

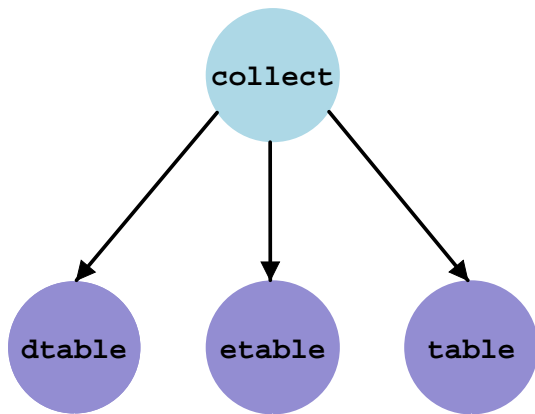
# Customizable Tables with Stata

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German Stata Conference

- Creating customized tables
  - Motivating examples
  - Creating tables of summary statistics with `dtable`
  - Creating tables of regression results with `etable`
  - Creating cross-tabulations and more with `table`
  - Creating fully customized tables using the `collect` suite
- Incorporating tables in complete reports



## You can create a table of summary statistics,

Table 1: Summary statistics

	Hypertension		Total
	No	Yes	
Age (years)	42.2 (16.8)	55.0 (14.9)	47.6 (17.2)
Body mass index (BMI)	24.2 (4.1)	27.4 (5.3)	25.5 (4.9)
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	4,915 (47.5%)
Female	3,364 (56.3%)	2,072 (47.3%)	5,436 (52.5%)
Race			
White	5,317 (89.0%)	3,748 (85.6%)	9,065 (87.6%)
Black	545 (9.1%)	541 (12.4%)	1,086 (10.5%)
Other	113 (1.9%)	87 (2.0%)	200 (1.9%)
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	2,407 (23.3%)
Very good	1,666 (27.9%)	925 (21.2%)	2,591 (25.1%)
Good	1,572 (26.4%)	1,366 (31.2%)	2,938 (28.4%)
Fair	766 (12.8%)	904 (20.7%)	1,670 (16.2%)
Poor	310 (5.2%)	419 (9.6%)	729 (7.1%)
Serum cholesterol (mg/dL)	208.7 (47.3)	229.9 (49.6)	217.7 (49.4)
Serum triglycerides (mg/dL)	129.2 (83.9)	166.0 (109.2)	143.9 (96.5)
High density lipids (mg/dL)	49.9 (14.1)	49.2 (14.5)	49.6 (14.3)

Mean and standard deviation reported for continuous variables.

And export your table to a Word document,

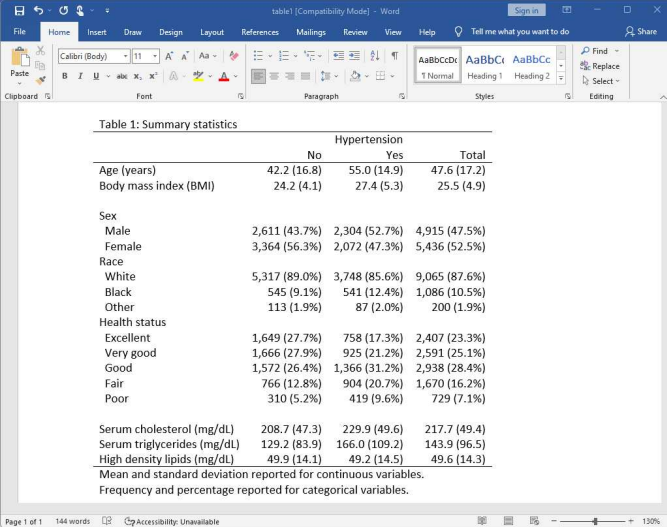


Table 1: Summary statistics

	Hypertension		Total
	No	Yes	
Age (years)	42.2 (16.8)	55.0 (14.9)	47.6 (17.2)
Body mass index (BMI)	24.2 (4.1)	27.4 (5.3)	25.5 (4.9)
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	4,915 (47.5%)
Female	3,364 (56.3%)	2,072 (47.3%)	5,436 (52.5%)
Race			
White	5,317 (89.0%)	3,748 (85.6%)	9,065 (87.6%)
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Poor	310 (5.2%)	419 (9.6%)	729 (7.1%)
Serum cholesterol (mg/dL)	208.7 (47.3)	229.9 (49.6)	217.7 (49.4)
Serum triglycerides (mg/dL)	129.2 (83.9)	166.0 (109.2)	143.9 (96.5)
High density lipids (mg/dL)	49.9 (14.1)	49.2 (14.5)	49.6 (14.3)

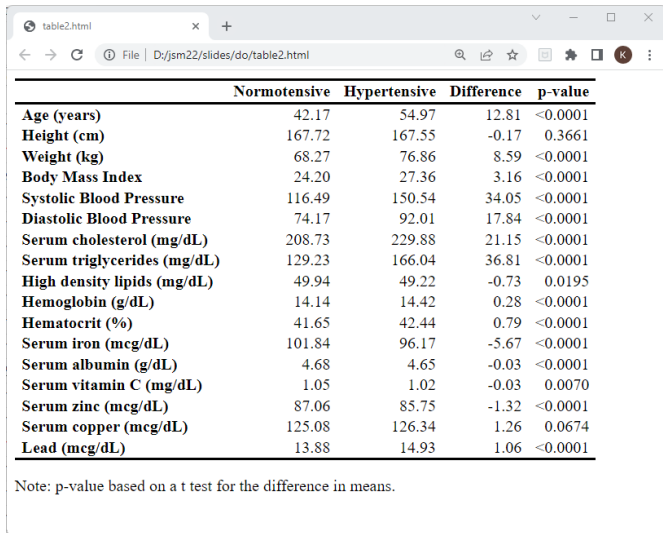
Mean and standard deviation reported for continuous variables.  
Frequency and percentage reported for categorical variables.

You can create a table of means and  $t$  tests of differences,

	Normotensive	Hypertensive	Difference	p-value
Age (years)	42.17	54.97	12.81	<0.0001
Height (cm)	167.72	167.55	-0.17	0.3661
Weight (kg)	68.27	76.86	8.59	<0.0001
Body Mass Index	24.20	27.36	3.16	<0.0001
Systolic Blood Pressure	116.49	150.54	34.05	<0.0001
Diastolic Blood Pressure	74.17	92.01	17.84	<0.0001
Serum cholesterol (mg/dL)	208.73	229.88	21.15	<0.0001
Serum triglycerides (mg/dL)	129.23	166.04	36.81	<0.0001
High density lipids (mg/dL)	49.94	49.22	-0.73	0.0195
Hemoglobin (g/dL)	14.14	14.42	0.28	<0.0001
Hematocrit (%)	41.65	42.44	0.79	<0.0001
Serum iron (mcg/dL)	101.84	96.17	-5.67	<0.0001
Serum albumin (g/dL)	4.68	4.65	-0.03	<0.0001
Serum vitamin C (mg/dL)	1.05	1.02	-0.03	0.0070
Serum zinc (mcg/dL)	87.06	85.75	-1.32	<0.0001
Serum copper (mcg/dL)	125.08	126.34	1.26	0.0674
Lead (mcg/dL)	13.88	14.93	1.06	<0.0001

Note: p-value based on a  $t$  test for the difference in means.

