

### **Homework1 Questions – (Answers at the end)**

Please answer the following questions using Stata and submit your output. You may find the **log** command useful for saving your output.

The charts can be either copied and pasted into MS Word (right click on the chart for the options), or saved on disk (.png format is recommended) and inserted into MS Word manually, through Insert->Picture menu. The latter method is sometimes more reliable, while the former one is faster.

1. Compare the distributions of the book/market ratio, market capitalization, dividend/price ratio, price/earnings between 1998 and 1990:
  - a. Compute summary statistics for these variables across all companies. Compare with the results for 1990.
  - b. Create 10 level groups, based on market capitalization. Compute average book/market ratio, dividend/price ratio, price/earnings ratio for each group. Create bar charts (see tutorial for an example).
  - c. Comment on differences between 1990 and 1998.
2. Consider a strategy of buying stock of companies with high book/market ratio and selling stock of those with low such ratio. Compute the profits such a strategy would have produced during 1999-2003.
3. List all of the most closely related companies in 1998 (`hsicig=739`). Display the company name, ticker, main line of business, book value, market capitalization and past year return.
4. Repeat the analysis in p.1 for the entire business services industry alone in 1990 and 1998 (`hsicmg=73`). What can you tell about these companies? How has the industry changed since 1990, compared to the rest of the economy? Hint: to avoid tedious repeating of the **if** conditions, you may check the Stata help on **keep** and **drop** commands and utilize one of them.
5. Make a scatter plot of the earnings/price ratio and market value of the companies with `hsicig==739`. Put ticker marks on the chart. Do any companies stand out? What can you tell about HGC?

## Stata tutorial

### **Contact Information**

With all questions concerning Stata, this tutorial and exercise questions, or other related issues, please contact Andrei Jirnyi, [a-jirnyi@northwestern.edu](mailto:a-jirnyi@northwestern.edu), phone (847) 491-4270(o), (847)858-2537(c).

### **Why Stata?**

There are many ways to do numerical analysis. Most of you are probably familiar with Microsoft Excel. Why learn another tool, and what advantages does it have?

1. Fast and easy data transformation
2. Fast and easy data analysis
3. Ability to handle large datasets
4. Easy automation of repeated operations
5. A large number of built-in statistical procedures

### **How to get Stata**

Kellogg has established a Stata "GradPlan" -- this is an agreement with StataCorp that allows Kellogg faculty, students and staff to buy Stata at the lowest possible cost. The details can be found at <http://www.kellogg.northwestern.edu/rc/sec/manuals/stata/stata-gradplan.htm>. Intercooled Stata is the recommended version, its' price is \$145.

The software is also available on the PCs located in the computing labs on the Evanston and Chicago campuses. Intercooled Stata version 8.2 is installed in the "special software" workstations at the Jacobs Center (ten workstations) and Chicago campus (six workstations) computer laboratories.

### **Some preliminary steps**

#### **Conventions in this document**

In this handout, I'll put a dot in front of Stata commands in this text, and display them in **bold courier**, like this:

```
.help
```

Note that you do not need to type the dot!

The parameters of the commands are sometimes given in text in *italic*, required parameters in **<angle brackets>**, and optional ones in **[square brackets]**. The actual Stata output is given in plain courier.

## Stata windows

Once you start Stata for Windows, you'll see four main windows:

1. **Stata Results:** (on Windows typically with black background): this is where you can see the estimation results and all other output that you ask Stata to provide.
2. **Stata Command:** this is where you enter all commands
3. **Review:** this window keeps a complete history of the commands you've entered
4. **Variables:** once the dataset is loaded, all the available variables are listed here.

The commands you enter are repeated in the results window in white color, while Stata output is typically in green and yellow.

## The working directory

Whenever you ask Stata to read or write something on disk, it will try to do it in your *working directory* (unless otherwise specified). To see the current working directory, you may type the following command:

```
.pwd
```

You probably want to change this directory to the one you keep the data in, to do so use a **cd** command:

```
.cd "c:\aj\work\value-investing"
```

both forward (/) and reverse (\) slashes work, and the quotation marks are required only when you have spaces in the path name.

To see the contents of the directory, use **ls** or **dir**:

```
.ls
```

```
<dir> 3/14/06 17:27 .
<dir> 3/14/06 17:27 ..
4040.7k 3/13/06 12:54 piotroski00.pdf
 5.8k 3/10/06 12:59 spore_stata.csv
28.0k 3/14/06 17:27 statatut.doc
```

To get help on a command, type **.man <command>**

```
.man ls
```

Some of you might have recognized these commands from DOS or UNIX. In fact, many other DOS and UNIX commands work in Stata, such as **copy**, **erase**, **mkdir**, **cat**, **type**, **shell**.

## ***Keeping track of your work***

One of the advantages Stata and other similar software has over the menu-driven systems, such as spreadsheets, is the ability to keep track of all your work and of all the results that you've obtained, and to quickly and easily repeat it all on a different dataset or with some other modifications.

