

Lab5: Tables and OLS

Introduction to Econometrics, Spring 2023

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Section 1

T-Test and Table

T-Test and Table

- Review the Theory

Hypothesis Test of \bar{Y}

- Specify H_0 and H_1

$$H_0 : E[Y] = \mu_{Y,0} \quad H_1 : E[Y] \neq \mu_{Y,0}$$

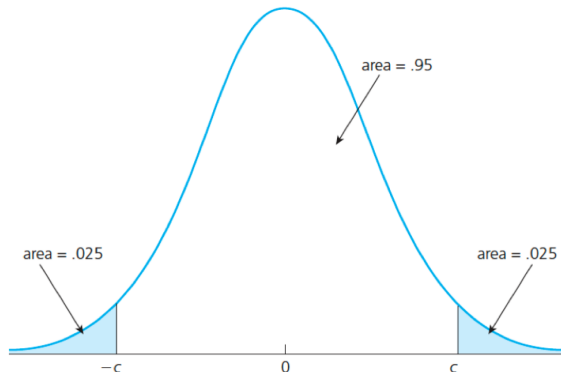
- Choose the significance level α and define a decision rule (critical region or critical value)
 - eg. if we choose $\alpha = 0.05$, then the critical value is 1.96, then the region is $(-\infty, -1.96]$ and $[1.96, +\infty)$

T-Test and Table

- Review the Theory

Hypothesis Test of \bar{Y}

FIGURE C.4 The 97.5th percentile, c , in a t distribution.



T-Test and Table

- Review the Theory

Hypothesis Test of \bar{Y}

- Given the data compute the test statistic
 - Step1: Compute the sample average \bar{Y}
 - Step2: Compute the **standard error** of \bar{Y}

$$SE(\bar{Y}) = \frac{s_Y}{\sqrt{n}}$$

- Step3: Compute the **t-statistic**

$$t^{act} = \frac{\bar{Y} - \mu_{Y,0}}{SE(\bar{Y})}$$

- Step4: Reject the null hypothesis if
 - $|t^{act}| > \text{critical value}$
 - or if $p\text{-value} < \text{significance level}$

- Review the Theory

Hypothesis Tests for the Difference Between Two Means

- To illustrate a test for the difference between two means, let μ_w be the mean hourly earning in the population of women recently graduated from college and let μ_m be the population mean for recently graduated men.
- Then the **null hypothesis** and **the two-sided alternative hypothesis** are

$$H_0 : \mu_m = \mu_w$$

$$H_1 : \mu_m \neq \mu_w$$

- Consider the null hypothesis that mean earnings for these two populations differ by a certain amount, say d_0 . The null hypothesis that men and women in these populations have the same mean earnings corresponds to $H_0 : d_0 = \mu_m - \mu_w = 0$

T-Test and Table

- Review the Theory

Assuming Case: (the California School)

Comparing Means from Different Populations

The Difference Between Two Means

- Suppose we have samples of n_m men and n_w women drawn at random from their populations. Let the sample average annual earnings be \bar{Y}_m for men and \bar{Y}_w for women. Then an estimator of $\mu_m - \mu_w$ is $\bar{Y}_m - \bar{Y}_w$.
- Let us discuss the distribution of $\bar{Y}_m - \bar{Y}_w$.

$$\sim N(\mu_m - \mu_w, \frac{\sigma_m^2}{n_m} + \frac{\sigma_w^2}{n_w})$$

- if σ_m^2 and σ_w^2 are known, then this approximate normal distribution can be used to compute p-values for the test of the null hypothesis. In practice, however, these population variances are typically unknown so they must be estimated.
- Thus the *standard error* of $\bar{Y}_m - \bar{Y}_w$ is

$$SE(\bar{Y}_m - \bar{Y}_w) = \sqrt{\frac{s_m^2}{n_m} + \frac{s_w^2}{n_w}}$$

T-Test and Table

- Review the Theory

Assuming Case: the California School

Comparing Means from Different Populations

The Difference Between Two Means

- The t-statistic for testing the null hypothesis is constructed analogously to the t-statistic for testing a hypothesis about a single population mean, thus *t-statistic* for comparing two means is

$$t_{act} = \frac{\bar{Y}_m - \bar{Y}_w - d_0}{SE(\bar{Y}_m - \bar{Y}_w)}$$

- If both n_m and n_w are large, then this t-statistic has a standard normal distribution when the null hypothesis is true, thus $\bar{Y}_m - \bar{Y}_w = 0$.

T-Test and Table

- Review the Theory

Assuming Case: the California School

Comparing Means from Different Populations

Confidence Intervals for the Difference Between Two Means

- the 95% two-sided confidence interval for d consists of those values of d within ± 1.96 standard errors of $\bar{Y}_m - \bar{Y}_w$, thus $d = \mu_m - \mu_w$ is

$$(\bar{Y}_m - \bar{Y}_w) \pm 1.96SE(\bar{Y}_m - \bar{Y}_w)$$

T-Test and Table

- 单样本 t 检验

```
. sysuse auto,clear  
(1978 Automobile Data)  
. ttest price == 6000 if foreign == 0 ,level(90)
```

One-sample t test

Variable	Obs	Mean	Std. Err.	Std. Dev.	[90% Conf. Interval]	
price	52	6072.423	429.4911	3097.104	5352.903	6791.943

```
mean = mean(price) t = 0.1686  
Ho: mean = 6000 degrees of freedom = 51  
Ha: mean < 6000 Ha: mean != 6000 Ha: mean > 6000  
Pr(T < t) = 0.5666 Pr(|T| > |t|) = 0.8668 Pr(T > t) = 0.4334
```

- * level默认95%的水平
- * 结果p值大于0.1, 不能拒绝H0

T-Test and Table

● 独立样本 t 检验

- ▶ 一个变量利用另一个变量来分组比较

```
. sdtest price, by(foreign)
```

```
Variance ratio test
```

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
Domestic	52	6072.423	429.4911	3097.104	5210.184	6934.662
Foreign	22	6384.682	558.9942	2621.915	5222.19	7547.174
combined	74	6165.257	342.8719	2949.496	5481.914	6848.6

```
ratio = sd(Domestic) / sd(Foreign) f = 1.3953
Ho: ratio = 1 degrees of freedom = 51, 21
Ha: ratio < 1 Ha: ratio != 1 Ha: ratio > 1
Pr(F < f) = 0.7963 2*Pr(F > f) = 0.4073 Pr(F > f) = 0.2037
```

- * 方差齐性检验(F检验)
- * 对两个独立样本进行比较的时候, 首先要判断两总体方差是否相同, 即方差齐性。
- * 若两总体方差相等equal variances(方差齐), 则直接用t检验;
- * 若方差不齐, 选择unequal variances(方差不齐)的均值T检验去做, 加unequal选项。

T-Test and Table

- 独立样本 t 检验

- ▶ 一个变量利用另一个变量来分组比较

```
. ttest price, by(foreign)
```

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

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
Domestic	52	6072.423	429.4911	3097.104	5210.184	6934.662
Foreign	22	6384.682	558.9942	2621.915	5222.19	7547.174
combined	74	6165.257	342.8719	2949.496	5481.914	6848.6
diff		-312.2587	754.4488		-1816.225	1191.708

```
diff = mean(Domestic) - mean(Foreign)          t = -0.4139
Ho: diff = 0                                     degrees of freedom = 72
Ha: diff < 0                                     Ha: diff != 0          Ha: diff > 0
Pr(T < t) = 0.3401                               Pr(|T| > |t|) = 0.6802 Pr(T > t) = 0.6599
```

T-Test and Table

- 独立样本 t 检验

- ▶ 在两个变量间进行比较

```
. webuse fuel,clear
```

```
. sdtest mpg1 == mpg2
```

```
Variance ratio test
```

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
mpg1	12	21	.7881701	2.730301	19.26525	22.73475
mpg2	12	22.75	.9384465	3.250874	20.68449	24.81551
combined	24	21.875	.6264476	3.068954	20.57909	23.17091

```
ratio = sd(mpg1) / sd(mpg2)
```

```
f = 0.7054
```

```
Ho: ratio = 1
```

```
degrees of freedom = 11, 11
```

```
Ha: ratio < 1
```

```
Ha: ratio != 1
```

```
Ha: ratio > 1
```

```
Pr(F < f) = 0.2862
```

```
2*Pr(F < f) = 0.5725
```

```
Pr(F > f) = 0.7138
```

T-Test and Table

- 独立样本 t 检验

▶ 在两个变量间进行比较

```
. ttest mpg1 == mpg2, unpaired
```

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

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
mpg1	12	21	.7881701	2.730301	19.26525	22.73475
mpg2	12	22.75	.9384465	3.250874	20.68449	24.81551
combined	24	21.875	.6264476	3.068954	20.57909	23.17091
diff		-1.75	1.225518		-4.291568	.7915684

```
diff = mean(mpg1) - mean(mpg2)                                t = -1.4280
Ho: diff = 0                                                    degrees of freedom = 22
Ha: diff < 0                                                    Ha: diff != 0          Ha: diff > 0
Pr(T < t) = 0.0837        Pr(|T| > |t|) = 0.1673        Pr(T > t) = 0.9163
```

* unpaired 表示对两个不同变量检验，不是配对检验

T-Test and Table

- 配对样本 t 检验 (单样本 t 检验的扩展)

- ▶ 检验对象是配对样本观测值之差

```
. ttest mpg1==mpg2
```

```
Paired t test
```

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
mpg1	12	21	.7881701	2.730301	19.26525	22.73475
mpg2	12	22.75	.9384465	3.250874	20.68449	24.81551
diff	12	-1.75	.7797144	2.70101	-3.46614	-.0338602

```
mean(diff) = mean(mpg1 - mpg2)                                t = -2.2444
Ho: mean(diff) = 0                                           degrees of freedom = 11
Ha: mean(diff) < 0                                           Ha: mean(diff) != 0
Pr(T < t) = 0.0232                                           Pr(|T| > |t|) = 0.0463
                                                              Ha: mean(diff) > 0
                                                              Pr(T > t) = 0.9768
```

- * 没有 unpaired 选项
- * 结果 p 值小于 0.05, 拒绝 H0

- -ttest-的局限

- ▶ 每次只能对一个变量进行检验，无法批量对多个变量检验。
- ▶ 汇报结果过于详细，有时我们只需要一个相对精简的结果，如两组各自均值，均值差异，T-Statistic 或者 P-Value。
- ▶ 当待检验变量增加，ttest 命令费时费力。

T-Test and Table

- 多变量均值比较表格输出-ttable2-

```
ssc install ttable2
```

```
. sysuse auto,clear  
(1978 Automobile Data)
```

```
. ttable2 price wei len mpg, by(foreign) f(%6.2f)
```

Variables	G1(Domestic)	Mean1	G2(Foreign)	Mean2	MeanDiff
price	52	6072.42	22	6384.68	-312.26
weight	52	3317.12	22	2315.91	1001.21***
length	52	196.13	22	168.55	27.59***
mpg	52	19.83	22	24.77	-4.95***

T-Test and Table

- 多变量均值比较表格输出-ttable2-

```
. tab rep78
      Repair
Record 1978      Freq.      Percent      Cum.
-----+-----+-----+-----
      1             2         2.90         2.90
      2             8        11.59        14.49
      3            30        43.48        57.97
      4            18        26.09        84.06
      5            11        15.94       100.00
-----+-----+-----+-----
      Total          69        100.00

. ttable2 price wei len mpg if rep78==3|rep78==4, by(rep78)
```

Variables	G1(3)	Mean1	G2(4)	Mean2	MeanDiff
price	30	6429.233	18	6071.500	357.733
weight	30	3299.000	18	2870.000	429.000*
length	30	194.000	18	184.833	9.167
mpg	30	19.433	18	21.667	-2.233*

* 当组类别大于两类时，可以通过指定样本范围进行比较

- 结果导出-logout-

```
^^Issc install logout
^^I
^^Ilogout, save(ttable) excel replace : ttable2 price ///
^^I      wei len mpg, by(foreign) f(%6.2f)
^^Ilogout, save(ttable) word replace : ttable2 price ///
^^I      wei len mpg, by(foreign) f(%6.2f)
^^Ilogout, save(ttable) tex replace : ttable2 price ///
^^I      wei len mpg, by(foreign) f(%6.2f)
```

T-Test and Table

- 结果导出-t2docx-

```
ssc install t2docx

t2docx price weight length mpg ///
      using ttable1.docx,replace ///
      by(foreign)                ///
      title("表1: t检验")
```

T-Test and Table

- 结果导出-esttab-

```
. sysuse auto,clear
(1978 Automobile Data)

. local var price wei len mpg
. qui estpost ttest `var', by(foreign)
. esttab ., cell("mu_1(fmt(2)) mu_2(fmt(2)) b(star fmt(2)) t(fmt(2))") ///
> starlevels(* 0.10 ** 0.05 *** 0.01) replace noobs compress ///
> title(esttab_Table: T_test)

esttab_Table: T_test by group
```

(1)

	mu_1	mu_2	b	t
price	6072.42	6384.68	-312.26	-0.41
weight	3317.12	2315.91	1001.21***	6.25
length	196.13	168.55	27.59***	5.89
mpg	19.83	24.77	-4.95***	-3.63

- 结果导出-esttab-

```
sysuse auto,clear

local var price wei len mpg
qui estpost ttest `var', by(foreign)
esttab using ttable2.rtf, cell("mu_1(fmt(2)) mu_2(fmt(2)) b(star fmt(2)) t(fmt(2))") ///
      starlevels(* 0.10 ** 0.05 *** 0.01) replace noobs compress ///
      title(esttab_Table: T_test)
```

Section 2

Descriptive Statistics Table

Descriptive Statistics Table

- 描述性统计表格导出

- ▶ -logout-

```
logout, save(Desc1) word replace:      ///  
tabstat price wei len mpg rep78,      ///  
^^I ^^Istats(mean sd min p50 max) c(s) f(%6.2f)
```


Descriptive Statistics Table

- 描述性统计表格导出

- ▶ -sum2docx-

```
sum2docx price wei len mpg rep78 using Desc2.docx,replace ///
      stats(N mean(%9.2f) sd(%9.3f) min(%9.2f) median(%9.2f) max(%9.2f)) ///
      title(sum2docx_Table: Descriptive statistics)
```

- *仅sum2docx支持中文,其余命令不支持

- *能设置每个统计量的小数点位数

Descriptive Statistics Table

- 描述性统计表格导出

- ▶ -outreg2-

```
outreg2 using Desc3, sum(detail) replace word          ^^I      ///  
      keep(price wei len mpg rep78) eqkeep(N mean sd min p50 max) ///  
      fmt(f) sortvar(wage age grade)                 ///  
      title(outreg2_Table: Descriptive statistics)
```

- *若变量里有字符串变量, `outreg2`命令的处理最智能化:
- *会在窗口说明什么变量是字符型, 并在报告列表中自动剔除该变量
- *支持变量排序

Descriptive Statistics Table

- 描述性统计表格导出

- ▶ -esttab-

```
estpost summarize price wei len mpg rep78, detail
esttab using Desc4.rtf,                                     ///
    cells("count mean(fmt(2)) sd(fmt(2)) min(fmt(2)) p50(fmt(2)) max(fmt(2))") ///
    noobs compress replace title(esttab_Table: Descriptive statistics)
```

*能设置每个统计量的小数点位数

Section 3

Correlation Matrix Table

Correlation Matrix Table

- 相关系数矩阵导出

```
*<方法一> -logout-
logout, save(Corr1) word replace: pcorr price wei len mpg rep78, star(.05)

*<方法二> -esttab-
estpost correlate price wei len mpg rep78, matrix
esttab using Corr2.rtf,                                     ///
    unstack not noobs compress nogaps replace star(* 0.1 ** 0.05 *** 0.01) ///
    b(%8.3f) p(%8.3f) title(esttab_Table: correlation coefficient matrix)

*<方法三> -corr2docx-
corr2docx price wei len mpg rep78 using Corr3.docx,       ///
^^I^^I           replace spearman(ignore) pearson(pw) star          ///
^^I^^I           title(corr2docx_Table: correlation coefficient matrix)
```

Section 4

OLS Regression-Estimation

Subsection 1

Data Analysis Flow

- Data Analysis Flow

- ▶ Open the data, find the variables, and see the base case.
- ▶ Data Cleaning.
- ▶ Summary Statistics: Figures and Tables.
- ▶ Model Estimation and Hypothesis Testing.
- ▶ Report results, explain and analyze.

Subsection 2

Review the Theory

- Review the Theory

OLS Estimation: Simple Regression

Terminology for Simple Regression Model

- The linear regression model with one regressor is denoted by

$$Y_i = \beta_0 + \beta_1 X_i + u_i$$

- Where
 - Y_i is the **dependent variable**(Test Score)
 - X_i is the **independent variable** or regressor(Class Size or Student-Teacher Ratio)
 - $\beta_0 + \beta_1 X_i$ is the **population regression line** or the **population regression function**

- Review the Theory

Review for the previous lectures

The OLS Estimator

- The estimators of the slope and intercept that *minimize the sum of the squares* of \hat{u}_i , thus

$$\arg \min_{b_0, b_1} \sum_{i=1}^n \hat{u}_i^2 = \min_{b_0, b_1} \sum_{i=1}^n (Y_i - b_0 - b_1 X_i)^2$$

are called the **ordinary least squares (OLS) estimators** of β_0 and β_1 .

OLS estimator of β_1 :

$$b_1 = \hat{\beta}_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})}$$

- Review the Theory

Least Squares Assumptions

- ① Assumption 1: Conditional Mean is Zero
 - ② Assumption 2: Random Sample
 - ③ Assumption 3: Large outliers are unlikely
- If the 3 least squares assumptions hold the OLS estimators will be
 - **unbiased**
 - **consistent**
 - **normal sampling distribution**

- Review the Theory

Multiple regression model with k regressors

- The multiple regression model is

$$Y_i = \beta_0 + \beta_1 X_{1,i} + \beta_2 X_{2,i} + \dots + \beta_k X_{k,i} + u_i, i = 1, \dots, n \quad (4.1)$$

where

- Y_i is the **dependent variable**
- X_1, X_2, \dots, X_k are the **independent variables (includes one is our of interest and some control variables)**
- $\beta_j, j = 1 \dots k$ are slope coefficients on X_j corresponding.
- β_0 is the estimate *intercept*, the value of Y when all $X_j = 0, j = 1 \dots k$
- u_i is the regression *error term*, still all other factors affect outcomes.

- Review the Theory

Multiple Regression: Assumption

Multiple Regression: Assumption

- Assumption 1: The conditional distribution of u_i given X_{1i}, \dots, X_{ki} has mean zero, thus

$$E[u_i | X_{1i}, \dots, X_{ki}] = 0$$

- Assumption 2: $(Y_i, X_{1i}, \dots, X_{ki})$ are i.i.d.
- Assumption 3: Large outliers are unlikely.
- Assumption 4: No perfect multicollinearity.

- Review the Theory

- The OLS estimators $\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_k$ are *unbiased*.
- The OLS estimators $\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_k$ are *consistent*.
- The OLS estimators $\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_k$ are *normally distributed* in large samples.

- Multiple OLS estimator

$$\hat{\beta}_j = \frac{\sum_{i=1}^n \tilde{X}_{j,i} Y_i}{\sum_{i=1}^n \tilde{X}_{j,i}^2} \text{ for } j = 1, 2, \dots, k$$

Subsection 3

OLS in stata

OLS Regression-Estimation

- 普通最小二乘法 (OLS)

```
. *help reg  
. *regress depvar [indepvars] [if] [in] [weight] [, options] //因变量, 自变量
```

```
. sysuse auto, clear  
(1978 Automobile Data)
```

```
. reg price weight mpg turn foreign
```

Source	SS	df	MS	Number of obs	=	74
Model	334771309	4	83692827.3	F(4, 69)	=	19.23
Residual	300294087	69	4352088.22	Prob > F	=	0.0000
				R-squared	=	0.5271
				Adj R-squared	=	0.4997
Total	635065396	73	8699525.97	Root MSE	=	2086.2

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
weight	4.284532	.7404967	5.79	0.000	2.807282	5.761783
mpg	-.4660076	73.51407	-0.01	0.995	-147.1226	146.1905
turn	-229.2059	114.2423	-2.01	0.049	-457.1131	-1.298676
foreign	3221.415	706.4847	4.56	0.000	1812.017	4630.813
_cons	1368.197	4887.597	0.28	0.780	-8382.292	11118.69

OLS Regression-Estimation

- 普通最小二乘法 (OLS)

```
. regress weight length, noconstant //不包括截距项 (constant)
```

Source	SS	df	MS	Number of obs	=	74
Model	703869302	1	703869302	F(1, 73)	=	3450.13
Residual	14892897.8	73	204012.299	Prob > F	=	0.0000
Total	718762200	74	9713002.7	R-squared	=	0.9793
				Adj R-squared	=	0.9790
				Root MSE	=	451.68
weight	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
length	16.29829	.2774752	58.74	0.000	15.74528	16.8513

OLS Regression-Estimation

- 普通最小二乘法 (OLS)

```
. reg price weight mpg turn foreign, robust //稳健标准误 (robust)
Linear regression                               Number of obs   =          74
                                                F(4, 69)       =         12.46
                                                Prob > F       =         0.0000
                                                R-squared     =         0.5271
                                                Root MSE     =         2086.2
```

price	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
weight	4.284532	.9164881	4.67	0.000	2.456188	6.112876
mpg	-.4660076	84.34373	-0.01	0.996	-168.7271	167.7951
turn	-229.2059	136.4962	-1.68	0.098	-501.5084	43.09658
foreign	3221.415	690.7001	4.66	0.000	1843.506	4599.324
_cons	1368.197	6008.419	0.23	0.821	-10618.27	13354.66

OLS Regression-Estimation

- 回归结果

```
. regress price mpg weight foreign
```

Source	SS	df	MS	Number of obs	=	74
Model	317252881	3	105750960	F(3, 70)	=	23.29
Residual	317812515	70	4540178.78	Prob > F	=	0.0000
				R-squared	=	0.4996
				Adj R-squared	=	0.4781
Total	635065396	73	8699525.97	Root MSE	=	2130.8

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
mpg	21.8536	74.22114	0.29	0.769	-126.1758	169.883
weight	3.464706	.630749	5.49	0.000	2.206717	4.722695
foreign	3673.06	683.9783	5.37	0.000	2308.909	5037.212
_cons	-5853.696	3376.987	-1.73	0.087	-12588.88	881.4934

- 回归结果

```
predict yhat, xb           //price的拟合值
predict e, residual       //残差
vce                        //获取变量的方差—协方差矩阵

. test mpg = 20           //单变量检验
( 1)  mpg = 20
      F( 1, 70) = 0.00
      Prob > F = 0.9801

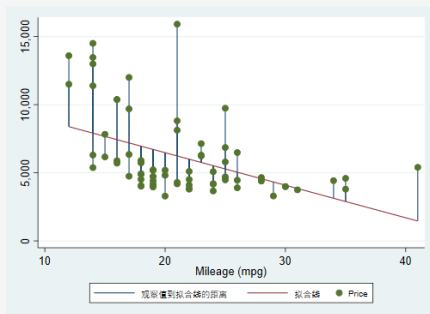
. test mpg weight foreign //联合检验
( 1)  mpg = 0
( 2)  weight = 0
( 3)  foreign = 0
      F( 3, 70) = 23.29
      Prob > F = 0.0000
```

OLS Regression-Estimation

● 回归结果

```
. qui reg price mpg
. predict yhat_p, xb
(option xb assumed; fitted values)
. twoway (rspike price yhat_p mpg )    ///
>      (lfit price mpg)                ///
>      (scatter price mpg ),          ///
>      legend(label(1 "观察值到拟合线的距离") label(2 "拟合线") row(1) size(small))

. graph export olsf.png, width(500) replace
(note: file olsf.png not found)
(file olsf.png written in PNG format)
```



Subsection 4

OLS Result Table

- 回归结果输出-esttab-

- ▶ word 文档

```
sysuse nlsw88, clear

reg wage age married occupation
est store m1
reg wage age married collgrad occupation
est store m2
xi: reg wage age married collgrad occupation i.race
est store m3

esttab m1 m2 m3 using ols.rtf, scalar(r2 r2_a N F) compress ///
      star(* 0.1 ** 0.05 *** 0.01) ///
      b(%6.3f) t(%6.3f) r2(%9.3f) ar2 ///
      mtitles("OLS-1" "OLS-2" "OLS-3") ///
      title(esttab_Table: regression result)
```


● 回归结果输出-esttab-

▶ Tex 文档

```
esttab m1 m2 m3 using ols.tex, replace    ///
      star( * 0.10 ** 0.05 *** 0.01 ) compress  ///
      b(%6.3f) t(%6.3f) r2(%9.3f) ar2        ///
      mtitles("OLS-1" "OLS-2" "OLS-3")      ///
      title(esttab_Table: regression result)  ///
      booktabs page width(\hsize)
```

/*

esttab 的 LaTeX 输出的专有选项:

1. `booktabs`: 用 `booktabs` 宏包输出表格(三线表格)。
2. `page[(packages)]`: 创建完成的 LaTeX 文档以及添加括号里的宏包
3. 如果写了 `booktabs` 选项, 则 `page[(packages)]` 将自动添加 `\usepackagebooktabs`。
4. `alignment(cccccc)`: 定义从第二列开始的列对齐方式(默认居中)。
5. `width(\hsize)`: 可以使得表格宽度为延伸至页面宽度
6. `fragment`: 不输出表头表尾, 只输出表格本身内容, 其不能与 `page[(packages)]` 选项共存。

*/

OLS Regression-Estimation

表 1: esttab_Table: regression result

	(1) OLS-1	(2) OLS-2	(3) OLS-3
age	-0.064 (-1.637)	-0.059 (-1.579)	-0.067* (-1.796)
married	-0.469* (-1.873)	-0.472** (-1.983)	-0.629** (-2.578)
occupation	-0.284*** (-8.055)	-0.384*** (-11.251)	-0.370*** (-10.756)
collgrad		4.220*** (15.444)	4.133*** (15.051)
_lrace_2			-0.784*** (-2.897)
_lrace_3			-0.224 (-0.210)
_cons	11.910*** (7.654)	11.168*** (7.545)	11.753*** (7.878)
<i>N</i>	2237	2237	2237
<i>R</i> ²	0.031	0.125	0.128
adj. <i>R</i> ²	0.030	0.123	0.126

t statistics in parentheses

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Section 5

Nonlinear Regression

Subsection 1

Review the Theory

- Review the Theory

Nonlinear in X s

Nonlinear Regression Regression Functions

- So far our regression model is

$$Y_i = \beta_0 + \beta_1 X_{1,i} + \dots + \beta_k X_{k,i} + u_i$$

- The effect of Y on a change in X_j by 1 (unit) is constant and equals β_j :

$$\beta_j = \frac{\partial Y_i}{\partial X_{ji}}$$

- But if a relation between Y and X is nonlinear:
 - The effect on Y of a change in X depends on the value of X – that is, the *marginal effect of X is not constant*.
 - A linear regression is misspecified – the functional form is wrong.
 - The estimator of the effect on Y of X is biased (a special case of OVB).
- The solution to this is to estimate a regression function that is nonlinear in X .

Nonlinear Regression

- Review the Theory

Nonlinear in X_s

Two Cases:

- 1 The effect of change in X_1 on Y depends on X_1 itself.
 - eg. the effect of a change in class size on test scores is bigger when initial class size is small.
- 2 The effect of change in X_1 on Y depends on another variable, like X_2 .
 - eg. the effect of class size on test scores depends on the percentage of disadvantaged pupils in the class.

- Review the Theory

Two Complementary Approaches:

1 Polynomials in X

- The population regression function is approximated by a quadratic, cubic, or higher-degree polynomial.

2 Logarithmic transformations

- Y and/or X is transformed by taking its logarithm
- this gives a *percentages* interpretation that makes sense in many applications

- Review the Theory

Introduction

- The product of two variables is called an **interaction term**.
- Try to answer *how the effect on Y of a change in an independent variable depends on the value of another independent variable*.
- Consider three cases:
 - Interactions between two binary variables.
 - Interactions between a binary and a continuous variable.
 - Interactions between two continuous variables.

Subsection 2

In practice

Nonlinear Regression

- In practice

```
. use caschool.dta, clear
```

```
. ***** 数据转化 *****

. gen avginc2 = avginc^2    //avginc平方项
. gen avginc3 = avginc^3    //avginc立方项

. gen loginc = ln(avginc)    //avginc对数
. gen logtest = ln(testscr) //testscr对数

. gen loginc2 = loginc^2    //avginc对数平方项
. gen loginc3 = loginc^3    //avginc对数立方项

. gen histr = (str>=20)     //histr = 1 if the str > 20
. gen hiel = (el_pct >= 10) //hiel = 1 if the el_pct > 10%
```

Nonlinear Regression

- In practice

```
. gen hisxhie = histr*hiel //D1i × D2i
. gen strxhie = str*hiel //Xi × Di
. gen strxelpc = str*el_pct //X1i × X2i

. gen sttr2 = str^2 //str平方项
. gen sttr3 = str^3 //str立方项

. gen str2hie = sttr2*hiel //Xi × Di
. gen str3hie = sttr3*hiel //Xi × Di
```

Nonlinear Regression

- In practice

$$\text{Test Score} = \beta_0 + \beta_1 STR + \beta_2 \text{English} + \beta_3 \text{Lunch} + u$$

```
. reg testscr str el_pct meal_pct, r //多元线性回归模型
Linear regression                               Number of obs   =       420
                                                F(3, 416)       =       453.48
                                                Prob > F        =       0.0000
                                                R-squared       =       0.7745
                                                Root MSE      =       9.0801
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	-.9983092	.2700799	-3.70	0.000	-1.529201	-.4674178
el_pct	-.1215733	.0328317	-3.70	0.000	-.18611	-.0570366
meal_pct	-.5473456	.0241072	-22.70	0.000	-.5947328	-.4999583
_cons	700.15	5.56845	125.74	0.000	689.2042	711.0958

```
. est store m1
```

Nonlinear Regression

- In practice

$$\text{Test Score} = \beta_0 + \beta_1 STR + \beta_2 \text{ English} + \beta_3 \text{ Lunch} + \beta_4 \ln(\text{Income}) + u$$

```
. reg testscr str el_pct meal_pct loginc, r //控制对数收入的影响
```

```
Linear regression                Number of obs   =          420
                                F(4, 415)       =          417.20
                                Prob > F             =           0.0000
                                R-squared            =           0.7962
                                Root MSE        =           8.6426
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	-.734326	.2567803	-2.86	0.004	-1.239078	-.2295738
el_pct	-.1755344	.0336606	-5.21	0.000	-.241701	-.1093678
meal_pct	-.3982342	.0331741	-12.00	0.000	-.4634443	-.333024
loginc	11.56897	1.818811	6.36	0.000	7.993736	15.1442
_cons	658.552	8.641528	76.21	0.000	641.5653	675.5386

```
. est store m2
```

Nonlinear Regression

- In practice

$$\text{Test Score} = \beta_0 + \beta_1 \text{STR} + \beta_2 \text{HiEL} + \beta_3 (\text{STR} \times \text{HiEL}) + u$$

```
. reg testscr str hiel strxhiel, r //考虑生师比与英语学习者百分比的交互影响
Linear regression                               Number of obs   =       420
                                                F(3, 416)      =       63.67
                                                Prob > F       =       0.0000
                                                R-squared     =       0.3103
                                                Root MSE     =       15.88
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	-.9684601	.5891016	-1.64	0.101	-2.126447	.1895268
hiel	5.639141	19.51456	0.29	0.773	-32.72029	43.99857
strxhiel	-1.276613	.9669194	-1.32	0.187	-3.17727	.6240436
_cons	682.2458	11.86781	57.49	0.000	658.9175	705.5742

```
. est store m3
```

- In practice

```
. test str strxhiel //联合假设的F 统计量和P 值
( 1)  str = 0
( 2)  strxhiel = 0
      F( 2, 416) = 5.64
      Prob > F = 0.0038
```

Nonlinear Regression

- In practice

$$\text{Test Score} = \beta_0 + \beta_1 \text{STR} + \beta_2 \text{HiEL} + \beta_3 (\text{STR} \times \text{HiEL}) \\ + \beta_4 \text{Lunch} + \beta_5 \ln(\text{Income}) + u$$

```
. reg testscr str hiel strxhiel meal_pct loginc, r //控制学生特征的基础上考虑交互影响  
Linear regression
```

```
Number of obs   =      420  
F(5, 414)       =     335.85  
Prob > F        =     0.0000  
R-squared       =     0.7974  
Root MSE      =     8.6286
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	-.5310319	.3418468	-1.55	0.121	-1.203004	.14094
hiel	5.498208	9.795	0.56	0.575	-13.75593	24.75234
strxhiel	-.5776664	.4957779	-1.17	0.245	-1.552222	.3968896
meal_pct	-.4113776	.0288276	-14.27	0.000	-.4680443	-.3547109
loginc	12.12447	1.797513	6.75	0.000	8.591078	15.65786
_cons	653.6661	9.869378	66.23	0.000	634.2658	673.0665

```
. est store m4
```


- In practice

```
. test str strxhiel //联合假设的F统计量和P值
( 1)  str = 0
( 2)  strxhiel = 0
      F( 2, 414) = 5.92
      Prob > F = 0.0029
```

Nonlinear Regression

- In practice

$$\text{TestScore} = \beta_0 + \beta_1 \times \text{STR} + \beta_2 \times \text{STR}^2 + \beta_3 \times \text{STR}^3 \\ + \beta_4 \text{HiEL} + \beta_5 \text{Lunch} + \beta_6 \ln(\text{Income}) + u$$

```
. reg testscr str sttr2 sttr3 hiel meal_pct loginc, r //考虑师生比的非线性影响
Linear regression      Number of obs   =      420
                      F(6, 413)         =     281.14
                      Prob > F          =     0.0000
                      R-squared         =     0.8011
                      Root MSE       =     8.5593
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	64.33964	24.86053	2.59	0.010	15.47069	113.2086
sttr2	-3.423925	1.249901	-2.74	0.006	-5.880886	-.9669636
sttr3	.0592895	.020763	2.86	0.005	.0184752	.1001037
hiel	-5.473991	1.033762	-5.30	0.000	-7.506081	-3.4419
meal_pct	-.4200571	.0285258	-14.73	0.000	-.4761309	-.3639833
loginc	11.74819	1.771437	6.63	0.000	8.266032	15.23035
_cons	252.0458	163.6341	1.54	0.124	-69.61384	573.7055

```
. est store m5
```

Nonlinear Regression

- In practice

```
. test str sttr2 sttr3 //联合假设的F统计量和P值
( 1) str = 0
( 2) sttr2 = 0
( 3) sttr3 = 0
      F( 3, 413) = 6.31
      Prob > F = 0.0003
. test sttr2 sttr3 //联合假设的F统计量和P值
( 1) sttr2 = 0
( 2) sttr3 = 0
      F( 2, 413) = 6.17
      Prob > F = 0.0023
```

Nonlinear Regression

- In practice

$$\begin{aligned} \text{TestScore} = & \beta_0 + \beta_1 \times \text{STR} + \beta_2 \times \text{STR}^2 + \beta_3 \times \text{STR}^3 \\ & + \beta_4 \text{HiEL} + \beta_5 (\text{STR} \times \text{HiEL}) + \beta_6 (\text{STR}^2 \times \text{HiEL}) + \\ & \beta_7 (\text{STR}^3 \times \text{HiEL}) + \beta_8 \text{Lunch} + \beta_9 \ln(\text{Income}) + u \end{aligned}$$

```
. reg testscr str sttr2 sttr3 hiel strxhiel str2hiel str3hiel meal_pct loginc, r  
//考虑所有因素  
Linear regression
```

```
Number of obs   =    420  
F(9, 410)       =   199.88  
Prob > F        =    0.0000  
R-squared       =    0.8031  
Root MSE      =    8.5475
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	83.70272	28.4967	2.94	0.003	27.68485	139.7206
sttr2	-4.380841	1.441017	-3.04	0.003	-7.213544	-1.548139
sttr3	.0749236	.024008	3.12	0.002	.0277295	.1221178
hiel	816.0904	327.6745	2.49	0.013	171.9587	1460.222
strxhiel	-123.2842	50.21281	-2.46	0.014	-221.9909	-24.57757
str2hiel	6.12124	2.541978	2.41	0.016	1.124303	11.11818
str3hiel	-.1006	.0425094	-2.37	0.018	-.1841636	-.0170364
meal_pct	-.4177876	.0287011	-14.56	0.000	-.4742073	-.3613679
loginc	11.80035	1.778008	6.64	0.000	8.305197	15.2955
_cons	122.3464	185.518	0.66	0.510	-242.3388	487.0315

```
. est store m6
```

Nonlinear Regression

- In practice

```
. test str sttr2 sttr3 strxhiel str2hiel str3hiel //联合假设的F 统计量和P 值
( 1)  str = 0
( 2)  sttr2 = 0
( 3)  sttr3 = 0
( 4)  strxhiel = 0
( 5)  str2hiel = 0
( 6)  str3hiel = 0
      F( 6, 410) = 4.96
      Prob > F = 0.0001
```

- In practice

```
. test sttr2 sttr3 //联合假设的F 统计量和P 值
( 1) sttr2 = 0
( 2) sttr3 = 0
      F( 2, 410) = 5.81
      Prob > F = 0.0033
```

- In practice

```
. test strxh1el str2h1el str3h1el //联合假设的F 统计量和P 值
( 1) strxh1el = 0
( 2) str2h1el = 0
( 3) str3h1el = 0
      F( 3, 410) = 2.69
      Prob > F = 0.0460
```

Nonlinear Regression

- In practice

$$\text{TestScore} = \beta_0 + \beta_1 \times \text{STR} + \beta_2 \times \text{STR}^2 + \beta_3 \times \text{STR}^3 \\ + \beta_4 \text{ English} + \beta_5 \text{ Lunch} + \beta_6 \ln(\text{Income}) + u$$

```
. reg testscr str sttr2 sttr3 el_pct meal_pct loginc, r //除交互影响外所有因素
Linear regression                Number of obs   =       420
                                F(6, 413)         =       280.81
                                Prob > F             =       0.0000
                                R-squared            =       0.8007
                                Root MSE         =       8.5679
```

testscr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
str	65.28595	25.25864	2.58	0.010	15.63443	114.9375
sttr2	-3.465567	1.270893	-2.73	0.007	-5.963793	-.9673414
sttr3	.059909	.0211205	2.84	0.005	.0183919	.1014262
el_pct	-.165687	.0343657	-4.82	0.000	-.2332405	-.0981334
meal_pct	-.4024177	.0332667	-12.10	0.000	-.4678108	-.3370246
loginc	11.50894	1.806403	6.37	0.000	7.958044	15.05983
_cons	244.8025	165.7221	1.48	0.140	-80.9614	570.5665

```
. est store m7
```


- In practice

```
. test str sttr2 sttr3 //联合假设的F 统计量和P 值
( 1)  str = 0
( 2)  sttr2 = 0
( 3)  sttr3 = 0
      F( 3, 413) = 5.91
      Prob > F = 0.0006
```

- In practice

```
. test sttr2 sttr3 //联合假设的F 统计量和P 值
( 1) sttr2 = 0
( 2) sttr3 = 0
      F( 2, 413) = 5.96
      Prob > F = 0.0028
```

- In practice

```
*导出到Latex (或Word)
. esttab m* using nl_ols.tex, replace           ///
      star( * 0.10 ** 0.05 *** 0.01 ) compress  ///
      b(%6.3f) t(%6.3f) r2(%9.3f) ar2         ///
      booktabs page width(\\hsize)
(output written to nl_ols.tex)
```

Nonlinear Regression

- In practice

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	testsc	testsc	testsc	testsc	testsc	testsc	testsc
str	-0.998*** (-3.696)	-0.734*** (-2.860)	-0.968 (-1.644)	-0.531 (-1.553)	64.340*** (2.588)	83.703*** (2.937)	65.286** (2.585)
el_pct	-0.122*** (-3.703)	-0.176*** (-5.215)					-0.166*** (-4.821)
meal_pct	-0.547*** (-22.705)	-0.398*** (-12.004)		-0.411*** (-14.270)	-0.420*** (-14.726)	-0.418*** (-14.556)	-0.402*** (-12.097)
loginc		11.569*** (6.361)		12.124*** (6.745)	11.748*** (6.632)	11.800*** (6.637)	11.509*** (6.371)
hiel			5.639 (0.289)	5.498 (0.561)	-5.474*** (-5.295)	816.090** (2.491)	
strxhiel			-1.277 (-1.320)	-0.578 (-1.165)		-123.284** (-2.455)	
sttr2					-3.424*** (-2.739)	-4.381*** (-3.040)	-3.466*** (-2.727)
sttr3					0.059*** (2.856)	0.075*** (3.121)	0.060*** (2.837)
str2hiel						6.121** (2.408)	
str3hiel						-0.101** (-2.367)	
_cons	700.150*** (125.735)	658.552*** (76.208)	682.246*** (57.487)	653.666*** (66.232)	252.046 (1.540)	122.346 (0.659)	244.803 (1.477)
<i>N</i>	420	420	420	420	420	420	420
<i>R</i> ²	0.775	0.796	0.310	0.797	0.801	0.803	0.801
adj. <i>R</i> ²	0.773	0.794	0.305	0.795	0.798	0.799	0.798

Section 6

Nonlinear Regression (continued)

Subsection 1

Dummy Variables

Nonlinear Regression (continued)

- Dummy Variables

```
. use Nations2.dta,clear
(UN Human Development Indicators)
. des
Contains data from Nations2.dta
  obs:          194
  vars:          13
```

```
UN Human Development Indicators
2 Jul 2012 06:11
```

variable name	storage type	display format	value label	variable label
country	str21	%21s		Country
region	byte	%8.0g	region	Region
gdp	float	%9.0g		Gross domestic product per cap 2005\$, 200
school	float	%9.0g		Mean years schooling (adults) 2005/2010
adfert	float	%8.0g		Adolescent fertility: births/1000 fem 15-
				2010
chldmort	float	%9.0g		Prob dying before age 5/1000 live births
				2005/2009
life	float	%9.0g		Life expectancy at birth 2005/2010
pop	float	%9.0g		Population 2005/2010
urban	float	%9.0g		Percent population urban 2005/2010
femlab	float	%9.0g		Female/male ratio in labor force 2005/200
literacy	float	%9.0g		Adult literacy rate 2005/2009
co2	float	%9.0g		Tons of CO2 emitted per cap 2005/2006
gini	float	%9.0g		Gini coef income inequality 2005/2009

```
Sorted by: region country
```

Nonlinear Regression (continued)

- Dummy Variables

```
. tab region,gen(reg) //为每一类创建虚拟变量
```

Region	Freq.	Percent	Cum.
Africa	52	26.80	26.80
Americas	35	18.04	44.85
Asia	49	25.26	70.10
Europe	43	22.16	92.27
Oceania	15	7.73	100.00
Total	194	100.00	

```
. gen loggdp=ln(gdp)  
(15 missing values generated)
```

```
. label values reg1 reg1
```

```
. label define reg1 0 "others" 1 "Africa" //reg1 = 1 (Africa); = 0 (elsewhere)
```


Nonlinear Regression (continued)

- Dummy Variables

```
. eststo : qui reg life reg1
. eststo : qui reg life loggdp chldmort
. eststo : qui reg life reg1 reg2 reg3 reg4 loggdp chldmort
. eststo : qui reg life reg1 reg2 reg4 loggdp chldmort
. eststo : qui reg life reg1 loggdp chldmort

. esttab using e2.tex, replace nonumbers           ///
  title(Dummy_Var_Example Table)                 ///
  mtitles(m1 m2 m3 m4)                           ///
  star( * 0.10 ** 0.05 *** 0.01 ) compress       ///
  b(%6.3f) t(%6.3f) r2(%9.3f)                    ///
  booktabs page
(output written to e2.tex)
```

Nonlinear Regression (continued)

- Dummy Variables

Table 1: Dummy_Var_Example Table

	m1	m2	m3	m4	est5
reg1	-16.721*** (-15.175)		-2.151* (-1.775)	-2.927*** (-3.570)	-3.144*** (-3.979)
loggdp		1.525*** (4.834)	1.407*** (4.412)	1.422*** (4.468)	1.483*** (4.894)
chldmort		-0.146*** (-19.041)	-0.127*** (-14.605)	-0.127*** (-14.610)	-0.128*** (-14.781)
reg2			1.487 (1.264)	0.692 (0.933)	
reg3			0.984 (0.871)		
reg4			1.455 (1.213)	0.649 (0.852)	
_cons	73.211*** (128.333)	62.286*** (20.464)	62.162*** (20.043)	62.821*** (20.901)	62.657*** (21.430)
<i>N</i>	194	178	178	178	178
<i>R</i> ²	0.545	0.880	0.891	0.891	0.890

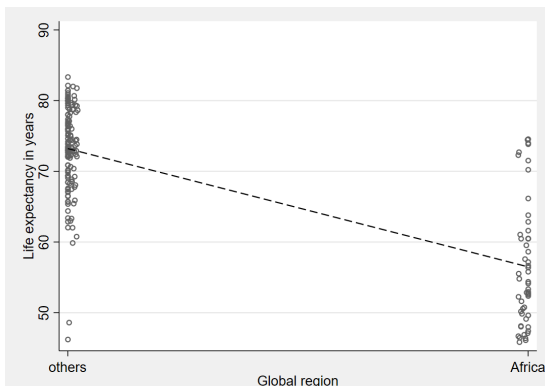
t statistics in parentheses

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Nonlinear Regression (continued)

• Dummy Variables

```
. predict lifehat
. graph twoway scatter life reg1, msymbol(oh S) jitter(5) ///
  || lfit life reg1 ///
  || , legend(off) xlabel(0 "others" 1 "Africa") scheme(s2mono) ///
  xtitle("Global region") ytitle("Life expectancy in years")
. cap graph export nat1.png, replace
```



Nonlinear Regression (continued)

- Dummy Variables

- ▶ Explanation:

m1 : The difference between the two means equals -16.72 years.

m2 : **loggpd** and **chldmort**(child mortality rate) together explain about 88% of the variance in life expectancy(m2). Including four dummy variables for regions 1–4 raises this only to about 89% (m3).

m3 : It is not possible to include all five in one regression because of **multicollinearity**. None of the regional dummy variables have significant effects.

m4 : Dropping reg3, the weakest of these predictors. The coefficient on reg1 now appears significant.

m5 : A reduced model.

Conclusion : The differences in life expectancy among **other regions** of the world are largely accounted for by variations in **wealth** and **child mortality**, but in **Africa** there are circumstances at work (such as **wars**) that further depress life expectancy.

Subsection 2

Interaction Effects

- Interaction Effects

- ▶ consider some different variables:

- per capita carbon dioxide emissions (`co2`)

- percent of the population living in urban areas (`urban`)

- dummy variable `reg4` defined as 1 for European countries and 0 for all others

- form an interaction term named `urb_reg4` by multiplying the dummy variable `reg4` times the measurement variable `urban`

- Interaction Effects

```
. label val reg4 reg4
. label define reg4 0 "others" 1 "Europe"
. gen logco2 = log10(co2)
(9 missing values generated)
. label var logco2 "log10(per cap CO2)"

. gen urb_reg4 = urban * reg4
. label variable urb_reg4 "interaction urban*reg4 (Europe)"
```

- Interaction Effects

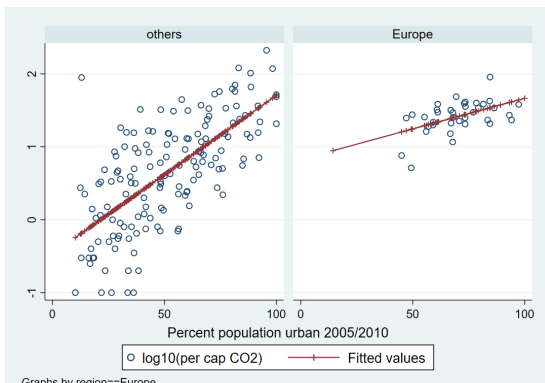
```
. qui reg logco2 urban reg4 urb_reg4
. qui reg logco2 c.urban i.reg4 c.urban#i.reg4
. qui reg logco2 c.urban##i.reg4

* factor-variable :
* i. indicator variables
* c. continuous variables
* # an interaction between two variables
* ## factorial interaction which automatically includes all the lower-level
interactions involving those variables
```


