

CLASSES - MICROECONOMETRIA - LABORATORIO 5 &

BINARY DTA

$$P_{1i} = E(\text{CONSUME}_i) = F(\beta_1 + \beta_2 \text{SPA}_i + \beta_3 \text{TUCE}_i + \beta_4 \text{PS}_i)$$

$$P_i \nearrow \text{PS}_i \in (0, 1) \quad i = 1 \dots N$$

$$\hat{U}_i = \text{GRADE}_i - \hat{p}_i \quad i=1 \sim N$$

$\swarrow \searrow$
 $0 \quad 1$

$\in (0,1)$

RESOLVA : $\hat{y}_i = 1$ SE $\hat{p}_i > 0.5$

$\xrightarrow{0}$ SE $\hat{p}_i \leq 0.5$

$$\hat{\hat{U}}_i = \text{GRADE}_i - \hat{y}_i$$

$\xrightarrow{0}$ SE $\hat{p}_i \leq 0.5$
 $\swarrow \searrow$
 $0 \quad 1$

$$MSE_1 = \frac{1}{H} \sum_{i=1}^H \widehat{U}_i^2 = \frac{1}{N} \sum_{i=1}^N (y_i - \widehat{y}_i)^2$$

Modello di riferimento (D_0):

$$D_0: E(y_{obs,i}) = F(\beta_2)$$

$$\hookrightarrow \widehat{p} = \frac{N_0}{N}$$

$$MSFE_0 = \frac{1}{N} \sum_i (e_{nois}^i - \hat{y}_i)^2$$

N.B. SE $\hat{p} > 0.5$, Always $\hat{y}_i = 1$ $\forall i$

$$\begin{aligned} MSFE_0 &= \frac{1}{N} \sum_i (e_{nois}^i - 1)^2 = \\ &= \frac{1}{N} \sum_i (e_{nois}^i{}^2 + 1 - 2e_{nois}^i) = \hat{p} - \hat{p} \\ &= \frac{1}{N} (N_1 + N - 2N_1) = 1 - \left(\frac{N_1}{N}\right) = 1 - \hat{p} \end{aligned}$$

SE $\hat{p} \leq 0.5$, allora $\hat{y}_i = 0$ V_i

$$MSFE_0 = \frac{1}{N} \sum_i (e_{nois}_i - 0)^2 = \frac{1}{N} \sum_i e_{nois}_i^2 =$$

$$\downarrow \quad = \frac{1}{N} \sum_i e_{nois}_i^2 = \frac{N/2}{N} \hat{p}$$

$$MSFE_0 = \begin{cases} 1 - \hat{p} & \text{SE } \hat{p} > 0.5 \\ \hat{p} & \text{SE } \hat{p} \leq 0.5 \end{cases}$$

SUM

SCALE n $NN = n(N)$

SUM IF GRADE $== 1$

SCALE $N1 = n(N)$

SCALE $N0 = NN - N1$

SCALE $POST = N1/N$

SCALAR LIST POST

PREDICT GRADE FROM PSI

PREDICT ~~PSI~~ $\frac{P_{HAT} - P_B}{P_{HAT} - P_B}$

GEN $y_{HAT} = P_{HAT} > 0.5$

GEN ERRORMEN² = $(\text{GRADE} - y_{HAT})^2$

SUM ERROR²

SCALAR MSFE₁ = $n(\text{MEAN})$

SCALAR LIST MSFE₂ MSFE₃

LJVA-DTA

$$y_i, X_i \quad i = 1 - N = 200$$

$$\text{DATA} \downarrow \text{GENERATE} \downarrow \text{GDP} \text{ : } \text{process} \quad y_i = \underbrace{(-800)}_{\beta_0} + \underbrace{300 \cdot X_i}_{\beta_1} + u_i$$

(DARI SIMULASI)

SCATTER y X , $y_{\text{line}}(0)$

GENCODE: $y = y^*$ $\text{SE } y^* > 0$
 $= 0$ $\text{SE } y^* \leq 0$

GEN $y_{\text{GEN}} = 0$

PERIODE $y_{\text{GEN}} = y$ $\text{SE } y > 0$

TRANSFORMS: $y = y^*$ SE $y^* > 0$
NULL SE $y^* \leq 0$

GEN $y_{TRU} = \cdot$

MERGE $y_{TRU} = y$ IF $y > 0$

REG y X

↳ TUTO 12 CAPICORTE

OLS "Non Distanto"

REG y X

↳ OLS SU DATA CENSURATI (Distanto)
↳ OLS (Non censored)

REC 1700 X
↳ as su dari Transisi (Disianjo)

SOLUTION 1 : MLE

$$\text{CENSURA} : \log L = \sum_{i \in y_{>0}} \log f(y_i) + \sum_{i \in y_{=0}} \log(1 - p_i)$$
$$\text{TABIT } y, X, \mathcal{L}(\theta)$$

$$\text{TRUNCAMENTO} : \log L = \sum_{i \in y_{>0}} \log f(y_i | y_{i>0})$$
$$\text{TRUNCAR } y, X, \mathcal{L}(\theta)$$

SOLUTIONS 2 : αS "Garnern" " (Mills Ratio)

TRANSFORMED : $E(y_i^* | y_i^* = 0) = \alpha_i \beta + \sigma \frac{\phi_i}{\Phi_i}$

POVE $\phi_i \equiv \phi\left(\frac{x_i \beta}{\sigma}\right)$
 $\Phi_i \equiv \Phi\left(\frac{x_i \beta}{\sigma}\right)$

GENOVA: $E(y_i^*) = \Phi_i x_i \beta + \sigma \phi_i$

PROBIT PER SINDONE $\phi_i \in \Phi_i$

$$D_i = 1 \quad \text{SE } y_i^* > 0 \quad \rightarrow \text{PROBIT}$$

$$= 0 \quad \text{SE } y_i^* \leq 0 \quad \downarrow \sqrt{\frac{\beta}{\sigma}} \quad \downarrow$$

$$E(D_i) = \Phi(x_i \beta)$$

$$\begin{aligned} \hat{\Phi}_i &\equiv \Phi(x_i, \hat{\beta}_i) \\ \Phi_i &\equiv \Phi(x_i, \beta_i) \end{aligned}$$

$$\begin{aligned} \text{GEN DUMMY} &= 1 \text{ SE } y > 0 \\ \text{P11.815 DUMMY } X &\Rightarrow \hat{\beta}_0 \in \hat{\beta}_1 \end{aligned}$$

Preis X_{BP} , X_B ↴

$$X_{BP,i} = \hat{\beta}_1 + \hat{\beta}_2 X_{i,}$$

GEN $\hat{\beta}_i = \text{KANNALDEN}(X_{BP})$

GEN $\hat{\beta}_i = \text{KANNAL}(X_{BP})$

$$\text{GEN minus} = f_i / F_i$$

$$\text{REG YTRU X minus} \quad (\text{Transfers})$$

$$\text{GEN Fix} = F_i * X$$

$$\text{REG YGEN Fix } f_i F_i, \text{ No } \text{KANS}$$

THE END _

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