Planned comparisons using contrasts in R

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1 Planned comparisons using contrasts in R

```
[1]: # Load needed packages
     library(tidyverse)
     library(ggplot2)
      Attaching packages
                                                 tidyverse
    1.3.0
                                   0.3.4
      ggplot2 3.3.0
                          purrr
      tibble 3.0.1
                          dplyr
                                  0.8.5
      tidyr
              1.0.2
                          stringr 1.4.0
      readr
              1.3.1
                          forcats 0.5.0
      Conflicts
    tidyverse_conflicts()
      dplyr::filter() masks stats::filter()
      dplyr::lag()
                      masks stats::lag()
```

This module uses data simulated to match means and SDs of data analyzed in A. Parenti, L. Guerrini, P. Masella, S. Spinelli, L. Calamai, P. Spugnoli (2014). "Comparison of Espresso Coffee Brewing Techniques," *Journal of Food Engineering*, Vol. 121, pp. 112-117.

The goal of this experiment in Parenti et al. is to "evaluate and compare the differences in terms of quality between espresso coffee made using three different extraction procedures." Quality, in this exercise, is measured in terms of the "foam index". Foam, according to the authors of this study, is a "distinctive feature of espresso coffee, as it is absent in other coffee brews and is required for consumer acceptance."

More compactly, here are the variables in our experimental data set:

- 1. foam Indx: The foam index is defined as the ratio between the foam and liquid volume (vol vol $^{-1}\%$) measured 30 seconds after extraction
- 2. method: Three methods were studied: a. Method 1=Bar Machine(BM); b. Method 2=Hyper-Espresso Method (HIP); and c. Method 3= I-Espresso System (IT).

Suppose that, in our espresso study, we pre-planned to study the hypothesis test:

$$H_0: \mu_2 = \frac{1}{2}(\mu_1 + \mu_3)$$

 $H_1: \mu_2 > \frac{1}{2}(\mu_1 + \mu_3)$

```
[2]: # Load the data
    esp = read.csv("espresso1.txt", sep="\t")
    esp$method = as_factor(esp$method)
     esp$method = recode(esp$method, "1" = "Bar Machine", "2" = "Hyper-Espresso_
     →Method", "3" = "I-Espresso System")
    summary(esp)
    lmod = lm(foamIndx ~ method, data = esp)
    summary(lmod)
        foamIndx
                                       method
     Min.
           :21.02
                     Bar Machine
                                          :9
     1st Qu.:35.66
                     Hyper-Espresso Method:9
     Median :38.52
                     I-Espresso System
     Mean :44.47
     3rd Qu.:55.23
     Max. :73.19
    Call:
    lm(formula = foamIndx ~ method, data = esp)
    Residuals:
       Min
               1Q Median
                             ЗQ
                                   Max
    -14.62 -6.60
                   0.41 5.73 16.49
    Coefficients:
                                Estimate Std. Error t value Pr(>|t|)
    (Intercept)
                                  32.400
                                              2.819 11.492 3.04e-11 ***
    methodHyper-Espresso Method
                                  28.900
                                              3.987
                                                     7.248 1.73e-07 ***
    methodI-Espresso System
                                  7.300
                                              3.987
                                                     1.831
                                                             0.0796 .
    Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
    Residual standard error: 8.458 on 24 degrees of freedom
    Multiple R-squared: 0.7031, Adjusted R-squared: 0.6783
```

F-statistic: 28.41 on 2 and 24 DF, p-value: 4.699e-07

Note that:

- 1. The test was pre-planned.
- 2. The full F-test is significant (p-value: 4.7×10^{-7})
- 3. & 4. There is only one contrast, so we've met the conditions of orthogonality.

Let's first "hard code" our contrast hypothesis test. Then, we'll use an R package to perform the test automatically.

```
[5]: c = c(-0.5, 1, -0.5) #constants in the contrast specified above
     b = coef(lmod); #ANOVA regression model coefficients
     n = length(resid(lmod)) #total number of espressos brewed
     n_method = with(esp, c(length(foamIndx[method == "Bar Machine"]),
                 length(foamIndx[method == "Hyper-Espresso Method"]),
                 length(foamIndx[method == "I-Espresso System"])))
                 #vector with number of espressos brewed by each method
     J = length(unique(esp$method)); J
     ybar = c(b[1], b[1] + b[2], b[1] + b[3]); #vector of sample means of foam index
     → for each method
     ybar[2] - 0.5*(ybar[1] + ybar[3])
     rss = sum(resid(lmod)^2); #residual sum of squares
     sighat = sqrt(rss/(n-J)); #estimate of sigma^2 hat
     gammahat = ybar[2] - 0.5*(ybar[1] + ybar[3]);
     cat("The estimate of the contrast is is", as.numeric(gammahat))
     se = (sqrt(sighat^2*(1/(4*n method[1]) + 1/(n method[2]) + 1/(4*n method[3])))_{11}
     →#standard error of gamma hat
     z = gammahat/se # test statistic
     pval = 1-pnorm(z) #p-value for upper tailed test
     cat(". The test statistic is ", as.numeric(z), ". The p value for the test is",_{\sqcup}
      →pval,".")
```

3

(Intercept): 25.25

The estimate of the contrast is is 25.25. The test statistic is 7.312531. The p value for the test is 1.311173e-13.

Luckily, R has a function that will compute contrasts for us. This isn't a substitution for understanding contrasts, but it is helpful in saving time! The glht() function will allow us to conduct "general linear hypotheses and multiple comparisons for parametric models, including [ANOVA] generalized linear models, linear mixed effects models, and survival models."

```
[6]: #install.packages("multcomp") #if we need to install the multcomp package library(multcomp) #load the multicomp package
```

```
contrast = glht(lmod, linfct = mcp(method = c(-0.5, 1, -0.5))) #glht = 

→ generalized linear hypothesis test
summary(contrast)
```

Simultaneous Tests for General Linear Hypotheses

```
Multiple Comparisons of Means: User-defined Contrasts
```

[]: