

Case Study



Gas Mileage for Classes of Vehicles

Data analysis

Means (\bar{x} s):

Midsize: 27.903

SUV: 22.677

Pickup: 21.286

- ◆ Mean gas mileage for SUVs and pickups appears less than for midsize cars
- ◆ Are these differences statistically significant?

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Data analysis

Means (\bar{x} s):

Midsize: 27.903

SUV: 22.677

Pickup: 21.286

Null hypothesis:
The true means (for gas mileage) are the same for all groups (the three vehicle classifications)

For example, could look at separate t tests to compare each pair of means to see if they are different:

27.903 vs. 22.677, 27.903 vs. 21.286, & 22.677 vs. 21.286

$H_0: \mu_1 = \mu_2$

$H_0: \mu_1 = \mu_3$

$H_0: \mu_2 = \mu_3$

Problem of *multiple comparisons*!

Multiple Comparisons

- ◆ Problem of how to do many comparisons at the same time with some overall measure of confidence in all the conclusions
- ◆ Two steps:
 - overall test to test for any differences
 - follow-up analysis to decide which groups differ and how large the differences are
- ◆ Follow-up analyses can be quite complex; we will look at only the overall test for a difference in several means, and examine the data to make follow-up conclusions

Analysis of Variance F Test

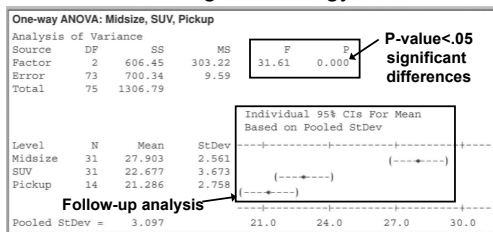
- ◆ $H_0: \mu_1 = \mu_2 = \mu_3$
- ◆ H_a : not all of the means are the same
- ◆ To test H_0 , compare how much variation exists among the sample means (how much the \bar{x} s differ) with how much variation exists within the samples from each group
 - is called the **analysis of variance F test**
 - test statistic is an F statistic
 - ◆ use F distribution (F table) to find P -value
 - analysis of variance is abbreviated **ANOVA**

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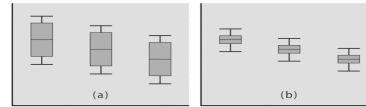
Data analysis

- ◆ $F = 31.61$
- ◆ P -value = 0.000 (rounded) (is < 0.001)
 - there is significant evidence that the three types of vehicle do not all have the same gas mileage
 - from the confidence intervals (and looking at the original data), we see that SUVs and pickups have similar fuel economy and both are distinctly poorer than midsize cars

ANOVA Idea

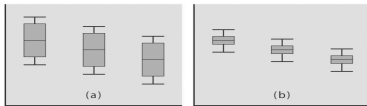
- ◆ ANOVA tests whether several populations have the same mean by comparing how much variation exists among the sample means (how much the \bar{x} 's differ) with how much variation exists within the samples from each group
 - the decision is not based only on how far apart the sample means are, but instead on how far apart they are relative to the variability of the individual observations within each group

ANOVA Idea



- ◆ Sample means for the three samples are the same for each set (a) and (b) of boxplots (shown by the center of the boxplots)
 - variation among sample means for (a) is identical to (b)
- ◆ Less spread in the boxplots for (b)
 - variation among the individuals within the three samples is much less for (b)

ANOVA Idea

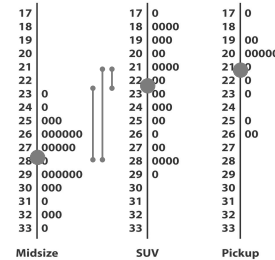


- ◆ CONCLUSION: the samples in (b) contain a larger amount of variation among the sample means relative to the amount of variation within the samples, so ANOVA will find more significant differences among the means in (b)
 - assuming equal sample sizes here for (a) and (b)
 - larger samples will find more significant differences

