

Assignment 3

Anna Hipp Kaplan, Jona Gavazi

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```
# First, we will want to install the appropriate packages "tidyr" and "ggplot2"  
# and then utilize the library() function to load in the packages and their  
# dependencies. We also will name an object "text" and read the .csv file  
# TextMessages unto it, upon which we will generate our visual summaries  
# and run the relevant descriptive statistics.
```

```
text <- read.csv("TextMessages.csv")  
#install.packages("ggplot2")  
library(ggplot2)  
#install.packages("tidyr")  
library(tidyr)  
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'  
  
## The following objects are masked from 'package:stats':  
##  
##   filter, lag  
  
## The following objects are masked from 'package:base':  
##  
##   intersect, setdiff, setequal, union
```

Visualization 1

```
# To get a quick understanding of the exact names of the variables involved,  
# we can simply run the names() function, followed by the glimpse() command, to  
# ensure that the data are in order and accurately reflect the csv file.
```

```
names(text)
```

```
## [1] "Group"      "Baseline"   "Six_months" "Participant"
```

```
glimpse(text)
```

```
## Rows: 50  
## Columns: 4  
## $ Group      <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1~  
## $ Baseline    <int> 52, 68, 85, 47, 73, 57, 63, 50, 66, 60, 51, 72, 77, 57, 79~  
## $ Six_months  <int> 32, 48, 62, 16, 63, 53, 59, 58, 59, 57, 60, 56, 61, 52, 9,~  
## $ Participant <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17,~
```

```
# We want to first covert the variables Baseline and Six_months into numeric  
# format, to ensure that they can be utilized for quantitative analyses and
```

```
# visualizations. We then need to reshape the dataset from wide format to
# long format, using the pivot_longer() command. In the original wide structure,
# each participant featured separate columns for the number of text messages
# sent at Baseline and at Six months. After reshaping, these two columns are
# combined into a single variable called TextMessages, with a corresponding
# Time variable, which indicates whether each observation came from the
# Baseline or Six-month time point. This transformation will serve to make
# the data tidy, which is ideal for plotting/faceting and conducting
# group/time comparisons within ggplot2.
```

```
text_long <- text %>% mutate(Baseline = as.numeric(Baseline),
  Six_months = as.numeric(Six_months)) %>% pivot_longer(
  cols = c(Baseline, Six_months),
  names_to = "Time",
  values_to = "TextMessages")
```

```
# We want to refine the structure by explicitly converting key variables into
# factors with defined levels. The Time variable (split between indication of
# Baseline or Six months) will be converted into a factor and ordered, so that
# Baseline appears first, ensuring consistency in the plots. The Group variable
# will also be converted into a factor, so that it represents categorical data
# rather than numeric values. In defining these variables as factors, we
# ensure that ggplot2 treats them as categorical axes rather than continuous
# scales.
```

```
text_long <- text_long %>% mutate(Time = factor(Time,
  levels = c("Baseline", "Six_months")),
  Group = as.factor(Group))
```

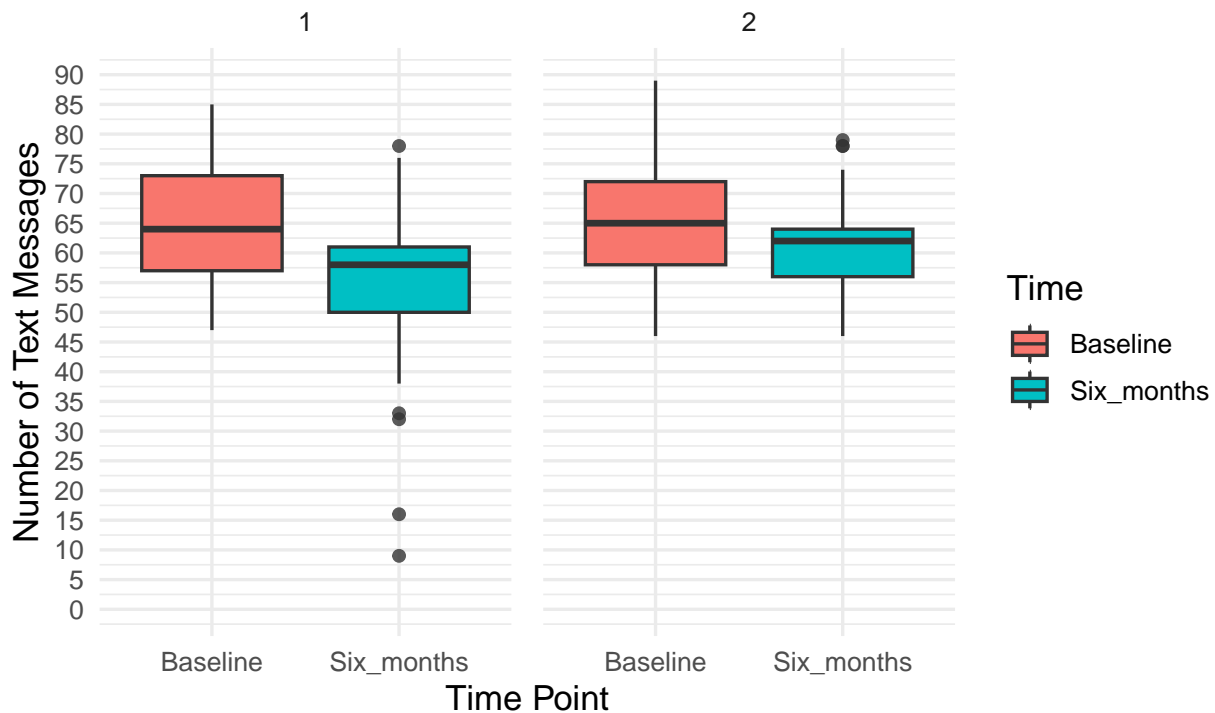
```
# VISUALIZATION 1:
```

```
# We want to generate a set of faceted boxplots that display the distribution
# of texts across the two time points for each participant group. The function
# ggplot() will map Time to the x-axis, TextMessages to the y-axis, and Time
# to the fill color, to create distinction between the periods. We will then
# utilize the geom_boxplot() layer to create boxplots that show the central
# tendency, variability, and potential outliers. The facet_wrap(~ Group)
# function produces separate panels for each group, allowing for side-by-side
# comparisons for texting behaviors.
```

