

#### REFERENCES

Abel, A.B. 1979. Investment and the Value of Capital. New York: Garland Publishing Company.

\_\_\_\_\_. 1980. Empirical Investment Equations: An Integrative Framework. Carnegie Rochester Conference Series on Public Policy, 12: 39-91.

Allen, C.B. 1997. A Survey of Flexible Functional Forms Applied to Production Analysis. In Allen, C.B. and Hall, S.G. (Eds.), *Macroeconomic Modelling in a Changing World: Towards a Common Approach*. Chichester: John Wiley and Sons.

\_\_\_\_\_. 1997. Modelling Investment with Adjustment Costs Using Flexible Functional. In Allen, C.B. and Hall, S.G. (Eds.), *Macroeconomic Modelling in a Changing World: Towards a Common Approach*. Chichester: John Wiley and Sons.

\_\_\_\_\_. 1997. A Supply-Side Model of the UK Economy: An Application of Non-Linear Cointegration. In Allen, C.B. and Hall, S.G. (Eds.), *Macroeconomic Modelling in a Changing World: Towards a Common Approach*. Chichester: John Wiley and Sons.

Allen, C.B. and Nixon, J. 1997. Two concepts of the NAIRU. In Allen, C.B. and Hall, S.G. (Eds.), *Macroeconomic Modelling in a Changing World: Towards a Common Approach*. Chichester: John Wiley and Sons.

Allen, C.B. and Robertson, D. 1990. Recovering the Aggregate Production Functions: Supply-Side Modelling of Three Industrialised Economies. London Business School, Centre for Economic Forecasting Discussion Paper, No. 28-90.

Allen, C.B. and Urga, G. 1995. Derivation and Estimation of Interrelated Factor Demands from a Dynamic Cost Function. London Business School, Centre for Economic Forecasting Discussion Paper, No. 10-95.

Allen, C., Hall, S. G. and Nixon, J. 1994. *The New London Business School model of the UK economy*. London Business School, Centre for Economic Forecasting Discussion Paper, No. 18-94.

Anon. 1979. Why Supply-Side Economics is Suddenly Popular. *Business Week*, September 17, 1979.

Anon. 1980. A Guide to Understanding the Supply-Siders. Business Week, December 22, 1980.

Arrow, K.J., Chenery, H.B., Minhas, B.S. and Solow, R.M. 1961. Capital-labour Substitution and Economic Efficiency. *Review of Economics and Statistics*, 43: 225-50.



Banerjee, A., Dolado, J.J., Gailbraith, J.W. and Hendry, D.F. 1993. Cointegration, Error Correction, and the Econometric Analysis of Non-Stationary Data. Oxford: Oxford University Press.

Barker, F.S. 1992. The South African Labour Market - Critical Issues for Transition. Pretoria: J.L van Schaik.

Bernanke, B. and Gertler, M. 1996. Inside the Black Box: The Credit Channel of Monetary Policy Transmission. *Journal of Economic Perspectives*, 9(4): 27-48.

Berndt, E.R. 1991. The Practice of Econometrics: Classic and Contemporary. New York: Addison-Wesley Publishing Company.

Berndt, E.R. and Christensen, L.R. 1973. The Internal Structure of Functional Relationships: Separability, Substitution, and Aggregation. Review of Economic Studies, 40: 402-10.

\_\_\_\_\_ 1973. The Translog Function and The Substitution of Equipment, Structures and Labor in US Manufacturing. *Journal of Econometrics*, 18: 235-52.

Berndt, E.R. and Wood, D.O. 1975. Technology, Prices and the Derived Demand for Energy. Review of Economics and Statistics, 57: 259-68.

Bodkin, R.G., Klein, L.R. and Marwah, K. 1991. A History of Macroeconometric Model Building. Aldershot: Edward Elgar.

Boskin, M. 1982. An Empirical Evaluation of Supply-Side Economics. In Supply-Side Economics in the 1980's: Conference Proceedings. Westport: Quorum Books.

Boskin, M. and Lau, L.J. 1978. Taxation, Social Security, and Aggregate Factor Supply in the United States. In *Compendium of Tax Research*. U.S. Treasury, Office of Tax Analysis, Washington.

Bruno, M. and Sachs, J.D. 1985. The Economics of Worldwide Stagflation. Oxford: Blackwell Publishers.

Budd, A., Dicks, G., Holly, S., Keating, G. and Robinson, B. 1984. The London Business School Econometric Model of the UK. *Economic Modelling*, 1(4)

Burda, M. and Wyplosz, C. 1993. *Macroeconomics: A European Text*. Oxford: Oxford University Press.



Bureau for Economic Policy and Analysis (BEPA), 1998. Improving the Labour Absorption Capacity of the South African Economy. Bepa Economic Paper, No. 29. Pretoria: University of Pretoria.

\_\_\_\_\_. 1999. The Potential for Improving the Labour Absorption Capacity of SMMEs through the Institutionalisation of the Informal Sector. Bepa Economic Paper, No. 30. Pretoria: University of Pretoria.

Burgess, D.F. 1974. A Cost Minimization Approach to Import Demand Functions. Review of Economics and Statistics, 56: 234-55.

\_\_\_\_\_ 1975, Duality Theory and Pitfalls in the Specification of Technologies. *Journal of Econometrics*, 3: 105-121.

Burrows, L. and Smit, B.W. 1999. Estimating Potential Output and Output Gaps for the South African Economy. Paper presented at the Biennial Conference of the Economic Society of South Africa, September 1999, Pretoria.

Butler, I. 1996. A Semi-Structural Method to Estimate Potential Output: Combining Economic Theory with a Time-Series Filter. The Bank of Canada's New Quarterly Projection Model, Part 4. Technical Report, No. 77. Ottawa: Bank of Canada.

Canova, F. 1993. Detrending and Business Cycle Facts. CEPR Discussion Paper, No. 782.

Charemza, W.W. and Deadman, D.F. 1997. New Directions in Econometric Practice. Cheltenham: Edward Elgar.

Christensen, L.R., Jorgenson, D.W. and Lau, L.J. 1973. Transcendental Logarithmic Production Frontiers. Review of Economics and Statistics, 55: 28-45.

Chung, J.W. 1994. Utility and Production Functions. Oxford: Blackwell Publishers.

Cobb, C.W. and Douglas, P.H. 1928. A Theory of Production. American Economic Review, 18: 139-65.

Coen, R.M. and Hickman, B.G. 1995. A Stochastic Model of Potential Output. Paper presented at Project LINK, September 1995, Pretoria.

Conway, P. and Hunt, B. 1997. Estimating Potential Output: A Semi-Structural Approach. Discussion Paper, No. G97/9. Wellington: Bank of New Zealand.

Côté, D. and Hostland, D. 1996. An Econometric Examination of the Trend Unemployment Rate in Canada, Working Paper, No. 96-7. Ottawa: Bank of Canada.



Gapinski, J.H. 1997. Economic Growth in the Asian-Pacific Region. Paper presented at Project LINK, September 1997, Kuala Lumpur.

Giorno, C., Richardson, P. Roseveare, D. and Van den Noord, P. 1995. Estimating Potential Output, Output Gaps and Structural Balances. *OECD Economics Department Working Papers*. No. 152. Paris: Organisation for Economic Co-operation and Development (OECD).

Golden, S. 1980. Super Star of the New Economics. New York Times Magazine, March 23, 1980.

Gramley, L.E. 1980. Supply-side Economics: Its role during Inflation. *Voice*, Federal Reserve Bank of Dallas, November 1980.

Greene, W.H. 1995. Econometric Analysis New Jersey: Prentice-Hall International.

Griliches, Z. and Ringstad, V. 1971. Economies of Scale and the Form of the Production Function. Amsterdam: North-Holland.

Grunfeld, Y. 1960. The Determinants of Corporate Investment. In Harberger, C. (Ed.), The Demand for Durable Goods. Chicago: University of Chicago Press.

Gurley, J.G. and Shaw, E.S. 1955. Financial Aspects of Economic Development. American Economic Review, 45(4): 515-38.

1960, Money in a Theory of Finance. Washington: Brookings Institution.

Hailstones, T.J. 1982. A Guide to Supply-Side Economics. New York: Robert F. Dame.

Hall, R.E. 1991. Labour Demand. Labour Supply, and Employment Volatility. In Blanchard, O.J. and Fischer, S. (Eds.), NBER Macroeconomics Annual 1991. Cambridge: MIT Press.

Hall, R.E. and Jones, C.I. 1996. The Productivity of Nations. *NBER Working Paper Series*, No. 5812. Cambridge: NBER.

Hall, S.G. 1995. Macroeconomics and a Bit More Reality. The Economic Journal, July. 974-988.

1997. Modelling Economic Policy Responses. Paper presented at Project LINK, September 1997, Kuala Lumpur.

\_\_\_\_\_. 1997. Macroeconomic Modelling: A Perspective In Allen, C.B. and Hall, S.G. (Eds.), Macroeconomic Modelling in a Changing World: Towards a Common Approach, Chichester: John Wiley and Sons.