And export your table to HTML,



	<b>Normotensive</b>	<b>Hypertensive</b>	<b>Difference</b>	<b>p-value</b>
<b>Age (years)</b>	42.17	54.97	12.81	<0.0001
<b>Height (cm)</b>	167.72	167.55	-0.17	0.3661
<b>Weight (kg)</b>	68.27	76.86	8.59	<0.0001
<b>Body Mass Index</b>	24.20	27.36	3.16	<0.0001
<b>Systolic Blood Pressure</b>	116.49	150.54	34.05	<0.0001
<b>Diastolic Blood Pressure</b>	74.17	92.01	17.84	<0.0001
<b>Serum cholesterol (mg/dL)</b>	208.73	229.88	21.15	<0.0001
<b>Serum triglycerides (mg/dL)</b>	129.23	166.04	36.81	<0.0001
<b>High density lipids (mg/dL)</b>	49.94	49.22	-0.73	0.0195
<b>Hemoglobin (g/dL)</b>	14.14	14.42	0.28	<0.0001
<b>Hematocrit (%)</b>	41.65	42.44	0.79	<0.0001
<b>Serum iron (mcg/dL)</b>	101.84	96.17	-5.67	<0.0001
<b>Serum albumin (g/dL)</b>	4.68	4.65	-0.03	<0.0001
<b>Serum vitamin C (mg/dL)</b>	1.05	1.02	-0.03	0.0070
<b>Serum zinc (mcg/dL)</b>	87.06	85.75	-1.32	<0.0001
<b>Serum copper (mcg/dL)</b>	125.08	126.34	1.26	0.0674
<b>Lead (mcg/dL)</b>	13.88	14.93	1.06	<0.0001

Note: p-value based on a t test for the difference in means.

## You can create a table of regression results,

Table 3: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Female		0.617 (0.028)	0.152 (0.023)
Female x Age (years)			1.028 (0.003)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00

Odds ratio (Std. Err.)



And export your table to a PDF,

table3.pdf - Adobe Acrobat Reader DC (64-bit)

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Table 3: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Female		0.617 (0.028)	0.152 (0.023)
Female x Age (years)			1.028 (0.003)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00
Odds ratio (Std. Err.)			

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We can use the `dtable`, `etable`, `table`, and `collect` commands to create tables of

- Summary statistics, including a classic Table 1
- Results of classical hypothesis tests
- Cross-tabulations
- Regression results
- Postestimation tests
- Combinations of the above
- Results returned by any Stata commands

We can customize tables by changing

- Table layout
- Numeric formats
- Labels appearing on rows and columns
- Stars and other added text
- Font type, size, and color
- Shading, borders, margins, alignment, and more

We can export the customized tables to

- Word
- Excel
- L<sup>A</sup>T<sub>E</sub>X
- PDF
- Markdown
- HTML
- SMCL
- Plain text

# dtable

The `dtable` command produces tables of summary statistics.

It can report separate summary statistics for groups and perform tests comparing groups.

To demonstrate, we will use NHANES II data.

```
. webuse nhanes2l, clear
(Second National Health and Nutrition Examination Survey)
. describe age sex race height weight bmi highbp
>          bpsystol bpdiaast tcresult hdresult
```

Variable name	Storage type	Display format	Value label	Variable label
age	byte	%9.0g		Age (years)
sex	byte	%9.0g	sex	Sex
race	byte	%9.0g	race	Race
height	float	%9.0g		Height (cm)
weight	float	%9.0g		Weight (kg)
bmi	float	%9.0g		Body mass index (BMI)
highbp	byte	%8.0g		* High blood pressure
bpsystol	int	%9.0g		Systolic blood pressure
bpdiaast	int	%9.0g		Diastolic blood pressure
tcresult	int	%9.0g		Serum cholesterol (mg/dL)
hdresult	int	%9.0g		High-density lipids (mg/dL)

The most basic syntax for `dtable` is

```
. dtable [varlist]
```

where *varlist* includes both continuous and factor variables, specified in the order we would like to see them in the table.



```
. dtable age bmi i.sex i.hlthstat tresult tresult tresult
```

---

	Summary
N	10,351
Age (years)	47.580 (17.215)
Body mass index (BMI)	25.538 (4.915)
Sex	
Male	4,915 (47.5%)
Female	5,436 (52.5%)
Health status	
Excellent	2,407 (23.3%)
Very good	2,591 (25.1%)
Good	2,938 (28.4%)
Fair	1,670 (16.2%)
Poor	729 (7.1%)
Serum cholesterol (mg/dL)	217.670 (49.387)
Serum triglycerides (mg/dL)	143.896 (96.496)

---

We specify the `by()` option to request summary statistics for multiple groups.

```
. dtable age bmi i.sex i.hlthstat tresult tresult,
> by(highbp)
```

	High blood pressure		
	0	1	Total
N	5,975 (57.7%)	4,376 (42.3%)	10,351 (100.0%)
Age (years)	42.165 (16.772)	54.973 (14.909)	47.580 (17.215)
Body mass index (BMI)	24.202 (4.100)	27.361 (5.332)	25.538 (4.915)
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	4,915 (47.5%)
Female	3,364 (56.3%)	2,072 (47.3%)	5,436 (52.5%)
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	2,407 (23.3%)
Very good	1,666 (27.9%)	925 (21.2%)	2,591 (25.1%)
Good	1,572 (26.4%)	1,366 (31.2%)	2,938 (28.4%)
Fair	766 (12.8%)	904 (20.7%)	1,670 (16.2%)
Poor	310 (5.2%)	419 (9.6%)	729 (7.1%)
Serum cholesterol (mg/dL)	208.727 (47.287)	229.880 (49.583)	217.670 (49.387)
Serum triglycerides (mg/dL)	129.228 (83.930)	166.043 (109.200)	143.896 (96.496)

The `by()` option allows suboptions for customizing the table:

- `nototals` suppresses the column of summary statistics for the full dataset.
- `tests` requests that tests be reported comparing summary statistics across groups.

```
. dtable age bmi i.sex i.hlthstat tresult tresult,
> by(highbp, nototals tests)
note: using test regress across levels of highbp for age, bmi, tresult, and
      tresult.
note: using test pearson across levels of highbp for sex and hlthstat.
```

	High blood pressure		Test
	0	1	
N	5,975 (57.7%)	4,376 (42.3%)	
Age (years)	42.165 (16.772)	54.973 (14.909)	<0.001
Body mass index (BMI)	24.202 (4.100)	27.361 (5.332)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
Good	1,572 (26.4%)	1,366 (31.2%)	
Fair	766 (12.8%)	904 (20.7%)	
Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dL)	208.727 (47.287)	229.880 (49.583)	<0.001
Serum triglycerides (mg/dL)	129.228 (83.930)	166.043 (109.200)	<0.001

The first line of the table provides the frequency and percentage of the sample size in each group.