Nonlinear Regression (continued)

• Interaction Effects

```
. predict co2hat
(option xb assumed; fitted values)
. graph twoway scatter logco2 urban, msymbol(Oh)   ///
  || connect co2hat urban, msymbol(+)           ///
  || , by(reg4)
. cap graph export nat2.png, replace
```



Nonlinear Regression (continued)

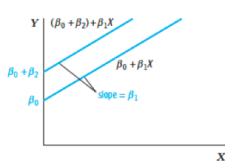
- Interaction Effects

- ▶ Explanation:

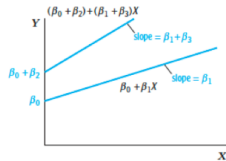
Interactions Between Independent Variables

Interactions: a Continuous and a Binary Variable

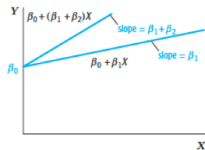
FIGURE 8.8 Regression Functions Using Binary and Continuous Variables



(a) Different intercepts, same slope



(b) Different intercepts, different slopes



(c) Same intercept, different slopes

Interactions of binary variables and continuous variables can produce three different population regression functions:

(a) $\beta_0 + \beta_1 X + \beta_2 D$ allows for different intercepts but has the same slope, (b) $\beta_0 + \beta_1 X + \beta_2 D + \beta_3 (X \times D)$ allows

- Interaction Effects

- ▶ Explanation:

The line in the left-hand ($\text{reg4} = 0$) panel has a slope of .0217 and y-intercept $-.4682$.

The line in the right panel ($\text{reg4} = 1$) has a less-steep slope (.0084) and a higher y-intercept (.826).

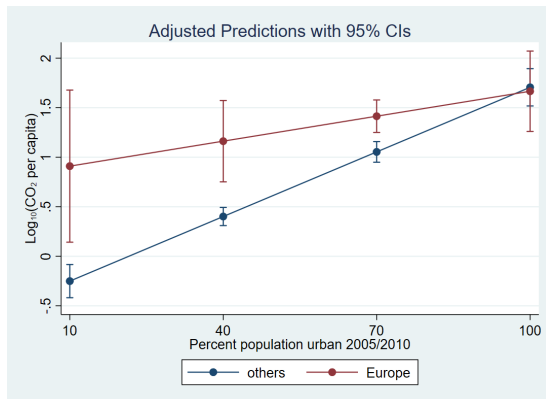
No European countries exhibit the low-urbanization, low-CO2 profile seen in other parts of the world.

Even European nations with middling urbanization have relatively high CO2 emissions.

Nonlinear Regression (continued)

• Interaction Effects

```
. qui margins, at(urban = (10(30)100) reg4 = (0 1)) vsquish  
. marginsplot, ytitle("Log10{CO2 per capita}") xlabel(10(30)100)  
  
. cap graph export nat3.png, replace
```



Nonlinear Regression (continued)

- Interaction Effects

- ▶ Interactions Between Two Continuous Variables
- ▶ **centering** makes their main effects easier to interpret
- ▶ Interacting variables have been centered, can be interpreted as the effect of each variable when the other is at its mean.

```
. summarize urban loggdp
```

Variable	Obs	Mean	Std. Dev.	Min	Max
urban	194	55.43488	23.4391	10.25	100
loggdp	179	8.693936	1.297024	5.634075	11.22399

Nonlinear Regression (continued)

- Interaction Effects

```
. sum urban
  Variable |          Obs          Mean    Std. Dev.      Min      Max
-----|-----
    urban |          194    55.43488    23.4391    10.25    100
. gen urban0 = urban - r(mean)

. sum loggdp
  Variable |          Obs          Mean    Std. Dev.      Min      Max
-----|-----
   loggdp |          179     8.693936    1.297024    5.634075    11.22399
. gen loggdp0 = loggdp - r(mean)
(15 missing values generated)

. gen urb_gdp = urban0 * loggdp0
(15 missing values generated)
```

Nonlinear Regression (continued)

- Interaction Effects

```
. reg logco2 c.loggdp0 c.urban0 c.loggdp0#c.urban0
```

Source	SS	df	MS	Number of obs	=	175
Model	83.4990751	3	27.833025	F(3, 171)	=	371.66
Residual	12.8060512	171	.074889188	Prob > F	=	0.0000
				R-squared	=	0.8670
				Adj R-squared	=	0.8647
Total	96.3051263	174	.553477737	Root MSE	=	.27366

logco2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
loggdp0	.4848767	.0242429	20.00	0.000	.4370228	.5327306
urban0	.0025141	.00137	1.84	0.068	-.0001903	.0052185
c.loggdp0#c.urban0	-.0035963	.0007565	-4.75	0.000	-.0050895	-.0021031
_cons	.8903587	.0267759	33.25	0.000	.8375049	.9432124

Nonlinear Regression (continued)

- Interaction Effects

- ▶ Explanation :

Interactions Between Independent Variables

Interactions Between Two Continuous Variables

- Thus the effect on Y of a change in X_1 , holding X_2 constant, is

$$\frac{\Delta Y}{\Delta X_1} = \beta_1 + \beta_3 X_2$$

- A similar calculation shows that the effect on Y of a change ΔX_1 in X_2 , holding X_1 constant, is

$$\frac{\Delta Y}{\Delta X_2} = \beta_2 + \beta_3 X_1$$

- That is, if X_1 changes by ΔX_1 and X_2 changes by ΔX_2 , then the expected change in Y

$$\Delta Y = (\beta_1 + \beta_3 X_2)\Delta X_1 + (\beta_2 + \beta_3 X_1)\Delta X_2 + \beta_3 \Delta X_1 \Delta X_2$$

Nonlinear Regression (continued)

- Interaction Effects

- ▶ Explanation :

predicted $\log\text{co2}$ rises by 0.65 with each 1-unit increase in $\log\text{gdp}$, when urban is at its mean

predicted $\log\text{co2}$ rises by only a small amount, .0025, with each 1-unit increase in urban when $\log\text{gdp}$ is at its mean.

each 1-unit increase in urbanization, the effect of $\log\text{gdp}$ on $\log\text{co2}$ becomes weaker, decreasing by $-.004$.

二氧化碳排放量随着财富的增加而增加，但在城市化程度较高的国家，二氧化碳排放量的增加幅度较小。

Section 7

Making Regression Tables in Stata : estout package

Subsection 1

Description

Making Regression Tables in Stata : estout package

- Description

- ▶ The estout package provides tools for making regression tables in Stata. The package currently contains the following commands.
 - `esttab` : A command for **publication-style regression tables** that display nicely in Stata's results window, and can be exported to various formats such as **CSV, RTF, HTML, LaTeX**.
 - `estout` : A **generic program for making a table** from one or more sets of estimation results. `estout` is **the engine behind esttab**.
 - `eststo` : A utility command to **store estimation results** for later tabulation.
 - `estadd` : A utility command to **add additional results** to an existing estimation set.
 - `estpost` : A utility command to **post results** from various **non-eclass commands** as estimation results (so that they can be tabulated).

Making Regression Tables in Stata : estout package

- Installation

```
ssc install estout, replace  
help esttab
```

- Reference

Making regression tables in Stata

Subsection 2

-esttab-

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ -esttab- is a wrapper for -estout-

```
esttab [ namelist ] [ using filename ] [, options estout_opt
```

- The procedure :

- ▶ Store a number of models
- ▶ Apply esttab to these stored estimation sets to compose a regression
- ▶ Table produces a fully formatted right away

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Example

```
. sysuse auto, clear
(1978 Automobile Data)
. qui eststo: reg price weight mpg
. qui eststo: reg price weight mpg foreign
. esttab //Models stored are automatically picked up by esttab
```

	(1)	(2)	(3)	(4)
	price	price	price	price
weight	1.747** (2.72)	3.465*** (5.49)	1.747** (2.72)	3.465*** (5.49)
mpg	-49.51 (-0.57)	21.85 (0.29)	-49.51 (-0.57)	21.85 (0.29)
foreign		3673.1*** (5.37)		3673.1*** (5.37)
_cons	1946.1 (0.54)	-5853.7 (-1.73)	1946.1 (0.54)	-5853.7 (-1.73)

N 74 74 74 74

```
t statistics in parentheses
* p<0.05, ** p<0.01, *** p<0.001
. eststo clear //removes the models from memory

* eststo : Store the regression models.
```


Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Or ->

```
. sysuse auto, clear
(1978 Automobile Data)
. qui reg price weight mpg
. est store model1
. qui reg price weight mpg foreign
. est store model2
. esttab model1 model2
```

	(1)	(2)
	price	price
weight	1.747** (2.72)	3.465*** (5.49)
mpg	-49.51 (-0.57)	21.85 (0.29)
foreign		3673.1*** (5.37)
_cons	1946.1 (0.54)	-5853.7 (-1.73)
N	74	74

```
t statistics in parentheses
* p<0.05, ** p<0.01, *** p<0.001
. est clear
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Standard errors, p-values, and summary statistics

* Default : (t-statistics) and the number of observations in the table footer

```
. sysuse auto, clear
(1978 Automobile Data)
. qui eststo: reg price weight mpg
. qui eststo: reg price weight mpg foreign
. esttab, se ar2 //replace by standard errors and add the adjusted R-squared
```

	(1)	(2)
	price	price
weight	1.747** (0.641)	3.465*** (0.631)
mpg	-49.51 (86.16)	21.85 (74.22)
foreign		3673.1*** (684.0)
_cons	1946.1 (3597.0)	-5853.7 (3377.0)
N	74	74
adj. R-sq	0.273	0.478

Standard errors in parentheses

* p<0.05, ** p<0.01, *** p<0.001

* t-statistics can also be replaced by : p, ci, aux

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Standard errors, p-values, and summary statistics

* Further summary statistics options : pr2, bic, scalars()