Hall, S.G and Henry, S.G.B. 1985. Macroeconomic Modelling. Amsterdam: North-Holland



Hall, S.G. and Nixon, J. 1997. Modelling Monetary Policy in the 1990s. In Allen, C.B. and Hall, S.G. (Eds.), *Macroeconomic Modelling in a Changing World: Towards a Common Approach*. Chichester: John Wiley and Sons.

\_\_\_\_\_ 1997. Modelling Wages and the Supply-Side of the UK Economy. Paper presented at Project LINK, September 1997, Kuala Lumpur.

Hamilton, J.D. 1994. Time Series Analysis. Princeton: Princeton University Press.

Hanoch, G. 1975. Production and Demand Models with Direct or Indirect Activity. Econometrica, 43(May): 395-420.

Harris, R.I.D. 1995. Using Cointegration Analysis in Econometric Modelling. New Jersey: Prentice-Hall International.

Hayashi, F. 1982. Tobin's Marginal q and Average q: A Neoclassical Interpretation. Review of Economic Studies, 37: 499-516.

Heathfield, D.F. 1971. Production Functions. London: MacMillan Education.

Heathfield, D.F. and Wibe, S. 1987. An Introduction to Cost and Production Functions. London: MacMillan Education.

Helliwell, J.F. 1986. Supply-Side Macro-Economics. The Canadian Journal of Economics, 19: 597-625.

\_\_\_\_\_. 1995. Modelling the Supply Side: What are the Lessons from Recent Research on Growth and Globalization? Paper presented at Project LINK, September 1995, Pretoria.

Helliwell, J. Sturm, P., Jarrett, P. and Salou, G. 1985. Aggregate Supply in Interlink: Model Specification and Empirical Results. *OECD Department of Economics and Statistics Working Paper*, No. 26. Paris: Organisation for Economic Co-operation and Development (OECD).

Hubbard, R.G. 1998. Capital-Market Imperfections and Investment Journal of Economic Literature, 36: 193-225.

Ireland, P.N. 1994. Supply-Side Economics and Endogenous Growth. *Journal of Monetary Economics*, 33: 559-571.

Janssen, R.F. 1980. IMF Sally Into Supply-Side Economics Raises Eyebrows and Risks of Friction. Wall Street Journal, August 5, 1980.



Jiang, Y. and Pauly, P. 1997. International Growth Convergence with Heterogeneous Technology. Paper presented at the Second Annual Conference on Econometric Modelling for Africa, September 1997, Pretoria.

Jorgenson, D.W. 1963. Capital Theory and Investment Behavior. American Economic Review, 53(2): 247-249.

\_\_\_\_\_ 1971. Econometric Studies of Investment Behavior: A Survey Journal of Economic Literature, 9(4): 1111-1147.

Jorgenson, D.W. and Siebert, C.D. 1968. A Comparison of Alternative Theories of Corporate Investment Behavior. *American Economic Review*, 58(4): 681-712.

Keleher, R.E. 1982. Evidence Relating to Supply-Side Tax Policy. In Fink, R.H. (Ed.), Supply-Side Economics: A Critical Appraisal. Maryland: University Publications of America.

Kemp, J. 1982. Supply-Side Economics: An American Renaissance? In Supply-Side Economics in the 1980's: Conference Proceedings. Westport: Quorum Books.

Keynes, J.M. 1935. The Theory of Employment, Interest and Money. London: Harcourt, Brace and Company.

King, M. 1980. Savings and Taxation. In Hughes, G.A. and Heal, G.M. (Eds.), Essays in Public Policy. London: Allen Urwin.

Klein, L.R. 1982. The Supply-Side of the Economy: A View from the Prospective of the Wharton Model. In Fink, R.F. (Ed.), Supply-Side Economics: A Critical Appraisal. Maryland: University Publications of America.

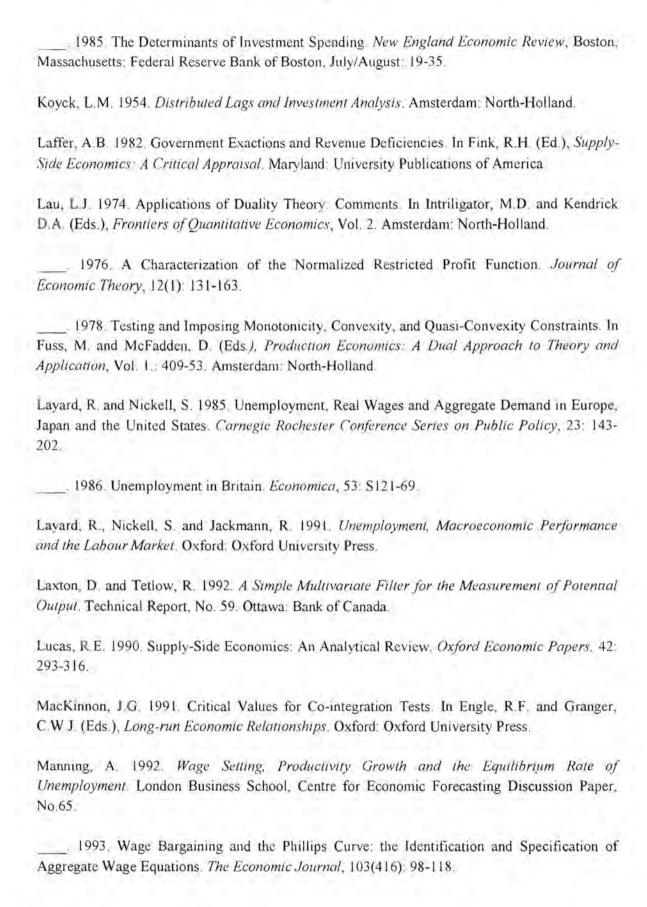
Klein, L.R. and Welfe, W. 1983. Lectures in Econometrics. Amsterdam: North-Holland.

Kmenta, J. 1967. On Estimation of the CES Production Function. *International Economic Review*, 8: 18-93.

Kopcke, R.W. 1977. The Behavior of Investment Spending during the Recession and Recovery, 1973-7. New England Economic Review, Boston, Massachusetts: Federal Reserve Bank of Boston, November/December: 5-41.

\_\_\_\_\_\_1982 Forecasting Investment Spending: The Performance of Statistical Models. New England Economic Review, Boston, Massachusetts: Federal Reserve Bank of Boston, November/December: 13-32.







\_\_\_\_\_. 1995. Development in Labour Market Theory and their Implications for Macroeconomic Policy. Scottish Journal of Political Economy, 42(3): 250-265.

McFadden, D. 1962. Factor Substitutability in the Economic Analysis of Production. Unpublished Ph.D. dissertation, Minneapolis: University of Minnesota.

McFadden, D. 1978. Cost, Revenue, and Profit Functions. In Fuss, M. and McFadden, D. (Eds.). Production Economics: A Dual Approach to Theory and Applications, Vol. 1., 3-110. Amsterdam, North-Holland.

Meyer, J.R. and Kuh, E.E. 1957. The Investment Decision: An Empirical Study. Cambridge. Massachusetts: Harvard University Press.

Modigliani, F. and Miller, M.H. 1958. The Cost of Capital, Corporation Finance, and the Theory of Investment. *American Economic Review*, 48(3): 261-97.

Molinas, C. 1986. A Note on Spurious Regressions with Integrated Moving Average Errors. Oxford Bulletin of Economics and Statistics, 48(3): 279-282.

Nelson, C.R. and Plosser, C. 1982. Trend and Random Walks in Macro-economic Time Series. Journal of Monetary Economics, 10: 139-167.

Nickell, S. 1978. Investment Decisions of Firms. Cambridge: Cambridge University Press.

\_\_\_\_\_\_. 1988. The Supply Side and Macroeconomic Modelling. In Bryant, R.C. et. al. (Eds.), Empirical Macroeconomics for Interdependent Economies. Washington: Brookings Institution: 202-221.

Nicholson, W. 1995. Microeconomic Theory; Basic Principles and Extensions. USA: Harcourt, Brace and Company.

Pencavel, J. 1985. Wages and Employment under Trade Unionism: Microeconomic Models and Macroeconomic Applications. Scandinavian Journal of Economic, 87(2).

Perron, P. 1989. The Great Crash, the Oil Price Shock and the Unit root Hypothesis. *Econometrica*, 57(6): 1361-1401.

Pichelmann, K. and Schuh, A.U. 1997. The NAIRU-Concept: A Few Remarks. *OECD Economics Department Working Papers*. No. 178. Paris: Organisation for Economic Cooperation and Development (OECD).

Pindyck, R.S. and Rubinfeld, D.L. 1991. Econometric Models and Economic Forecasts. New York: McGraw-Hill.



Raboy, D. 1980. Norman B. Turé on Supply-side Economics. Enterprise, June 1980.

Roberts, P.C. 1982. The Breakdown of the Keynesian Model. In Fink, R.H. (Ed.), Supply-Side Economics: A Critical Appraisal. Maryland: University Publications of America.

