

Intelligent Research: Hands-on Training on EViews Application

Presentation at the Interactive Session with UG and PG students held at PES University, Bangalore South Campus, Electronic City, Bangalore 560100, India, on December 21, 2015

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December 21, 2015 @ PES University, Bangalore, India

Objectives...

1. To introduce participants into the use of *EViews* econometric software
2. To experiment with the use of the *EViews* software in performing uncomplicated time series analysis, focusing on single time series data.

Anticipated learning experience...

1. At the end of this workshop, you should have appreciable knowledge of EViews and related basic econometric concepts
2. You should also be able to use *EViews*, to perform simple time series analysis.

Additional information

- This presentation is made to support BBA / MBA students in their use of EViews econometric software.
- It is not a complete guide on how to use the software, but only meant to help the students to perform basic econometric analysis needed for research and statistical data analysis.
- This presentation is not a statistics theory guide, and should not be seen as a substitute for textbook on the subject. The presentation assumes students' understanding of basic underlying statistical principles.
- We will be using six different work files to illustrate the use of EViews. These work files have been distributed to the students ahead of the workshop.

What EViews is about basically...

- EViews is a spread sheet software used for various types of data analysis.
- EViews has some similarity to the MS Excel
- EViews can be used to perform sophisticated data analysis, regression, and forecasting tools on Windows based computers.
- EViews enables you to do more than the conventional Excel analysis, (e.g. descriptive statistics); with EViews, you can do more advanced calculations, regressions and simulations, which you won't find in Excel.
- Additionally, it has increased functionality, and operates at a much faster pace, easy to use.
- Students / workshop participants are expected to have a full version of EViews 6.0/7.0 is in-stalled in their respective laptops.
- **So, explore & explore!**

Key application areas...

1. Asset returns
2. Asset pricing – technical analysis & fundamental analysis
3. Portfolio spreads
4. Interest rates / yields
5. Yield of a portfolio of bonds
6. GDP and other economic series, etc.

Econometrics

- A primer

Resources

- ❑ *Introductory Econometrics for Finance, 2nd edn.* by Chris Brooks (2008). Cambridge University Press.
- ❑ *Statistics for Business: Decision-making and analysis* by Robert Stine & Dean Foster (2012).
- ❑ *Business Statistics*, by S. L. Aggarwal & S. L. Bhardwaj (2010). New Delhi: Kalyani Publishers

Great inputs from:

The Workshop on Financial Econometrics, Correlation, Causation and Co-integration
@SDMIMD, Mysore, India on 20th August 2015
delivered by



Dr Kuldeep Kumar
Professor, Department of Economics & Statistics,
Bond University, Australia. Email: kkumar@bond.edu.au

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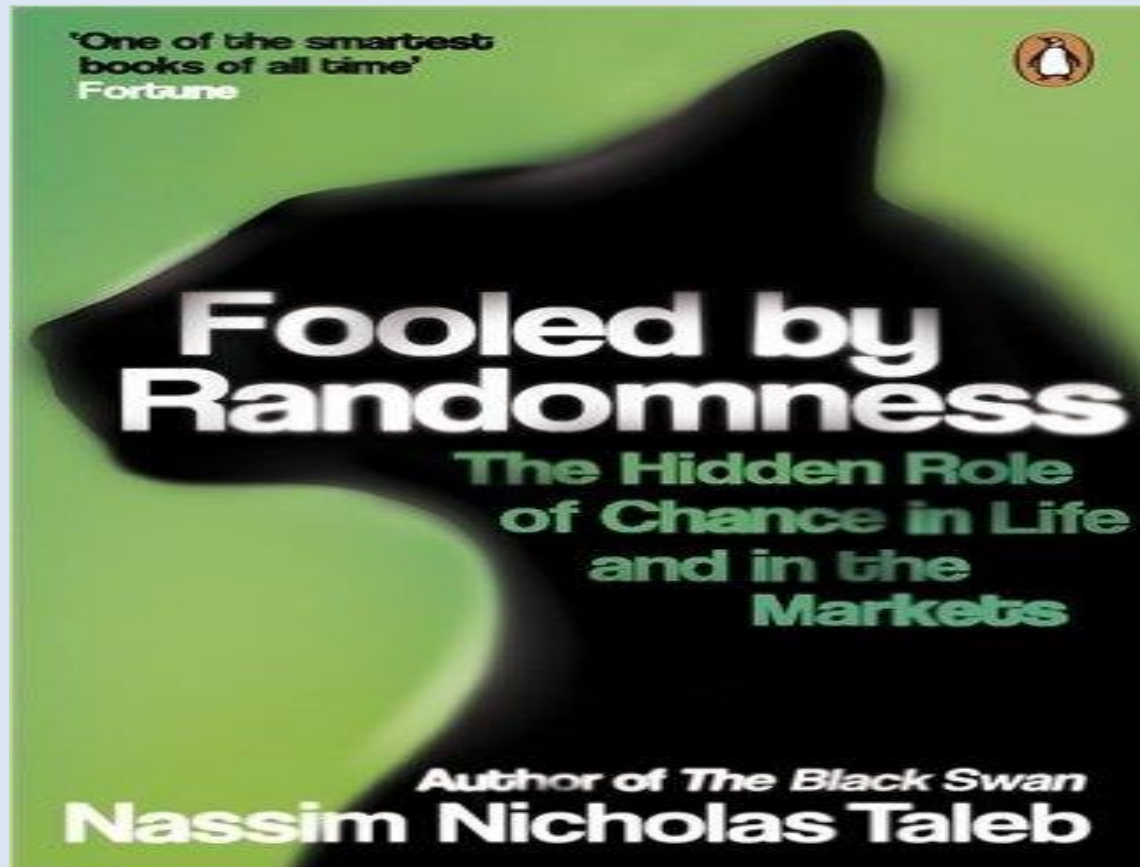
***EViews:
A helpful resource***



**□ *Introduction to EViews
6.0*** by Anders Thomsen *et al*
(January 2013), Analytics
Group, Aarhus University,
Denmark.

***Interesting econometrics!
A compulsory read:***

BAZE
UNIVERSITY
ABUJA



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On the underlying maths...

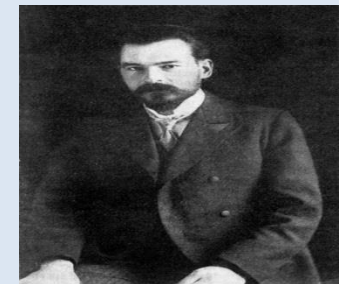
- **Observe:** We have reduced the mathematical aspects to the barest minimum.
 - We are focusing on practical principles and applications of financial econometrics.
 - However, workshop participants are advised to consult the primary references for more mathematical and technical details.

*Appreciate
the contributions of these thinkers?*

- ❑ Polish Economist Pawel Ciompa
- ❑ Norwegian Economist Ragnar Frisch →
- ❑ Thorvald N. Thiele – 1880 paper
on *least squares*

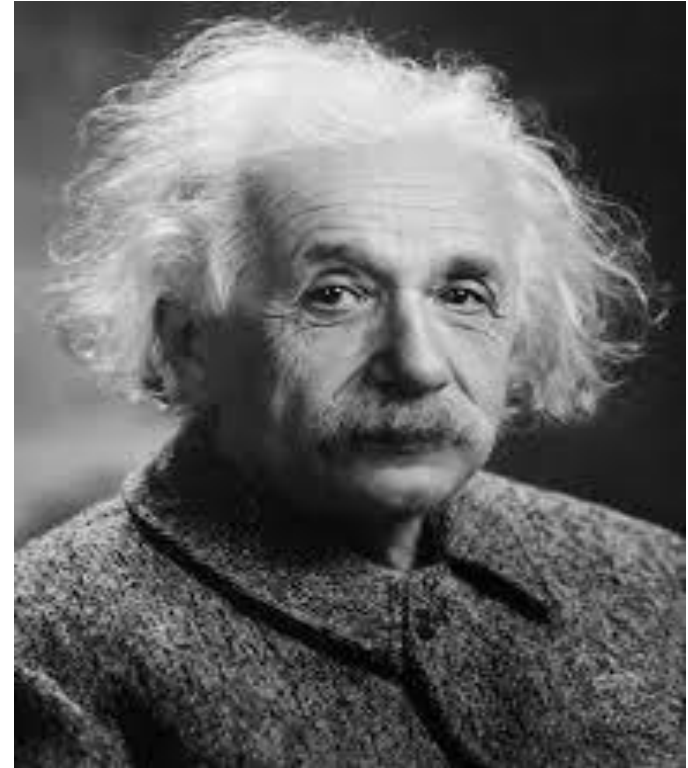


- ❑ Louis Bachalier – 1900 PhD thesis on ‘*Theory of Speculation*’
- ❑ Albert Einstein – 1905 paper (see excerpt in the next slide)
- ❑ Marian Smoluchwski on statistical physics.



Digest this excerpt from Albert Einstein's 1905 paper

- *"It must clearly be assumed that each individual particle executes a motion which is independent of the motions of all other particles; it will also be considered that the movements of one and the same particle in different time intervals are independent processes, as long as these time intervals are not chosen too small..."*
- https://en.wikipedia.org/wiki/Stochastic_process
[21/12/2015]



“... as long as these time intervals are not chosen too small...” - Einstein

“Time makes more converts than reason.”

- Thomas Paine

What is 'Econometrics'?



What 'Econometrics' is all about...

□ Some useful definitions:

1. = application of mathematics, statistical methods, and computer science to economic data
2. = a branch of economics that aims to give empirical content to economic relationships, e.g.

- **Unemployment & economic growth**
- **Wages & years of education**
- **Income & Spending**

□ Basic econometric methods / estimators / tools:

1. **Linear regression model** – OLS (ordinary least squares) based on Gauss-Markov assumptions
2. When 'normality' assumption is violated, other estimation techniques are applied:
 - Maximum likelihood estimation
 - Generalized methods of moments
 - Generalized OLS
 - Bayesian statistics

What 'Econometrics' is all about...

1. = Quantitative analysis of actual economic phenomena on the basis of theory, observation, and appropriate methods of inference.
2. = Translating data into models to make forecasts and to support decision-making

3. = Sifting through massive data to extract simple relationships

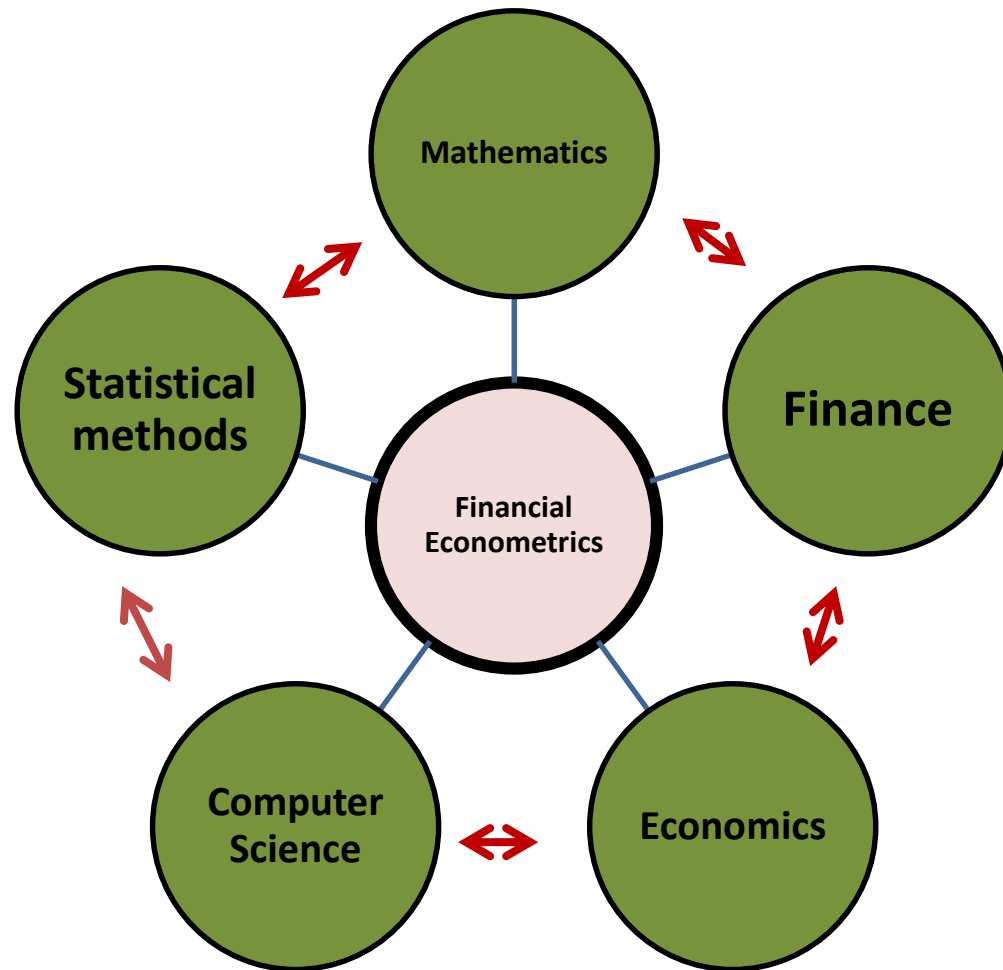
What about Financial Econometrics?

