



Data Preparation & Descriptive Statistics

(ver. 2.7)

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Basic definitions...

For statistical analysis we think of **data** as a collection of different pieces of information or facts. These pieces of information are called variables. A **variable** is an identifiable piece of data containing one or more values. Those values can take the form of a number or text (which could be converted into number)

In the table below variables var1 thru var5 are a collection of seven values, 'id' is the identifier for each observation. This dataset has information for seven cases (in this case people, but could also be states, countries, etc) grouped into five variables.

id	var1	var2	var3	var4	var5
1	7.3	32.27	0.1	Yes	Male
2	8.28	40.68	0.56	No	Female
3	3.35	5.62	0.55	Yes	Female
4	4.08	62.8	0.83	Yes	Male
5	9.09	22.76	0.26	No	Female
6	8.15	90.85	0.23	Yes	Female
7	7.59	54.94	0.42	Yes	Male

Data structure...

For data analysis your data should have variables as columns and observations as rows. The first row should have the column headings. Make sure your dataset has **at least** one identifier (for example, individual id, family id, etc.)

id	var1	var2	var3	var4	var5
1	7.3	32.27	0.1	Yes	Male
2	8.28	40.68	0.56	No	Female
3	3.35	5.62	0.55	Yes	Female
4	4.08	62.8	0.83	Yes	Male
5	9.09	22.76	0.26	No	Female
6	8.15	90.85	0.23	Yes	Female
7	7.59	54.94	0.42	Yes	Male

← First row should have the variable names

← Cross-sectional data



At least one identifier

Cross-sectional time series data
or panel data →

	id	year	var1	var2	var3
Group 1	1	2000	7	74.03	0.55
	1	2001	2	4.6	0.44
	1	2002	2	25.56	0.77
Group 2	2	2000	7	59.52	0.05
	2	2001	2	16.95	0.94
	2	2002	9	1.2	0.08
Group 3	3	2000	9	85.85	0.5
	3	2001	3	98.85	0.32
	3	2002	3	69.2	0.76

Data format (ASCII)...

ASCII (American Standard Code for Information Interchange). The most universally accepted format. Practically any statistical software can open/read these type of files. Available formats:

- Delimited. Data is separated by comma, tab or space. The most common extension is *.csv (comma-separated value). Another type of extensions are *.txt for tab-separated data and *.prn for space-separated data. Any statistical package can read these formats.
- Record form (or fixed). Data is structured by fixed blocks (for example, var1 in columns 1 to 5, var2 in column 6 to 8, etc). You will need a codebook and to write a program (either in Stata, SPSS or SAS) to read the data. Extensions for the datasets could be *.dat, *.txt. For data in this format no column headings is available.

Data formats (comma-separated)...

Comma-separated value (*.csv)

```
ID,Last Name,First Name,City,State,Gender,Student Status,Major,Country,Age,SAT,Average score (grade),Height (in),Newspaper readership (times/wk),.....
1,DOE01,JANE01,Los Angeles,California,Female,Graduate,Politics,US,30,2263,67,61,5,.....
2,DOE02,JANE02,Sedona,Arizona,Female,Undergraduate,Math,US,19,2006,63,64,7,.....
3,DOE01,JOE01,Elmira,New York,Male,Graduate,Math,US,26,2221,78,73,6,.....
4,DOE02,JOE02,Lackawana,New York,Male,Graduate,Econ,US,33,1716,78,68,3,.....
5,DOE03,JOE03,Defiance,Ohio,Male,Graduate,Econ,US,37,1701,65,71,6,.....
6,DOE04,JOE04,Tel Aviv,Israel,Male,Graduate,Econ,Israel,25,1786,69,67,5,.....
7,DOE05,JOE05,Cimax,North Carolina,Male,Graduate,Politics,US,39,1577,96,70,5,.....
8,DOE03,JANE03,Liberal,Kansas,Female,Undergraduate,Politics,US,21,1842,87,62,5,.....
9,DOE04,JANE04,Montreal,Canada,Female,Undergraduate,Math,Canada,18,1813,91,62,6,.....
10,DOE05,JANE05,New York,New York,Female,Graduate,Math,US,33,2041,71,66,5,.....
11,DOE06,JOE06,Hot Coffe,Mississippi,Male,Undergraduate,Econ,US,18,1787,82,67,3,.....
12,DOE06,JANE06,Java,Virginia,Female,Graduate,Math,US,38,1513,79,59,5,.....
13,DOE07,JOE07,Varna,Bulgaria,Male,Graduate,Politics,Bulgaria,30,1637,79,63,4,.....
14,DOE08,JOE08,Moscow,Russia,Male,Graduate,Politics,Russia,30,1512,70,75,6,.....
15,DOE07,JANE07,Drunkard Creek,New York,Female,Undergraduate,Math,US,21,1338,82,64,5,.....
16,DOE08,JANE08,Mexican Hat,Utah,Female,Undergraduate,Econ,US,18,1821,80,63,3,.....
17,DOE09,JANE09,Amsterdam,Holland,Female,Undergraduate,Math,Holland,19,1494,75,60,3,.....
18,DOE10,JANE10,Mexico,Mexico,Female,Graduate,Politics,Mexico,31,2248,95,59,4,.....
19,DOE11,JANE11,Caracas,Venezuela,Female,Undergraduate,Math,Venezuela,18,2252,92,68,5,.....
20,DOE09,JOE09,San Juan,Puerto Rico,Male,Graduate,Politics,US,33,1923,95,63,7,.....
21,DOE12,JANE12,Remote,Oregon,Female,Undergraduate,Econ,US,19,1727,67,62,7,.....
22,DOE10,JOE10,New York,New York,Male,Undergraduate,Econ,US,21,1872,82,73,4,.....
23,DOE13,JANE13,The X,Massachusetts ,Female,Graduate,Politics,US,25,1767,89,68,6,.....
24,DOE14,JANE14,Beijing,China,Female,Undergraduate,Math,China,18,1643,79,65,6,.....
25,DOE11,JOE11,Stockholm,Sweden,Male,Undergraduate,Politics,Sweden,19,1919,88,64,4,.....
26,DOE12,JOE12,Embarrass,Minnesota,Male,Graduate,Econ,US,28,1434,96,71,4,.....
27,DOE13,JOE13,Inter course,Pennsylvania,Male,Undergraduate,Math,US,20,2119,88,71,5,.....
28,DOE15,JANE15,Loco,oklahoma,Female,Undergraduate,Econ,US,20,2309,64,68,6,.....
29,DOE14,JOE14,Buenos Aires,Argentina,Male,Graduate,Politics,Argentina,30,2279,85,72,3,.....
30,DOE15,JOE15,Acme,Louisiana,Male,Undergraduate,Econ,US,19,1907,79,74,3,.....
```

Data format (tab/space separated)...

Tab -
separated
value (*.txt)



ID	Last Name	First Name	City	State	Gender	Student Status	Major	Country	Age	SAT	Average score (grade)	Height (in)	Newspaper readership (times/wk)
1	DOE01	JANE01	Los Angeles	California	Female	Graduate	Politics	US	30	2263	67	61	5
2	DOE02	JANE02	Sedona	Arizona	Female	Undergraduate	Math	US	19	2006	63	64	7
3	DOE01	JOE01	Elmira	New York	Male	Graduate	Math	US	26	2221	78	73	6
4	DOE02	JOE02	Lackawana	New York	Male	Graduate	Econ	US	33	1716	78	68	3
5	DOE03	JOE03	Defiance	Ohio	Male	Graduate	Econ	US	37	1701	65	71	6
6	DOE04	JOE04	Tel Aviv	Israel	Male	Graduate	Econ	Israel	25	1786	69	67	5
7	DOE05	JOE05	Climax	North Carolina	Male	Graduate	Politics	US	39	1577	96	70	5
8	DOE03	JANE03	Liberal	Kansas	Female	Undergraduate	Politics	US	21	1842	87	62	5
9	DOE04	JANE04	Montreal	Canada	Female	Undergraduate	Math	Canada	18	1813	91	62	6
10	DOE05	JANE05	New York	New York	Female	Graduate	Math	US	33	2041	71	66	5
11	DOE06	JOE06	Hot Coffe	Mississippi	Male	Undergraduate	Econ	US	18	1787	82	67	3
12	DOE06	JANE06	Java	Virginia	Female	Graduate	Math	US	38	1513	79	59	5
13	DOE07	JOE07	Varna	Bulgaria	Male	Graduate	Politics	Bulgaria	30	1637	79	63	4
14	DOE08	JOE08	Moscow	Russia	Male	Graduate	Politics	Russia	30	1512	70	75	6
15	DOE07	JOE07	Drunkard Creek	New York	Female	Undergraduate	Math	US	21	1338	82	64	5
16	DOE08	JANE08	Mexican Hat	Utah	Female	Undergraduate	Econ	US	18	1821	80	63	3
17	DOE09	JANE09	Amsterdam	Holland	Female	Undergraduate	Math	Holland	19	1494	75	60	3
18	DOE10	JANE10	Mexico	Mexico	Female	Graduate	Politics	Mexico	31	2248	95	59	4
19	DOE11	JANE11	Caracas	Venezuela	Female	Undergraduate	Math	Venezuela	18	2252	92	68	5
20	DOE09	JOE09	San Juan	Puerto Rico	Male	Graduate	Politics	US	33	1923	95	63	7
21	DOE12	JANE12	Remote	Oregon	Female	Undergraduate	Econ	US	19	1727	67	62	7
22	DOE10	JOE10	New York	New York	Male	Undergraduate	Econ	US	21	1872	82	73	4
23	DOE13	JANE13	The X	Massachusetts	Female	Graduate	Politics	US	25	1767	89	68	6
24	DOE14	JANE14	Beijing	China	Female	Undergraduate	Math	China	18	1643	79	65	6
25	DOE11	JOE11	Stockholm	Sweden	Male	Undergraduate	Politics	Sweden	19	1919	88	64	4
26	DOE12	JOE12	Embarrass	Minnesota	Male	Graduate	Econ	US	28	1434	96	71	4
27	DOE13	JOE13	Intercourse	Pennsylvania	Male	Undergraduate	Math	US	20	2119	88	71	5
28	DOE15	JANE15	Loco	Oklahoma	Female	Undergraduate	Econ	US	20	2309	64	68	6
29	DOE14	JOE14	Buenos Aires	Argentina	Male	Graduate	Politics	Argentina	30	2279	85	72	3
30	DOE15	JOE15	Acme	Louisiana	Male	Undergraduate	Econ	US	19	1907	79	74	3

Space -
separated
value (*.prn)



