2 A brief description of Stata

Stata is a statistical package for managing, analyzing, and graphing data.

Stata is available for a variety of platforms. Stata may be used either as a point-and-click application or as a command-driven package.

Stata's GUI provides an easy interface for those new to Stata and for experienced Stata users who wish to execute a command that they seldom use.

The command language provides a fast way to communicate with Stata and to communicate more complex ideas.

Here is an extract of a Stata session using the GUI:

(Throughout the Stata manuals, we will refer to various datasets. These datasets are all available from http://www.stata-press.com/data/r13/. For easy access to them within Stata, type webuse dataset_name, or select File > Example Datasets... and click on Stata 13 manual datasets.)

. webuse lbw (Hosmer & Lemeshow data)

We select **Data > Describe data > Summary statistics** and choose to summarize variables low, age, and smoke, whose names we obtained from the Variables window. We click on **OK**.

=		summa	arize - Summary statistics	-	□ ×
Main	by/if/in	Weights			
Variab	oles: (leav	e empty for all	variables)		
low a	ge smoke	•			×
Examp	ples:	yr*	all variables starting with "yr"		
		xyz-abc	all variables between xyz and abc		
– Opti	ons				
	Standard	display			
0	Display ac	lditional statist	ics		
0	No display	; just calculat	e mean		
	Use varial	ble's display fo	rmat		
	5 🜩	Separator line	e every N variables (set 0 for none)		
	Factor-v	variable display	y options		
00			OK Cancel		Submit

. :	ummarize	low	age	smoke
-----	----------	-----	-----	-------

Variable	Obs	Mean	Std. Dev.	Min	Max
low	189	.3121693	.4646093	0	1
age	189	23.2381	5.298678	14	45
smoke	189	.3915344	.4893898	0	1

Stata shows us the command that we could have typed in command mode—summarize low age smoke—before displaying the results of our request.

Next we fit a logistic regression model of low on age and smoke. We select Statistics > Binary outcomes > Logistic regression (reporting odds ratios), fill in the fields, and click on OK.

-8		logis	stic - Logi	stic regre	ssion, re	porting o	odds ra	atios	-		×
Model	by/if/in	Weights	SE/Robust	Reporting	Maximizat	ion					
Deper Iow	ndent varia	able:	Independe age smok Suppre	ent variables: e ss constant l	term					~	
Optio Offs	ons et variable Retain perf	ect predict	or variables								
Con	straints:							.	Mana	ле	
	Keep collin	ear variabl	es (rarely use	d)					Man		
00						OK		Cancel		Subi	mit

. logistic lou	<i>v</i> age smoke						
Logistic regre	ession			Number	of obs	=	189
				LR chi	2(2)	=	7.40
				Prob >	chi2	=	0.0248
Log likelihood	1 = -113.6381	5		Pseudo	R2	=	0.0315
low	Odds Ratio	Std. Err.	z	P> z	[95%	Conf.	Interval]
age	.9514394	.0304194	-1.56	0.119	.8936	482	1.012968
smoke	1.997405	.642777	2.15	0.032	1.063	027	3.753081
_cons	1.062798	.8048781	0.08	0.936	.2408	901	4.689025

Here is an extract of a Stata session using the command language:

```
. use http://www.stata-press.com/data/r13/auto
(1978 Automobile Data)
. summarize mpg weight
```

Variable	Obs	Mean	Std. Dev.	Min	Max
mpg	74	21.2973	5.785503	12	41
weight	74	3019.459	777.1936	1760	4840

The user typed summarize mpg weight and Stata responded with a table of summary statistics. Other commands would produce different results:

. generate gp100m = 100/mpg

- . label var gp100m "Gallons per 100 miles"
- . format gp100m %5.2f

```
. correlate gp100m weight (obs=74)
```

	gp100m	weight
gp100m weight	1.0000 0.8544	1.0000

. regress gp100m weight gear_ratio

Source	SS	df	MS		Number of obs	= 74
Model Residual	87.4543721 32.1218886	2 43.7 71 .452	271861 420967		Prob > F R-squared Adj R-squared	= 0.0000 = 0.7314 = 0.7238
Total	119.576261	73 1.63	803097		Root MSE	= .67262
gp100m	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
weight gear_ratio _cons	.0014769 .1566091 .0878243	.0001556 .2651131 1.198434	9.49 0.59 0.07	0.000 0.557 0.942	.0011665 3720115 -2.301786	.0017872 .6852297 2.477435

. scatter gp100m weight, by(foreign)



The user-interface model is type a little, get a little, etc., so that the user is always in control.

Stata's model for a dataset is that of a table—the rows are the observations and the columns are the variables:

mpg weight gp100m 1. 22 2,930 4.55 2. 17 3,350 5.88 3. 22 2,640 4.55 4. 20 3,250 5.00 5. 15 4,080 6.67 6. 18 3,670 5.56 7. 26 2,230 3.85 8. 20 3,280 5.00 9. 16 3,880 6.25	mpg weight gp100m 1. 22 2,930 4.55 2. 17 3,350 5.88 3. 22 2,640 4.55 4. 20 3,250 5.00 5. 15 4,080 6.67 6. 18 3,670 5.56 7. 26 2,230 3.85 8. 20 3,280 5.00 9. 16 3,880 6.25 10. 19 3,400 5.26				
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$		mpg	weight	gp100m
6. 18 3,670 5.56 7. 26 2,230 3.85 8. 20 3,280 5.00 9. 16 3,880 6.25	6. 18 3,670 5.56 7. 26 2,230 3.85 8. 20 3,280 5.00 9. 16 3,880 6.25 10. 19 3,400 5.26	1. 2. 3. 4.	22 17 22 20 15	2,930 3,350 2,640 3,250 4 080	4.55 5.88 4.55 5.00 6.67
10 10 3 400 5 26	10. 19 3,400 5.20	6. 7. 8. 9.	18 26 20 16	3,670 2,230 3,280 3,880	5.56 3.85 5.00 6.25

. list mpg weight gp100m in 1/10

Observations are numbered; variables are named.

Stata is fast. That speed is due partly to careful programming, and partly because Stata keeps the data in memory. Stata's file model is that of a word processor: a dataset may exist on disk, but the dataset in memory is a copy. Datasets are loaded into memory, where they are worked on, analyzed, changed, and then perhaps stored back on disk.

Working on a copy of the data in memory makes Stata safe for interactive use. The only way to harm the permanent copy of your data on disk is if you explicitly save over it.

Having the data in memory means that the dataset size is limited by the amount of computer memory. Stata stores the data in memory in an efficient format—you will be surprised how much data can fit. Nevertheless, if you work with extremely large datasets, you may run into memory constraints. You will want to learn how to store your data as efficiently as possible; see [D] compress.

2.1 Video example

Tour of the Stata 13 interface