W241 Final Project Google Forms Analysis

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```
# Load Libraries and Dataset
library(tidyverse)
## Warning: package 'ggplot2' was built under R version 4.3.3
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.2 v readr
                                   2.1.4
## v forcats 1.0.0 v stringr 1.5.0
## v ggplot2 3.5.0
                    v tibble
                                   3.2.1
                     v tidyr
## v lubridate 1.9.2
                                   1.3.0
## v purrr
              1.0.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(scales)
## Warning: package 'scales' was built under R version 4.3.3
##
## Attaching package: 'scales'
## The following object is masked from 'package:purrr':
##
      discard
## The following object is masked from 'package:readr':
##
      col_factor
library(effsize)
```

Warning: package 'effsize' was built under R version 4.3.3

```
file_path <- "C:/Users/anniy/Downloads/coffee_drinking_dataset_for_analysis.csv"
df <- read.csv(file_path, stringsAsFactors = FALSE)</pre>
df <- df %>% filter(rowSums(is.na(.)) < ncol(.))</pre>
df <- df[!is.na(df[, 7]), ]</pre>
glimpse(df)
## Rows: 44
## Columns: 13
## $ Timestamp
## $ Name
## $ Email
## $ Do.you.drink.at.least.one.cup.of.coffee.4.times.a.week.
## $ What.time.did.you.wake.up.this.morning.
## $ What.time.did.you.drink.your.coffee.....A..On.day.1.you.need.to.drink.coffee.20.minutes.after.waki:
## $ How.many.hours.has.it.been.since.you.woke.up...For.testing.accuracy..please.complete.this.survey.6
## $ How.awake.do.you.feel.right.now...scale.1.5.
## $ How.difficult.was.it.to.focus.on.tasks.today....scale.1.5.
## $ How.physically.tired.do.you.feel.right.now.....scale.1.5.
## $ How.much.coffee.did.you.drink.today.
## $ Did.you.notice.any.differences.in.your.alertness.today.compared.to.yesterday...Open.ended.
## $ Any.additional.comments.about.your.experience.with.drinking.coffee.at.this.time...Open.ended.
```

URL to the Coffee Study https://tinyurl.com/coffeestudy2025

Data Cleaning:

```
# Clean values from our excel sheet
colnames(df) <- make.names(colnames(df))</pre>
df <- df %>%
 rename(
   wake_time = What.time.did.you.wake.up.this.morning.,
    coffee_time = What.time.did.you.drink.your.coffee.....A..On.day.1.you.need.to.drink.coffee.20.minut
   hours_since_wake = How.many.hours.has.it.been.since.you.woke.up...For.testing.accuracy..please.comp
   alertness = How.awake.do.you.feel.right.now...scale.1.5.,
   focus_difficulty = How.difficult.was.it.to.focus.on.tasks.today....scale.1.5.,
    physical_tiredness = How.physically.tired.do.you.feel.right.now.....scale.1.5.
extract_number <- function(x) {</pre>
  as.numeric(gsub("[^1-5]", "", x))
df$coffee_amount_raw <- df$How.much.coffee.did.you.drink.today.</pre>
# Clean and convert it to numeric
df$coffee_cups <- df$coffee_amount_raw %>%
 tolower() %>%
 str_extract("\\d+(\\.\\d+)?") %>% # extract number like "1" or "1.5"
```

```
df$alertness <- extract_number(df$alertness)
df$focus_difficulty <- extract_number(df$focus_difficulty)
df$physical_tiredness <- extract_number(df$physical_tiredness)
df$hours_since_wake <- as.numeric(df$hours_since_wake)

# Convert time to datetime and calculate coffee delay
df$wake_time <- as.POSIXct(df$wake_time, format = "%I:%M:%S %p")
df$coffee_time <- as.POSIXct(df$coffee_time, format = "%I:%M:%S %p")</pre>
df$coffee_delay_mins <- as.numeric(difftime(df$coffee_time, df$wake_time, units = "mins"))
```

Identify Compliant Participants

```
# Define compliance flags
df$compliant_survey_time <- ifelse(df$hours_since_wake >= 5 & df$hours_since_wake <= 7, 1, 0)
df$compliant_coffee_time <- ifelse(
    df$coffee_delay_mins <= 30 | (df$coffee_delay_mins >= 90 & df$coffee_delay_mins <= 150),
    1, 0
)

# Define treatment vs control group
df$treatment_group <- ifelse(df$coffee_delay_mins >= 90, 1, 0)
df$treatment_label <- ifelse(df$treatment_group == 1, "Treatment", "Control")

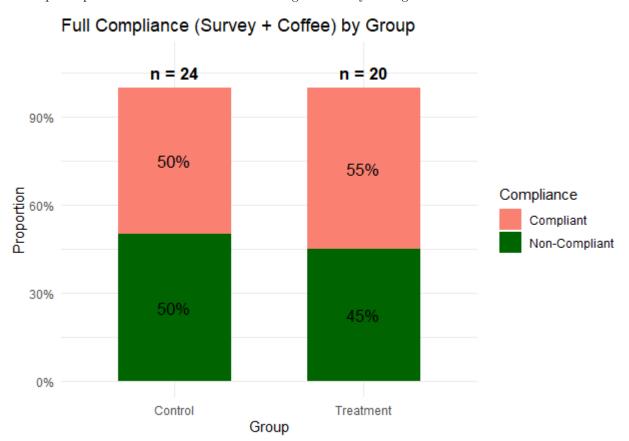
# Creating a new df to filter for compliant participants only
df_compliant <- df %>%
filter(compliant_survey_time == 1 & compliant_coffee_time == 1)
```

Note: We defined compliance with the survey timing as completing the survey between 5 and 7 hours after waking, rather than exactly at the 6-hour mark. This allowed us to include participants who followed the instructions closely, even if not perfectly. We believe this +- 1 hour window may help us maintain data quality and ensure a bigger sample size for our analysis.

Compliance Summary