ID	Last Name	First Name	City	State	Gender	Student Statu	Major	Country	Age	SAT	Average score	Height (in)	Newspaper readership (times
1000E01	JANE01		Los Angeles	California	Female	Graduate	Politics	US	30	2263	67	61	5
2000E02	JANE02		Sedona	Arizona	Female	Undergraduate	Math	US	19	2006	63	64	7
3000E01	JOE01		Elmira	New York	Male	Graduate	Math	US	26	2221	78	73	6
4000E02	JOE02		Lackawana	New York	Male	Graduate	Econ	US	33	1716	78	68	3
5000E03	JOE03		Defiance	Ohio	Male	Graduate	Econ	US	37	1701	65	71	6
6000E04	JOE04		Tel Aviv	Israel	Male	Graduate	Econ	Israel	25	1786	69	67	5
7000E05	JOE05		Climax	North Carolina	Male	Graduate	Politics	US	39	1577	96	70	5
8000E03	JANE03		Liberal	Kansas	Female	Undergraduate	Politics	US	21	1842	87	62	5
9000E04	JANE04		Montreal	Canada	Female	Undergraduate	Math	Canada	18	1813	91	62	6
1000E05	JANE05		New York	New York	Female	Graduate	Math	US	33	2041	71	66	5
1100E06	JOE06		Hot Coffe	Mississippi	Male	Undergraduate	Econ	US	18	1787	82	67	3
1200E06	JANE06		Java	Virginia	Female	Graduate	Math	US	38	1513	79	59	5
1300E07	JOE07		Varna	Bulgaria	Male	Graduate	Politics	Bulgaria	30	1637	79	63	4
1400E08	JOE08		Moscow	Russia	Male	Graduate	Politics	Russia	30	1512	70	75	6
1500E07	JANE07		Drunkard Cree	New York	Female	Undergraduate	Math	US	21	1338	82	64	5
1600E08	JANE08		Mexican Hat	Utah	Female	Undergraduate	Econ	US	18	1821	80	63	3
1700E09	JANE09		Amsterdam	Holland	Female	Undergraduate	Math	Holland	19	1494	75	60	3
1800E10	JANE10		Mexico	Mexico	Female	Graduate	Politics	Mexico	31	2248	95	59	4
1900E11	JANE11		Caracas	Venezuela	Female	Undergraduate	Math	Venezuela	18	2252	92	68	5
2000E09	JOE09		San Juan	Puerto Rico	Male	Graduate	Politics	US	33	1923	95	63	7
2100E12	JANE12		Remote	Oregon	Female	Undergraduate	Econ	US	19	1727	67	62	7
2200E10	JOE10		New York	New York	Male	Undergraduate	Econ	US	21	1872	82	73	4
2300E13	JANE13		The X	Massachusetts	Female	Graduate	Politics	US	25	1767	89	68	6
2400E14	JANE14		Beijing	China	Female	Undergraduate	Math	China	18	1643	79	65	6
2500E11	JOE11		Stockholm	Sweden	Male	Undergraduate	Politics	Sweden	19	1919	88	64	4
2600E12	JOE12		Embarrass	Minnesota	Male	Graduate	Econ	US	28	1434	96	71	4
2700E13	JOE13		Intercourse	Pennsylvania	Male	Undergraduate	Math	US	20	2119	88	71	5
2800E15	JANE15		Loco	Oklahoma	Female	Undergraduate	Econ	US	20	2309	64	68	6
2900E14	JOE14		Buenos Aires	Argentina	Male	Graduate	Politics	Argentina	30	2279	85	72	3
3000E15	JOE15		Acme	Louisiana	Male	Undergraduate	Econ	US	19	1907	79	74	3

Data format (record/fixed)...

Record form (fixed) ASCII (*.txt, *.dat). For this format you need a **codebook** to figure out the layout of the data (it indicates where a variable starts and where it ends). See next slide for an example. Notice that fixed datasets do not have column headings.

```
DOE01JANE01Los AngelesCaliforniaFemaleGraduatePoliticsUS302263676152DOE02JANE02SedonaArizonaFemaleUndergraduateMathUS192006636473
DOE01JOE01ElmiraNew YorkMaleGraduateMathUS262221787364DOE02JOE02LackawanaNew YorkMaleGraduateEconUS331716786835
DOE03JOE03DefianceOhioMaleGraduateEconUS371701657166DOE04JOE04Tel AvivIsraelMaleGraduateEconIsrael251786696757
DOE05JOE05CimaxNorth CarolinaMaleGraduatePoliticsUS391577967058
DOE03JANE03LiberalKansasFemaleUndergraduatePoliticsUS211842876259DOE04JANE04MontrealCanadaFemaleUndergraduateMathCanada1818139162
10
DOE05JANE05New YorkNew YorkFemaleGraduateMathUS3320417166511DOE06JOE06Hot CoffeMississippiMaleUndergraduateEconUS1817878267312
DOE06JANE06JavaVirginiaFemaleGraduateMathUS3815137959513
DOE07JOE07VarnaBulgariaMaleGraduatePoliticsBulgaria3016377963414DOE08JOE08MoscowRussiaMaleGraduatePoliticsRussia3015127075615
DOE07JANE07Drunkard CreekNew YorkFemaleUndergraduateMathUS2113388264516DOE08JANE08Mexican HatUtahFemaleUndergraduateEconUS181821
8063317
DOE09JANE09AmsterdamHollandFemaleUndergraduateMathHolland1914947560318DOE10JANE10MexicoMexicoFemaleGraduatePoliticsMexico31224895
9419
DOE11JANE11CaracasVenezuelaFemaleUndergraduateMathVenezuela1822529268520
DOE09JOE09San JuanPuerto RicoMaleGraduatePoliticsUS3319239563721DOE12JANE12RemoteOregonFemaleUndergraduateEconUS1917276762722
DOE10JOE10New YorkNew YorkMaleUndergraduateEconUS2118728273423
DOE13JANE13The XMassachusetts FemaleGraduatePoliticsUS2517678968624DOE14JANE14BeijingChinaFemaleUndergraduateMathChina18164379656
5
DOE11JOE11StockholmSwedenMaleUndergraduatePoliticsSweden1919198864426DOE12JOE12EmbarrassMinesotaMaleGraduateEconUS2814349671427
DOE13JOE13Inter coursePennsylvaniaMaleUndergraduateMathUS2021198871528DOE15JANE15LocooklahoffaFemaleUndergraduateEconUS20230964686
29DOE14JOE14Buenos AiresArgentinaMaleGraduatePoliticsArgentina3022798572330DOE15JOE15AcmeLouisianaMaleUndergraduateEconUS19190779
43
```


Codebook (ASCII to Stata using `infix`)

NOTE: The following is a small example of a codebook. Codebooks are like maps to help you figure out the structure of the data. Codebooks differ on how they present the layout of the data, in general, you need to look for: variable name, start column, end column or length, and format of the variable (whether is numeric and how many decimals (identified with letter 'F') or whether is a string variable marked with letter 'A')

Data Locations

Variable	Rec	Start	End	Format
var1	1	1	7	F7.2
var2	1	24	25	F2.0
var3	1	26	27	A2
var4	1	32	33	F2.0
var5	1	44	45	A2



In Stata you write the following to open the dataset.
In the command window type:

```
infix var1 1-7 var2 24-25 str2 var3 26-27 var4 32-33 str2 var5 44-45 using mydata.dat
```

Notice the '`str#`' before var3 and var5, this is to indicate that these variables are string (text). The number in `str` refers to the length of the variable.

If you get an error like `...cannot be read as a number for...` [click here](#)

From ASCII to Stata using a dictionary file/infile

Using notepad or the do-file editor type:

```
dictionary using c:\data\mydata.dat {  
    _column(1)          var1          %7.2f          "Label for var1 "  
    _column(24)         var2          %2f            "Label for var2 "  
    _column(26) str2 var3          %2s            "Label for var3 "  
    _column(32)         var4          %2f            "Label for var4 "  
    _column(44) str2 var5          %2s            "Label for var5 "  
}  
/*Do not forget to close the brackets and press enter after the last bracket*/
```

Notice that the numbers in `_column(#)` refers to the position where the variable starts based on what the codebook shows. The option 'str#' indicates that the variable is a string (text or alphanumeric) with two characters, here you need to specify the length of the variable for Stata to read it correctly.

Save it as *mydata.dct*

To read data using the dictionary we need to import the data by using the command `infile`. If you want to use the menu go to File – Import - “ASCII data in fixed format with a data dictionary”.

With `infile` we run the dictionary by typing:

```
infile using c:\data\mydata
```

NOTE: Stata commands sometimes do not work with copy-and-paste. If you get error try re-typing the commands

From ASCII to Stata using a dictionary file/`infile` (data with more than one record)

If your data is in more than one records using notepad or the do-file editor type:

```
dictionary using c:\data\mydata.dat {  
    _lines(2)  
    _line(1)  
    _column(1)      var1      %7.2f      "Label for var1 "  
    _column(24)     var2      %2f        "Label for var2 "  
    _line(2)  
    _column(26) str2 var3      %2s        "Label for var3 "  
    _column(32)     var4      %2f        "Label for var4 "  
    _column(44) str2 var5      %2s        "Label for var5 "  
}  
/*Do not forget to close the brackets and press enter after the last bracket*/
```

Notice that the numbers in `_column(#)` refers to the position where the variable starts based on what the codebook shows.

Save it as `mydata.dct`

To read data using the dictionary we need to import the data by using the command `infile`. If you want to use the menu go to File – Import - “ASCII data in fixed format with a data dictionary”.

With `infile` we run the dictionary by typing:

```
infile using c:\data\mydata
```

NOTE: Stata commands sometimes do not work with copy-and-paste. If you get error try re-typing the commands
For more info on data with records see http://www.columbia.edu/cu/lweb/indiv/dssc/eds/stata_write.html

From ASCII to Stata: error message

If running `infix` or `infile` you get errors like:

```
'1-1001-' cannot be read as a number for var1[14]  
'del11' cannot be read as a number for var2[11]  
'xvet-' cannot be read as a number for var3[15]  
'0---0' cannot be read as a number for var4[16]  
'A5' cannot be read as a number for var5[16]
```

Make sure you specified those variables to be read as strings (`str`) and set to the correct length (`str#`), see the codebook for these.

Double-check the data locations from the codebook. If the data file has more than one record make sure is indicated in the dictionary file.

If after checking for the codebook you find no error in the data locations or the data type, then depending of the type of variable, this may or may not be an error. Stata will still read the variables but those non-numeric observations will be set to missing.

From ASCII to SPSS

Using the syntax editor in SPSS and following the data layout described in the codebook, type:

```
FILE HANDLE FHAND /NAME='C:\data\mydata.dat' /LRECL=1003.
DATA LIST FILE=FHAND FIXED RECORDS = 1 TABLE /
    var1 1-7
    var2 24-25
    var3 26-27 (A)
    var4 32-33
    var5 44-45 (A) .
EXECUTE.
```

You get /LRECL from the codebook.

Select the program and run it by clicking on the arrow



If you have more than one record type:

```
FILE HANDLE FHAND /NAME='C:\data\mydata.dat' /LRECL=1003.
DATA LIST FILE=FHAND FIXED RECORDS = 2 TABLE
/1
    var1 1-7
    var2 24-25
    var3 26-27 (A)
/2
    var4 32-33
    var5 44-45 (A) .
EXECUTE.
```

PU/DSS/OTR Notice the '(A)' after var3 and var5, this is to indicate that these variables are string (text).

From SPSS/SAS to Stata

If your data is already in SPSS format (*.sav) or SAS(*.sas7bcat). You can use the command `usespss` to read SPSS files in Stata or the command `usesas` to read SAS files.

If you have a file in SAS XPORT format you can use `fduse` (or go to file-import).

