Bihan Q

Covariate, Pairwise, Anova, Tied Obs

Q1(a)

Consider the ChickWeight data in R. The body weights of the chicks were measured at birth (i.e., time=0) and every second day thereafter until day 20. They were also measured on day 21. There were four groups of chicks on different protein diets.

1.Determine whether there is a significant difference in the mean weights of the four groups on Day 18:

a)Without adjusting for Birth Weight

```
In [46]: # packages
import pandas as pd

In [47]: # get data
import statsmodels.api as sm
ChickWeight = sm.datasets.get_rdataset('ChickWeight','datasets')
chick = pd.DataFrame(ChickWeight.data)
print(chick.shape)
chick.head(40)

(578, 4)
```

	weight	Time	Chick	Diet
0	42	0	1	1
1	51	2	1	1
2	59	4	1	1
3	64	6	1	1
4	76	8	1	1
5	93	10	1	1
6	106	12	1	1
7	125	14	1	1
8	149	16	1	1
9	171	18	1	1
10	199	20	1	1
11	205	21	1	1
12	40	0	2	1
13	49	2	2	1
14	58	4	2	1
15	72	6	2	1
16	84	8	2	1
17	103	10	2	1
18	122	12	2	1
19	138	14	2	1
20	162	16	2	1
21	187	18	2	1
22	209	20	2	1
23	215	21	2	1
24	43	0	3	1
25	39	2	3	1
26	55	4	3	1
27	67	6	3	1
28	84	8	3	1
29	99	10	3	1
30	115	12	3	1
31	138	14	3	1
32	163	16	3	1
33	187	18	3	1

Out[47]:

	weight	Time	Chick	Diet
34	198	20	3	1
35	202	21	3	1
36	42	0	4	1
37	49	2	4	1
38	56	4	4	1
39	67	6	4	1

```
In [48]: chick['Diet'].value_counts()
```

220 Out[48]:

120 3 120

118

Name: Diet, dtype: int64

In [49]: chick.describe()

Ωı	1+ [710	7 .
υı	4 C [+2] .

	weight	Time	Chick	Diet
count	578.000000	578.000000	578.000000	578.000000
mean	121.818339	10.717993	25.750865	2.235294
std	71.071960	6.758400	14.568795	1.162678
min	35.000000	0.000000	1.000000	1.000000
25%	63.000000	4.000000	13.000000	1.000000
50%	103.000000	10.000000	26.000000	2.000000
75%	163.750000	16.000000	38.000000	3.000000
max	373.000000	21.000000	50.000000	4.000000

```
chick_18 = chick[chick['Time'] == 18]
In [50]:
         print(chick_18.shape)
         chick_18.head(3)
```

(47, 4)

Out[50]:

	weight	Time	Chick	Diet
9	171	18	1	1
21	187	18	2	1
33	187	18	3	1

```
In [51]: # with fixed time, weight is Y, diets are groups and independent
         from statsmodels.formula.api import ols
         model = ols('weight ~ Diet', data=chick_18).fit()
         anova_table = sm.stats.anova_lm(model)
```

Print the ANOVA table

```
print(anova_table)
          One-Way ANOVA Table:
                       df
                                                                  F
                                                                        PR(>F)
                                   sum_sq
                                                 mean sq
          Diet
                      1.0
                            23263.992032 23263.992032 8.161704 0.006455
                                           2850.384101
          Residual 45.0 128267.284564
                                                                NaN
                                                                           NaN
          #check non-normality
In [52]:
          import matplotlib.pyplot as plt
          # hist
          model = ols('weight ~ Diet', data=chick_18).fit()
          residuals = model.resid
          plt.figure(figsize=(12, 6))
          plt.subplot(1, 2, 1)
          plt.hist(residuals, bins=20, density=True, color='b', alpha=0.7)
          plt.title('Histogram of Residuals')
          plt.xlabel('Residuals')
          plt.ylabel('Density')
          # 99
          plt.subplot(1, 2, 2)
          sm.qqplot(residuals, line='s', ax=plt.gca())
          plt.title('QQ Plot of Residuals')
          plt.tight_layout()
          plt.show()
          # Bartlett's test
          import pingouin as pg
          bartlett test result = pg.homoscedasticity(chick 18, dv='weight', group='Diet')
          print("Bartlett's Test for Homogeneity of Variances:")
          print(bartlett_test_result)
                            Histogram of Residuals
                                                                            QQ Plot of Residuals
           0.010
                                                           100
           0.008
                                                            50
                                                         Sample Quantiles
           0.006
                                                             0
           0.004
                                                           -50
           0.002
                                                           -100
           0.000
                  -100
                          -50
                                         50
                                                                   -1.5
                                                                        -1.0
                                                                             -0.5
                                 Residuals
                                                                             Theoretical Quantiles
          Bartlett's Test for Homogeneity of Variances:
                                  pval equal_var
                          W
          levene 1.058972 0.376444
                                              True
         # shapiro
In [53]:
          from scipy.stats import shapiro
```

