Section 8.1: Analysis of Variance

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Outline

- ▶ Overview of Analysis of Variance
- ▶ The Basic ANOVA
- ▶ The ANOVA Model
- ▶ Checking Conditions

The Basic Idea of ANOVA

When we ask if a set of sample means gives evidence for differences among the population means, what matters is not how far apart the sample means are but how far apart they are relative to the variability of individual observations.
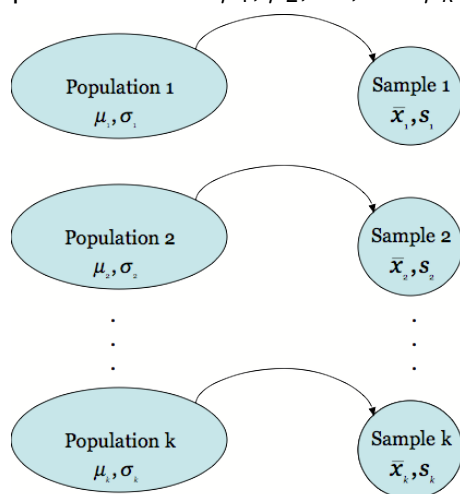
Baldi & Moore, 3rd ed., p.606

The Basic Idea of ANOVA

The basic idea is to compare measures of variability, both between the groups and within each group, as a way to assess how different the groups really are.

Lock5, p.540

Analysis of Variance Sampling Model

Draw simple random samples from k independent populations to compare population means μ_1, μ_2, ..., and μ_k:

Condition: $\sigma_1 = \sigma_2 = \cdots = \sigma_k = \sigma$

Conditions for Applying ANOVA

1. We have k independent SRSs, one from each of k populations.
2. Each of the k populations has a Normal distribution with an unknown mean μ_i.
3. All of the populations have the same standard deviation σ, whose value is unknown.

Case Study

Diet Restriction and Longevity
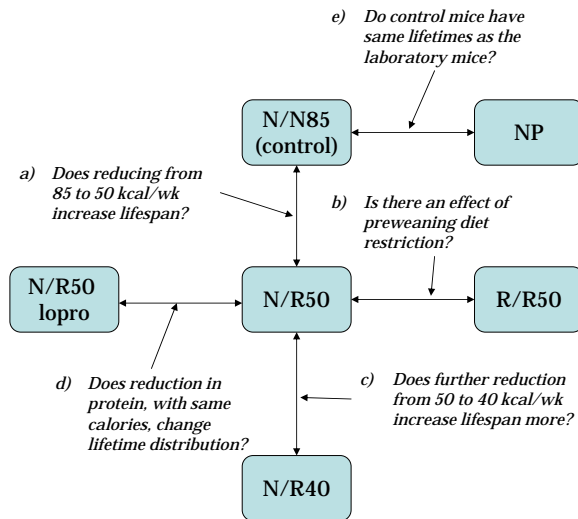
► Study of the effect of restricting caloric intake on life expectancy.
► Female mice randomly assigned to one of 6 treatment groups:
  1. NP: unlimited nonpurified standard diet for laboratory mice.
  2. N/N85 (control group): fed normally both before and after weaning. Caloric intake controlled at 85 kcal/wk after weaning.
  3. N/R50: normal diet before weaning and reduced-calorie diet of 50 kcal/wk after weaning.
  4. R/R50: reduced-calorie diet of 50 kcal/wk before and after weaning.
  5. N/R50 lopro: normal diet before weaning, restricted diet of 50 kcal/wk after weaning with dietary protein content decreased with advancing age.
  6. N/R40: normal diet before weaning, severely reduced-calorie diet of 50 kcal/wk after weaning.
► Several questions of interest to be addressed.

Case Study

Diet Restriction and Longevity: Study Citation

Weindruch R, Walford RL, Fligiel S, Guthrie D. The retardation of aging in mice by dietary restriction: longevity, cancer, immunity and lifetime energy intake. *J Nutr.* 1986;116(4):641-654. doi:10.1093/jn/116.4.641

Planned Comparisons Among Groups in the Diet Restriction Study

e) Do control mice have same lifetimes as the laboratory mice?

N/N85 (control)

NP

a) Does reducing from 85 to 50 kcal/wk increase lifespan?

b) Is there an effect of preweaning diet restriction?

N/R50 lopro

N/R50

R/R50

d) Does reduction in protein, with same calories, change lifetime distribution?

c) Does further reduction from 50 to 40 kcal/wk increase lifespan more?

N/R40

Diet Restriction Study
Exploratory Plots

Use R to create plots to explore the possibility of differences among the means.

The Big Picture of ANOVA

▶ Key question: Is variability across populations greater than variability within populations
  ▶ Variability between vs within
▶ Hypothesis test:
  ▶ $H_0 : \mu_1 = \mu_2 = \cdots = \mu_k$ versus
  ▶ H_a : At least one $\mu_i \neq \mu_j$
▶ Test statistic (F): ratio of the variability among group (or treatment) means over the variability within samples.