For SPSS and SAS, you may need to install it by typing

```
ssc install usespss  
ssc install usesas
```

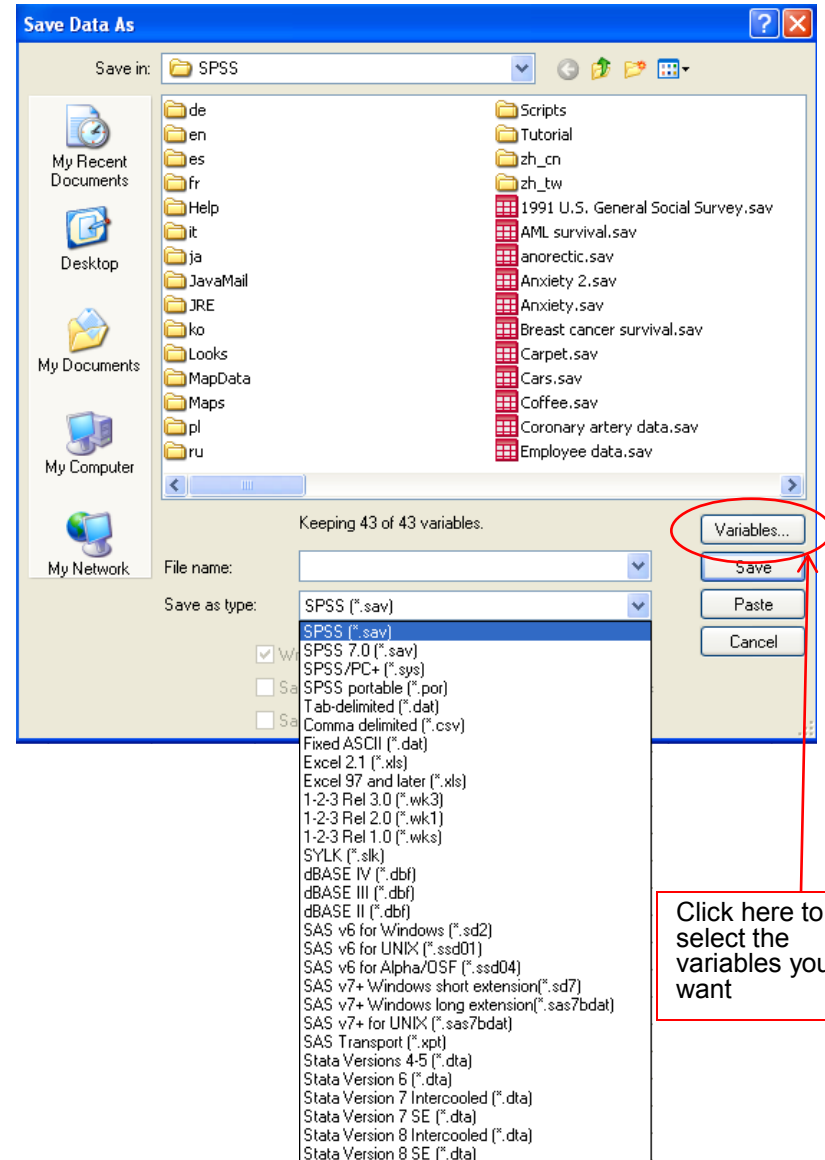
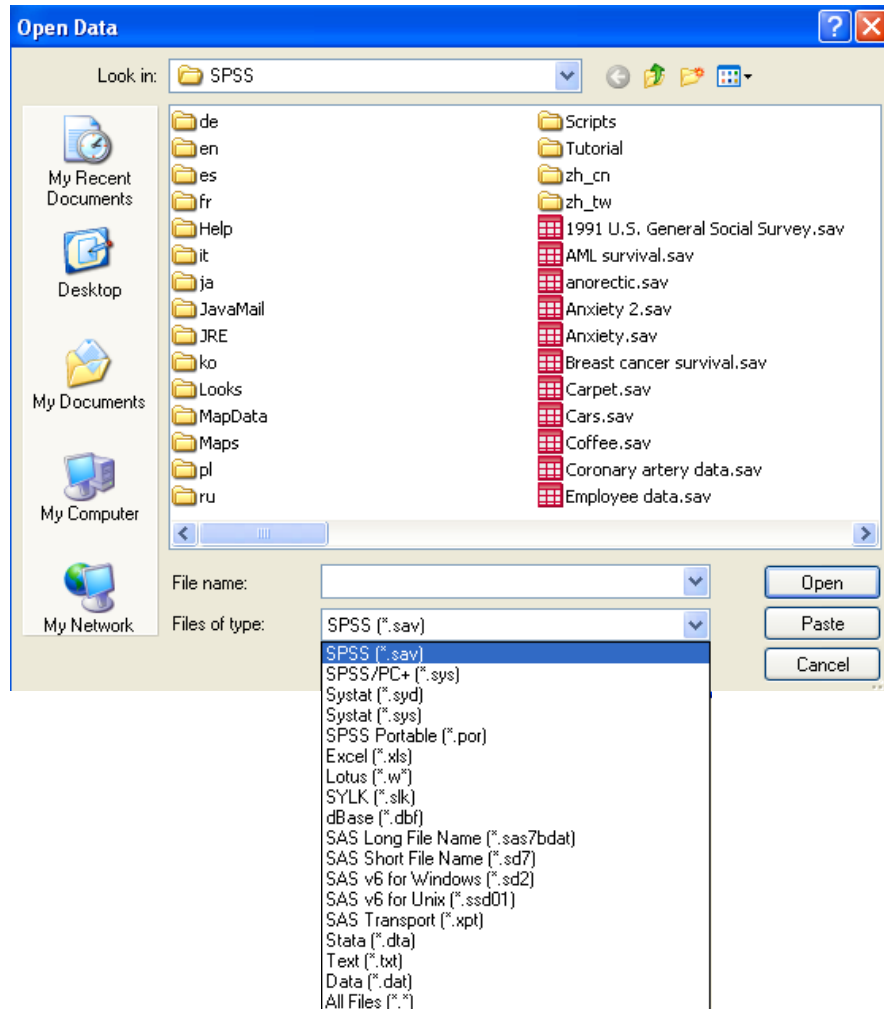
Once installed just type

```
usespss using "c:\mydata.sav"  
usesas using "c:\mydata.sas7bcat"
```

Type `help usespss` or `help usesas` for more details.

Loading data in SPSS

SPSS can read/save-as many proprietary data formats, go to file-open-data or file-save as



Loading data in R

1. tab-delimited (*.txt), type:

```
mydata <- read.table("mydata.txt")  
mydata <- read.table("mydata.txt", header = TRUE, na.strings = "-9") #If  
    missing data is coded as "-9"
```

2. space-delimited (*.prn), type:

```
mydata <- read.table("mydata.prn")
```

3. comma-separated value (*.csv), type:

```
mydata <- read.csv("mydata.csv")  
mydata <- read.csv("mydata.csv", header = TRUE) #With column headings
```

4. From SPSS/Stata to R use the foreign package, type:

```
library(foreign) # Load the foreign package.  
stata.data <- read.dta("mydata.dta") # For Stata.  
spss.data <- read.spss("mydata.sav", to.data.frame = TRUE) # For SPSS.
```

5. To load data in R format use

```
mydata <- load("mydata.RData")
```

Source: <http://gking.harvard.edu/zelig/docs/static/syntax.pdf>

Also check: http://www.ats.ucla.edu/stat/R/modules/raw_data.htm

Other data formats...

Features	Stata	SPSS	SAS	R
Data extensions	*.dta	*.sav, *.por (portable file)	*.sas7bcat, *.sas#bcat, *.xpt (xport files)	*.Rdata
User interface	Programming/point-and-click	Mostly point-and-click	Programming	Programming
Data manipulation	Very strong	Moderate	Very strong	Very strong
Data analysis	Powerful	Powerful	Powerful/versatile	Powerful/versatile
Graphics	Very good	Very good	Good	Good
Cost	Affordable (perpetual licenses, renew only when upgrade)	Expensive (but not need to renew until upgrade, long term licenses)	Expensive (yearly renewal)	Open source
Program extensions	*.do (do-files)	*.sps (syntax files)	*.sas	*.txt (log files)
Output extension	*.log (text file, any word processor can read it), *.smcl (formatted log, only Stata can read it).	*.spo (only SPSS can read it)	(various formats)	*.txt (log files, any word processor can read)

Compress data files (*.zip, *.gz)

If you have datafiles with extension *.zip, *.gz, *.rar you need file compression software to extract the datafiles. You can use Winzip, WinRAR or 7-zip among others.

7-zip (<http://7-zip.org/>) is freeware and deals with most compressed formats.

Stata allows you to unzip files from the command window.

```
unzipfile "c:\data\mydata.zip"
```

You can also zip file using zipfile

```
zipfile myzip.zip mydata.dta
```

Before you start

Once you have your data in the proper format, before you perform any analysis you need to explore and prepare it first:

1. Make sure variables are in columns and observations in rows.
2. Make sure you have all variables you need.
3. Make sure there is at least one id.
4. If times series make sure you have the years you want to include in your study.
5. Make sure missing data has either a blank space or a dot ('.')
6. Make sure to make a back-up copy of your original dataset.
7. Have the codebook handy.

Stata color-coded system

An important step is to make sure variables are in their expected format. Numeric should be numeric and text should be text.

Stata has a color-coded system for each type. Black is for numbers, red is for text or string and blue is for labeled variables.

Var2 is a string variable even though you see numbers. You can't do any statistical procedure with this variable other than simple frequencies

Var3 is a numeric You can do any statistical procedure with this variable

	var1	var2	var3	var4
1	Fairly well	2	2	Fairly well
2	Very well	1	1	Very well
3	Fairly badly	3	3	Fairly badly
4	Fairly well	2	2	Fairly well
5	Very badly	4	4	Very badly
6	Fairly badly	3	3	Fairly badly
7	Fairly well	2	2	Fairly well

For var1 a value 2 has the label "Fairly well". It is still a numeric variable

Var4 is clearly a string variable. You can do frequencies and crosstabulations with this but not any statistical procedure.

Cleaning your variables

If you are using datasets with [categorical](#) variables you need to clean them by getting rid of the non-response categories like 'do not know', 'no answer', 'no applicable', 'not sure', 'refused', etc.

Usually non-response categories have higher values like 99, 999, 9999, etc (or in some cases negative values). Leaving these will bias, for example, the mean age or your regression results as outliers.

In the example below the non-response is coded as 999 and if we leave this the mean age would be 80 years, removing the 999 and setting it to missing, the average age goes down to 54 years.

This is a frequency of age, notice the 999 value for the no response.

88	2	0.15	96.58
90	3	0.22	96.80
92	4	0.29	97.09
93	1	0.07	97.16
95	1	0.07	97.23
999	38	2.77	100.00
<hr/>			
Total	1,373	100.00	

```
. tabstat age age_w999
```

In Stata you can type

```
replace age=. if age==999
```

or

```
replace age=. if age>100
```

stats	age	age_w999
mean	54.58801	80.72615

Cleaning your variables

No response categories not only affect the statistics of the variable, it may also affect the interpretation and coefficients of the variable if we do not remove them.

