

6

Combining continuous and categorical variables

6.2 Combining continuous and categorical variables

To combine continuous and categorical variables, the word equation and continuous variables are specified as before (using the covariate subcommand or the covariates sub-dialog box). Minitab will assume that all variables not specified as continuous are categorical. This is illustrated below with the *leprosy* dataset.

MINITAB COMMANDS FOR BOX 6.1 A treatment for leprosy

Commands `glm BACAFTER = BACBEF + TREATMT;`
 `covariate BACBEF;`
 `brief 3.`

Menu route Stat > ANOVA > General Linear Model
 BACAFTER → Response
 BACBEF + TREATMT → Model

Covariates...

BACBEF → Covariates

Results...

⊙ In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.1 A treatment for leprosy

General Linear Model: BACAFTER versus TREATMT

Factor	Type	Levels	Values
TREATMT	fixed	3	1 2 3

Analysis of Variance for BACAFTER, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
BACBEF	1	587.48	515.01	515.01	35.22	0.000
TREATMT	2	83.35	83.35	41.67	2.85	0.076
Error	26	380.15	380.15	14.62		
Total	29	1050.98				

Term	Coef	SE Coef	T	P
Constant	-0.013	1.806	-0.01	0.994
BACBEF	0.8831	0.1488	5.93	0.000
TREATMT				
1	-1.590	1.012	-1.57	0.128
2	-0.726	1.002	-0.72	0.475

Unusual Observations for BACAFTER

Obs	BACAFTER	Fit	SE Fit	Residual	St Resid
13	16.0530	6.6802	1.3434	9.3728	2.62R

R denotes an observation with a large standardized residual

With the *fats* dataset, combining continuous and categorical variables can reveal hitherto hidden relationships, as discussed in the main text.

MINITAB COMMANDS FOR BOX 6.2 Both weight and sex as explanatory variables for fat content

Commands `glm FAT = WEIGHT + SEX;`
 `covariate WEIGHT;`
 `brief 3.`

Menu route Stat > ANOVA > General Linear Model
 FAT → Response
 WEIGHT + SEX → Model

Covariates...

FAT → Covariates

Results...

Ⓞ In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.2 Both weight and sex as explanatory variables for fat content

General Linear Model: FAT versus SEX

Factor	Type	Levels	Values
SEX	fixed	2	1 2

Analysis of Variance for FAT, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
WEIGHT	1	1.328	87.105	87.105	34.00	0.000
SEX	1	176.098	176.098	176.098	68.73	0.000
Error	16	40.995	40.995	2.562		
Total	18	218.421				

Term	Coef	SE Coef	T	P
Constant	13.010	2.679	4.86	0.000
WEIGHT	0.21715	0.03724	5.83	0.000
SEX				
1	3.9519	0.4767	8.29	0.000

Unusual Observations for FAT

Obs	FAT	Fit	SE Fit	Residual	St Resid
13	35.0000	31.9454	0.5769	3.0546	2.05R
16	26.0000	29.7740	0.5551	-3.7740	-2.51R

R denotes an observation with a large standardized residual

6.3 Orthogonality in the context of continuous and categorical variables

This example illustrates orthogonality between a continuous and categorical explanatory variable using the *bacterial growth* dataset.

MINITAB COMMANDS FOR BOX 6.3 Bacterial growth at two levels of lactose

Commands `glm BACTERIA = DAY + LACTOSE;`
 `covariate DAY;`
 `brief 3.`

Menu route Stat > ANOVA > General Linear Model
 BACTERIA → Response
 DAY + LACTOSE → Model

DAY → Covariates

⊙ In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.3 Bacterial growth at two levels of lactose						
General Linear Model: BACTERIA versus LACTOSE						
Factor	Type	Levels	Values			
LACTOSE	fixed	2	1 2			
Analysis of Variance for BACTERIA, using Adjusted SS for Tests						
Source	DF	Seq SS	Adj SS	Adj MS	F	P
DAY	1	297.97	297.97	297.97	130.56	0.000
LACTOSE	1	397.07	397.07	397.07	173.99	0.000
Error	17	38.80	38.80	2.28		
Total	19	733.85				
Term		Coef	SE Coef	T		P
Constant		3.0939	0.7922	3.91		0.001
DAY		2.7293	0.2389	11.43		0.000
LACTOSE						
1		-4.4557	0.3378	-13.19		0.000
Unusual Observations for BACTERIA						
Obs	BACTERIA	Fit	SE Fit	Residual		St Resid
15	24.1410	21.1964	0.6756	2.9446		2.18R
R denotes an observation with a large standardized residual						

Orthogonality may then be demonstrated more directly, by using DAY as the response variable.

