

When no interaction is present
the additive model is a

5. Analysis of Two-Factor Experiments Based on Additive Models

Simple & convenient way to
model the data

An Example Two-Factor Experiment

Researchers were interested in studying the effects of 2 diets (low fiber, high fiber) and 3 drugs (D1, D2, D3) on weight gained by Yorkshire hogs. A total of 12 hogs were assigned to the 6 diet-drug combinations using a balanced and completely randomized experimental design. Hogs were housed in individual pens, injected with their assigned drugs once per week, and fed their assigned diets for a 6-week period. The amount of weight gained during the 6-week period was recorded for each hog.

As discussed in the previous set of slides, this experiment involves the two factors: Diet (with levels low fiber and high fiber) and Drug (with levels D1, D2, and D3).

The Additive Model

When factors do not interact, it makes sense to consider the *additive* model:

$$y_{ijk} = \mu + \alpha_i + \beta_j + \epsilon_{ijk} \quad (i = 1, 2; j = 1, 2, 3; k = 1, 2) \quad \text{where}$$

$\mu, \alpha_1, \alpha_2, \beta_1, \beta_2, \beta_3$ are unknown real-valued parameters and

$$\epsilon_{111}, \epsilon_{112}, \epsilon_{121}, \epsilon_{122}, \epsilon_{131}, \epsilon_{132}, \epsilon_{211}, \epsilon_{212}, \epsilon_{221}, \epsilon_{222}, \epsilon_{231}, \epsilon_{232}$$

$$\stackrel{i.i.d.}{\sim} \mathcal{N}(0, \sigma^2)$$

for some unknown $\sigma^2 > 0$.

Table of Treatments and Means for the Additive Model

Treatment	Diet	Drug	Mean
1	1	1	$\mu + \alpha_1 + \beta_1$
2	1	2	$\mu + \alpha_1 + \beta_2$
3	1	3	$\mu + \alpha_1 + \beta_3$
4	2	1	$\mu + \alpha_2 + \beta_1$
5	2	2	$\mu + \alpha_2 + \beta_2$
6	2	3	$\mu + \alpha_2 + \beta_3$

Diet 1 = Low Fiber, Diet 2 = High Fiber

Drug 1 = D1, Drug 2 = D2, Drug 3 = D3

Additive Model in Matrix and Vector Form

$$\begin{bmatrix} y_{111} \\ y_{112} \\ y_{121} \\ y_{122} \\ y_{131} \\ y_{132} \\ y_{211} \\ y_{212} \\ y_{221} \\ y_{222} \\ y_{231} \\ y_{232} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu \\ \alpha_1 \\ \alpha_2 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} + \begin{bmatrix} \epsilon_{111} \\ \epsilon_{112} \\ \epsilon_{121} \\ \epsilon_{122} \\ \epsilon_{131} \\ \epsilon_{132} \\ \epsilon_{211} \\ \epsilon_{212} \\ \epsilon_{221} \\ \epsilon_{222} \\ \epsilon_{231} \\ \epsilon_{232} \end{bmatrix}$$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}, \quad \boldsymbol{\epsilon} \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I})$$

Additive Model in Matrix and Vector Form

$$\hat{y}_{ijk} = \hat{\mu} + \hat{\alpha}_i + \hat{\beta}_j$$

$$\hat{y}_{ijk} = \hat{\mu} + \hat{\alpha}_i + \hat{\beta}_j$$

$$\begin{bmatrix} y_{111} \\ y_{112} \\ y_{121} \\ y_{122} \\ y_{131} \\ y_{132} \\ y_{211} \\ y_{212} \\ y_{221} \\ y_{222} \\ y_{231} \\ y_{232} \end{bmatrix} = \begin{bmatrix} \mu + \alpha_1 + \beta_1 \\ \mu + \alpha_1 + \beta_1 \\ \mu + \alpha_1 + \beta_2 \\ \mu + \alpha_1 + \beta_2 \\ \mu + \alpha_1 + \beta_3 \\ \mu + \alpha_1 + \beta_3 \\ \mu + \alpha_2 + \beta_1 \\ \mu + \alpha_2 + \beta_1 \\ \mu + \alpha_2 + \beta_2 \\ \mu + \alpha_2 + \beta_2 \\ \mu + \alpha_2 + \beta_3 \\ \mu + \alpha_2 + \beta_3 \end{bmatrix} + \begin{bmatrix} \epsilon_{111} \\ \epsilon_{112} \\ \epsilon_{121} \\ \epsilon_{122} \\ \epsilon_{131} \\ \epsilon_{132} \\ \epsilon_{211} \\ \epsilon_{212} \\ \epsilon_{221} \\ \epsilon_{222} \\ \epsilon_{231} \\ \epsilon_{232} \end{bmatrix}$$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}, \quad \boldsymbol{\epsilon} \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I})$$