\_\_\_\_\_ 1982. Theoretical Foundations of Supply-Side Economics. In Supply-Side Economics in the 1980's: Conference Proceedings. Westport: Quorum Books.

Romer, D. 1996. Advanced Macroeconomics. New York: McGraw Hill.

Rose, D.E. 1988. The NAIRU in Canada: Concepts, Determinants and Estimates. Technical Report, No. 50. Ottowa: Bank of Canada.

Say, J.B. 1830. A Treatise on Political Economy. Philadelphia: John Grigg.

Scacciavillani, F. and Swagel, P. 1999. Measures of Potential Output: An Application to Israel. Working Paper, No. 104. Washington: International Monetary Fund (IMF).

Setterfield, M., Gordon, D. and Osberg, L. 1992. Searching for a Will o' the Wisp: An Empirical Study of the NAIRU in Canada. European Economic Review, 36: 119-136.

Shephard, R.W. 1953. Cost and Production functions. Princeton: Princeton University Press.

Smith, A. 1937. The Wealth of Nations. New York: The Modern Library.

Solow, R.M. 1956. The Production Function and the Theory of Capital. Review of Economic Studies, 23: 101-8.

South African Reserve Bank. Quarterly Bulletin, (various issues).

Sowell, T. 1972. Say's Law: A Historical Analysis. Princeton: Princeton University Press.

Stein, H., 1980. Some 'Supply-Side' Propositions. The AEI Economists, April 1980.

Strum, J.E., and De Haan, J. 1995. Is public expenditure really productive? *Economic Modelling*, 12(1): 60-72.

Studenmund, A.H. 1997. *Using Econometrics: A Practical Guide*. New York: Addison-Wesley Publishing Company.

Summers, L.H. 1989. Tax Policy in a Life Cycle Model. NBER Working Paper Series. Cambridge: NBER.



Thomas, R.L. 1993. Introductory Econometrics: Theory and Applications. New York: Longman Publishing.

Thomsen, T. 1998. Links between Short- and Long-Run Factor Demand. *Economic Modelling Working Paper Series*, 1998:2.

Tobin, J. 1969. A General Equilibrium Approach to Monetary Theory. Journal of Money, Credit and Banking, 1(1): 15-29.

Torres, R. and Martin, J.P. 1990. Potential Output in the Seven Major OECD Countries. OECD Economic Studies, No. 14. Paris: Organisation for Economic Co-operation and Development (OECD)

Torres, R. and Jarrett, P. 1987. A Revised Supply Block for the Seven Major Seven Countries in Interlink. *OECD Economics Department Working Papers*, No. 41. Paris: Organisation for Economic Co-operation and Development (OECD).

Torres, R., Jarrett, P. and Suyker, W. 1989. Modelling Business Sector Supply for the Smaller OECD Countries. *Department of Economics and Statistics Working Papers*, No. 71. Paris. Organisation for Economic Co-operation and Development (OECD).

Turé, N.B. 1980. Forecasting the Supply-Side of the Economy Hearings. Testimony before U.S. Congress, Joint Economic Committee, Ninety-sixth Congress, Second Session, May 21, 1980.

\_\_\_\_\_ 1982. The Economic Effects of Tax Changes. A Neoclassical Analysis. In Fink, R.F. (Ed.), Supply-Side Economics: A Critical Appraisal, Maryland: University Publications of America.

Turner, D., Richardson, P. and Rauffet, S. 1993. The Role of Real and Nominal Rigidities in Macroeconomic Adjustment, *OECD Economic Studies*, No. 21. Paris: Organisation for Economic Co-operation and Development (OECD).

\_\_\_\_\_ 1993. Modelling the Supply-Side of the Seven Major OECD Economics. OECD Economics Department Working Papers, No. 167 Paris: Organisation for Economic Cooperation and Development (OECD).

Uzawa, H. 1964. Duality Principles in the Theory of Cost and Production. *International Economic Review*, 5(2): 216-220.

Varian, H.R. 1992. Microeconomic Analysis. New York: W.W. Norton and Company

\_\_\_\_\_. 1996. Intermediate Microeconomics: A Modern Approach. New York: W.W. Norton and Company.



Wales, T.J. 1977. On the Flexibility of Functional Forms. Journal of Econometrics, 43: 183-93.

Wallis, K.F., Fisher, P.G., Longbotton, J.A., Turner, D.S. and Whitley, J.D. 1987. Models of the UK Economy: A Fourth Review by the ESRC Macroeconomic Modelling Bureau. Oxford: Oxford University Press.

Weidenbaum, M. 1982. Theoretical Foundations of a Growth-Oriented Macroeconomy. In Supply-Side Economics in the 1980's: Conference Proceedings. Westport: Quorum Books.

Whitley, J.D. 1994. A Course in Macroeconomic Modelling and Forecasting. Hertfordshire: Harvester Wheatsheaf.

Wojnilower, A.M. 1980. The Central Role of Credit Crunches in Recent Financial History. Brookings Pap. Economic Act., 2: 277-326.

Yoshikawa, H. 1980. On the 'q' Theory of Investment. American Economic Review, 70(4): 739-744.

Young, G. 1996. A New System of Factor Demand Equations for the NIESR Domestic Model. Paper presented at the Macroeconomic Modelling Bureau Seminar, July 1996, University of Warwick.



#### **APPENDIX 1: ESTIMATION TECHNIQUE**

The estimation technique used was the Engle and Yoo (1991) three-step procedure, which is an extension of the Engle-Granger (1987) two-step procedure. This approach consists first, of a simple test for the presence of cointegration between variables, indicating whether a particular set of variables represents a combination that is consistent with a long-run equilibrium relationship.

At the second stage, an ECM is constructed in order to estimate the short-run or dynamic adjustment process to long-run equilibrium. The ECM indicates the dynamic relationship between variables, i.e. fluctuations in the dependent variable around its long-run trend are explained by fluctuations in the explanatory variables around their long-term trends. "The idea is simply that a proportion of the disequilibrium from one period is corrected in the next" (Engle and Granger 1986: 254). Long-term effects are accounted for by the inclusion of the error term (e) of the long-term cointegration relationship in the ECM (Harris 1995: 24).

All the variables in an ECM were transformed (differenced) into stationary variables. A regression of the stationary form of the dependent variable on the other stationary variables and the first lag of the error term of the cointegrating regression was then performed. Other variables besides those included in the cointegrating regression may be considered for inclusion in the ECM. Examples of such variables are those which were discarded from the cointegrating regression, stationary variables (in levels) and variables, which theoretically will only influence the short-run trend of the dependent variable and were therefore also not considered for inclusion in the cointegrating regression. Since all the variables in the ECM are stationary, the assumptions of classical regression analysis are fulfilled. Therefore, standard diagnostic tests can be used to determine which variables should be included in the final specification of the ECM (Harris 1995: 24).

The third step is applied to adjust the coefficients and *t*-statistics so that they are closer to their true values. This ensures that the variables included in the long-run regression could be evaluated statistically. Engle and Yoo (1989) have proposed a "third step" in addition to the Engle and Granger two-step estimation technique, which step is computationally tractable and overcomes two of the disadvantages of the two-step procedure.

This step provides a correction to the parameter estimates of the first stage, static regression that makes them asymptotically equivalent to FIML and provides a set of standard errors, which allows the valid calculation of standard *t*-tests.

The third step consists simply of a further regression of the conditioning variable from the static regression multiplied by minus the error correction parameter, regressed on the errors from the second-stage error correction model. The coefficients from this model are the corrections to the parameter estimates while their standard errors are the relevant standard errors for inference.



The three steps are then:

i) First estimate a standard cointegrating regression of the form:

$$Y_t = \alpha^1 X_t + Z_t$$

where  $Z_t$  is the OLS (ordinary least square) residual to give first-stage estimates of  $\alpha^1$ .

- ii) Then estimate a second-stage dynamic model (ECM) using the residuals from the cointegrating regression to impose the long-run constraint:  $\Delta Y_t = \Phi(L)\Delta Y_{t-1} + \Omega(L)\Delta X_t + \delta Z_{t-1} + \mu_t.$
- iii) The third stage then consists of the regression:  $\mu_t = \varepsilon(-\hat{\delta}X_t) + v_t.$

The correction for the first-stage estimates is then simply:  $\alpha^3 = \alpha^1 + \varepsilon$  and the correct standard errors for  $\alpha^3$  are given by the standard errors for  $\varepsilon$  ( $SE_\varepsilon$ ) in the third-stage regression.

