

# 社会统计学及SPSS软件应用

## STATISTICS WITH SPSS

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(1) 嵌套模型

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## 归纳与总结

- 统计学包括两部分内容：描述统计和推论统计。
  - 描述是对样本的描述。
  - 推论是对总体的推论。

# 归纳与总结

测量水平	单个变量	分类变量+连续变量	分类变量+分类变量	多分类变量+连续变量
针对样本	$\geq <$	两个平均数差异的比较	两个分类变量的独立性	多个平均数差异的比较
针对总体	标准误、置信区间	双样本t检验	$\chi^2$ 检验	ANOVA: F检验

# A TYPOLOGY OF STATISTICAL TEST

<b>Statistical Test</b>	<b>Dependent Variable</b>	<b>Independent Variable</b>
One sample t-test	interval or ratio	
Independent samples t-test	interval or ratio	nominal with only two categories
Paired samples t-tests	interval or ratio	nominal with only two categories
ANOVA	interval or ratio	nominal with more than two categories
$\chi^2$	nominal or ordinal	nominal or ordinal

# ONE-SAMPLE T TEST

## Example 1.1

假设有人提出1988年全国城市居民平均收入为3800元，而在1988年的CHIP数据中，我们发现居民的年收入均值为3687元，标准差为3489元，那么在0.05的显著性水平下，这一样本结果与3800元的提法一致吗？

- 首先，建立 $H_0 : \mu = 3800$  和  $H_1 : \mu \neq 3800$
- 根据样本数据，计算t检验统计量
- stata 命令：ttest earn =3800
- 结论：否定 $H_0$ ，即认为全国城市居民平均收入不是3800元。

# ONE-SAMPLE T TEST

```
. ttest earn =3800
```

One-sample t test

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
earn	20421	3687.005	24.4163	3489.139	3639.147	3734.862

mean = mean(earn)

t = -4.6279

Ho: mean = 3800

degrees of freedom = 20420

Ha: mean < 3800

Ha: mean != 3800

Ha: mean > 3800

Pr(T < t) = 0.0000

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

Pr(T > t) = 1.0000

Figure 1.1:

# TWO-SAMPLE T TEST

## Example 1.2

在1988年的CHIP数据中，有女性9700人，其年收入均值为3329元，标准差为3197元；有男性10721人，其年收入均值为4011元，标准差为3704元。那么在0.05的显著性水平下，是否存在收入的性别差异？

- 首先，建立 $H_0: \mu_1 = \mu_2$  和  $H_1: \mu_1 \neq \mu_2$
- 检验方差是否相等
- 根据样本数据，计算t检验统计量
- stata 命令：ttest earn, by(sex) unequal
- 结论：否定 $H_0$ ，即认为男性和女性的年收入均值是不相等的。



# TWO-SAMPLE T TEST

```
. ttest earn, by(sex) unequal
```

Two-sample t test with unequal variances

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
male	10721	4010.848	35.77724	3704.456	3940.718	4080.978
female	9700	3329.074	32.45828	3196.77	3265.449	3392.699
combined	20421	3687.005	24.4163	3489.139	3639.147	3734.862
diff		681.7748	48.30684		587.0895	776.4601

```
diff = mean(male) - mean(female)          t = 14.1134
```

Ho: diff = 0 Satterthwaite's degrees of freedom = 20373.8

$H_a: \text{diff} < 0$	$H_a: \text{diff} \neq 0$	$H_a: \text{diff} > 0$
------------------------	---------------------------	------------------------

$$\Pr(T < t) = 1.0000$$
$$\Pr(|T| > |t|) = 0.0000$$
$$Pr(I > t) = 0.0000$$

Figure 1.2:

- 1 A variety of social science data come in the form of cross-classified tables of counts, commonly referred to as contingency tables.
- 2 With the log-linear approach, we model cell counts in a contingency table in terms of associations among the variables and marginal frequencies.