In the example below responses go from 'very well' to 'refused', with codes 1 to 6. Leaving the variable 'as-is' in a regression model will misinterpret the variable as going from quite positive to ... refused? This does not make sense. You need to clean the variable by eliminating the no response so it goes from positive to negative. Even more, you may have to reverse the valence so the variable goes from negative to positive for a better/easier interpretation.

```
. tab var1
```

Status of Nat'l Eco	Freq.	Percent	Cum.
Very well	149	10.85	10.85
Fairly well	670	48.80	59.65
Fairly badly	348	25.35	85.00
Very badly	191	13.91	98.91
Not sure	12	0.87	99.78
Refused	3	0.22	100.00
Total	1,373	100.00	

```
. tab var1, nolabel
```

Status of Nat'l Eco	Freq.	Percent	Cum.
1	149	10.85	10.85
2	670	48.80	59.65
3	348	25.35	85.00
4	191	13.91	98.91
5	12	0.87	99.78
6	3	0.22	100.00
Total	1,373	100.00	

=

Cleaning your variables (using `recode` in Stata)

First, **never** work with the original variable, *always keep originals original*.

The command `recode` in Stata lets you create a new variable without modifying the original.

```
recode var1 (1=4 "Very well") (2=3 "Fairly well") (3=2 "Fairly badly")  
(4=1 "Very badly") (else=.), gen(var1_rec) label(var1_rec)
```

Get frequencies of both variables: `var1` and `var1_rec` to verify:

. tab var1				. tab var1_rec			
Status of Nat'l Eco	Freq.	Percent	Cum.	RECODE of var1 (Status of Nat'l Eco)	Freq.	Percent	Cum.
Very well	149	10.85	10.85				
Fairly well	670	48.80	59.65				
Fairly badly	348	25.35	85.00				
Very badly	191	13.91	98.91	Very badly	191	14.06	14.06
Not sure	12	0.87	99.78	Fairly badly	348	25.63	39.69
Refused	3	0.22	100.00	Fairly well	670	49.34	89.03
				Very well	149	10.97	100.00
Total	1,373	100.00		Total	1,358	100.00	

Now you can use `var1_rec` in a regression since it is an ordinal variable where higher values mean positive opinions. This process is useful when combining variables to create indexes.

For additional help on data management, analysis and presentation please check:

<http://dss.princeton.edu/training/>

<http://dss.princeton.edu/>

Reshape wide to long (if original data in Excel)

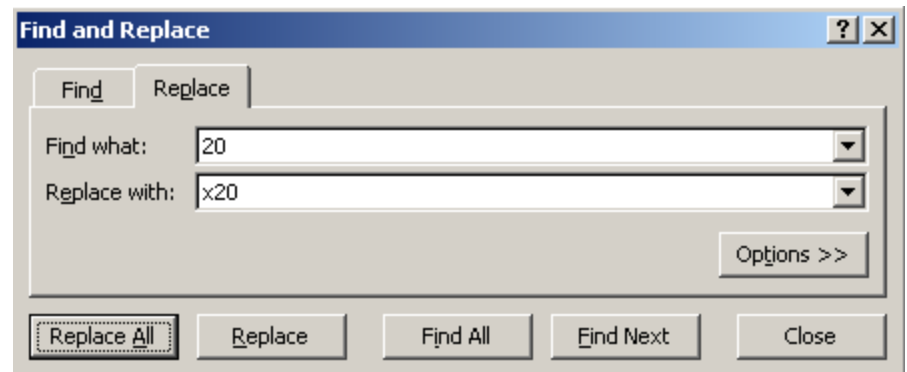
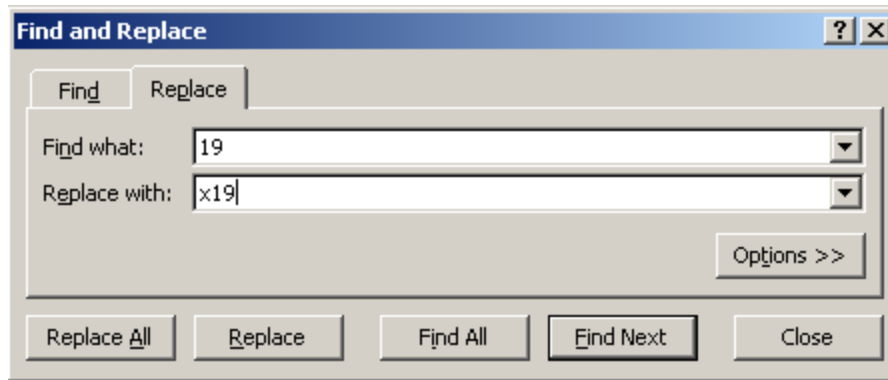
The following dataset is not ready for analysis, years are in columns and cases and variables are in rows ([click here to get it](#)). The ideal is for years and countries to be in rows and variables (var1 and var2) in columns. We should have four columns: Country, Year, var1 and var2

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Country	Variable	1995	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005
2	A	var1	8000.01	8212.90	7847.36	7702.89	7288.48	6430.98	6932.45	7486.24	8094.17
3	A	var2	6.83	2.66	-4.45	-1.84	-5.38	-11.77	7.80	7.99	8.12
4	B	var1	18268.01	18738.99	19360.46	20151.42	20715.54	20866.90	21364.02	21801.41	22404.59	22676.26	23039.43
5	B	var2	2.87	2.58	3.32	4.09	2.80	0.73	2.38	2.05	2.77	1.21	1.60
6	C	var1	21088.14	21608.14	21988.64	22739.28	23436.61	24194.85	24300.57	24411.48	24650.02	25076.01	25346.01
7	C	var2	1.60	2.47	1.76	3.41	3.07	3.24	0.44	0.46	0.98	1.73	1.08
8	D	var1	313.74	321.36	331.76	342.12	351.70	365.33	377.15	386.26	398.86	415.96	432.63
9	D	var2	2.66	2.43	3.24	3.12	2.80	3.87	3.24	2.42	3.26	4.29	4.01
10	E	var1	21123.66	21659.55	22299.13	22972.31	23613.87	24150.86	24788.69	25368.87	25885.48	26582.19	26890.73
11	E	var2	2.69	2.54	2.95	3.02	2.79	2.27	2.64	2.34	2.04	2.69	1.16
12	F	var1	29941.64	30703.73	31716.04	32671.27	33748.21	34599.47	34483.98	34669.47	35312.75	36450.55	37267.33
13	F	var2	1.32	2.55	3.30	3.01	3.30	2.52	-0.33	0.54	1.86	3.22	2.24
14	G	var1	4891.60	5063.81	5328.88	5512.59	5647.06	5934.98	5864.12	5852.99	5872.29	6055.92	6162.84
15	G	var2	-7.86	3.52	5.23	3.45	2.44	5.10	-1.19	-0.19	0.33	3.13	1.77

We can prepare this dataset using Stata but we need to do some changes in Excel.

Reshape wide to long (if original data in Excel)

First, you need to add a character to the column headings so Stata can read them. Stata does not take numbers as variable names. In this case we add an “x” to the years. In excel you do this by using the ‘replace’ function. For the 1900s we replace “19” for “x19”, same for the 2000s (make sure to select only the headings). See the following

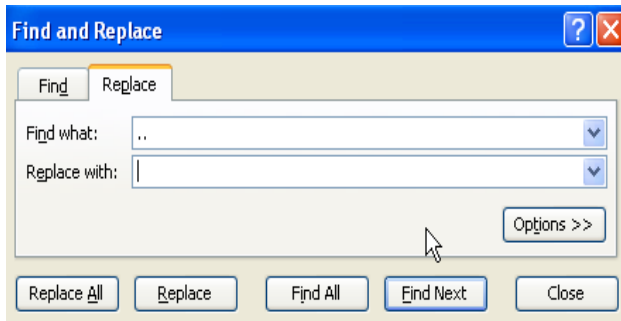


Reshape wide to long (if original data in Excel)

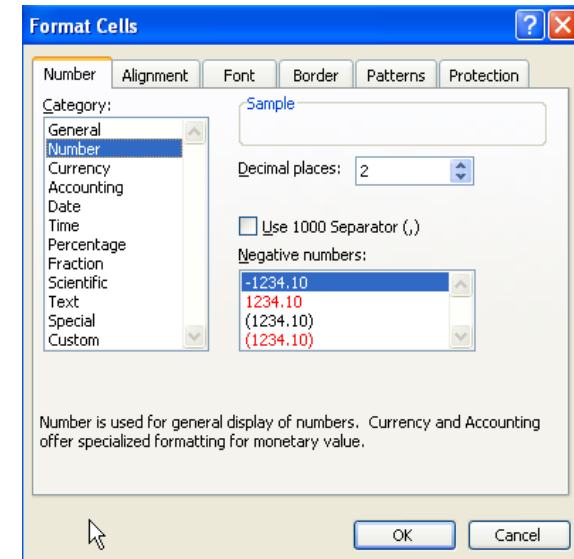
We have...

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Country	Variable	x1995	x1996	x1997	x1998	x1999	x2000	x2001	x2002	x2003	x2004	x2005
2	A	var1	8000.01	8212.90	7847.36	7702.89	7288.48	6430.98	6932.45	7486.24	8094.17
3	A	var2	6.83	2.66	-4.45	-1.84	-5.38	-11.77	7.80	7.99	8.12
4	B	var1	18268.01	18738.99	19360.46	20151.42	20715.54	20866.90	21364.02	21801.41	22404.59	22676.26	23039.43
5	B	var2	2.87	2.58	3.32	4.09	2.80	0.73	2.38	2.05	2.77	1.21	1.60
6	C	var1	21088.14	21608.14	21988.64	22739.28	23436.61	24194.85	24300.57	24411.48	24650.02	25076.01	25346.01
7	C	var2	1.60	2.47	1.76	3.41	3.07	3.24	0.44	0.46	0.98	1.73	1.08
8	D	var1	313.74	321.36	331.76	342.12	351.70	365.33	377.15	386.26	398.86	415.96	432.63
9	D	var2	2.66	2.43	3.24	3.12	2.80	3.87	3.24	2.42	3.26	4.29	4.01
10	E	var1	21123.66	21659.55	22299.13	22972.31	23613.87	24150.86	24788.69	25368.87	25885.48	26582.19	26890.73
11	E	var2	2.69	2.54	2.95	3.02	2.79	2.27	2.64	2.34	2.04	2.69	1.16
12	F	var1	29941.64	30703.73	31716.04	32671.27	33748.21	34599.47	34483.98	34669.47	35312.75	36450.55	37267.33
13	F	var2	1.32	2.55	3.30	3.01	3.30	2.52	-0.33	0.54	1.86	3.22	2.24
14	G	var1	4891.60	5063.81	5328.88	5512.59	5647.06	5934.98	5864.12	5852.99	5872.29	6055.92	6162.84
15	G	var2	-7.86	3.52	5.23	3.45	2.44	5.10	-1.19	-0.19	0.33	3.13	1.77

Replace the dots “..” (or any string character) with a blank



Make sure the numbers are numbers. Select all and format cells as numbers.



Reshape wide to long (from Excel to Stata)

The table should look like.

