

Econometrics III

(菅史彦. 九州大学 2019)

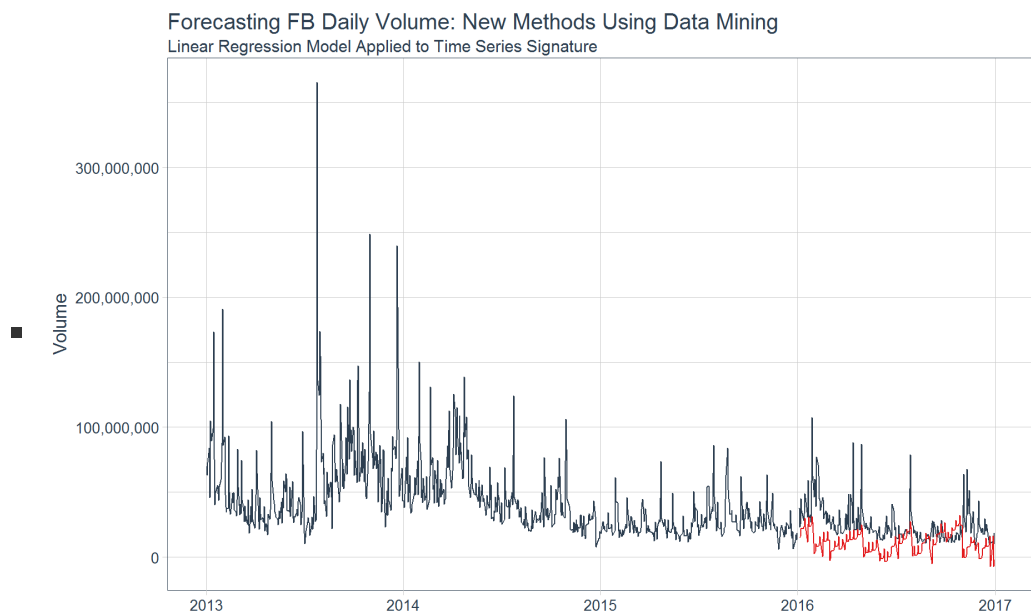
WEEK 1

1. Nonexperimental data

observational data, or social science data.

2. Data type

- cross-sectional data 横截面数据
 - sample of different individuals ,at a given point of time
 - often ignore minor time differences
 - observations are more or less independent
 - *applied microeconomics*
- time series data 时间序列数据
 - one variable(or several variables) over time



Data from Yahoo! Finance: 'FB' Daily Volume from 2013 to 2016.

- **serially correlated**. trends and seasonality.
- *applied macroeconomics and finance*
- pooled cross sections 混合截面数据
 - two or more cross sections combined
 - evaluate policy changes
- panel/longitudinal data 面板数据
 - same cross sections followed over time = *time series+cross sections*
 - can be used to account for time-invariant un-observables: eg. lags in decision making; city crime statistics 's un-observable city characteristics

3. r regression (SLR)

```

1 #install wooldridge package to use datasets
2 data("crime")
3 result <- lm(wage ~ educ+exper+training, data = crime)
4 summary(result)

```

4. definition of causal effect: *ceteris paribus*

how does y change if variable x is changed, but all other relevant factors are held constant?

5. topics

- OLS
- IV
- Fixed Effect & Random Effect Model
- MLE
- Probit/Logit
- Tobit
- Heckman Two Step

WEEK 2

1. SLR --- Simple Linear Regression

Focus: when is there a causal interpretation?