- ***Financial econometrics*** applies mathematical, statistical, and computer science techniques to problems in finance.
- ***Financial econometrics*** studies how the supply of ***capital*** and its use are considered and measured.
- **Note**: '***Capital***' is economic resource – a resource or resources that can be used to generate economic wealth (Encarta Dictionaries, 2009).

Aspects of Financial Econometrics?

- ❑ Asset valuation – real estate, stocks, bonds, derivatives, currencies and other financial assets – CAPM, APT, EMH, etc.
- Corporate Finance
- Tests of random walk hypothesis
- Term structure of interest rates
- Causality analysis
- Event analysis, etc.

*****Financial Econometrics...



Why we need to follow the protocol in Econometrics

- ❑ Observational data versus controlled experiments
 - Systems analysis and control theory
 - Observational study design is not peculiar to econometrics; other disciplines also use the approach
 - The approach basically allows the analyst to do model estimation and investigate a model's empirical consequences.
 - Hence, in econometrics, we use the regression methods because we cannot use controlled experiments.
 - Most importantly, data analysis on the basis of observational data should be guided by the study protocol / procedure / - taking models through statistical trials**

... *Utility*

1. Economics
2. Finance
3. Marketing
4. OB & HRM
5. OM
6. Machine performance

7. Engineering
8. Data analysis
9. *Climate change* –
e.g. verifying causal
relationship between
greenhouse-gas emissions
and higher temperatures
(*The Economist*, November 28th-
December 4th 2015, “Clear thinking on
climate change”, p. 10)

... a wide range of job opportunities for analysts

1. Role as Quantitative analysts in public and private organizations
2. Data analysis, structuring and transaction advisory
3. Credit analysis / scoring /provisioning
4. Investment banking/
Corporate finance / trading
5. Asset management /
portfolio optimization /
trading strategy
development
6. Credit cards
7. Risk management
8. Mortgage banks
9. Management Consulting
10. Derivatives pricing and
hedging
11. Business/asset valuation
12. Venture capital
13. Foreign exchange
services, etc.
14. Operations management
15. Research / academia

Still on job opportunities for analysts...

- Historical analysis of an organization
- Projecting an organization's financial performance
- Project finance
- Real estate
- Oil and Gas projects
- Banking & Financial Institutions
- Personal finances
- Non-profit organizations / NGOs
- Government – at Federal/central/national, State/Regional, local council
- Investment Banking
- Academia - research and educational centres

*The essence of financial econometrics -
The **BIG PICTURE***

- ❑ Well-being of the society...
- ❑ Internet age / smart phones & massive data availability...
- ❑ Rapid and monumental changes and implications on the future of the professions – e.g. technology ‘disruptions’ rendering traditional approaches antiquated, opaque, and unaffordable (Suskind & Suskind, 2015)

Financial econometrics: Wide variety of data sets...



- ❖ Fundamental economics
- ❖ Real estate
- ❖ Human resources management
- ❖ Accounting
- ❖ Advertising
- ❖ Agriculture
- ❖ Banking & Finance
- ❖ Business

- ❖ Finance & investments
- ❖ Marketing
- ❖ Opinion polls
- ❖ Transportation
- ❖ Sports
- ❖ Life sciences
- ❖ Physics and engineering, etc.

How 'straight' is our datasets?

- *“Capitalism does indeed have very bad times, and it goes off the rails temporarily.”*
 - **W. Easterly (2008)** in *Creative Capitalism*

Formulating your hypothesis...

□ **Null Hypothesis: H_0**

- H_0 is set up by the investigator with the intention of being rejected based on the available statistical evidence.

□ **Alternative Hypothesis: H_1**

- H_1 is the hypothesis accepted by the investigator after the H_0 has been rejected.

Recall:

Type I and Type II decision errors

Type I error – alpha (α) error:

- True H_0 is rejected
- False H_1 is accepted

Type II error – beta (β) error:

- True H_1 is rejected
- False H_0 is accepted
- Why is Type II error usually emphasized in statistical literature?

Type I and Type II decision errors

Continuation

- ❑ Pick hypothesis before looking at the data to avoid bias – *apriori* expectation
- ❑ Don't confuse statistical importance with substantive importance.
- ❑ The p -value is not necessarily the probability that the null hypothesis (H_0) is true; the p -value already assumes that H_0 is true.
- ❑ Rather, it is the probability of rejecting H_0 incorrectly on the basis of your results that is displayed in sample assessment.

- ❑ You have learnt about how to interpret the adequacy of statistical models, such as beta coefficients, R-squared, adjusted R-squared, finding the critical value of F statistic and verifying the F calculated value.
- ❑ But ensure that you are able to summarize your results in everyday language
- ❑ Less jargons, please!

Recall
Formal testing statistics...

- t-test statistics (the student *t*-test) – small sample test, i.e. $n < 30$
- Z-test statistics (normal test) – large sample test i.e. $n \geq 30$
- F-test statistics (joint test)
- Chi-square distribution χ^2

Recall 'confidence levels'...

- ❑ Confidence level – the extent of confidence or certainty that the investigator has; conventionally (2-tailed):
 - $Z_T = Z_\alpha = 1\% = 2.58$
 - **$Z_T = Z_\alpha = 5\% = 1.96$** (This is the widely used range in social & management sciences)
 - $Z_T = Z_\alpha = 10\% = 1.65$

*****Decision Rule:

- Calculated test statistic > tabular (critical) value: Accept H_1 and reject H_0
i.e. the parameter testes is statistically significant
- Calculated test statistic < tabular (critical) value: Accept H_0 and reject H_1
i.e. the parameter testes is statistically insignificant

******Statistical testing:*

On simpler (less confusing!) approaches...

- ❑ The p -value is the probability that the test statistic (z^* , t^* , x^* , F^*) will be exceeded, and thus p is called the observed level of significance, in contrast to the α -value which is a priori-level of significance.
- ❑ The default value of $\alpha = 0.05$, and the relationship between P and α is as stated below (Kothari & Garg, 2014):
 - i. **If $p \geq \alpha$, do not reject H_0**
 - ii. **If $p < \alpha$, reject H_0** (Chris Brooks, 2008).

Statistical significance test: ***On simpler (less confusing!) approaches...***

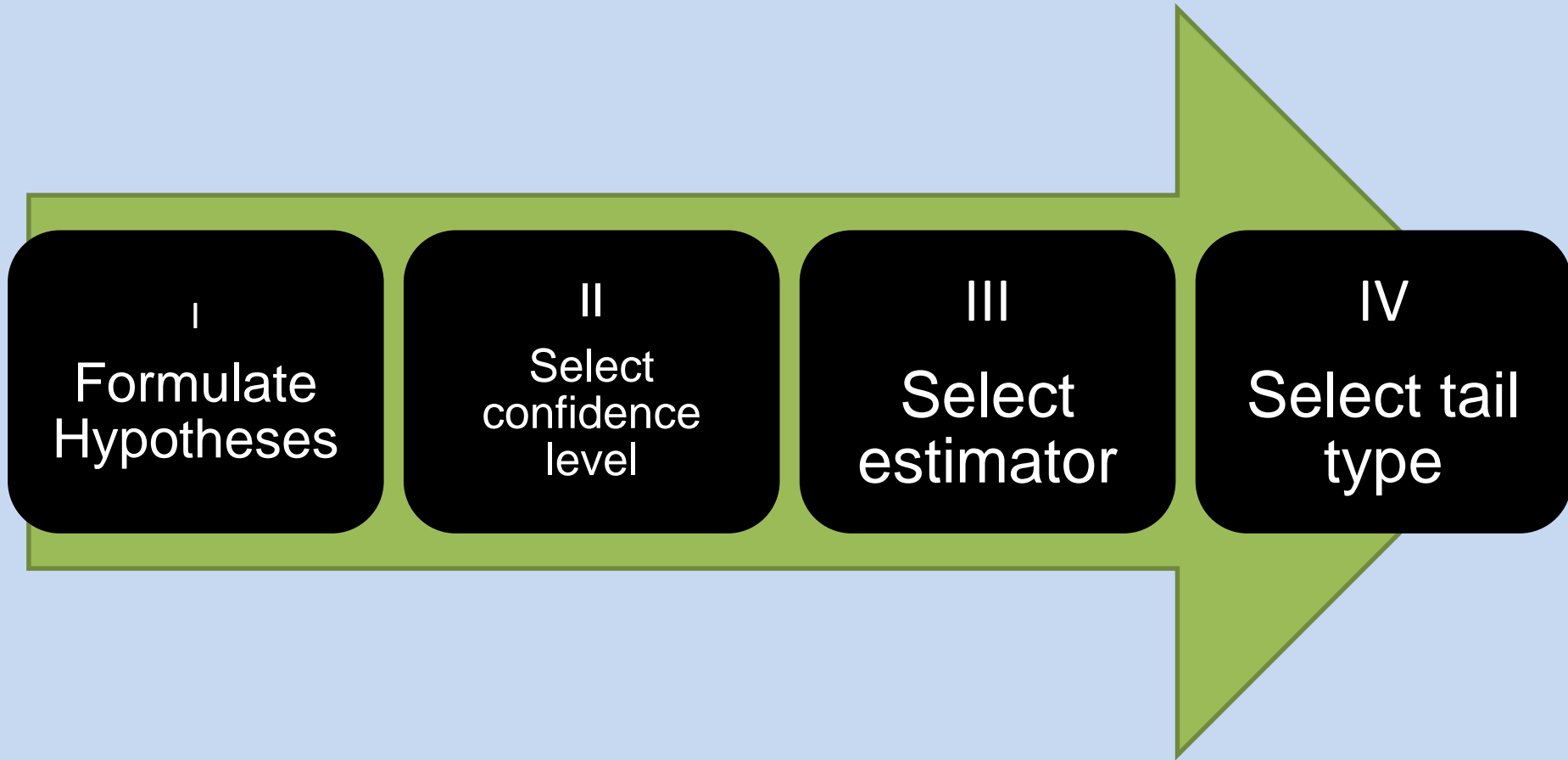
- In essence, using a table is not necessary when you have the exact probability for a statistic.
- Your econometric software can calculate exact probabilities for most test statistics. If you have an exact probability output from computer software, you simply compare it to your critical alpha level.
- If the exact probability is less than the critical alpha level, your finding is significant (i.e. H_0 rejected)
- if the exact probability is greater than your critical alpha level, your finding is not significant (i.e. H_0 accepted)

Formal statistical 'significance' test:

To sum up...

S/No	Observation	Interpretation rule
1	If <u>calculated</u> t-statistic < 1.96 (i.e. 5% significance level)	Accept Ho
2	If <u>calculated</u> $p > 0.05$ (i.e. 5% significance level)	Accept Ho

Recall hypothesis testing
The standard procedure revisited...



Hypothesis testing: Standard procedure revisited...

V

Calculate test
statistic

VI

Compare t-value to the
rejection region

VII

Make your
conclusions

Statistical testing:

...made easier by technology, but...

- ❑ The manual procedure for hypothesis testing can be tedious.
- ❑ Happily, there are statistical tools available these days to solve the problems with relative ease, once they have been carefully / properly defined.
- ❑ All you need to do is enter the given data and then issue the appropriate commands.
- ❑ Software supporting hypothesis tests: SPSS, MINITAB, EViews, MS Excel, and so on.
- ❑ When the results of manual solutions are compared to the computer solutions, we see that they are identical.
- ❑ But...

some provisos...

- ✓ Understand the methodology (which we described earlier).
- ✓ Be familiar with the output generated by the computer (you will see some examples of these in the practical session).
- ✓ Be knowledgeable enough to interpret the computer results meaningfully (again, the recurring imperative of proper understanding of the related theory, the business and the question at hand)

Time series analysis

A primer

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Time...

- ❑ **Time**: denoted t , is a dimension that enables two identical events occurring at the same point in space to be distinguished, measured by the interval between the events (Encarta Dictionary, 2009).
- ❑ *“Time with its continuity logically involves some other kind of continuity than its own. Time as the universal form of change, cannot exist unless there is something to undergo change and to undergo a change continuous in time, there must be continuity of changeable qualities.”*
- Charles Sanders Peirce

Spurious correlation...

*“First impressions
can often lead to
wrong conclusions.”*

-Our Daily Bread, 2015, p. 346.

Spurious Correlation...

- ❑ Correlation does not mean causation
- ❑ Consequently regression may be also spurious and interpretation may not be valid.

Non-stationary time series

Basic classifications / sources...

1. Seasonality – pervasive in economic time series
2. Deterministic trends (time trends)
3. Unit Roots (Stochastic trends)
4. Structural breaks

- ❑ **Seasonality** – data exhibiting a non-constant deterministic pattern with an annual frequency.
- ❑ **Hebdomadality** – data exhibiting day-of-the-week deterministic effects.
- ❑ **Diurnality** – Data that exhibit intra-day deterministic effects

Spurious regression...

- According to Kumar (2015), when nonstationary time series are used in a regression model, the results may spuriously indicate a significant relationship when there is none
- In these cases the least squares estimator and least squares predictor do not have their usual properties, and t -statistics are not reliable
- Since many macroeconomic and financial time series are nonstationary, it is crucial to exercise care when estimating regressions with dynamic variables.