The *t*-values for  $\alpha^3$  are given by:

$$t = \frac{\alpha^3}{SE_{\varepsilon}}$$



#### **APPENDIX 2: SIMULATION ERROR STATISTICS**

Root mean square simulation error:

RMSE = 
$$\sqrt{\frac{1}{T} \sum_{t=1}^{T} (y_t - \hat{y}_t)^2}$$

Mean absolute simulation error:

$$MAE = \frac{1}{T} |y_t - \hat{y}_t|$$

Root mean square percentage error:

$$RMSPE = \sqrt{\frac{1}{T} \sum_{t=1}^{T} \left( \frac{y_t - \hat{y}_t}{y_t} \right)^2}$$

Mean absolute percentage error:

$$MAPE = \frac{1}{T} \sum_{t=1}^{T} \left| \frac{y_t - \hat{y}_t}{y_t} \right|^2$$

Theil inequality coefficient:

$$U = \frac{\sqrt{\frac{1}{T} \sum_{t=1}^{T} (y_t - \hat{y}_t)^2}}{\sqrt{\frac{1}{T} \sum_{t=1}^{T} (\hat{y}_t)^2} \sqrt{\frac{1}{T} \sum_{t=1}^{T} (y_t)^2}},$$

that lies between zero and one, where zero indicates a perfect fit.

The Theil inequality coefficient can be decomposed into three proportions:

Bias, 
$$U^{M} = \frac{(\hat{y} - \bar{y})^{2}}{\frac{1}{T} \sum_{t=1}^{T} (y_{t} - \hat{y}_{t})^{2}}$$
,

measuring the extent to which the average values of the simulated and actual series deviate from each other,

Variance, U<sup>S</sup> = 
$$\frac{(\sigma_{\hat{y}} - \sigma_{y})^{2}}{\frac{1}{T} \sum_{t=1}^{T} (y_{t} - \hat{y}_{t})^{2}}$$
,

indicating the degree of variability in the series, and



Covariance, 
$$U^{C} = \frac{2(1-\rho)\sigma_{\hat{y}}\sigma}{\frac{1}{T}\sum_{t=1}^{T}(y_{t}-\hat{y}_{t})^{2}}$$
,

representing the remaining unsystematic error.

This is based on the fact that the mean squared error can be decomposed as:

$$\frac{1}{T} \sum_{t=1}^{T} (y_t - \hat{y}_t)^2 = (\overline{\hat{y}} - \overline{y})^2 + (\sigma_{\hat{y}} - \sigma_y)^2 + 2(1 - \rho)\sigma_{\hat{y}}\sigma_y$$

where  $\bar{\hat{y}}$ ,  $\bar{y}$ ,  $\sigma_{\hat{y}}$  and  $\sigma_{y}$  are the means and standard deviations of the series  $\hat{y}_{t}$  and  $y_{t}$ , respectively, and

$$\rho = (1/\sigma_y \sigma_y T) \sum_{t=1}^{T} \ (\hat{y}_t - \overline{\hat{y}}) (y_t - \overline{y})$$

is the correlation coefficient (Pindyck and Rubinfeld: 338-342).



#### APPENDIX 3: FULL IDEAL PRINCIPLES

The process of model selection involves the following five criteria:

#### (i) Consistency

The model should have logically possible signs and magnitudes of parameters and predictions. It should be consistent with existing long-run equilibria between the variables.

#### (ii) Significance

The estimated model should exhibit economic and statistical significance of its parameters, as examined through a vast array of statistical indicators (significance tests) as well as economic theory (economic significance). The model must be economically as well as statistically meaningful.

#### (iii) Data adequacy

Various indications of adequacy should suggest that the model provides an adequate representation of the data. This is where numerous statistical tests are applied, focusing on explanatory power as well as testing the adequacy of the underlying econometric assumptions. Tests of the latter can be classified into three groups (Likelihood ratio, Wald and Lagrange multipliers). These are asymptotically equivalent, but have different small-sample properties. The choice of tests is problem dependent. However, any given equation must be tested as thoroughly as possible.

#### (iv) Encompassing

The model should encompass the characteristics of rival models, so that the latter contain no information that would be useful to improve the preferred model. Several econometric procedures have been suggested to examine whether, in some sense, any alternative model could be expected to deliver superior properties.

#### (v) Sensitivity

Finally, the model should not display too much sensitivity to the sample size, the variable menu, or to other equations of the system. Sensitivity tests can be applied to individual equations and full systems.

The following table summarises the methods to test model selection criteria.



Table A3.1: Model selection criteria

Model selection criteria			
Category	Property	Examples/Methods	
Consistency	a) Inadmissibility	Signs, magnitude of parameters	
		and predictions	
	b) Poor operating characteristics	No flow equilibria	
		Inconsistent with long-run equilibrium	
Significance	a) Economic	Quantitative impact of unacceptable magnitude	
	b) Statistical	-	
	i) Nominal significance levels	F-tests, t-tests, Wald tests, etc	
	ii) Optimized significance levels	Various information criteria	
Indices of inadequacy	a) For the conditional mean	RESET, LM-test for serial	
		correlation	
	b) For the conditional variance	Various heteroscedasticity tests	
	c) For normality	Jarque-Bera test	
Encompassing	a) The mean	F-test	
	b) The variance	t-test	
Sensitivity	a) To sample size	Chow-test	
	b) To variable menu	Extreme bound analysis	
	c) To equations of model	Simulation	

Source:

Adrian Pagan's course Economics 517 Model Selection and Evaluation, taught at the University of Rochester in the Fall 1987.



# APPENDIX 4: IMPLICATIONS OF ADDITIVITY AND SEPARABILITY ON PRODUCTION FUNCTIONAL STRUCTURES

The notion of strong separability or strong additivity has major implications for the functional structure. In particular, a production function that is strongly separable must have a Cobb-Douglas or CES structure. Tests for strong separability are therefore equivalent to tests of flexible functional forms against restricted Cobb-Douglas or CES functions.

Given a function 
$$y = f(x_1, x_2, ...., x_n)$$
, it is strongly additive if  $\frac{\partial^2 y}{\partial x_i \partial x_j} = 0$ ,  $(i \neq j)$  and strongly separable (from  $x_k$ ) if  $\frac{\partial \left[ (\partial y / \partial x_i) / (\partial y / \partial x_j) \right]}{\partial x_k} = 0$ ,  $(i \neq j \neq k)$ .

If a monotonic transformation (e.g. taking the logarithmic form) of the function satisfies the additivity condition, it is referred to as being quasi-additive (Chung 1994: 112). Christensen *et al.* (1973) and Berndt and Christensen (1973) showed that a production function is homogeneous and strongly separable if and only if it can be represented as either a Cobb-Douglas or CES production function (Allen 1997: 20). If a function is additive and homothetic, the Allen-Uzuwa partial elasticities of substitution between any pair of inputs are equal and constant (Chung 1994: 189).

The notion of weak separability, on the other hand, is important because it is a necessary condition to enable aggregation, whether inputs (in the case of a cost function) or outputs (in the case of a production function). A production function, y = f(x), is said to be weakly separable if the marginal rate of substitution between any two inputs,  $x_i$  and  $x_j$ , belonging to the same group,  $N^s$ , is independent of the quantities of inputs outside of  $N^s$ :

$$\frac{\partial \left[\left(\partial y/\partial x_{i}\right)/\left(\partial y/\partial x_{j}\right)\right]}{\partial x_{k}}=0, \left(i\neq j\neq k; \forall i,j\in N^{s}, k\in N^{t}, s\neq t\right).$$

Berndt and Christensen (1973) showed that weak separability, combined with homotheticity, are necessary and sufficient for the equality of all Allen elasticities of substitution with respect to different subsets of partition, i.e.  $\sigma_{ik}^A = \sigma_{ik}^A$ ,  $(\forall i, j \in N^s, k \in N^t, s \neq t)$ .

Berndt and Christensen (1973) derived conditions for weak separability in the Translog function:

$$lnQ = ln\alpha_0 + \sum_{k=1}^{\infty} \alpha_k \ln V_k + \frac{1}{2} \sum_{k=1}^{\infty} \sum_{l=1}^{\infty} \beta_{kl} \ln V_k \ln V_l .$$

The necessary conditions for separability are:

$$\left(\beta_{ik}\alpha_j - \beta_{jk}\alpha_i\right) + \sum_{m=1} \left(\beta_{ik}\beta_{jm} - \beta_{jk}\beta_{im}\right) \ln V_m = 0, \forall i, j \in \mathbb{N}^s, k \in \mathbb{N}^t, s \neq t; m = 1, ..., n.$$

They showed that there are two possible sets of conditions under which these equations would be satisfied:



First, a set of linear restrictions:

$$\beta_{ik} = \beta_{jk} = 0, \forall i, j \in N^s, k \in N^t, s \neq t \;.$$

Alternatively, a set of non-linear restrictions:

$$\frac{\alpha_i}{\alpha_j} = \frac{\beta_{ik}}{\beta_{jk}} = \frac{\beta_{im}}{\beta_{jm}} = 0, \forall i, j \in \mathbb{N}^s, k \in \mathbb{N}^t, s \neq t; m = 1, ..., n$$

(Allen 1997: 36-37)

In the case of the two-input Translog production function:

$$lnQ = ln\alpha_0 + \alpha_1 lnV_1 + \alpha_2 lnV_2 + \beta_{11} (lnV_1)^2 + \beta_{22} (lnV_2)^2 + \beta_{12} lnV_1 lnV_2,$$

the assumption of a constant elasticity of substitution (i.e. assumptions of homotheticity and separability) implies the parameter restrictions of:

$$\beta_{11} = \beta_{22} = -\frac{1}{2}\beta_{12}$$

(Thomas 1993: 331).