Assumptions:

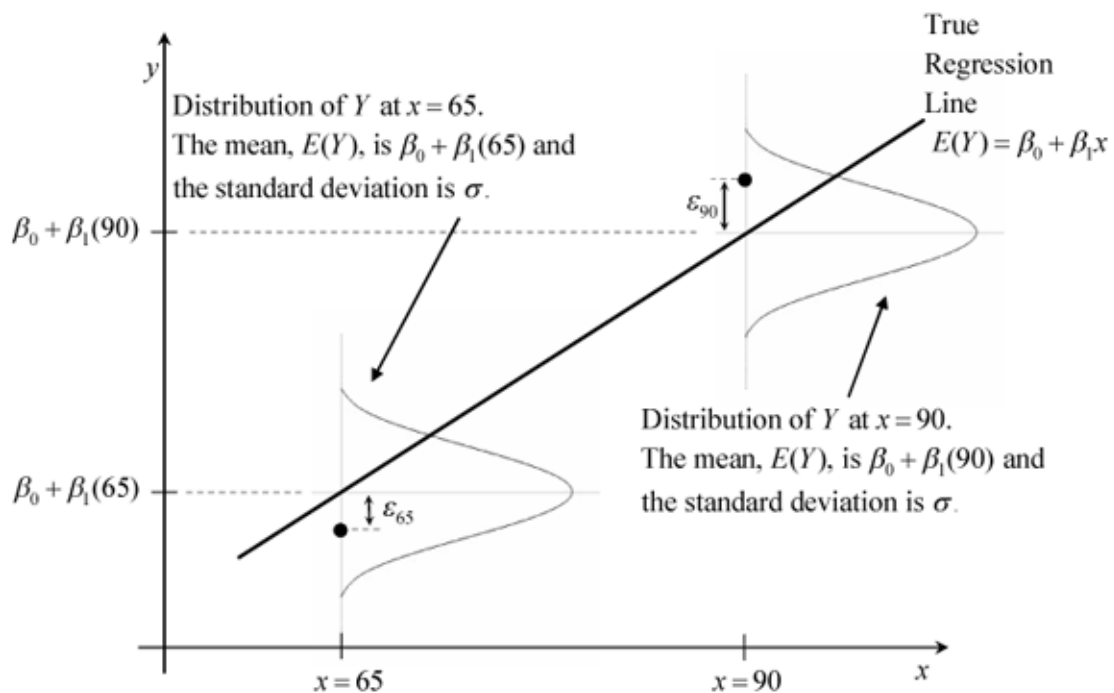
1. Conditional Expectation independence

$E(u|x) = 0$ (which is a function of x)

x does not contain any information about mean of u .

■ PRF Population Regression Function

$$\begin{aligned}
 E(y|x) &= E(\beta_0 + \beta_1 x + u|x) \\
 &= \beta_0 + \beta_1 x + E(u|x) \\
 &= \beta_0 + \beta_1 x
 \end{aligned}$$



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- Sample analogue
 - regression residuals

$$\hat{u} = y^i - \hat{y}_i$$

minimize sum of squared residuals:

$$\min \sum \hat{u}_i^2 \rightarrow \hat{\beta}_0, \hat{\beta}_1$$

- OLS estimates *Ordinary Least Squares Estimates*
by minimizing SSR, we estimates β_0 and β_1 .
- error terms
fitted regression line - **unknown population regression line.**
differ from residuals

2. R introduction

```

1 data("gpa1")
2 #subset dataset gpa1, we only use 3 columns
3 gpa1small <- gpa1[,c("colGPA", "hsGPA", "ACT")]
4 summary(gpa1small)
5 #variation and correlation index(pearson)
6 var(gpa1small)
7 cor(gpa1small)
8 #standard deviation
9 sd(gpa1small$ACT)
10 #histogram
11 hist(gpa1small$colGPA)
12 #scatter plots
13 plot(gpa1small$colGPA, gpa1small$hsGPA)
14 #rgl package
15 install.packages("rgl")
16 library("rgl")
17 plot3d(gpa1small$colGPA, gpa1small$hsGPA) # this function must
    contain 3 vars.
18 #linear regression model
19 result <- lm(gpa1small$colGPA~gpa1small$hsGPA+gpa1small$ACT)
20 result <- lm(colGPA~hsGPA+ACT, data=gpa1small)
21 summary(result)
22 #obtain bata1 by coefficients index
23 beta1<-result$coefficients[2]
```