

Ch12-13: Factorial ANOVA, Blocking ANOVA, and χ^2 Test

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Dr. Sean Ho

busi275.seanho.com

- Please download:
11-ANOVA.xls
- **HW8** this week
- **Presentations**
in two weeks!

Outline for today

- Factorial ANOVA (multiple nominal IVs)
 - Graphing
 - Main effects and interaction
- Randomized Complete Block ANOVA
 - Fixed vs. random effects
 - Post-hoc analysis: Fisher's LSD
- χ^2 goodness-of-fit test: observed vs. expected
 - Test against uniform distribution
 - ◆ CHIDIST(), CHITEST()
 - Test of normality
 - Test 2 nominal vars for independence
 - ◆ Marginal probabilities

Overview: regression / ANOVA

- Both: DV (outcome) quantitative
- Regression: IVs quant. (or use dummy coding)
 - ANOVA: IVs nominal
- Global F-test: do all IVs together impact DV?
 - R^2 : fraction of variability in DV explained
- Main effects: each IV's contribution
 - Significance (p-val) vs. effect size (unique R^2)
 - Follow-up (nominal IVs): Tukey-Kramer
 - ◆ See “Delivery” in 11-ANOVA.xls
- Interaction:
 - Multiple regr: include $X_1 * X_2$ as a predictor
 - Factorial ANOVA (2-way, 3-way, etc.)

2-way ANOVA

DV: Purch Amt

IV1: Gender

IV2:
Src

\$40 \$20 \$25 \$32	\$17 \$21 \$19 \$22
\$30 \$26 \$25	\$22 \$19
\$40 \$45 \$52 \$43 \$39 \$48	\$50 \$60 \$55

- Equivalent to multiple regression
 - Except with nominal predictors
 - N-way ANOVA for N predictors
- IVs are “between-groups” factors:
 - Divide up sample into cells
 - Each participant in only 1 cell
- If your IVs are mixed continuous / nominal, try regression using dummy-coded variables
 - Although this may result in many IVs!
- You can also try ANCOVA:
 - Continuous “covariates” are first factored out then regular ANOVA is done on residuals

2-way: assumptions

DV: Purch Amt

IV1: Gender

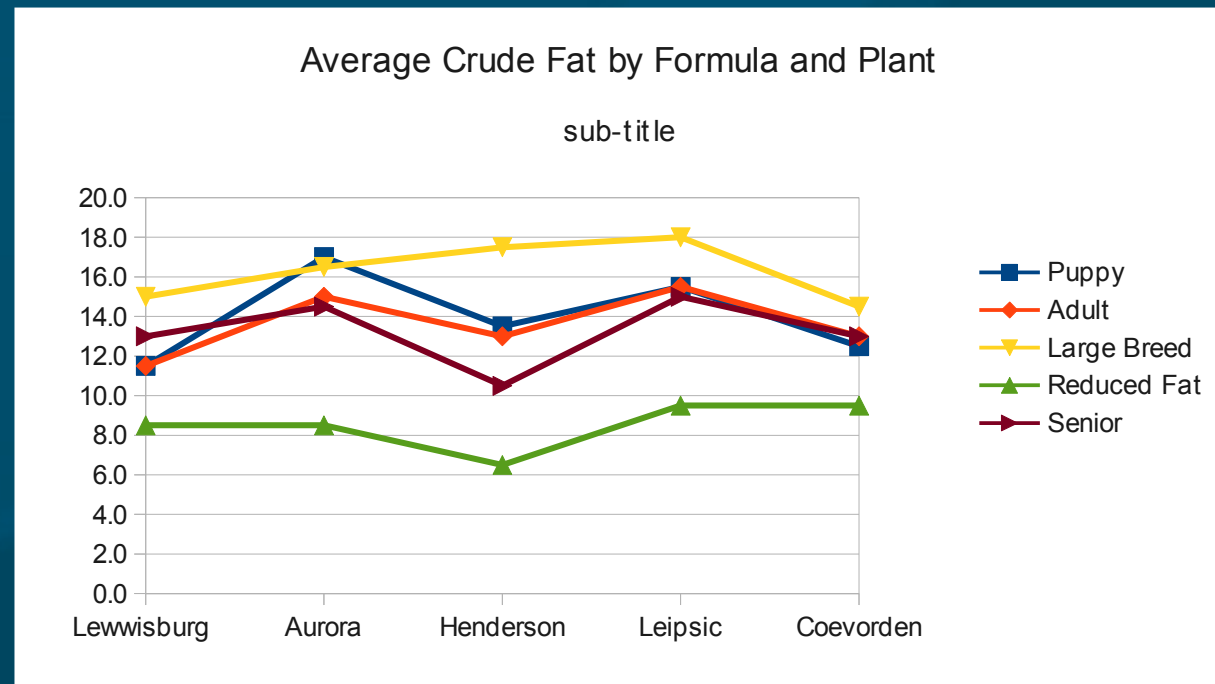
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IV2:
Src

