Guide to Computing Minitab commands used in labs (mtbcode.out)

A full listing of Minitab commands can be found by invoking the HELP command while running Minitab. A reference card, with listing of available commands, can be purchased in the University Bookstore.

Use the HELP command (or reference card) much as you would a dictionary or thesaurus, to find out how a command works or find a command to accomplish the calculation at hand.

The commands listed here are for common routines used in this course:

Reading Data into Minitab Writing Data and Output Files from Minitab Summarizing data Re-organizing Data for Analysis by the General Linear Model Executing the General Linear Model Calculating Residuals Using Residuals to Check Assumptions Randomization Tests with Minitab

These routines are built up out of commands, just as sentences are built out of words. Most of the assignments in this course can be accomplished by modifying one of the following routines, depending on the situation or goals of computation.

Many routines have been assigned a name, which is used in other parts of the lab manual. The name of the routine occurs in *boldface italics* to the right of the routine box.

Reading Data into Minitab.

Data are read from a file (in quotes) into columns. For the data files in this course, you will need to state the number of lines of data, because information about the data is listed below the data in the same file.

MTB > read 'srbx9_5.dat' c1 c2; SUBC> nobs = 7. MTB > name c1 'age(I)' c2 'age(II)'

Define Data from file

Data are also typed directly from the keyboard.

MTB > set into c1 DATA> 2.68 2.60 2.43 2.90 2.94 2.70 2.68 2.98 2.85 DATA> end MTB > set c2 DATA> 2.36 2.41 2.39 2.85 2.82 2.73 2.58 2.89 2.78 DATA> end MTB > name c1 'N(B)' c2 'N(13)'

Define Data from keyboard

Writing Data and Output Files from Minitab

 Data in columns can be saved by writing to a named system file, as in the box above.

 MTB > write [to] 'srbx1311.dat' c1 c2
 Define Data to file

 This command prints output directly to the screen.
 Define Data to file

 MTB > print c1 c2
 Output to screen

 Most commands send output to the screen automatically.
 Output to screen

 MTB > describe c1 c2
 MTB > histogram c1 c2

 MTB > let k2 = std(c1)
 Compute

MTB > let k2 = k2*k2 MTB > print k2 a variance Output to screen

Output that appears on the screen can be written to a named system file, this file can then be printed out, or moved to another computer.

MTB > outfile 'srbx9_5.out' MTB > print c1 c2 MTB > describe c1 c2 MTB > nooutfile Outfile open

Outfile closed

Summarizing Data

The following 4 commands calculate and display descriptive statistics on data in column 1

MTB > describe c1 MTB > histogram c1 MTB > dotplot c1 MTB > mean c1

MTB > let k1 = 1

MTB > print k3

MTB > let k2 = 11 MTB > let k3 = k1/k2

This next routine calculates a cumulative relative relative frequency CRF(C1 < 2) = 0.09from the following data: 1 2 3 3 4 4 4 5 5 6 7 MTB > histogram c1; MTB > start 2.

CRF(C1 < 2)

Re-organizing Data for Analysis by the General Linear Model

The following commands reorganize data from tabular format (7 columns) to model format (2 columns, response and explanatory variable).

<--from screen display

<-- from screen display

```
MTB > read 'srtab8_1.dat' c1-c7;
SUBC> nobs = 5.
MTB > stack c1-c7 c8
MTB > name c8 'wlength'
MTB > set c9
DATA> (1 2 3 4 5 6 7)5
DATA> end
MTB > name c9 'groups'
MTB > print c8 c9
```

Reorganize

Executing the General Linear Model, Calculating Residuals

ANOVA designs

MTB > anova 'wlength' = 'groups'; SUBC> fits c10; SUBC> residuals c11. MTB > name c10 'fits' c11 'res'

MTB >	glm 'wlength' = 'groups';
SUBC>	fits c10;
SUBC>	residuals c11.
MTB >	name c10 'fits' c11 'res'

Run GLM ANOVA

Run GLM ANOVA

Regression designs

MTB >	read 'ryder.dat' c1 c2;	
SUBC>	nobs = 23.	
	name c1 'area' c2 'yield'	
MTB >	regress 'yield' 1 predictor 'area';	
SUBC>	residuals c10.	
MTB >	name c10 'res'	
MTB >	<pre>> let k1 = 837382 <from display<="" pre="" screen=""></from></pre>	
MTB >	<pre>let k2 = 1.45 <from display<="" pre="" screen=""></from></pre>	
MTB >	let c11 = k1 + k2*'area'	
MTB >	name c11 'fits'	

Run GLM Regression

MTB	>	read 'ryder.dat' c1 c2;
SUB	C>	nobs = 23.
MTB	>	name c1 'area' c2 'yield'
MTB	>	glm 'yield' = 'area';
SUB	C>	covariate 'area';
SUB	C>	fits c9;
SUB	C>	residuals c10.
MTB	>	name c9 'fits' c10 'res'

Run GLM Regression

Using Residuals to Check Assumptions

A. linear relation of response to explanatory ? (bowls and arches) MTB > plot 'res' vs 'fits'

GLM linear?