# CHI-SQUARE TEST(OBSERVED)

		性别		总数
		男	女	
是否经理	否	290	100	390
	是	80	4	84
总数		370	104	474

## CHI-SQUARE TEST(EXPECTED)

		性别		总数
		男	女	
是否经理	否	304.4	85.6	390
	是	65.6	18.4	84
总数		370	104	474

单元格的预期值，等于该单元格所在行的边缘值（marginals）乘以它所在列的边缘值，除以大总数（grand total）

$$\chi^2 = \sum \frac{(\text{observed value} - \text{expected value})^2}{\text{expected value}}$$

Association between child abuse and substance abuse by parent of abusers (n=49)

Substance Abuse by Parent of Abusers	Child Abuse	
	Yes	No
Yes	18	19
No	2	10

$\chi^2 = 3.84(1), p < 0.05$ , read as "Chi-square with one degree of freedom equals 3.84".

- 1  $H_0$  两个变量无关。
- 2  $H_1$  两个变量有关。

# ONE-WAY ANOVA

## Example 1.3

在2003年的CGSS数据中，对收入（取对数）进行方差分析，检验不同教育程度的收入水平是否存在差异。

- 首先，建立 $H_0: \mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5$  和  $H_1: \mu_1 \neq \mu_2 \neq \mu_3 \neq \mu_4 \neq \mu_5$
- 根据样本数据，计算F检验统计量
- stata 命令：anovallinc edu
- 结论：否定 $H_0$ ，即认为不同教育程度的收入水平存在差异。

# ONE-WAY ANOVA

```
. tab edu,sum(lninc)
```

RECODE of educ (教育程度)	Summary of lninc		Freq.
	Mean	Std. Dev.	
0	5.8131884	.77492148	9
1	6.2615059	.7317007	52
2	6.3786467	.83584724	138
3	6.594358	.6006642	167
4	7.0642606	.64383691	92
Total	6.5706113	.75885336	458

Figure 1.3:

# ONE-WAY ANOVA

```
anova lninc edu
```

```
Number of obs =      458      R-squared      =  0.1434
Root MSE      =  .705445      Adj R-squared =  0.1358
```

Source	Partial SS	df	MS	F	Prob > F
Model	37.730583	4	9.43264576	18.95	0.0000
edu	37.730583	4	9.43264576	18.95	0.0000
Residual	225.436718	453	.497652798		
Total	263.167301	457	.575858426		

Figure 1.4:



## TWO-WAY OR MULTI-WAY ANOVA

### Example 1.4

在2003年的CGSS数据中，对收入（取对数）进行方差分析，检验不同教育程度、不同性别、不同党员身份的收入水平是否存在差异。

- 根据样本数据，计算F检验统计量
- stata 命令：anovallnc edu sex party

# MULTI-WAY ANOVA

```
. anova lninc edu sex party
```

```
Number of obs =      452      R-squared      =  0.1762
Root MSE      =  .694816      Adj R-squared =  0.1651
```

Source	Partial SS	df	MS	F	Prob > F
Model	45.9401838	6	7.65669731	15.86	0.0000
edu	32.4662959	4	8.11657397	16.81	0.0000
sex	3.87457706	1	3.87457706	8.03	0.0048
party	2.95558709	1	2.95558709	6.12	0.0137
Residual	214.832278	445	.482769164		
Total	260.772462	451	.578209449		

Figure 1.5:

# COMPARISONS OF CONDITIONAL MEANS

- 1 Categorical independent variables can be easily handled in a regression model. We can create dummy variables.
- 2 Analysis of variance (ANOVA) is statistically equivalent to regression analysis with dummy variable.

## 嵌套模型

```
/* 回归分析 */  
xi: reg lninc age age2 i.sex  
outreg2 using example1, replace  
xi: reg lninc age age2 i.sex edu  
outreg2 using example1  
xi: reg lninc age age2 i.sex*edu  
outreg2 using example1  
xi: reg lninc age age2 i.sex*edu i.party  
outreg2 using example1  
xi: reg lninc age age2 i.sex*edu i.party fedu i.fparty  
outreg2 using example1  
xi: reg lninc age age2 i.sex*edu i.party fedu i.fparty*i.sex  
outreg2 using example1  
xi: reg lninc age age2 i.sex*edu i.party fedu i.fparty*i.sex if age<=60  
outreg2 using example1  
|
```

Figure 1.6:

## 嵌套模型

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
VARIABLES	1ninc	1ninc	1ninc	1ninc	1ninc	1ninc	1ninc
age	-0.0146 (0.0184)	-0.00831 (0.0170)	-0.00610 (0.0170)	-0.00557 (0.0170)	0.00383 (0.0172)	0.00235 (0.0171)	-0.00462 (0.0264)
age2	0.000146 (0.000201)	0.000149 (0.000186)	0.000124 (0.000185)	0.000109 (0.000185)	2.19e-05 (0.000188)	3.70e-05 (0.000187)	0.000137 (0.000324)
o._Isex_2						-	-
edu		0.301*** (0.0343)	0.228*** (0.0449)	0.207*** (0.0460)	0.188*** (0.0462)	0.201*** (0.0463)	0.236*** (0.0558)
_IsexXedu_2			0.166** (0.0660)	0.176*** (0.0662)	0.167** (0.0662)	0.135** (0.0674)	0.135** (0.0801)
_Iparty_2				-0.172** (0.0857)	-0.192** (0.0851)	-0.210** (0.0850)	-0.215** (0.0969)
fedu					0.0690** (0.0310)	0.0740** (0.0309)	0.0750** (0.0349)
_Ifparty_2					0.101 (0.0767)	0.277** (0.109)	0.295** (0.116)
_Isex_2	-0.201*** (0.0707)	-0.223*** (0.0655)	-0.657*** (0.185)	-0.672*** (0.186)	-0.669*** (0.187)	-0.348 (0.234)	-0.337 (0.269)
_IfpaXsex_2_2						-0.330** (0.145)	-0.333** (0.157)
Constant	7.005*** (0.400)	5.944*** (0.389)	6.086*** (0.391)	6.276*** (0.408)	5.949*** (0.426)	5.825*** (0.427)	5.827*** (0.563)

- 1 Although in principle we could design our models to best predict any population parameter(e.g. the median), in practice we commonly use the regression to denote the problem of predicting conditional means.
- 2 When the regression function is a linear combination of independent variables, we have so-called linear regression, which are widely used for continuous dependent variables.

# A TYPOLOGY OF REGRESSION MODELS

Dependent Variable	Independent	Method of Analysis
Continuous	Continuous	Regression, correlation
Continuous	Categorical	Regression, ANOVA
Binary	Categorical	Logit/probit, loglinear
Binary	Continuous	Logit/probit
Unordered polytomous	Categorical	Multinomial logit, loglinear
Unordered polytomous	Continuous	Multinomial logit
Ordered polytomous	Categorical	Order logit/probit, loglinear
Ordered polytomous	Continuous	Order logit/probit
Cross-classified data	Categorical	Loglinear
Censored duration data	Categorical, continuous	Loglinear, logit/comp. log-log

回归分析的目的在于利用变量间的简单函数关系，用自变量对因变量进行“预测”，让“预测值”尽可能地接近因变量的“观测值”。

In regression analysis, the objective is to predict, as closely as possible, an array of observed values of the dependent variable based on a simple function of independent variables.

Obviously, predicted values from regression models are not exactly the same as observed ones. Characteristically, regression partitions an observation into two parts:



回归分析的特点就在于它把观测值分解成两部分：结构部分和扰动部分。

$$\begin{aligned}\text{观测项} &= \text{结构项} + \text{随机项} \\ \text{observed} &= \text{structural} + \text{stochastic}\end{aligned}$$

- 1 观测部分代表因变量的实际取值；
- 2 结构项部分代表因变量和自变量之间的结构关系，表现为“预测值”；

The structural part denotes the relationship between the dependent and independent variable.

- 3 随机项部分表示观测项中未被结构项解释的剩余部分。  
The stochastic part is the random component unexplained by the structural part.

描述性关系：观测项=概括项+残差项

Description: Observed=summary+residual

- The model serves to summarize the basic feature of data. The question is whether the model corresponds to the facts.
- The primary goal of statistical modeling is to summarize massive amounts of data with simple structures and few parameters.