Table of Cell Means for the Additive Model

- All Interactions are Zero for the Additive Model.
- The simple effect of Diet is $\alpha_1 - \alpha_2$ for all levels of Drug.
- The simple effect of Drug j vs. Drug j' is $\beta_j - \beta_{j'}$ regardless of Diet.

marginal mean difference for diet :
 $\bar{\mu}_{1.} - \bar{\mu}_{2.}$

	Drug 1	Drug 2	Drug 3
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$

$= \alpha_1 - \alpha_2$
for estimating
 α_1 & α_2 we
will use all

available data (across all 3 levels of drug)

Marginal Means for the Additive Model

difference is $\alpha_1 - \alpha_2$

	Drug 1	Drug 2	Drug 3	
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$	$\mu + \alpha_1 + \bar{\beta}.$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \bar{\beta}.$
	$\mu + \bar{\alpha} + \beta_1$	$\mu + \bar{\alpha} + \beta_2$	$\mu + \bar{\alpha} + \beta_3$	$\mu + \bar{\alpha} + \bar{\beta}.$

$\beta_1 - \beta_2$ $\beta_2 - \beta_3$

averaging over diet

averaging over drug

Simple Effects = Main Effects for the Additive Model

The main effect of Diet is

$$(\mu + \alpha_1 + \bar{\beta}.) - (\mu + \alpha_2 + \bar{\beta}.) = \alpha_1 - \alpha_2$$

	Drug 1	Drug 2	Drug 3	
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$	$\mu + \alpha_1 + \bar{\beta}.$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \bar{\beta}.$
	$\mu + \bar{\alpha}. + \beta_1$	$\mu + \bar{\alpha}. + \beta_2$	$\mu + \bar{\alpha}. + \beta_3$	$\mu + \bar{\alpha}. + \bar{\beta}.$

Simple Effects = Main Effects for the Additive Model

The main effect for Drug 1 vs. Drug 2 is

$$(\mu + \bar{\alpha}_{.} + \beta_1) - (\mu + \bar{\alpha}_{.} + \beta_2) = \beta_1 - \beta_2$$

	Drug 1	Drug 2	Drug 3	
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$	$\mu + \alpha_1 + \bar{\beta}_{.}$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \bar{\beta}_{.}$
	$\mu + \bar{\alpha}_{.} + \beta_1$	$\mu + \bar{\alpha}_{.} + \beta_2$	$\mu + \bar{\alpha}_{.} + \beta_3$	$\mu + \bar{\alpha}_{.} + \bar{\beta}_{.}$

Tests for Main Effects in the Additive Model

No Diet main effect $\iff \underline{\alpha_1 = \alpha_2}$

No Drug main effects $\iff \underline{\beta_1 = \beta_2 = \beta_3}$

	Drug 1	Drug 2	Drug 3	
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$	$\mu + \alpha_1 + \bar{\beta}.$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \bar{\beta}.$
	$\mu + \bar{\alpha}_. + \beta_1$	$\mu + \bar{\alpha}_. + \beta_2$	$\mu + \bar{\alpha}_. + \beta_3$	$\mu + \bar{\alpha}_. + \bar{\beta}.$

H_0 : No Diet Main Effect ($\alpha_1 = \alpha_2$)

$$\begin{array}{c} \alpha_1 - \alpha_2 \\ | \\ [0 \quad 1 \quad -1 \quad 0 \quad 0 \quad 0] \end{array} \begin{bmatrix} \mu \\ \alpha_1 \\ \alpha_2 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = [0]$$

H_0 : No Drug Main Effects ($\beta_1 = \beta_2 = \beta_3$)

$$C = \begin{pmatrix} C_1^T \\ C_2^T \end{pmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & \underbrace{1 \quad -1}_{\beta_2 - \beta_3} \end{bmatrix} \begin{bmatrix} \mu \\ \alpha_1 \\ \alpha_2 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

LSMEANS for the Additive Model

LSMEANS are OLS estimators of the quantities in the margins below.

