Analysis of the randomized plan

35 marks

1 Problem

Recall the physical laboratory involving the plane. (See the file background.pdf for more information.) The equation of the plane can be written as

$$y = \alpha + \beta x + \gamma z$$

which has no error. All of the points lie exactly on the plane and both β and γ are unknown. The planes were arranged so that $\beta = 0.5$ for every run.

Values of x and y were recorded by teams for three different experimental protocols or plans. The third value z was not. In practice, the values of z are never known – there is always some variate that is not measured, perhaps not even thought of, that might be part of relating y to x. These are called **lurking variables** and they will always exist.

The purpose of this question is to investigate and compare the different experimental plans. Of particular interest is whether or not x causes y, that is testing $H_0: \beta = 0$. And if so, to estimate the value of β defining the causal relationship.

2 Plan

Three experimental plans were considered:

- "observational" where six (x, y) pairs were observed in a particular configuration by each team.
- "randomized" where three tower-markers were randomly allocated to each of only two different x values, from which the ys were determined. Each team produced two replicates here.
- "randomizedBlock" where tower markers were sorted into pairs by height and one marker of each pair were randomly assigned the lower of the two x values and the other to the higher x value. Again, y was determined after allocation for all six markers.

In this question, you will be working only with the data collected using the **randomized** plan. See the file background.pdf for more information.

3 Data

Set up the following:

```
## Set this up for your own directory
imageDirectory <- "MyAssignmentDirectory/img" # e.g. in current "./img"
dataDirectory <- "MyAssignmentDirectory/data" # e.g. in current "./data"
path_concat <- function(path1, ..., sep="/") paste(path1, ..., sep = sep)</pre>
```

The full data set is then read in as:

```
labData <- read.csv(file = path_concat(dataDirectory, "labData.csv"))</pre>
```

The data can be subsetted according to the three different experimental plans.

a. (1 mark) Select that subset of the data corresponding to the randomized plan. Assign it to the variable randomized. Show your code.

4 Analysis

b. (4 marks) Plot the (x, y) pairs from all of the randomized data

Use xlim = c(0, 30), ylim = c(0,40), pch = 19, col = adjustcolor("black", 0.3) in the call to plot().

Label the plot meaningfully.

Fit a straight line model of y on x and add this fitted line to the plot. Save the fit object. Report the value of the slope estimate.

Show your code.

- c. Learning from repetition. Each team executed the same plan. Moreover, each team replicated that execution. To gain a better appreciation of the qualities of that plan, we investigate the individual team estimates of β .
 - i. (2 marks) Separate the data into two subsets, one for each rep. Assign the two subsets to the variables rand1 and rand2 for replicates 1 and 2. Show your code.
 - ii. (4 marks) For each replication, fit a separate line for each team's data. For each replication, capture the slope estimates of each team's fit and collect these into a single vector. Call the vector for replication 1's slope estimates slopes1 and the same for replication 2's betas2.

Show your code.

ii. (4 marks) Plot the (betas1, betas2) pairs from the randomized data

Use xlim = c(-1, 1), ylim = c(-1, 1), pch = 19, col = adjustcolor("black", 0.3) in the call to plot().

Label the plot meaningfully.

Show your code.

iii. (4 marks) Test the hypothesis that the team paired slope estimators, $(\tilde{\beta}_1, \tilde{\beta}_2)$, based on replicates 1 and 2, are independently distributed. That is test $H_0: \tilde{\beta}_1 \perp \tilde{\beta}_2$.

Use numericalTest() with the appropriate choices of discrepancy measure and generation function.

Show your code.

Write up your conclusion about the independence.

iv. (3 marks) Draw a meaningfully labelled histogram of the individual slope coefficient estimates for all teams for **replicate 1** only.

Show your code.

Use xlim = c(-1, 1), col = "lightgrey" in hist() and an appropriate main title and xlab.

Add a vertical red dashed line at the average of the slope estimates.

Add a vertical blue dashed line at the true value of β .

Print the average and standard deviation of the slope estimates.

v. (3 marks) Draw a meaningfully labelled histogram of the individual slope coefficient estimates for all teams for **replicate 2** only.

Show your code.

Use xlim = c(-1, 1), col = "lightgrey" in hist() and an appropriate main title and xlab.

Add a vertical red dashed line at the average of the slope estimates.

Add a vertical blue dashed line at the true value of β .

Print the average and standard deviation of the slope estimates.

vi. (3 marks) For all teams, draw a meaningfully labelled histogram of the average of the two individual slope coefficient estimates (over the two replicates).

Show your code.

Use xlim = c(-1, 1), col = "lightgrey" in hist() and an appropriate main title and xlab.

Add a vertical red dashed line at the average of the slope estimates.

Add a vertical blue dashed line at the true value of β .

Print the average and standard deviation of the slope estimates.

5 Conclusion

- e. (3 marks) What do you conclude about the quality of team slope estimates from the randomized study?
- f. (2 marks) What do you conclude about the value of having each team average their replicates from the randomized study?
- g. (2 marks) What effect, if any, has been produced by a lurking variable? Explain.