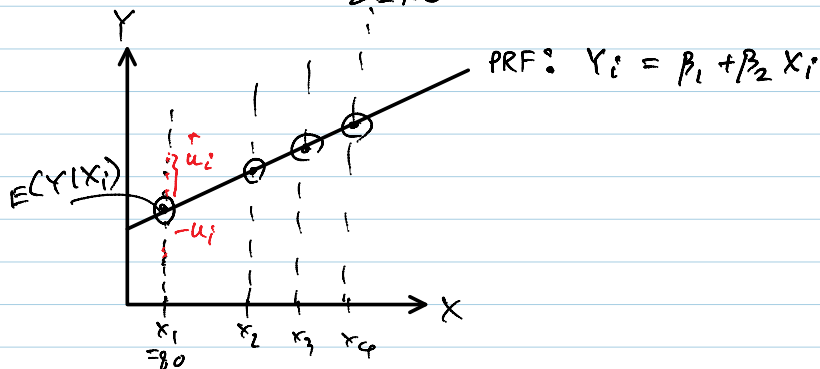


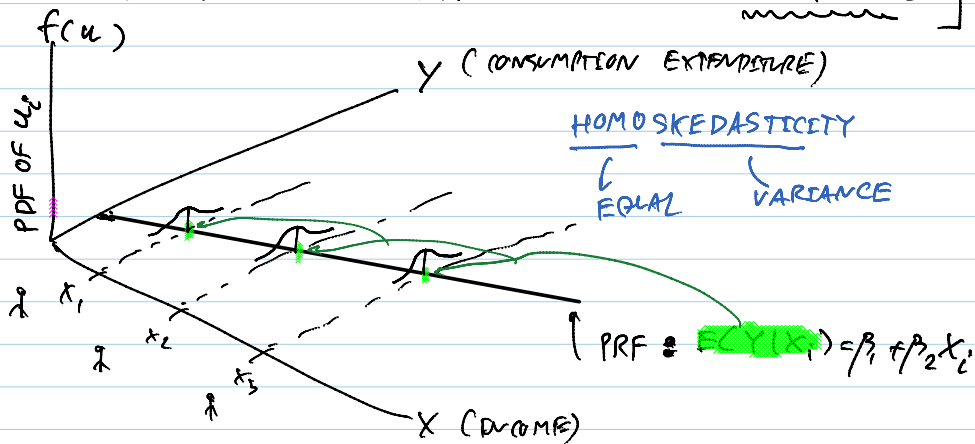
ASSUMPTIONS UNDERLYING THE OLS (10 ASSUMPTIONS)

- ① LINEAR MODEL : $Y_i = \beta_1 + \beta_2 X_i + u_i$.
- ② X IS **NONSTOCHASTIC**, i.e., X IS FIXED IN REPEATED SAMPLING.
- ③ $E(u_i) = 0$, i.e., MEAN OR AVERAGE VALUES OF DISTURBANCE TERM IS EQUAL TO ZERO

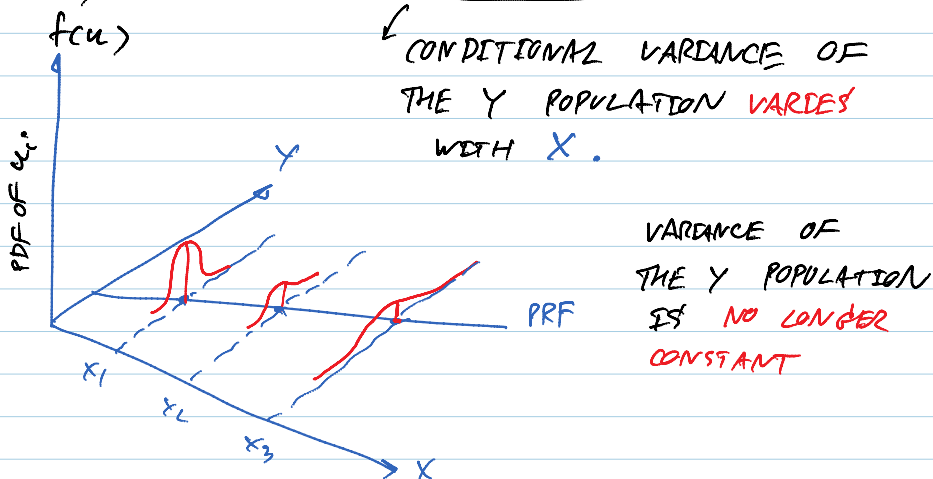


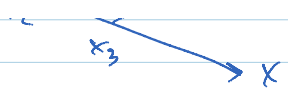
$$\textcircled{4} \text{VAR}(u_i) = E(u_i - E(u_i))^2 = E(u_i^2) = \sigma_u^2$$