The `sample()` option allows us to specify the title for this first row, which sample statistics to report, and where to place the statistics.

- The `statistics()` suboption specifies whether frequencies, percentages, or proportions are reported.
- The `place()` suboption specifies where the sample statistics are reported.

The `nformat()` and `sformat()` options specify the numeric and string formats for the sample statistics.

```
. dtable age bmi i.sex i.hlthstat tresult tresult,
> by(highbp, nototals tests)
> sample(, place(seplabels)) sformat("N=%s" frequency)
note: using test regress across levels of highbp for age, bmi, tresult, and
      tresult.
note: using test pearson across levels of highbp for sex and hlthstat.
```

	High blood pressure		Test
	0 N=5,975 (57.7%)	1 N=4,376 (42.3%)	
Age (years)	42.165 (16.772)	54.973 (14.909)	<0.001
Body mass index (BMI)	24.202 (4.100)	27.361 (5.332)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
Good	1,572 (26.4%)	1,366 (31.2%)	
Fair	766 (12.8%)	904 (20.7%)	
Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dL)	208.727 (47.287)	229.880 (49.583)	<0.001
Serum triglycerides (mg/dL)	129.228 (83.930)	166.043 (109.200)	<0.001

By default, the frequency and percentage are reported for each factor variables while the mean and standard deviation are reported for each continuous variable.

Pearson's chi-squared tests are reported by default for factor variables, and linear-regression-based tests are reported for continuous variables.

The `factor()` and `continuous()` options allow you to change these defaults for individual variables or for all variables of one type.

- The `statistics()` suboption specifies the statistics to be reported.
- The `test()` suboption specifies the test to be reported.

The `nformat()` and `sformat()` options specify the numeric and string formats for the selected statistics.



```
. dtable age bmi i.sex i.hlthstat tcresult tgresult,
> by(highbp, nototals tests)
> sample(, place(seplabels)) sformat("N=%s" frequency)
> continuous(, statistics(median iqr) test(kwallis)) sformat("(%)s" iqr)
note: using test kwallis across levels of highbp for age, bmi, tcresult, and
      tgresult.
note: using test pearson across levels of highbp for sex and hlthstat.
```

	High blood pressure		Test
	0 N=5,975 (57.7%)	1 N=4,376 (42.3%)	
Age (years)	38.000 (33.000)	60.000 (22.000)	<0.001
Body mass index (BMI)	23.690 (5.002)	26.528 (6.364)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
Good	1,572 (26.4%)	1,366 (31.2%)	
Fair	766 (12.8%)	904 (20.7%)	
Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dL)	203.000 (63.000)	226.000 (64.000)	<0.001
Serum triglycerides (mg/dL)	108.000 (77.000)	142.000 (103.000)	<0.001

We can make the table publication ready by adding a title and notes explaining the tests.

- `title()` adds a title to the top of the table.
- `note()` adds a note below the table. This option may be repeated to add multiple notes on separate lines.

```
. dtable age bmi i.sex i.hlthstat tresult tresult,
> by(highbp, nototals tests notestnotes)
> sample(, place(seplabels)) sformat("N=%s" frequency)
> continuous(, statistics(median iqr) test(kwallis)) sformat("(%s)" iqr)
> title("Table 1: Descriptive statistics")
> note("Frequency, percentage, and Pearson's chi-squared test reported for factor
> variables.")
> note("Median, IQR, and Kruskal-Wallis test reported for continuous variables.")
```

Table 1: Descriptive statistics

	High blood pressure		Test
	0 N=5,975 (57.7%)	1 N=4,376 (42.3%)	
Age (years)	38.000 (33.000)	60.000 (22.000)	<0.001
Body mass index (BMI)	23.690 (5.002)	26.528 (6.364)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
Good	1,572 (26.4%)	1,366 (31.2%)	
Fair	766 (12.8%)	904 (20.7%)	
Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dL)	203.000 (63.000)	226.000 (64.000)	<0.001
Serum triglycerides (mg/dL)	108.000 (77.000)	142.000 (103.000)	<0.001

Frequency, percentage, and Pearson's chi-squared test reported for factor variables.

Median, IQR, and Kruskal-Wallis test reported for continuous variables.

Finally, we can use the `export()` option to export our table to a variety of formats.

```
. dtable age bmi i.sex i.hlthstat tcresult tgresult,
> by(highbp, nototals tests notestnotes)
> sample(, place(seplabels)) sformat("N=%s" frequency)
> continuous(, statistics(median iqr) test(kwallis)) sformat("(%s)" iqr)
> title("Table 1: Descriptive statistics")
> note("Frequency, percentage, and Pearson's chi-squared test reported for factor
> variables.")
> note("Median, IQR, and Kruskal-Wallis test reported for continuous variables.")
> export(sumtable.docx)
```

sumtable.txt [Compatibility Mode] - Word

File Home Insert Design Layout References Mailings Review View Help Table Design Layout Tell me

**Table 1: Descriptive statistics**

	High blood pressure		Test
	0 N=5,975 (57.7%)	1 N=4,376 (42.3%)	
Age (years)	38.000 (33.000)	60.000 (22.000)	<0.001
Body mass index (BMI)	23.690 (5.002)	26.528 (6.364)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
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Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dL)	203.000 (63.000)	226.000 (64.000)	<0.001
Serum triglycerides (mg/dL)	108.000 (77.000)	142.000 (103.000)	<0.001

Frequency, percentage, and Pearson's chi-squared test reported for factor variables.  
Median, IQR, and Kruskal-Wallis test reported for continuous variables.

Page 1 of 1 106 words 110%

Tables created by `dtable` can be customized in many additional ways. Type

```
. help dtable
```

in Stata to see the full syntax of this command.

# `etable`



After fitting any regression model—whether a linear regression, logistic regression, Cox proportional hazards model, multilevel model, or any other regression model available in Stata—we can use the `etable` command to create a table of regression results.

We can also use `etable` to create a table comparing results from multiple models.

To create a table, we simply type `etable` after fitting a model to create the default table. To demonstrate, we first fit a logistic regression model.

```
. logistic highbp bmi age
```

```
Logistic regression
```

```
Number of obs = 10,351
```

```
LR chi2(2) = 2295.05
```

```
Prob > chi2 = 0.0000
```

```
Pseudo R2 = 0.1628
```

```
Log likelihood = -5903.2393
```

highbp	Odds ratio	Std. err.	z	P> z	[95% conf. interval]	
bmi	1.145469	.0058173	26.74	0.000	1.134124	1.156928
age	1.046239	.0014489	32.64	0.000	1.043403	1.049083
_cons	.002455	.000372	-39.66	0.000	.0018242	.0033038

Note: `_cons` estimates baseline odds.

Then we create our table of results.

```
. etable
```

	highbp
Body mass index (BMI)	1.145 (0.006)
Age (years)	1.046 (0.001)
Intercept	0.002 (0.000)
Number of observations	10351

By default, `etable` reports the coefficients or related statistics reported by the regression command (odds ratios in this case). The corresponding standard errors are reported in parentheses.