```
. esttab, p scalars(F df_m df_r)
```

	(1)	(2)
	price	price
weight	1.747** (0.008)	3.465*** (0.000)
mpg	-49.51 (0.567)	21.85 (0.769)
foreign		3673.1*** (0.000)
_cons	1946.1 (0.590)	-5853.7 (0.087)
N	74	74
F	14.74	23.29
df_m	2	3
df_r	71	70

p-values in parentheses

* p<0.05, ** p<0.01, *** p<0.001

Making Regression Tables in Stata : estout package

• -esttab-

► Numerical formats

- * Default :
- * t-statistics are printed using 2 decimal places.
- * R-squared measures are printed using 3 decimal places.
- * point estimates, standard errors using an adaptive display format.(a3)

```
. esttab, b(a6) p(4) r2(4) nostar
```

	(1)	(2)
	price	price
weight	1.746559 (0.0081)	3.464706 (0.0000)
mpg	-49.51222 (0.5673)	21.85360 (0.7693)
foreign		3673.060 (0.0000)
_cons	1946.069 (0.5902)	-5853.696 (0.0874)
N	74	74
R-sq	0.2934	0.4996

p-values in parentheses

- * increase precision for the point estimates
- * display p-values and the R-squared using 2 decimal places

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Labels, titles, and notes

```
. esttab, label ///  
    title(This is a regression table) ///  
    nonumbers mtitles("Model A" "Model B") ///  
    addnote("Source: auto.dta")
```

This is a regression table

	Model A	Model B
Weight (lbs.)	1.747** (2.72)	3.465*** (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1*** (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Source: auto.dta

* p<0.05, ** p<0.01, *** p<0.001

```
. eststo clear
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Labels, titles, and notes

- * About factor variables and interactions

```
. sysuse auto, clear
(1978 Automobile Data)
. qui eststo: reg price mpg i.foreign
. qui eststo: reg price c.mpg##i.foreign
. esttab, varwidth(25) label nobaselevels interaction(" X ")
```

	(1)	(2)
	Price	Price
Mileage (mpg)	-294.2*** (-5.28)	-329.3*** (-4.39)
Foreign	1767.3* (2.52)	-13.59 (-0.01)
Foreign X Mileage (mpg)		78.89 (0.70)
Constant	11905.4*** (10.28)	12600.5*** (8.25)

Observations	74	74
--------------	----	----

```
t statistics in parentheses
* p<0.05, ** p<0.01, *** p<0.001
```

```
. eststo clear
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Compressed table

```
. sysuse auto, clear
(1978 Automobile Data)
. qui eststo: reg price weight
. qui eststo: reg price weight mpg
. qui eststo: reg price weight mpg foreign
. qui eststo: reg price weight mpg foreign displacement
. esttab, compress
```

	(1)	(2)	(3)	(4)
	price	price	price	price
weight	2.044*** (5.42)	1.747** (2.72)	3.465*** (5.49)	2.458** (2.82)
mpg		-49.51 (-0.57)	21.85 (0.29)	19.08 (0.26)
foreign			3673.1*** (5.37)	3930.2*** (5.67)
displace_t				10.22 (1.65)
_cons	-6.707 (-0.01)	1946.1 (0.54)	-5853.7 (-1.73)	-4846.8 (-1.43)
N	74	74	74	74

t statistics in parentheses

* p<0.05, ** p<0.01, *** p<0.001

* reduces horizontal spacing to fit more models on screen without line breaking

```
. eststo clear
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Significance stars

```
. sysuse auto, clear
(1978 Automobile Data)
. qui eststo: reg price weight mpg
. qui eststo: reg price weight mpg foreign
. esttab, star(+ 0.10 * 0.05)
```

	(1)	(2)
	price	price
weight	1.747* (2.72)	3.465* (5.49)
mpg	-49.51 (-0.57)	21.85 (0.29)
foreign		3673.1* (5.37)
_cons	1946.1 (0.54)	-5853.7+ (-1.73)
N	74	74

t statistics in parentheses

+ p<0.10, * p<0.05

* default symbols and thresholds are: * for p<.05, ** for p<.01, and *** for p<.001.

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With Excel

```
esttab using example.csv
esttab using example.csv,replace wide plain
```

- ▶ With Word

```
esttab using example.rtf
esttab using example.rtf, append wide label modelwidth(8)
```

* varwidth() and modelwidth() change the column widths

```
lab var mpg "The mpg variable has a really long label"
esttab using example.rtf, replace label nogap onecell
```

* onecell : placed beneath one another in the same table cell

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

```
. esttab using example.tex, label nostar replace page
```

```
* page[(packages)] adds opening and closing code to define a whole LaTeX document
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

	(1)	(2)
	Price	Price
Weight (lbs.)	1.747 (2.72)	3.465 (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1 (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

```
. esttab using example.tex, label nostar replace page booktabs
```

```
* produces a LaTeX formatted table for use with LaTeX's booktabs package
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

	(1)	(2)
	Price	Price
Weight (lbs.)	1.747 (2.72)	3.465 (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1 (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

```
. esttab using example.tex, label nostar replace page booktabs ///  
    width(0.8\hsize)
```

```
* width(\hsize) in LaTeX or width(100%) in HTML to span the whole page
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

	(1)	(2)
	Price	Price
Weight (lbs.)	1.747 (2.72)	3.465 (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1 (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

```
. esttab using example.tex, label nostar replace page booktabs ///  
    width(0.8\hsize) alignment(11)
```

```
* specify the alignment of the models' columns in LaTeX
```


Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

	(1)	(2)
	Price	Price
Weight (lbs.)	1.747 (2.72)	3.465 (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1 (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

```
. esttab using example.tex, label nostar replace page booktabs ///  
    width(0.8\hsize) alignment(l1) title(Regression table)
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX

表 2: Regression table

	(1)	(2)
	Price	Price
Weight (lbs.)	1.747 (2.72)	3.465 (5.49)
Mileage (mpg)	-49.51 (-0.57)	21.85 (0.29)
Car type		3673.1 (5.37)
Constant	1946.1 (0.54)	-5853.7 (-1.73)
Observations	74	74

t statistics in parentheses

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX—分组样本回归 mgroup() 选项

```
sysuse auto, clear
eststo clear
eststo: qui reg weight mpg
eststo: qui reg weight mpg foreign
eststo: qui reg price weight mpg
eststo: qui reg price weight mpg foreign
esttab using mgroups.tex, replace          ///
      star(* 0.1 ** 0.05 *** 0.01)        ///
      compress nogaps                      ///
      title(An Illustration of mgroup() in esttab) ///
      mgroups("Group A" "Group B",        ///
      pattern(1 0 1 0) span              ///
      prefix(\multicolumn{@span}{c}{}) suffix{}) ///
      erepeat(\cmidrule(lr){@span}) )    ///
      booktabs \page{dcolumn} alignment(D{.}{.}{-1})
```

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX—分组样本回归 `mgroup()` 选项

- * "Group A" "Group B" : 组别名称
- * `mgroup()` 子选项 :
- * `pattern(1 0 1 0)` : 1--该组别的第一个模型, 0--该组别的其他模型
- * `span` : 定组别名能在表格中跨列
- * `prefix(\multicolumn{@span}{c}{})` 和 `suffix()` : 组别名在LaTeX代码中跨行
- * `erepeat(\cmidrule(lr){@span})` : 设定跨行代码, 下面加底部表格线
- * `page(dcolumn)` : 添加加载宏包 `dcolumn`
- * `alignment(D{.}{.}{-1})` : 调整单元格对齐方式, 小数点对齐

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ With LaTeX—分组样本回归 `mgroup()` 选项

Table 1: An Illustration of `mgroup()` in `esttab`

	Group A		Group B	
	(1)	(2)	(3)	(4)
	weight	weight	price	price
mpg	-108.4*** (-11.60)	-91.22*** (-10.34)	-49.51 (-0.57)	21.85 (0.29)
foreign		-550.1*** (-4.96)		3673.1*** (5.37)
weight			1.747*** (2.72)	3.465*** (5.49)
_cons	5328.8*** (25.85)	5125.7*** (27.93)	1946.1 (0.54)	-5853.7* (-1.73)
<i>N</i>	74	74	74	74

t statistics in parentheses

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$

- -esttab-

- ▶ Non-standard contents

Sometimes it is necessary to include parameter statistics in a table

`main()` option : replacing the point-estimates

`aux()` option : replacing the t-statistics

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Non-standard contents

For example, to include VIF(variance inflation factors) instead of t-statistics after reg

```
. sysuse auto, clear
. eststo clear
. reg price weight mpg foreign
. estadd vif
```

Variable	VIF	1/VIF
weight	3.86	0.258809
mpg	2.96	0.337297
foreign	1.59	0.627761
Mean VIF	2.81	

added matrix:

e(vif) : 1 x 4

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Non-standard contents

```
. esttab, aux(vif 2) wide nopar
```

	(1)	
	price	
weight	3.465***	3.86
mpg	21.85	2.96
foreign	3673.1***	1.59
_cons	-5853.7	

N	74	
---	----	--

```
vif in second column
```

```
* p<0.05, ** p<0.01, *** p<0.001
```

- -esttab-

- ▶ Non-standard contents

- More than two kinds of parameter statistics

- Switch to **estout syntax** and make use of the `cells()` option

- `cells()` disables `b()`, `beta()`, `main()`, `t()`, `abs`, `se()`, `p()`, `ci()`...