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Country	Variable	x1995	x1996	x1997	x1998	x1999	x2000	x2001	x2002	x2003	x2004	x2005
2	A	var1			8000.01	8212.90	7847.36	7702.89	7288.48	6430.98	6932.45	7486.24	8094.17
3	A	var2			6.83	2.66	-4.45	-1.84	-5.38	-11.77	7.80	7.99	8.12
4	B	var1	18268.01	18738.99	19360.46	20151.42	20715.54	20866.90	21364.02	21801.41	22404.59	22676.26	23039.43
5	B	var2	2.87	2.58	3.32	4.09	2.80	0.73	2.38	2.05	2.77	1.21	1.60
6	C	var1	21088.14	21608.14	21988.64	22739.28	23436.61	24194.85	24300.57	24411.48	24650.02	25076.01	25346.01
7	C	var2	1.60	2.47	1.76	3.41	3.07	3.24	0.44	0.46	0.98	1.73	1.08
8	D	var1	313.74	321.36	331.76	342.12	351.70	365.33	377.15	386.26	398.86	415.96	432.63
9	D	var2	2.66	2.43	3.24	3.12	2.80	3.87	3.24	2.42	3.26	4.29	4.01
10	E	var1	21123.66	21659.55	22299.13	22972.31	23613.87	24150.86	24788.69	25368.87	25885.48	26582.19	26890.73
11	E	var2	2.69	2.54	2.95	3.02	2.79	2.27	2.64	2.34	2.04	2.69	1.16
12	F	var1	29941.64	30703.73	31716.04	32671.27	33748.21	34599.47	34483.98	34669.47	35312.75	36450.55	37267.33
13	F	var2	1.32	2.55	3.30	3.01	3.30	2.52	-0.33	0.54	1.86	3.22	2.24
14	G	var1	4891.60	5063.81	5328.88	5512.59	5647.06	5934.98	5864.12	5852.99	5872.29	6055.92	6162.84
15	G	var2	-7.86	3.52	5.23	3.45	2.44	5.10	-1.19	-0.19	0.33	3.13	1.77

Copy and paste the table from Excel to Stata. In Stata go to Data -> Data Editor



Data Editor													
country [1] =													
	country	variable	x1995	x1996	x1997	x1998	x1999	x2000	x2001	x2002	x2003	x2004	x2005
1	A	var1	.	.	8000.01	8212.9	7847.36	7702.89	7288.48	6430.98	6932.45	7486.24	8094.17
2	A	var2	.	.	6.83	2.66	-4.45	-1.84	-5.38	-11.77	7.8	7.99	8.12
3	B	var1	18268.01	18738.99	19360.46	20151.42	20715.54	20866.9	21364.02	21801.41	22404.59	22676.26	23039.43
4	B	var2	2.87	2.58	3.32	4.09	2.8	.73	2.38	2.05	2.77	1.21	1.6
5	C	var1	21088.14	21608.14	21988.64	22739.28	23436.61	24194.85	24300.57	24411.48	24650.02	25076.01	25346.01
6	C	var2	1.6	2.47	1.76	3.41	3.07	3.24	.44	.46	.98	1.73	1.08
7	D	var1	313.74	321.36	331.76	342.12	351.7	365.33	377.15	386.26	398.86	415.96	432.63
8	D	var2	2.66	2.43	3.24	3.12	2.8	3.87	3.24	2.42	3.26	4.29	4.01
9	E	var1	21123.66	21659.55	22299.13	22972.31	23613.87	24150.86	24788.69	25368.87	25885.48	26582.19	26890.73
10	E	var2	2.69	2.54	2.95	3.02	2.79	2.27	2.64	2.34	2.04	2.69	1.16
11	F	var1	29941.64	30703.73	31716.04	32671.27	33748.21	34599.47	34483.98	34669.47	35312.75	36450.55	37267.33
12	F	var2	1.32	2.55	3.3	3.01	3.3	2.52	-.33	.54	1.86	3.22	2.24
13	G	var1	4891.6	5063.81	5328.88	5512.59	5647.06	5934.98	5864.12	5852.99	5872.29	6055.92	6162.84
14	G	var2	-7.86	3.52	5.23	3.45	2.44	5.1	-1.19	-.19	.33	3.13	1.77

Reshape wide to long (summary)

id	x2001	x2002	x2003
1	2	7	1
2	3	5	9
3	1	1	8

```
gen id = _n  
order id  
reshape long x , i(id) j(year)
```



id	year	x
1	1	2
1	2	7
1	3	1
2	1	3
2	2	5
2	3	9
3	1	1
3	2	1
3	3	8

date	x_var1	x_var2	x_var3
1	2	7	1
2	3	5	9
3	1	1	8

```
reshape long x_var , i(date) j(id) str
```



date	id	x_var
1	1	2
1	2	7
1	3	1
2	1	3
2	2	5
2	3	9
3	1	1
3	2	1
3	3	8

Reshape (Stata, 1)

Back to the example, create a unique id for each observation, type:

```
gen id = _n  
order id
```



Variables	
Name	Label
id	
country	Country
variable	Variable
x1995	
x1996	
x1997	
x1998	
x1999	
x2000	
x2001	
x2002	
x2003	
x2004	
x2005	

To reshape from wide to long, type

```
reshape long x, i(id) j(year)
```

```
. reshape long x, i(id) j(year)  
(note: j = 1995 1996 1997 1998 1999 2000 2001 2002 2003 2004 2005)
```

Data	wide	->	long
Number of obs.	14	->	154
Number of variables	14	->	5
j variable (11 values)		->	year
xij variables:			
	x1995 x1996 ... x2005	->	x

Where:

- **long** – Goes from wide to long format.
- **x** – The variables with the prefix “x” (x1960, x1961, x1962, etc.) are to be converted from wide to long.
- **i(id)** – A unique identifier for the wide format is in variable “id”.
- **j(year)** – Indicates that the suffix of “x” (x1961, x1962, x1963, ...), the years, should be put in variable called “year”.

NOTE: If you have more than one variable you can list them as follows:

```
reshape long x y z, i(id) j(year)
```

Reshape wide to long (Stata, 2)

The data it should look like the picture below. Notice that `var1` and `var2` are together in one column as variable '`x`' (the prefix we originally had for the years). If we had one variable we are done, in this example we have **two** and we need to separate them into two columns, `var1` and `var2`. Basically we need to reshape again but this time from long to wide.

	id	year	country	variable	x
1	1	1995	A	var1	.
2	1	1996	A	var1	.
3	1	1997	A	var1	8000.01
4	1	1998	A	var1	8212.9
5	1	1999	A	var1	7847.36
6	1	2000	A	var1	7702.89
7	1	2001	A	var1	7288.48
8	1	2002	A	var1	6430.98
9	1	2003	A	var1	6932.45
10	1	2004	A	var1	7486.24
11	1	2005	A	var1	8094.17
12	2	1995	A	var2	.
13	2	1996	A	var2	.
14	2	1997	A	var2	6.83
15	2	1998	A	var2	2.66
16	2	1999	A	var2	-4.45
17	2	2000	A	var2	-1.84
18	2	2001	A	var2	-5.38
19	2	2002	A	var2	-11.77
20	2	2003	A	var2	7.8
21	2	2004	A	var2	7.99

Reshape (Stata, 3)

To separate `var1` and `var2` we need to do a little bit of work.

First we need to create a new variable with the labels of each variable, type

```
encode variable, gen(varlabel)
```



```
. encode variable, gen(varlabel)
```

```
. tab varlabel
```

Variable	Freq.	Percent	Cum.
var1	77	50.00	50.00
var2	77	50.00	100.00
Total	154	100.00	

```
. tab varlabel, nolabel
```

Variable	Freq.	Percent	Cum.
1	77	50.00	50.00
2	77	50.00	100.00
Total	154	100.00	

Create a do-file with the labels for each variable. This comes in handy when dealing with lots of variables.

```
label save varlabel using varname, replace
```

You will notice that a file `varname.do` is created.



```
. label save varlabel using varname, replace  
file varname.do saved
```

Open the do-file with the do-file editor and do the following changes...

```
Stata Do-File Editor - varname.do  
File Edit Search Tools  
label define varlabel 1 `var1', modify  
label define varlabel 2 `var2', modify
```

- Change "label define" to "label variable"
- Change "varlabel 1" to "x1" and "varlabel 2" to "x2"
- Delete ", modify"
- Save the do-file



```
Stata Do-File Editor - varname.do  
File Edit Search Tools  
label variable x1 `var1'  
label variable x2 `var2'
```

Reshape (Stata, 4)

To separate `var1` and `var2` we need to reshape again, this time from long to wide. First we need to create another id to identify the groups (country and years), type

```
egen id2 = group(country year)
move id2 year
drop id
drop variable
```



Variables		X	
Name	Label		
id2			
year			
x			
varlabel			

Reshape the data by typing

```
reshape wide x, i(id2) j(varlabel)
order id2 country year x1 x2
. reshape wide x, i(id2) j(varlabel)
(note: j = 1 2)
```



Data	long	->	wide
Number of obs.	154	->	77
Number of variables	5	->	5
j variable (2 values)	varlabel	->	(dropped)
xij variables:			
	x	->	x1 x2

Variables		X	
Name	Label		
id2	group(country year)		
x1	1 x		
x2	2 x		
year			
country	Country		

Where:

wide – Indicates long to wide format.

x – The variable of interest to go from long to wide is called “data”.

i(id2) – A unique identifier for the wide format is in variable “id2”.

j(varlabel) – Indicates that the suffix of “data” has to be taken from “varlabel” (“varlabel” has two categories: 1 –var1- and 2 – var2).

NOTE: If “j” is not available in your dataset, you may be able to generate one using the following command:


```
bysort id: gen jvar=_n
```

Then reshape

```
reshape wide data, i(id) j(jvar)
```


Reshape (Stata, 5)

Run the do-file `varname.do` by selecting all and clicking on the last icon, this will change the labels for `x1` and `x2`



Variables

Name	Label
id2	group(country year)
country	Country
year	
x1	var1
x2	var2



Stata Do-File Editor - varname.do

File Edit Search Tools

Untitled1.do varname.do

```
label variable x1 `"'var1"'`  
label variable x2 `"'var2"'`
```

The final dataset will look like...

	id2	country	year	x1	x2
1	1	A	1995	.	.
2	2	A	1996	.	.
3	3	A	1997	8000.01	6.83
4	4	A	1998	8212.9	2.66
5	5	A	1999	7847.36	-4.45
6	6	A	2000	7702.89	-1.84
7	7	A	2001	7288.48	-5.38
8	8	A	2002	6430.98	-11.77
9	9	A	2003	6932.45	7.8
10	10	A	2004	7486.24	7.99
11	11	A	2005	8094.17	8.12
12	12	B	1995	18268.01	2.87
13	13	B	1996	18738.99	2.58
14	14	B	1997	19360.46	3.32
15	15	B	1998	20151.42	4.09
16	16	B	1999	20715.54	2.8
17	17	B	2000	20866.9	.73
18	18	B	2001	21364.02	2.38
19	19	B	2002	21801.41	2.05
20	20	B	2003	22404.59	2.77
21	21	B	2004	22676.26	1.21

Reshape long to wide (Stata, 1)

You want to go from...

id	time	r
1	1	2
1	2	7
1	3	1
2	1	3
2	2	5
2	3	9
3	1	1
3	2	1
3	3	8

`reshape wide r, i(id) j(time)`



to...