[VARIANCE OF ERROR TERM IS EQUAL EVERYWHERE]



IN CONTRAST, LET'S SEE **HETEROSKEDASTICITY**





$$\textcircled{5} \text{ COV}(u_i, u_j) = E(u_i - E(u_i))(u_j - E(u_j)) \\ = E(u_i u_j) = 0$$

[IT SAYS THAT NO AUTOCORRELATION BETWEEN THE DISTURBANCE TERMS]

EX: CONSIDER $Y_t = \beta_1 + \beta_2 X_t + u_t \Rightarrow$ OUR PRF
 SUPPOSE THAT u_t IS POSITIVELY LINKED W/ u_{t-1} .

THEN, Y_t DEPENDS NOT ONLY ON X_t BUT ALSO u_{t-1} !
 BECAUSE u_t AND u_{t-1} ARE POSITIVELY CORRELATED.

- THIS CASE IS NOT ALLOWED TO HAPPEN IF ASSUMPTION 4 IS VALID.
- WE WILL SEE LATER (A/F MEDIUM) ON WHAT WOULD BE THE CONSEQUENCES IF u_t AND u_{t-1} ARE CORRELATED. (THIS PROBLEM CALLED "AUTO CORRELATION")

$$\textcircled{6} \text{ COV}(x_i, u_i) = E(x_i - E(x_i))(u_i - E(u_i)) \\ = E(x_i u_i) = 0$$

[IT SAYS THAT EXPLANATORY VARIABLE x_i AND DISTURBANCE TERMS ARE NOT CORRELATED]

INTUITION? $\Rightarrow Y_i = \beta_1 + \beta_2 x_i + u_i$ (PRF)
 $\beta_2 = 0.8$

AS WE WANT TO SEPARATE THE EFFECT OF x ON y (INDIVIDUAL EFFECT)

FROM THE EFFECT OF ANY OTHER VARIABLES ON y ,

WE NEED THIS ASSUMPTION OF $\text{COV}(x_i, u_i) = 0$.

$\textcircled{7}$ NUMBER OF OBSERVATIONS $>$ NUMBER OF PARAMETERS
 (OR $n > k$)
 \downarrow # OF EXPLANATORY VARIABLES.

EX: IF $n = k = 2 \Rightarrow$ CANNOT ESTIMATE THIS MODEL.

(8) X HAS VARIABILITY, i.e., $\text{var}(x_i) = \sigma_x^2 \neq 0$

(9) THE MODEL HAS A CORRECT SPECIFICATION.

MEANS THAT

- (1) CORRECT FUNCTIONAL FORM
- (2) NO OMITTED VARIABLES
- (3) NO REDUNDANT VARIABLES

CH. 13

A/F
MIDTERM

(10) NO PERFECT MULTICOLLINEARITY

IT MEANS THAT THERE ARE NO PERFECT "LINEAR" RELATIONSHIPS BETWEEN EXPLANATORY VARIABLES.

FOR EXAMPLE:

Y_i	X_{2i}	$X_{3i} (= 3X_{2i} + 2)$
⋮	1	5
⋮	3	11
⋮	5	17
⋮	7	23
⋮	9	29

↓
PERFECT
LINEAR
RELATIONSHIP

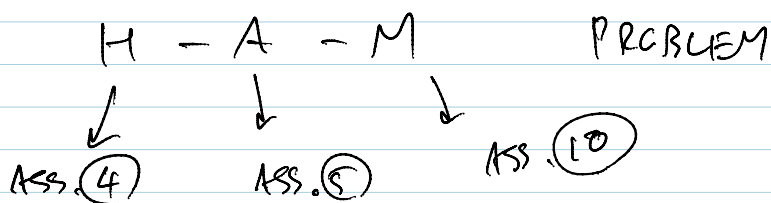
↓
IF THIS IS SO,

THE MODEL CANNOT
BE SOLVED!