```
boxplot1 <- ggplot(text_long, aes(x = Time, y = TextMessages, fill = Time)) +
  geom_boxplot(outlier.alpha = 0.8) + facet_wrap(~ Group) + labs(
  title = "Distribution of Text Messages by Time and Group",
  subtitle = "Faceted by Group; boxplots show spread at Baseline vs Six_months",
  x = "Time Point",
  y = "Number of Text Messages",
  fill = "Time") + scale_y_continuous(limits = c(0, 90),
  breaks = seq(0, 90, by = 5)) + theme_minimal(base_size = 13) + theme(
  plot.title = element_text(face = "bold"),
  panel.spacing = unit(12, "pt"))
print(boxplot1)
```

Distribution of Text Messages by Time and Group

Faceted by Group; boxplots show spread at Baseline vs Six_months



*# The faceted boxplots illustrate the distribution of text counts, for each
participant Group (1, 2) across two time points (Baseline, Six_months).
In both groups, the median number of text messages is moderately high, with
slightly greater variability at Baseline, as compared to Six months (indicated
by the wider interquartile range and the presence of a few lower outliers).
Across groups, the pattern suggests a small decline in text frequency over
time, with the central tendencies remaining fairly consistent.
The consistency of medians across groups and the overlapping interquartile
ranges thus indicate that no dramatic behavioral shift occurred, with a rather
gradual convergence toward more uniform texting patterns over time (as
indicated by the compactness of the boxes at the Six_months mark). The
whiskers demonstrate that most of the values fall within a similar overall
range.*

Visualization 2

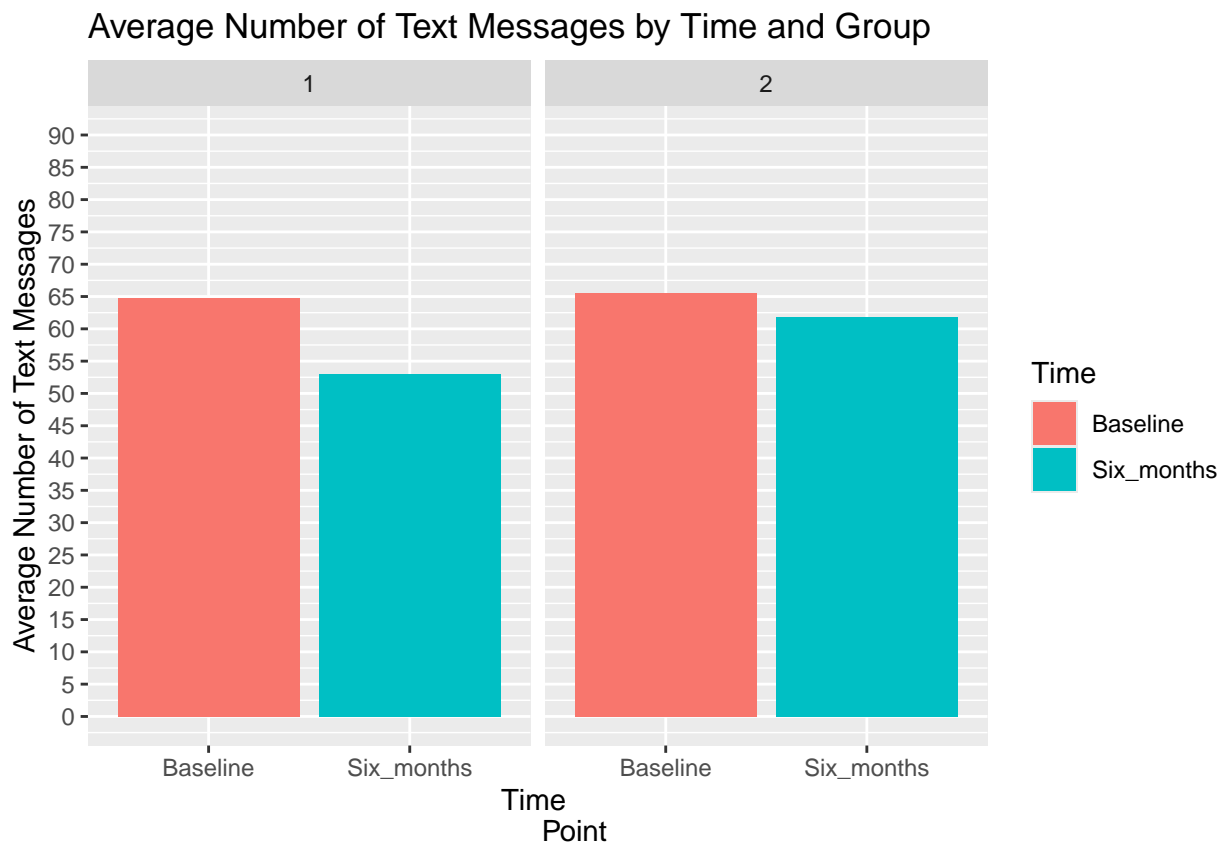
*# create stratified bar charts of text messages Group and Time using our
converted data file text_long from above. We want to generate a set of faceted
bar charts that display the distribution of texts across the two time points
for each participant group. Using the function ggplot to map the variable
Time that we created to the x-axis and the variable TextMessages to the
y-axis, we generate these bar charts. We also utilized Time to differentiate
the fill color of each time period (Baseline and Six_months). geom_bar()
creates bar charts where the height of the bar proportional to the number of
cases in each group, showing us the average number of text messages sent by
each group at each time period. Finally, using facet_wrap(~ Group) function*