We can request that different statistics related to each coefficient be reported by specifying the `cstat()` option.

Instead of reporting odds ratios and standard errors, perhaps we want to report odds ratios and confidence intervals. We type

```
. etable, cstat(_r_b) cstat(_r_ci)
```

	highbp	
Body mass index (BMI)		1.145
	[1.134	1.157]
Age (years)		1.046
	[1.043	1.049]
Intercept		0.002
	[0.002	0.003]
Number of observations		10351

Each result is specified separately to allow different formatting options. Similar to `table`, we can specify an `nformat()` and `sformat()` for each result. We can also specify the delimiter to be used between the upper and lower bound of the confidence interval with `cidelimiter()`.

```
. etable, cstat(_r_b) cstat(_r_ci, cidelimiter(", ")) nformat(%4.2f)
```

---

	highbp
Body mass index (BMI)	1.145
	[1.13, 1.16]
Age (years)	1.046
	[1.04, 1.05]
Intercept	0.002
	[0.00, 0.00]
Number of observations	10351

---

We could request that the  $p$ -value be displayed alongside each odds ratio. However, another alternative is to request that stars corresponding to specific  $p$ -value levels be displayed.

```
. etable, showstars showstarsnote
```

---

	highbp
Body mass index (BMI)	1.145 ** (0.006)
Age (years)	1.046 ** (0.001)
Intercept	0.002 ** (0.000)
Number of observations	10351

---

\*\* p<.01, \* p<.05

We can also request a variety of overall model statistics by specifying the `mstat()` option.

Rather than the number of observations, we request that AIC and BIC be reported so that we can later compare this model to others.

```
. etable, mstat(aic) mstat(bic)
```

---

	highbp
Body mass index (BMI)	1.145 (0.006)
Age (years)	1.046 (0.001)
Intercept	0.002 (0.000)
AIC	11812.48
BIC	11834.21

---

Suppose we want to compare the model we fit above to two other models. We can build our table in two ways. The first is to use the `append` option with `etable` after fitting the second and third models.



We begin by fitting the first model and using `etable` just as we did previously.

```
. logistic highbp bmi age
```

```
Logistic regression
```

```
Number of obs = 10,351
```

```
LR chi2(2) = 2295.05
```

```
Prob > chi2 = 0.0000
```

```
Pseudo R2 = 0.1628
```

```
Log likelihood = -5903.2393
```

highbp	Odds ratio	Std. err.	z	P> z	[95% conf. interval]	
bmi	1.145469	.0058173	26.74	0.000	1.134124	1.156928
age	1.046239	.0014489	32.64	0.000	1.043403	1.049083
_cons	.002455	.000372	-39.66	0.000	.0018242	.0033038

Note: `_cons` estimates baseline odds.

```
. etable, mstat(aic) mstat(bic)
```

---

	highbp
Body mass index (BMI)	1.145 (0.006)
Age (years)	1.046 (0.001)
Intercept	0.002 (0.000)
AIC	11812.48
BIC	11834.21

---

Next, we fit our second model, which includes sex as a regressor, and then specify the append option with etable.

```
. logistic highbp bmi age i.sex
```

```
Logistic regression
```

```
Number of obs = 10,351
```

```
LR chi2(3) = 2410.54
```

```
Prob > chi2 = 0.0000
```

```
Pseudo R2 = 0.1709
```

```
Log likelihood = -5845.4948
```

highbp	Odds ratio	Std. err.	z	P> z	[95% conf. interval]	
bmi	1.147006	.005827	27.00	0.000	1.135642	1.158484
age	1.047058	.0014632	32.91	0.000	1.044194	1.04993
sex						
Female	.6172715	.0278625	-10.69	0.000	.5650079	.6743695
_cons	.0029059	.0004428	-38.33	0.000	.0021556	.0039174

Note: \_cons estimates baseline odds.

```
. etable, append
```

---

	highbp	highbp
Body mass index (BMI)	1.145	1.147
	(0.006)	(0.006)
Age (years)	1.046	1.047
	(0.001)	(0.001)
Sex		
Female		0.617
		(0.028)
Intercept	0.002	0.003
	(0.000)	(0.000)
AIC	11812.48	11698.99
BIC	11834.21	11727.97

---

Similarly, we fit our third model, which includes an interaction between age and sex, and again specify append with etable.

```
. logistic highbp bmi c.age##i.sex
```

```
Logistic regression
```

```
Number of obs = 10,351
```

```
LR chi2(4) = 2505.76
```

```
Prob > chi2 = 0.0000
```

```
Pseudo R2 = 0.1777
```

```
Log likelihood = -5797.8856
```

highbp	Odds ratio	Std. err.	z	P> z	[95% conf. interval]	
bmi	1.147615	.0058948	26.80	0.000	1.136119	1.159226
age	1.034097	.0019132	18.12	0.000	1.030354	1.037854
sex						
Female	.1522122	.0232902	-12.30	0.000	.1127732	.2054439
sex#c.age						
Female	1.028	.0029402	9.66	0.000	1.022253	1.033779
_cons	.0052605	.0008567	-32.22	0.000	.003823	.0072384

Note: \_cons estimates baseline odds.

```
. etable, append
```

	highbp	highbp	highbp
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Sex			
Female		0.617 (0.028)	0.152 (0.023)
Sex # Age (years)			
Female			1.028 (0.003)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00

Finally, we modify our table by adding a title, notes, and numbers for the column headings. We specify the `replay` option to make these modifications to the existing table.

```
. etable, replay column(index)
> title("Table 2: Logistic regression results")
> notes("Odds ratio (Std. Err.)")
```

Table 2: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Sex			
Female		0.617 (0.028)	0.152 (0.023)
Sex # Age (years)			
Female			1.028 (0.003)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00

Odds ratio (Std. Err.)



We could have fit all three models and then created the same table with a single etable command as follows:

```
. logistic highbp bmi age
. estimates store M1
. logistic highbp bmi age i.sex
. estimates store M2
. logistic highbp bmi c.age##i.age
. estimates store M3
. etable, estimates(M1 M2 M3) mstat(aic) mstat(bic)
      title("Table 2: Logistic regression results")
      notes("Odds ratio (Std. Err.)")
```

The `etable` command also allows us to create tables of marginal means and marginal effects produced by `margins` and more. Type

```
. help etable
```

in Stata to see the full syntax of this command.

table

The `table` command is a flexible command that can produce cross-tabulations; one-way, two-way, and multiway tables of summary statistics; and tables or results from other Stata commands.

