

BIO 226, Spring 2015: Lab 1

Contrasts in SAS

Yoonyoung Park

February 3, 2015

Outline

- Guinea pigs ANOVA example
- How to change the reference level for class variables.
- Writing hypotheses in terms of β 's.
- Writing contrasts.
- Testing contrasts in SAS.

Example: Dosages of four cardiotoxic drugs at death of infused guinea pigs



- Evaluating potencies of four cardiac treatments
- Observe dosage at which animals (guinea pigs) die for each treatment
- 10 guinea pigs per treatment (40 observations in all)

Example: Dosages of four cardiotoxic drugs at death of infused guinea pigs



- Assess any differences in toxicity of four treatments, i.e. differences in mean dosage required to kill animal

$$\bar{y}_1 = 25.9, \bar{y}_2 = 22.2, \bar{y}_3 = 20.0, \bar{y}_4 = 19.6$$

SAS Syntax for One-Way ANOVA

```
DATA toxic;  
INFILE 'tox.txt';  
INPUT y drug;  
RUN;  
  
PROC GLM DATA=toxic;  
CLASS drug;  
MODEL y=drug / solution;  
RUN;
```

SAS Syntax for One-Way ANOVA

Parameter		Estimate	Standard Error	t Value	Pr > t
Intercept		19.60000000 B	0.98728021	19.85	<.0001
DRUG	1	6.30000000 B	1.39622507	4.51	<.0001
DRUG	2	2.60000000 B	1.39622507	1.86	0.0708
DRUG	3	0.40000000 B	1.39622507	0.29	0.7761
DRUG	4	0.00000000 B	.	.	.

Changing the reference level

```
proc sort data=toxic; by descending drug; run;
```

```
PROC GLM DATA=toxic order=data;
```

```
CLASS drug;
```

```
MODEL y=drug / solution;
```

```
RUN;
```

Recall from Lecture: SAS Syntax for One-Way ANOVA

The GLM Procedure

Class Level Information

Class	Levels	Values
DRUG	4	1 2 3 4

Dependent Variable: Y

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	249.8750000	83.2916667	8.55	0.0002
Error	36	350.9000000	9.7472222		
Corrected Total	39	600.7750000			

R-Square	Coeff Var	Root MSE	Y Mean
0.415921	14.23970	3.122054	21.92500

Recall from Lecture: SAS Syntax for One-Way ANOVA

Parameter		Estimate	Standard Error	t Value	Pr > t
Intercept		25.90000000 B	0.98728021	26.23	<.0001
DRUG	4	-6.30000000 B	1.39622507	-4.51	<.0001
DRUG	3	-5.90000000 B	1.39622507	-4.23	0.0002
DRUG	2	-3.70000000 B	1.39622507	-2.65	0.0119
DRUG	1	0.00000000 B	.	.	.

The model

$$Y_i = \beta_1 + \beta_2 X_{i2} + \beta_3 X_{i3} + \beta_4 X_{i4} + \epsilon_i$$

$$X_{i2} = \begin{cases} 1 & \text{DRUG 2} \\ 0 & \text{otherwise} \end{cases}$$

$$X_{i3} = \begin{cases} 1 & \text{DRUG 3} \\ 0 & \text{otherwise} \end{cases}$$

$$X_{i4} = \begin{cases} 1 & \text{DRUG 4} \\ 0 & \text{otherwise} \end{cases}$$

The model

$$Y_i = \beta_1 + \beta_2 X_{i2} + \beta_3 X_{i3} + \beta_4 X_{i4} + \epsilon_i$$

$$\mu_1 = \beta_1$$

$$\mu_2 = \beta_1 + \beta_2$$

$$\mu_3 = \beta_1 + \beta_3$$

$$\mu_4 = \beta_1 + \beta_4$$

Hypotheses to test

H_0 : Drug 2 is as potent as drug 3.

How would we write that hypothesis in terms of the β 's?

$$\mu_1 = \beta_1$$

$$\mu_2 = \beta_1 + \beta_2$$

$$\mu_3 = \beta_1 + \beta_3$$

$$\mu_4 = \beta_1 + \beta_4$$

Hypotheses to test

H_0 : Drug 2 is as potent as drug 3.

How would we write that hypothesis in terms of the β 's?

$$\beta_2 = \beta_3$$

$$\mu_1 = \beta_1$$

$$\mu_2 = \beta_1 + \beta_2$$

$$\mu_3 = \beta_1 + \beta_3$$

$$\mu_4 = \beta_1 + \beta_4$$

Hypotheses to test

H_0 : Drug 2 is as potent as drug 3.

$$H_0: \beta_2 = \beta_3$$

$$H_0: \beta_2 - \beta_3 = 0$$

$$\begin{bmatrix} 0 & 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix} = 0 * \beta_1 + 1 * \beta_2 + -1 * \beta_3 + 0 * \beta_4$$
$$= \beta_2 - \beta_3$$

Hypotheses to test

H_0 : Drug 2 is as potent as drug 3.

$$H_0: \beta_2 = \beta_3$$

$$H_0: \beta_2 - \beta_3 = 0$$

$$\underbrace{\begin{bmatrix} 0 & 1 & -1 & 0 \end{bmatrix}}_{\mathbf{L}} \underbrace{\begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix}}_{\boldsymbol{\beta}} = 0 * \beta_1 + 1 * \beta_2 + -1 * \beta_3 + 0 * \beta_4$$
$$= \beta_2 - \beta_3$$

$$H_0: \mathbf{L}\boldsymbol{\beta} = 0$$

Hypotheses to test

H_0 : Drugs 2, 3 and 4 are equally potent.