- Same as for regular ANOVA, per cell:
 - DV continuous
 - Independent observations and independent cells (groups)
 - DV normal within each cell
 - Variance of DV similar across all cells:
 - (largest SD) / (smallest SD) < 2
 - Better: Levene's test of homoscedasticity
- The last two are less important as long as:
 - Total sample size is reasonably large (>50)
 - Balanced design: all cells similar sample size
 - No rows/cols are completely empty

Graphing 2-way ANOVA data

- Dataset: “Eukanuba” in **11-ANOVA.xls**
- The DV has a different **distribution** in each cell
- One way to **visualize**: condense it down to the **average** of DV within each cell
- Pivot Table:
 - Formula (row)
 - Plant (col)
 - Average of Fat (data)
- Try a **line** chart:



2-way ANOVA: model

- Main effects on each IV, plus interactions:
 - $\text{Fat} = b_0 + (\text{Formula effect}) + (\text{Plant effect}) + (\text{Formula*Plant effect}) + \varepsilon$
- Decomposition of variance:
 - $SS_{\text{tot}} = SS_{\text{For}} + SS_{\text{Plt}} + SS_{\text{For*Plt}} + SS_{\text{resid}}$
- Global F-test looks for any effect of IVs on DV
 - If not significant, check for violations of assumptions
- Effect size η^2 is akin to R^2 : $1 - (SS_{\text{resid}} / SS_{\text{tot}})$

2-way ANOVA: calculating

-	IV ₁ (a levels)	IV ₂ (b levels)	IV ₁ *IV ₂ (Interaction)
SS	$bn \sum_{i=1}^a (\bar{x}_i - \bar{x})^2$	$an \sum_{j=1}^b (\bar{x}_j - \bar{x})^2$	$n \sum_{i=1}^a \sum_{j=1}^b (\bar{x}_{ij} - \bar{x}_i - \bar{x}_j + \bar{x})^2$
df	a - 1	b - 1	(a - 1) * (b - 1)

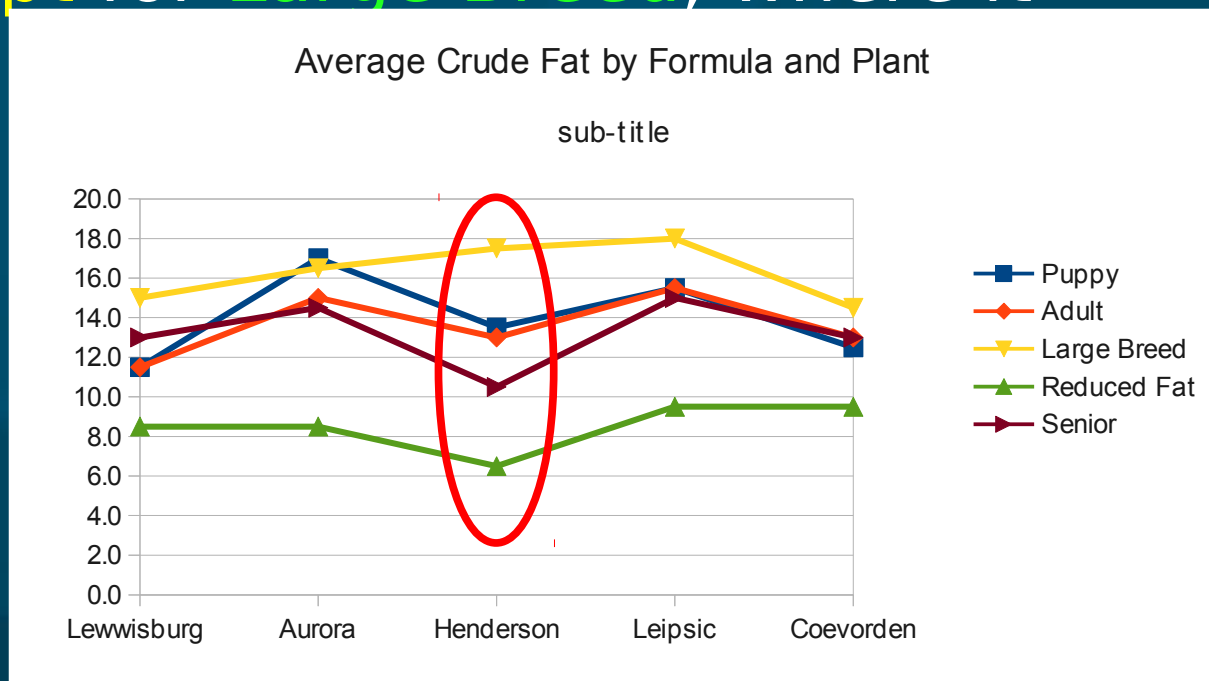
- Also find SS_{tot} as before, and SS_{res}
 - $df_{\text{tot}} = n - 1$, and $df_{\text{res}} = n - ab$
 - The SS and df always add up:
 - ◆ $\text{Tot} = IV_1 + IV_2 + (IV_1 * IV_2) + \text{Resid}$
- 3 F-tests: IV_1 , IV_2 , and interaction
 - e.g., main effect on IV_1 : $F = MS_1 / MS_{\text{res}}$