A simplified syntax for the `table` command is

```
. table ( row variables ) ( column variables )
```

To create a one-way tabulation, we can specify a single row variable in the first set of parenthesis or a single column variable in the second set of parentheses.

```
. table (highbp) ()
```

	Frequency
High blood pressure	
0	5,975
1	4,376
Total	10,351

```
. table () (highbp)
```

	High blood pressure		
	0	1	Total
Frequency	5,975	4,376	10,351

To create a two-way tabulation, we specify both a row and a column variable.

```
. table (sex) (highbp)
```

	High blood pressure		
	0	1	Total
Sex			
Male	2,611	2,304	4,915
Female	3,364	2,072	5,436
Total	5,975	4,376	10,351

We can add the `nototals` option to remove the row frequencies, column frequencies, and overall frequencies from the tabulation.

```
. table (sex) (highbp), nototals
```

	High blood pressure	
	0	1
Sex		
Male	2,611	2,304
Female	3,364	2,072

Alternatively, we could request totals only for columns (`highbp`) or only for rows (`sex`).

```
. table (sex) (highbp), totals(highbp)
```

	High blood pressure	
	0	1
Sex		
Male	2,611	2,304
Female	3,364	2,072
Total	5,975	4,376

```
. table (sex) (highbp), totals(sex)
```

	High blood pressure		
	0	1	Total
Sex			
Male	2,611	2,304	4,915
Female	3,364	2,072	5,436



By default, the `table` command reports frequencies. Many additional summary statistics can be specified using the `statistic()` option.

```
. table ( row variables ) ( column variables ),  
        statistic( statspec )
```

We can specify the `statistic()` option repeatedly to display multiple statistics in a table. For example, we can request both frequencies and percentages for `highbp`.

```
. table () (highbp),  
>     statistic(frequency)  
>     statistic(percent)
```

	High blood pressure		
	0	1	Total
Frequency	5,975	4,376	10,351
Percent	57.72	42.28	100.00

We can also request multiple statistics for our two-way table. This time, we omit the totals.

```
. table (sex) (highbp),
>     statistic(frequency)
>     statistic(percent)
>     nototals
```

	High blood pressure	
	0	1
Sex		
Male		
Frequency	2,611	2,304
Percent	25.22	22.26
Female		
Frequency	3,364	2,072
Percent	32.50	20.02

We might instead want to compare the means and standard deviations of some continuous variables, here age and bmi, across the levels of highbp.

```
. table () (highbp),
>     statistic(mean age bmi)
>     statistic(sd age bmi)
```

	High blood pressure		
	0	1	Total
<b>Mean</b>			
Age (years)	42.16502	54.97281	47.57965
Body mass index (BMI)	24.20231	27.36081	25.5376
<b>Standard deviation</b>			
Age (years)	16.77157	14.90897	17.21483
Body mass index (BMI)	4.100279	5.332119	4.914969

We can request different statistics for different variables. In addition to the means and standard deviations of continuous variables, we can compare the frequencies and percentages of the factor variable `sex` across the levels of `highbp`.

```

. table () (highbp),
>     statistic(mean age bmi)
>     statistic(sd age bmi)
>     statistic(fvfrequency sex)
>     statistic(fvpercent sex)

```

	High blood pressure		
	0	1	Total
Mean			
Age (years)	42.16502	54.97281	47.57965
Body mass index (BMI)	24.20231	27.36081	25.5376
Standard deviation			
Age (years)	16.77157	14.90897	17.21483
Body mass index (BMI)	4.100279	5.332119	4.914969
Factor-variable frequency			
Sex=Male	2,611	2,304	4,915
Sex=Female	3,364	2,072	5,436
Factor-variable percent			
Sex=Male	43.70	52.65	47.48
Sex=Female	56.30	47.35	52.52

We can change the numeric format and the string format of the results in our table by using the `nformat()` and `sformat()` options.

```
. table ( row variables ) ( column variables ),  
        nformat() sformat()
```

We can modify the numeric and string formats of the reported statistics by specifying the `nformat()` and `sformat()` options.

```
. table () (highbp),  
    statistic(mean age bmi)  
    statistic(sd age bmi)  
    statistic(fvfrequency sex)  
    statistic(fvpercent sex)  
    nformat(%5.2f mean sd)  
    sformat("(%s%%)" fvpercent)  
    sformat("(%s)" sd)
```



	High blood pressure		
	0	1	Total
Mean			
Age (years)	42.17	54.97	47.58
Body mass index (BMI)	24.20	27.36	25.54
Standard deviation			
Age (years)	(16.77)	(14.91)	(17.21)
Body mass index (BMI)	(4.10)	(5.33)	(4.91)
Factor variable frequency			
Sex=Male	2,611	2,304	4,915
Sex=Female	3,364	2,072	5,436
Factor variable percent			
Sex=Male	(43.70%)	(52.65%)	(47.48%)
Sex=Female	(56.30%)	(47.35%)	(52.52%)

The `table` command also allows us to specify keywords in the row and column specifications to further control the layout of the table.

```
. table (rowspec) (colspec), options
```

<code>result</code>	requested statistics
<code>var</code>	variables from <code>statistic()</code> option
<code>across</code>	<code>index across()</code> specifications
<code>colname</code>	column names for matrix statistics
<code>rowname</code>	row names for matrix statistics
<code>coleq</code>	column equation names for matrix statistics
<code>roweq</code>	row equation names for matrix statistics
<code>command</code>	<code>index option command()</code>
<code>statcmd</code>	<code>index options statistic()</code> and <code>command()</code>

We modify our table so that the rows to are organized by variables (`var`) and then statistics (`result`) nested within variables.

```
. table (var result) (highbp),  
      statistic(mean age bmi)  
      statistic(sd age bmi)  
      statistic(fvfrequency sex)  
      statistic(fvpercent sex)  
      nformat(%5.2f mean sd)  
      sformat("(%s%)" fvpercent)  
      sformat("(%s)" sd)
```

	High blood pressure		
	0	1	Total
Age (years)			
Mean	42.17	54.97	47.58
Standard deviation	(16.77)	(14.91)	(17.21)
Body mass index (BMI)			
Mean	24.20	27.36	25.54
Standard deviation	(4.10)	(5.33)	(4.91)
Sex=Male			
Factor variable frequency	2,611	2,304	4,915
Factor variable percent	(43.70%)	(52.65%)	(47.48%)
Sex=Female			
Factor variable frequency	3,364	2,072	5,436
Factor variable percent	(56.30%)	(47.35%)	(52.52%)

The `table` command also allows us to create three-way and higher-way tables, include results from other Stata commands, add titles and notes, and more. Type

```
. help table
```

in Stata to see the full syntax of this command.

# collect

`collect` is a suite of commands designed to collect results from various Stata commands; place the selected results into a table; customize the labels, formats, and style; and export the final table to Word, Excel,  $\text{\LaTeX}$ , PDF, Markdown, HTML, and more.

Because the `dtable`, `etable`, and `table` commands are based on the `collect` system, we can also use the `collect` commands to further modify tables originally created by one of these commands.

