

Lecture 21: more 11.3 (ANOVA) Categorical & Quantitative Variable Begin Ch.12 Inf. for 2 Categorical Vars.

- ANOVA: Table, Test Stat, P -value
- 1st Step in Practice: Displays, Summaries
- ANOVA Output
- Guidelines for Use of ANOVA
- Formulating Hypotheses about 2 Cat. Vars.
- Test Based on Proportions or Counts: z or ChiSq

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Looking Back: Review

- **4 Stages of Statistics**
 - Data Production (discussed in Lectures 1-3)
 - Displaying and Summarizing (Lectures 3-8)
 - Probability (discussed in Lectures 9-14)
 - Statistical Inference
 - 1 categorical (discussed in Lectures 14-16)
 - 1 quantitative (discussed in Lectures 16-18)
 - cat and quan: paired, 2-sample, several-sample
 - 2 categorical
 - 2 quantitative

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The F Statistic (Review)

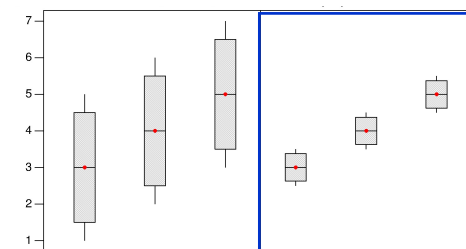
$$F = \frac{[n_1(\bar{x}_1 - \bar{x})^2 + n_2(\bar{x}_2 - \bar{x})^2 + \cdots + n_I(\bar{x}_I - \bar{x})^2] / (I - 1)}{[(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \cdots + (n_I - 1)s_I^2] / (N - I)}$$

- **Numerator:** variation **among** groups
 - How different are $\bar{x}_1, \dots, \bar{x}_I$ from one another?
- **Denominator:** variation **within** groups
 - How spread out are samples? (sds s_1, \dots, s_I)

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Role of Variations on Conclusion (Review)

Boxplots with same variation *among* groups (3, 4, 5) but different variation *within*: sds large (left) or small (right)



Scenario on right: smaller s.d.s \rightarrow larger $F = \frac{\text{var among}}{\text{var within}}$
 \rightarrow smaller P -value \rightarrow likelier to reject $H_0 \rightarrow$ conclude **pop means differ**

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ANOVA Table

Source	Degrees of Freedom	Sum of Squares	Mean Sum of Squares	F	P
Factor	$DFG = I - 1$	SSG	$MSG = SSG/DFG$	$F = \frac{MSG}{MSE}$	p-value
Error	$DFE = N - I$	SSE	$MSE = SSE/DFE$		
Total	$N - 1$	SST			

Organizes calculations

- “Source” refers to source of variation:
 - “Factor” refers to variation **among** groups (expl var)
This variation is from the numerator.
 - “Error” refers to individuals differing **within** groups
This variation is from the denominator.

ANOVA Table

Source	Degrees of Freedom	Sum of Squares	Mean Sum of Squares	F	P
Factor	$DFG = I - 1$	SSG	$MSG = SSG/DFG$	$F = \frac{MSG}{MSE}$	p-value
Error	$DFE = N - I$	SSE	$MSE = SSE/DFE$		
Total	$N - 1$	SST			

Organizes calculations

- “Source” refers to source of variation
- DF: use I = no. of groups, N = total sample size
 - $DFG = I - 1$
 - $DFE = N - I$

ANOVA Table

Source	Degrees of Freedom	Sum of Squares	Mean Sum of Squares	F	P
Factor	$DFG = I - 1$	SSG	$MSG = SSG/DFG$	$F = \frac{MSG}{MSE}$	p-value
Error	$DFE = N - I$	SSE	$MSE = SSE/DFE$		
Total	$N - 1$	SST			

Organizes calculations

- “Source” refers to source of variation
- DF: use I = no. of groups, N = total sample size
- SSG measures overall variation **among** groups
- SSE measures overall variation **within** groups
- SSG and SSE tedious to calculate; other table entries straightforward, except for P -value**