	Drug 1	Drug 2	Drug 3	
Diet 1	$\mu + \alpha_1 + \beta_1$	$\mu + \alpha_1 + \beta_2$	$\mu + \alpha_1 + \beta_3$	$\mu + \alpha_1 + \bar{\beta}.$
Diet 2	$\mu + \alpha_2 + \beta_1$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \bar{\beta}.$
	$\mu + \bar{\alpha} . + \beta_1$	$\mu + \bar{\alpha} . + \beta_2$	$\mu + \bar{\alpha} . + \beta_3$	$\mu + \bar{\alpha} . + \bar{\beta}.$

LSMEANS for the Additive Model (continued)

For example, the LSMEAN for Diet 1 is

$$c^T \hat{\beta} = \begin{bmatrix} 1, 1, 0, \frac{1}{3}, \frac{1}{3}, \frac{1}{3} \end{bmatrix} \begin{bmatrix} \hat{\mu} \\ \hat{\alpha}_1 \\ \hat{\alpha}_2 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \\ \hat{\beta}_3 \end{bmatrix} = \underline{\hat{\mu}} + \underline{\hat{\alpha}_1} + \boxed{\frac{\hat{\beta}_1 + \hat{\beta}_2 + \hat{\beta}_3}{3}},$$

estimate of
marginal mean $\bar{Y}_{1.}$

where $\hat{\beta}$ is any solution to the Normal Equations.

Although $\hat{\beta}$ will depend on which of infinitely many solutions to the Normal Equations is used, $c^T \hat{\beta}$ will be the same for all solutions.

SAS Code for the Additive Model

```
proc import datafile='C:\dietdrug.txt'  
    dbms=TAB replace out=d;  
run;
```

```
proc mixed;  
    class diet drug;  
    model weightgain=diet drug;  
    lsmeans diet drug / cl;  
    estimate 'diet effect' diet 1 -1 / cl;  
    estimate 'drug 1 - drug 2' drug 1 -1 0;  
    estimate 'drug 1 - drug 3' drug 1 0 -1;  
    estimate 'drug 2 - drug 3' drug 0 1 -1;  
    contrast 'drug main effects' drug 1 -1 0  
                                                drug 1 0 -1;  
run;
```

confidence
limits

use column
to
separate
several
comparisons

Fitting the Additive Model in R

```
d=read.delim("http://.../S510/dietdrug.txt")
```

```
d$diet=factor(d$diet)
```

```
d$drug=factor(d$drug)
```

```
o=lm(weightgain~diet+drug, data=d)
```

coef(o) is $\hat{\beta}_{\mathbf{R}}$

vcov(o) is $\widehat{\text{Var}}(\hat{\beta}_{\mathbf{R}}) = \hat{\sigma}^2(\mathbf{X}_{\mathbf{R}}^{\top} \mathbf{X}_{\mathbf{R}})^{-1}$.

o\$df is $n - r$.

this model
matrix is full
rank
(the $\mathbf{X}_{\mathbf{R}}$ sets up
internally)

The Additive Model Matrix is Not Full-Column Rank

$$\begin{bmatrix} y_{111} \\ y_{112} \\ y_{121} \\ y_{122} \\ y_{131} \\ y_{132} \\ y_{211} \\ y_{212} \\ y_{221} \\ y_{222} \\ y_{231} \\ y_{232} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu \\ \alpha_1 \\ \alpha_2 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} + \begin{bmatrix} \epsilon_{111} \\ \epsilon_{112} \\ \epsilon_{121} \\ \epsilon_{122} \\ \epsilon_{131} \\ \epsilon_{132} \\ \epsilon_{211} \\ \epsilon_{212} \\ \epsilon_{221} \\ \epsilon_{222} \\ \epsilon_{231} \\ \epsilon_{232} \end{bmatrix}$$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}, \quad \boldsymbol{\epsilon} \sim N(\mathbf{0}, \sigma^2 \mathbf{I})$$