The basic workflow for collecting results and building tables is

- Collect results from Stata commands—`collect` command, `collect: prefix`, `dtable`, `etable`, or `table`
- Explore the collection—`collect dims`, `collect levelsof`, and `collect label list`
- Define the rows and columns of the table—`collect layout`, `dtable`, `etable`, or `table`
- Modify items within the collection—`collect composite`, `collect recode`, `collect remap`, and `collect addtags`



- Customize the table, specifying formats, labels, font, shading, and more—`collect label`, `collect style`, `collect stars`, ...
- Export the table to Word, Excel,  $\text{\LaTeX}$ , PDF, Markdown, HTML, SMCL, or plain text—`collect export`
- Save the style, labels, and collection to use and modify later—`collect label save`, `collect style save`, and `collect save`

Suppose we want to create a table that displays the results from  $t$  tests performed using the `ttest` command. We first take a look at the results that this command returns by typing `return list`.

```
. ttest age, by(highbp)
```

```
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	42.16502	.2169725	16.77157	41.73968	42.59037
1	4,376	54.97281	.2253767	14.90897	54.53095	55.41466
Combined	10,351	47.57965	.1692044	17.21483	47.24798	47.91133
diff		-12.80779	.3185604		-13.43223	-12.18335

```
diff = mean(0) - mean(1)                                t = -40.2052
HO: diff = 0                                           Degrees of freedom = 10349
Ha: diff < 0                                           Ha: diff != 0                                           Ha: diff > 0
Pr(T < t) = 0.0000                                     Pr(|T| > |t|) = 0.0000                                     Pr(T > t) = 1.0000
```

```
. return list
```

```
scalars:
```

```
  r(level) = 95  
    r(sd) = 17.21482923023818  
  r(sd_2) = 14.9089715191102  
  r(sd_1) = 16.77156676799842  
    r(se) = .3185603831285  
  r(p_u) = 1  
  r(p_1) = 0  
    r(p) = 0  
  r(t) = -40.20520433030012  
  r(df_t) = 10349  
  r(mu_2) = 54.97280621572212  
  r(N_2) = 4376  
  r(mu_1) = 42.16502092050209  
  r(N_1) = 5975
```

Before collecting results, we use `collect clear` to make sure any previous results are cleared from the default collection. Then we are ready to store results in the collection. We do this with the `collect` prefix. We want our table to include the mean of each group (stored in `r(mu_1)` and `r(mu_2)`) as well as the  $p$ -value (stored in `r(p)`).

```
. collect clear
. collect r(mu_1) r(mu_2) r(p): ttest age, by(highbp)
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	42.16502	.2169725	16.77157	41.73968	42.59037
1	4,376	54.97281	.2253767	14.90897	54.53095	55.41466
Combined	10,351	47.57965	.1692044	17.21483	47.24798	47.91133
diff		-12.80779	.3185604		-13.43223	-12.18335

```
diff = mean(0) - mean(1)                                t = -40.2052
HO: diff = 0                                             Degrees of freedom = 10349
Ha: diff < 0                                           Ha: diff != 0
Pr(T < t) = 0.0000                                     Pr(|T| > |t|) = 0.0000
                                                         Ha: diff > 0
                                                         Pr(T > t) = 1.0000
```

```
. collect r(mu_1) r(mu_2) r(p): ttest weight, by(highbp)
```

```
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	68.26626	.1755737	13.57152	67.92207	68.61045
1	4,376	76.85565	.2455179	16.24134	76.37431	77.33699
Combined	10,351	71.89752	.1509381	15.35642	71.60165	72.19339
diff		-8.589386	.2936615		-9.165019	-8.013753

```
diff = mean(0) - mean(1)
```

```
t = -29.2493
```

```
H0: diff = 0
```

```
Degrees of freedom = 10349
```

```
Ha: diff < 0
```

```
Ha: diff != 0
```

```
Ha: diff > 0
```

```
Pr(T < t) = 0.0000
```

```
Pr(|T| > |t|) = 0.0000
```

```
Pr(T > t) = 1.0000
```

```
. collect r(mu_1) r(mu_2) r(p): ttest bmi, by(highbp)
```

```
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	24.20231	.053045	4.100279	24.09832	24.30629
1	4,376	27.36081	.0806049	5.332119	27.20279	27.51884
Combined	10,351	25.5376	.0483092	4.914969	25.4429	25.63229
diff		-3.158506	.0927376		-3.34029	-2.976723

```
diff = mean(0) - mean(1)
```

```
t = -34.0585
```

```
H0: diff = 0
```

```
Degrees of freedom = 10349
```

```
Ha: diff < 0
```

```
Ha: diff != 0
```

```
Ha: diff > 0
```

```
Pr(T < t) = 0.0000
```

```
Pr(|T| > |t|) = 0.0000
```

```
Pr(T > t) = 1.0000
```

Values in a collection are organized according to their *tags*. A tag consists of a *dimension* and a *level* within the dimension. A tag is written as `dimension[level]`. Consider the means and  $p$ -value from the previous `ttest` command.

Value	Tag 1	Tag 2
24.202	<code>cmdset [3]</code>	<code>result [mu_1]</code>
27.361	<code>cmdset [3]</code>	<code>result [mu_2]</code>
0.000	<code>cmdset [3]</code>	<code>result [p]</code>

Our collection includes dimensions `cmdset` and `result`, among others. These results all come from the third `ttest` command, so they have level 3 of dimension `cmdset`. The level of dimension `result` corresponds to the name in the `r()` result.



How do we find out what is in the collection? We can use `collect dims` to see a list of the dimensions.

```
. collect dims
```

```
Collection dimensions
```

```
Collection: default
```

	Dimension	No. levels
Layout, style, header, label		
	cmdset	3
	program_class	1
	result	14
	result_type	1
Style only		
	border_block	4
	cell_type	4

We can see the levels of a specified dimension with `collect levelsof` or `collect label list`. We add option `all` to see the levels that are not labeled in addition to those that are.

```
. collect label list result, all
Collection: default
Dimension: result
Label: Result
Level labels:
  N_1 Sample size n1
  N_2 Sample size n2
  df_t Degrees of freedom
  level Confidence level
  mu_1 x1 mean for population 1
  mu_2 x2 mean for population 2
  p Two-sided p-value
  p_l Lower one-sided p-value
  p_u Upper one-sided p-value
  sd Combined std. dev.
  sd_1 Standard deviation for first variable
  sd_2 Standard deviation for second variable
  se Std. error
  t t statistic
```

```
. collect label list cmdset, all
      Collection: default
      Dimension: cmdset
      Label: Command results index
Level labels:
      1
      2
      3
```

The levels of the `cmdset` dimension correspond to our three `ttest` commands. The levels of the `result` dimension correspond to the statistics. We use the dimension names to specify a table layout with commands on the rows and statistics on the columns by typing

```
. collect layout (cmdset) (result)
```

```
Collection: default
```

```
  Rows: cmdset
```

```
  Columns: result
```

```
Table 1: 3 x 3
```

	x1 mean for population 1	x1 mean for population 2	Two-sided p-value
1	42.16502	54.97281	0
2	68.26626	76.85565	9.1e-181
3	24.20231	27.36081	4.5e-241

We could label the values of `cmdset` to give more information about the rows. Another option is to create meaningful tags when we collect results.