B1. Do errors sum to zero?

Note that this assumption is listed for completeness. There is no need to check it if fitted values are calculated via least squares, as in statistical packages such as Minitab.

MTB > let k1 = sum('res') MTB > print k1

B2. Are errors independent?

	let c20 = lag('res') plot c20 'res'	<graphical analysis<="" th=""></graphical>
MTB >	runs 'res'	<runs (optional)<="" test="" th=""></runs>
MTB >	corr c20 'res'	<autocorrelation, 1(optional)<="" lag="" th=""></autocorrelation,>
MTB >	acf 'res'	<autocorrelation, (optional)<="" lags="" many="" th=""></autocorrelation,>

Errors independent?

B3. Are the errors homogeneous ?(no cones facing left or right)

MTB > plot 'res' vs 'fits'

Errors homogeneous?

B4 Are the errors normally distributed ?

MTB > hist 'res' MTB > nscores 'res' c30	<graphical analysis<="" th=""></graphical>
MTB > plot c30 'res'	< 2nd graphical analysis
MTB > rootogram 'res'	<3rd graphical analysis, shows confidence limits

Errors normal ?

Randomization Tests with Minitab

Statistic is mean (compared to zero)	
MTB > let k1 = mean(c1)	Calculate statistic
MTB > nooutfile MTB > set into c2 DATA> -1 1 DATA> end MTB > let k2 = 0	Set up
MTB > store 'ran.ctl'	1
MIB > store 'ran.ctl' STOR> let $k^2 = k^2 + 1$ STOR> sample 12 c2 c3; STOR> replace. STOR> let c4 = c1*c3 STOR> let k3 = mean(c4) - k1 STOR> let c5(k2) = k3 STOR> end	Create control file
	I
MTB > execute 'ran.ctl' MTB > execute 'ran.ctl' 1000	Check file
MED	Run file
MTB > outfile MTB > histogram c5; SUBC> start k1.	CRF(c5 > k1)

p-value is % of distribution greater than statistic in k1

Calculate

statistic

Randomization Tests with Minitab

MTB > let k10 = mean(c6) - mean(c7) MTB > print k10 MTB > let c25(1) = k10 MTB > name c25 'F(st)'

MTB > nooutfile

MTB > store into 'srbx9_5.ctl'
STOR> stack c1 c2 into c3;
STOR> subscripts c4.
STOR> sample 14 times from c3 into c5
STOR> unstack c5 c6 c7;
STOR> subscripts c4.
STOR> let k3 = mean(c6) - mean(c7)
STOR> stack c25 k3 into c25
STOR> end

MTB > execute 'srbx9_5.ctl'
MTB > print 'F(st)'
MTB > execute 'srbx9_5.ctl' 10 times
MTB > print 'F(st)'
MTB > execute 'srbx9 5.ctl' 500

Check file

CRF(c25 > k10)

Run file

MTB	>	outfi	le
MTB	>	hist	'F(st)'
MTB	>	hist	'F(st)';
SUBC	2>	start	k10.

p-value is cumulative relative frequency (%) of outcomes in c25 that are **larger** than value in k10 CRF(c25 > k10) = %

set up

Create control_file

Charl

Randomization Tests with Minitab

Statistic is mean squared error (MSE)

	Statistic is mean squared error (MSE)
Calculate statistic	<pre>MTB > regress 'wloss' on 1 predictor 'humidity'; SUBC> residuals c5. MTB > name c5 'res' MTB > let k10 = std('res')**2 MTB > print k10</pre>
Set up	<pre>MTB > let c25(1) = k10 MTB > name c25 'F(st)' MTB > nooutfile</pre>
Create control file	<pre>MTB > store into 'srbx14_1.ctl' STOR> sample 9 times from 'wloss' into c8 STOR> regress c8 1 'humidity'; SUBC> residuals c5. STOR> let k3 = std('res')**2 STOR> stack c25 k3 into c25 STOR> end</pre>
Check file	<pre>MTB > execute 'srbx14_1.ctl' MTB > print 'F(st)' MTB > execute 'srbx14_1.ctl' 10 times MTB > print 'F(st)' MTB > execute 'srbx14_1.ctl' 500</pre>
Run file	
CRF(c25 < k10)	MTB > outfile MTB > hist 'F(stat)' MTB > hist 'F(stat)'; SUBC> start k10.

p-value is cumulative relative frequency (%) of outcomes in c25 that are **smaller** than value in k10. CRF(c25 < k10) = %