The Full-Rank Formulation Used by R

$$\begin{bmatrix} y_{111} \\ y_{112} \\ y_{121} \\ y_{122} \\ y_{131} \\ y_{132} \\ y_{211} \\ y_{212} \\ y_{221} \\ y_{222} \\ y_{231} \\ y_{232} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu \\ \alpha_2 \\ \beta_2 \\ \beta_3 \end{bmatrix} + \begin{bmatrix} \epsilon_{111} \\ \epsilon_{112} \\ \epsilon_{121} \\ \epsilon_{122} \\ \epsilon_{131} \\ \epsilon_{132} \\ \epsilon_{211} \\ \epsilon_{212} \\ \epsilon_{221} \\ \epsilon_{222} \\ \epsilon_{231} \\ \epsilon_{232} \end{bmatrix}$$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}, \quad \boldsymbol{\epsilon} \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I})$$

The Full-Rank Formulation Used by R

$$\begin{bmatrix} y_{111} \\ y_{112} \\ y_{121} \\ y_{122} \\ y_{131} \\ y_{132} \\ y_{211} \\ y_{212} \\ y_{221} \\ y_{222} \\ y_{231} \\ y_{232} \end{bmatrix} = \begin{bmatrix} \mu \\ \mu \\ \mu + \beta_2 \\ \mu + \beta_2 \\ \mu + \beta_3 \\ \mu + \beta_3 \\ \mu + \alpha_2 \\ \mu + \alpha_2 \\ \mu + \alpha_2 + \beta_2 \\ \mu + \alpha_2 + \beta_2 \\ \mu + \alpha_2 + \beta_3 \\ \mu + \alpha_2 + \beta_3 \end{bmatrix} + \begin{bmatrix} \epsilon_{111} \\ \epsilon_{112} \\ \epsilon_{121} \\ \epsilon_{122} \\ \epsilon_{131} \\ \epsilon_{132} \\ \epsilon_{211} \\ \epsilon_{212} \\ \epsilon_{221} \\ \epsilon_{222} \\ \epsilon_{231} \\ \epsilon_{232} \end{bmatrix}$$

mean diet & drug1

β_2 reflects the difference between \bar{y}_{ijk} when going from drug 1 to drug 2

$$y = X\beta + \epsilon, \quad \epsilon \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I})$$

Table of Means for the R Full-Rank Formulation

	Drug 1	Drug 2	Drug 3	
Diet 1	μ	$\mu + \beta_2$	$\mu + \beta_3$	$\mu + \frac{\beta_2 + \beta_3}{3}$
Diet 2	$\mu + \alpha_2$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \frac{\beta_2 + \beta_3}{3}$
	$\mu + \frac{\alpha_2}{2}$	$\mu + \frac{\alpha_2}{2} + \beta_2$	$\mu + \frac{\alpha_2}{2} + \beta_3$	$\mu + \frac{\alpha_2}{2} + \frac{\beta_2 + \beta_3}{3}$

Main Effects in the R Additive Model Formulation

No Diet main effect $\iff \alpha_2 = 0$

this differs from the set-up using the less-than-2-12-25

No Drug main effects $\iff \beta_2 = \beta_3 = 0$

full rank model matrix

	Drug 1	Drug 2	Drug 3	
Diet 1	μ	$\mu + \beta_2$	$\mu + \beta_3$	$\mu + \frac{\beta_2 + \beta_3}{3}$
Diet 2	$\mu + \alpha_2$	$\mu + \alpha_2 + \beta_2$	$\mu + \alpha_2 + \beta_3$	$\mu + \alpha_2 + \frac{\beta_2 + \beta_3}{3}$
	$\mu + \frac{\alpha_2}{2}$	$\mu + \frac{\alpha_2}{2} + \beta_2$	$\mu + \frac{\alpha_2}{2} + \beta_3$	$\mu + \frac{\alpha_2}{2} + \frac{\beta_2 + \beta_3}{3}$

all the same when $\beta_2 = \beta_3 = 0$