(WE WILL SEE
WHEN WE DO STATA)

$$Y_i = \beta_1 + \beta_2 X_{2i} + \beta_3 X_{3i} + u_i$$

(MULTIPLE REGRESSION MODEL)

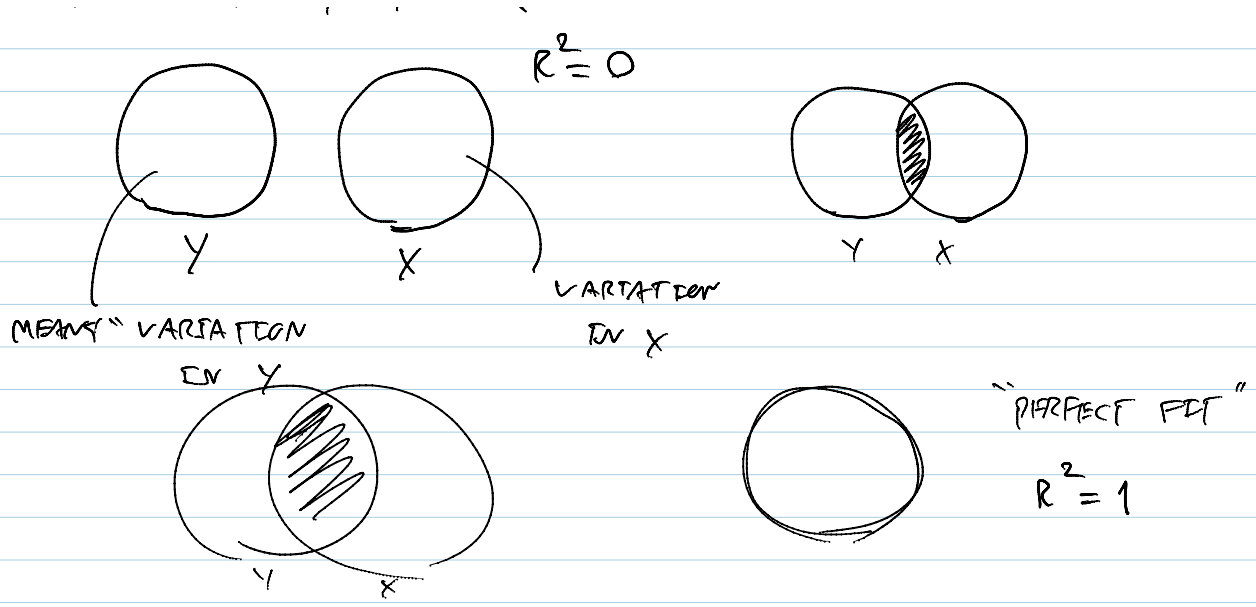


21.09.2012

THE COEFFICIENT OF DETERMINATION R^2 :
A MEASURE OF "GOODNESS OF FIT"