#### **APPENDIX 5: A REVIEW ON PRODUCTION FUNCTIONS**

The production function, in traditional theory of the firm, expresses output (Q) as a function of inputs, that is:

$$Q = Q(V_1; V_2; \dots, V_n), (A5.1)$$

with

Q = Output / Level of production; and

 $V_i = i$ 'th production factor (i = 1, 2, ....n).

The variables Q, and  $V_i$  are flow variables so that the production function above expresses a flow of output as a function of the flow of services provided by the factor inputs. All the variables are assumed to be continuously variable, infinitely divisible and the inputs are assumed to be continuously substitutable at all levels of production. For given levels of inputs  $(V_i)$ , equation A5.1 defines a maximum possible level of output, Q. This means that the technical problem of how to achieve the greatest output from given levels of inputs is assumed to be solved. However, because of factor substitutability, a given output can be produced by many alternative combinations of inputs. The problem of deciding which combinations of inputs provide the given level of output at minimum cost is therefore an economic problem. The production function is thus not confined to the least cost combinations of all the inputs utilised (Heathfield 1987: 45).

A production function exhibits the following economic properties:

(i) The marginal product of each input is always positive, that is:

$$\frac{\partial Q}{\partial V_n} > 0. \tag{A5.2}$$

(ii) However, although the production function must be such that the marginal products of the inputs are always positive, they have to be diminishing, illustrating the law of diminishing marginal productivity, that is:

$$\frac{\partial^2 Q}{\partial V_n^2} < 0. \tag{A5.3}$$

(iii) It exhibits increasing, decreasing or constant returns to scale. The returns to scale implied by a production function depend on the response of output to an equiproportionate change in all the inputs. If equation A5.1 is homogenous of degree n, then depending on whether n is less than, equal to, or greater than unity, equi-proportionate increases in the inputs will lead to less than proportionate, equi-proportionate, or more than proportionate increases in output. In terms of equation A5.1,

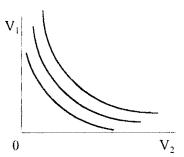
$$Q(\lambda V_n) = \lambda^n Q(V_n) \tag{A5.4}$$



where  $\lambda$  is the equi-proportionate change in factor inputs. It must be remembered, however, that all production functions are not homogeneous and that some, while exhibiting increasing returns to scale at, for example, low levels of inputs, may show decreasing returns to scale at higher input and output levels.

(iv) The fact that inputs are assumed to be continuously substitutable means that there are an infinite number of possible combinations of factor inputs (implying a wide variety of alternative techniques) which may be used to produce a given output. The slope of an isoquant yields the rate at which one factor can be substituted for another without altering the level of output. (Isoquants for the two-input case are illustrated in Figure 1). The absolute value of this slope is known as the marginal (also known as technical) rate of substitution (MRS), with

Figure 1: An isoquant map



Source: Nicholson 1994: 317

$$MRS = -\frac{dV_1}{dV_2}. (A5.5)$$

Taking the total derivative of equation A5.5 and assuming output is constant along the isoquant, result in

$$dQ = \frac{\partial Q}{\partial V_1} dV_1 + \frac{\partial Q}{\partial V_2} dV_2 = 0.$$
 (A5.6)

Thus

$$MRS = -\frac{dV_I}{dV_2} = \frac{\partial Q}{\partial V_2} / \frac{\partial Q}{\partial V_1}$$
(A5.7)

The MRS is therefore equal to the ratio of the two marginal products. The greater the ratio of  $V_2$  inputs to  $V_1$  inputs the greater the quantity of  $V_2$  needed to replace one unit of  $V_1$  without reducing output. This means that isoquants are convex to the origin. The more convex the isoquants, the more limited, generally, the substitution possibilities.

(v) The MRS and therefore also the degree of convexity or shape of the isoquants, influence the elasticity of substitution ( $\sigma$ ), which serves as an independent measure of the substitution possibilities of the production function. The elasticity of substitution ( $\sigma$ ), defined as

$$\sigma = -\frac{d(V_1/V_2)}{V_1/V_2} / \frac{d(MRS)}{MRS}$$
(A5.8)



therefore measures the proportionate change in the marginal rate of technical substitution along an isoquant. Thus, if the isoquants are relatively flat (i.e. substitution is relatively easy), then movements along an isoquant (i.e. changes in the input ratio) are accompanied by little change in the MRS and hence the elasticity of substitution is high. However, if the isoquants have a pronounced curvature, implying that substitution possibilities are more limited, then  $\sigma$  will be low.

Given, however, the marginal productivity conditions

$$\frac{\partial Q}{\partial V_1} = \frac{p_1}{p} \text{ and } \frac{\partial Q}{\partial V_2} = \frac{p_2}{p}$$
(A5.9)

with  $p_1$ ,  $p_2$  and p the prices of input<sub>1</sub>, input<sub>2</sub> and output respectively, the cost-minimising condition yields

$$MRS = \frac{\partial Q}{\partial V_2} / \frac{\partial Q}{\partial V_1} = \frac{p_2}{p_1}$$
 (A5.10)

Redefining the elasticity of substitution ( $\sigma$ ) as

$$\sigma = -\frac{d(V_1/V_2)}{V_1/V_2} / \frac{d(p_2/p_1)}{p_2/p_1} . \tag{A5.11}$$

Thus  $\sigma$  is given by the proportionate change in the input ratio divided by the proportionate change in the factor price ratio. When the price of input<sub>2</sub> rises relative to that of input<sub>1</sub>, firms will attempt to substitute input<sub>1</sub> for input<sub>2</sub> and increase the input<sub>1</sub>/input<sub>2</sub> ratio. When such substitution is possible only to a limited extent, a given proportionate change in p<sub>2</sub>/p<sub>1</sub> will lead to only small changes in the input<sub>1</sub>/ and the elasticity of substitution will be small. However, when considerable substitution is possible,  $\sigma$  will be large.

At this point it is important to take note of the concept of homotheticity. A homothetic production function is a function where the technical (marginal) rate of substitution depends only on the ratio between the inputs utilised, and not on the level of total production. The slope of the curves depends only on the ratio of the inputs used, and does not depend on the distance from the origin. Isoquants representing higher levels of production are simply copies of those for lower production. Cobb-Douglas and CES production functions are homothetic functional forms. More specifically, the Cobb-Douglas technology may be defined as a homothetic, unitary elasticity of substitution production technology (Nicholson 1994: 98).

Several functional forms of production functions, which represent the technology of production, have evolved during the history of estimating production behaviour by means of mathematical and statistical modelling. Although there are many functional forms, the most commonly used are the Cobb-Douglas, constant elasticity of substitution (CES), variable elasticity of substitution (VES), Transcendental logarithmic (Translog) and the generalised Leontief functions. In order to



Solving equations A6.2 and A6.3 with respect to  $x_1$  and  $x_2$  gives:

$$x_1 = x_1(p_1, p_2, C)$$
 (A6.4)

$$x_2 = x_2(p_1, p_2, C).$$
 (A6.5)

Equations A6.4 and A6.5 can be substituted into equation A6.1 to give:

$$C = C(p_1, p_2, y).$$
 (A6.6)

Compare equations A6.1 and A6.6. Normalising output (y = 1), there exits a duality between the production function and the corresponding cost function. The duality observed is called Shephard's duality (Chung 1994:201-202).

Given the relationship between a production and cost function, estimation of a production function can, therefore, be carried out directly or indirectly. The indirect approach is based upon the duality theory whereby a production function is derived from an estimated cost function. An alternative approach can be followed, by means of which the production function is estimated directly. The indirect approach is used in this study to overcome the obstacles posed by data constraints associated with the direct approach.

By evaluating different functional forms to estimate a production function and by using duality theory, the production function can be empirically estimated. This estimation will consist of an estimation of the Cobb-Douglas cost function. The cost function will then be used to derive the Cobb-Douglas production function. The validity of the Cobb-Douglas functional form, as representation of the production technology of the South African economy, is tested. The restrictive assumptions of a constant and unitary elasticity of substitution are tested by estimating a Translog production function, imposing the restrictions step-by-step and testing their validity to determine whether the Translog collapses to the Cobb-Douglas functional form.



## APPENDIX 7: AN EXPOSITION OF THE DATA UTILISED IN THE SUPPLY-SIDE MODEL

The necessary data was obtained from the South African Reserve Bank's Quarterly Bulletin, South African Statistics (SA Statistics) and the Development Bank of Southern Africa (DBSA). Annual, constant price data from 1970 to 1995 was used to estimate the parameters of the model. Where appropriate data was transformed, 1990 served as base year. Series affected by inflation were deflated by the appropriate price index.

