

## EC226 Econometrics 1: Formula Sheet

**Two Variable Regression Model:**  $Y_i = \alpha + \beta X_i + \epsilon_i$

Least Squares estimates:  $b = \frac{\sum_i (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_i (X_i - \bar{X})^2}$   $a = \bar{Y} - b\bar{X}$

Estimation of the error variance:  $s_e^2 = \frac{RSS}{n-2} = \frac{\sum_i (Y_i - \bar{Y})^2 - b^2 \sum_i (X_i - \bar{X})^2}{n-2}$

Test on the regression slope coefficient:  $H_0 : \beta = \beta_0$ ,  $t = \frac{(b - \beta_0)}{s_b}$ , where  $s_b^2 = \frac{s_e^2}{\sum_i (X_i - \bar{X})^2}$

Standard error of prediction of  $Y_{n+1}$  given  $X_{n+1}$ :  $se(Y_{n+1}) = s_e \sqrt{1 + \frac{1}{n} + \frac{(X_{n+1} - \bar{X})^2}{\sum_i (X_i - \bar{X})^2}}$

**Multiple Regression Model:**  $Y_i = \alpha + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots + \beta_k X_{ki} + \epsilon_i$

Estimation of the error variance:  $s_e^2 = \frac{RSS}{n-k-1} = \frac{\sum_i e_i^2}{n-k-1}$

R-squared:  $R^2 = 1 - \frac{RSS}{TSS}$ ; R-bar-squared:  $\bar{R}^2 = 1 - \frac{RSS/(n-k-1)}{TSS/(n-1)}$

Test on regression coefficients:

(i) Single coefficient is equal to some hypothesised value:

$H_0 : \beta_i = \beta_{i0}$ ,  $t = \frac{(b_i - \beta_{i0})}{s_{b_i}} \sim t_{n-k-1}$

(ii) All slope coefficients are equal to zero:

$H_0 : \beta_1 = \beta_2 \dots = \beta_k = 0$ ,  $F = \frac{R^2}{1-R^2} \cdot \frac{n-k-1}{k} \sim F_{k, n-k-1}$

(iii) A subset of coefficients are equal to zero:

$H_0 : \beta_1 = \beta_2 \dots = \beta_j = 0$ ,  $F = \frac{(RSS^R - RSS^U)}{RSS^U} \cdot \frac{n-k-1}{j} \sim F_{j, n-k-1}$

(iv) The coefficients from some sub-set of observations are equal to those of some other sub-set of observations:

$H_0 : \alpha^1 = \alpha^2, \beta_1^1 = \beta_2^2 \dots = \beta_k^1 = \beta_k^2$ ,  $F = \frac{[RSS^R - (RSS^1 + RSS^2)]/(k+1)}{(RSS^1 + RSS^2)/[n-2(k+1)]} \sim F_{k+1, n-2(k+1)}$

Durbin-Watson Test statistic: 
$$d = \frac{\sum_{t=2}^n (\hat{u}_t - \hat{u}_{t-1})^2}{\sum_{t=1}^n \hat{u}_t^2}$$

Durbin's h-statistic:  $h = \phi \sqrt{\frac{n}{1-n s_c^2}}$ , where  $s_c^2 =$  Estimated variance of coefficient on lagged dependent variable and  $\phi =$  Estimated 1st autocorrelation term.

**Unit Root Tests:**

ADF:  $H_0 : \gamma = 0, H_1 : \gamma < 0$

Model A: 
$$\Delta y_t = \gamma y_{t-1} + \sum_{j=1}^q \Delta y_{t-j} + \epsilon_t$$

Model B: 
$$\Delta y_t = \mu + \gamma y_{t-1} + \sum_{j=1}^q \Delta y_{t-j} + \epsilon_t$$

Model C: 
$$\Delta y_t = \mu + \alpha t + \gamma y_{t-1} + \sum_{j=1}^q \Delta y_{t-j} + \epsilon_t$$

Test statistic:  $\frac{\hat{\gamma}}{se(\hat{\gamma})} \sim$  MacKinnon critical values

Error Correction Model (ECM):  $\Delta y_t = \alpha + \gamma_1 \Delta x_{t-1} + \alpha_1 \Delta y_{t-1} + \delta s_{t-1} + u_t$  where  $s_t = y_t - (a + b x_t)$

**Limited Dependent Variable Model:**

Linear Probability Model:  $Pr[Y_i = 1] = X_i' \beta$

Logit Model:  $Pr[Y_i = 1] = F(X_i' \beta) = \frac{\exp(X_i' \beta)}{1 + \exp(X_i' \beta)} = \Lambda(X_i' \beta)$

Probit Model:  $Pr[Y_i = 1] = F(X_i' \beta) = \int_{-\infty}^{X_i' \beta} (2\pi)^{-\frac{1}{2}} \exp(-z^2/2) dz = \Phi(X_i' \beta)$

Interpreting Coefficients:  $\frac{\partial E(Y_i)}{\partial X_{ji}} = \frac{\partial F(X_i' \beta)}{\partial (X_i' \beta)} \cdot \beta_j$

where  $\frac{\partial F(X_i' \beta)}{\partial (X_i' \beta)} = f(X_i' \beta)$  and  $f(X_i' \beta)$  is the pdf.

pdf for logit model:  $f(X_i' \beta) = \lambda(X_i' \beta) = \Lambda(X_i' \beta)[1 - \Lambda(X_i' \beta)]$

pdf for probit model:  $f(X_i' \beta) = \phi(X_i' \beta) = (2\pi)^{-\frac{1}{2}} \exp(-z^2/2)$