id	r.time1	r.time2	r.time3
1	2	7	1
2	3	5	9
3	1	1	8

EXAMPLE: If you have a dataset like this one ([click here to get it](#)), we need to change the date variable as follows:

```
tostring month year, replace
gen date=year+"_0"+month if length(month)==1
replace date=year+"_"+month if date==" "
drop year month
order id date
```



Variables		X
Name	Label	
id		
date		
return		
interest		

	id	year	month	return	interest
1	105.1	2002	11	1.307071	.87494
2	105.1	2002	12	1.403008	1.019082
3	105.1	2003	1	1.570926	1.152942
4	105.1	2003	2	1.894784	1.307366
5	105.1	2003	3	1.798847	1.235295
6	105.1	2003	4	1.7628	1.173506
7	105.1	2003	5	2.026655	1.297084
8	105.1	2003	6	2.302488	1.708849
9	105.1	2003	7	2.968058	1.749977
10	105.1	2003	8	3.027948	2.161742
11	105.1	2003	9	3.117896	2.238954
12	105.1	2003	10	5.036636	2.753636
13	105.1	2003	11	5.000024	3.542246
14	105.1	2003	12	7.469865	4.266157
15	105.1	2004	1	8.072268	5.145268
16	105.1	2004	2	7.95181	5.015967
17	105.1	2004	3	8.192726	5.843377
18	105.1	2004	4	8.493984	4.395458
19	105.1	2004	5	5.843343	3.542246
20	105.1	2004	6	5.458126	3.602774
21	105.1	2004	7	5.456205	3.911331

Reshape long to wide (Stata, 2)

The data will look like...



	id	date	return	interest
1	105.1	2002_11	1.307071	.87494
2	105.1	2002_12	1.403008	1.019082
3	105.1	2003_01	1.570926	1.152942
4	105.1	2003_02	1.894784	1.307366
5	105.1	2003_03	1.798847	1.235295
6	105.1	2003_04	1.7628	1.173506
7	105.1	2003_05	2.026655	1.297084
8	105.1	2003_06	2.302488	1.708849
9	105.1	2003_07	2.968058	1.749977
10	105.1	2003_08	3.027948	2.161742
11	105.1	2003_09	3.117896	2.238954
12	105.1	2003_10	5.036636	2.753636
13	105.1	2003_11	5.000024	3.542246
14	105.1	2003_12	7.469865	4.266157
15	105.1	2004_01	8.072268	5.145268
16	105.1	2004_02	7.95181	5.015967
17	105.1	2004_03	8.192726	5.843377
18	105.1	2004_04	8.493984	4.395458
19	105.1	2004_05	5.843343	3.542246
20	105.1	2004_06	5.458126	3.602774
21	105.1	2004_07	5.456205	3.911331

To reshape type

```
reshape wide return interest, i(id) j(date) str
```

```
. reshape wide return interest, i(id) j(date) str
(note: j = 1998_11 1998_12 1999_01 1999_02 1999_03 1999_04 1999_05 1999_06 1999_07 1999_08 1999_09 1999_10 1999_11
> 1999_12 2000_01 2000_02 2000_03 2000_04 2000_05 2000_06 2000_07 2000_08 2000_09 2000_10 2000_11 2000_12 2001_01 2
> 001_02 2001_03 2001_04 2001_05 2001_06 2001_07 2001_08 2001_09 2001_10 2001_11 2001_12 2002_01 2002_02 2002_03 20
> 02_04 2002_05 2002_06 2002_07 2002_08 2002_09 2002_10 2002_11 2002_12 2003_01 2003_02 2003_03 2003_04 2003_05 200
> 3_06 2003_07 2003_08 2003_09 2003_10 2003_11 2003_12 2004_01 2004_02 2004_03 2004_04 2004_05 2004_06 2004_07 2004
> _08 2004_09 2004_10 2004_11 2004_12 2005_01 2005_02 2005_03 2005_04 2005_05 2005_06 2005_07 2005_08 2005_09 2005_
> 10 2005_11 2005_12 2007_01 2007_02 2007_03 2007_04 2007_05 2007_06 2007_07 2007_08 2007_09 2007_10 2007_11)
```

Data	long	->	wide
<hr/>			
Number of obs.	802	->	25
Number of variables	4	->	195
j variable (97 values)	date	->	(dropped)
xij variables:			
	return	->	return1998_11 return1998_12 ... return2007_11
	interest	->	interest1998_11 interest1998_12 ... interest2007_11

Where:

- wide** – Indicates the type of reshape, in this case from long to wide format.
- return interest** – The variables of interest from long to wide are “return” and “interest” (prefix for the new variables).
- i(id)** – A unique identifier for the wide format is in variable “id”.
- j(date)** – Indicates the suffix of “return” and “interest” taken from “date” (notice “xij” variables:” above)

Reshape long to wide (Stata, 3)

The variable window and the data will look like

Variables	
Name	Label
id	
return1998_11	1998_11 return
interest1998_11	1998_11 interest
return1998_12	1998_12 return
interest1998_12	1998_12 interest
return1999_01	1999_01 return
interest1999_01	1999_01 interest
return1999_02	1999_02 return
interest1999_02	1999_02 interest
return1999_03	1999_03 return
interest1999_03	1999_03 interest
return1999_04	1999_04 return
interest1999_04	1999_04 interest
return1999_05	1999_05 return
interest1999_05	1999_05 interest

	id	return1998~1	interes~8_11	return1998~2	interes~8_12	return199~01
1	105.1
2	121.1	3.4126	2.592616	3.108856	2.331589	3.139705
3	143.1
4	161.2
5	162.1
6	162.2
7	167.1	19.20548	15.14606	18.16995	13.73898	18.71529

If you want to sort all returns and interest together, run the following commands:

```
xpose, clear varname
sort _varname
xpose, clear
order id
```



Variables	
Name	Label
id	
interest1998_11	
interest1998_12	
interest1999_01	
interest1999_02	
interest1999_03	
interest1999_04	
interest1999_05	
interest1999_06	
interest1999_07	
interest1999_08	
interest1999_09	
interest1999_10	
interest1999_11	
interest1999_12	

Renaming variables (using `renvars`)

You can use the command `renvars` to shorten the names of the variables...

```
renvars interest1998_11-interest2007_11, presub(interest i)
```

```
renvars return1998_11-return2007_11, presub(return r)
```

Before

Variables	
Name	Label
id	
interest1998_11	
interest1998_12	
interest1999_01	
interest1999_02	
interest1999_03	
interest1999_04	
interest1999_05	
interest1999_06	
interest1999_07	
interest1999_08	
interest1999_09	
interest1999_10	
interest1999_11	
interest1999_12	



After

Variables	
Name	Label
id	
i1998_11	
i1998_12	
i1999_01	
i1999_02	
i1999_03	
i1999_04	
i1999_05	
i1999_06	
i1999_07	
i1999_08	
i1999_09	
i1999_10	
i1999_11	
i1999_12	

NOTE: You may have to install `renvars` by typing:

```
ssc install renvars
```

Type `help renvars` for more info. Also `help rename`

Descriptive statistics (definitions)

Descriptive statistics are a collection of measurements of two things: *location* and *variability*.

Location tells you the central value of your variable (the mean is the most common measure).

Variability refers to the spread of the data from the center value (i.e. variance, standard deviation).

Statistics is basically the study of what causes variability in the data.

Location	Variability
Mean	Variance
Mode	Standard deviation
Median	Range

Descriptive statistics (location)...

Indicator	Definition	Formula	In Excel	In Stata	In R
Location					
Mean	The mean is the sum of the observations divided by the total number of observations. It is the most common indicator of central tendency of a variable	$\bar{X} = \frac{\sum X_i}{n}$	=AVERAGE(range of cells) For example: =AVERAGE(J2:J31)	-tabstat var1, s(mean) or - sum var1	summary(x) mean(x) supply(x, mean, na.rm=T)
Median	The median is another measure of central tendency. To get the median you have to order the data from lowest to highest. The median is the number in the middle. If the number of cases is odd the median is the single value, for an even number of cases the median is the average of the two numbers in the middle. It is not affected by outliers. Also known as the 50 th percentile. <div style="text-align: center;"> 2 6 <u>7</u> 8 9 2 6 <u>7 8</u> 9 10 </div>		=MEDIAN(range of cells)	- tabstat var1, s(median) or - sum var1, detail	summary(x) median(x) supply(x, median, na.rm=T) #median
Mode	The mode refers to the most frequent, repeated or common number in the data		=MODE(range of cells)	mmodes var1	table(x) (frequency table)

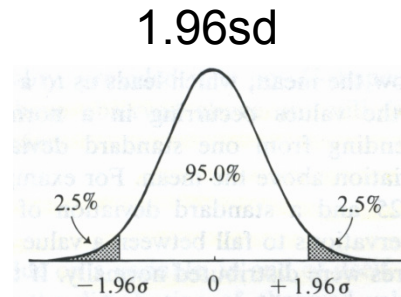
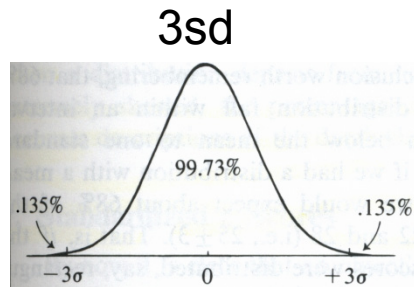
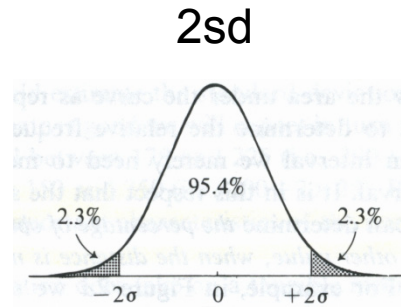
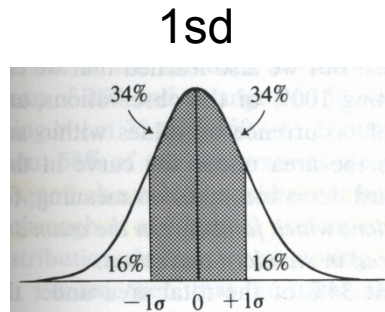
NOTE: For `mmodes` you may have to install it by typing `ssc install mmodes`. You can estimate all statistics in Excell using “Descriptive Statistics” in “Analysis Toolpack”. In Stata by typing all statistics in the parenthesis `tabstat var1, s(mean median)`. In R see http://www.ats.ucla.edu/stat/r/faq/basic_desc.htm

Descriptive statistics (variability)...

