_Time_5 | 1.999 | 1.336 | 748 |
| Hobbies_Time_6 | 5.352 | 1.501 | 748 |
| Hobbies_Time_7 | 1.488 | 0.840 | 748 |
| Hobbies_Time_8 | 2.485 | 1.370 | 748 |
| Imp_overall    | 2.877 | 0.468 | 748 |
| Time_overall   | 2.592 | 0.495 | 748 |

## Variables Relation to DERS\_mean

What is important is which hobby variables are most related to DERS\_mean.

Computing a correlation between DERS\_mean and each hobby variable, then look for variables with a correlation with statistically significant p-values.

```
screen_results <- tibble(
  Variable = all_hobby_vars,
  r        = NA_real_,
  p_value  = NA_real_,
  n        = NA_integer_
)

for (i in seq_along(all_hobby_vars)) {
```

```

v <- all_hobby_vars[i]
tmp <- Data_analysis[, c("DERS_mean", v)]
tmp <- tmp[complete.cases(tmp), ]
test <- cor.test(tmp$DERS_mean, tmp[[v]])

screen_results$r[i] <- unname(test$estimate)
screen_results$p_value[i] <- test$p.value
screen_results$n[i] <- nrow(tmp)
}

screen_results <- screen_results %>%
  mutate(abs_r = abs(r)) %>%
  arrange(desc(abs_r))

kable(screen_results,
  digits = 3,
  caption = "Correlations Between DERS_mean and All Hobby Variables (Screening Table)")

```

Table 2: Correlations Between DERS\_mean and All Hobby Variables (Screening Table)

| Variable       | r      | p_value | n   | abs_r |
|----------------|--------|---------|-----|-------|
| Hobbies_Time_1 | -0.120 | 0.001   | 748 | 0.120 |
| Hobbies_Imp_4  | 0.108  | 0.003   | 748 | 0.108 |
| Hobbies_Imp_1  | -0.067 | 0.065   | 748 | 0.067 |
| Hobbies_Imp_8  | 0.067  | 0.066   | 748 | 0.067 |
| Hobbies_Time_4 | 0.063  | 0.084   | 748 | 0.063 |
| Imp_overall    | 0.050  | 0.174   | 748 | 0.050 |
| Hobbies_Imp_3  | 0.046  | 0.206   | 748 | 0.046 |
| Time_overall   | -0.034 | 0.357   | 748 | 0.034 |
| Hobbies_Time_2 | -0.027 | 0.466   | 748 | 0.027 |
| Hobbies_Time_5 | -0.025 | 0.488   | 748 | 0.025 |
| Hobbies_Time_6 | -0.021 | 0.564   | 748 | 0.021 |
| Hobbies_Imp_2  | 0.019  | 0.608   | 748 | 0.019 |
| Hobbies_Imp_6  | 0.017  | 0.643   | 748 | 0.017 |
| Hobbies_Imp_5  | 0.011  | 0.767   | 748 | 0.011 |
| Hobbies_Time_7 | 0.011  | 0.770   | 748 | 0.011 |
| Hobbies_Time_8 | -0.005 | 0.884   | 748 | 0.005 |
| Hobbies_Imp_7  | -0.005 | 0.896   | 748 | 0.005 |
| Hobbies_Time_3 | 0.003  | 0.944   | 748 | 0.003 |

To decide what to focus on, variables with a correlation ( $\text{abs\_r} \geq 0.10$ ) and  $\text{p\_value} < 0.05$  are designated significant:

```

screen_focus <- screen_results %>%
  filter(!is.na(r),
    abs_r >= 0.10,
    p_value < 0.05)

kable(screen_focus,
  digits = 3,
  caption = "Hobby Variables With A Correlation (abs_r >= 0.10) With DERS_mean")

```

Table 3: Hobby Variables With A Correlation ( $\text{abs\_r} \geq 0.10$ )  
With DERS\_mean

| Variable       | r      | p_value | n   | abs_r |
|----------------|--------|---------|-----|-------|
| Hobbies_Time_1 | -0.120 | 0.001   | 748 | 0.120 |
| Hobbies_Imp_4  | 0.108  | 0.003   | 748 | 0.108 |

## Correlations and Linear Models

- Hobbies\_Time\_1 - “How many hours per week do you spend participating in athletics, such as varsity sports or intramurals?”
- Hobbies\_Imp\_4 - “How important is watching online recreational content such as on Netflix or Youtube to you?”

### Correlation Tests

```
cor_time1 <- cor.test(~ DERS_mean + Hobbies_Time_1, data = Data_analysis)
cor_imp4 <- cor.test(~ DERS_mean + Hobbies_Imp_4, data = Data_analysis)
```

```
cor_time1
```

```
##
## Pearson's product-moment correlation
##
## data: DERS_mean and Hobbies_Time_1
## t = -3.3048, df = 746, p-value = 0.0009959
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.1901676 -0.0488571
## sample estimates:
## cor
## -0.1201208
```

```
cor_imp4
```

```
##
## Pearson's product-moment correlation
##
## data: DERS_mean and Hobbies_Imp_4
## t = 2.969, df = 746, p-value = 0.003083
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.03666589 0.17836881
## sample estimates:
## cor
## 0.1080662
```

### Simple Linear Models

```
lm_time1 <- lm(DERS_mean ~ Hobbies_Time_1, data = Data_analysis)
lm_imp4 <- lm(DERS_mean ~ Hobbies_Imp_4, data = Data_analysis)

summary(lm_time1)
```

```
##
## Call:
## lm(formula = DERS_mean ~ Hobbies_Time_1, data = Data_analysis)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9011 -0.7136 -0.0261  0.6964  2.1840
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.06112    0.06065  50.471 < 2e-16 ***
## Hobbies_Time_1 -0.09752    0.02951  -3.305 0.000996 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9123 on 746 degrees of freedom
## Multiple R-squared:  0.01443,    Adjusted R-squared:  0.01311
## F-statistic: 10.92 on 1 and 746 DF,  p-value: 0.0009959
```

```
anova(lm_time1)
```

```
## Analysis of Variance Table
##
## Response: DERS_mean
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Hobbies_Time_1  1   9.09  9.0907   10.922 0.0009959 ***
## Residuals      746 620.94  0.8324
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(lm_imp4)
```

```
##
## Call:
## lm(formula = DERS_mean ~ Hobbies_Imp_4, data = Data_analysis)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.00416 -0.71903 -0.03153  0.69110  2.22623
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    2.57850    0.11130  23.167 < 2e-16 ***
## Hobbies_Imp_4  0.09763    0.03288   2.969 0.00308 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9136 on 746 degrees of freedom
## Multiple R-squared:  0.01168,    Adjusted R-squared:  0.01035
## F-statistic: 8.815 on 1 and 746 DF,  p-value: 0.003083
```

```
anova(lm_imp4)
```

```
## Analysis of Variance Table
##
## Response: DERS_mean
```

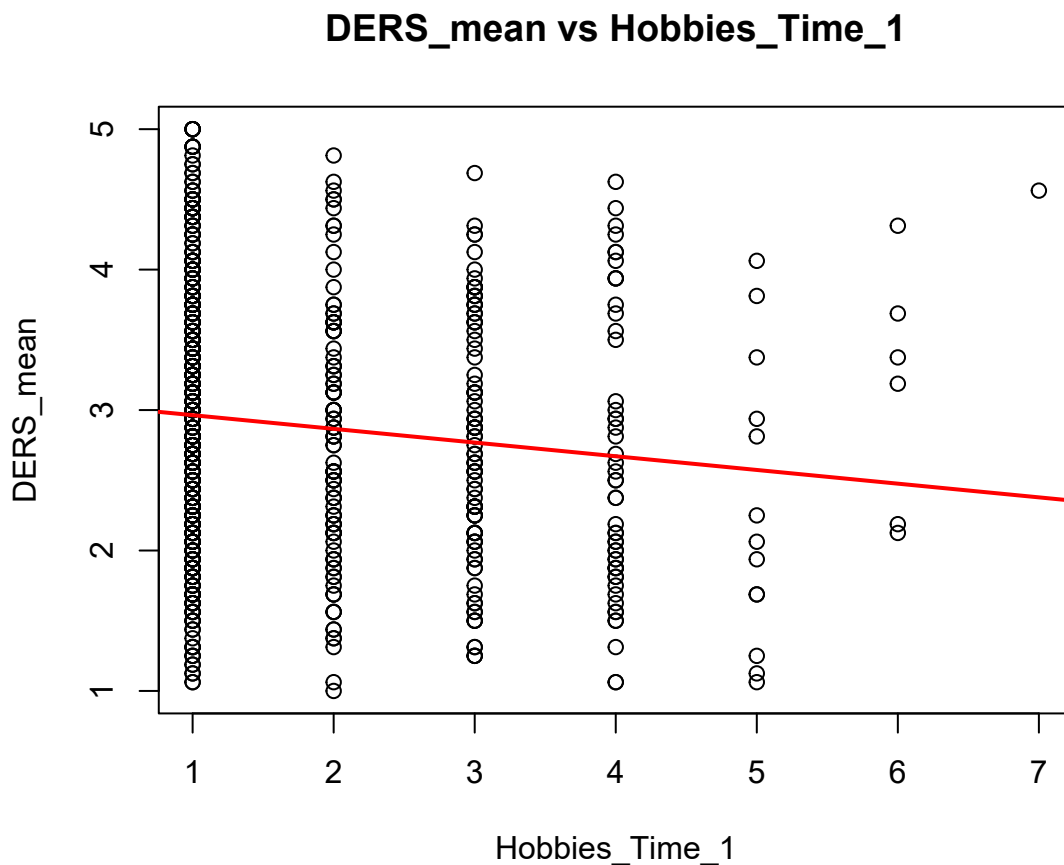
```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## Hobbies_Imp_4    1    7.36   7.3576    8.815 0.003083 **
## Residuals      746 622.67   0.8347
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## Graphs

### DERS\_mean vs Hobbies\_Time\_1

```
plot(DERS_mean ~ Hobbies_Time_1,
     data = Data_analysis,
     xlab = "Hobbies_Time_1",
     ylab = "DERS_mean",
     main = "DERS_mean vs Hobbies_Time_1")

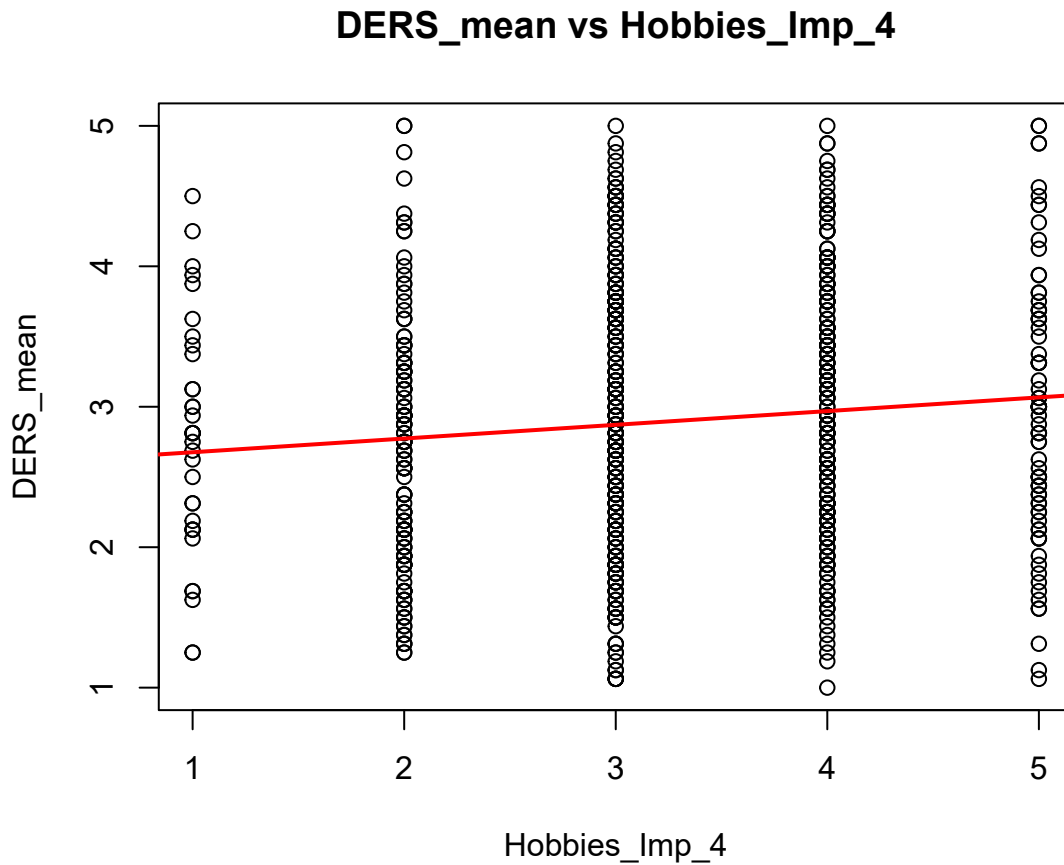
abline(lm_time1, col = "red", lwd = 2)
```



### DERS\_mean vs Hobbies\_Imp\_4

```
plot(DERS_mean ~ Hobbies_Imp_4,
     data = Data_analysis,
     xlab = "Hobbies_Imp_4",
     ylab = "DERS_mean",
     main = "DERS_mean vs Hobbies_Imp_4")
```

```
abline(lm_imp4, col = "red", lwd = 2)
```



## Conclusion

From comparing all 16 hobby items individually along with `Imp_overall` and `Time_overall` to the DERS, two variables stood out as having small but statistically significant relationships to Difficulties in Emotional Regulation Scale (DERS). These are:

Time spent on `Hobbies_Time_1`, which asked “How many hours per week do you spend participating in athletics, such as varsity sports or intramurals?”, showed a slight negative association with DERS. This suggests that students who spend more time in athletic activities tend to less often have difficulties in emotional regulation.

In contrast, the perceived importance of `Hobbies_Imp_4`, which asked “How important is watching online recreational content such as on Netflix or YouTube to you?”, showed a small positive association. This suggests that students who view watching online recreational content as more important tend to more often have difficulties in emotional regulation.

While these correlations are not large, they suggest that both the amount of time invested in certain hobbies and the personal value assigned to them contribute to mental well-being.