Here, we use the `tags()` option with the `collect` prefix to create a new dimension named `varname`. Its levels correspond to the names of the variables specified in `ttest`.

```
. collect clear
. collect r(mu_1) r(mu_2) r(p), tags(varname[age]): ttest age, by(highbp)
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	42.16502	.2169725	16.77157	41.73968	42.59037
1	4,376	54.97281	.2253767	14.90897	54.53095	55.41466
Combined	10,351	47.57965	.1692044	17.21483	47.24798	47.91133
diff		-12.80779	.3185604		-13.43223	-12.18335

```
diff = mean(0) - mean(1)                                t = -40.2052
HO: diff = 0                                             Degrees of freedom = 10349
Ha: diff < 0                                           Ha: diff != 0
Pr(T < t) = 0.0000                                     Pr(|T| > |t|) = 0.0000
                                                         Ha: diff > 0
                                                         Pr(T > t) = 1.0000
```

```
. collect r(mu_1) r(mu_2) r(p), tags(varname[weight]): ttest weight, by(highbp)
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	68.26626	.1755737	13.57152	67.92207	68.61045
1	4,376	76.85565	.2455179	16.24134	76.37431	77.33699
Combined	10,351	71.89752	.1509381	15.35642	71.60165	72.19339
diff		-8.589386	.2936615		-9.165019	-8.013753

```
diff = mean(0) - mean(1)                                t = -29.2493
HO: diff = 0                                             Degrees of freedom = 10349
Ha: diff < 0                                           Ha: diff != 0                                           Ha: diff > 0
Pr(T < t) = 0.0000                                     Pr(|T| > |t|) = 0.0000                                   Pr(T > t) = 1.0000
```

```
. collect r(mu_1) r(mu_2) r(p), tags(varname[bmi]): ttest bmi, by(highbp)
```

```
Two-sample t test with equal variances
```

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	5,975	24.20231	.053045	4.100279	24.09832	24.30629
1	4,376	27.36081	.0806049	5.332119	27.20279	27.51884
Combined	10,351	25.5376	.0483092	4.914969	25.4429	25.63229
diff		-3.158506	.0927376		-3.34029	-2.976723

```
diff = mean(0) - mean(1)
```

```
t = -34.0585
```

```
H0: diff = 0
```

```
Degrees of freedom = 10349
```

```
Ha: diff < 0
```

```
Ha: diff != 0
```

```
Ha: diff > 0
```

```
Pr(T < t) = 0.0000
```

```
Pr(|T| > |t|) = 0.0000
```

```
Pr(T > t) = 1.0000
```



We see that the new varname dimension has been added to the collection.

```
. collect dims
Collection dimensions
Collection: default
```

---

	Dimension	No. levels
<hr/>		
Layout, style, header, label		
	cmdset	3
	program_class	1
	result	14
	result_type	1
	varname	3
Style only		
	border_block	4
	cell_type	4

---

The levels of this dimension are age, bmi, and weight, as we requested.

```
. collect label list varname, all
      Collection: default
      Dimension: varname
      Label:
Level labels:
      age
      bmi
      weight
```

We can now specify that the varname dimension be used to construct the rows of our table.

```
. collect layout (varname) (result)
Collection: default
  Rows: varname
  Columns: result
Table 1: 3 x 3
```

	x1 mean for population 1	x2 mean for population 2	Two-sided p-value
age	42.16502	54.97281	0
weight	68.26626	76.85565	9.1e-181
bmi	24.20231	27.36081	4.5e-241

We now have more meaningful labels on the rows, but the column headers have default labels that we might want to modify.

We can use the `collect label` command to specify our preferred labels for the levels of the result dimension.

```
. collect label levels result  
> mu_1 "Normotensive" mu_2 "Hypertensive" p "p-value", modify  
. collect preview
```

	Normotensive	Hypertensive	p-value
age	42.16502	54.97281	0
weight	68.26626	76.85565	9.1e-181
bmi	24.20231	27.36081	4.5e-241

`collect style cell` allows us to make a wide variety of formatting modifications to our table. First, we modify the numeric format of the  $p$ -values, and we request that numbers less than 0.001 be reported as <0.001 in the table.

```
. collect style cell result[p], nformat(%5.3f) min(0.001)
. collect preview
```

	Normotensive	Hypertensive	p-value
age	42.16502	54.97281	<0.001
weight	68.26626	76.85565	<0.001
bmi	24.20231	27.36081	<0.001

We also request that only two decimal places be reported for the means.

```
. collect style cell result[mu_1 mu_2], nformat(%5.2f)  
. collect preview
```

	Normotensive	Hypertensive	p-value
age	42.17	54.97	<0.001
weight	68.27	76.86	<0.001
bmi	24.20	27.36	<0.001

We specify that the values (`cell_type[item]`) be centered horizontally within in the columns.

```
. collect style cell cell_type[item], halign(center)  
. collect preview
```

	Normotensive	Hypertensive	p-value
age	42.17	54.97	<0.001
weight	68.27	76.86	<0.001
bmi	24.20	27.36	<0.001

Finally, we request that the border be removed from the right side of the row headers.

```
. collect style cell border_block, border(right, pattern(nil))  
. collect preview
```

---

	Normotensive	Hypertensive	p-value
age	42.17	54.97	<0.001
weight	68.27	76.86	<0.001
bmi	24.20	27.36	<0.001

---



Once the table is formatted to our liking, we can export to a number of different formats by specifying the appropriate suffix. Below, we export the table to HTML.

```
. collect export ttest_tab.html, replace  
(collection default exported to file ttest_tab.html)
```

If we create tables of a similar style repeatedly, there is no need to remember the `collect` commands each time we create a new table. Instead, we can save the style we like and apply it to tables we create in the future.

```
. collect style save ttest_style, replace  
(style from default saved to file ttest_style.stjson)
```

Now, we can apply this style to tables using different variables and even different datasets.

```
. clear all
. sysuse auto, clear
. collect r(mu_1) r(mu_2) r(p), tags(varname[price]): ttest price, by(foreign)
. collect r(mu_1) r(mu_2) r(p), tags(varname[weight]): ttest weight, by(foreign)
. collect r(mu_1) r(mu_2) r(p), tags(varname[length]): ttest length, by(foreign)
. collect style use ttest_style, replace
. collect preview
```

---

	x1 mean for population 1	x2 mean for population 2	Two-sided p-value
price	6072.42	6384.68	0.680
weight	3317.12	2315.91	<0.001
length	196.13	168.55	<0.001

---

Now, we need only to make modifications specific to this dataset, such as the labels in the column headers.

```
. collect label levels result  
> mu_1 "Domestic cars" mu_2 "Foreign cars" p "p-value", modify  
. collect preview
```

	Domestic cars	Foreign cars	p-value
price	6072.42	6384.68	0.680
weight	3317.12	2315.91	<0.001
length	196.13	168.55	<0.001

As we have seen previously, `dtable`, `etable`, and `table` make it easy to create tables. They have many options for formatting, but when we don't get exactly what we want from one of these commands, we can often use just a few `collect` commands to finish the customization.