Indicator	Definition	Formula	In Excel	In Stata	In R
Variability					
Variance	<p>The variance measures the dispersion of the data from the mean.</p> <p>It is the simple mean of the squared distance from the mean.</p>	$s^2 = \frac{\sum (X_i - \bar{X})^2}{(n-1)}$	=VAR(range of cells)	- tabstat var1, s(variance) or - sum var1, detail	var(x) sapply(x, var, na.rm=T)
Standard deviation	<p>The standard deviation is the squared root of the variance. Indicates how close the data is to the mean. Assuming a normal distribution:</p> <ul style="list-style-type: none"> • 68% of the values are within 1 sd (.99) • 95% within 2 sd (1.96) • 99% within 3 sd (2.58). 	$s = \sqrt{\frac{\sum (X_i - \bar{X})^2}{(n-1)}}$	=STDEV(range of cells)	- tabstat var1, s(sd) or - sum var1, detail	sd(x) sapply(x, sd, na.rm=T)
Range	Range is a measure of dispersion. It is simple the difference between the largest and smallest value, “max” – “min”.		=MAX(range of cells) - MIN(same range of cells)	tabstat var1, s(range)	range=(max(x)-min(x));range

NOTE: You can estimate all statistics in Excell using “Descriptive Statistics” in “Analysis Toolpack”. In Stata by typing all statistics in the parenthesis `tabstat var1, s(mean median variance sd range)`. In R see

Descriptive statistics (standard deviation)



Source: Kachigan, Sam K., *Statistical Analysis. An Interdisciplinary Introduction to Univariate & Multivariate Methods*, 1986, p.61

Descriptive statistics (z-scores)...

z-scores show how many standard deviations a single value is from the mean. Having the mean is not enough.

$$z = \frac{x_i - \mu}{\sigma}$$

Student	x_i	Mean SAT score	sd	z-score	% (below)	%(above)
A	1842	1849	275	-0.03	49.0%	51.0%
B	1907	1849	275	0.21	58.4%	41.6%
C	2279	1849	275	1.56	94.1%	5.9%

Student	x_i	Mean SAT score	sd	z-score	% (below)	%(above)
A	1842	1849	162	-0.04	48.3%	51.7%
B	1907	1849	162	0.36	64.0%	36.0%
C	2279	1849	162	2.65	99.6%	0.4%

Student	x_i	Mean SAT score	sd	z-score	% (below)	%(above)
A	1855	1858	162	-0.02	49.3%	50.7%
B	1917	1858	162	0.36	64.2%	35.8%
C	2221	1858	162	2.24	98.7%	1.3%

NOTE: To get the %(below) you can use the tables at the end of any statistics book or in Excel use =normsdist(z-score) . %(above) is just 1-%(below) .

In Stata type:

```
egen z_var1=std(var1)
gen below=normal(z_var1)
gen above=1-below
```


Descriptive statistics (distribution)...

Indicator	Definition	Formula	In Excel	In Stata	In R
Standard error (deviation) of the mean	Indicates how close the sample mean is from the 'true' population mean. It increases as the variation increases and it decreases as the sample size goes up. It provides a measure of uncertainty.	$SE_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$	=(STDEV(range of cells))/(SQRT(COUNT(sample range of cells))).	tabstat var1, s(emean)	sem=sd(x)/sqrt(length(x)); sem
Confidence intervals for the mean	The range where the 'true' value of the mean is likely to fall most of the time	$CI_{\bar{X}} = \bar{X} \pm SE_{\bar{X}} * Z$	Use "Descriptive Statistics" in the "Data Analysis" tab (1)	ci var1	Use package "pastecs"
Skewness	Measures the symmetry of the distribution (whether the mean is at the center of the distribution). The skewness value of a normal distribution is 0. A negative value indicates a skew to the left (left tail is longer than the right tail) and a positive value indicates a skew to the right (right tail is longer than the left one)	$Sk = \frac{\sum (X_i - \bar{X})^3}{(n-1)s^3}$	=SKEW(range of cells)	-tabstat var1, s(skew) - sum var1, detail	Custom estimation
Kurtosis	Measures the peakedness (or flatness) of a distribution. A normal distribution has a value of 3. A kurtosis >3 indicates a sharp peak with heavy tails closer to the mean (leptokurtic). A kurtosis < 3 indicates the opposite a flat top (platykurtic).	$K = \frac{\sum (X_i - \bar{X})^4}{(n-1)s^4}$	=KURT(range of cells)	-tabstat var1, s(k) - sum var1, detail	Custom estimation kurtosis(x)

Notation:

X_i = individual value of X

\bar{X} = mean of X

n = sample size

s^2 = variance

s = standard deviation

$SE_{\bar{X}}$ = standard error of the mean

Z = critical value (Z=1.96 give a 95% certainty)

For more info check the module "Descriptive Statistics with Excel/Stata" in <http://dss.princeton.edu/training/>

Confidence intervals...

Confidence intervals are ranges where the true mean is expected to lie.

Student	x_i	Mean SAT score	sd	N	SE	Lower(95%)	Upper(95%)
A	1842	1849	275	30	50	1751	1947
B	1907	1849	275	30	50	1751	1947
C	2279	1849	275	30	50	1751	1947

Student	x_i	Mean SAT score	sd	N	SE	Lower(95%)	Upper(95%)
A	1842	1849	162	30	30	1791	1907
B	1907	1849	162	30	30	1791	1907
C	2279	1849	162	30	30	1791	1907

Student	x_i	Mean SAT score	sd	N	SE	Lower(95%)	Upper(95%)
A	1855	1858	162	30	30	1800	1916
B	1917	1858	162	30	30	1800	1916
C	2221	1858	162	30	30	1800	1916

lower(95%) = (Mean SAT score) – (SE*1.96)

upper(95%) = (Mean SAT score) + (SE*1.96)

Coefficient of variation (CV)...

Measure of dispersion, helps compare variation across variables with different units. A variable with higher coefficient of variation is more dispersed than one with lower CV.

	A	B	B/A
	Mean	Standard Deviation	Coefficient of variation
<i>Age (years)</i>	25	6.87	27%
<i>SAT</i>	1849	275.11	15%
<i>Average score (grade)</i>	80	10.11	13%
<i>Height (in)</i>	66	4.66	7%
<i>Newspaper readership (times/wk)</i>	5	1.28	26%

CV works only with variables with positive values.

[Click here to get the table](#)

Examples (Excel)

	A	B	C	D	E	F	G	H	I	J	K	L	M	N
	ID	Last Name	First Name	City	State	Gender	Student Status	Major	Country	Age	SAT	Average score (grade)	Height (in)	Newspaper readership (times/wk)
1														
2	1	DOE01	JANE01	Los Angeles	California	Female	Graduate	Politics	US	30	2263	67	61	5
3	2	DOE02	JANE02	Sedona	Arizona	Female	Undergraduate	Math	US	19	2006	63	64	7
4	3	DOE01	JOE01	Elmira	New York	Male	Graduate	Math	US	26	2221	78	73	6
5	4	DOE02	JOE02	Lackawana	New York	Male	Graduate	Econ	US	33	1716	78	68	3
6	5	DOE03	JOE03	Defiance	Ohio	Male	Graduate	Econ	US	37	1701	65	71	6
7	6	DOE04	JOE04	Tel Aviv	Israel	Male	Graduate	Econ	Israel	25	1786	69	67	5
8	7	DOE05	JOE05	Cimax	North Carolina	Male	Graduate	Politics	US	39	1577	96	70	5
9	8	DOE03	JANE03	Liberal	Kansas	Female	Undergraduate	Politics	US	21	1842	87	62	5
10	9	DOE04	JANE04	Montreal	Canada	Female	Undergraduate	Math	Canada	18	1813	91	62	6
11	10	DOE05	JANE05	New York	New York	Female	Graduate	Math	US	33	2041	71	66	5
12	11	DOE06	JOE06	Hot Coffe	Mississippi	Male	Undergraduate	Econ	US	18	1787	82	67	3
13	12	DOE06	JANE06	Java	Virginia	Female	Graduate	Math	US	38	1513	79	59	5
14	13	DOE07	JOE07	Varna	Bulgaria	Male	Graduate	Politics	Bulgaria	30	1637	79	63	4
15	14	DOE08	JOE08	Moscow	Russia	Male	Graduate	Politics	Russia	30	1512	70	75	6
16	15	DOE07	JANE07	Drunkard Creek	New York	Female	Undergraduate	Math	US	21	1338	82	64	5
17	16	DOE08	JANE08	Mexican Hat	Utah	Female	Undergraduate	Econ	US	18	1821	80	63	3
18	17	DOE09	JANE09	Amsterdam	Holland	Female	Undergraduate	Math	Holland	19	1494	75	60	3
19	18	DOE10	JANE10	Mexico	Mexico	Female	Graduate	Politics	Mexico	31	2248	95	59	4
20	19	DOE11	JANE11	Caracas	Venezuela	Female	Undergraduate	Math	Venezuela	18	2252	92	68	5
21	20	DOE09	JOE09	San Juan	Puerto Rico	Male	Graduate	Politics	US	33	1923	95	63	7
22	21	DOE12	JANE12	Remote	Oregon	Female	Undergraduate	Econ	US	19	1727	67	62	7
23	22	DOE10	JOE10	New York	New York	Male	Undergraduate	Econ	US	21	1872	82	73	4
24	23	DOE13	JANE13	The X	Massachusetts	Female	Graduate	Politics	US	25	1767	89	68	6
25	24	DOE14	JANE14	Beijing	China	Female	Undergraduate	Math	China	18	1643	79	65	6
26	25	DOE11	JOE11	Stockholm	Sweden	Male	Undergraduate	Politics	Sweden	19	1919	88	64	4
27	26	DOE12	JOE12	Embarrass	Minnesota	Male	Graduate	Econ	US	28	1434	96	71	4
28	27	DOE13	JOE13	Intercourse	Pennsylvania	Male	Undergraduate	Math	US	20	2119	88	71	5
29	28	DOE15	JANE15	Loco	Oklahoma	Female	Undergraduate	Econ	US	20	2309	64	68	6
30	29	DOE14	JOE14	Buenos Aires	Argentina	Male	Graduate	Politics	Argentina	30	2279	85	72	3
31	30	DOE15	JOE15	Acme	Louisiana	Male	Undergraduate	Econ	US	19	1907	79	74	3

	Age	SAT	Average score (grade)	Height (in)	Newspaper readership (times/wk)
Mean	25.2	Mean	1848.9	Mean	66.43333333
Standard Error	1.254325848	Standard Error	50.22838301	Standard Error	0.850535103
Median	23	Median	1817	Median	66.5
Mode	19	Mode	#N/A	Mode	68
Standard Deviation	6.870225615	Standard Deviation	275.112184	Standard Deviation	4.658572619
Sample Variance	47.2	Sample Variance	75686.71379	Sample Variance	21.70229885
Kurtosis	-1.049751548	Kurtosis	-0.846633469	Kurtosis	-1.066828463
Skewness	0.557190515	Skewness	0.155667999	Skewness	0.171892733
Range	21	Range	971	Range	16
Minimum	18	Minimum	1338	Minimum	59
Maximum	39	Maximum	2309	Maximum	75
Sum	756	Sum	55467	Sum	1993
Count	30	Count	30	Count	30

Use “Descriptive Statistics” in the “Data Analysis” tab.

