## Math 4720: Statistical Methods

## $12^{th}$ Week Summary (04/10/25)

- Multiple Comparisons .....
- The problem of multiplicity is serious when we are testing many hypotheses
- Instead of using Type-I error rate, we use familywise error rate (FEW):

 $\alpha_F = P(\text{Falsely reject at least one hypotheses})$ 

• Bonferroni Method:

If there are 
$$m$$
 hypotheses, then Reject  $H_0^{ij}: \mu_i = \mu_j$  if  $|\bar{y}_i - \bar{y}_j| > t_{\alpha/2m} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$ 

**Pros:** This will guarantee that  $\alpha_F \leq 0.05$ 

Cons: The chance of rejecting  $H_0$  is small. In other word, the power of Bonferroni is very poor

R: with(exmp9.3, pairwise.t.test(yield, agent, p.adj = "bonf"))

• Fisher's Least Significant Difference (LSD):

Reject 
$$H_0^{ij}: \mu_i = \mu_j$$
 if  $|\bar{y}_i - \bar{y}_j| > t_{\alpha/2} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$ 

**Pros:** Easy to discover differences (reject  $H_0^{ij}$ ). In other word, high power

Cons: This DOES NOT control the familywise error

R: with(exmp9.3, pairwise.t.test(yield, agent, p.adj = "none"))

• Tukey's Method:

Reject 
$$H_0^{ij}: \mu_i = \mu_j$$
 if  $|\bar{y}_i - \bar{y}_j| > \frac{q_\alpha(t, df_E)}{\sqrt{2}} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$ 

**Pros:** Control the familywise error rate

Cons: Lower power comparing to Fisher LSD and Dunnett methods

R: model = aov(yield agent, data = exmp9.3); TukeyHSD(model)

• Dunnett's Method:

Reject 
$$H_0^{ij}: \mu_i = \mu_j \text{ if } |\bar{y}_i - \bar{y}_j| > d_{\alpha}(t - 1, df_E) \sqrt{\frac{2MSE}{n}}$$

Pros: Control the familywise error rate. Higher power comparing to Bonferroni and Tukey's

**Cons:** It's just for comparing with a control group.

R: library("DescTools"); with(exmp9.3, DunnettTest(yield, agent, control = "None"))

- Population proportion .....
- Draw a large random sample of size n from a large population having unknown proportion p of successes. To test the hypothesis  $H_0: \pi = \pi_0$ , compute the Z statistic:  $z = \frac{\hat{\pi} \pi_0}{\sqrt{\pi_0(1 \pi_0)}}$
- Use this test when the sample size n is so large that both  $n\pi_0$  and  $n(1-\pi_0)$  are 5 or more.
- The p-value for a test of  $H_0: \pi = \pi_0$  against

$$H_a: \pi > \pi_0$$
 is p-value  $= P(Z \ge z)$ 

$$H_a: \pi < \pi_0 \text{ is p-value} = P(Z \leq z)$$

 $H_a: \pi \neq \pi_0$  is p-value =  $2P(|Z| \geq |z|)$ 

• Confidence interval for  $\pi$  can be obtained by  $\hat{\pi} \pm z_{\alpha/2v} \sqrt{\frac{\hat{\pi}(1-\hat{\pi})}{n}}$ . This interval should not be used unless  $n\hat{\pi} \geq 5$  and  $n(1-\hat{\pi}) \geq 5$ .