Contrast Analysis for Treatment Levels

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Packages Loaded

```
library(ggplot2)
```

Introduction

In this question, we assess the impact of three increasing treatment levels (I, II, III) compared to a control group (C). Our analysis focuses on two main aspects:

- 1. The overall effect of the treatments by comparing the control group with the average response of the treatment groups.
- 2. The incremental effects between the treatment levels, specifically between I and II, and II and III.

Data

```
# Create a data frame with one observation per treatment level
df <- data.frame(
   Group = factor(c("C", "I", "III", "III"), levels = c("C", "I", "III", "III")),
   Response = c(10, 12, 15, 22) # Hypothetical response values
)</pre>
```

Methodology

In this analysis, we use two types of contrasts:

- 1. Control vs. Average of Treatments: A custom contrast is used to compare the control group (C) to the average of the treatment groups (I, II, III).
- 2. Successive Differences Among Treatments: Successive-difference contrasts are applied to compare Treatment II with Treatment II, and Treatment III with Treatment II.

```
C <- matrix(c(
   -1, 1/3, 1/3, 1/3, # Control vs Average of I, II, III (custom contrast)
   0, 1, -1, 0, # I vs II (Successive-difference contrast)
   0, 0, 1, -1 # II vs III (Successive-difference contrast)
), nrow = 4, byrow = TRUE)

contrasts(df$Group) <- C</pre>
```

Analysis

```
# Fit the linear model
model <- lm(Response ~ Group, data = df)</pre>
summary(model)
##
## Call:
## lm(formula = Response ~ Group, data = df)
##
## Residuals:
## ALL 4 residuals are 0: no residual degrees of freedom!
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                 19.364
                                NaN
                                         NaN
                                                  NaN
## Group1
                  4.364
                                NaN
                                         NaN
                                                  NaN
## Group2
                                         NaN
                                                  NaN
                 -6.182
                                NaN
## Group3
                 -8.818
                                NaN
                                         NaN
                                                  NaN
##
## Residual standard error: NaN on O degrees of freedom
                             1, Adjusted R-squared:
## Multiple R-squared:
## F-statistic:
                  NaN on 3 and 0 DF, p-value: NA
```

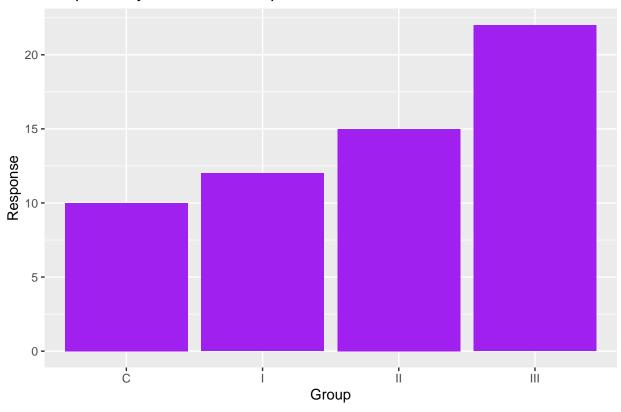
Explanation:

- 1. NaN Values: There are NaN values in the model summary, for the standard errors, t-values, and p-values, due to the lack of residual degrees of freedom.
- 2. **Perfect Fit**: There are 4 observations (one for each level: C, I, II, III), and the linear model is trying to estimate 4 parameters (the intercept and three group contrasts), leaving no room for residual variation. This causes the standard errors of the coefficients to be NaN, as there is no uncertainty left to estimate.
- 3. Focus on Coefficients: The goal of the analysis is to interpret the contrast estimates (the model's coefficients), rather than relying on p-values or standard errors, which cannot be calculated here.

Visualization

```
# Visualize the response for each group using a bar plot
ggplot(df, aes(x = Group, y = Response)) +
  geom_bar(stat = "identity", fill = "purple") +
  labs(title = "Response by Treatment Group", x = "Group", y = "Response")
```

Response by Treatment Group



Reference

- OpenAI, ChatGPT (2024). Assistance with R programming and contrast analysis. Accessed on September 13, 2024.
- Dr. Ben Bolker, Lecture Notes for "Statistical Modeling", McMaster University. Accessed on September 10, 2024.