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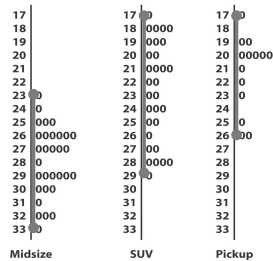
Variation among sample means (how much the \bar{x} 's differ from each other)



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Variation within the individual samples



ANOVA F Statistic

- ◆ To determine statistical significance, we need a test statistic that we can calculate
 - ANOVA F Statistic:

$$F = \frac{\text{variation among the sample means}}{\text{variation among individuals in the same sample}}$$
 - must be zero or positive
 - ✦ only zero when all sample means are identical
 - ✦ gets larger as means move further apart
 - large values of F are evidence against H_0 : equal means
 - the F test is upper one-sided (like the chi-square test)

ANOVA F Test

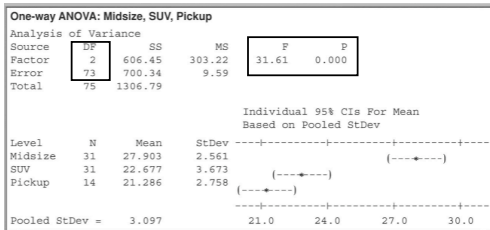
- ◆ Calculate value of F statistic
 - by hand (cumbersome)
 - using technology (computer software, etc.)
- ◆ Find P -value in order to reject or fail to reject H_0
 - F table (not provided in book. Will provide on website)
 - from computer output
- ◆ If significant relationship exists (small P -value):
 - follow-up analysis
 - ✦ observe differences in sample means in original data
 - ✦ formal multiple comparison procedures (not covered here)

ANOVA F Test

- ◆ F test for comparing I populations, with an SRS of size n_i from the i^{th} population (thus giving $N = n_1 + n_2 + \dots + n_I$ total observations) uses critical values from an F distribution with the following *numerator* and *denominator degrees of freedom*:
 - *numerator df* = $I - 1$
 - *denominator df* = $N - I$
- ◆ P -value is the area to the right of F under the density curve of the F distribution

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$F = 31.61$ $I = 3$ classes of vehicle
 $n_1 = 31$ midsize, $n_2 = 31$ SUVs, $n_3 = 14$ trucks
 $N = 31 + 31 + 14 = 76$
 $df_{num} = (I-1) = (3-1) = 2$
 $df_{den} = (N-I) = (76-3) = 73$
 P -value from technology output is **0.000**. This probability is not 0, but is very close to 0 and is smaller than 0.001, the smallest value the technology can record.
 ** P -value < .05, so we conclude significant differences **

ANOVA Model, Assumptions

- ◆ Conditions required for using ANOVA F test to compare population means
 - 1) have I *independent* SRSs, one from each population.
 - 2) the i^{th} population has a *Normal distribution* with unknown mean μ_i (means may be different).
 - 3) all of the populations have the *same standard deviation* σ , whose value is unknown.

Robustness

- ◆ ANOVA F test is not very sensitive to lack of Normality (is **robust**)
 - what matters is Normality of the sample means
 - ANOVA becomes safer as the sample sizes get larger, due to the Central Limit Theorem
 - if there are no outliers and the distributions are roughly symmetric, can safely use ANOVA for sample sizes as small as 4 or 5

Robustness

- ◆ ANOVA F test is not too sensitive to violations of the assumption of equal standard deviations
 - especially when all samples have the same or similar sizes and no sample is very small
 - statistical tests for equal standard deviations are very sensitive to lack of Normality (not practical)
 - check that sample standard deviations are similar to each other (next slide)

Checking Standard Deviations

- ◆ The results of ANOVA F tests are approximately correct when the largest sample standard deviation (s) is no more than twice as large as the smallest sample standard deviation

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One-way ANOVA: Midsize, SUV, Pickup			
Analysis of Variance			
Source	DF	SS	MS
Factor	2	606.45	303.22
Error	73	700.34	9.59
Total	75	1306.79	

Level	N	Mean	StDev
Midsize	31	27.903	2.561
SUV	31	22.677	3.673
Pickup	14	21.286	2.758

