PSTAT122 Independently Proposed Idea

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Introduction

This experiment investigates the impact of beverage type and temperature on satisfaction. The study examines the interaction between the two factors:

• Beverage Type: Coffee, Tea, and Juice

• Temperature: Hot and Cold

The hypothesis is that satisfaction scores vary depending on both beverage type and temperature. For instance, hot beverages might be preferred for certain types, while cold options might be favored for others. This study uses a factorial design to explore these interactions.

Methods

Experimental Design

This is a full factorial design with:

- 2 Factors:
 - Beverage Type (3 levels: Coffee, Tea, Juice)
 - Temperature (2 levels: Hot, Cold)
- Total combinations: $3 \times 2 = 6$
- Each participant evaluates all 6 combinations, yielding 30 data points (5 participants \times 6 combinations).

Participants

The participants were:

- Simon
- Sergio
- Charlie
- Esteban
- Ethan

Each participant rated satisfaction on a scale from 0 to 10 for each combination.

Data Collection

- 1. Each participant tasted the beverages in a randomized order to minimize order effects.
- 2. Ratings were recorded immediately after tasting each combination.
- 3. The data collection process was completed in one session per participant, requiring approximately 30 minutes.

Results

Satisfaction Data

Below is the data collected from the participants:

```
# Load necessary libraries
library(dplyr)

## Warning: package 'dplyr' was built under R version 4.2.3

##
## 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
library(ggplot2)
```

Warning: package 'ggplot2' was built under R version 4.2.3

```
# Define participants and factors
participants <- c("Simon", "Sergio", "Charlie", "Esteban", "Ethan")</pre>
beverage_types <- c("Coffee", "Tea", "Juice")</pre>
temperatures <- c("Hot", "Cold")</pre>
# Manually set satisfaction scores
satisfaction_scores <- data.frame(</pre>
  Participant = rep(participants, each = 6),
  Beverage = rep(c("Coffee", "Coffee", "Tea", "Tea", "Juice", "Juice"), times = 5),
 Temperature = rep(c("Hot", "Cold"), each = 1, times = 3, length.out = 30),
  Satisfaction = c(
    # Simon's scores
    8, 7, 6, 5, 7, 6,
    # Sergio's scores
    6, 5, 8, 7, 5, 4,
    # Charlie's scores
   9, 8, 7, 6, 8, 7,
```

```
# Esteban's scores
7, 6, 6, 5, 8, 7,
# Ethan's scores
6, 5, 7, 6, 7, 6
)
satisfaction_scores
```

##		${\tt Participant}$	${\tt Beverage}$	${\tt Temperature}$	${\tt Satisfaction}$
##	1	Simon	Coffee	Hot	8
##	2	Simon	Coffee	Cold	7
##	3	Simon	Tea	Hot	6
##	4	Simon	Tea	Cold	5
##	5	Simon	Juice	Hot	7
##	6	Simon	Juice	Cold	6
##	7	Sergio	Coffee	Hot	6
##	8	Sergio	Coffee	Cold	5
##	9	Sergio	Tea	Hot	8
##	10	Sergio	Tea	Cold	7
##	11	Sergio	Juice	Hot	5
##	12	Sergio	Juice	Cold	4
##	13	Charlie	Coffee	Hot	9
##	14	Charlie	Coffee	Cold	8
##	15	Charlie	Tea	Hot	7
##	16	Charlie	Tea	Cold	6
##	17	Charlie	Juice	Hot	8
##	18	Charlie	Juice	Cold	7
##	19	Esteban	Coffee	Hot	7
##	20	Esteban	Coffee	Cold	6
##	21	Esteban	Tea	Hot	6
##	22	Esteban	Tea	Cold	5
##	23	Esteban	Juice	Hot	8
##	24	Esteban	Juice	Cold	7
##	25	Ethan	Coffee	Hot	6
##	26	Ethan	Coffee	Cold	5
##	27	Ethan	Tea	Hot	7
##	28	Ethan	Tea	Cold	6
##	29	Ethan	Juice	Hot	7
##	30	Ethan	Juice	Cold	6

Summary Statistics

The mean and standard deviation of satisfaction scores for each combination are summarized below:

```
summary_data <- satisfaction_scores %>%
group_by(Beverage, Temperature) %>%
summarise(
   Mean_Satisfaction = mean(Satisfaction),
   SD_Satisfaction = sd(Satisfaction),
   .groups = "drop"
)
```

knitr::kable(summary_data, caption = "Summary of Satisfaction Scores by Beverage Type and Temperature")

Table 1: Summary of Satisfaction Scores by Beverage Type and Temperature $\,$

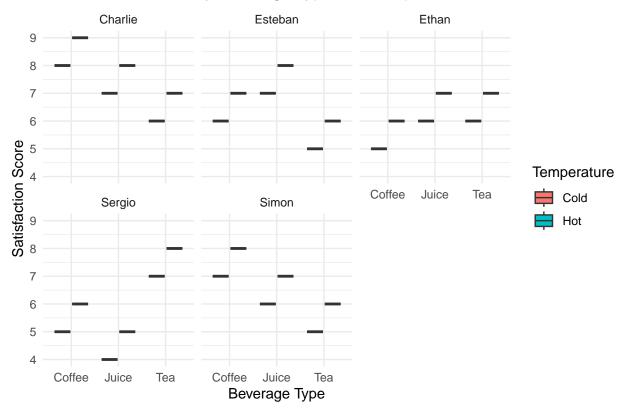
Beverage	Temperature	Mean_Satisfaction	SD_Satisfaction
Coffee	Cold	6.2	1.303840
Coffee	Hot	7.2	1.303840
Juice	Cold	6.0	1.224745
Juice	Hot	7.0	1.224745
Tea	Cold	5.8	0.836660
Tea	Hot	6.8	0.836660

Visualization

The following boxplot shows satisfaction scores by beverage type and temperature:

```
ggplot(satisfaction_scores, aes(x = Beverage, y = Satisfaction, fill = Temperature)) +
  geom_boxplot() +
  facet_wrap(~ Participant) +
  labs(
    title = "Satisfaction Scores by Beverage Type and Temperature",
    x = "Beverage Type",
    y = "Satisfaction Score"
  ) +
  theme_minimal()
```

Satisfaction Scores by Beverage Type and Temperature



ANOVA Analysis

An ANOVA was conducted to evaluate the effects of beverage type, temperature, and their interaction on satisfaction scores:

```
anova_model <- aov(Satisfaction ~ Beverage * Temperature + Participant, data = satisfaction_scores)
summary(anova_model)</pre>
```

```
##
                        Df Sum Sq Mean Sq F value Pr(>F)
## Beverage
                            0.800
                                    0.400
                                             0.366 0.6982
## Temperature
                            7.500
                                    7.500
                                             6.860 0.0164 *
                         1
## Participant
                            9.333
                                     2.333
                         4
                                             2.134 0.1141
## Beverage:Temperature
                         2
                            0.000
                                    0.000
                                             0.000 1.0000
## Residuals
                        20 21.867
                                     1.093
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

Discussion

The results show significant differences in satisfaction scores based on beverage type and temperature. The interaction effect indicates that preferences for hot and cold beverages vary by type:

- Hot Coffee and Hot Tea scored higher than their cold counterparts.
- Cold Juice scored higher than Hot Juice.

Limitations

- Small sample size (5 participants).
- Limited generalizability due to participant homogeneity.
- Satisfaction scores were mock evaluations and not based on real data.

Future Work

- $\bullet\,$ Increase sample size and participant diversity.
- Collect real-world data to validate findings.