How would we write that hypothesis in terms of the β 's?

$$\mu_1 = \beta_1$$

$$\mu_2 = \beta_1 + \beta_2$$

$$\mu_3 = \beta_1 + \beta_3$$

$$\mu_4 = \beta_1 + \beta_4$$

Hypotheses to test

H_0 : Drugs 2, 3 and 4 are equally potent.

How would we write that hypothesis in terms of the β 's?

$$\beta_2 = \beta_3 = \beta_4$$

$$\mu_1 = \beta_1$$

$$\mu_2 = \beta_1 + \beta_2$$

$$\mu_3 = \beta_1 + \beta_3$$

$$\mu_4 = \beta_1 + \beta_4$$

Hypotheses to test

H_0 : Drugs 2, 3 and 4 are equally potent.

$H_0: \beta_2 = \beta_3 = \beta_4$

$H_0: \beta_2 = \beta_3$ and $\beta_3 = \beta_4$

$H_0: \beta_2 - \beta_3 = 0$ and $\beta_3 - \beta_4 = 0$

$$\underbrace{\begin{bmatrix} 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & -1 \end{bmatrix}}_{\mathbf{L}} \underbrace{\begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix}}_{\boldsymbol{\beta}} = \begin{bmatrix} 0 * \beta_1 + 1 * \beta_2 + -1 * \beta_3 + 0 * \beta_4 \\ 0 * \beta_1 + 0 * \beta_2 + 1 * \beta_3 + -1 * \beta_4 \end{bmatrix}$$
$$= \begin{bmatrix} \beta_2 - \beta_3 \\ \beta_3 - \beta_4 \end{bmatrix}$$

$H_0: \mathbf{L}\boldsymbol{\beta} = \mathbf{0}$

Hypotheses to test

H_0 : All 4 drugs are equally potent.

H_0 : $\beta_2 = \beta_3 = \beta_4 = 0$

H_0 : $\beta_2 = 0$ and $\beta_3 = 0$ and $\beta_4 = 0$

$$\underbrace{\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}}_{\mathbf{L}} \underbrace{\begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix}}_{\boldsymbol{\beta}} = \begin{bmatrix} 0 * \beta_1 + 1 * \beta_2 + 0 * \beta_3 + 0 * \beta_4 \\ 0 * \beta_1 + 0 * \beta_2 + 1 * \beta_3 + 0 * \beta_4 \\ 0 * \beta_1 + 0 * \beta_2 + 0 * \beta_3 + 1 * \beta_4 \end{bmatrix}$$
$$= \begin{bmatrix} \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix}$$

H_0 : $\mathbf{L}\boldsymbol{\beta} = 0$

How to test these Hypotheses

We know that the least squares (LS) estimate $\hat{\beta}$, which is also the maximum likelihood (ML) estimator, has distribution

$$\hat{\beta} \sim N(\beta, \text{Cov}(\hat{\beta}))$$

(see Lecture 2 slide 22)

then

$$\mathbf{L}\hat{\beta} \sim N(\mathbf{L}\beta, \mathbf{L}\text{Cov}(\hat{\beta})\mathbf{L}')$$

Thus, we can perform inference to test $H_0: \mathbf{L}\beta = 0$.

Luckily, SAS does this for us!

Contrast Statement

CONTRAST 'label' effect values / E;

label - identifies the contrast on the output.

effect - the variable in the model

values - the \mathbf{L} matrix

E - displays the \mathbf{L} matrix. This option is useful in confirming the ordering of parameters for specifying.

H_0 : Drug 2 is as potent as drug 3.

$$H_0: \beta_2 - \beta_3 = 0$$

```
PROC GLM DATA=toxic order=data;
```

```
CLASS drug;
```

```
MODEL y=drug / solution;
```

```
CONTRAST 'Drug 2 vs Drug 3' DRUG 0 -1 1 0 / E;
```

```
RUN;
```

Contrast Statement

H_0 : Drug 2 is as potent as drug 3.

$$H_0: \beta_2 - \beta_3 = 0$$

E - displays the **L** matrix. This option is useful in confirming the ordering of parameters for specifying .

Coefficients for Contrast Drug 2 vs Drug 3

Row 1		
Intercept		0
DRUG	4	0
DRUG	3	-1
DRUG	2	1
DRUG	1	0

Contrast Statement

H_0 : Drug 2 is as potent as drug 3.

Contrast	DF	Contrast SS	Mean Square	F Value	Pr > F
Drug 2 vs Drug 3	1	24.20000000	24.20000000	2.48	0.1238

Contrast Statement

H_0 : Drugs 2, 3 and 4 are equally potent.

```
PROC GLM DATA=toxic order=data;  
CLASS drug;  
MODEL y=drug / solution;  
CONTRAST 'Drug 2 vs Drug 3 vs Drug 4' DRUG 0 -1 1 0,  
        DRUG -1 1 0 0 / E;  
RUN;
```


Contrast Statement

H_0 : Drugs 2, 3 and 4 are equally potent.

E - displays the displays the **L** matrix. This option is useful in confirming the ordering of parameters for specifying .

Coefficients for Contrast Drug 2 vs Drug 3 vs Drug 4

		Row 1	Row 2
Intercept		0	0
DRUG	4	0	-1
DRUG	3	-1	1
DRUG	2	1	0
DRUG	1	0	0

Contrast Statement

H_0 : Drugs 2, 3 and 4 are equally potent.

Contrast	DF	Contrast SS	Mean Square	F Value	Pr > F
Drug 2 vs Drug 3 vs Drug 4	2	39.2000	19.600000	2.01	0.1486