Pooled StDev = 3.097

$s_1 = 2.561$
 $s_2 = 3.673$
 $s_3 = 2.758$
 $\frac{\text{largest } 3.673}{\text{smallest } 2.561} = 1.434$
 ⇒ safe to use ANOVA F test

ANOVA Details

- ◆ ANOVA F statistic:

$$F = \frac{\text{variation among the sample means}}{\text{variation among individuals in the same sample}}$$
 - the measures of variation in the numerator and denominator are **mean squares**
 - ◆ general form of a sample variance
 - ◆ ordinary s^2 is “an average (or mean) of the squared deviations of observations from their mean”

ANOVA Details

- ◆ Numerator: Mean Square for Groups (**MSG**)
 - an average of the I squared deviations of the means of the samples from the overall mean \bar{X}
$$MSG = \frac{n_1(\bar{x}_1 - \bar{x})^2 + n_2(\bar{x}_2 - \bar{x})^2 + \dots + n_I(\bar{x}_I - \bar{x})^2}{I - 1}$$
 - ◆ n_i is the number of observations in the i^{th} group
 - ◆ $\bar{x} = \frac{n_1\bar{x}_1 + n_2\bar{x}_2 + \dots + n_I\bar{x}_I}{N}$

ANOVA Details

- ◆ Denominator: Mean Square for Error (**MSE**)
 - an average of the individual sample variances (s_i^2) within each of the I groups
$$MSE = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \dots + (n_I - 1)s_I^2}{N - I}$$
 - ◆ MSE is also called the **pooled sample variance**, written as s_p^2 (s_p is the **pooled standard deviation**)
 - ◆ s_p^2 estimates the common variance σ^2

ANOVA Details

- the numerators of the mean squares are called the **sums of squares** (*S_S* and *S_{S_E}*)
- the denominators of the mean squares are the two degrees of freedom for the *F* test, (*I*-1) and (*N*-1)
- usually results of ANOVA are presented in an **ANOVA table**, which gives the *source of variation*, *df*, *SS*, *MS*, and *F* statistic

◆ ANOVA *F* statistic: $F = \frac{MSG}{MSE} = \frac{SSG/dfG}{SSE/dfE}$

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One-way ANOVA: Midsize, SUV, Pickup

Analysis of Variance					
Source	DF	SS	MS	F	P
Factor	2	606.45	303.22	31.61	0.000
Error	73	700.34	9.59		
Total	75	1306.79			

For detailed calculations, see Examples 24.7 and 24.8 on pages 652-654 of the textbook.

THE ANOVA F TEST Summary

Draw an independent SRS from each of *I* populations. The *i*th population has the $N(\mu_i, \sigma)$ distribution, where σ is the common standard deviation in all the populations. The *i*th sample has size n_i , sample mean \bar{x}_i , and sample standard deviation s_i .

The ANOVA *F* statistic tests the null hypothesis that all *I* populations have the same mean:

$H_0: \mu_1 = \mu_2 = \dots = \mu_I$
 H_a : not all of the μ_i are equal

The statistic is

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

The numerator of *F* is the mean square for groups

$$MSG = \frac{n_1(\bar{x}_1 - \bar{x})^2 + n_2(\bar{x}_2 - \bar{x})^2 + \dots + n_I(\bar{x}_I - \bar{x})^2}{I - 1}$$

The denominator of *F* is the mean square for error

$$MSE = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \dots + (n_I - 1)s_I^2}{N - I}$$

When H_0 is true, *F* has the *F* distribution with *I* - 1 and *N* - *I* degrees of freedom.

ANOVA Confidence Intervals

- ◆ **Confidence interval** for the mean μ_i of any group:

$$\bar{x}_i \pm t^* \frac{S_p}{\sqrt{n_i}}$$

- t^* is the critical value from the *t* distribution with *N*-1 degrees of freedom
- S_p (pooled standard deviation) is used to estimate σ because it is better than any individual s_i

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