Examples (Stata)

[Click here to get the table](#)

	id	lastname	firstname	city	state	gender	studentstatus	major	country	age	sat	averagescore	heightin	newspaperread
1	1	DOE01	JANE01	Los Angeles	California	Female	Graduate	Politics	US	30	2263	67	61	5
2	2	DOE02	JANE02	Sedona	Arizona	Female	Undergraduate	Math	US	19	2006	63	64	7
3	3	DOE01	JOE01	Elmira	New York	Male	Graduate	Math	US	26	2221	78	73	6
4	4	DOE02	JOE02	Lackawana	New York	Male	Graduate	Econ	US	33	1716	78	68	3
5	5	DOE03	JOE03	Defiance	Ohio	Male	Graduate	Econ	US	37	1701	65	71	6
6	6	DOE04	JOE04	Tel Aviv	Israel	Male	Graduate	Econ	Israel	25	1786	69	67	5
7	7	DOE05	JOE05	Cimax	North Carolina	Male	Graduate	Politics	US	39	1577	96	70	5
8	8	DOE03	JANE03	Liberal	Kansas	Female	Undergraduate	Politics	US	21	1842	87	62	5
9	9	DOE04	JANE04	Montreal	Canada	Female	Undergraduate	Math	Canada	18	1813	91	62	6
10	10	DOE05	JANE05	New York	New York	Female	Graduate	Math	US	33	2041	71	66	5
11	11	DOE06	JOE06	Hot Coffe	Mississippi	Male	Undergraduate	Econ	US	18	1787	82	67	3
12	12	DOE06	JANE06	Java	Virginia	Female	Graduate	Math	US	38	1513	79	59	5
13	13	DOE07	JOE07	Varna	Bulgaria	Male	Graduate	Politics	Bulgaria	30	1637	79	63	4
14	14	DOE08	JOE08	Moscow	Russia	Male	Graduate	Politics	Russia	30	1512	70	75	6
15	15	DOE07	JANE07	Drunkard Creek	New York	Female	Undergraduate	Math	US	21	1338	82	64	5
16	16	DOE08	JANE08	Mexican Hat	Utah	Female	Undergraduate	Econ	US	18	1821	80	63	3
17	17	DOE09	JANE09	Amsterdam	Holland	Female	Undergraduate	Math	Holland	19	1494	75	60	3
18	18	DOE10	JANE10	Mexico	Mexico	Female	Graduate	Politics	Mexico	31	2248	95	59	4
19	19	DOE11	JANE11	Caracas	Venezuela	Female	Undergraduate	Math	Venezuela	18	2252	92	68	5
20	20	DOE09	JOE09	San Juan	Puerto Rico	Male	Graduate	Politics	US	33	1923	95	63	7
21	21	DOE12	JANE12	Remote	Oregon	Female	Undergraduate	Econ	US	19	1727	67	62	7
22	22	DOE10	JOE10	New York	New York	Male	Undergraduate	Econ	US	21	1872	82	73	4
23	23	DOE13	JANE13	The X	Massachusetts	Female	Graduate	Politics	US	25	1767	89	68	6
24	24	DOE14	JANE14	Beijing	China	Female	Undergraduate	Math	China	18	1643	79	65	6
25	25	DOE11	JOE11	Stockholm	Sweden	Male	Undergraduate	Politics	Sweden	19	1919	88	64	4
26	26	DOE12	JOE12	Embarrass	Minnesota	Male	Graduate	Econ	US	28	1434	96	71	4
27	27	DOE13	JOE13	Intercourse	Pennsylvania	Male	Undergraduate	Math	US	20	2119	88	71	5
28	28	DOE15	JANE15	Loco	Oklahoma	Female	Undergraduate	Econ	US	20	2309	64	68	6
29	29	DOE14	JOE14	Buenos Aires	Argentina	Male	Graduate	Politics	Argentina	30	2279	85	72	3
30	30	DOE15	JOE15	Acme	Louisiana	Male	Undergraduate	Econ	US	19	1907	79	74	3

stats	age	sat	score	heightin	read
mean	25.2	1848.9	80.36667	66.43333	4.866667
se(mean)	1.254326	50.22838	1.846079	.8505351	.2335795
p50	23	1817	79.5	66.5	5
sd	6.870226	275.1122	10.11139	4.658573	1.279368
variance	47.2	75686.71	102.2402	21.7023	1.636782
skewness	.5289348	.1477739	-.1017756	.1631759	-.049278
kurtosis	1.923679	2.094488	1.966325	1.909319	1.988717
N	30	30	30	30	30
sum	756	55467	2411	1993	146
range	21	971	33	16	4
min	18	1338	63	59	3
max	39	2309	96	75	7

```

rename averagescoregrade score
rename newspaperreadshiptimeswk read

tabstat age sat score heightin read,
s(mean semean median sd var skew k
count sum range min max )

```

Examples (R)

```
students<-read.table("H:/students.csv", sep=";", header=T)
```

```
> students
```

	ID	last	first	City	State	Gender	status	Major	Country	Age	SAT	score	height	read
1	1	DOE01	JANE01	Los Angeles	California	Female	Graduate	Politics	US	30	2263	67	61	5
2	2	DOE02	JANE02	Sedona	Arizona	Female	Undergraduate	Math	US	19	2006	63	64	7
3	3	DOE01	JOE01	Elmira	New York	Male	Graduate	Math	US	26	2221	78	73	6
4	4	DOE02	JOE02	Lackawana	New York	Male	Graduate	Econ	US	33	1716	78	68	3
5	5	DOE03	JOE03	Defiance	Ohio	Male	Graduate	Econ	US	37	1701	65	71	6
6	6	DOE04	JOE04	Tel Aviv	Israel	Male	Graduate	Econ	Israel	25	1786	69	67	5
7	7	DOE05	JOE05	Cimax	North Carolina	Male	Graduate	Politics	US	39	1577	96	70	5
8	8	DOE03	JANE03	Liberal	Kansas	Female	Undergraduate	Politics	US	21	1842	87	62	5
9	9	DOE04	JANE04	Montreal	Canada	Female	Undergraduate	Math	Canada	18	1813	91	62	6
10	10	DOE05	JANE05	New York	New York	Female	Graduate	Math	US	33	2041	71	66	5
11	11	DOE06	JOE06	Hot Coffe	Mississippi	Male	Undergraduate	Econ	US	18	1787	82	67	3
12	12	DOE06	JANE06	Java	Virginia	Female	Graduate	Math	US	38	1513	79	59	5
13	13	DOE07	JOE07	Varna	Bulgaria	Male	Graduate	Politics	Bulgaria	30	1637	79	63	4
14	14	DOE08	JOE08	Moscow	Russia	Male	Graduate	Politics	Russia	30	1512	70	75	6
15	15	DOE07	JANE07	Drunkard Creek	New York	Female	Undergraduate	Math	US	21	1338	82	64	5
16	16	DOE08	JANE08	Mexican Hat	Utah	Female	Undergraduate	Econ	US	18	1821	80	63	3
17	17	DOE09	JANE09	Amsterdam	Holland	Female	Undergraduate	Math	Holland	19	1494	75	60	3
18	18	DOE10	JANE10	Mexico	Mexico	Female	Graduate	Politics	Mexico	31	2248	95	59	4
19	19	DOE11	JANE11	Caracas	Venezuela	Female	Undergraduate	Math	Venezuela	18	2252	92	68	5
20	20	DOE09	JOE09	San Juan	Puerto Rico	Male	Graduate	Politics	US	33	1923	95	63	7
21	21	DOE12	JANE12	Remote	Oregon	Female	Undergraduate	Econ	US	19	1727	67	62	7
22	22	DOE10	JOE10	New York	New York	Male	Undergraduate	Econ	US	21	1872	82	73	4
23	23	DOE13	JANE13	The X	Massachusetts	Female	Graduate	Politics	US	25	1767	89	68	6
24	24	DOE14	JANE14	Beijing	China	Female	Undergraduate	Math	China	18	1643	79	65	6
25	25	DOE11	JOE11	Stockholm	Sweden	Male	Undergraduate	Politics	Sweden	19	1919	88	64	4
26	26	DOE12	JOE12	Embarrass	Minnesota	Male	Graduate	Econ	US	28	1434	96	71	4
27	27	DOE13	JOE13	Intercourse	Pennsylvania	Male	Undergraduate	Math	US	20	2119	88	71	5
28	28	DOE15	JANE15	Loco	Oklahoma	Female	Undergraduate	Econ	US	20	2309	64	68	6
29	29	DOE14	JOE14	Buenos Aires	Argentina	Male	Graduate	Politics	Argentina	30	2279	85	72	3
30	30	DOE15	JOE15	Acme	Louisiana	Male	Undergraduate	Econ	US	19	1907	79	74	3

```
> |
```

```
> library(pastecs)
```

```
Loading required package: boot
```

```
> stat.desc(students[10:14])
```

	Age	SAT	score	height	read
nbr.val	30.000000	3.000000e+01	30.0000000	3.000000e+01	30.0000000
nbr.null	0.000000	0.000000e+00	0.0000000	0.000000e+00	0.0000000
nbr.na	0.000000	0.000000e+00	0.0000000	0.000000e+00	0.0000000
min	18.000000	1.338000e+03	63.0000000	5.900000e+01	3.0000000
max	39.000000	2.309000e+03	96.0000000	7.500000e+01	7.0000000
range	21.000000	9.710000e+02	33.0000000	1.600000e+01	4.0000000
sum	756.000000	5.546700e+04	2411.0000000	1.993000e+03	146.0000000
median	23.000000	1.817000e+03	79.5000000	6.650000e+01	5.0000000
mean	25.200000	1.848900e+03	80.3666667	6.643333e+01	4.8666667
SE.mean	1.254326	5.022838e+01	1.8460790	8.505351e-01	0.2335795
CI.mean.0.95	2.565384	1.027286e+02	3.7756555	1.739540e+00	0.4777237
var	47.200000	7.568671e+04	102.2402299	2.170230e+01	1.6367816
std.dev	6.870226	2.751122e+02	10.1113911	4.658573e+00	1.2793677
coef.var	0.272628	1.487978e-01	0.1258157	7.012402e-02	0.2628838

```
install.packages("pastecs")
```



```
> |
```

Useful links / Recommended books/References

- DSS Online Training Section <http://dss.princeton.edu/training/>
- UCLA Resources <http://www.ats.ucla.edu/stat/>
- DSS help-sheets for STATA http://dss/online_help/stats_packages/stata/stata.htm
- *Introduction to Stata* (PDF), Christopher F. Baum, Boston College, USA. “A 67-page description of Stata, its key features and benefits, and other useful information.”
<http://fmwww.bc.edu/GStat/docs/StataIntro.pdf>
- STATA FAQ website <http://stata.com/support/faqs/>
- Princeton DSS Libguides <http://libguides.princeton.edu/dss>

Books

- *Introduction to econometrics* / James H. Stock, Mark W. Watson. 2nd ed., Boston: Pearson Addison Wesley, 2007.
- *Data analysis using regression and multilevel/hierarchical models* / Andrew Gelman, Jennifer Hill. Cambridge ; New York : Cambridge University Press, 2007.
- *Econometric analysis* / William H. Greene. 6th ed., Upper Saddle River, N.J. : Prentice Hall, 2008.
- *Designing Social Inquiry: Scientific Inference in Qualitative Research* / Gary King, Robert O. Keohane, Sidney Verba, Princeton University Press, 1994.
- *Unifying Political Methodology: The Likelihood Theory of Statistical Inference* / Gary King, Cambridge University Press, 1989
- *Statistical Analysis: an interdisciplinary introduction to univariate & multivariate methods* / Sam Kachigan, New York : Radius Press, c1986
- *Statistics with Stata (updated for version 9)* / Lawrence Hamilton, Thomson Books/Cole, 2006