MINITAB COMMANDS FOR BOX 6.4 Illustrating orthogonality between continuous and categorical variables	
Commands	glm DAY = LACTOSE; brief 3.
Menu route	Stat > ANOVA > General Linear Model DAY → Response LACTOSE → Model
	<input type="button" value="Results..."/>
	⊙ In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.4 Illustrating orthogonality between continuous and categorical variables
General Linear Model: DAY versus LACTOSE

Factor	Type	Levels	Values
LACTOSE	fixed	2	1 2

Analysis of Variance for DAY, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
LACTOSE	1	0.000	0.000	0.000	0.00	1.000
Error	18	40.000	40.000	2.222		
Total	19	40.000				

Term	Coef	SE Coef	T	P
Constant	3.0000	0.3333	9.00	0.000
LACTOSE 1	-0.0000	0.3333	-0.00	1.000

6.4 Treating variables as continuous or categorical

Some variables may be legitimately treated as continuous or categorical. DAY in the *bacterial growth* dataset is an example of this.

MINITAB COMMANDS FOR BOX 6.5 Analysing bacterial growth with DAY as a categorical

Commands `glm BACTERIA = DAY + LACTOSE;`
`brief 3.`

Menu route Stat > ANOVA > General Linear Model
 BACTERIA → Response
 DAY + LACTOSE → Model

Results...

⊙ In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.5 Analysing bacterial growth with DAY as a categorical variable
General Linear Model: BACTERIA versus DAY, LACTOSE

Factor	Type	Levels	Values
DAY	fixed	5	1 2 3 4 5
LACTOSE	fixed	2	1 2

Analysis of Variance for BACTERIA, using Adjusted SS for Tests						
Source	DF	Seq SS	Adj SS	Adj MS	F	P
DAY	4	298.51	298.51	74.63	27.30	0.000
LACTOSE	1	397.07	397.07	397.07	145.28	0.000
Error	14	38.26	38.26	2.73		
Total	19	733.85				

Term	Coef	SE Coef	T	P
Constant	11.2820	0.3697	30.52	0.000
DAY				
1	-5.2729	0.7393	-7.13	0.000
2	-2.8055	0.7393	-3.79	0.002
3	-0.2044	0.7393	-0.28	0.786
4	2.6236	0.7393	3.55	0.003
LACTOSE				
1	-4.4557	0.3697	-12.05	0.000

6.7 Exercises

Conservation and its influence on biomass

These data are stored in the *conservation* dataset.

MINITAB COMMANDS FOR BOX 6.7 Conservation and biomass analysis	
Commands	<pre>glm BIOMASS = CONS + ALT + SOIL; covariate ALT; brief 3.</pre>
Menu route	Stat > ANOVA > General Linear Model BIOMASS → Response CONS + ALT + SOIL → Model <div style="border: 1px solid black; padding: 2px; display: inline-block; margin: 5px;">Covariates...</div> ALT → Covariates <div style="border: 1px solid black; padding: 2px; display: inline-block; margin: 5px;">Results...</div> <input checked="" type="radio"/> In addition, coefficients for all terms

MINITAB OUTPUT FOR BOX 6.7 Conservation and biomass analysis			
General Linear Model: BIOMASS versus CONS, SOIL			
Factor	Type	Levels	Values
CONS	fixed	2	1 2
SOIL	fixed	3	1 2 3

Analysis of Variance for BIOMASS, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
CONS	1	0.7176	0.0249	0.0249	2.80	0.101
ALT	1	5.8793	4.4273	4.4273	498.10	0.000
SOIL	2	0.3953	0.3953	0.1977	22.24	0.000
Error	45	0.4000	0.4000	0.0089		
Total	49	7.3922				

Term	Coef	SE Coef	T	P
Constant	2.21156	0.02486	88.97	0.000
CONS				
1	-0.02443	0.01460	-1.67	0.101
ALT	-0.002907	0.000130	-22.32	0.000
SOIL				
1	0.10574	0.02057	5.14	0.000
2	0.01952	0.01889	1.03	0.307

Unusual Observations for BIOMASS

Obs	BIOMASS	Fit	SE Fit	Residual	St Resid
7	0.98000	1.16311	0.03508	-0.18311	-2.09R
18	1.53000	1.34099	0.02896	0.18901	2.11R
23	2.14000	1.94560	0.03599	0.19440	2.23R
50	1.18000	0.98815	0.04317	0.19185	2.29R

R denotes an observation with a large standardized residual

Determinants of the Grade Point Average

See Minitab output for this exercise in the answers to exercises in Chapter 14.