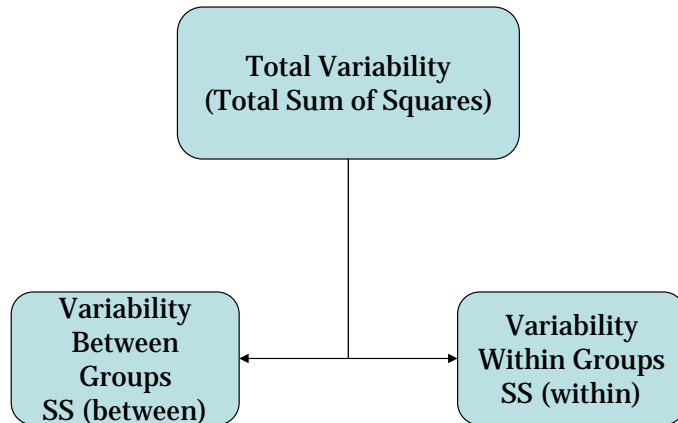$$F = \frac{\text{variation among the sample means}}{\text{variation among individuals in the same sample}}$$

  ▶ Large test statistic $\Rightarrow$ evidence against H_0
▶ Results are summarized in the ANOVA table.

Notation

Key notation used in calculations for comparing variability between groups and within groups:

$$
\begin{aligned}
k &= \text{number of groups} \\
n_i &= \text{sample size for group } i \\
\bar{x}_i &= \text{sample mean for group } i \\
s_i &= \text{standard deviation for group } i \\
n &= \sum_{i=1}^{k} n_i = \text{total sample size} \\
\bar{x} &= \frac{\sum_{i=1}^{k} n_i \bar{x}_i}{n} = \text{overall mean.}
\end{aligned}
$$

Apportioning Variability

Variability within Groups

Measure variability within groups by the sum of squared deviations from the group mean:

$$SSE = (n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \cdots + (n_k - 1)s_k^2$$

Degrees of Freedom

▶ The degrees of freedom within samples is the sum of degrees of freedom for each sample.

$$
\begin{aligned}
df(error) &= (n_1 - 1) + (n_2 - 1) + \cdots + (n_k - 1) \\
&= n - k
\end{aligned}
$$

▶ This is also equal to the total sample size minus the number of groups.

Mean Square within Groups (Error)

In ANOVA, the mean square measures variability as the ratio of sum of squares to degrees of freedom.

$$
\begin{aligned}
MSE = MS(error) &= \frac{SSE}{df(error)} \\
&= \frac{(n_1 - 1)s_1^2 + \cdot + (n_l - 1)s_k^2}{n - k}
\end{aligned}
$$

▶ The pooled standard deviation (s_p) estimates the common group-specific standard deviation:

$$s_p = \sqrt{MS(error)}$$

Variability between Groups

Measure variability between groups by SS(between), a weighted sum of squared deviations of group means from the grand mean:

$$SSG = n_1(\bar{x}_1 - \bar{x})^2 + n_2(\bar{x}_2 - \bar{x})^2 + \cdots + n_k(\bar{x}_k - \bar{x})^2$$

The corresponding degrees of freedom between groups is one less than the number of groups:

$$df(groups) = k - 1$$

Mean Square between Groups

The mean square between groups, MSG or MS (treatment) is the ratio of the sum of squares over the degrees of freedom:

$$
\begin{aligned}
MSG &= \frac{SSG}{df(groups)} \\
&= \frac{n_1(\bar{x}_1 - \bar{x})^2 + n_2(\bar{x}_2 - \bar{x})^2 + \cdots + n_k(\bar{x}_k - \bar{x})^2}{k - 1}
\end{aligned}
$$

The F Statistic

▶ The F statistic is the ratio of the mean square between to the mean square within.

$$F = \frac{MSG}{MSE}$$

▶ If the null hypothesis is true, F has an F distribution with $k - 1$ and $n - k$ degrees of freedom

▶ Use the StatKey F distribution web applet to find p-values when necessary or obtain them from R output for ANOVA using the **lm()** command we also used for regression.
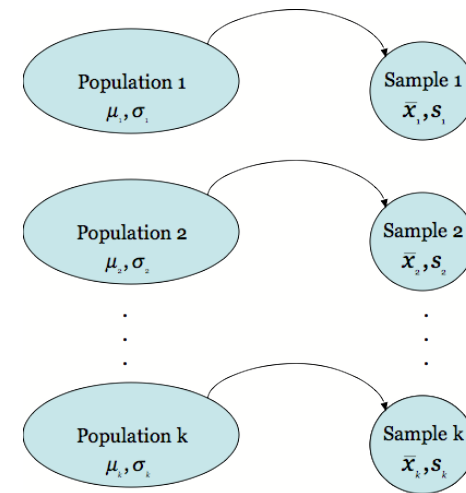
Diet Restriction Study

See R Commands for ANOVA for R help.

Checking Conditions

Analysis of Variance Sampling Model

Draw samples from k independent populations to compare population means μ_1, μ_2, ..., and μ_k:

Condition: $\sigma_1 = \sigma_2 = \cdots = \sigma_k = \sigma$

Standard Conditions for ANOVA

- ▶ Design conditions:
  - ▶ Random samples: reasonable to consider observations a random sample from respective populations
  - ▶ Independent samples: the k samples are independent of each other
- ▶ Population:
  - ▶ Normal: Population distributions are normal (not crucial if n_i are large and similar)
  - ▶ Equal standard deviations:

$$\sigma_1 = \sigma_2 = \cdots = \sigma_k = \sigma$$

Rules of Thumb for Checking ANOVA Conditions

- ▶ Inference requires random samples.
- ▶ Outliers are always problematic! Plot your data.
- ▶ ANOVA methods are robust if group samples sizes are similar and not too small.
- ▶ Ratio of largest sample SD to smallest should not be much greater than 2.
  - ▶ Biggest problem is if sample sizes are unequal and SD from a small sample is much larger than others.
- ▶ Normality is not critical if sample sizes n_i are large and approximately equal.