# Math 133: Statistical Learning Methods - Module 1 Assessment

Pranav Jayakumar

February 19, 2025

#### Abstract

In this module assessment, we will draw from survey data and fit simple linear models to understand the effects of various predictors on students' stress.

## 1 Data Analysis

For the following analysis, we will fit simple linear models of the form

$$Y = \beta_0 + \beta_1 X$$

where  $Y = \mathtt{stress\_score}$  and X is one of the following numeric features:  $\mathtt{maximum\_alcohol\_consumed}$ ,  $\mathtt{gpa}$ , or  $\mathtt{screen\_time}$ .

### 1.1 Data Visualization

We will begin by creating scatterplots of students' stress scores vs. each of the three aforementioned predictors. We will use a 70-30 training-testing split for our models.

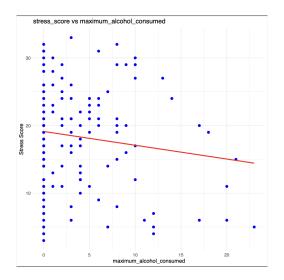


Figure 1: Scatterplot of Stress Score vs. Maximum Alcohol Consumed

The linear model for stress score vs. maximum alcohol consumed is represented by:

$$\hat{y} = 18.8543 - 0.2999x$$

We observe an  $\mathbb{R}^2$  score of 0.0417.

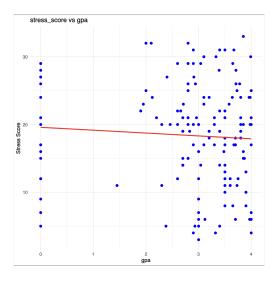


Figure 2: Scatterplot of Stress Score vs. GPA

The linear model for stress score vs. GPA is represented by:

$$\hat{y} = 19.8029 - 0.3343x$$

We observe an  $R^2$  score of 0.0027.

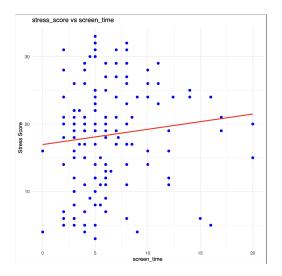


Figure 3: Scatterplot of Stress Score vs. Screen Time

The linear model for stress score vs. screen time is represented by:

$$\hat{y} = 16.7664 + 0.3193x$$

We observe an  $\mathbb{R}^2$  score of 0.0222.

### 1.2 Insights

We observe that the linear model  $stress\_score^-screen\_time$  has the highest  $R^2$  score. Therefore, there is evidence to suggest that screen time is the best out of the three predictors of stress.

Plotting screen time data in a histogram shows that the data is approximately normally distributed.

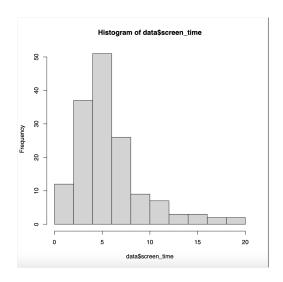


Figure 4: Histogram of Screen Time

## 2 R Coding

```
#!/usr/bin/env Rscript
  library(ggplot2)
  set.seed(123)
fit_linear_model <- function(y, x, raw_data) {</pre>
    # Train-test split
    n <- nrow(raw_data)</pre>
    trainIndex <- sample(n, round(0.7 * n, 0))</pre>
    train <- raw_data[trainIndex, ]</pre>
10
    test <- raw_data[-trainIndex, ]</pre>
11
12
    # Construct formula
13
    formula <- as.formula(paste(y, "~", x))</pre>
14
15
    # Fit model
16
    model <- lm(formula, data = train)</pre>
17
18
    overview = summary(model)
19
    print(overview)
20
21
    # Predict on testing data
22
    y_test <- test[[y]]</pre>
23
    y_hat <- predict(model, newdata = test)</pre>
24
25
    # Analyze accuracy
26
    SSE <- sum((y_test - y_hat)^2)</pre>
27
    MSE <- SSE / nrow(test)</pre>
28
    RMSE <- sqrt(MSE)
```

```
SST <- sum((y_test - mean(y_test))^2)</pre>
    R2 <- 1 - SSE / SST
31
32
    return(list(SSE = SSE, MSE = MSE, RMSE = RMSE, SST = SST, R2 = R2))
33
34 }
35
36 main <- function() {
    # Load data
37
    data <- read.csv("../../data/survey_fall2023.csv")</pre>
38
    # Define predictors
40
    predictors <- c("maximum_alcohol_consumed", "gpa", "screen_time")</pre>
41
    results <- list()
42
43
    # Open PDF device for saving all plots
44
    pdf("Rplots.pdf") # This will save all plots sequentially to Rplots.
45
        pdf
46
    # Iterate over predictors
47
    for (predictor in predictors) {
48
      if (!predictor %in% colnames(data)) {
        cat("\nWarning: Predictor", predictor, "not found in dataset.
            Skipping...\n")
51
        next
      }
52
53
      cat("\nLinear_model_for_stress_score_", predictor, "\n")
54
      result <- fit_linear_model("stress_score", predictor, data)</pre>
55
56
57
      # Store results for later comparison
58
      results[[predictor]] <- result</pre>
      # Format and print results
60
      formatted_results <- lapply(result[1:5], function(x) format(round(x,</pre>
61
          4), nsmall = 4))
      print(formatted_results)
62
63
      # Create scatterplot for the predictor
64
      ggplot(data, aes_string(x = predictor, y = "stress_score")) +
65
        geom_point(color = "blue", size = 2) +
66
        geom_smooth(method = "lm", color = "red", se = FALSE) +
67
        ggtitle(paste("stress_score_uvs", predictor)) +
68
        xlab(predictor) +
        ylab("Stress<sub>□</sub>Score") +
70
        theme_minimal() -> plot # Assign ggplot to a variable
71
72
      print(plot) # Ensure the plot is sent to the PDF file
73
74
75
    hist(data$screen_time, breaks = "Freedman-Diaconis") -> plot1
76
    print(plot1)
77
78
    # Close PDF device
```

```
dev.off()

dev.off()

cat("\nAll_uplots_usaved_uto_uRplots.pdf.\n")

Run the script if executed directly

if (interactive() || identical(Sys.getenv("R_SCRIPT"), "")) {
    main()
}
```