All data, except the real capital stock<sup>1</sup> in the real fixed investment function and the technology index included in the cost/production function, was transformed into logarithmic form. Apart from the fact that the structure of the transcendental functional form requires that logarithmic variables be used, there are certain advantages to the estimation of logarithmic models. Logarithmic models transform non-linear models (relationships) into linear form, which allows the use of linear estimation procedures. The estimated regressors are coefficients of elasticity (partial multipliers), rather than coefficients of marginal effects. Logarithmic transformations also help to overcome problems due to non-stationarity.

The following time series had to be derived for the variables defined in the various theoretical models:

#### (i) Real gross domestic cost of production at factor cost

An investigation of the South African input-output table concluded that only 5 percent of the gross operating surplus, published by the South African Reserve Bank, relates to production (capital and entrepreneurial) costs. The definition for the real gross domestic cost of production at factor cost (c) can therefore be expressed as:

$$c = \frac{labour \ remuneration \ + \ 0.05*gross \ operating \ surplus}{production \ price \ index}$$

with the series for nominal labour remuneration, nominal gross operating surplus and the production price index obtained from the SARB Quarterly Bulletin.

#### (ii) Cost of labour

The cost of labour represents the nominal wage bill of the country and is calculated by adding the wages of skilled workers and unskilled workers in South Africa. Unskilled workers are classified as those workers without matriculation or an equivalent

Investment decisions are made with the purpose of profit maximisation, i.e. investment decisions (the extent to which capital stock needs to be increased) are made to obtain an optimal level of capital, i.e. the point where the marginal product of capital ( $\Delta Y/\Delta K$ ) equals real user-cost-of-capital. Therefore, in the empirical estimation of an investment function, although the natural logarithms of all relevant variables are used, capital stock must be used in its level-form. It implies that the relative change in investment is not dependent on the relative change in capital stock, but on its absolute level. Remember that the logarithmic form of any variable represents the relative change of that variable over time.



qualification. Skilled workers are classified as those workers with matriculation or a higher qualification. The cost of labour ( $\ell$ ) can be represented by the following expression:

$$\ell = \frac{(wage \ rate \ of \ unskilled \ labour + wage \ rate \ of \ skilled \ labour)}{production \ price \ index}$$

Data for skilled and unskilled wage rates was obtained from the Development Bank of South Africa.

#### (iii) User-cost-of-capital

While the cost of labour can be calculated by adding directly observable series, the cost of capital has to be inferred. The neo-classical approach by Hall and Jorgenson's (1967) is used to derive a measure for the user-cost-of-capital (r). See Appendix 14 for a discussion of the theory. The measure for user-cost-of-capital (r) can be expressed as:

$$r = price \ of \ capital \left( \frac{(interest \ rate) - (inflation \ rate) + (rate \ of \ depreciation)}{1 - tax \ ratio} \right)$$

with the price of capital approximated by the deflator for gross domestic fixed investment; the production price index is utilised as the appropriate inflation rate; the rate of depreciation is set at 20 percent per annum<sup>2</sup>, and the tax ratio is defined as the direct tax rate of incorporated business enterprises relative to the gross domestic product at factor cost.

#### (iv) Technical progress

Instead of resorting to either the Hicks-neutral  $(A = A_0)$  or the more commonly used, although still restrictive, Harrod-neutral (labour-augmenting) technical progress specifications, an attempt was made to at least partially endogenise technical progress in the cost/production relation. See Appendix 12 for a discussion on the various kinds of technical progress. A technological index is included in the function to represent technical progress. With the construction of the index, recognition is given to work done by Budd and Hobbis (1989) by capturing the role of imported technical innovation, proxied by foreign investment relative to domestic investment. The index is extended to allow for human capital augmentation (through the inclusion of an education index), labour augmentation (through the inclusion of a productivity index) and exogenous factors such as once-off innovations. A time trend is also included to allow for a degree of neutrality.

The technological index (*t*) can therefore be represented by the following expression:

This rate is assumed to be 20 percent, based on the fact that firms write their capital stock off over a period of usually 5 years in an attempt to optimise the tax deductions based on the depreciation of capital goods.



$$t = \left(\frac{education\ index}{1.51}\right) * 0.15 + \left(\frac{productivity}{100}\right) * 0.15 + \left(\frac{relative\ foreign\ investment}{1.805383}\right) * 0.05 + \left(\frac{time\ +\ technical\ innovation}{1}\right) * 0.65$$

#### (v) Financing of gross domestic investment

The financial constraint variable (fc), i.e. the "financing of gross domestic investment" according to the system of national accounts, enters the system of equations as an identity, adding together private savings (sp), government savings (sg), corporate savings (sc), replacement investment or depreciation in real capital stock (depr), net foreign capital flow (capflow) and the change in gold and other foreign reserves (reserv):

$$fc = sp + sg + depr + sc + capflow + reserv$$
.

However, since the financing of gross domestic investment equals total gross domestic investment in a general equilibrium framework such as the system of national accounts, data series for both these entities may be utilised. These series are only published in nominal terms and had to be deflated with the production price index to obtain the constant price values.

An explanatory list of all the variables encountered in the long-run cointegration relationship and the error correction model (ECM) is presented in Appendix 8.

Tests for non-stationarity (the presence of a unit root) were subsequently performed. Inferences concerning the stationarity properties of data were based on a range of evidence, of which the formal ADF test is merely one component. Economic theory, structural properties of the data, a priori information concerning political and other exogenous shocks to the economy over the sample period (more often than not resulting in structural breaks), data plots (Appendix 9), residual plots, plots of the autocorrelation function and spectral analysis all form part of the evidence used to draw conclusions on the stationarity of the data. This comprehensive set of inference guidelines prompted the conclusion that all relevant variables are integrated of order 1, i.e. stationary in first differences.

The results of the Augmented Dickey-Fuller tests as well as an exposition of the testing procedure are reported in Appendix 17.

Apart from the long-run explanatory variables also included in the ECM, a few dummy variables were included in several individual models to avoid a bias in the estimation which may arise if significant events in the South African economy are ignored. These variables smooth out the significant and substantial effects of the following incidents:



#### (i) Economic sanctions (sanction dum)

During the latter half of the eighties and the early nineties, South Africa was heavily affected by sanctions as a result of growing international discontent with the country's political dispensation. Sanctions had an adverse effect on the domestic cost of production. A higher demand for local inputs increased the prices of these production factors, while imported inputs became increasingly scarce and expensive. It also affected production directly in that local producers no longer had access to several, more productive imported inputs.

#### (ii) Drought (drought dum)

South Africa experienced a severe drought in 1983 and a prolonged one from 1992 to 1995. These droughts had a direct, negative effect on agricultural production and increased the production cost of the secondary industries dependent on agricultural inputs.

#### (iii) IMF support (imf dum)

South Africa benefited from IMF assistance from 1975 to 1977. These contributions stimulated the demand side of the economy and, at least in theory, also production. These funds were also used to assist production in South Africa directly

#### (iv) Opec oil shocks (opec dum)

The oil price shocks of the 1970s resulted in dramatic increases in domestic fuel prices and as a consequence activated a chain of price and cost surges in the economy. The economy in general and firms in particular could not escape the negative effects of increasing costs, decreasing profits and savings, and slumping economic growth.

#### (v) Emigration of skilled labour (braindrain dum)

International sanctions and disinvestment during the mid-1980s and early 1990s, coupled with a new political dispensation endorsing labour acts of affirmative action, created an unstable and unprofitable environment for especially skilled labour in South Africa. This led to a substantial outflow of highly qualified and professional labour which the domestic economy is in serious need of.



### APPENDIX 8: VARIABLE LIST

series	natural logarithms	Variable names	Data source
bbb_cp		Gross domestic expenditure at current prices	KB6012jj
bbp_90p	ln_bbp_90p	Gross domestic product at market prices at constant 1990 prices	KB6006yj
bbp_cp		Gross domestic product at market prices at current prices	KB6006jj
bbp_min_90p	ln_bbp_min_90p	Gross domestic product: mining and quarrying at constant 1990 prices	KB6032yj
bbpfact_90p	ln_bbpfact_90p	Gross domestic product at factor cost at constant 1990 prices	KB6003yj
bbpfact_cp		Gross domestic product at factor cost at current prices	KB6003jj
bpppot_90p	ln_bbppot_90p	Potential gross domestic product at factor cost at constant 1990 prices	Solved by production function
braindrain_dum		Dummy: emigration of skilled labour	na
c90p_lrem_gos5	ln_c90p_lrem5	Gross domestic cost of production at constant 1990 prices	cost_lrem_gos95/ppi
cost_lrem_gos5		Gross domestic cost of production at current prices	(lab_rem - 0.05*gos_cp)
crime_index	ln_crimeind	Crime index	crime incidents/total_pop/1990-number/1000 (South African survey)
cu	ln_cu	Percentage utilisation of production capacity: manufacturing sector	KB7078jj
de_cp		Provision for depreciation at current prices	КВ6002јј
depend_rat		Dependency ratio	(total_pop - s*1000)/(s*1000)
dìpi		Deflator : gross domestic fixed investment	(if_cp/if_r_sarb)
dirinves_r	ln_dirinves_r	Total foreign direct investment at constant 1990 prices	dirinvest/ppi
dirinvest		Total foreign direct investment at current prices	KB5162jj
drought_dum		Dummy: periods of severe drought	na
educ_index	ln_educind	Education index	((matrics passed/total_pop/1990-number)*100)
empl_rat	ln_empl_rat	Employment ratio	n/s
empl_rat_s	ln_empl_rat_s	Employment ratio: skilled labour sector	n_s/s_s
empl_rat_u	ln_empl_rat_u	Employment ratio: unskilled labour sector	n_u/s_u
energy_ind	ln_energy_ind	Energy index = electricity index	KB7068jj
excess_dem_cp	ln_exces_dem_cp	Excess demand at current prices	bbb_cp/bbp_cp
exchange_rate		Dollar/rand exchange rate	100/r_\$