print("One-Way ANOVA Table:")

```
model = ols('weight ~ Diet', data=chick_18).fit()
         residuals model with covariate = model.resid
         shapiro_test_statistic, shapiro_test_pvalue = shapiro(residuals_model_with_covariate)
         print("Shapiro-Wilk Test Statistic:", shapiro_test_statistic)
         print("p-value:", shapiro test pvalue)
         Shapiro-Wilk Test Statistic: 0.980424702167511
         p-value: 0.6098649501800537
In [54]:
        # Bartlett's test
         bartlett_test_result = pg.homoscedasticity(chick_18, dv='weight', group='Diet')
         print("Bartlett's Test for Homogeneity of Variances:")
         print(bartlett_test_result)
         Bartlett's Test for Homogeneity of Variances:
                             pval equal_var
                       W
         levene 1.058972 0.376444 True
```

Answer:

1. Under satisfied normality and equal variance assumption, 26.098205 as F-stats and 4.418863e-07 as p-value indicate a significant difference in the mean weights of the four groups on Day 18.

Q1(b)

b)Adjusting for Birth Weight. Give the LS Means (i.e., adjusted for Birth Weight).

```
In [55]: # get inital weight
         birth weight = chick[chick['Time'] == 0][['Chick', 'weight']]
         birth_weight.rename(columns={'weight': 'weight_inital'}, inplace=True)
         chick ajusted = pd.merge(chick, birth weight, on='Chick', how='left')
         chick_ajusted_18 = chick_ajusted[chick_ajusted['Time'] == 18]
         print(chick_ajusted_18.shape[0]==chick_18.shape[0])
         chick ajusted 18.head(3)
         chick_ajusted_18['weight_inital'].value_counts()
         True
               18
         41
Out[55]:
         42
               14
         39
                6
         40
                5
         43
         Name: weight inital, dtype: int64
In [56]: chick_ajusted_18['Diet'].value_counts()
              17
Out[56]:
         2
              10
         3
              10
         4
              10
         Name: Diet, dtype: int64
```

```
In [57]: # two-anova (python)
         model2 = ols('weight ~ Diet+weight_inital', data=chick_ajusted_18).fit()
         anova table2 = sm.stats.anova lm(model2)
         print(anova table2)
                         df
                                                                   PR(>F)
                                   sum_sq
                                               mean_sq
         Diet
                       1.0 23263.992032 23263.992032 8.573318 0.005384
         weight inital 1.0 8871.784187 8871.784187 3.269457 0.077419
         Residual
                      44.0 119395.500377 2713.534099
                                                             NaN
                                                                      NaN
In [58]: # anova (r)
         from IPython.display import Image
         image file path = f"C:/Users/11139/Desktop/STAT5391/anova2.png"
         Image(filename=image_file_path)
           300 # anova
Out[58]:
           301 library(car)
           302 m <- aov(data=chick_adjusted_18, weight ~ Diet+weight_initial)
           303 summary(m)
           304 -
                                Df Sum Sq Mean Sq F value Pr(>F)
                                3 36690 12230 4.729 0.00623 **
                 Diet
                 weight_initial 1 6229
                                             6229
                                                    2.409 0.12818
                 Residuals 42 108612
                                             2586
                 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
In [59]:
         # pairwise test
         import pingouin as pg
         from statsmodels.stats.anova import AnovaRM
         # Ls means = pq.pairwise ttests(data=chick ajusted 18, dv='weight', between='Diet', co
         from IPython.display import Image
         image_file_path = f"C:/Users/11139/Desktop/STAT5391/pairwise_test.png"
         Image(filename=image_file_path)
```

```
292 library(stats)
Out[59]:
          293 library(stats)
          294 pairwise_tests <- pairwise.t.test(</pre>
          295  x = chick_adjusted_18$weight,
          g = chick_adjusted_18$Diet,
p.adjust.method = "fdr",
          298
                 covariates = chick_adjusted_18$weight_initial
          299 )
          300
          301 # Print the results
          302 print(pairwise_tests)
          303
          304 -
                        Pairwise comparisons using t tests with pooled SD
                data: chick_adjusted_18$weight and chick_adjusted_18$Diet
                  1
                         2
                                3
                2 0.2379 -
                3 0.0049 0.1119 -
                4 0.1119 0.5142 0.2379
                P value adjustment method: fdr
In [60]: # LSM
        image_file_path = f"C:/Users/11139/Desktop/STAT5391/LSM.png"
        Image(filename=image file path)
Out[60]:
           314 model <- lm(weight ~Diet+weight_initial, chick_adjusted_18)
           315 anova_result <- anova(model)</pre>
           316 | lsmeans_result <- lsmeans(model, "Diet", cov.reduce=TRUE)
           317 lsmeans_result
           318
          320:1 C Chunk 8 $
          Console Background Jobs ×
          R 4.2.1 · ~/ ≈
                   202 16.1 42
                                     169
                                               234
         Confidence level used: 0.95
         > lsmeans_result <- lsmeans(model, "Diet", cov.reduce=TRUE)</pre>
         > lsmeans_result
          Diet 1smean SE df lower.CL upper.CL
                   164 12.8 42
                                    138
                                               190
          2
                  183 16.3 42
                                     150
                                               216
          3
                  230 16.2 42
                                     197
                                               262
                   202 16.1 42
                                    169
                                               234
         Confidence level used: 0.95
         > |
```

