

## BIA 654 Homework 9

Recall the case study on “*Direct Mail Sales*,” using Plackett-Burman (PB) design; see recent lecture slides numbered 11–18 and the published paper “*Experimental design on the front lines of marketing: Testing new ideas to increase direct mail sales*” uploaded on Canvas.

- (a) Consider the Plackett-Burman design in Table shown on slides 14 and 15. Check that the design is indeed orthogonal for the main effects. That is, pick any two factors, say A and F, and check that the design includes 5 runs at each of the four factor-level combinations. (Just check this for yourself and no need to put your answer here.)
  - (b) Reanalyze the data shown on slide 15 by the Plackett-Burman design; Excel data file is uploaded. More specifically,
    - i. Produce the normal probability plot of 19 main effects and see what factors seem to be significant.
    - ii. Produce also the Pareto plot as shown on slide number 16. Using JMP: DOE-Effect Screening will do the work. It will also produce all effect estimates that you will need in part iii below.
    - iii. Screen out the 19 factors by examining their respective 95% confidence intervals.
- Here, I provide some details for part iii; see also page 314 in the uploaded paper. Notice that the response variable is a *proportion* (in percent!) of customers who have responded (accepted the offer) to a particular mail sale.
  - Note that there are  $N = 20$  test cells (i.e., versions of mails), and each of the 20 test cells was sent to  $n = 5,000$  people, resulting in the response rates listed in the last column. Therefore, the total  $nN = 100,000$  number of people have received mails in this study.
  - For each effect,  $nN/2 = 50,000$  people will see the  $+$  level, and another  $nN/2 = 50,000$  people will see the  $-$  level. Then the “effect” is the difference,

$$p_+ - p_-$$

where  $p_+$  is the proportion of responses resulting in orders among the  $nN/2$  subjects exposed to the  $+$  level, and  $p_-$  is the proportion of responses resulting in order among the  $nN/2$  subjects exposed to the  $-$  level. That is,  $p_+ = (\text{total number of orders in the } + \text{ level group})/50,000$ .

- Now, what is the standard error of effect?

$$\begin{aligned} \text{standard error (effect)} &= \sqrt{\text{estimate of } \text{Var}(p_+ - p_-)} \\ &= \sqrt{\text{estimate of } \text{Var}(p_+) + \text{estimate of } \text{Var}(p_-)}. \end{aligned}$$

Let  $\pi_+$  be the *true* (unknown) proportion of successes for customers exposed to the  $+$  level. Then,

$$\text{Var}(p_+) = \text{Var}\left(\frac{\text{Binomial}(nN/2, \pi_+)}{nN/2}\right) = \frac{(nN/2)\pi_+(1 - \pi_+)}{(nN/2)^2} = \frac{\pi_+(1 - \pi_+)}{nN/2}.$$

- Finally, from the above and under the null hypothesis  $H_0 : \pi_+ = \pi_-$  (saying there’s no effect),

$$\text{standard error (effect)} = \sqrt{\frac{\bar{p}(1 - \bar{p})}{nN/2} + \frac{\bar{p}(1 - \bar{p})}{nN/2}},$$

where  $\bar{p}$  is the *overall* success proportion, pooled over all runs; it is an estimate of  $\pi_+ (= \pi_-)$ . That is, in the case study,  $\bar{p} = 1.298\%$  by noting there are 1,298 people placed an order (positive response) out of  $nN = 100,000$  people.

- You can use  $z_{0.05/2} = 1.96$  in this case due to large sample size (instead of  $t$ -critical values). Recall the confidence intervals formed this way can be used to *reject* the null  $H_0 : \pi_+ = \pi_-$  if the interval does *not* contain zero.
- **Caution:** When you construct confidence intervals, be careful about the unit, e.g., if the response is in percent %, then standard error should also be in percent unit.