ANOVA Table

Source	Degrees of Freedom	Sum of Squares	Mean Sum of Squares	F	P
Factor	$DFG = I - 1$	SSG	$MSG = SSG/DFG$	$F = \frac{MSG}{MSE}$	p-value
Error	$DFE = N - I$	SSE	$MSE = SSE/DFE$		
Total	$N - 1$	SST			

Organizes calculations

- “Source” refers to source of variation
- DF: use I = no. of groups, N = total sample size
- SSG measures overall variation among groups
- SSE measures overall variation within groups
- Mean Sums: Divide Sums by DFs
- F : Take quotient of MSG and MSE
- P -value: Found with software or tables

Example: Key ANOVA Values

- **Background:** Compare mileages for 8 sedans, 8 minivans, 12 SUVs; find $SSG=42.0$, $SSE=181.4$.
- **Question:** What are the following values for table:
 - DFG? DFE? MSG? MSE? F ?
- **Response:**
 - $DFG = 3 - 1 = \underline{\hspace{1cm}}$
 - $DFE = N - I = (8+8+12) - 3 = \underline{\hspace{1cm}}$
 - $MSG = SSG/DFG = 42/2 = \underline{\hspace{1cm}}$
 - $MSE = SSE/DFE = 181.4/25 = \underline{\hspace{1cm}}$
 - $F = MSG/MSE = 21/7.256 = \underline{\hspace{1cm}}$

Example: Completing ANOVA Table

- **Background:** Found these values for ANOVA:
 - $DFG=3-1=2$
 - $DFE=N-I=(8+8+12)-3=25$
 - $MSG=SSG/DFG=42/2=21$
 - $MSE=SSE/DFE=181.4/25=7.256$
 - $F=MSG/MSE=21/7.256=2.89$
- **Question:** Complete ANOVA table?
- **Response:** Software $\rightarrow P\text{-val}=0.0743 \rightarrow \underline{\hspace{1cm}}$

Source	DF	SS	MS	F	P
Factor					
Error					

ANOVA F Statistic and P -Value

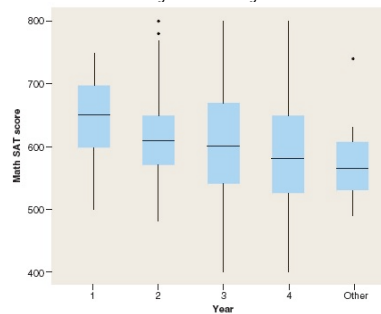
- Sample means **very different** \rightarrow
 F **large** \rightarrow
 P -value small \rightarrow
Reject claim of equal population means.
- Sample means **relatively close** \rightarrow
 F **not large** \rightarrow
 P -value **not small** \rightarrow
Believe claim of equal population means.

How Large is “Large” F

Particular F distribution determined by
DFG, DFE
(these determined by sample size, number of groups)
 P -value in software output lets us know if F is large.
Note: P -value is “bottom line” of test; “top line” is examination of display and summaries.

Example: Examining *Boxplots*

- **Background:** For all students at a university, are Math SATs related to what year they're in?



- **Question:** What do the boxplots suggest?
- **Response:** As year goes up, mean _____
(Suggests _____ students scored better in Math.)

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Example: Examining *Summaries*

- **Background:** For all students at a university, are Math SATs related to what year they're in?

Level	N	Mean	StDev
1	32	643.75	63.69
2	233	613.91	61.00
3	87	601.84	89.79
4	28	581.79	89.73
other	10	578.00	72.08

- **Question:** What do the summaries suggest?
- **Response:** Means decrease by about _____ points for each successive year 1 to 4. Standard deviations are around _____, and sample sizes are _____.

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Example: *ANOVA Output*

- **Background:** For all students at a university, are Math SATs related to what year they're in?

Analysis of Variance for Math

Source	DF	SS	MS	F	P
Year	4	78254	19563	3.87	0.004
Error	385	1946372	5056		
Total	389	2024626			

- **Question:** What does the output suggest?
- **Response:** Test H_0 :
 P -value=0.004. Small? _____ Reject H_0 ?
_____ Conclude all 5 population means may be equal?
_____ Year and Math SAT related in population? _____

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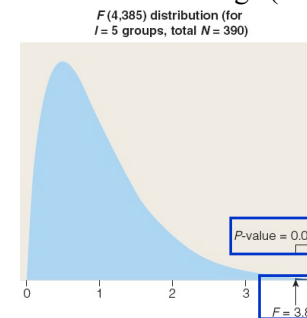
How Large is "Large" F (Review)

Particular F dist determined by DFG, DFE

(these determined by sample size, number of groups)

P -value in software output lets us know if F is large.

P -value = 0.004 $\rightarrow F = 3.87$ is large (in given situation)



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Example: ANOVA Output

- **Background:** A test for a relationship between Math SAT and year of study, based on data from a large sample of intro stats students at a university, produced a large F and a small P -value.
- **Question:** What issues should be considered before we use these results to draw conclusions about the relationship between year of study and Math SAT for all students at that university?
- **Response:**

Guidelines for Use of ANOVA Procedure

- Need random samples taken independently from several populations.
- Confounding variables should be separated out.
- Sample sizes must be large enough to offset non-normality of distributions.
- Need populations at least 10 times sample sizes.
- Population variances must be equal.

Pooled Two-Sample t Procedure (Review)

If we can assume $\sigma_1 = \sigma_2$, standardized difference between sample means follows a pooled t distribution.

- Some apply **Rule of Thumb:** use pooled t if larger sample s.d. not more than twice smaller.

The F distribution is in a sense “pooled”: our standardized statistic follows the F distribution only if population variances are equal (same as equal s.d.s)

Example: Checking Standard Deviations

- **Background:** For all students at a university, are Math SATs related to what year they're in?

Level	N	Mean	StDev
1	32	643.75	63.69
2	233	613.91	61.00
3	87	601.84	89.79
4	28	581.79	89.73
other	10	578.00	72.08

- **Question:** Is it safe to assume equal population variances?

- **Response:**

Largest s.d. = _____ > 2(smallest s.d.) _____?

_____ Assumption of equal variances OK? _____

- **Background:** For all students at a university, are Verbal SATs related to what year they're in?

- **Questions:** Are conditions met? Do the data provide evidence of a relationship?

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Inference for Relationship (Review)

- H_0 and H_a about **variables**: not related or related
 - Applies to all three $C \rightarrow Q$, $C \rightarrow C$, $Q \rightarrow Q$
- H_0 and H_a about **parameters**: equality or not
 - $C \rightarrow Q$: pop **means** equal?
 - $C \rightarrow C$: pop **proportions** equal?
 - $Q \rightarrow Q$: pop **slope** equals zero?

Example: 2 Categorical Variables: Hypotheses

- **Background:** We are interested in whether or not smoking plays a role in alcoholism.
- **Question:** How would H_0 and H_a be written
 - in terms of **variables**?
 - in terms of **parameters**?
- **Response:**
 - in terms of **variables**
 - H_0 : smoking and alcoholism _____ related
 - H_a : smoking and alcoholism _____ related
 - in terms of **parameters**
 - H_0 : Pop proportions alcoholic _____ for smokers, non-smokers
 - H_a : Pop. proportions alcoholic _____ for smokers, non-smokers

The word "not" appears
in H_0 about variables,
in H_a about parameters.

Example: Summarizing with Proportions

- **Background:** Research Question: Does smoking play a role in alcoholism?
- **Question:** What statistics from this table should we examine to answer the research question?
- **Response:** Compare proportions _____ (response) for _____ (explanatory).

	Alcoholic	Not Alcoholic	Total
Smoker	30	200	230
Nonsmoker	10	760	770
Total	40	960	1,000

Example: Test Statistic for Proportions

- **Background:** One approach to the question of whether smoking and alcoholism are related is to compare proportions.
- | | Alcoholic | Not Alcoholic | Total |
|-----------|-----------|---------------|-------|
| Smoker | 30 | 200 | 230 |
| Nonsmoker | 10 | 760 | 770 |
| Total | 40 | 960 | 1,000 |
- $\hat{p}_1 = \frac{30}{230} = 0.130$
 $\hat{p}_2 = \frac{10}{770} = 0.013$
- **Question:** What would be the next step, if we've summarized the situation with the difference between sample proportions 0.130-0.013?
 - **Response:** _____ the difference between sample proportions 0.130-0.013.
Stan. diff. is normal for large n : _____

z Inference for 2 Proportions: Pros & Cons

Advantage:

Can test against *one-sided* alternative.

Disadvantage:

2-by-2 table: comparing proportions straightforward

Larger table: comparing proportions complicated,
can't just standardize one difference $\hat{p}_1 - \hat{p}_2$

Another Comparison in Considering Categorical Relationships (*Review*)

- Instead of considering how different are the *proportions* in a two-way table, we may consider how different the *counts* are from what we'd expect if the “explanatory” and “response” variables were in fact unrelated.
- Compared observed, expected counts in wasp

study:	NA	T
B	16	31
U	24	31
T	40	62

Exp	A	NA	T
B	20	11	31
U	20	11	31
T	40	22	62

Inference Based on Counts

To test hypotheses about relationship in *r*-by-*c* table, compare *counts observed* to *counts expected* if H_0 (equal proportions in response of interest) were true.

Example: *Table of Expected Counts*

- **Background:** Data on smoking and alcoholism:

	Alcoholic	Not Alcoholic	Total
Smoker	30	200	230
Nonsmoker	10	760	770
Total	40	960	1,000

- **Question:** What counts are expected if H_0 is true?

- **Response:** Overall proportion alcoholic is _____

If proportions alcoholic were same for S and NS, expect

- $(40/1,000)(230) = \underline{\hspace{1cm}}$ smokers to be alcoholic
- $(40/1,000)(770) = \underline{\hspace{1cm}}$ non-smokers to be alcoholic; also
- $(960/1,000)(230) = \underline{\hspace{1cm}}$ smokers not alcoholic
- $(960/1,000)(770) = \underline{\hspace{1cm}}$ non-smokers not alcoholic

Example: Table of Expected Counts

- **Background:** If proportions alcoholic were same for S and NS, expect
 - $(40/1,000)(230) = 9.2$ smokers to be alcoholic
 - $(40/1,000)(770) = 30.8$ non-smokers to be alcoholic; also
 - $(960/1,000)(230) = 220.8$ smokers not alcoholic
 - $(960/1,000)(770) = 739.2$ non-smokers not alcoholic
- **Question:** Where do they appear in table of expected counts?
- **Response:**

	Alcoholic	Not Alcoholic	Total
Smoker			230
Non-smoker			770
Total	40	960	1,000

Note:
 $9.2/230 =$
 $30.8/770 =$
 $40/1,000$

Example: Table of Expected Counts

	Alcoholic	Not Alcoholic	Total
Smoker	9.2	220.8	230
Non-smoker	30.8	739.2	770
Total	40	960	1000

- **Note:** Each expected count is $\frac{\text{Column total} \times \text{Row total}}{\text{Table total}}$
- Expect:**
 - $(40)(230)/1,000 = 9.2$ smokers to be alcoholic
 - $(40)(770)/1,000 = 30.8$ non-smokers to be alcoholic; also
 - $(960)(230)/1,000 = 220.8$ smokers not alcoholic
 - $(960)(770)/1,000 = 739.2$ non-smokers not alcoholic

Chi-Square Statistic

- Components to compare observed and expected counts, one table cell at a time:

$$\text{component} = \frac{(\text{observed} - \text{expected})^2}{\text{expected}}$$

Components are **individual standardized squared differences**.

- **Chi-square** test statistic χ^2 combines all components by summing them up:

$$\text{chi-square} = \text{sum of } \frac{(\text{observed} - \text{expected})^2}{\text{expected}}$$

Chi-square is **sum** of standardized squared differences.

Example: Chi-Square Statistic

- **Background:** Observed and Expected Tables:

Obs	A	NA	Total
S	30	200	230
NS	10	760	770
Total	40	960	1000

Exp	A	NA	Total
S	9.2	220.8	230
NS	30.8	739.2	770
Total	40	960	1000

- **Question:** What is the chi-square statistic?
- **Response:** Find $\text{chi-square} = \text{sum of } \frac{(\text{observed} - \text{expected})^2}{\text{expected}}$

Example: Assessing Chi-Square Statistic

- **Background:** We found chi-square = 64.
- **Question:** Is the chi-square statistic (64) large?
- **Response:**

Chi-Square Distribution

chi-square = sum of $\frac{(\text{observed} - \text{expected})^2}{\text{expected}}$ follows a predictable pattern (assuming H_0 is true) known as

chi-square distribution with $df = (r-1) \times (c-1)$

- r = number of rows (possible explanatory values)
- c = number of columns (possible response values)

Properties of chi-square:

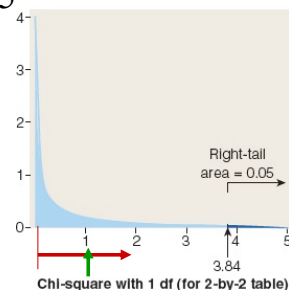
- Non-negative (based on squares)
- Mean = df [=1 for smallest (2×2) table]
- Spread depends on df
- Skewed right

Chi-Square Density Curve

For chi-square with 1 df, $P(\chi^2 \geq 3.84) = 0.05$
→ If $\chi^2 > 3.84$, $P\text{-value} < 0.05$

Properties of chi-square:

- Non-negative
- Mean = df
 $df=1$ for smallest [2×2] table
- Spread depends on df
- Skewed right



Example: Assessing Chi-Square (Continued)

- **Background:** In testing for relationship between smoking and alcoholism in 2×2 table, found $\chi^2 = 64$
- **Question:** Is there evidence of a relationship in general between smoking and alcoholism (not just in the sample)?
- **Response:** For $df=(2-1) \times (2-1)=1$, chi-square considered “large” if greater than 3.84
→ chi-square=64 large? _____ $P\text{-value}$ small? _____
Evidence of a relationship between smoking and alcoholism? _____

Inference for 2 Categorical Variables; z or χ^2

For 2×2 table, $z^2 = \chi^2$

- z statistic (comparing proportions) → combined tail probability = 0.05 for $z = 1.96$
- chi-square statistic (comparing counts) → right-tail prob = 0.05 for $\chi^2 = 1.96^2 = 3.84$

Example: Relating Chi-Square & z

- **Background:** We found chi-square = 64 for the 2-by-2 table relating smoking and alcoholism.
- **Question:** What would be the z statistic for a test comparing proportions alcoholic for smokers vs. non-smokers?
- **Response:**

Assessing Size of Test Statistics (*Summary*)

When test statistic is “large”:

- z : greater than 1.96 (about 2)
- t : depends on df; greater than about 2 or 3
- F : depends on DFG, DFE
- χ^2 depends on $df = (r-1) \times (c-1)$; greater than 3.84 (about 4) if $df=1$

Lecture Summary (Inference for Cat → Quan; More About ANOVA)

- ANOVA for several-sample inference
 - ANOVA table
 - F statistic and P -value
- 1st step in practice: displays and summaries
 - Side-by-side boxplots
 - Compare means, look at sds and sample sizes
- ANOVA output
- Guidelines for use of ANOVA

Lecture Summary

(Inference for Cat \rightarrow Cat; Chi-Square)

- Hypotheses in terms of variables or parameters
- Inference based on proportions or counts
- Chi-square test
 - Table of expected counts
 - Chi-square statistic, chi-square distribution
 - Relating z and chi-square for 2×2 table
 - Relative size of chi-square statistic