Answer:

Two-way Anova:

- 1. Compared to before adjusted, after we add inital_birth_weight as second variable, diet with p-value 0.00623 indicates that there is a statistically difference in the mean weights among diet groups on Day18.
- 2. From p-value = 0.12818, different inital weight groups does not have a significant difference on mean weight on Day18.
- 3. Python and R gives little different numbers, but similar in result significance.

Pairwise:

1. Only p-value for the comparison between **Group 1 and Group 3** is 0.0049. **There is statistically significant difference in the mean weights between Group 1 and Group 3 after adjusting for the covariate 'weight_inital' at Day18.** For other groups, there are no significant difference.

LSM:

1. Compared to original weights for four groups 1-4 (158.94,187.70,233.10,202.90), we get new adjusted weight for group 1-4 is 164,183,230,and 202. **Four groups of diet are different.**

1(c)

Image(filename=image file path)

c)Check the validity of your assumptions, including parallelism. Suggest measures that you would take if the assumptions are not satisfied.

```
In [61]: # Shapiro-Wilk
          # from scipy.stats import shapiro
          # shapiro_test_statistic, shapiro_test_pvalue = shapiro(residuals_model_with_covariate
         image file path = f"C:/Users/11139/Desktop/STAT5391/shapiro.png"
         Image(filename=image file path)
           300 # anova
Out[61]:
           301 library(car)
           302 m <- aov(data=chick_adjusted_18, weight ~ Diet+weight_initial)</pre>
           303 summary(m)
           304
           305 shapiro.test(resid(aov(data=chick_adjusted_18, weight ~ Diet+weight_initial)))
           306 -
                         Shapiro-Wilk normality test
                 data: resid(aov(data = chick_adjusted_18, weight ~ Diet + weight_initial))
                 W = 0.98613, p-value = 0.8441
In [62]:
         # Levene's and Box's test
         image_file_path = f"C:/Users/11139/Desktop/STAT5391/Box.png"
```

319 leveneTest(weight ~Diet, chick_adjusted_18) Out[62]: 320 fligner.test(weight ~Diet, chick_adjusted_18) #alternative to Box, robust to normality 321 323:1 C Chunk 8 \$ Console Background Jobs × R 4.2.1 · ~/ ≈ > leveneTest(weight ~Diet, chick_adjusted_18) Levene's Test for Homogeneity of Variance (center = median) Df F value Pr(>F) group 3 1.059 0.3764 > fligner.test(weight ~Diet, chick_adjusted_18) #alternative to Box, robust to normality Fligner-Killeen test of homogeneity of variances data: weight by Diet Fligner-Killeen:med chi-squared = 2.6903, df = 3, p-value = 0.4419 # Two-way ANOVA with interaction In [63]: two_way_anova_result = sm.stats.anova_lm(ols('weight ~ Diet * weight_inital' , data=ch print("Two-Way ANOVA with Interaction:") print(two_way_anova_result) Two-Way ANOVA with Interaction: df sum_sq F PR(>F) mean_sq Diet 1.0 23263.992032 23263.992032 8.393622 0.005901 8871.784187 8871.784187 3.200930 0.080642 weight inital 1.0 Diet:weight_inital 1.0 215.522671 215.522671 0.077760 0.781693 Residual 43.0 119179.977706 2771.627389 NaN NaN

Answer:

Shapiro-Wilk:

• Shapiro-Wilk p-value (0.8441) shows normality assumption is **satisfied**.

Bartlett's test:

Assumption of constant variance is satisfied.

Interaction(parallelism):

• The p-value for the interaction term (0.781693), suggesting **no interaction** effect between diet and weight_inital.