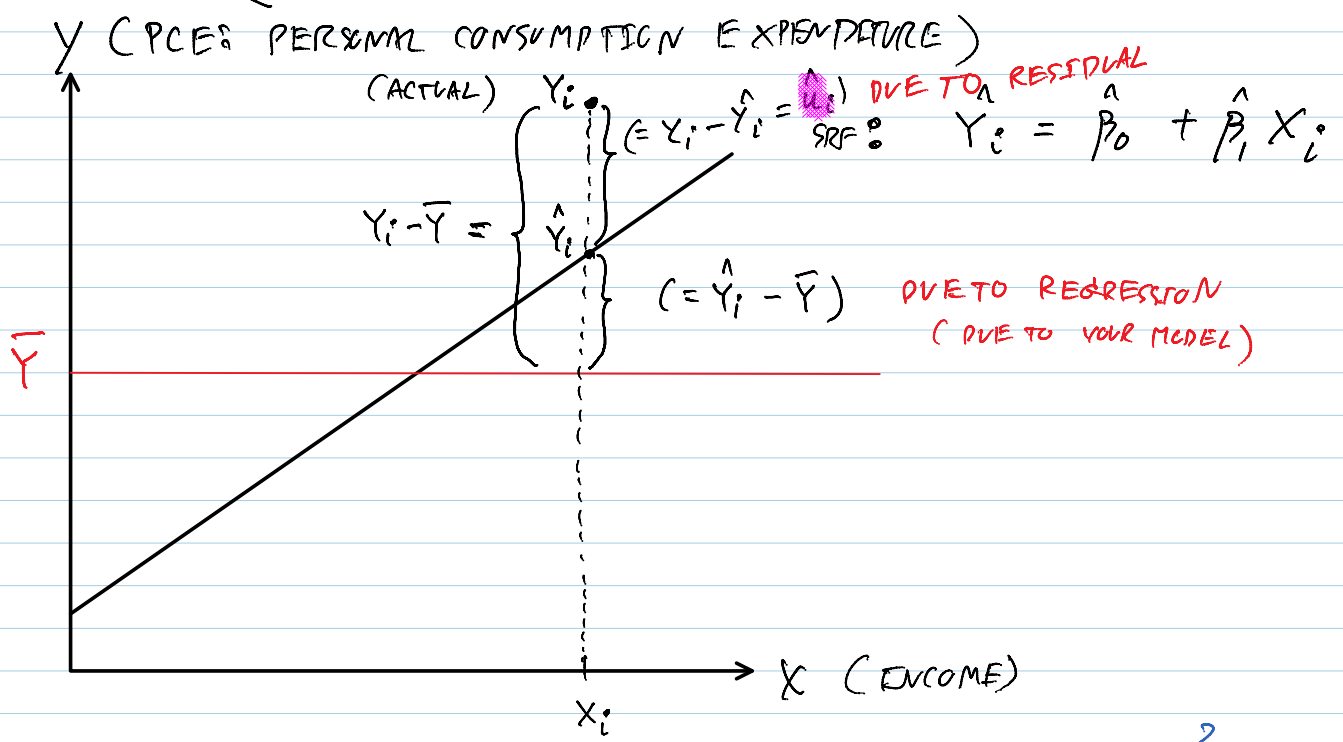
Q: HOW "WELL" THE SAMPLE REGRESSION LINE FITS THE DATA?

A: WE USE R^2 TO MEASURE THE DEGREE OF EXPLANATORY POWER



• THE HIGHER THE OVERLAPPING AREA, THE HIGHER PROPORTION THAT VARIATION IN Y IS EXPLAINED BY VARIATION IN X.

• $0 \leq R^2 \leq 1$



LET'S LOOK AT HOW TO COMPUTE OR TO DERIVE R^2

CONSIDER $Y_i = \hat{Y}_i + u_i$ — (1) (TRUE EVERYWHERE)

(ACTUAL) (ESTIMATED) (RESIDUAL)

OR IN DEVIATION FORM: $y_i = \hat{y}_i + u_i$ — (2)

TAKING SQUARE FOR BOTH SIDES AND SUMMING OVER ALL OBSERVATION

$$\begin{aligned}
 \sum_{i=1}^n y_i^2 &= \sum_{i=1}^n (\hat{y}_i + \hat{u}_i)^2 \\
 &= \sum_{i=1}^n [\hat{y}_i^2 + 2\hat{y}_i\hat{u}_i + \hat{u}_i^2] \\
 &= \sum_{i=1}^n \hat{y}_i^2 + 2 \sum_{i=1}^n \hat{y}_i\hat{u}_i + \sum_{i=1}^n \hat{u}_i^2 \\
 &= \sum_{i=1}^n \hat{y}_i^2 + \sum_{i=1}^n \hat{u}_i^2 \quad \text{BY ASSUMPTION} \\
 \sum_{i=1}^n y_i^2 &= \sum_{i=1}^n \hat{y}_i^2 + \sum_{i=1}^n \hat{u}_i^2 \\
 \sum_{i=1}^n (Y_i - \bar{Y})^2 &= \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2 + \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 \quad [\sum_{i=1}^n \hat{Y}_i = \bar{Y}]
 \end{aligned}$$

TSS

TOTAL SUM OF
SQUARE =
TOTAL VARIATION
OF ACTUAL Y
FROM THEIR SAMPLE
MEAN