统计模型的主要目标在于用最简单的结构和尽可能少的参数来概括大量数据所包含的主要信息。

- For the  $i$ th observation, we have

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_k x_{ik} + \epsilon_i$$

$$= \sum \beta_k x_{ik} + \epsilon_i$$

$$= x_i' \beta + \epsilon_i$$

- $y$  is decomposed into a linear function of  $x$ 's with unknown parameters ( $\beta$ ) and a residual term ( $\epsilon_i$ ).

- 1 Since  $\epsilon_i$  is intrinsically unobservable, simplifying assumptions about the characteristics of  $\epsilon_i$  are necessary.
- 2 A key assumption that yields the identification of the unknown parameters is the independence between  $\epsilon$  and  $x$  variables.
- 3 Other assumptions are often invoked to improve efficiency. For example, it is common to assume  $\epsilon_i$  to be independent of one another and identically distributed (i.i.d).
  - (1) The independence assumption implies that the correlation in  $\epsilon$  between a pair of observations is zero.
  - (2) The identical distribution assumption assures a common variance of  $\sigma_\epsilon^2$  (i.e. homoscedasticity).

# ESTIMATION

## 1 Least squares estimation

- (1) The goal of LS estimation is to find estimates of  $\beta$ 's that make the sum of squared errors around the conditional mean as small as possible.

## 2 Maximum likelihood estimation

- (1) The principal aim of ML estimation is to find parameter values that maximize the sampling likelihood,  $L$ , which may be thought of as the formula for the joint probability distribution (or joint density) of the sample.

## 参数点估计的三个评价标准

- 1 unbiased: the expected value of an estimator equals the value of the true parameter being estimated  
(1) 无偏性：估计值的平均数恰好等于总体的参数值。
- 2 efficiency: the estimator with the smallest variance in its sampling distribution  
(1) 有效性：抽样分布方差最小的那个估计就是对总体参数最有效率的估计。
- 3 consistent: 随着样本规模扩大，估计的偏差随之减小，或者说估计值以更大的概率趋近总体参数值。

# GAUSS-MARKOV THEOREM

- 1 Mean zero errors:  $E(\epsilon|x) = 0$
- 2 Homoscedasticity:  $var(\epsilon|x) = \sigma^2$
- 3 Noncorrelated errors:  $cov(\epsilon_i, \epsilon_j) = 0, i \neq j$
- 4 Exogeneity of explanatory variables:  $cov(x_i, \epsilon) = 0$

The Gauss-Markov theorem states that OLS estimator is the best (means most efficient) among all linear unbiased estimators (BLUE).

# TENSION BETWEEN ACCURACY AND PARSIMONY

- 1 By "accuracy" we mean the ability to reproduce the data, measured by goodness-of-fit statistics.
  - (1) saturated model
- 2 By "parsimony" we commonly mean statistical models with few parameters.
  - (1)  $y_i = \beta_0 + \epsilon_i$



- 1 The objective in model searching is to find models that describe the essential characteristics of the data using as few parameters as possible.
- 2 Nested models provide a way to motivate model searching.
- 3 Two models are nested if one model is a special case of the other.
  - (1) Model1:  $y_i = \beta_0 + \beta_1 x_{i1} + \epsilon_i$
  - (2) Model2:  $y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \epsilon_i$
- 4 Model1 is said to be nested within Model2, as Model1 is a special case of Model2 with the constraint that  $\beta_2 = 0$ .

- 1 The key question is whether the unconstrained model fits significantly better than the constrained model.
- 2 We need to assess the improvement in fit.
- 3 For linear regressions, such assessments are conducted using F-tests based on reductions in residual sums of squares, or other proportionate reduction in error (PRE) criteria.
  - (1) (削减误差比例)  $PRE = \frac{TSS - RSS}{TSS}$
- 4 For the nonlinear models, most assessments are chi-squared based on different criteria, such as reductions in the log-likelihood-ratio statistics.

# MLE

- 1 If we regard  $f(x|\theta)$  as a function of  $\theta$  for given observed data  $X$ , then  $L(x|\theta) = f(x|\theta)$  is called a likelihood function.
- 2 The ML principles states that an admissible  $\theta$  that maximizes likelihood function probability (discrete case) or density (continuous case), relative to alternative values of  $\theta$ , provides the  $\theta$  that is the most "likely" to have generated the observed data  $X$ , given the assumed parametric form.