

Business Forecasting

Optional Advanced Notes on Regression



Checking Regression Assumptions



Before using any regression models for interpretation or forecasting you need to ensure the assumptions of the regression model are valid

Requires checks of model adequacy involving model significance tests and the residuals of the model

This includes visual inspection of **residual plots** against **time** and against **predicted Y** and **all X's**

There will also be several **statistical tests** on the **residuals** to further test the validity of error assumptions

Common Problems with Regression



<u>Model Mis-specification</u>: Exclusion of relevant variables, inclusion of irrelevant variables and incorrect functional form. Mis-specified models likely to lead to **erroneous conclusions** and **forecasts**. Detection: Logic, economic theory, model significance tests and residual plots. Remedy: Use correct model

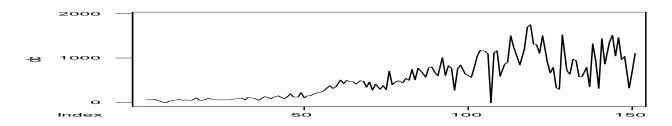
Heteroskedasticity: Constant variance assumption of the error term is violated. Commonly, residuals increase in magnitude over time

Violation of homoscedasticity (constant error variance) may lead to <u>inefficient estimates and forecasts</u> (still <u>unbiased</u>) and <u>invalidity of the standard inferential tests</u>. Detection: White Test and residual plots. Remedy: use logarithms of Y, X

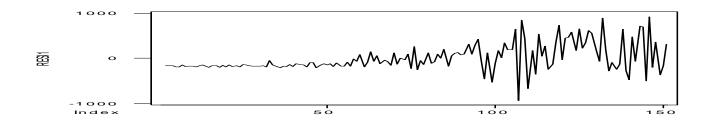
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Heteroscedasticity

<u>Detection:</u> Mostly evidence from <u>time series plots</u> will establish if there is heteroscedasticity.



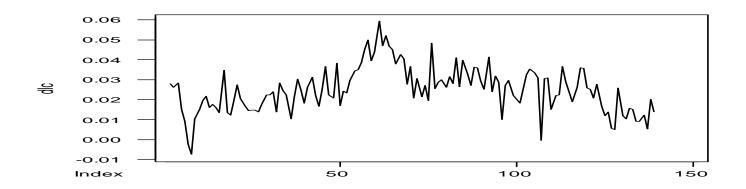
From the example above, <u>volatility of the time series</u> (<u>random component variance</u>) is increasing over time. Alternatively, <u>plots of residuals</u> from estimated regression model may also show heteroscedastic nature of the time series

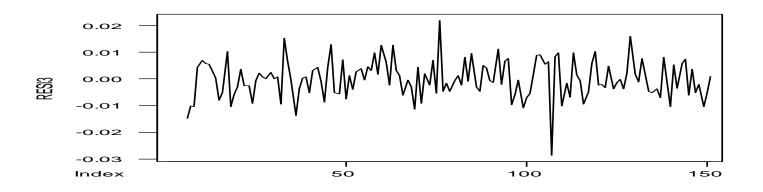




Remedy (example)

Below are <u>logarithmic transformations</u> (natural logs) of the previous graphs.





Autocorrelation



A key assumption for <u>OLS estimation efficiency</u> is <u>independence</u> of the error term with respect to error terms at other observations

May be violated with time series data (<u>autocorrelation</u>)

Error in one period may be related to error terms in other periods

$$\varepsilon_{t} = f(\varepsilon_{t-1}, \varepsilon_{t-2}, \dots \varepsilon_{t-p})$$

Severe autocorrelation may lead to <u>inefficient and/or biased</u> estimation/forecasts

<u>Detection:</u> Residual ACF and PACF, DW test (1st order), LM test (any order)

Remedy: Correct specifications, Autocorrelation corrections



Time Series Regressions

Time Series regressions present specific estimation problems due to the likely **connection** between observations over time

For example, two unconnected (in theory) time series that are both trending or have similar structure are likely to produce an apparently significant relationship in a regression even though in theory (and in practice) they are not connected

In general, problems in <u>estimating and interpreting</u>
<u>regressions</u> <u>may</u> arise when the levels of the series used (Y and X's) vary with time or the time series are <u>non-stationary</u>

This is called spurious regression and can lead to erroneous conclusions and forecasts



Spurious Regression

Regression of non-stationary variables may lead to erroneous estimation results

Experiments show two time series that are <u>non-stationary</u> but are <u>independently generated</u> (ie no relationship exists between the time series) if regressed against one another <u>may</u> lead to results which suggest the variables are related <u>(large t and F statistics, R² close to 1)</u>

The relevant regression tests and inferential procedures will suggest incorrectly the two variables have a relationship and the estimated regression model is valid

Forecasts and predictions are likely to be poor



Spurious Regression (cont)

The consequences of spurious regression results cannot be overstated

Results will suggest the model will be a useful predictive tool (R² high, F high). Yet, since **no relationship** exists between the variables **predictive performance will be poor**

Therefore great caution must be exercised in regression of time series to ensure **estimated results are valid**

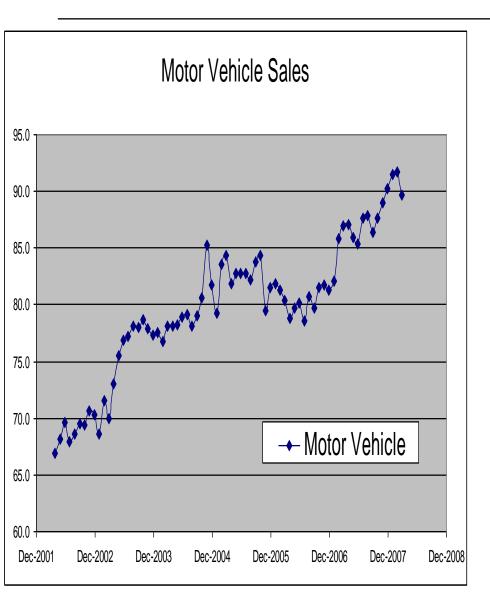
Remedy: Economic Theory, Logic, Co-integration, Use

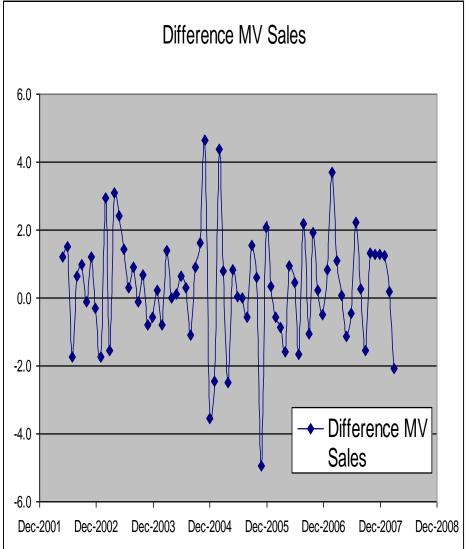
Stationary series (possibly difference original time series)

Detection: Line Plots, ADF test (Non-stationarity)

Comparing Original and Differenced









Dynamic Structure

It is likely impacts of independent variables on the dependent variable will not be entirely **instantaneous**

The modeller needs to consider a dynamic structure

The typical dynamic models used are the **distributed lag models** or models using **lagged dependent variables** or models with both

Selection of the appropriate dynamic model is a complex task with many pitfalls for the unwary. There may possibly be **estimation problems or theoretical problems** with any type of dynamic structure chosen. It is best to consult an expert.