series	natural logarithms	Variable names	Data source
fincond_cp	ln_fincond_cp	Financing of gross domestic investment at current prices	(sp_cp + sc_cp + sg_cp + de_cp + netcapfl + gold_reserv_cp) = totdomi_cp
fincond_ppi	ln_fincond_ppi	Financing of gross domestic investment at constant 1990 prices : PPI deflated	(sp_cp + sc_cp + sg_cp + de_cp - netcapfl - gold_reserv_cp)/ppi = totdomi_cp/ppi
gold_reserv_cp		Change in gold and other foreign reserves at current prices	KB6205jj
gos_cp		Gross operating surplus at current prices	KB6212jj
gostc_cp	ln_gostc_cp	After-taxed gross operating surplus at current prices	gos_cp - tc_cp
gprys_\$	ln_gprys_\$	London gold price in US dollar	KB5357jj*100
gprys_r	ln_gprys_r	London gold price in SA rand	KB5356jj *100 (gprys_\$*r_\$)
h_transf_receiv	ln_h_transf	Net current transfers received by households from government, incorporated business enterprises and the rest of the world	h_tran_rec_c (KB6231jj)+ h_tran_rec_g (KB6257jj) + h_tran_rec_w (KB6243jj) - h_tran_pay_g (KB6252jj) - h_tran_pay_w (KB6248jj)
health_index	ln_healthind	Health index	(number of medical practisions / total_pop)/1990- number/1000 (SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ health)
if	ln_if	Gross domestic fixed investment at constant 1990 prices	if_cp/ppi
if_cp		Gross domestic fixed investment at current prices	KB6009jj
if_r_sarb		Gross domestic fixed investment at constant 1990 prices	KB6009yj
imf_dum		Dummy: IMF contributions and loans	na
income_index	ln_incomeind	Income index: per capita personal disposable income at constant 1990 prices	KB6272yj
indinves_r	ln_indinves_r	Total foreign non-direct investment at constant 1990 prices	indinvest/ppi
indinvest		Total foreign non-direct investment at current prices	KB5180jj
intern_comp	ln_intern_comp	International competitiveness of South African economy	domestic export prices relative to world export prices (IFS)
intern_posind	ln_interposind	International position index	((exchange_rate/1990-number)*0.2 + (intern_comp/1990-number)*0.05 + (rel_diri_r/1990-number)*0.5 + (rel_indi_r/1990-number)*0.2 + (worldshare/1990-number)*0.05)
inv_cp		Total change in inventories at current prices	KB6010jj
kap_lab_rat	ln_kap_lab_rat	Capital-labour ratio	kap_r/(lab_rem_cp/ppi)*1000



series	natural logarithms	Variable names	Data source
kap_r	ln_kap_r	Capital stock at constant 1990 prices	KB6149yj
lab_rem_cp		Remuneration of employees at current prices	KB6000jj
n	ln_n	Demand for labour (Employment)	SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ labour or DBSA: Development Information Business Unit – standardised employment series
n_diff		Difference between employment figures from SA Statistics and employment figures from sarb	n - n_sarb
n_ind_sarb		Demand for labour in noon-agricultural sectors : 1990 = 100	KB7009jj
n_informal	ln_n_informal	Labour participants in the informal sector	DBSA (n_informal (DBSA) + n (SA Statistics) + unempl (SA Statistics)= s (SA Statistics))
n_pot	ln_n_pot	Potential demand for labour (employment)	(s_smooth*(1 - nawru_hpf))
n_s	ln_ns	Demand for skilled labour (skilled employment)	(percentages of DBSA applied to n of SA Statistics)/100
n_sarb		Demand for labour (excluding agricultural sector)	(n_ind_sarb)*1990-number/100 000 000
n_u	ln_nu	Demand for unskilled labour (unskilled employment)	(percentages of DBSA applied to n of SA Statistics)/100
nawru		Non-accelerating wage rate of unemployment	unempl_rat - (d(unempl_rat)/d <sup>3</sup> (ln_wtot_vpi))*d <sup>2</sup> (ln_wtot_vpi)
nawru_hpf		Hodrick-Prescott filter used to smooth nawru	na
netcapfl		Net capital flow at current prices	KB5024jj
netcapfl_r		Net capital flow at constant 1990 prices	netcapfl/ppi
opec_dum		Dummy: OPEC oil price shocks	na
p_manuf	ln_p_manuf	Production prices: manufacturing (1990 = 100)	KB7046jj/63.6
pk		Price of capital	KB2004jj (escom stock)
pk_ppi		Price of capital at constant 1990 prices : PPI deflated	pk - ppigrowth
pnettx		Deflator : net indirect taxes	(txind - subs)/(bbp_90p - bbpfact_90p) (vpigrowth in future)
ppi	ln_ppi	Production prices: total (1990 = 100)	KB7048jj/65.4



series	natural logarithms	Variable names	Data source
ppigrowth	ln_ppigrowth	Production inflation	((ppi - ppi(-1)/ppi(-1))*100
prime_rate	ln_prime_rate	Predominant prime overdraft rate of clearing banks at current prices	KB1403g
product_sarb	ln_product_sarb	Labour productivity according to SARB statistics, excluding agricultural sector	(bbp_90p/n_sarb)/1990-number
product	ln_product	Labour productivity	(bbp_90p/n/1990-number
px	ln_px	Deflator: exports of goods and non-factor services	xgnfd_cp/xgnfd_90p
px_sa_\$		Deflator: South African exports in US dollar prices	IFS: 19974DZF; WT028
px_world_\$		Deflator: World exports in US dollar prices	IFS: 00174DZF; WT028
pz	ln_pz	Deflator: imports of goods and non-factor services	zgnfd_cp/zgnfd_90p
pz_\$	ln_pz\$	Import prices in US dollar	pz*exchange_rate
r_\$	ln_r\$	Rand-dollar exchange rate	KB5339jj
rel_diri_r	ln_rel_diri_r	Direct foreign investment relative to gross domestic investment at constant prices	(dirinvest/if_cp)/ppi
rel_i_r	ln_rel_i_r	Total foreign investment relative to gross domestic investment at constant 1990 prices	((dirinvest + indinvest)/if_cp)/ppi
rel_indi_r	ln_rel_indi_r	Indirect foreign investment relative to gross domestic investment at constant 1990 prices	(indinvest/if_cp)/ppi
rel_wscost_ucc	ln_rel_wscost_u	Skilled labour cost relative to user-cost-of-capital	ws_cost_ppi/ucc2_90p
rel_wsu	ln_rel_wsu	Skilled wages relative to unskilled wages	w_s/w_u
rel_wsu_rat	ln_rel_wsu_rat	Skilled wage rate relative to the unskilled wage rate	(w_s/n_s)/(w_u/n_u)
rel_wucost_ucc	ln_rel_wucost_u	Unskilled labour cost relative to user-cost-of-capital	wu_cost_ppi/ucc2_90p
S	ln_s	Total labour supply	SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ labour
s_rat	ln_srat	Total labour supply ratio	s/total_pop
s_rat_hpf		Hodrick-Prescott filter used to smooth total supply of labour	na
s_s	ln_ss	Skilled labour supply	SA Statistics (www.statssa.gov.za/ publications/ statistics_in_brief/ labour) data for s, but ratio according to DBSA
s_smooth		Smoothed labour supply	na