Still on spurious regression...

- ❑ Usually the regression only tells us there is some 'relationship' between x and y , and does not tell the nature of the relationship, such as whether x causes y or y causes x .
- ❑ Granger test can be used in investigating whether or not Y causes X . (Kumar, 2015)

Understanding time-based data...

- ❑ In time-based data, the change in a variable is an important concept.
- ❑ The change in a variable y_t , also known as its first difference, is given by $\Delta y_t = y_t - y_{t-1}$.
- ❑ Δy_t is the change in the value of the variable y from period $t - 1$ to period t

Understanding time series analysis...

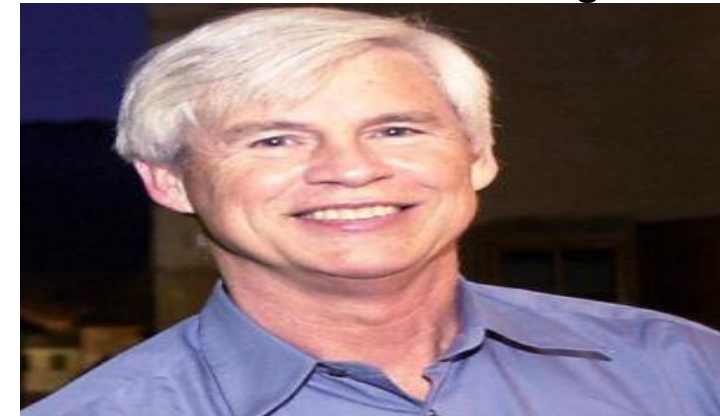
- ❑ Time-based data – showing the dynamic movement of a phenomenon over a period of time, usually at equal intervals.
- ❑ Used in any domain of applied sciences and engineering that involve temporal measurements.
- ❑ Sequence of data points – successive movements over time, e.g. ocean tides, closing stock prices on Indian stock markets, etc.

Salute to knowledge-building in modern Econometrics **Robert F. Engle & Sir Clive William John Granger**

- ❑ Sir Clive William John **Granger** was a British economist, who taught in Britain at the University of Nottingham and in the United States at the University of California, San Diego. In 2003, Granger was awarded the Nobel Prize in Economic Sciences in recognition of what he and his co-winner, Robert F. **Engle** (picture below) had made contributions to the analysis of time series data that had changed fundamentally the way in which economists analyse financial and macroeconomic data, enabling economists to make more accurate stock-market forecasts.
- ❑ Engle co-founded the **Society for Financial Econometrics** (SoFiE), a network of experts committed to the field of financial econometrics. In 2009, he founded the **Volatility Institute** at NYU. This organization promotes research on the issue of risk in financial markets. Engle also directs the NYU V-Lab, which provides forecasting and analysis of market trends using both classic models and newer tools.
- ❑ <http://www.biography.com/people/robert-f-engle-13607503#major-contributions> - 28/11/2015]



Sir Clive William John Granger



Stationarity & its essence

- ❑ ***Stationarity*** - a concept / tool aimed at transforming raw data to become stationary (so that joint probability distribution parameters such as mean and variance do not change when shifted in time).
- ❑ ***Stationary time series*** is one whose statistical properties such as mean, variance, autocorrelation, and so on are all constant over time.
- ❑ *****The world of science is cautious about trying to extrapolate *regression* models fitted to non-stationary data.

Stationarity & its essence

- ❑ **Stationarity** acknowledges **stochastic processes** – sequence of scenarios along the course of time (Taleb, 2004); i.e. stochastic process is an arbitrary sequence of random data
- ❑ Random walk: $y_t = y_{t-1} + \epsilon_t$
- ❑ Hence, realizing randomness of time series, stationarity is a measure of regularity exploited to allow us to make estimation of unknown parameters and characterize the dependence between observations across time.
- ❑ If dataset is allowed to change in an unpredictable manner, constructing a meaningful model would be difficult or impossible.

Random run (Taleb, 2007)

- A random sample path (*random run*) is the mathematical name for such a succession of virtual historical events, starting at a given date and ending at another, except that they are subjected to some varying level of uncertainty (e.g. body temperature, blood pressure, exchange rate fluctuations, stock price, corporate / personal wealth, etc.)
- *Stochastic process* (Greek name for 'random')- the dynamics of events unfolding with the course of time.

Random life...

○ *“There is no man
for all seasons.”*

- Prasanna Chandra

Stationarity - continuation...

- Formally, a time series y_t is stationary if its mean and variance are constant over time, and if the covariance between two values from the series depends only on the length of time separating the two values, and not on the actual times at which the variables are observed (Kumar, 2015).
- That is, the time series y_t is stationary if for all values, and every time period, it is true that:
 - $E(y_t) = \mu$ (constant mean)
 - $\text{var}(y_t) = \sigma^2$ (constant variance)
 - $\text{cov}(y_t, y_{t+s}) = \text{cov}(y_t, y_{t-s}) = \gamma_s$ (covariance depends on s , not t)

Explaining stationarity (Kumar, 2015) continuation

□ The autoregressive model of order one, the AR(1) model, is a useful univariate time series model for explaining the difference between stationary and non-stationary series:

$$y_t = \rho y_{t-1} + v_t, \quad |\rho| < 1$$

- The errors v_t are independent, with zero mean and constant variance, and may be normally distributed
- The errors are sometimes known as “shocks” or “innovations”

Explaining stationarity (Kumar, 2015) continuation

- ❑ The main reason why it is important to know whether a time series is stationary or non-stationary before one embarks on a regression analysis is that there is a danger of obtaining apparently significant regression results from unrelated data when non-stationary series are used in regression analysis.
- ❑ Such regressions are said to be **spurious**
- ❑ Assume two independent random walks:

$$rW_1 : y_t = y_{t-1} + v_{1t}$$

$$rW_2 : x_t = x_{t-1} + v_{2t}$$

- ❑ These series were generated independently and, in truth, have no relation to one another, yet this may not be apparent from its graph (see next slide)

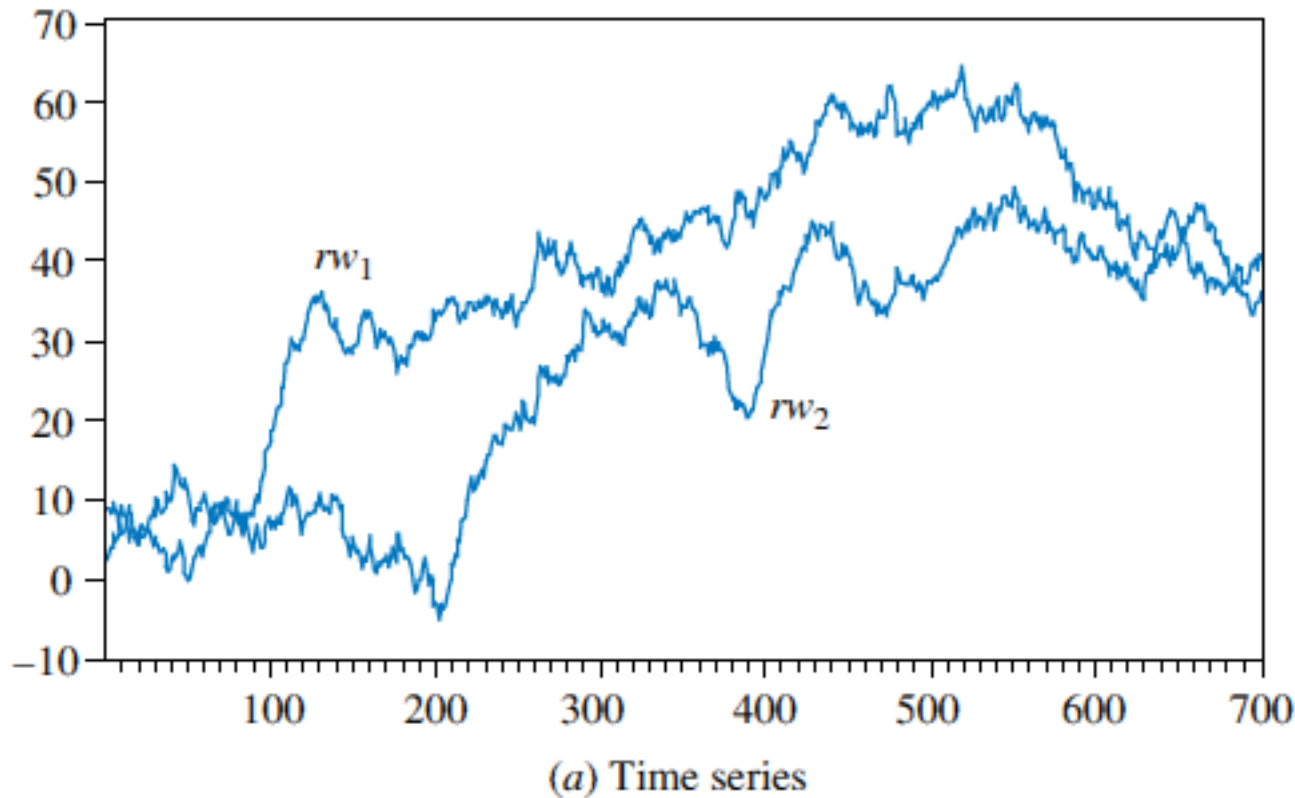
The rationale behind stationarity tests...

- ❑ We need to stationarize a time series to be able to obtain meaningful sample statistics such as means, variances, and correlations with other variables.
- ❑ ***Such statistics are useful as descriptors of future behaviour only if the series is stationary.***

- **Other motivations:**

1. Most statistical forecasting methods are based on the assumption that the time series can be rendered approximately stationary (i.e., "stationarized") through the use of mathematical transformations.
2. A stationarized series is relatively easy to predict: you simply predict that its statistical properties will be the same in the future as they have been in the past. Your computer software normally takes care of the requisite computations and transformation details.
3. Stationarizing a time series through differencing (if required) is an important part of the process of fitting an ARIMA model. (Hatemi, 2004).

Still on stationarity (Kumar, 2015) continuation



‘Correlation’ – A reminder...

□ **Correlation is relatedness of variables - the degree to which two or more variables are related and change together (Encarta Dictionary, 2009)**

Time series analysis: Understanding other key terms...

- ❑ **Cointegration** (long-run relationship between two moving variables) is a statistical property of a collection (X_1, X_2, \dots, X_k) of time series variables. First, all of the series must be integrated of order 1. Thereafter, if a linear combination of this collection is integrated of statistical order zero, then the collection is said to be co-integrated.
- ❑ **Autocorrelation** - a mathematical tool for finding repeating patterns, such as the presence of a periodic signal obscured by noise – serial dependence

- ❑ **Lags:** a period of time between one event and another.
- ❑ **Random walk** – each value is completely a random step away from the previous value (not auto-correlated)

Understanding ‘differencing’

- ❑ **Differencing** – this is a viable method of transforming a nonstationary series to become stationary
- ❑ **First difference** of a time series is the series of changes from one period to the next. If Y_t denotes the value of the time series Y at period t , then the first difference of Y at period t is equal to $Y_t - Y_{t-1}$.
- ❑ First difference is useful filter to separate a “trend” from “cyclic” component in a series.
- ❑ Don’t over-difference – applying the difference operator to a stationary series.

Understanding Lags... & Autocorrelations...

- ❑ **Distributed lag model** is a model for time series data in which a regression equation is used to predict current values of a dependent variable based on both the current values of an explanatory variable and the **lagged** (past period) values of this explanatory variable.
- ❑ **Autocorrelations** are to autocovariances as correlations are to covariances.
- ❑ The **autocorrelation function (ACF)** relates the **lag length** and the parameters of the model to the autocorrelation.
- ❑ The ACF is a function of the population parameters that defines the relationship between the autocorrelations of a process and lag length.

Still on Lags...

- ❑ The standard practice when working with non-stationary / seasonal data is to conduct model selection over **two sets of lags** by choosing a maximum lag to capture the seasonal dynamics and by selecting a maximum lag to capture non-seasonal ones

On time series models dealing with conditional variances

- ❑ Dealing with non-normal /non-linear distributions and ultra-high frequency data. - the availability of more and better data and the availability of low-cost high-performance computers allowed the development of a vast family of ARCH/GARCH models originally developed by Robert F. Engle (1982) [in his “Autoregressive conditional heteroskedasticity with estimates of the variance of United Kingdom inflation”. *Econometrical* 50, 4: 987–1007]
- ❑ While forecasting of expected returns perhaps still remains a rather elusive task, analyzing and predicting the level of uncertainty and the strength of co-movements between asset returns has become a fundamental pillar of financial econometrics.

- ❑ Rationale:
 - Precision analysis: More precise analysis and forecasting.
 - Towards finding steady investment strategies.
 - Determining the usefulness and reliability of trading strategies.
 - Increased usage where volatility of returns is a key issue, e.g. with electronic trading comes massive amount of data – measurement of intraday risk and discovery of trading profit opportunities
 - There are phenomena that exist at some time horizon and disappear at other time horizon.

Nonlinear models for Financial time series analysis....

- ❑ Autoregressive models (**AR**)
- ❑ Integrated models (**I**)
- ❑ The moving average models (**MA**)
 - **ARMA**: Autoregressive moving average
 - **ARIMA**: Autoregressive integrated moving average. In time series analysis, ARIMA model is a generalization of ARMA model. These models are fitted to time series data set either to better understand the data or to forecast – i.e. to predict future points in the series. They are applied in some cases where data show evidence of non-stationarity, where an initial differencing step (corresponding to the "integrated" part of the model) can be introduced to reduce the non-stationarity.
 - **ARCH**: Autoregressive conditional heteroskedasticity
 - **GARCH**: Generalized autoregressive conditional heteroskedasticity
 - **TARCH** – Threshold autoregressive conditional heteroskedasticity

- ❑ Markov Switching
Autoregression (MSAR)
- ❑ Threshold Autoregression (TAR)
- ❑ Self-Exciting Threshold
Autoregression (SETAR).

On time-series models dealing with conditional variances

<i>S/No</i>	<i>Model</i>	<i>Rationale</i>
1	ARCH	Models volatility; deals with time-varying variances (heteroskedasticity) that depend on lagged effects (autocorrelation) commonly observed in many financial market variables.
2	GARCH	Captures long-lagged effects of fewer parameters. This fits many data series well.
3	TARCH	Treats positive and negative news asymmetrically.
4	ANN	Artificial Neural Networks – dealing with large, complex, non-linear or interconnected data sets that are hard to solve using conventional approaches

Models for time series data (continuation)

- ❑ **Error Correction Model (ECM)** is a theoretically-based time series models that directly estimates the speed at which a dependent variable (Y) returns to equilibrium after a change in an independent variable (X). ECMs are useful for estimating both short-term and long-term effects of one time series on another when dealing with stationary data and co-integrated data.
- ❑ The **Vector Autoregression (VAR)** is used to capture the linear interdependencies among multiple time series VAR models generalize the univariate autoregressive (AR model) by allowing for more than one evolving variable.
- A **VAR model** describes the evolution of a set of k variables (called endogenous *variables*) over the same sample period ($t = 1 \dots T$) as a linear function of only their past values. The variables are collected in a $k \times 1$ vector y_t , which has as the i^{th} element, $y_{i,t}$ the observation at time "t" of the i^{th} variable. For example, if the i^{th} variable is GDP, then $y_{i,t}$ is the value of GDP at time t .

The Box-Jenkins Methodology

- ❑ The Box-Jenkins methodology is one of the widely used approaches in financial econometrics
- ❑ Has two aspects:
 - **Identification** – visual inspection of the series – the autocorrelations and partial autocorrelation
 - **Estimation** – candidate models are identified by relating the sample autocorrelations and partial autocorrelations to the autocorrelation function and partial autocorrelation function of the ARMA models

- ❑ **The BJ procedure relies on two principles:**
 - i. **Parsimony** – the specification with the fewest parameter capable of capturing the dynamics of a time series is preferred to other representations equally capable of capturing the same dynamics.
 - ii. **Invertibility** – a technical requirement stemming from the use of the autocorrelogram and partial autocorrelogram to choose a model – achieving a unique identification of the moving average component of a model.

Granger causality test

- ❑ In time series analysis, you would like to know whether changes in a variable will have an impact on changes other variables.
- ❑ Granger causality test is used to address such phenomena more accurately.

Granger causality, yes but...

- ❑ ***Granger causality*** does not equal to what is usually meant by causality.
- ❑ Even if A does not ‘cause’ B, (in the ordinary sense of the word ‘cause’), it may still help to predict B, and thus Granger-causes B if changes in A precedes that of B for various reasons.

Granger Causality continuation...

- ❑ In principle, if X causes Y , then, changes of X happened first then followed by changes of Y .
- ❑ If X causes Y , there are two conditions to be satisfied:
 1. X can help in predicting Y . Regression of X on Y has a big R^2
 2. Y can not help in predicting X .

Granger Causality continuation...

- ❑ In the context of two variables, x and y , y is said to Granger-cause x if current or lagged values of y helps to predict future values of x .
- ❑ On the other hand, y fails to Granger-cause x if for all $s > 0$, the mean squared error of a forecast of x_{t+s} based on (x_t, x_{t-1}, \dots) is the same as that is based on (y_t, y_{t-1}, \dots) and (x_t, x_{t-1}, \dots) .

***Understanding causality test:
Two broad possibilities***

- ❑ Unidirectional causality
- ❑ Bidirectional causality
- Uni-directionality: That all millionaires were persistent, hardworking people does not make hard workers become millionaires (does it?); plenty of failed entrepreneurs were persistent, hard working people.
- Similarly, risk-taking is necessary for large success, but it is also necessary for failure. [Nassim Nicholas Taleb, 2004].

Causality test

NOTE: 4 possible specific outcomes...

1. X Granger causes Y but Y does not Granger cause X
2. Y Granger causes X but X does not Granger cause Y
3. X Granger causes Y and Y Granger causes X (i.e., there is a feedback system)
4. X does not Granger cause Y and Y does not Granger cause X

Granger causality testing procedure



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*Granger causality testing procedure:
Three conditions (Kumar, 2015)*

1. Establish correlation first.
2. There must be issue of timing – the independent variable must have changed for the dependent variable to react
3. Third or other factors must be isolated.

Granger causality testing process...

- Stationary and non-stationary variables
- Spurious regressions
- Unit root tests - for non-stationarity
- Cointegration
- When there is no cointegration – what do you do?
- Granger's causality test

Order of integration of the variables

- Note that all variables have to be of the same order of integration; the following are possible cases (Hatemi, 2004):
 1. *All the variables are $I(0)$ (stationary):* one is in the standard case, i.e. a VAR in level.
 2. All the variables are $I(d)$ (non-stationary) with $d > 0$
 3. **The variables are co-integrated: the error correction term has to be included in the VAR. The model becomes a Vector Error Correction Model (VECM).**
 4. The variables are not co-integrated: the variables have first to be differenced d times and one has a VAR in difference.

Granger causality testing process continuation...

STEPS	PURPOSE
Graphs	Visualization to 'see' if there is any possibility of correlation to begin with.
Unit Root Test (Augmented Dickey Fuller method is commonly used)	Testing for non-stationarity to ensure the validity of empirical results. Note: We can test for non-stationarity by testing the null hypothesis that $\rho = 1$ against the alternative that $ \rho < 1$ (Kumar, 2015)
Co-integration (Johansen System Cointegration test is commonly used)	Testing for short-run relationship between two moving variables. The testing statistic is $\tau = \hat{\varphi} / se(\hat{\varphi})$
Vector Error Correction Model (VECM)	Fitting an error correction model if co-integration is established - to check whether error correction mechanism takes place if some disturbance comes in the equilibrium relationship, i.e. to measure the speed of convergence to the long-run steady state of equilibrium.
Granger Causality	Establishes presence of causality <u>and</u> its direction – i.e. to examine if the correlation coefficients have causal relationship

A word about 'Cointegration'

- If the two series are cointegrated, it means that the spot prices and futures have a long-term relationship, that prevents them from wandering apart without bound (Chris Books, 2008).

What Unit Root Test is all about

- A primer

What do we mean by 'unit root' and 'unit root test' (URT)?

- ❑ A **URT** is a statistical test for the idea or proposition that in a autoregressive statistical model of a time series, the autoregressive parameter is one.
- ❑ A unit root is an attribute of a statistical model of a time series whose autoregressive parameter is one.
- ❑ *In $y_t = y_{t-1} + ut$, where $-1 \leq p \leq 1$, and ut is a white noise error term, if p is in fact 1, then we face what is called the unit root problem, i.e. a situation of non-stationarity (a non-stationary stochastic process).*

Checking for Unit Root in time series data...

- ❑ Recall that if $p = 1$, we have a unit root problem, i.e. a non-stationary stochastic process.
- ❑ We check for presence of unit root by regressing y_t on its lagged value y_{t-1} to establish whether the calculated p is equal to 1.
- ❑ If $p = 1$, then y_t (i.e. the time series under analysis) is non-stationary.
- ❑ Note: If $p = 0$, then y_t is stationary (i.e. the first differences of a random walk time series are stationary).

*Obtaining the right order of integration:
Unit Root Test (URT) for stationarity*

- ❑ Unit root processes are generalizations of the classic random walk – described mathematically earlier
- ❑ A process is said to have a unit root if the distribute lag polynomial can be factored so that one of the roots is exactly one.
- ❑ Many economic / financial time series have roots close to 1; thus, it is important to maximize the power of a unit root test so that models possess the right order of integration.

Types of Unit Root Test

1. Dickey-Fuller (**DF**)
2. Augmented Dickey-Fuller (**ADF**)
3. Phillip Perron (**PP**) – modification of ADF

- Note:** In time series analysis, we have to specify which model of the three URT models we wish to use.
- We also need to specify the number of lagged dependent variables to be included in the model in order to correct the presence of serial correlation.

Unit Root Test of Stationarity

Dickey–Fuller test

- ❑ The most popular URT is the Dickey–Fuller test - You want to test whether ρ is equal to one or significantly less than one
- ❑ The AR(1) process $y_t = \rho y_{t-1} + v_t$ is stationary when $|\rho| < 1$
- ❑ But, when $\rho = 1$, it becomes the non-stationary random walk process

Unit Root Test (URT) of stationarity

Dickey–Fuller method...

- The most popular URT is the **Augmented Dickey–Fuller** (ADF) test - You want to test whether ρ is equal to one or significantly less than one
- **NOTE:** The AR(1) process $y_t = \rho y_{t-1} + v_t$ is stationary when $|\rho| < 1$
- But, when $\rho = 1$, it becomes the non-stationary random walk process

Unit Root Test of stationarity

Dickey–Fuller test

- ❑ The most popular URT is the Dickey–Fuller test - You want to test whether ρ is equal to one or significantly less than one
- ❑ The AR(1) process $y_t = \rho y_{t-1} + v_t$ is stationary when $|\rho| < 1$
- ❑ But, when $\rho = 1$, it becomes the non-stationary random walk process

DF Test: The hypotheses

□ Consider the AR(1) model:

$$y_t = \rho y_{t-1} + v_t$$

We can test for non-stationarity by testing the null hypothesis that $\rho = 1$ against the alternative that $|\rho| < 1$ (Kumar, 2015).

DF Test: The hypotheses continuation...

□ An alternative format is:

$$y_t - y_{t-1} = \rho y_{t-1} - y_{t-1} + v_t$$

$$\Delta y_t = (\rho - 1) y_{t-1} + v_t$$

$$= \gamma y_{t-1} + v_t$$

– The hypotheses are:

$$H_0 : \rho = 1 \quad \Leftrightarrow \quad H_0 : \gamma = 0$$

$$H_1 : \rho < 1 \quad \Leftrightarrow \quad H_1 : \gamma < 0$$

The Dickey-Fuller testing procedure

- ❑ First plot the time series of the variable and select a suitable Dickey-Fuller test based on a visual inspection of the plot
- ❑ If the series appears to be wandering or fluctuating around a sample average of zero, use test equation **(a)**
- ❑ If the series appears to be wandering or fluctuating around a sample average which is non-zero, use test equation **(b)**
- ❑ If the series appears to be wandering or fluctuating around a linear trend, use test equation **(c)**

DF Test with constant but no trend

- The second Dickey–Fuller test includes a constant term in the test equation:

(b)
$$\Delta y_t = \alpha + \gamma y_{t-1} + v_t$$

The null and alternative hypotheses are the same as before

- The third Dickey–Fuller test includes a constant and a trend in the test equation:

(c)
$$\Delta y_t = \alpha + \gamma y_{t-1} + \lambda t + v_t$$

–The null and alternative hypotheses are

– $H_0: \gamma = 0$ and $H_1: \gamma < 0$

DF Test Critical value

- To test the hypothesis in all three cases, we simply estimate the test equation by least squares and examine the t -statistic for the hypothesis that

$$\gamma = 0$$

- Unfortunately this t -statistic no longer has the t -distribution
- Instead, we use the statistic often called a τ (tau) statistic (Kumar, 2015).

Critical values (Kumar, 2015)

To carry out a one-tail test of significance, if τ_c is the critical value obtained from Table, we reject the null hypothesis of non-stationarity if $\tau \leq \tau_c$
If $\tau > \tau_c$ then we do not reject the null hypothesis that the series is non-stationary

Model	1%	5%	10%
$\Delta y_t = \gamma y_{t-1} + v_t$	-2.56	-1.94	-1.62
$\Delta y_t = \alpha + \gamma y_{t-1} + v_t$	-3.43	-2.86	-2.57
$\Delta y_t = \alpha + \lambda t + \gamma y_{t-1} + v_t$	-3.96	-3.41	-3.13
Standard critical values	-2.33	-1.65	-1.28

Note: These critical values are taken from R. Davidson and J. G. MacKinnon (1993), *Estimation and Inference in Econometrics*, New York: Oxford University Press, p. 708.

Mathematics of unit root test (Kumar, 2015)

- Hence, testing for a unit root is equivalent to testing $\phi=1$ in the following model

$$ADF \text{ test equation : } Y_t = \phi Y_{t-1} + \sum_{j=1}^{p-1} \varphi_j \Delta Y_{t-j} + \theta_0 + a_t$$

$$\Delta Y_t = \underbrace{(\phi - 1)}_{\delta} Y_{t-1} + \sum_{j=1}^{p-1} \varphi_j \Delta Y_{t-j} + \theta_0 + a_t$$

$$ADF \text{ test equation : } \Delta Y_t = \delta Y_{t-1} + \sum_{j=1}^{p-1} \varphi_j \Delta Y_{t-j} + \theta_0 + a_t$$

$$H_0 : \phi = 1$$

$$H_0 : \delta = 0$$

$$H_1 : |\phi| < 1$$

$$H_1 : \delta < 0$$

Reject H_0 if $t_{\theta=1} < CV$

Reject H_0 if $t_{\delta=0} < CV$

Usage of Granger test: A brief Illustration (Kumar, 2015)

- ❑ **World Oil Price and Growth of US Economy – a study by James Hamilton** (using 91 – 95 observations)
- ❑ **The enquiry:** Does the increase of world oil price influence the growth of US economy or does the growth of US economy affect the world oil price?
 - There are two causalities that need to be observed:
 - (i) H_0 : Growth of US Economy does not influence world oil price
 - (ii) H_0 : World oil price does not influence growth of US Economy
 - **James Hamilton's F Tests Results:**
 1. Hypothesis that US economy does not effect world oil price is not rejected. It means that the US economy does not have effect on world oil price.
 2. Hypothesis that world oil price does not influence US economy is rejected. It means that the world oil price does influence US economy.

Chicken vs. Egg

Which first came?



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Chicken vs. Egg

- ❑ Thurman and Fisher (1988) using yearly data of chicken population and egg productions in the USA from 1930 to 1983 concludes that:
 1. Egg causes the chicken.
 2. There is no evidence that chicken causes egg.
- ❑ Hypothesis that egg has no effect on chicken population is thus rejected; while the other hypothesis that chicken has no effect on egg is not rejected.
- ❑ Why?

On chicken vs. egg continuation...

- ❑ Thurman and Fisher (1988) using yearly data of chicken population and egg productions in the USA from 1930 to 1983 concludes that:
 1. Egg causes the chicken.
 2. There is no evidence that chicken causes egg.
- ❑ Hypothesis that egg has no effect on chicken population is thus rejected; while the other hypothesis that chicken has no effect on egg is not rejected.
- ❑ Why?

***Further application:
Efficient Market Hypothesis (EMH)***

- ❑ The ***Efficient Market Hypothesis (EMH)***, which suggests that *returns of a stock-market are unpredictable from historical price changes*, is satisfied when stock prices are characterized by a random walk (***unit root***) process.
- ❑ A finding of unit root implies that stock returns cannot be predicted (Munir *et al*, 2012)

*****Applications: Further perspectives...

- ❑ If two variables are co-integrated, in the long-run these two variables will have a common trend. If we have market efficient efficiency, co-integration must exist in the futures and spot market. This means that the futures prices will not be consistently above or below the spot prices (see Hakkio & Mark, 1989, in Market efficiency and co-integration: An application to the sterling and Deutschmark exchange markets, *Journal of International Money and Finance*, 8, pp. 75-88)
- ❑ If co-integration exists between two variables, this means that the futures prices can be used to predict spot prices. (see Granger, 1986, in Developments in the study of co-integrated economic variables, *Oxford Bulletin of Economics and Statistics*, 48, pp. 213 - 228)

Applications: Further perspectives...

- The next few slides provide further causality illustrations provided by Professor Kumar (2015) based on this enquiry:
 - *Does the US economy influence Australia economy or does the Australia economy influence the US economy?*

Unit Root Test - ADF @ level: Example using E-Views
(Kumar, 2015)

The screenshot shows the EViews software interface. A data series window for 'AUS' is open, displaying a time series of values from 1 to 20. A 'Unit Root Test' dialog box is overlaid on the series window, configured for an Augmented Dickey-Fuller test at the level.

Series: AUS Workfile: COPY OF GDP::Untitled

Year	Value
1	38.2355
2	38.7551
3	38.7706
4	38.8948
5	39.5621
6	39.6402
7	40.8614
8	40.6741
9	40.2642
10	41.1712
11	40.9545
12	41.4527
13	42.2458
14	42.5829
15	42.8018
16	44.1979
17	44.2659
18	42.9056
19	43.6730
20	

Unit Root Test

- Test type: Augmented Dickey-Fuller
- Test for unit root in:
 - Level
 - 1st difference
 - 2nd difference
- Include in test equation:
 - Intercept
 - Trend and intercept
 - None
- Lag length:
 - Automatic selection: Schwarz Info Criterion
 - Maximum lags: 12
 - User specified: 4

Buttons: OK, Cancel

Path = \\staff\home\kkumar\documents DB = none WF = copy of gdp

Unit Root Test - ADF @ level:

Continuation - E-Views output (Kumar, 2015)

The screenshot displays the EViews software interface. The main window shows a workfile titled 'COPY OF GDP' with a range of 1 to 124 observations. A series named 'AUS' is selected. A dialog box titled 'Series: AUS' is open, showing the results of an Augmented Dickey-Fuller Unit Root Test. The test results indicate that the null hypothesis of a unit root is not rejected at the 1%, 5%, or 10% significance levels.

Workfile: COPY OF GDP - (\\staff\home\kkumar\documents\copy of gdp....

View Proc Object Save Freeze Details+/- Show Fetch Store Delete Genr Sample

Range: 1 124 -- 124 obs Filter: *
Sample: 1 124 -- 124 obs Order: Name

aus
c
resid
usa

Series: AUS Workfile: COPY OF GDP::Untitled\

View Proc Object Properties Print Name Freeze Sample Genr Sheet Graph Stats I

Augmented Dickey-Fuller Unit Root Test on AUS

Null Hypothesis: AUS has a unit root
Exogenous: Constant
Lag Length: 0 (Automatic - based on SIC, maxlag=12)

	t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic	2.977928	1.0000
Test critical values:		
1% level	-3.484198	
5% level	-2.885051	
10% level	-2.579386	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation
Dependent Variable: D(AUS)
Method: Least Squares
Date: 08/03/15 Time: 14:04
Sample (adjusted): 2 124
Included observations: 123 after adjustments

Path = \\staff\home\kkumar\documents DB = none WF = copy of gdp

Unit Root Test - ADF @ 1st difference:

Continuation - E-Views output (Kumar, 2015)

Workfile: COPY OF GDP - (\\staff\home\kkumar\documents\copy of gdp....

View Proc Object Save Freeze Details+/- Show Fetch Store Delete Genr Sample

Range: 1 124 -- 124 obs Filter: *
Sample: 1 124 -- 124 obs Order: Name

aus
c
resid
usa

Series: AUS Workfile: COPY OF GDP::Untitled\
View Proc Object Properties Print Name Freeze Sample Genr Sheet Graph Stats I

Augmented Dickey-Fuller Unit Root Test on D(AUS)

Null Hypothesis: D(AUS) has a unit root
Exogenous: Constant
Lag Length: 0 (Automatic - based on SIC, maxlag=12)

	t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic	-9.435297	0.0000
Test critical values:		
1% level	-3.484653	
5% level	-2.885249	
10% level	-2.579491	

*Mackinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation
Dependent Variable: D(AUS,2)
Method: Least Squares
Date: 08/03/15 Time: 14:06
Sample (adjusted): 3 124
Included observations: 122 after adjustments

Path = \\staff\home\kkumar\documents DB = none WF = copy of gdp

Stationarity test continuation (Kumar, 2015)

Regression model	1%	5%	10%
(1) $y_t = \beta x_t + e_t$	-3.39	-2.76	-2.45
(2) $y_t = \beta_1 + \beta_2 x_t + e_t$	-3.96	-3.37	-3.07
(3) $y_t = \beta_1 + \delta t + \beta_2 x_t + e_t$	-3.98	-3.42	-3.13

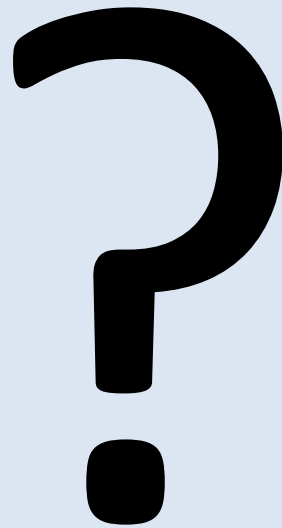
Note: These critical values are taken from J. Hamilton (1994), *Time Series Analysis*, Princeton University Press, p. 766.

Some information:

The Phillips-Perron (PP) unit root tests

- ❑ The Phillips-Perron (PP) unit root tests developed by Phillips and Perron (1988) are similar to ADF tests.
- ❑ Kumar (2015) suggests that the PP unit root tests differ from the ADF tests mainly in complexity and how they deal with serial correlation and heteroskedasticity in the errors.
- ❑ Notably, where the ADF tests use a parametric autoregression to approximate the ARMA structure of the errors in the test regression, the PP tests ignore any serial correlation in the test regression.
- ❑ **The PP tests usually give the same conclusions as the ADF tests** (Kumar, 2015).

Next... how to conduct the
Cointegration test?



Co-integration test...

- Conventionally, non-stationary time-series variables should not be used in regression models to avoid the problem of spurious regression
 - There is an exception to this rule when $e_t = y_t - \beta_1 - \beta_2 x_t$ is a stationary $I(0)$ process
- In this case y_t and x_t are said to be **co-integrated**, i.e. y_t and x_t share similar stochastic trends, and, since the difference e_t is stationary, they never diverge too far from each other (Kumar, 2015).

Johansen Cointegration Test: E-Views window(Kumar, 2015) – PRESS OK!

The screenshot displays the EViews software interface. In the background, a data window titled 'Group: UNTITLED' shows a time series for 'AUS' and 'USA' from 1970 to 2000. The 'Johansen Cointegration Test' dialog box is open in the foreground, showing the 'Cointegration Test Specification' tab. The dialog includes options for deterministic trend assumptions, lag intervals, and critical values.

Group: UNTITLED Workfile: COPY OF GDP::Untitled

	AUS	USA
1	38.2355	38.3011
2	38.7551	38.3734
3	38.7706	38.7137
4	38.8948	38.2991
5	39.5621	39.3615
6	39.6402	39.5836
7	40.8614	39.8973
8	40.6741	40.0114
9	40.2642	40.7224
10	41.1712	41.6840
11	40.9545	42.0813
12	41.4527	42.7699
13	42.2458	43.8558
14	42.5829	44.3631
15	42.8018	44.1267
16	44.1979	44.5485
17	44.2659	44.1624
18	42.9056	44.2897
19	43.6730	43.8609
20	43.6216	43.6887
21	43.7671	43.1662
22	45.1109	43.4819
23	44.8353	44.2184
24	44.8000	44.7000

Johansen Cointegration Test

Cointegration Test Specification

Deterministic trend assumption of test

Assume no deterministic trend in data:

- 1) No intercept or trend in CE or test VAR
- 2) Intercept (no trend) in CE - no intercept in VAR

Allow for linear deterministic trend in data:

- 3) Intercept (no trend) in CE and test VAR
- 4) Intercept and trend in CE - no intercept in VAR

Allow for quadratic deterministic trend in data:

- 5) Intercept and trend in CE - intercept in VAR

Summary:

- 6) Summarize all 5 sets of assumptions

* Critical values may not be valid with exogenous variables; do not include C or Trend.

Exog variables*

Lag intervals

1 4

Lag spec for differenced endogenous

Critical Values

- MHM
- Osterwald-Lenum

Size 0.05

OK Cancel

Path = \\staff\home\kkumar\documents DB = none WF = copy of gdp

Johansen Cointegration Test:

E-Views output (Kumar, 2015) – At most 1 cointegration observed

The screenshot displays the EViews software interface. The main window shows the 'Johansen Cointegration Test' results for a group named 'UNTITLED'. The test was performed on 08/03/15 at 14:13, using a sample of 6 to 124 observations (adjusted from 124). The included observations are 119 after adjustments, with a linear deterministic trend assumption. The series are AUS and USA, with a lags interval of 1 to 4 in first differences.

The results are presented in two tables: the Unrestricted Cointegration Rank Test (Trace) and the Unrestricted Cointegration Rank Test (Maximum Eigenvalue). Both tests indicate no cointegration at the 0.05 level, as the p-values (0.0833 and 0.2816 respectively) are greater than the significance level. The asterisk (*) denotes rejection of the hypothesis at the 0.05 level, which is not the case here.

Unrestricted Cointegration Rank Test (Trace)

Hypothesized No. of CE(s)	Eigenvalue	Trace Statistic	0.05 Critical Value	Prob.**
None	0.073266	13.98836	15.49471	0.0833
At most 1 *	0.040613	4.933864	3.841466	0.0263

Trace test indicates no cointegration at the 0.05 level
 * denotes rejection of the hypothesis at the 0.05 level
 **MacKinnon-Haug-Michelis (1999) p-values

Unrestricted Cointegration Rank Test (Maximum Eigenvalue)

Hypothesized No. of CE(s)	Eigenvalue	Max-Eigen Statistic	0.05 Critical Value	Prob.**
None	0.073266	9.054500	14.26460	0.2816

Path = \\staff\home\kkumar\documents DB = none WF = copy of gdp

Next... how to conduct further tests using

VECM?

BAZE
UNIVERSITY
ABUJA



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Recall:

Order of integration of the variables....

- Note that all variables have to be of the same order of integration; the following are possible cases (Hatemi, 2004):
 1. *All the variables are $I(0)$ (stationary):* one is in the standard case, i.e. a VAR in level.
 2. All the variables are $I(d)$ (non-stationary) with $d > 0$
 3. **The variables are co-integrated: the error correction term has to be included in the VAR. The model becomes a Vector Error Correction Model (VECM).**
 4. The variables are not co-integrated: the variables have first to be differenced d times and one has a VAR in difference.

Granger causality – VAR Equation (Kumar, 2015)

- In the VAR equation, the example we proposed above implies a lower triangular coefficient matrix:

$$\begin{bmatrix} x_t \\ y_t \end{bmatrix} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} + \begin{bmatrix} \phi_{11}^1 & 0 \\ \phi_{21}^1 & \phi_{22}^1 \end{bmatrix} \begin{bmatrix} x_{t-1} \\ y_{t-1} \end{bmatrix} + \dots + \begin{bmatrix} \phi_{11}^p & 0 \\ \phi_{21}^p & \phi_{22}^p \end{bmatrix} \begin{bmatrix} x_{t-p} \\ y_{t-p} \end{bmatrix} + \begin{bmatrix} a_{1t} \\ a_{2t} \end{bmatrix}$$

Or if we use MA representations,

$$\begin{bmatrix} x_t \\ y_t \end{bmatrix} = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix} + \begin{bmatrix} \Psi_{11}(B) & 0 \\ \Psi_{21}(B) & \Psi_{22}(B) \end{bmatrix} \begin{bmatrix} a_{1t} \\ a_{2t} \end{bmatrix}$$

where $\Psi_{ij}(B) = \phi_{ij}^0 + \phi_{ij}^1 B + \phi_{ij}^2 B^2 + \dots$, $\phi_{11}^0 = \phi_{22}^0 = 1$, $\phi_{21}^0 = 0$.

General procedure for testing Granger Causality... (Kumar, 2015)

- 1) Check that both series are stationary in mean, variance and covariance (if necessary transform the data via logs, differences to ensure this)
- 2) Estimate $AR(p)$ models for each series, where p is large enough to ensure white noise residuals. F tests and other criteria can be used to establish the maximum lag p that is needed.
- 3) Re-estimate both model, now including all the lags of the other variable
- 4) Use F tests to determine whether, after controlling for past Y , past values of X can improve forecasts Y (and vice versa)

*Testing Granger causality - continuation:
(Kumar, 2015)*

i. The simplest test is to estimate the regression which is based on

$$x_t = c_1 + \sum_{i=0}^p \alpha_i x_{t-i} + \sum_{j=1}^p \beta_j y_{t-j} + u_t$$

using **OLS** and then conduct a F -test of the null hypothesis

$$H_0 : \beta_1 = \beta_2 = \dots = \beta_p = 0.$$

Testing Granger causality continuation...

ii. Run the following regression, and calculate RSS (full model)

$$x_t = c_1 + \sum_{i=0}^p \alpha_i x_{t-i} + \sum_{j=1}^p \beta_j y_{t-j} + u_t$$

iii. Run the following limited regression, and calculate RSS (Restricted model).

$$x_t = c_1 + \sum_{i=0}^p \alpha_i x_{t-i} + u_t$$

Testing Granger causality continuation...

iv. Do the following F -test using RSS obtained from stages 2 and 3:

$$F = \left[\frac{(n-k)}{q} \cdot \frac{(RSS_{restricted} - RSS_{full})}{RSS_{full}} \right]$$

n : number of observations

k : number of parameters from full model

q : number of parameters from restricted model

TESTING GRANGER CAUSALITY

(Kumar, 2015)

v. If H_0 rejected, then X causes Y .

□ Granger test can be used in investigating whether or not Y causes X .

Workfile: COPY OF GDP - (\\staff\ho

View Proc Object Save Freeze Details

Range: 1 124 -- 124 obs

Sample: 1 124 -- 124 obs

 aus
 c
 resid
 usa

G Group: UNTITLED Workfile: COPY OF GDP::Untitled\

View Proc Object Print Name Freeze Default Sort Edit+/- Smpl+/- Compare+/-

- Group Members
- Spreadsheet
- Dated Data Table
- Graph...

- Descriptive Stats ▶
- Covariance Analysis...
- N-Way Tabulation...
- Tests of Equality...
- Principal Components...

- Correlogram (1) ...
- Cross Correlation (2) ...
- Long-run Covariance...
- Unit Root Test...
- Cointegration Test ▶
- Granger Causality...**

- Label

USA

011

734

137

991

615

836

973

114

224

840

813

699

558

631

267

485

624

897

609

887

662

819

184

 ◀ ▶ **Untitled** New Page

Workfile: COPY OF GDP - (\\staff\ho

View Proc Object Save Freeze Details

Range: 1 124 -- 124 obs

Sample: 1 124 -- 124 obs

- aus
- c
- resid
- usa

Untitled New Page

G Group: UNTITLED Workfile: COPY OF GDP::Untitled\

View Proc Object Print Name Freeze Sample Sheet Stats Spec

Pairwise Granger Causality Tests

Date: 08/03/15 Time: 14:21

Sample: 1 124

Lags: 4

Null Hypothesis:	Obs	F-Statistic	Prob.
USA does not Granger Cause AUS	120	4.50841	0.0021
AUS does not Granger Cause USA		0.94253	0.4422

Hands-on training session

Exploring some applications using
EViews econometric software

Note: some helpful technology hints
(EViews software)

For what?	Where to go on the EViews package
i. Graph Options	Select VIEW
ii. Unit Root Test	Select VIEW
iii. Cointegration Test	Select VIEW
iv. Vector Error Correction	Select PROC
v. Granger Causality	Select VIEW

***Beyond analytics:
Some important workplace skills...***



- ✓ Written & oral communication
- ✓ Problem-solving abilities – analytical reasoning
- ✓ Creative thinking
- ✓ Interpersonal skills (team work)
- ✓ Time management
- ✓ General professionalism, personal effectiveness, and work ethics

Some helpful technology hints (EViews software)

For what?	Where to go on the EViews package
i. Graphs	Select VIEW
ii. Unit Root test	Select VIEW
iii. Cointegration test	Select VIEW
iv. Vector Error Correction	Select PROC
v. Granger Causality	Select VIEW

Granger causality analysis

Now, let's test these pairs of time series...

1. Econodata file_1: DGR & GDP growth in Nigeria
2. Econodata file_2: Agriculture & GDP in Nigeria
3. Econodata file_3: Crude oil price and growth in Nigeria
4. Econodata file_4: Infosys Ltd & NSE Nifty - India
5. Econodata file_5: Agriculture & GDP in India
6. Econodata file_6: India economic growth & Global growth

*Econodata file_1:
DGR & GDP growth in Nigeria*



Situation analysis of Debt-to-GDP Ratio (DGR) and GDP in Nigeria?

Causality analysis

Experimenting with Case Nigeria...

- ❑ **CONTEXT:** The assumption that the developing countries can borrow with expectation of economic growth is open to question.
- ❑ It's controversial – Government needs to borrow to fulfil huge developmental goals for their citizens, but if government becomes a dominant debtor in a financial system, there is concern that the private sector may become 'growth at the end.
- ❑ Besides the economic implications and associated debate on the subject, the degree of stability or volatility of government's fiscal policies such as Debt-to-GDP ratio (DGR) will have remarkable influence on business performance because **companies are not immuned from the macroeconomic environment in which they operate.**

Formulate your hypotheses

- H_1 : GDP growth rate has a unit root.
- H_2 : DGR has a unit root.
- H_3 : There is no co-integration between GDP and DGR.
- H_4 : GDP growth rate does not Granger-cause DGR.
- H_5 : DGR does not Granger-cause GDP growth rate.

DGR-GDP growth causality test: ***An application of the Granger methodology...***

□ **Data:** (i) **GDP growth rate** was used as the proxy for economic growth- time series from 1981 to 2014; i.e. 34 years of secondary data sourced primarily from The World Bank and the Central bank of Nigeria (CBN). (ii) **Debt** means ‘total debt stock’, i.e. it includes long-term and short-term domestic and foreign liabilities.

□ **Econometrics with *EViews* Software:**

Graphics

ADF

JCiT

VECM

Granger

Next...

Upload your data for analysis



Nigeria DGR-GDP growth time series - Microsoft Excel

File Home Insert Page Layout Formulas Data Review View

Clipboard Font Alignment Number Styles Cells Editing

V37

Nigeria: Debt-to-GDP ratio (DGR) and GDP growth rate, 1981 - 2014

Year	DGR	GDP Growth %
1981	46.47	-13.13
1982	61.78	-1.05
1983	128.38	-5.05
1984	188.05	-2.02
1985	186.94	8.32
1986	171.17	-8.75
1987	168.90	-10.75
1988	159.92	7.54
1989	150.77	6.47
1990	124.15	12.77
1991	136.70	-0.62
1992	135.38	0.43
1993	129.33	2.09
1994	104.14	0.91
1995	51.87	-0.31
1996	32.83	5.0
1997	31.36	2.80
1998	34.40	2.72
1999	141.55	0.47
2000	68.58	5.32
2001	67.03	4.41
2002	49.54	3.78
2003	47.35	10.35
2004	37.12	33.74
2005	19.47	3.44
2006	8.06	8.21
2007	8.22	6.83
2008	7.29	6.27
2009	10.52	6.93
2010	10.03	7.84
2011	10.89	4.65
2012	11.03	6.75
2013	11.32	7.31
2014	10.15	6.83

Sheet1 Sheet2 Sheet3

Ready 59% 09:33 27/11/2015

*First, follow the commands
to upload your data into the EViews ...*



❑ Click-open your *EViews* –
FILE-OPEN-FOREIGN DATA
AS WORK FILE-locate your
file in MS Excel-OPEN-
NEXT-FINISH

You're ready!

Your EViews now opened...



EViews

File Edit Object View Proc Quick Options Window Help



Welcome to EViews

Path = c:\users\stephen aro gordon\documents DB = none WF = none

07:27 01/12/2015

Next, open your time series data



The screenshot shows the EViews software interface with an 'Open' dialog box. The dialog is set to the 'Documents' folder. The file list shows several folders, and the file 'Nigeria DGR-GDP growth time series.xlsx' is selected in the 'File name' field. The 'File type' is set to 'All files (*.*)'. The 'Open' button is highlighted.

Date modified	Type
01/12/2014 ...	File folder
29/08/2015 ...	File folder
20/04/2015 ...	File folder
01/12/2014 ...	File folder
01/12/2014 ...	File folder
19/09/2014 ...	File folder
01/12/2014 ...	File folder
07/07/2015 ...	File folder
01/12/2014 ...	File folder

File name: Nigeria DGR-GDP growth time series.xlsx

All files (*.*)

Open Cancel

Next,

Obtain the time series' graphical visuals

- Highlight the two time series (DGR&GDP growth)-VIEW-OPEN SELECTED-ONE WINDOW-OPEN GROUP (to display the two time series)-VIEW(again)-GRAPH-GRAPH OPTIONS-LINE&SYMBOL-PRESS OK.

EViews graphic visualization example

Remember to highlight / select the two time series...



Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES - (c... - □ ×

View Proc Object Print Save Details+/- Show Fetch Store Delete Genr Sample

Range: 1981 2014 -- 34 obs Filter: *

Sample: 1981 2014 -- 34 obs

- c
- dgr
- gdp_growth__
- resid
- year

< > Untitled New Page



The two time series displayed...



EViews
File Edit Object View Proc Quick Options Window Help

Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES - (c... - □ ×

View: [G] Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWT... - □ ×

View	Proc	Object	Print	Name	Freeze	Default	Sort	Transpose	Edit+/-	Smpl
obs		DGR		GDP_GRO...						
1981		46.47		-13.13						
1982		61.78		-1.05						
1983		128.38		-5.05						
1984		188.05		-2.02						
1985		186.94		8.32						
1986		171.17		-8.75						
1987		168.90		-10.75						
1988		159.92		7.54						
1989		150.77		6.47						
1990		124.15		12.77						
1991		136.70		-0.62						
1992		135.38		0.43						
1993		129.33		2.09						
1994		104.14		0.91						
1995		51.87		-0.31						
1996		32.83		4.99						
1997		31.36		2.80						
1998		34.40		2.72						
1999		141.55		0.47						
2000		68.58		5.32						
2001		67.03		4.41						
2002		49.54		3.78						
2003		47.35		10.35						
2004		37.12		33.74						
2005										

Selecting your graph options...

Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES - (c... - □ ×

View: [G] Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWT... - □ ×

Graph Options

Option Pages

- Graph Type
 - Basic type**
- Frame & Size
- Axes & Scaling
- Legend
- Graph Elements
- Quick Fonts
- Templates & Objects

Graph type

General:
Basic graph

Specific:
Line & Symbol
Bar
Spike
Area
Area Band
Mixed with Lines
Dot Plot
Error Bar
High-Low (Open-Close)
Scatter
XY Line
XY Area
Pie
Distribution
Quantile - Quantile
Boxplot

Details

Graph data: Raw data

Orientation: Normal - obs axis on bottom

Axis borders: None

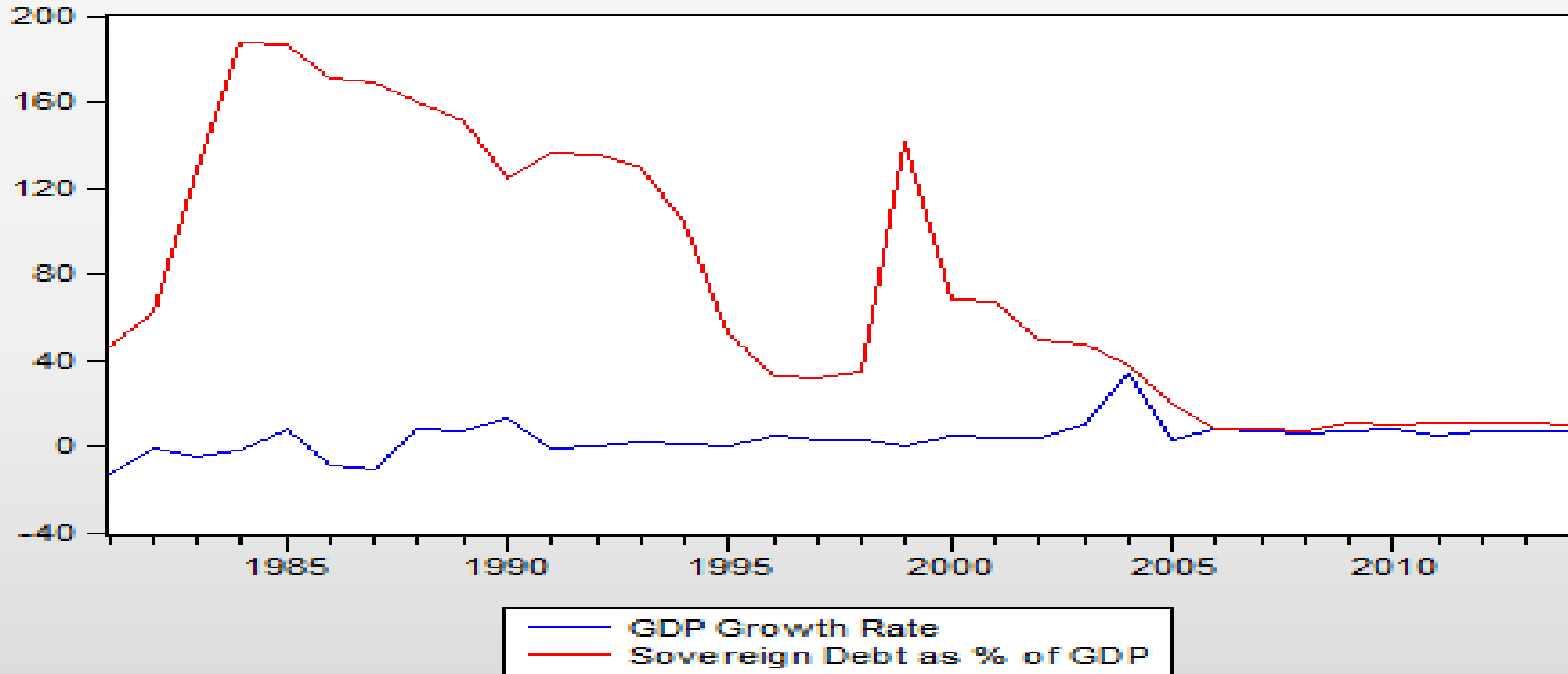
Multiple series: Single graph

Undo Page Edits

OK Cancel

2005 < > ..

EViews graphic output



Next, perform the Unit Root Test for each series
i.e. individually – let's start with DGR...

1. AT LEVEL

- HIGHLIGHT/SELECT THE TIME SERIES (DGR/GDP GROWTH-**VIEW**-OPEN SELECTED-**VIEW** (again)-UNIT ROOT TEST-AUGMENTED DICKEY-FULLER-**LEVEL**-PRESS OK

2. FIRST DIFFERENCE

- HIGHLIGHT/SELECT THE TIME SERIES (DGR/GDP GROWTH-**VIEW**-OPEN SELECTED-**VIEW** (again)-UNIT ROOT TEST-AUGMENTED DICKEY-FULLER-**1ST DIFFERENCE**-PRESS OK (*check for stationarity: $p \approx 0$*)

EViews window - at level (default)



EViews
File Edit Object View Proc Quick Options Window Help

Series: DGR Workfile: NIGERIA DGR-GDP GROWTH TIM... - x

View Proc Object Properties Print Name Freeze Default Sort Edit+/- Smpl+

DGR

Unit Root Test

Test type: Augmented Dickey-Fuller

Test for unit root in:

- Level
- 1st difference
- 2nd difference

Lag length:

- Automatic selection: Schwarz Info Criterion
- User specified: 1

Maximum lags: 8

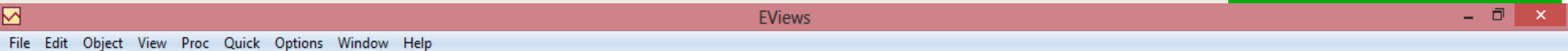
Include in test equation:

- Intercept
- Trend and intercept
- None

OK Cancel

1981				
1982				
1983				
1984				
1985				
1986				
1987				
1988				
1989				
1990				
1991				
1992				
1993				
1994				
1995				
1996				
1997		31.36		
1998		34.40		
1999		141.55		
2000				

EViews window – URT output for DGR at level



Series: DGR Workfile: NIGERIA DGR-GDP GROWTH TIM... - □ ×

View Proc Object Properties Print Name Freeze Sample Genr Sheet Graph Stats |

Augmented Dickey-Fuller Unit Root Test on DGR

Null Hypothesis: DGR has a unit root
Exogenous: Constant
Lag Length: 0 (Automatic - based on SIC, maxlag=8)

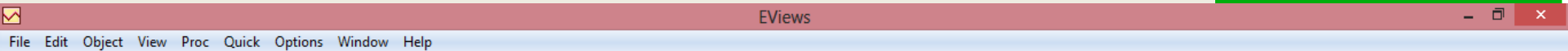
	t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic	-1.262839	0.6348
Test critical values:		
1% level	-3.646342	
5% level	-2.954021	
10% level	-2.615817	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation
Dependent Variable: D(DGR)
Method: Least Squares
Date: 12/01/15 Time: 07:42
Sample (adjusted): 1982 2014
Included observations: 33 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DGR(-1)	-0.110410	0.087430	-1.262839	0.2161
C	7.432881	8.626774	0.861606	0.3955

EViews window – URT output for DGR at first difference



Series: DGR Workfile: NIGERIA DGR-GDP GROWTH TIM... - x

View Proc Object Properties Print Name Freeze Sample Genr Sheet Graph Stats |

Augmented Dickey-Fuller Unit Root Test on D(DGR)

Null Hypothesis: D(DGR) has a unit root
Exogenous: Constant
Lag Length: 0 (Automatic - based on SIC, maxlag=8)

	t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic	-5.414109	0.0001
Test critical values:		
1% level	-3.653730	
5% level	-2.957110	
10% level	-2.617434	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation
Dependent Variable: D(DGR,2)
Method: Least Squares
Date: 12/01/15 Time: 07:44
Sample (adjusted): 1983 2014
Included observations: 32 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
D(DGR(-1))	-0.983901	0.181729	-5.414109	0.0000
C	-1.595753	5.653770	-0.282246	0.7797

DGR-GDP growth unit root test

Summary of results

Particulars	DGR			GDP			
	t-statistic	Critical Value	P-value	t-statistic	Critical Value	P-value	
At level	-1.2628	1%	-3.6463	-4.6899	1%	-3.6463	0.0006
		5%	-2.9540		5%	-2.9540	
		10%	-2.6158		10%	-2.6158	
At first difference	-5.4141	1%	-3.6537	-8.5980	1%	-3.6537	0.0000
		5%	-2.9571		5%	-2.9571	
		10%	-2.6174		10%	-2.6174	

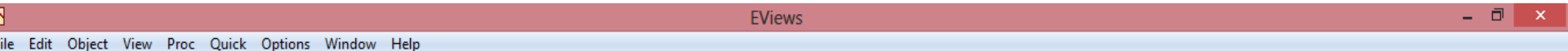
*An important word before
performing cointegration test...*

- ❑ In some situations, it may be desirable to determine the short-run dynamics or long-run dynamics exclusively.
- ❑ For instance, in technical analysis (as opposed to fundamental analysis) asset prices are believed to be long-run unpredictable but may have some short- or medium-run predictability

Next,
Perform Cointegration test...

□ Highlight the two time series
(DGR&GDP growth)-VIEW-OPEN
SELECTED-ONE WINDOW-OPEN
GROUP (to display the two time
series)-VIEW(again)-COINTEGRATION
TEST-JOHANSEN SYSTEM
COINTEGRATION-PRESS OK.

EViews windows now opened for Johansen system Cointegration Test



Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES - (c... - □ ×

View Proc Object Print Name Freeze Default Sort Transpose Edit+/- Smp1

Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWT... - □ ×

- Group Members
- Spreadsheet
- Dated Data Table
- Graph...
- Descriptive Stats
- Covariance Analysis...
- N-Way Tabulation...
- Tests of Equality...
- Principal Components...
- Correlogram (1) ...
- Cross Correlation (2) ...
- Long-run Covariance...
- Unit Root Test...
- Cointegration Test
 - Johansen System Cointegration Test...
 - Single-Equation Cointegration Test...
- Granger Causality...
- Label

Object Name	Freeze	Default	Sort	Transpose	Edit+/-	Smp1
Group Members	0...					
Spreadsheet	.13					
Dated Data Table	.05					
Graph...	.05					
Descriptive Stats	.02					
Covariance Analysis...	.32					
N-Way Tabulation...	.75					
Tests of Equality...	.75					
Principal Components...	.54					
Correlogram (1)47					
Cross Correlation (2)77					
Long-run Covariance...	.62					
Unit Root Test...	.43					
Cointegration Test	.09					
Granger Causality...	.91					
Label	.31					
	.99					
	.80					
	.72					
	.78					
	.35					
2004	37 12	33 74				
2005	<	>				

Johansen system Cointegration Test EViews Output

EViews

File Edit Object View Proc Quick Options Window Help

Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWT...

View Proc Object Print Name Freeze Sample Sheet Stats Spec

Johansen Cointegration Test

Date: 12/01/15 Time: 07:49
 Sample (adjusted): 1983 2014
 Included observations: 32 after adjustments
 Trend assumption: Linear deterministic trend
 Series: DGR GDP_GROWTH_
 Lags interval (in first differences): 1 to 1

Unrestricted Cointegration Rank Test (Trace)

Hypothesized No. of CE(s)	Eigenvalue	Trace Statistic	0.05 Critical Value	Prob.**
None *	0.439511	20.00285	15.49471	0.0098
At most 1	0.045094	1.476571	3.841466	0.2243

Trace test indicates 1 cointegrating eqn(s) at the 0.05 level
 * denotes rejection of the hypothesis at the 0.05 level
 **Mackinnon-Haug-Michelis (1999) p-values

Unrestricted Cointegration Rank Test (Maximum Eigenvalue)

Hypothesized No. of CE(s)	Eigenvalue	Max-Eigen Statistic	0.05 Critical Value	Prob.**
None *	0.439511	18.52628	14.26460	0.0100
At most 1	0.045094	1.476571	3.841466	0.2243

Workfile: Untitled

- View Proc Object
- Range: 1983 2014
- Sample: 1983 2014
- c
- dgr
- gdp_grow
- resid
- year

Path = c:\users\stephen aro gordon\documents DB = none WF = nigeria dgr-gdp growth time series

Johansen system Cointegration Test Summary results

Level	Eigen Value	Trace Statistic	Critical Value at 5%	P-values
$H_0: r = 0$ (none)*	0.4395	20.0028	15.4947	0.0098
$H_1: r = 1$ (at most 1)	0.0451	1.4766	3.8415	0.2243

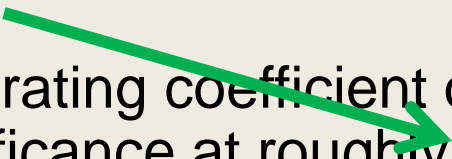
Johansen system Cointegration Test

Summary results - continuation

- ❑ Scroll down to the bottom of the EVIEWS output where you have
 - ALPHA
 - COEFFICIENTS
- ❑ Observe: The estimated co-integrating coefficient for the GDP growth is as follows:

$$LGDP = -7.7045 - 0.1134DGR$$

[0.08]



- ❑ The t-statistic (standard error) of the co-integrating coefficient of DGR is given in the bracket suggesting significance at roughly 10%
- ❑ ***The coefficient for DGR is negative, at least in the short-run, which means that increase in DGR can be associated with decline in the country's economic growth.***

Next,

Perform Vector Error Correction test...

- Highlight the two time series (DGR&GDP growth)-VIEW-OPEN
SELECTED-ONE WINDOW-PROC-
MAKE VECTOR AUTOCORRECTION-
VECTOR ERROR CORRECTION--
PRESS OK.

EViews windows now opened for

Vector Error Correction Test



EViews
File Edit Object View Proc Quick Options Window Help

Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES - (c:...

Group: UNTITLED

View	Proc	Object	Print	...
obs		DGR	G	
1981		46.47		
1982		61.78		
1983		128.38		
1984		188.05		
1985		186.94		
1986		171.17		
1987		168.90		
1988		159.92		
1989		150.77		
1990		124.15		
1991		136.70		
1992		135.38		
1993		129.33		
1994		104.14		
1995		51.87		
1996		32.83		
1997		31.36		
1998		34.40		
1999		141.55		
2000		68.58	5.32	
2001		67.03	4.41	
2002		49.54	3.78	
2003		47.35	10.35	
2004		37.12	33.74	
2005				

VAR Specification

Basics | Cointegration | VEC Restrictions

VAR Type

Unrestricted VAR

Vector Error Correction

Endogenous Variables

dgr gdp_growth__

Estimation Sample

1981 2014

Lag Intervals for Endogenous: 1, 2

1 2

Exogenous Variables

Do NOT include C or Trend in VEC's

OK Cancel

Vector Error Correction Test

EViews Output

Workfile: NIGERIA DGR-GDP GROWT...
View Proc Object
Range: 1981 2014
Sample: 1981 2014

- c
- dgr
- gdp_growth
- resid
- year

Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWT...
Var: UNTITLED Workfile: NIGERIA DGR-GDP GROWT...
View Proc Object Print Name Freeze Estimate Stats Impulse Resids

Vector Error Correction Estimates

Date: 12/01/15 Time: 07:53
Sample (adjusted): 1984 2014
Included observations: 31 after adjustments
Standard errors in () & t-statistics in []

Cointegrating Eq:		CointEq1	
DGR(-1)		1.000000	
GDP_GROWTH__(-1)		17.11125 (3.86788) [4.42394]	
C		-154.8679	

Error Correction:	D(DGR)	D(GDP_GR...
CointEq1	-0.113375 (0.08142) [-1.39254]	-0.054897 (0.01971) [-2.78568]
D(DGR(-1))	-0.043143 (0.17706) [-0.24367]	0.038671 (0.04286) [0.90231]

Summary Results



Document6 (Autosaved) - Microsoft Word

Table Tools

File Home Insert Page Layout References Mailings Review View Design Layout

Clipboard Font Paragraph Styles Editing

Vector Error Correction Estimates

CoIntegrating Eq:	CoIntEq1	
DGR(-1)	1	
GDP_GROWTH_(-1)	17.11128060891794	3.867876205049176
	[4.42394]	
C	154.8678962559113	-
Error Correction:	D(DGR)	D(GDP_GROWTH_(-1))
CoIntEq1	0.113375130923563 7	0.054897476067876 78
	0.081415955540611 55	0.019707018739284 58
	[-1.39254] [-2.78568]	
D(DGR(-1))	0.043143045338128 78	0.038671071441652 21
	0.177033554583280 3	0.042857545583742 08
	[-0.24367] [0.90231]	
D(DGR(-2))	0.078149665113711 98	0.006627627807399 84
	0.178021358210124 8	0.04309096591785 06
	[0.43899] [-0.15381]	
D(GDP_GROWTH_(-1))	1.000290430940833 84	0.075953511218838 64
	1.120216515418953 [0.89294]	0.271152351989411 [0.28011]
D(GDP_GROWTH_(-2))	0.446394634469309 2	0.042828010582574 31
	0.791264644775412 5	0.191528393416586 5
	[0.58415] [-0.22381]	
C	4.350778297965078 9	0.445577670840113
	5.596058445934629	1.354544925148543

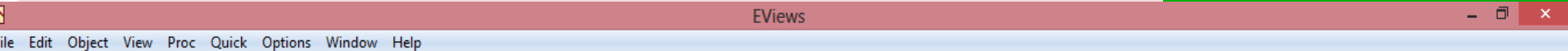
VECM tests - Findings

- **Note:** t-statistics in []
- **Scroll through the output / carefully observe:** In all cases $t < 1.96$ (alpha)
- **Decision:** Null Hypothesis is accepted – this means that there may be no long-run cointegration between DGR and GDP growth rate.

Finally,
Perform your Granger Causality test...

□ Highlight the two time series
(DGR&GDP growth)-VIEW-OPEN
SELECTED-ONE WINDOW-OPEN
GROUP (to display the two time
series)-VIEW(again)-GRANGER
CAUSALITY-LAG SPECIFICATION(2)-
PRESS OK.

Opening the EViews windows for Granger causality test operation



The screenshot displays the EViews software interface. On the left, a "Workfile: NIGERIA DGR-GDP GROWT..." window is open, showing a list of objects: c, dgr, gdp_growth, resid, and year. The "View" tab is selected. In the center, a "Group: UNTITLED" window is open, showing a list of menu options. The "Granger Causality..." option is highlighted. The menu options include: Group Members, Spreadsheet, Dated Data Table, Graph..., Descriptive Stats, Covariance Analysis..., N-Way Tabulation..., Tests of Equality..., Principal Components..., Correlogram (1) ..., Cross Correlation (2) ..., Long-run Covariance..., Unit Root Test..., Cointegration Test, Granger Causality..., and Label. The "Granger Causality..." option is highlighted in blue.



Microsoft Word ribbon showing font and paragraph options. The 'Normal' style is highlighted in a yellow box.

EViews - [Group: UNTITLED Workfile: NIGERIA DGR-GDP GROWTH TIME SERIES::Untitled]

EViews software menu bar: File Edit Object View Proc Quick Options Window Help

EViews software sub-menu bar: View Proc Object Print Name Freeze Sample Sheet Stats Spec

Pairwise Granger Causality Tests
Date: 12/03/15 Time: 12:26
Sample: 1981 2014
Lags: 2

Null Hypothesis:	Obs	F-Statistic	Prob.
GDP_GROWTH__ does not Granger Cause DGR	32	1.91626	0.1666
DGR does not Granger Cause GDP_GROWTH__		0.89783	0.4193

DGR and GDP Growth in Nigeria: Results of Granger Causality Test

Null Hypotheses	Observations	F-Statistic	Probability	Decision
GDP growth does not Granger-cause DGR	26	1.91626	0.1666	Accept Ho
DGR does not Granger-cause GDP growth	26	0.89783	0.4193	Accept Ho

Summary of findings

- i. Both DGR and GDP are stationary based on Augmented Dickey Fuller (ADF) test.
- ii. The trace test under Johansen co-integration method indicates one co-integrating equation at 5 percent level of significance.
- iii. From the VECM result, it is evident that DGR has significant long-run negative impact on economic growth of Nigerian economy. The long-run negative relationship between DGR and GDP growth tested statistically significant (approx. 10%) by a negative coefficient of DGR.
- iv. The Granger causality test results showed weak nexus between DGR and GDP. This indicates that there is no guarantee that high level of DGR significantly will lead to a slow-down in the economy and vice-versa as previously observed in a similar analysis (Shehu, 2006).

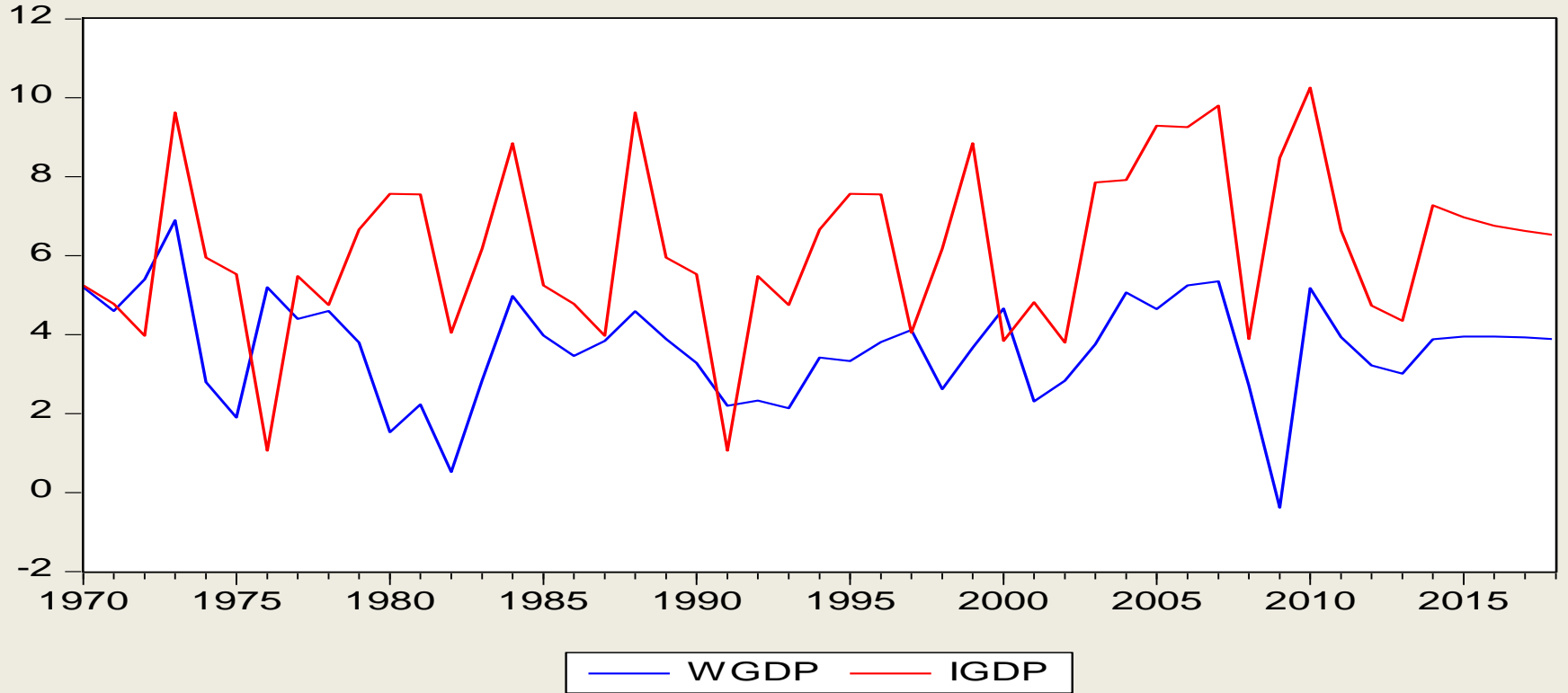
Insights / implications

- ❑ Government may need to revisit its mind-set for increasing debt stock so that it can create more value sustainably.
- ❑ Stop piling up debts; it may not necessarily grow your economy.
- ❑ Rather, promote efficiency by sealing wastages / leakages; diversify your economy by promoting SMEs across sectors, quality education, rebuilding fiscal buffers, external reserves, introducing investor-friendly policies, could provide better options for achieving macroeconomic stability, sustainable and inclusive growth.

Further application ...

India & Global growth: any causal nexus?

(Econodata File_6)



December 21, 2015 @ PES University, Bangalore, India

EViews applications: Know your limits!

- Plurality of models often with similar explanatory ability.
 - Analysts' fixation on numbers – non or inadequate discussion of economic / management importance / implications of statistical results.
 - Qualitative facts may not be immediately reflected in numbers.
 - “Not all problems have solutions” – Microsoft
 - It is not a mistake to use logic without statistics; logic does not need empirical verification
- [Nassim Nicholas Taleb (2004) in his *Fooled by Randomness*, Penguin Books]

Keywords

- ADF, ANNs, ARCH, Autoregression, heteroskedasticity, Bias, Computer Science, Correlation, Econometrics, Descriptive statistics, DF, Financial modeling, Granger causality, EMH, Hypothesis testing, Inferential statistics, Johansen system cointegration test, Macroeconomic model, Mathematics, MSAR, Time series, GARCH, TARARCH, Sample period, SETAR, Regression, Technology, Unit root test, VAR, VECM.

Concluding remarks

- ❑ Social science research issues are wide-ranging. The *EViews* software offers a versatile tool for intelligent application mathematical, statistical, and computer science techniques to facilitate innovation and smart policies for solving myriads of problems confronting the society.
- ❑ Time series analysis is an increasingly widely used econometric tool supported by technology to help in gaining faster, top quality, and useful insights from observational data on diverse phenomena, not just on management and social science issues, hence, the imperative for protocol /procedure compliance in conducting the requisite statistical trials was stressed.
- ❑ In this presentation, hands-on, practical *EViews* applications in finance and macroeconomic problems showed the capacity of technology to make things relatively easy for the budding social scientist and practitioner. The need to adopt the intelligent parsimony principles in making sense out of the numerous computer outputs was emphasized.

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Caveat / feedback ...

- All the materials used in this presentation are solely for educational purpose, advancement of knowledge and improvement of educational practices to move the society forward. The author will be pleased to make good any omissions or rectify any mistakes brought to his attention at the earliest possible opportunity.