Chapter 9: Analysis of Variance (ANOVA)

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Recall that the two-sample t-test is used to compare the means of two populations or two experimental groups. In many experiments there are more than two conditions or treatments to compare in which case the two sample *t*-test will not suffice. This chapter introduces *analysis of variance* (ANOVA) and the associated test of equality of more than two means.

Terminology

Example: An experiment is carried out to see how genotype and the amount of phosphorus affect plant growth. The biomass of the plant is the measured response. Three genotypes and two phosphorus levels (high and low) were used in the experiment. There were 10 plants grown at each experimental combination.

▶ This is an example of a 3×2 factorial experiment.

Terminology

- ▶ **Response variable:** The response (or dependent) variable in this example is the biomass of the plant.
- ► **Factors:** The independent variables which *may* influence the response variable (genotype and phosphorus in this example).
- ▶ **Levels:** The individual levels of each factor. For example, genotype has three levels and phosphorus has two.
- ► **Treatment:** A combination of factor levels. In this example, there are six possible treatments:
 - genotype 1 and low phosphorus;
 - genotype 2 and low phosphorus;
 - genotype 3 and low phosphorus;
 - genotype 1 and high phosphorus;
 - genotype 2 and high phosphorus;
 - genotype 3 and high phosphorus.

Terminology

- Replicates: The number of experimental units (plants in this example) per treatment. In this example, there are a total of ten replicates because there are ten plants grown at each treatment combination.
- ▶ **Balanced design:** This occurs when each treatment combination uses the same number of replicates.

Single factor ANOVA

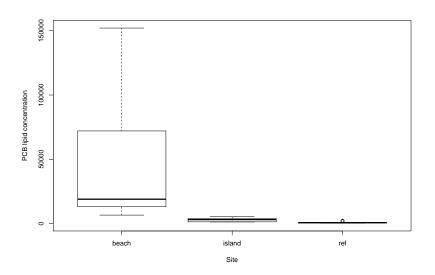
- An experiment with a single factor is called a single factor ANOVA or a one-way ANOVA.
- ► The two-sample t-test was used to compare the means of two populations (think og this as a one-way ANOVA with only two factor levels).
- ▶ In a one-way ANOVA, the classical statistical inference is to test if the means across the different levels of the factor are equal or not:

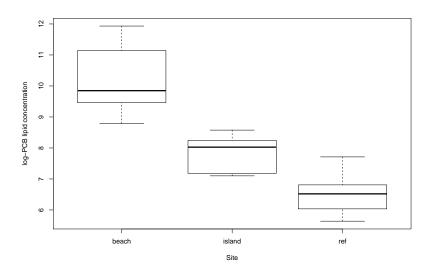
$$H_0: \mu_1 = \mu_2 = \dots \mu_k$$

Because we are typically dealing with factors with three or more levels, the two-sample t-test will not suffice.

A study was done on the liver PCB concentration (lipid) in Black Guillemot birds in Canada at three different sites: a reference site (ref), a nearby island (island), and the beach (beach). A boxplot of the PCB lipid concentration for the birds at the three sites shows strong positive skewness (i.e., right skewed data) which is quite common for data on concentrations of toxins. Because the data is strongly skewed, a log transformation was applied. The goal of the study is to compare the mean log-PCB lipid concentration in the birds at the three sites.

Note that these data are observational, not experimental. That is, the data were collected by observing the birds at the three sites. In an experiment, the experimenter would assign units at random to the different treatments.





Let $\mu_{\textit{beach}},~\mu_{\textit{island}},$ and $\mu_{\textit{ref}}$ be the mean log-PCB lipid concentration of Black Guillemot birds at the three site. We are interested in testing

$$H_0: \mu_{beach} = \mu_{island} = \mu_{ref}$$

The alternative (or research) hypothesis is that the means are not all equal (but two of them could be).

To set up the one-way ANOVA hypothesis test procedure, we need to partition the overall variability in the response variable into two components: a component due to the variability **within** each factor level and the variability **between** the factor level means:

$$SS(Total) = SS(Between) + SS(Within)$$

Let y_{ij} denote the j-th observation at level i and \bar{y}_i . be the sample mean of the response at level i. Then, we define the following:

- ▶ Total sum of squares: $SS(Total) = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (y_{ij} \bar{y})^2$.
- ▶ Between sum of squares: $SS(Between) = \sum_{i=1}^{k} n_i (\bar{y}_{i\cdot} - \bar{y})^2$.
- ► Within sum of squares: $SS(Within) = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2.$

- ▶ If the factor level means are all equal, then the within group variance and between group variance should be equal.
- ► The F-test of the one-way ANOVA model is based on a comparison of these two variances:

$$F = \frac{MS (Between)}{MS (Within)} = \frac{SS (Between) / (k - 1)}{SS (Within) / (n - k)}$$

▶ Where $n = n_1 + n_2 + \cdots + n_k$ is the total sample size.

▶ If the null hypothesis (i.e., equality of means) is true, then the *F* statistic

$$F = \frac{MS (Between)}{MS (Within)}$$

has an F distribution with k-1 and n-k degrees of freedom, denoted $F_{k-1,n-k}$.

- ▶ Correspondingly, we will reject the null hypothesis whenever F is greater than the $1-\alpha$ percentile from an F distribution with k-1 and n-k degrees of freedom.
- The p-value for the test is given by the area to the right of F under an F_{k-1,n-k} distribution: Pr {F_{k-1,n-k} > F}.

To test the null hypothesis

$$H_0$$
: $\mu_{beach} = \mu_{island} = \mu_{ref}$

in R, we could use the aov function:

```
## Call:
## aov(formula = log(lipid) ~ site)
##
## Terms:
## site Residuals
## Sum of Squares 78.67164 19.06337
## Deg. of Freedom 2 29
##
## Residual standard error: 0.8107758
## Estimated effects may be unbalanced
```

Just like in regression, we can use the residuals to construct model diagnostic plots:

