Chemometrics MA4605

Week 4. Lecture 8. ANOVA

September 27, 2011

ANalysis Of VAriance

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Example Three different groups of people such as

- smoker
- nonsmoker(exposed to tobacco)
- nonsmoker(nonexposed)

are measured for nicotine. We can test whether on average they have the same level of nicotine.

$$H_0: \mu_1 = \mu_2 = \mu_3$$

$$H_a: \mu_1 \neq \mu_2 \neq \mu_3$$



- One-way ANOVA. It is used when the sample data are separated into groups according to 1 characteristic or factor.
- Two-way ANOVA. it is used to compare population means when the sample is separated into groups by 2 factors.

One way ANOVA tests the claim that the means are equal for *k* independent groups

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- Each population from which a sample is taken is assumed to be normal.
- Each sample is randomly selected and independent.
- The populations have approximately equal variances(standard deviations).



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- \mathbf{s}_{j}^{2} is the variance of group j.
- $\overline{\overline{x}}$ is the overall mean for all the observations
- \blacksquare n_j is the number of observations in group j.

Example

The results obtained in an investigation into the stability of of a fluorescent reagent stored under three different conditions. The values for the fluorescence signals are:

	Group1	Group2	Group3
	23	27	24
	23	29	26
	20	25	24
	21	23	
		24	
sample size n	4	5	3
sample means	21.75	25.6	24.67
sample variance	2.25	5.8	1.33



ANOVA distinguishes between two sources of variation: between groups and within groups.

To distinguish between groups, the variability between groups must be greater than the variability within groups.

This way we decide if the groups are significantly different.

Total variation = Total Sum of Squares= TSS =
$$\sum_{i=1}^{n} (x_i - \overline{x})^2$$
 = $(23 - 24.08)^2 + (23 - 24.08)^2 + ... + (24 - 24.08)^2$ = 66.92 **Degrees of freedom**= n-1=12-1=11

Number of groups is k=3.

Variation between groups = Sum of Squares Between groups

= SSB =
$$\sum_{j=1}^{k} n_j (\overline{x}_j - \overline{\overline{x}})^2$$

-4(21.75 - 24.08)² + 5(2

$$=4(21.75-24.08)^2+5(25.6-24.08)^2+3(24.67-24.08)^2$$

=34.3

Degrees of freedom= Number of groups-1=k-1=3-1=2



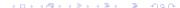
Variation within groups = Sum of Squares Within groups

$$= \sum_{j=1}^{k} (n_j - 1) s_j^2$$

=3(2.25)+5(5.8)+2(1.33)

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Degrees of freedom=
$$\sum_{j=1}^{k} (n_j - 1) = 3 + 4 + 2 = 9$$



Rules

- Total variation= Variation between groups = Variation within groups 66.92 = 34.3+ 32.62
- Total df= Between groups df+ Within groups df 11= 2+9



Calculate the test statistic

- Mean Square Between groups= MSB = $\frac{SSB}{df \ between} = \frac{34.3}{2} = 17.15$
- Mean Square Within groups= MSE = $\frac{SSE}{df \ within} = \frac{32.62}{9} = 3.62$
- Test Statistics= $F = \frac{MSB}{MSE} = \frac{17.15}{3.62} = 4.7$
- $F > F_{2,9} = 4.26$, hence we reject the null hypothesis.