Main effects

- A **main effect** is a **one-way** ANOVA on one IV, after **controlling** for the other predictors
 - Analogous to **t-tests on slope** for each IV in multiple regression
 - Here, the main effects are themselves **F-tests**
- E.g., do **females** spend **more** at your site, after accounting for **source**?
 - 2-way ANOVA on both Gender and Source, then look at main effect of Gender
- E.g., do different **formulas** have different **fat** content, across all **plants**?

Interactions

- When the **effect** of one IV on the DV changes, depending on the **level** of the moderator
- e.g., **females** spend more in response to **print** ads, but **males** spend more in response to **web**
- e.g., **Henderson** generally has lower fat than the other plants, **except** for **Large Breed**, where it has the second-highest fat:
- Plot **means**, note change in **shape** of the curves



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Randomized Complete Block

- Special case of 2-way ANOVA, where each **cell** has only **1** observation
- **IV1** is the **factor**: typically a **fixed** effect
 - Fixed: **levels** are set in the **hypothesis**: e.g., gender, province, store branch, plant
- **IV2** is the **block**: typically a **random** effect
 - Random: levels are **sampled** from a **population**: e.g., customer, truck, day
- e.g, “Applebees” in **11-ANOVA.xls**
 - **Factor**: Restaurant
 - **Block**: Week
 - **DV**: Revenue

Week	Restaurant		
	8.34	6.79	9.18
	10.7	10.0	12.8

Randomized Block model

- A complete 2-way ANOVA on this data would have **zero residual** in each cell
 - So the **interaction** term serves as “residual”
 - $Tot = Factor + Blocking + Residual$
 - ◆ $df_{res} = (a - 1)(b - 1)$
- **Factor effect** (IV_1): $F = MS_1 / MS_{res}$
 - This is usually what we're most interested in
- **Blocking effect** (IV_2): $F = MS_2 / MS_{res}$
 - If **non-sig**, then blocking was **not** necessary and we could've just done a **1-way** ANOVA

Post-hoc: Fisher's LSD test

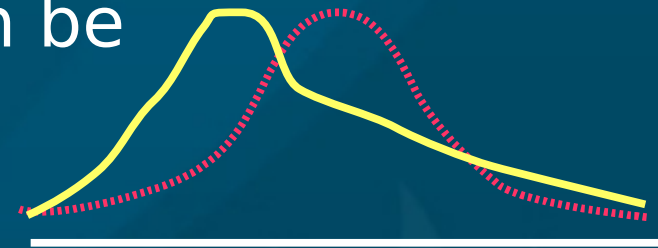
- If the factor effect is significant, one post-hoc test we can use is Fisher's least sig. diff. test
 - Like Tukey-Kramer, but for equal-size cells
- Critical range:
 - t : 2-tails, use df_{res}
 - b : # blocks (IV_2)
$$LSD = t \sqrt{\frac{2 MS_{res}}{b}}$$
- For all pairs of levels of the main factor, if the difference of means $|\bar{x}_i - \bar{x}_j|$ exceeds LSD , then those two groups differ significantly
 - Use the results to cluster the factor levels

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Goodness of fit

- The χ^2 (chi-squared) test is one way to assess goodness of fit:
 - How well an **observed** distribution fits a **hypothesized** distribution
 - Hypothesized distribution can be **uniform, normal, etc.**
- χ^2 can also be applied to test if two **nominal** variables are **independent**
 - Compare pivot table (**contingency table**) with hypothesized results if vars independent
 - Analogous to **correlation** for continuous vars

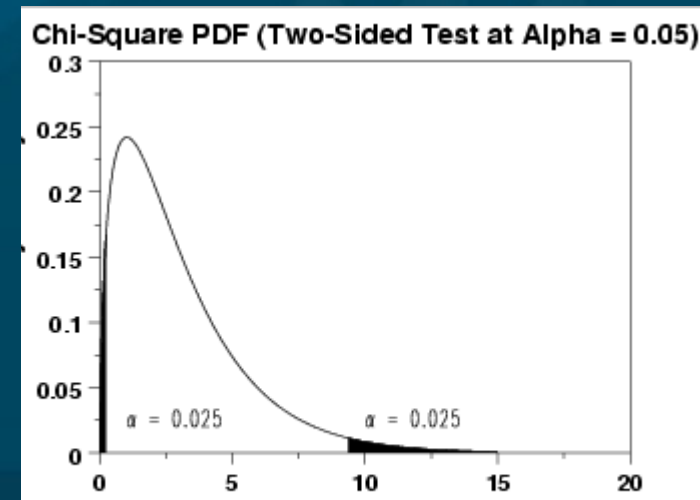


χ^2 vs. uniform distribution

- e.g., are technical support calls evenly distributed across the weekdays?
 - H_0 : evenly distributed, matches uniform dist.
- Expected # calls per day (uniform distribution):
 - Total observed calls (1300), divided by 5

Observed	290	250	238	257	265
Expected	260	260	260	260	260

- Test statistic: $\chi^2 = \sum \frac{(o_i - e_i)^2}{e_i}$
- Use CHIDIST(χ^2 , #cells - 1)
 - Or CHITEST(obs, exp)

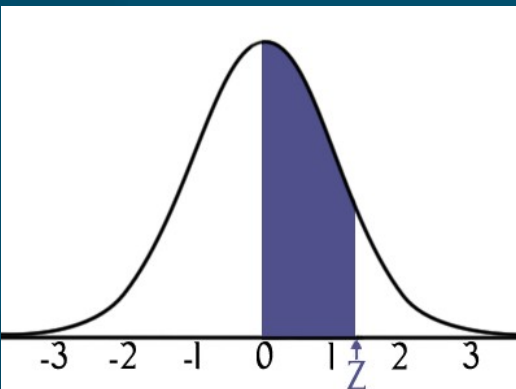


χ^2 vs. normal distribution

- e.g., are student **test** scores **normally** distrib?
- Other normality tests: Shapiro-Wilk, K-S
- Count **frequency** of test scores by **bins**
- How to find **expected** frequencies?
 - Find **mean**, **SD** of the data
 - Use **NORMDIST()** to find percentage of the data that would lie within each **bin** on the ideal **normal**:

Bin	Freq	Norm Freq
40		
45	1	1.30
50	4	2.77
55	15	5.14
60	7	8.28
65	5	11.60
70	7	14.11
75	11	14.93
80	15	13.72
85	22	10.97
90	9	7.62
95	4	4.60
100	0	2.41

◆ $\text{NORMDIST}(80, \mu, \sigma, 1) - \text{NORMDIST}(75, \mu, \sigma, 1)$



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Contingency tables

- Joint **freq.** distribs for **multiple** nominal variables
 - Each **cell** of the table holds the **#** (frequency) of observations that match that combo
 - **Pivot** tables, with **Count** in the Data field
- E.g., **Handedness** vs. **Gender**
 - H_0 : handedness is **independent** of gender: the **probability** of being left-handed stays the same, regardless of the gender
 - ◆ $P(\text{left} \mid M) = P(\text{left} \mid F) = P(\text{left})$

Observed	Left	Right
Male	17	163
Female	33	167

χ^2 on 2-way contingency

- Expected values assume independence
- Calculate marginal probabilities:
 - $P(\text{female}) = 200/380 \approx 52.6\%$
 - $P(\text{left}) = 50/380 \approx 13.2\%$
- Assuming independence,
 - $P(F \cap L) = P(F) * P(L) = (.526)(.132)$
- Thus the expected count for $(F \cap L)$ is
 - $P(F) * P(L) * (\text{total}) = (.526)(.132)(380)$
- Calculate χ^2 summed over all cells
 - $df = (\#rows - 1) (\#cols - 1)$
 - $= 1$ in this case!

	L	R	Tot
M	17	163	180
F	33	167	200
Tot	50	330	380

Summary on χ^2

- Test of **goodness-of-fit**: **observed** vs. **expected**
- May apply to a **single** nominal variable:
 - Expected distrib. may be **uniform**, **normal**, ...
- May apply to **two** nominal variables:
 - Expected distrib. is that vars are **independent**
 - Akin to **correlation** on continuous variables
 - ◆ Large $\chi^2 \leftrightarrow |r| \approx 1$
- But only an **approximation** to the true distrib:
 - Results may be **invalid** if cell counts are **<5**
 - May need to **combine** levels of a var

TODO

- HW8 due Thu
- Projects: be pro-active and self-led
 - All groups have passed REB by now
 - Presentations on 10Apr (2 weeks from now!)
 - Remember your potential clients:
what questions would they like answered?
 - Tell a story/narrative in your presentation
- Email me your preferences for presentation slot