```
# produces separate panels for each group, allowing us to generate side-by-side
# comparisons.
```

```
barchart <- ggplot(text_long, aes(x = Time, y = TextMessages, fill = Time)) +
  geom_bar(stat = "summary", fun = "mean", position = "dodge") +
  facet_wrap(~ Group) +
  labs(title = "Average Number of Text Messages by Time and Group", x = "Time
    Point", y = "Average Number of Text Messages", fill = "Time") +
  scale_y_continuous(limits = c(0,90), breaks = seq(from = 0, to = 90, by =
    5))
```

```
# view bar chart
```

```
barchart
```



```
# This plot shows the average number of text messages sent in each group, at
# both the baseline and six month mark of the observational period. On the left
# side, we see Group 1, who sent an approximate average of 65 texts at the
# beginning of the observational period and an approximate average of 53 text
# messages sent at the six month mark. On the right side, we see Group 2, who
# also sent an approximate average of 65 text messages at the beginning of the
# observational period, and an approximate average of 62 text messages at the
# six month mark. From this plot, it seems that both groups had a decline in
# the amount of texting they did, with Group 1 having a steeper decline between
# the beginning of the observational period and the six month mark.
```

Summary Statistics

```
# compute summary statistics by Group and Time to examine the number of text  
# messages sent by participants in each group at baseline and at six months.
```

```
summary_stats <- text_long %>%  
  group_by(Group, Time) %>%  
  summarise(  
    n = n(),  
    mean = mean(TextMessages),  
    sd = sd(TextMessages),  
    min = min(TextMessages),  
    max = max(TextMessages))
```

```
## `summarise()` has grouped output by 'Group'. You can override using the  
## `.groups` argument.
```

```
# view results  
print(summary_stats)
```

```
## # A tibble: 4 x 7  
## # Groups:   Group [2]  
##   Group Time      n mean   sd   min   max  
##   <fct> <fct>   <int> <dbl> <dbl> <dbl> <dbl>  
## 1 1     Baseline    25  64.8 10.7   47    85  
## 2 1     Six_months  25  53.0 16.3    9    78  
## 3 2     Baseline    25  65.6 10.8   46    89  
## 4 2     Six_months  25  61.8  9.41  46    79
```

```
# For Group 1, the mean number of text messages decreased from 64.84  
# (SD = 10.68) at baseline to 52.96 (SD = 16.33) at six months, suggesting a  
# reduction in texting activity over time. Message counts ranged from 47 to 85  
# at baseline and from 9 to 78 at six months, the largest range in our data. For  
# Group 2, the mean number of text messages showed a smaller decline, from 65.60  
# (SD = 10.84) at baseline to 61.84 (SD = 9.41) at six months. Message counts  
# ranged from 46 to 89 at baseline and from 46 to 79 at six months. Overall,  
# both groups exhibited a decrease in texting activity over the six-month  
# period, with Group 1 seeming to show a larger reduction on average than Group  
# 2. This could be a point of further analysis.
```

```
# The collective of the findings from the graphics and summary statistics  
# present a consistent narrative about participants' texting behaviors over the  
# six month observation period. Both visualizations reveal that text message  
# frequency declined slightly from Baseline to Six_months in both groups, with  
# Group 1 showing a more pronounced decrease. The summary statistics reinforce  
# these patterns, as Group 1's mean text count dropped from ~64.84 to 52.96,  
# while Group 2's mean declined more modestly from 65.60 to 61.84. The boxplots  
# further illustrate that the variability in texting behavior decreased at the  
# Six_month mark, pointing to activity levels becoming more consistent over time.  
# The results suggest that while both groups maintained broadly similar  
# communication habits, there was a subtle overall reduction in texting activity,  
# and a convergence toward greater behavioral consistency across the two time  
# points, with only mild group-level differences in degree of change.
```