2(a)

2.For 1a), perform pairwise comparisons among the 4 groups using each of the following, and comment on the results a)Bonferroni method

```
In [64]: from IPython.display import Image
   image_file_path = f"C:/Users/11139/Desktop/STAT5391/Bonferroni.png"
   Image(filename=image_file_path)
```

```
336 # bonferroni
337 model <- aov(weight ~ Diet, data = chick_18)</pre>
Out[64]:
              338 pairwise_results <- pairwise.t.test(chick_18$weight, chick_18$Diet, p.adjust.method =
                   "bonferroni")
              339 print(pairwise_results)
              340
              341
             342:1 Chunk 9 $
            Console Background Jobs ×
            R 4.2.1 · ~/ ≈
            > print(pairwise_results)
                     Pairwise comparisons using t tests with pooled SD
            data: chick_18$weight and chick_18$Diet
              1
                      2
            2 1.0000 -
            3 0.0049 0.3358 -
            4 0.2313 1.0000 1.0000
            P value adjustment method: bonferroni
```

Answer:

- There is significant difference in the mean weights of the diet1 and diet3 on Day 18.
- All other groups are not significantly different from others on Day18..

2(b)

2.For 1a), perform pairwise comparisons among the 4 groups using each of the following, and comment on the results b)Tukey method

```
In [65]: from statsmodels.stats.multicomp import pairwise_tukeyhsd

post_hoc = pairwise_tukeyhsd(chick_18['weight'], chick_18['Diet'], alpha=0.05)
#table
print("ANOVA Table:")
print(anova_table)
print("\nPairwise Comparisons (Bonferroni Correction):")
print(post_hoc)
```

ANOVA Table: df sum_sq mean_sq F PR(>F) 1.0 23263.992032 23263.992032 8.161704 0.006455 Diet Residual 45.0 128267.284564 2850.384101 NaN NaN Pairwise Comparisons (Bonferroni Correction): Multiple Comparison of Means - Tukey HSD, FWER=0.05 _____ group1 group2 meandiff p-adj lower upper reject -----2 28.7588 0.5085 -26.2809 83.7986 False 1 3 74.1588 0.0044 19.1191 129.1986 True 1 4 43.9588 0.1586 -11.0809 98.9986 False 2 3 45.4 0.2173 -16.3638 107.1638 False 4 15.2 0.9123 -46.5638 76.9638 False 3 4 -30.2 0.5638 -91.9638 31.5638 False

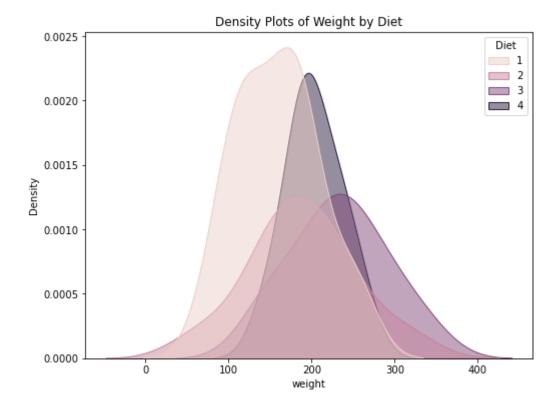
Answer:

• Same as Bonferroni correction, by Tukey, there is significant difference in the mean weights of the diet1 and diet3 on Day 18.

Q3

Repeat 1a) using the Kurskal-Wallis test

```
187
                 2
Out[66]:
          163
                 2
          234
                 2
                 2
          146
          184
                 2
                 2
          185
                 1
          171
          227
                 1
          204
                 1
          263
                 1
          294
                 1
          332
                 1
          232
                 1
          157
                 1
          230
                 1
          214
                 1
                 1
          262
          198
                 1
          174
                 1
          210
                 1
          261
                 1
          151
                 1
          231
                 1
          207
                 1
          205
                 1
          154
                 1
          199
                 1
          160
                 1
          250
                 1
          134
                 1
          100
                 1
          112
                 1
          81
                 1
          248
                 1
          123
                 1
          120
                 1
          107
                 1
          307
                 1
          148
                 1
          72
                 1
          203
                 1
          Name: weight, dtype: int64
          # density plot for each group
In [67]:
          import matplotlib.pyplot as plt
          import seaborn as sns
          plt.figure(figsize=(8, 6))
          sns.kdeplot(data=chick_18, x='weight', hue='Diet', fill=True, alpha=0.5)
          plt.title('Density Plots of Weight by Diet')
          plt.show()
```



```
import pandas as pd
from scipy.stats import kruskal
result = kruskal(*[group["weight"] for name, group in chick_18.groupby("Diet")])
print(result)
```

KruskalResult(statistic=10.623444357500071, pvalue=0.01394646461394125)

Answer:

Discussion for assumption of Kruskal:

- Total sample size larger than 5.
- We test with numerical data weight.
- Deal with tied values, correction automatically applied.
- The distributions are similar across groups from density plot.

Result:

• The result of Kruskal test suggests that there is a significant difference in the mean weights of the four groups on Day 18.