```
total_counts <- df %>%
  filter(!is.na(treatment_group)) %>%
  group_by(treatment_label) %>%
  summarise(total = n())
# Count compliant participants per group
compliant_counts <- df %>%
  filter(compliant_survey_time == 1 & compliant_coffee_time == 1) %>%
  group_by(treatment_label) %>%
  summarise(compliant = n())
# Merge the two summaries
group_summary <- left_join(total_counts, compliant_counts, by = "treatment_label") %>%
  mutate(compliant = replace_na(compliant, 0),
         compliance_rate = paste0(compliant, "/", total, " (", round(100 * compliant / total), "%)"))
print(group_summary)
## # A tibble: 2 x 4
     treatment_label total compliant compliance_rate
##
                               <int> <chr>
                     <int>
## 1 Control
                        24
                                  12 12/24 (50%)
## 2 Treatment
                        20
                                  11 11/20 (55%)
```

Our compliance rates were 50% in the control group and 55% in the treatment group. Only about half of participants followed both the coffee timing and survey timing instructions.



How can we improve this?

Our group discussed a few ways after the experiment to improve future compliance rates.

For any future experiments with human participants, we can:

- 1. Send Automated text/email reminders right before their target coffee time or survey time
- 2. Include a Mobile survey link with push notifications (although I'm not sure if google forms would have that feature)
- 3. Add Incentives for full compliance, like a gift card raffle or bonus points. Anni has offered a \$25 starbucks giftcard to only one random participant.
- 4. We can use a third party App-based logging tool with built-in timers to guide them through Day A and Day B. We would need to search other survey tools to identify the right one

Linear Regression Model

```
model <- lm(alertness ~ treatment_group + focus_difficulty + physical_tiredness + hours_since_wake + co
summary(model)
##
## Call:
## lm(formula = alertness ~ treatment_group + focus_difficulty +
       physical_tiredness + hours_since_wake + coffee_cups, data = df_compliant)
##
##
## Residuals:
                      Median
##
       Min
                  1Q
                                    3Q
  -0.54051 -0.27362 0.02837
                              0.22219
                                        0.67300
##
##
## Coefficients:
                      Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                       -0.3326
                                   1.3857
                                           -0.240 0.81355
## treatment_group
                        0.3554
                                   0.2024
                                            1.756 0.09951 .
## focus difficulty
                        0.3756
                                   0.1394
                                            2.695
                                                   0.01661 *
## physical_tiredness
                        0.5710
                                   0.1717
                                            3.325
                                                   0.00462 **
## hours_since_wake
                        0.0649
                                   0.2081
                                            0.312 0.75939
## coffee_cups
                        0.1284
                                   0.1392
                                            0.922 0.37095
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3958 on 15 degrees of freedom
     (2 observations deleted due to missingness)
## Multiple R-squared: 0.8299, Adjusted R-squared: 0.7731
## F-statistic: 14.63 on 5 and 15 DF, p-value: 2.595e-05
```

The updated linear regression model shows that participants who delayed their coffee had 0.36 higher alertness scores on average compared to the control group, although this effect was still only marginally significant (p = 0.10), with a standard error of 0.20.

 focus_difficulty and physical_tiredness goes into the Alert index score and should have somewhat of a correlation.

- Adding how much cups of coffee participants drank that day to the model didn't improve predictive power significantly. This suggests that timing of caffeine may be more important than the quantity consumed
- While we observed a possible trend that delaying caffeine may improve alertness, our sample size of compliant participants using the google sheets was small. This limited our ability to detect a statistically significant effect with our models.
- Based on our earlier power calculation, we would definitely **need more participants** (especially compliant ones) to reliably detect an effect of this size. With more data, the standard errors would likely shrink, making it easier to confirm whether this trend holds up.

```
t.test(alertness ~ treatment_group, data = df_compliant)
```

```
##
## Welch Two Sample t-test
##
## data: alertness by treatment_group
## t = -0.28814, df = 17.922, p-value = 0.7765
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
## -0.8796327    0.6675115
## sample estimates:
## mean in group 0 mean in group 1
##    4.166667    4.272727
```

We ran a Welch's t-test and found no significant difference in alertness between the control group (mean = 4.17) and the treatment group (mean = 4.27), with a p-value of 0.78. The 95% confidence interval ranged from -0.88 to +0.67, meaning the true effect could go either way.

```
library(effsize)
cohen.d(alertness ~ treatment_group, data = df_compliant)

## Warning in cohen.d.formula(alertness ~ treatment_group, data = df_compliant):
## Cohercing rhs of formula to factor

##
## Cohen's d
##
## d estimate: -0.1220871 (negligible)
## 95 percent confidence interval:
## lower upper
## -0.9909737 0.7467995
```

Cohen's d was **-0.12**, which is considered a **negligible effect size**. This shows us that the difference in alertness between groups was very small overall.