Making Regression Tables in Stata : estout package

- -esttab-

- ▶ Non-standard contents

For example, to print point estimates, t-statistics, and variance inflation factors in one table

```
. esttab, cells("b(fmt(a3) star) vif(fmt(2))" t(par fmt(2)))
```

	(1)	
	price	
	b/t	vif
weight	3.465*** (5.49)	3.86
mpg	21.85 (0.29)	2.96
foreign	3673.1*** (5.37)	1.59
_cons	-5853.7 (-1.73)	

N	74	
---	----	--

- -esttab-

- ▶ Non-standard contents

- complicated summary statistics section in the table footer

- `r2`, `ar2`, `pr2`, `aic`, `bic`, `scalars()`...

- estout's `stats()` option, equivalently

- -esttab-
 - ▶ Non-standard contents

```
return list  
dis r(cmdline)
```

Subsection 3

-estout-

Making Regression Tables in Stata : estout package

- -estout-

- ▶ The full syntax of estout is rather complex
- ▶ The most important options:

```
estout [ namelist ] [ using filename ] [, cells(array)  
                                     stats(scalarlist) style(style) option
```

- ▶ cells() and stats() options : determine the primary contents of the table
- ▶ style() option : determines the basic formatting of the table.

Making Regression Tables in Stata : estout package

- -estout-

- ▶ Choosing an output format

```
. sysuse auto, clear
. eststo: qui reg price weight mpg
. eststo: qui reg price weight mpg foreign
. estout, style(tex)
```

	&	est1&	est2&	est3&	est4\\
	&	b&	b&	b&	b\\
weight	&	1.746559&	3.464706&	1.746559&	3.464706\\
mpg	&	-49.51222&	21.8536&	-49.51222&	21.8536\\
foreign	&	&	3673.06&	&	3673.06\\
_cons	&	1946.069&	-5853.696&	1946.069&	-5853.696\\

Making Regression Tables in Stata : estout package

- -estout-

- ▶ The cells option
- ▶ The stats option

```
. eststo clear
. estout, stats(r2 bic N)
```

	.
	b
weight	3.464706
mpg	21.8536
foreign	3673.06
_cons	-5853.696

r2	.4995594
bic	1357.414
N	74

Making Regression Tables in Stata : estout package

- -estout-

- ▶ Using labels

```
. sysuse auto, clear
. eststo clear
. eststo, title("Model 1"): qui reg price weight mpg
. eststo, title("Model 2"): qui reg price weight mpg foreign
. label variable foreign "Car type (1=foreign)"
. estout, cells("b(star label(Coef.)) se(label(Std. err.))") ///
      stats(r2 N, labels(R-squared "N. of cases")) ///
      label legend varlabels(_cons Constant)
```

	Model 1		Model 2	
	Coef.	Std. err.	Coef.	Std. err.
Weight (lbs.)	1.746559**	.6413538	3.464706***	.630749
Mileage (mpg)	-49.51222	86.15604	21.8536	74.22114
Car type (1=foreign)			3673.06***	683.9783
Constant	1946.069	3597.05	-5853.696	3376.987

R-squared	.2933891		.4995594	
N. of cases	74		74	

* p<0.05, ** p<0.01, *** p<0.001

Subsection 4

-eststo-

Making Regression Tables in Stata : estout package

- -eststo-

- ▶ Stores a copy of the active estimation results for later tabulation.
- ▶ Analogous to official Stata's **estimates store**.
- ▶ Does not require the user to specify a name for the stored estimates.

```
sysuse auto, clear
reg price weight mpg
eststo
reg price weight mpg foreign
eststo
esttab
```

- ▶ As a prefix

```
sysuse auto, clear
eststo: qui reg price weight mpg
eststo: qui reg price weight mpg foreign
esttab
```

Making Regression Tables in Stata : estout package

- -eststo-

- ▶ Using by

```
. sysuse auto, clear  
. eststo clear  
. by foreign : eststo: qui reg price weight mpg
```

```
-> Domestic  
(est1 stored)
```

```
-> Foreign  
(est2 stored)  
. esttab, label nodelvar nonumber
```

	Domestic	Foreign
Weight (lbs.)	4.415*** (4.66)	5.156*** (5.85)
Mileage (mpg)	237.7 (1.71)	-19.78 (-0.34)
Constant	-13285.4* (-2.32)	-5065.8 (-1.58)

Observations	52	22
--------------	----	----

```
t statistics in parentheses  
* p<0.05, ** p<0.01, *** p<0.001
```

Making Regression Tables in Stata : estout package

- -eststo-

- ▶ Adding additional statistics

```
. sysuse auto, clear
. eststo clear
. qui reg price weight mpg
. test weight = mpg
( 1) weight - mpg = 0
    F( 1, 71) = 0.36
    Prob > F = 0.5514
. eststo, addscalars(p_diff r(p))
(e(p_diff) = .55138216 added)
(est1 stored)
. esttab, scalars(p_diff) obslast
```

	(1)
	price
weight	1.747** (2.72)
mpg	-49.51 (-0.57)
_cons	1946.1 (0.54)
p_diff	0.551
N	74

t statistics in parentheses
* p<0.05, ** p<0.01, *** p<0.001

Subsection 5

-estadd-

- -estadd-

- ▶ Add results to (stored) estimates
- ▶ Results that are included in the e()-returns for the models can be tabulated by estout or esttab
- ▶ 举例：向内存中添加两个统计量：

一个是文字类型的返回值 `Industry`，采用暂元 (`local`) 存储

一个是数值类型的返回值 `Mean_Wage`，采用单值 (`scalar`) 来存储

Making Regression Tables in Stata : estout package

- -estadd-

```
. eststo clear
. sysuse nlsw88.dta, clear
(NLSW, 1988 extract)
. qui reg wage ttl_exp married
. estadd local Industry "Yes"
added macro:
      e(Industry) : "Yes"
```

```
. qui sum wage
. estadd scalar Mean_Wage = r(mean)
added scalar:
      e(Mean_Wage) = 7.766949

. qui ereturn list
```

Making Regression Tables in Stata : estout package

- -estadd-

```
^^Isysuse nlsw88.dta, clear

^^Iglobal xx "ttl_exp married south hours tenure age i.industry"
^^I
^^Iqui reg wage $xx if race==1
^^Iestadd local Industry "Yes"
^^Iestadd local Occupation "No"
^^Iest store m1
^^I
^^Iqui reg wage $xx if race==2
^^Iestadd local Industry "Yes"
^^Iestadd local Occupation "No"
^^Iest store m2
^^I
^^Iqui reg wage $xx i.occupation if race==1
^^Iestadd local Industry "Yes"
^^Iestadd local Occupation "Yes"
^^Iest store m3
^^I
^^Iqui reg wage $xx i.occupation if race==2
^^Iestadd local Industry "Yes"
^^Iestadd local Occupation "Yes"
^^Iest store m4
```

Making Regression Tables in Stata : estout package

• -estadd-

```
local m "m1 m2 m3 m4"  
esttab `m', mtitle(White Black White Black) b(%6.3f) nogap compress ///  
star(* 0.1 ** 0.05 *** 0.01) ///  
drop(*.industry *.occupation) ///  
ar2 scalar(N Industry Occupation)
```

	(1) White	(2) Black	(3) White	(4) Black
t1l_exp	0.251*** (6.49)	0.271*** (4.82)	0.176*** (4.61)	0.193*** (3.61)
married	-0.737** (-2.31)	0.082 (0.21)	-0.657** (-2.12)	0.099 (0.27)
south	-0.814*** (-2.72)	-2.038*** (-4.92)	-0.758*** (-2.61)	-1.791*** (-4.56)
hours	0.051*** (3.81)	0.036 (1.35)	0.021 (1.56)	0.007 (0.25)
tenure	0.025 (0.76)	-0.003 (-0.08)	0.039 (1.25)	-0.014 (-0.34)
age	-0.073 (-1.58)	-0.216*** (-3.12)	-0.058 (-1.31)	-0.148** (-2.23)
_cons	5.576** (2.22)	9.490** (2.53)	9.607*** (3.82)	8.420** (2.39)
N	1615	572	1612	571
adj. R-sq	0.112	0.166	0.176	0.281
Industry	Yes	Yes	Yes	Yes
Occupation	No	No	Yes	Yes

t statistics in parentheses
* p<0.1, ** p<0.05, *** p<0.01

Subsection 6

-estpost-

Making Regression Tables in Stata : estout package

- -estpost-

- ▶ esttab and estout tabulate the e()-returns of a command, but not all commands return their results in e()
- ▶ Posting results from **non-eclass** commands
- ▶ Review

- 结果导出-esttab-

```
sysuse auto,clear

local var price wei len mpg
qui estpost ttest `var', by(foreign)
esttab using ttable2.rtf, cell("mu_1(fmt(2)) mu_2(fmt(2)) b(star fmt(2)) t(fmt(2))" ) ///
      starlevels(* 0.10 ** 0.05 *** 0.01) replace noobs compress ///
      title(esttab_Table: T_test)
```

- 描述性统计表格导出

- ▶ -esttab-

```
estpost summarize price wei len mpg rep78, detail
esttab using Desc4.rtf, ///
      cells("count mean(fmt(2)) sd(fmt(2)) min(fmt(2)) p50(fmt(2)) max(fmt(2))" ) ///
      noobs compress replace title(esttab_Table: Descriptive statistics)
```

Making Regression Tables in Stata : estout package

- -estpost-

- ▶ Review

- 相关系数矩阵导出

```
*<方法一> -logout-
logout, save(Corr1) word replace: pcorr price wei len mpg rep78, star(.05)

*<方法二> -esttab-
estpost correlate price wei len mpg rep78, matrix
esttab using Corr2.rtf,                                     ///
    unstack not noobs compress nogaps replace star(* 0.1 ** 0.05 *** 0.01) ///
    b(%8.3f) p(%8.3f) title(esttab_Table: correlation coefficient matrix)

*<方法三> -corr2docx-
corr2docx price wei len mpg rep78 using Corr3.docx,       ///
    replace spearman(ignore) pearson(pw) star             ///
    title(corr2docx_Table: correlation coefficient matrix)
```