There are two main commands that I recommend that you do at the very beginning of your session: **log** and **cmdlog**:

```
.log using session.log  
.cmdlog using commands.txt
```

The first will create a file named `session.log` in your home directory<sup>1</sup>. This file is a simple text file, that could be opened in any text editor, such as Microsoft Word or Notepad.

The second command will create a file `commands.txt`, containing all the commands that you have entered. This file can later be easily edited into a Stata script for automatic execution.

Whenever you do something that you would like to make a particular note of in the script or the output file, you may just enter a comment in the command line by typing an asterisk (\*) first:

```
.* This is a comment
```

This can greatly simplify future parsing of the command file in particular.

Once your work is done, you should close the log files by typing

```
.log close  
.cmdlog close
```

## ***Loading the data into Stata***

The first step in any analysis is to read the dataset. There are many data formats, such as Excel spreadsheets, text files, or SAS datasets.

This time we'll start with an existing Stata dataset, `nyse90.dta`. It contains data on accounting information and average future 5-year returns on publicly traded stocks with a closing price of \$5 or above on December 31, 1990. Please download the file from the class website and save it into your working directory. After that, start Stata and load the dataset by typing the following command:

```
.*Loading the data...  
.clear all  
.set mem 100m  
.use stocks90
```

---

<sup>1</sup> It is essential to explicitly specify the extension `.log` for the output log – otherwise Stata will create a file `session.smcl` in its' own format, that can only be viewed from within Stata by entering a command **view session.smcl**.

The first command clears any possibly open datasets from memory, so that you can load a new one. This step is particularly recommended to include in any script files that you may write.

Once the data is loaded, inspect it by choosing **Data->Data Editor** in the main Stata menu. You'll see a spreadsheet, much like an Excel one, of all the data in memory. Each row represents the data for one company in 1990. From now on I'll refer to the rows of data as "observations". It is a fairly large data set including many variables that we will not use in this tutorial.

If you scroll around the spreadsheet a bit, you might notice that some apparently numerical variables have a value of "." in the cell. This means that this particular variable is missing for this company. For example, we do not have data on 1990 earnings for M.L. Macadamia Orchards, LP.

## Summarizing the data

First, we would like to list the variables:

. d

Contains data from stocks90.dta

```
obs:      1,811
vars:      352                               29 Mar 2006 07:45
size:      2,571,620 (97.5% of memory free)
```

variable name	storage type	display format	value label	variable label
name	str32	%32s		company name
ticker	str5	%5s		exchange ticker symbol
permno	long	%12.0g		crsp permanent number
gvkey	long	%12.0g		standard and poor's identifier
hsiccd	int	%8.0g		header sic code
hsicig	int	%8.0g		header sic industry group
ret90	float	%9.0g		annual return on common stock in 1990
retstd	float	%9.0g		holding period return
ret5	float	%9.0g		average annual return on common stock for 1991-1995
vol	double	%12.0g		share volume
shrout	double	%12.0g		shares outstanding
prc	float	%9.0g		closing price or bid/ask average
linkdt	int	%dD_m_Y		first effective date of link
linkenddt	int	%dD_m_Y		last effective date of link
fyr	byte	%8.0g		fiscal yearend month of data
data1	float	%9.0g		cash and short-term investments (mm\$)
data2	double	%10.0g		receivables - total (mm\$)
data3	float	%9.0g		inventories - total (mm\$)
assets_cur	float	%9.0g		current assets - total (mm\$)
.....				
be	float	%9.0g		book equity
me	float	%9.0g		market equity
mv	float	%9.0g		
z50	float	%9.0g		
ebi	float	%9.0g		
z19	float	%9.0g		
ecomm	float	%9.0g		
bme	float	%9.0g		book/market equity ratio
hsicmg	float	%9.0g		header sic major group

Sorted by:

At the beginning we can see some summary information on this data file. We know that it is named `stocks90.dta`, that it has 1811 observations and 352 variables (or, in other words, that it contains 352 data items, describing 1811 different companies), and that it occupies about 2.5M of memory.

Below we can see, among other things, the *variable names* – short names for different data items, that we'll use to refer to them in Stata commands, and *variable labels* – short descriptions of each item. We'll soon learn how to create both.

