

# 12.2-3: Factorial ANOVA and Blocking

15 Nov 2011  
BUSI275  
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- **HW8** due next Tue
- Please download:  
19-Eukanuba.xls  
19-Applebees.xls

# Outline for today

- Factorial ANOVA (multiple nominal IVs)
  - Assumptions
  - Graphing
  - Model and calculations
  - Main effects
  - Interaction / moderation
- Randomized Complete Block ANOVA
  - Fixed vs. random effects
  - Model and calculations
  - Post-hoc analysis: Fisher's LSD

# Factorial ANOVA

DV: Purch Amt

IV1: Gender

- 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 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

IV2:  
Src

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

# Assumptions

DV: Purch Amt

IV1: Gender

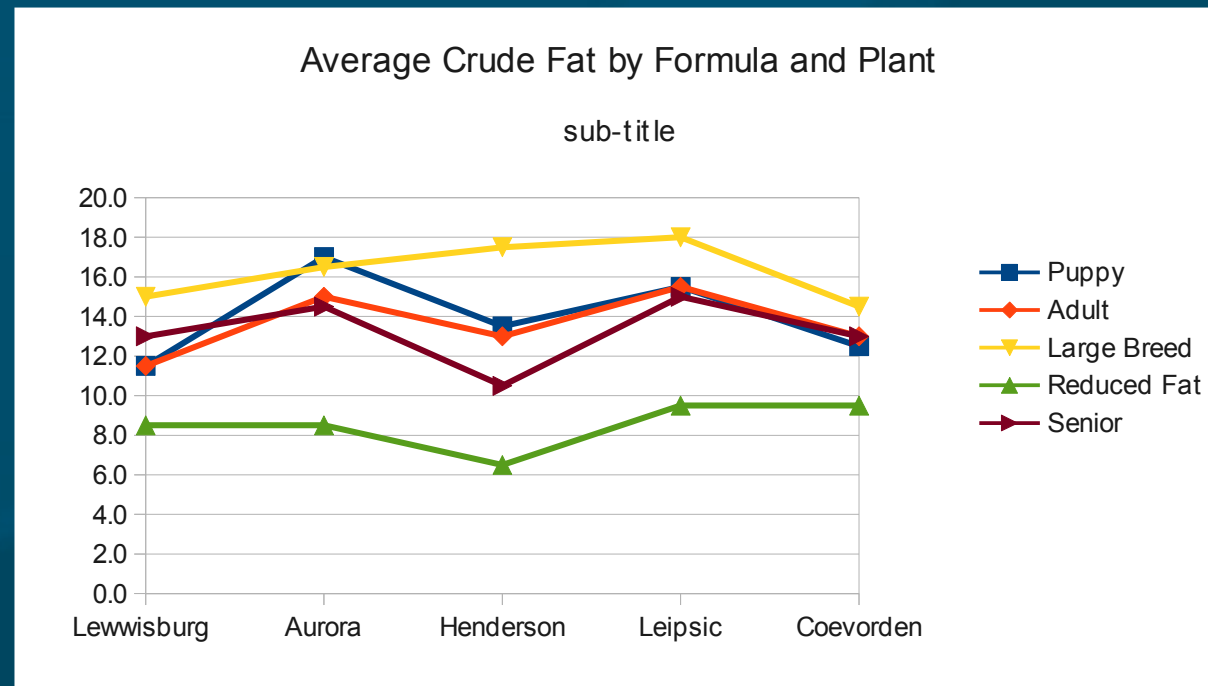
\$40 \$20 \$25 \$32	\$17 \$21 \$19 \$22
\$30 \$26 \$25	\$22 \$19
\$40 \$45 \$52 \$43 \$39 \$48	\$50 \$60 \$55

IV2:  
Src

- Same as for regular ANOVA, per cell:
  - DV continuous
  - Independent observations, independent cells (groups)
  - DV normal within each cell
  - Variance of DV similar across all cells:
    - ◆  $(\text{largest SD}) / (\text{smallest SD}) < 2$
- 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: 19-Eukanuba.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:
  - $\text{Purchase} = b_0 + (\text{Gender effect}) + (\text{Src effect}) + (\text{Gender*Source effect}) + \varepsilon$
- Decomposition of variance:
  - $SS_{\text{tot}} = SS_{\text{Gen}} + SS_{\text{Src}} + SS_{\text{Gen*Src}} + SS_{\text{resid}}$
- Global F-test looks for any effect of IVs on DV
  - If not significant, check for violations of assumptions
- Effect size  $\eta^2$  is akin to  $R^2$ :  $1 - (SS_{\text{resid}} / SS_{\text{tot}})$