We can put the finishing touches on our table of logistic regression results.

```
. quietly logistic highbp bmi age  
. quietly etable  
. quietly logistic highbp bmi age i.sex  
. quietly etable, append  
. quietly logistic highbp bmi c.age##i.sex
```

```
. etable, append column(index)
      mstat(aic) mstat(bic)
      title("Table 2: Logistic regression results")
      notes("Odds ratio (Std. Err.)")
```

Table 2: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Sex			
Female		0.617 (0.028)	0.152 (0.023)
Sex # Age (years)			
Female			1.028 (0.003)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00

Odds ratio (Std. Err.)

Because `sex` is a binary variable, we do not need the variable label and the level labels. We hide the title for this dimension. We also specify a split row style with an `x` for the interaction delimiter instead of `#`. We again add some vertical space to the table.



```
. collect style header sex, title(hide)
. collect style row split, dups(first) delimiter(" x ") spacer
. collect preview
```

Table 2: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145	1.147	1.148
	(0.006)	(0.006)	(0.006)
Age (years)	1.046	1.047	1.034
	(0.001)	(0.001)	(0.002)
Female		0.617	0.152
		(0.028)	(0.023)
Female x Age (years)			1.028
			(0.003)
Intercept	0.002	0.003	0.005
	(0.000)	(0.000)	(0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00

Odds ratio (Std. Err.)

We now have a clean table that can be used alone or included in complete reports.

# Complete reports

Stata offers the following suites of commands for creating full reports that include customized text, formatted tables, Stata results, and images:

- `putdocx` creates Word documents
- `putpdf` creates PDFs
- `putexcel` creates Excel files
- `dyndoc` and Markdown together create HTML files and Word documents

A basic outline for commands used to create a Word document is shown below.

```
putdocx begin

    putdocx paragraph
    putdocx text = ("mytext")

    putdocx textblock begin
        my paragraph text
    putdocx textblock end

    putdocx paragraph
    putdocx image myimage.png

    putdocx table tablename = etable

    putdocx collect

putdocx save myfile.docx
```

We demonstrate how to incorporate tables in a Word document by using `putdocx` to generate a complete Word document with text, the tables we created previously, and a graph.

```
version 18.5
clear all
putdocx begin

putdocx paragraph, style(Heading1)
putdocx text ("Introduction")

putdocx textblock begin, hardbreak
This study examines relationships between high blood pressure and
a number of demographic and health measures.
putdocx textblock end

putdocx paragraph, style(Heading1)
putdocx text ("Data")

putdocx textblock begin, hardbreak
We use data from the Second National Health and Nutrition
Examination Survey. Table 1 below reports summary statistics
for individuals with and without high blood pressure.
putdocx textblock end
```

```
webuse nhanes21
```

```
dtable age bmi i.sex i.hlthstat tresult tresult,  
      by(highbp, nototals tests notestnotes)  
      sample(, place(seplabels)) sformat("N=%s" frequency)  
      title("Table 1: Descriptive statistics")  
      note("Frequency, percentage, and Pearson's chi-squared  
           test reported for factor variables.")  
      note("Mean, SD, and t test reported for continuous variables.")
```

```
putdocx collect
```

```
putdocx paragraph, style(Heading1)  
putdocx text ("Results")
```

```
putdocx textblock begin, hardbreak
```

```
We fit three logistic regression models of high blood pressure on BMI,  
age, sex, and the interaction of age and sex. Results of the three  
models are presented in Table 2.
```

```
putdocx textblock end
```



```
logistic highbp bmi age
etable
logistic highbp bmi age i.sex
etable, append
logistic highbp bmi c.age##i.sex
etable, append column(index)
      mstat(aic) mstat(bic)
      title("Table 2: Logistic regression results")
      notes("Odds ratio (Std. Err.)")

collect style header sex, title(hide)
collect style row split, dups(first) delimiter(" x ") spacer
putdocx collect

putdocx textblock begin, hardbreak
The model with the interaction provided best fit. The graph below
plots explores the interaction effect. Probability of high blood
pressure increases with age. For younger individuals, males
have a higher probability of having high blood pressure, but this
changes with age.
putdocx textblock end
```

```
margins sex, at(age=(20(5)75))

marginsplot, ytitle("Probability of high blood pressure")
  title("Interaction of age and sex")
  graphregion(fcolor(white))
  legend(cols(1) bplacement(5) ring(0))
  ylabel(, angle(horizontal)) xlabel(, grid)

graph export marginsplot.png, replace

putdocx paragraph, halign(center)
putdocx image marginsplot.png, height(3.5in)

putdocx save rpt1.docx, replace
```

When we run the do-file, we create the following Word document.

**Introduction**

This study examines relationships between high blood pressure and a number of demographic and health measures.

**Data**

We use data from the Second National Health and Nutrition Examination Survey. Table 1 below reports summary statistics for individuals with and without high blood pressure.

**Table 1: Descriptive statistics**

	High blood pressure		Test
	0 N=5,975 (57.7%)	1 N=4,376 (42.3%)	
Age (years)	42.165 (16.772)	54.973 (14.909)	<0.001
Body mass index (BMI)	24.202 (4.100)	27.361 (5.332)	<0.001
Sex			
Male	2,611 (43.7%)	2,304 (52.7%)	<0.001
Female	3,364 (56.3%)	2,072 (47.3%)	
Health status			
Excellent	1,649 (27.7%)	758 (17.3%)	<0.001
Very good	1,666 (27.9%)	925 (21.2%)	
Good	1,572 (26.4%)	1,366 (31.2%)	
Fair	766 (12.8%)	904 (20.7%)	
Poor	310 (5.2%)	419 (9.6%)	
Serum cholesterol (mg/dl)	208.727 (47.287)	229.880 (49.583)	<0.001
Serum triglycerides (mg/dl)	129.228 (83.930)	166.043 (109.200)	<0.001

Frequency, percentage, and Pearson's chi-squared test reported for factor variables. Mean, SD, and t test reported for continuous variables.

Page 1 of 3    280 words    100%

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## Results

We fit three logistic regression models of high blood pressure on BMI, age, sex, and the interaction of age and sex. Results of the three models are presented in Table 2.

Table 2: Logistic regression results

	1	2	3
Body mass index (BMI)	1.145 (0.006)	1.147 (0.006)	1.148 (0.006)
Age (years)	1.046 (0.001)	1.047 (0.001)	1.034 (0.002)
Female		0.617 (0.028)	0.152 (0.023)
Female x Age (years)			1.028 (0.009)
Intercept	0.002 (0.000)	0.003 (0.000)	0.005 (0.001)
AIC	11812.48	11698.99	11605.77
BIC	11834.21	11727.97	11642.00
Odds ratio (Std. Err.)			

The model with the interaction provided best fit. The graph below plots explores the interaction effect. Probability of high blood pressure increases with age. For younger individuals, males have a higher probability of having high blood pressure, but this changes with age.

Interaction of age and sex

Probability of high blood pressure

Age (years)

Male  
Female

Page 2 of 2 280 words

100%

We used `putdocx` as an example of incorporating tables in complete reports. We can similarly create

- PDFs by using `putpdf` or `docx2pdf` after `putdocx`
- Excel files by using `putexcel`
- HTML files by using `dyndoc`

Thank you!