We do not need to go to the spreadsheet data browser to get a quick look at the data. Just use the **.list** command (usually abbreviated as **.l**):

```
. l name ticker assets_tot sales_net in 1/5
```

	name	ticker	assets~t	sales_~t
1.	CHIQUITA BRANDS INTL INC	CQB	2174.437	4272.66
2.	DEKALB GENETICS CORP	SEEDB	238.1	274
3.	DOLE FOOD INC	DOL	2423.27	3003.213
4.	M L MACADAMIA ORCHARDS LP	NUT	74.567	14.186
5.	PIONEER HI BRED INTL INC	PHB	1005.992	964.453

(your output may look a little differently, depending on the size of your results window). This command lists several columns

A few things to note here:

- **"in 1/3"** at the end instructed Stata to only use the observations 1 through 3
- Some variable names have been shortened to fit into the table (`sales_~t` instead of `sales_net`). When you are entering Stata commands, it is sufficient to enter only as many characters as to uniquely specify the variable:

```
. l name ticker sales in 1/10
```

	name	ticker	sales_~t
1.	CHIQUITA BRANDS INTL INC	CQB	4272.66
2.	DEKALB GENETICS CORP	SEEDB	274
3.	DOLE FOOD INC	DOL	3003.213
4.	M L MACADAMIA ORCHARDS LP	NUT	14.186
5.	PIONEER HI BRED INTL INC	PHB	964.453
6.	SERVICEMASTER CO	SVM	1825.75
7.	I P TIMBERLANDS LTD	IPT	195.461
8.	RAYONIER TIMBERLANDS L P	LOG	107.562
9.	B H P LTD	BHP	11853
10.	CLEVELAND CLIFFS INC	CLF	309.9

Let us also see some statistics:

`. su`

Variable	Obs	Mean	Std. Dev.	Min	Max
name	0				
ticker	0				
permno	1811	48897.9	22724.1	10064	93105
gvkey	1811	10164.17	11280.82	1004	160785
hsiccd	1811	4751.088	1892.273	0	9511
hsicig	1811	474.7929	189.21	0	951
ret90	1811	-.0621237	.5000388	-.8105354	14.3767
retstd	1804	.0999556	.0467829	.0071527	.4288884
ret5	1809	.1655348	.222925	-.7883611	4.006406
vol	1811	2.156981	3.939127	.0009	47.0501
shrout	1811	48.89816	94.83598	.073	1246.185
prc	1811	26.7208	157.644	5.125	6675
.....					
be	1584	1372.901	3349.589	-2587.88	46693
me	1753	1794.967	4814.561	8.870875	64567.18
mv	1566	6104.499	16204.48	12.584	211465.5
z50	1804	-.7642056	75.11449	-2424.8	671
ebi	1361	271.226	771.631	-1390.251	11600
z19	1810	3.975649	19.92937	-.526	559
ecomm	1576	137.8852	439.4368	-4448.7	6433
bme	1566	.9123388	.6511382	-5.262994	6.456691
hsicmg	1811	47.13418	18.93703	0	95

`.su` is a shortcut for `.summarize` here. The first column of the output gives the number of *non-missing numerical observations* (note that `ticker` and `name` have 0 in it since they are not numerical), followed by sample statistics.

We can do some very quick analysis already now. For example, one may be interested in what was the average relationship between past 1-year return and future 5-year returns in 1990. For a brief look, let us display a correlation table, using `.cor`:

`. cor ret90 ret5`  
(obs=1809)

	ret90	ret5
ret90	1.0000	
ret5	-0.0548	1.0000

We can see that the correlation is negative (-5%) here – that is, companies, whose stock did poorly in 1990, on average did better for the following 5 years.

## Generating new variables

While this file contains a few useful data items, we would like to look at other numbers as well – in particular, at some useful financial ratios. Just in Excel we would create new columns, in Stata to do that we need to create new variables. This is done with a command `.generate`.

```
. gen eps=earn/shrout
(269 missing values generated)
. gen epr=eps/prc
(269 missing values generated)
. gen dp=div/prc
(43 missing values generated)
```

We have just created five new variables: `eps`<sup>2</sup>, `epr`, and `dp`. Note that in all cases some missing values were created. This happened because we do not have data on all items for all companies. For example, because we do not have total earnings for NUT, we were not able to compute earnings per share either.

```
. l ticker earn shrout eps in 1/5
```

```
+-----+
| ticker   earn   shrout   eps   |
+-----+-----+-----+-----+
1. |   CQB   348.996  45.219  7.717906 |
2. |  SEEDB    20.7   5.581  3.709013 |
3. |   DOL   448.802  59.27  7.572161 |
4. |   NUT     .     7.5   . |
5. |   PHB   646.341  30.924  20.90095 |
+-----+-----+-----+-----+
```

Now our dataset contains 355 variables:

```
. d
Contains data from stocks90.dta
obs:           1,811
vars:           355                29 Mar 2006 07:45
size:         2,593,352 (97.5% of memory free)
```

```
-----+-----+-----+-----+
variable name  storage  display  value  variable label
              type   format   label
-----+-----+-----+-----+
name           str32   %32s                company name
ticker         str5    %5s                exchange ticker symbol
permno         long    %12.0g             crsp permanent number
.....
eps           float   %9.0g
epr           float   %9.0g
dp            float   %9.0g
-----+-----+-----+-----+
```

```
Sorted by:
Note: dataset has changed since last saved
```

---

<sup>2</sup> One thing that we need to watch out for here is measurement units: in our case both earnings are recorded in millions of dollars, and the number of shares outstanding in millions of shares, so it is OK to divide them. If, say, number of shares were recorded in thousands, we'd have to multiply it by 1000.

