## $2/2/2010$  36-402/608 ADA-II H. Seltman Handout  $\#7$ : One-way ANOVA

- 1. Analysis of Variance is a "workhorse" of data analysis, rather than an advanced technique.
- 2. Concept of the F test for equal group means

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- (a) Focuses on categorical explanatory variables (with Normal outcomes)
- (b) Uses variance-like-quantities as a tool to investigate mean differences.
- (c) Decompose deviations from the grand mean into meaningful components that are orthogonal. This leads to SS (sums of squared deviations) components that meaningfully add up to the total SS.
- (d)  $A + B = C \Rightarrow A^2 + B^2 = C^2$  only when A and B are orthogonal, which corresponds to statistically independent.
- (e) One way ANOVA with I groups each with  $n_i$  observations  $Y_{ij}$ ,  $j \in \{1, \ldots, n_i\}$ :

$$
Y_{ij} - \bar{Y} = (\bar{Y}_i - \bar{Y}) + (Y_{ij} - \bar{Y}_i)
$$

$$
\sum_{ij} (Y_{ij} - \bar{Y}) = \sum_{ij} [(\bar{Y}_i - \bar{Y}) + (Y_{ij} - \bar{Y}_i)] = 0
$$

$$
\sum (Y_{ij} - \bar{Y})^2 = \sum (\bar{Y}_i - \bar{Y})^2 + \sum (Y_{ij} - \bar{Y}_i)^2 + 0
$$

$$
SS_{\text{Total}} = SS_{\text{Between Groups}} + SS_{\text{Within Groups}}
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- (f) Definition:  $MS = SS/df$ , applies to  $s_Y^2 = SS_T/(n-1)$ .
- (g) Well known fact: For  $Y_i \sim N(\mu, \sigma^2)$ ,  $Z_i = \frac{Y_i \mu_i}{\sigma^2}$  $\frac{-\mu_i}{\sigma}$  is standard normal. If  $Z_1, \ldots, Z_p$  are independent standard normal variables, then  $\sum_{i=1}^p Z^2$  is a chisquare random variable with p df.
- (h) Intuitively similar fact: If a sum of squares of normal deviates (with variance  $\sigma^2$ ) has d degrees of freedom, then with  $SS/\sigma^2$  has a distribution that is chisquared with  $d$  degrees of freedom.
- (i) Well known fact: If  $U \sim \chi^2_u$  and  $V \sim \chi^2_v$  are independent, then  $\frac{U/u}{V/v} \sim F_{u,v}$ .
- (j) Conclusion:  $\frac{SS_B/df_B}{SS_W/df_W} = MS_B/MS_W \sim F_{df_B,df_W}$  if both mean squares have the same variance.  $MS_W$  has variance  $\sigma_2$ , but  $MS_B$  has that variance only if  $H_0: \mu_1 = \mu_2 = \ldots = \mu_I$ , otherwise it has a larger variance.
- 3. Concept of the planned test of contrasts:
	- (a) Let  $C_1, ..., C_I$  be a set of coefficients adding to 0.
	- (b) C defines a "contrast null hypothesis":  $\gamma = C_1 \mu_1 + \cdots + C_I \mu_i = 0$ . E.g., with I=4 and  $C=(1,0,-1/2,-1/2)$  this is a test of whether the mean of treatment 1 differs from the average of the means of treatments 3 and 4.
	- (c) The obvious estimate of  $\gamma$  is  $G = C_1 \overline{Y}_1 + \cdots + C_I \overline{Y}_I$ .
	- (d) If  $\sigma^2$  is known, the variance pf  $a\bar{Y}_1$  is  $a^2\frac{\sigma^2}{n_1}$  $\frac{\sigma^2}{n_1}$ . The variance of G is  $\left(\sum_{i=1}^I C_i^2 \frac{\sigma^2}{n_i}\right)$ ni .
	- (e) If  $\sigma^2$  is unknown, the best estimate of G's variance is  $s_p^2\left(\sum_{i=1}^I C_i^2/n_i\right)$
	- (f)  $s_p^2$  is  $MS_W$  in ANOVA or the "residual standard error" in regression.
	- (g) As usual  $SE(G) = \sqrt{Var(G)}$ .
	- (h) Test  $H_0: \gamma = 0$  using  $T = \frac{G}{SE(G)}$  with df<sub>W</sub> degrees of freedom.
- 4. Breakout and Discussion