# 2-way ANOVA: calculating

-	IV <sub>1</sub> (a levels)	IV <sub>2</sub> (b levels)	IV <sub>1</sub> *IV <sub>2</sub> (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  $SS_{res}$ 
  - $df_{tot} = n - 1$ , and  $df_{res} = n - ab$
  - The SS and df always add up:
    - ◆  $Tot = IV_1 + IV_2 + (IV_1 * IV_2) + Resid$
- 3 F-tests: IV<sub>1</sub>, IV<sub>2</sub>, and interaction
  - e.g., main effect on IV<sub>1</sub>:  $F = MS_1 / MS_{res}$

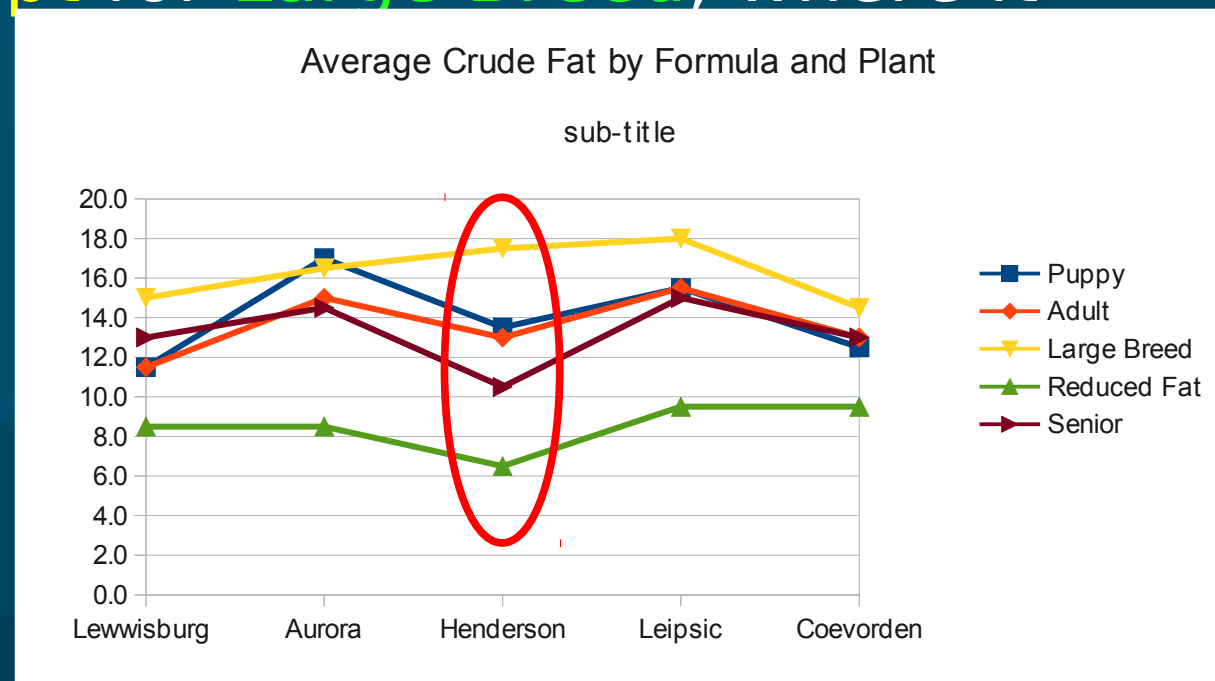
# Main effects

- 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



# 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, 19-Applebees.xls:
  - **Factor**: Restaurant
  - **Block**: Week
  - **DV**: Revenue

	<b>Restaurant</b>		
<b>Week</b>	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

# TODO

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- HW8 (ch15,12): due next Tues
- Projects:
  - Presentations in two weeks!
  - If you don't know what analysis to perform, or how to perform it, ask me for help