The five new variables are at the end. Note that unlike other variables, they do not yet have labels. It's a good idea to add them right now, so that we won't forget what they are. This is done with `.label var` command:

```
. label var eps "Earnings per share"  
. label var epr "Earnings to price ratio"  
. label var dp "Dividend to price ratio"
```

```
.d
```

Contains data from stocks90.dta

```
obs:      1,811  
vars:      355                29 Mar 2006 07:45  
size:     2,593,352 (97.5% of memory free)
```

```
-----  
variable name  storage  display  value  variable label  
               type    format    label  
-----  
name           str32   %32s          company name  
ticker         str5    %5s          exchange ticker symbol  
permno        long   %12.0g       crsp permanent number
```

```
.....
```

```
eps           float  %9.0g       Earnings per share  
epr           float  %9.0g       Earnings to price ratio  
dp            float  %9.0g       Dividend to price ratio
```

```
-----  
Sorted by:
```

```
Note: dataset has changed since last saved
```

The note in the last line reminds us that we have modified the dataset. It might be a good idea to `.save` it now, so that the work we've just done wouldn't be lost in case something unexpected happens:

```
.save, replace
```

The `[,replace]` option specifies that if a file with the current name already exists, it would be overwritten. To save it under a different name (e.g. `nyse90_new`) instead, we'd have to specify it like this:

```
.save stocks90_new, replace
```

### A few more statistics

Now, once we created our new variables, let us take a look at them. Here are some useful variables, including those we just computed:

```
. d bme me dp epr eps
```

variable name	storage type	display format	value label	variable label
bme	float	%9.0g		book/market equity ratio
me	float	%9.0g		market equity
dp	float	%9.0g		Dividend to price ratio
epr	float	%9.0g		Earnings to price ratio
eps	float	%9.0g		Earnings per share

First, compute some summary statistics:

```
. su bme me dp epr eps
```

Variable	Obs	Mean	Std. Dev.	Min	Max
bme	1566	.9123388	.6511382	-5.262994	6.456691
me	1753	1794.967	4814.561	8.870875	64567.18
dp	1768	.048625	.1420018	0	4.351515
epr	1542	2.000054	24.07292	-27.17606	765.8135
eps	1542	55.94497	632.4668	-268.3636	19911.15

Now let us check how they are related to each other and to returns. First, compute a correlation matrix:

```
. cor ret5 ret90 epr me bme dp  
(obs=1525)
```

	ret5	ret90	epr	me	bme	dp
ret5	1.0000					
ret90	-0.0769	1.0000				
epr	-0.0113	-0.0137	1.0000			
me	-0.0397	0.0919	0.0624	1.0000		
bme	0.1374	-0.3386	0.0385	-0.1394	1.0000	
dp	0.0026	-0.0376	0.0035	-0.0117	0.0630	1.0000

First, one thing you probably noticed that the correlation between `ret5` and `ret90` has changed – it is now -7.7%. This happened because all companies that had missing values in *any* of the variables were excluded from computations here.

Other than that, we can see that there is a small negative correlation between future returns and E/P ratio and the market value. On the other hand, companies with a higher D/P and especially B/M ratios did better than others.

One problem with these simple statistics is that they do not tell us much about how significant the relationships are, how large are the deviations from the average, etc.

We can see that there are a few large outliers that confuse the picture a little. To make it a little more transparent, we can break the companies into several groups, based on their B/M ratio, market capitalization and other variables:

### **Regression analysis**

To evaluate the significance of the potential relationships, let us run a linear regression of future returns on potential predictor variables:

```
. regress ret5 ret90 epr me bme dp
```

Source	SS	df	MS			
Model	1.69165501	5	.338331003	Number of obs =	1525	
Residual	80.568012	1519	.053040166	F( 5, 1519) =	6.38	
				Prob > F =	0.0000	
				R-squared =	0.0206	
				Adj R-squared =	0.0173	
				Root MSE =	.2303	
Total	82.259667	1524	.053976159			

ret5	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ret90	-.0200095	.0161318	-1.24	0.215	-.0516525	.0116336
epr	-.0001481	.0002445	-0.61	0.545	-.0006276	.0003314
me	-8.32e-07	1.17e-06	-0.71	0.476	-3.12e-06	1.46e-06
bme	.0444557	.009725	4.57	0.000	.0253798	.0635315
dp	-.0102525	.0393889	-0.26	0.795	-.0875149	.06701
_cons	.1304556	.0109711	11.89	0.000	.1089356	.1519757

The strongest and most significant predictor of future returns was clearly the B/M ratio. But, while other variables are not very significant in such a regression, this does not necessarily mean that a strategy, based on them, would not be profitable. One thing that a regression like this one fails to account for is the potential non-linear relationships between returns and predicting variables.

