

# Diagnostic Tests

Some practical ways to check the residual assumptions of the simple Classical Linear Regression Model (CLRM)

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# How to check graphically The Assumptions of the Simple Classical Linear Regression Model (CLRM)

- Consider the simple classical linear regression model  $y_t = \alpha + \beta x_t + u_t$  estimated from a sample by  $y_t = \hat{\alpha} + \hat{\beta} x_t + \hat{u}_t$ ,  $t=1, 2, \dots, T$

- Let  $\hat{y}_t = \hat{\alpha} + \hat{\beta} x_t$  be the predicted value for observation  $t$

<u>Assumption</u>	<u>Graphical method</u>
1. $E(u_t) = 0$	Plot $\hat{u}_t$ against $t$ (always true if estimated using OLS)
2. $\text{Var}(u_t) = \sigma^2$	Homoscedasticity: plot $(t, \hat{u}_t)$ and $(x_t, \hat{u}_t)$ ; look at the variation
3. $\text{Cov}(u_i, u_j) = 0$	Plot $(\hat{u}_t, \hat{u}_{t+1})$ to look for possible autocorrelation of first order May also plot $(\hat{u}_t, \hat{u}_{t+k})$ for higher order autocorrelation $k > 1$
4. $\text{Cov}(u_t, x_t) = 0$	Plot $(x_t, \hat{u}_t)$ ; look for possible dependence
5. $u_t \sim N(0, \sigma^2)$	Histogram compared to normal curve or QQ-plot (straight line if normal)

# How to test The Assumptions of the Simple Classical Linear Regression Model (CLRM)

- Consider the simple classical linear regression model  $y_t = \alpha + \beta x_t + u_t$  estimated from a sample by  $y_t = \hat{\alpha} + \hat{\beta} x_t + \hat{u}_t$ ,  $t=1, 2, \dots, T$
- Let  $\hat{y}_t = \hat{\alpha} + \hat{\beta} x_t$  be the predicted value for observation  $t$
- | <u>Assumption</u>               | <u>Tests</u>   |
|---------------------------------|--|
| 1. $E(u_t) = 0$                 | Always true if estimated using OLS; good to check that $\sum \hat{u}_t = 0$  |
| 2. $\text{Var}(u_t) = \sigma^2$ | White's test for heteroscedasticity:<br>Estimate quadratic regression $\hat{u}_t^2 = \hat{\alpha}_0 + \hat{\beta}_0 x_t + \hat{\beta}_1 x_t^2 + \hat{\varepsilon}_t$<br>(should be insignificant, otherwise $\text{Var}(u_t)$ depends on $x_t$ ) |
| 3. $\text{Cov}(u_i, u_j) = 0$   | Durbin-Watson test for possible autocorrelation of first order<br>Breusch-Godfrey for higher order autocorrelation   |
| 4. $\text{Cov}(u_t, x_t) = 0$   | Calculate correlation (should be insignificant) or<br>Estimate linear regression $\hat{u}_t = \hat{\alpha}_1 + \hat{\beta}_1 x_t + \hat{\varepsilon}_t$ (should be insignificant)  |
| 5. $u_t \sim N(0, \sigma^2)$    | Jarque-Bera-test or some other normality test (e.g. Shapiro-Wilk test)   |