### **Portfolio construction**

To further explore the potential relationships, let us construct simulation portfolios, based on our potential variables. To do so, we shall rank the stocks in accordance with the value of each such variable and compute the returns to a strategy of buying potential winners and selling short potential losers. Let us start with the book/market ratio, as the most potentially significant predictor.

First, let us create a *categorical variable* that forms 10 ranked groups of stocks, based on the book/market ratio. So far we've seen only one way of constructing new variables, that by using **gen.** Another such command is **xtile**:

```
. xtile me10=me, nq(10)
```

This command creates a new variable `me10`, based on values of `me`, grouped into ten ("**q(10)**") groups. This last statement is an *option* for a command – an additional instruction on how exactly to do the

computations. The 10 groups are created to be approximately equal in size, to check that, let us tabulate the values of this new variable:

```
. tab me10
```

10 quantiles of me	Freq.	Percent	Cum.
1	176	10.04	10.04
2	175	9.98	20.02
3	175	9.98	30.01
4	176	10.04	40.05
5	175	9.98	50.03
6	175	9.98	60.01
7	176	10.04	70.05
8	175	9.98	80.03
9	175	9.98	90.02
10	175	9.98	100.00
Total	1,753	100.00	

The new variable takes values from 1 to 10, indicating that the corresponding stock belongs to that particular ranked group. Let us also check how the average values of different variables differ across these 10 groups:

```
. tabstat eps epr mktval bme dp ret5, s(mean) by(me10)
```

Summary statistics: mean  
 by categories of: me10 (10 quantiles of me)

me10	eps	epr	mktval~s	bme	dp	ret5
1	6.791798	.6965072	54.43373	1.387439	.0722566	.219612
2	6.202881	.4946279	77.86297	1.052264	.0488254	.1886405
3	7.762573	.551816	122.7272	1.041249	.0533842	.1833389
4	9.226702	.5470097	150.6228	.8518151	.0402081	.1514274
5	7.394273	.3534503	205.8057	.7774228	.0558928	.1510664
6	11.4845	.5100749	386.4259	.9004726	.0537308	.1481498
7	9.05238	.3185214	589.5316	.886191	.0423823	.1346627
8	30.48633	1.324385	1031.318	.8770901	.0432942	.1924418
9	73.11799	3.636735	1936.98	.8267867	.0419831	.1585319
10	341.6869	9.889199	6106.672	.6531582	.0371551	.1376024
Total	56.45648	2.01975	1169.715	.9123388	.0489159	.1665508

The **tabstat** command computes the statistic, specified in the option “s()”, for each value of a categorical variable, specified in “by()”.

We can see that the average future return on common stock for the smallest companies is 22%, much higher than that of the largest companies (13.7%). Note that these average returns can be interpreted as returns on an equally-weighted portfolio of all stocks in that group.

Similarly, we can create groups based, for example, on the book/market ratio, and compute similar statistics there:

```
. tabstat eps epr mktval bme dp ret5, s(mean) by(bme10)
```

Summary statistics: mean  
by categories of: bme10 (10 quantiles of bme)

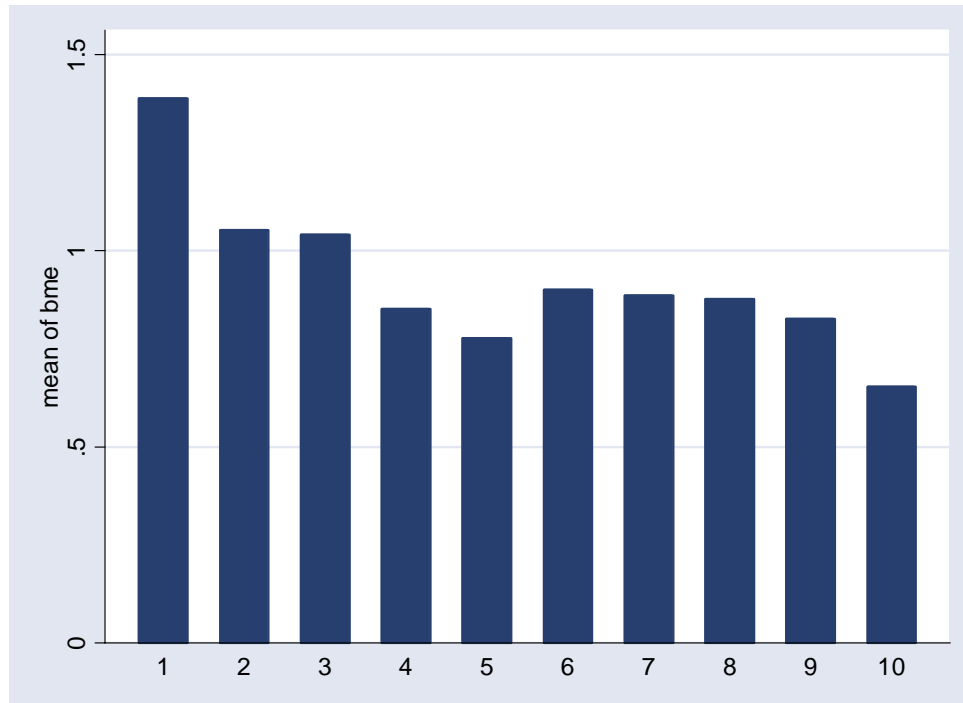
bme10	eps	epr	mktval~s	bme	dp	ret5
1	10.54549	.1328178	610.1211	.0863635	.0429207	.1347002
2	13.84976	.2927945	684.6179	.3563609	.0183622	.1117098
3	71.83744	2.400372	1109.809	.481466	.0255522	.1372841
4	57.49137	1.728079	1147.17	.612406	.0249824	.1524704
5	129.1861	4.570877	1930.534	.7276851	.0442441	.2038519
6	69.98866	1.585766	1449.554	.8738779	.0661087	.1737286
7	23.81782	1.114036	908.648	1.02227	.0567815	.1698623
8	11.39654	.4380677	1050.622	1.22347	.0481891	.1733748
9	21.31581	1.215803	1559.935	1.487134	.0547627	.2073983
10	154.0027	6.683219	1250.361	2.258993	.0706334	.235436
Total	56.45648	2.01975	1169.715	.9123388	.0452488	.1699369

Here we can see that the average future return on common stock for the companies with the highest B/M ratio is 23.5%, much higher than that of the lowest group (13.5%).

### ***A summary bar plot***

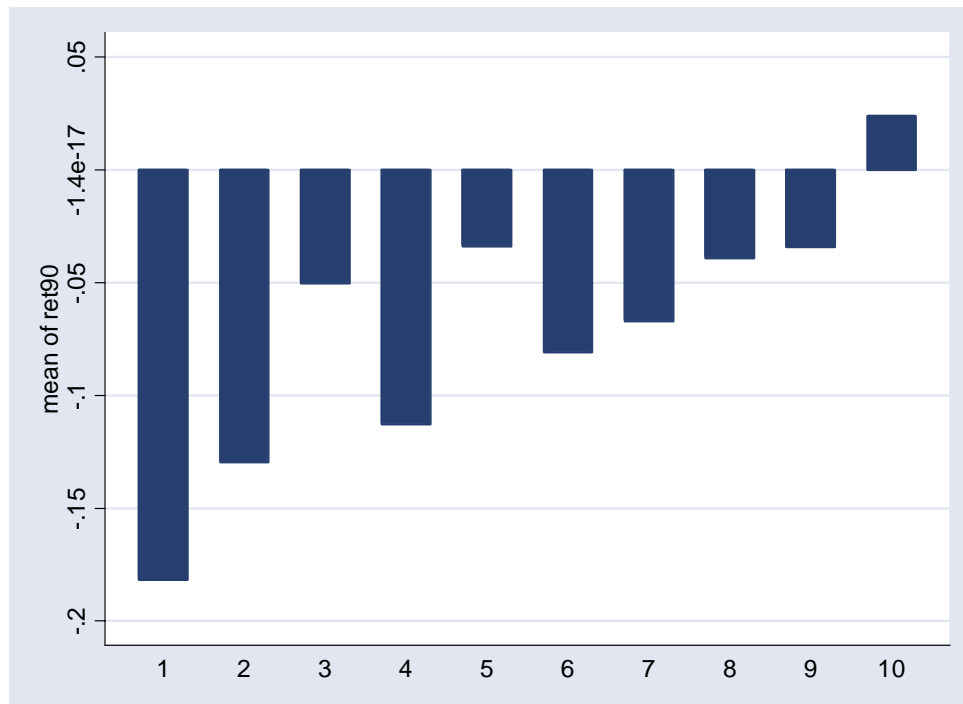
Of course, the book/market ratio and the market cap are correlated, as one can easily see from the summary tables we produced earlier. We can also look at their relationship on a plot:

```
. gr bar (mean) bme, over(me10)
```



The **gr bar (mean)** command automatically computed the mean of bme and plotted it over all the categories of me10. As expected, we see a negative relationship, especially pronounced at the lowest market cap levels: these are the stocks that have recently experienced poor returns.

```
. gr bar (mean) ret90, over(me10)
```



## Hudson General analysis

Now let us look at some statistics for Hudson General.

First, load the data file for 1998:

```
.use all198.dta
```

Generate some additional statistics:

```
. gen eps=earn/shrout
. gen epr=eps/prc
. gen dp=div/prc
```

Take a look at Hudson General (HGC):

```
. l name ticker iname hsiccd if ticker=="HGC"
```

	name	ticker	iname	hsiccd
19.	HUDSON GENERAL CORP	HGC	AIRPORTS & TERMINAL SERVICES	7394

Here, the **if ticker=="HGC"** statement has restricted the **list** command to the observations that satisfy the condition (note that to specify an equality, the "=" sign has to be repeated). `hsiccd` is the SIC industry code of the company, in this case 7394. The first three digits of this code (739) specify an industry group, and the first two (73) specify a "major industry group", in this case Business Services. Let us see what other companies are there in the same industry group:

```
. l name ticker iname if hsicig==739
```

	name	ticker	iname
1.	I B P INC	IBP	MEAT PACKING PLANTS
19.	HUDSON GENERAL CORP	HGC	AIRPORTS & TERMINAL SERVICES
304.	LASON INC	LSON	CMP PROGRAMMING,DATA PROCESS
372.	GEOTEL COMMUNICATIONS CORP	GEOC	PREPACKAGED SOFTWARE
565.	ARCADIS NV	ARCAF	ENGINEERING SERVICES

Now let's see some statistics on those companies:

```
. l ticker eps epr mktval bme dp ret98 ret5 if hsicig==739
```

	ticker	eps	epr	mktval	bme	dp	ret98	ret5
1.	IBP	11.32418	.3888131	2703.79	.5023689	.0034335	.3974356	-.0685974
19.	HGC	16.75358	.2659299	109.935	.7996095	.015873	.3359825	2.08204
304.	LSON	2.082292	.0357859	902.9536	.335761	0	1.185446	-.854272
372.	GEOC	.3261969	.008757	1009.14	.0556112	0	3.768	1.603755
565.	ARCAF	3.313642	.42078	148.6406	.6761166	.0392889	-.2409649	.127984

We can also display data on a scatter plot, for example like this:

```
. scatter bme me if hsicig==739, mlabel(ticker)
```



The last parameter **"mlabel(ticker)"** instructed Stata to place near each data point a label – in this case, the company's ticker symbol.

## Homework1 Answers

```
. /* Part 1a */
. su bme me dp epr
```

Variable	Obs	Mean	Std. Dev.	Min	Max
bme	5240	.5686128	.4472267	-2.075402	5.267616
me	5731	2907.297	13112.76	4.20525	333672.2
dp	5930	.0185453	.0484703	0	2.271852
epr	5274	1.230452	16.39802	-55.38419	668.0734

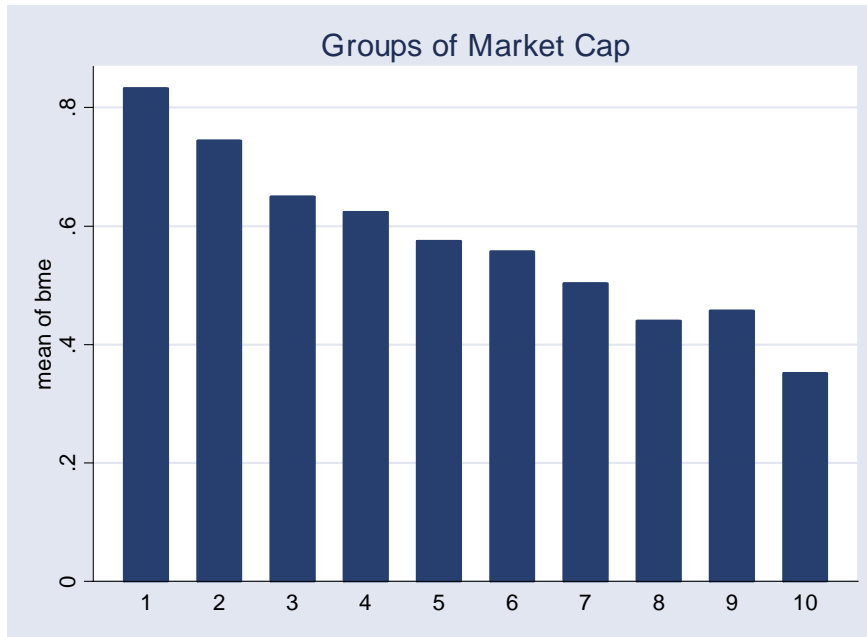
```
. /* Part 1b */
. xtile me10=me, nq(10)

. tabstat eps epr mktval bme dp, s(mean) by(me10)
```

Summary statistics: mean  
 by categories of: me10 (10 quantiles of me)

me10	eps	epr	mktval	bme	dp
1	3.08183	.2147178	32.36216	.8327547	.0215133
2	2.176366	.1434744	65.96726	.7439956	.0178496
3	2.215206	.1106453	102.8396	.6500231	.0208976
4	3.090467	.1480313	154.3239	.6234538	.0188834
5	4.386719	.2248164	238.7836	.5744724	.0177955
6	5.530824	.2404611	364.5323	.5568566	.0185632
7	5.938522	.2599116	580.7353	.5036616	.0181217
8	17.85195	.8504423	1042.505	.4400238	.0170886
9	123.1223	4.060059	2276.379	.4573416	.0156977
10	286.6091	5.458061	18484.25	.3514821	.0170179
Total	48.78835	1.248927	2333.867	.5686128	.0183414

```
.gr bar (mean) bme, over(me10) title("Groups of Market Cap.")
```



```

. /* Part 2 */
. xtile bme10=bme, nq(10)

. tabstat ret5 eps epr mktval bme dp, s(mean) by(bme10)
    
```

Summary statistics: mean  
 by categories of: bme10 (10 quantiles of bme)

bme10	ret5	eps	epr	mktval	bme	dp
1	.1114811	-9.46297	-.4852172	7784.574	-.0065247	.0059963
2	.0213247	5.851604	.0932791	5261.255	.1719405	.0065261
3	.0553223	34.56831	.768163	2896.32	.2665535	.0064367
4	.0162675	40.98703	.8380865	3362.296	.3572687	.0096505
5	.1016196	26.10176	.5733465	1616.691	.4517523	.0109517
6	.172112	102.555	3.134946	1051.511	.5468151	.0142634
7	2.736393	76.46099	.9267213	1045.78	.6534455	.0165595
8	.2607621	44.77619	1.512045	719.8845	.7879832	.0203103
9	.1332738	115.362	3.086587	661.424	.9677376	.0263889
10	.149084	37.04397	1.737113	383.291	1.489156	.027886
Total	.3756401	47.51699	1.220517	2478.303	.5686128	.0145021

```

. /* Part 3 */
. l name ticker iname be me eps if hsicig==739
    
```

	name	ticker	iname	be	me	eps
171.	I B P INC	IBP	MEAT PACKING PLANTS	1350.692	2688.645	11.32418
1532.	HUDSON GENERAL CORP	HGC	AIRPORTS & TERMINAL SERVICES	70.638	88.34062	16.75358
5199.	LASON INC	LSON	CMP PROGRAMMING,DATA PROCESS	287.039	854.8907	2.082292
5280.	GEOTEL COMMUNICATIONS CORP	GEOC	PREPACKAGED SOFTWARE	55.757	1002.621	.3261969
6025.	ARCADIS NV	ARCAF	ENGINEERING SERVICES	102.527	151.641	3.313642

```
. /* Part 4 */
.
. save nyse98a, replace
file nyse98a.dta saved

. keep if hsicmg==73
(5463 observations deleted)

.
. /* Part 4.1a */
. su bme me dp epr
```

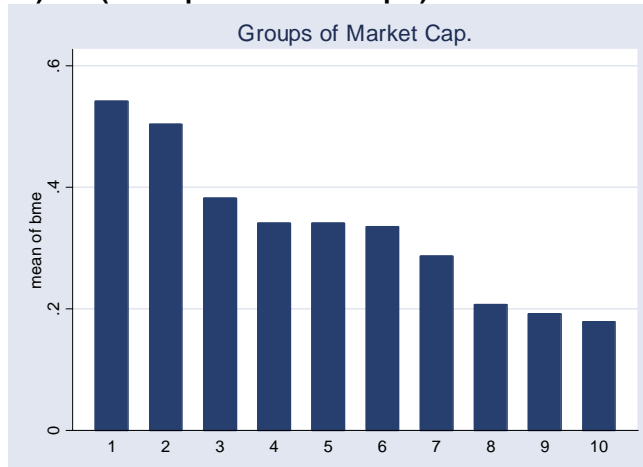
Variable	Obs	Mean	Std. Dev.	Min	Max
bme	536	.3317828	.3033717	-.3734393	2.601382
me	539	1850.572	12264.12	9.25	267686.3
dp	558	.001977	.0110414	0	.1877807
epr	546	.4061718	9.494692	-55.38419	202.2024

```
. /* Part 4.1b */
. xtile me10=me, nq(10)
. tabstat eps epr mktval bme dp, s(mean) by(me10)
```

Summary statistics: mean  
 by categories of: me10 (10 quantiles of me)

me10	eps	epr	mktval	bme	dp
1	.1309977	-.0085133	34.65505	.5413646	.0008217
2	.4968812	-.0254595	75.62775	.5034658	.003104
3	-.026774	-.0303351	125.114	.3820832	.0007111
4	-.7313988	-.0809841	186.313	.3404816	.0008445
5	2.345342	.1149984	275.2439	.3406725	.0002394
6	4.770126	.0891176	376.9049	.3350131	.0010716
7	.2362984	-.0891747	557.8234	.2871116	.0031214
8	.8485455	.0244454	1010.179	.2074182	.0012348
9	.6089149	.0107427	1914.736	.1919285	.0010746
10	304.5731	5.233272	15175.14	.1789721	.0061833
Total	30.87461	.5160783	1948.681	.3317828	.0018326

```
.gr bar (mean) bme, over(me10) title("Groups of Market Cap.")
```



```
. /* Part 5 */  
. scatter epr me if hsicig==739, mlabel(ticker)
```