series	natural logarithms	Variable names	Data source
s_u	ln_su	Unskilled labour supply	SA Statistics (www.statssa.gov.za/ publications/ statistics_in_brief/ labour) data for s, but ratio according to DBSA
sanction_dum		Dummy : sanctions	па
sc_cp	ln_sc_cp	Corporate saving at current prices	KB6201jj
sc_cp_rat		Corporate savings rate at current prices	sc_cp/bbp_cp
sg_cp		Saving of the general government at current prices	KB6202jj
share_ns		Demand for labour share of skilled workers	OHS: Workers by level of education (grade 10/NTC I + grade 11/NTC II + grade 12/NTC III + diplomas + degrees), SA Statistics (www.statssa.gov.za/publications/ statistics_in_brief/ labour)
share_nu		Demand for labour share of unskilled workers	(1 - share_ns)
share_ss		Labour supply share of skilled workers	OHS: Level of education (grade 10/NTC I + grade 11/NTC II + grade 12/NTC III + diplomas + degrees), SA Statistics (www.statssa.gov.za/ publications/ statistics_in_brief/ labour)
share_su		Labour supply share of unskilled workers	(1 - share_ss)
share_ws		Wage share of skilled labour	DBSA: high level + mid level
share_wu		Wage share of unskilled labour	DBSA: semi-level + unskilled
soc_index	ln_socind	Socio-economic index	(educ_index/1990-number)*0.20 + (energy_ind/1990-number)*0.15 + (health_index/1990-number)*0.15 + (1/(crime_index/1990-number))*0.1 + (income_index/1990-number)*0.2 + (h_transf_receiv/1990-number)*0.1 + (unpl_ben/1990-number)*0.1
sp_cp		Personal saving at current prices	KB6200jj
ss_rat	ln_ssrat	Skilled labour supply ratio	s_s/s
subs		Subsidies at current prices	KB6005jj
tc_cp		Direct taxes of incorporated business enterprises at current prices	KB6230jj
tc_ppi		Direct taxes of incorporated business enterprises at constant 1990 prices: PPI deflated	tc_cp/ppi



series	natural logarithms	Variable names	Data source
tc_rate_cp		Direct tax rate of incorporated business enterprises at current prices	tc_cp/bbpfact_cp
tc_rate_ppi	ln_tcrate_ppi	Direct tax rate of incorporated business enterprises at constant 1990 prices : PPI deflated	tc_ppi/bbpfact_90p
tecno_hpf		Hodrick-Prescott filter used to smooth technology index	na
tecno_index		Technology index	(educ_index/1990-number)*0.15 + (product/1990-number)*0.15 + (rel_i_r/1990-number)*0.05 + (time/10 + tecno_innov)*0.65
tecno_innov		Dummy: technology innovation	na
total_pop	ln_total_pop	Total population	SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ demography
totdomi_cp		Gross domestic investment at current prices	KB6180jj
txind	ln_txind	Indirect taxes at current prices	KB6004jj
ucc2_90p	ln_ucc2_90p	User-cost-of-capital at constant 1990 prices	dipi*((pk_ppi/100) + 0.2)/(1 - tc_rate_ppi)
ucc2_nom	ln_ucc2_nom	User-cost-of-capital at current prices	dipi*((pk/100) + 0.2)/(1 - tc_rate_cp)
unempl		Total unemployment	unempl_rat*s
unempl_rat	ln_unempl_rat	Unemployment rate	SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ labour
unempl_rat_s	ln_unemplrat_s	Unemployment rate: skilled labour	unempl_s/n_s
unempl_rat_u	ln_unemplrat_u	Unemployment rate: unskilled labour	unempl_u/n_u
unempl_s		Skilled labour force unemployment	s_s - n_s
unempl_u		Unskilled labour force unemployment	s_u-n_u
union_members		Union members	ILO Review or NMC, Department of labour or South African Institute of Race Relations Survey
union_mil	ln_unionmil	Union militancy	union_work_lost/n
union_power	ln_unionpower	Union power	union_members/n
union_pres_ind	ln_uniopresind	Union pressure index	((union_power/1990-number)*0.60 + (union_mil/1990-number)*0.10 + (depend_rat/1990- number)*0.30)
union_work_lost		Strikes and stoppages: man-days lost	KB7018jj



series	natural logarithms	Variable names	Data source
unpl_ben		Unemployment benefit	SA Statistics: SA labour statistics, table 4.2.1.1 (SA Statistics, www.statssa.gov.za/ publications/ statistics_in_brief/ labour)
vpi	ln_vpi	Consumer price index	KB7,032jj
vpigrowth	ln_vpigrowth	Consumption inflation	(vpi - vpi(-1))/vpi(-1)*100
w_cost_cp		Cost of labour at current prices	((wtot_rate)*(1+0.16))/(1-(tc_rate_cp)*(1+0.3)) with pension contribution = 0.06 (6%) medical contribution = 0.1 (10%) non-wage contributions = medical + pension = 0.16 (16%) corporate income tax rate = 0.3 (30%)
w_prod	ln_w_prod	Wage productivity	(w_tot/product)/100
w_s		Skilled wages at current prices	w_tot*share_ws
w_tot		Total wages at current prices	(w_s + w_u) <i>or</i> KB6000jj
w_u		Unskilled wages at current prices	w_tot*share_wu
worldshare	ln_worldshare	World market share	(x_sa_\$/x_world_\$)/10 (IFS)
ws_cost_cp		Cost of skilled labour at current prices	w_cost_cp*share_ws
ws_cost_ppi	ln_wscost_ppi	Cost of skilled labour at constant 1990 prices : PPI deflated	ws_cost_cp/ppi
ws_rate	ln_ws_rate	Skilled wage rate at current prices	w_s/n_s
ws_vpi		Skilled wages at constant 1990 prices : CPI deflated	w_s/vpi
ws_vpi_rate	ln_wsvpi_rat	Skilled wage rate at constant 1990 prices : CPI deflated	ws_vpi/n_s
wtot_ppi_rat	ln_wtot_ppi_rat	Total wage rate at constant 1990 prices : PPI deflated	(wtot_rate/ppi)/n
wtot_rate	ln_wtot_rate	Total wage rate at current prices	w_tot/n
wtot_vpi	ln_vvtot_vpi	Total wages at constant 1990 prices : CPI deflated	(w_tot)/vpi
wtot_vpi_rat	ln_wtot_vpi_rat	Total wage rate at constant 1990 prices : CPI deflated	(wtot_rate/vpi)/n
wu_cost_cp		Cost of unskilled labour at current prices	w_cost_cp*share_wu
wu_cost_ppi	ln_wucost_ppi	Cost of unskilled labour at constant 1990 prices: PPI deflated	wu_cost_cp/ppi
wu_ppi		Unskilled wages at constant 1990 prices : PPI deflated	w_u/ppi

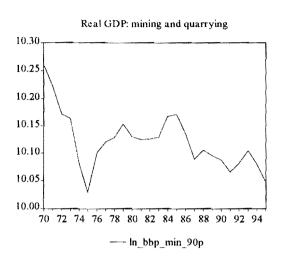


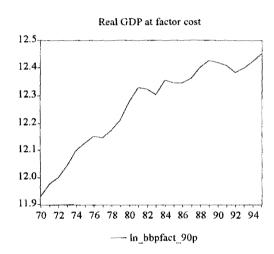
series	natural logarithms	Variable names	Data source
wu_ppi_rate	ln_wuppi_rat	Unskilled wage rate at constant 1990 prices : PPI deflated	wu_ppi/n_u
wu_rate		Unskilled wage rate at current prices	w_u/n_u
wu_vpi		Unskilled wages at constant 1990 prices : CPI deflated	w_u/vpi
wu_vpi_rate	ln_wuvpi_rat	Unskilled wage rate at constant 1990 prices : CPI deflated	wu_vpi/n_u
x_sa_\$		South African exports at US dollar prices	IFS: 19970DZF; WT025
x_world_\$		World exports at US dollar prices	IFS: 00170DZF; WT025
xgnfd_90p		Exports of goods and non-factor services at constant 1990 prices	KB6013yj
xgnfd_cp		Exports of goods and non-factor services at current prices	KB6013jj
xgoud_cp		Net gold exports at current prices	KB5001jj
xgoud_ppi	In_xgoud_ppi	Net gold exports at constant 1990 prices : PPI deflated	xgoud_cp/ppi
xgoud_px	ln_xgoud_px	Net gold exports at constant 1990 prices : deflated by export prices	xgoud_cp/px
zgnfd_90p		Imports of goods and non-factor services at constant 1990 prices	KB6014yj
zgnfd_cp		Imports of goods and non-factor services at current prices	KB6014jj

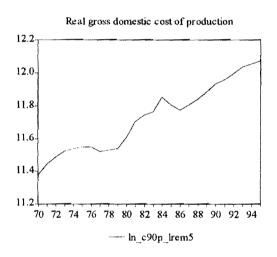


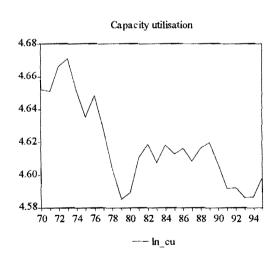
#### APPENDIX 9: GRAPHICAL REPRESENTATION OF THE DATA

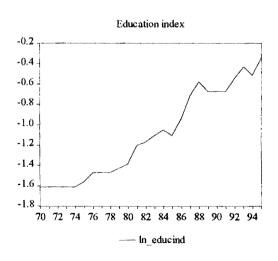




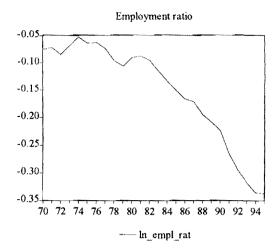


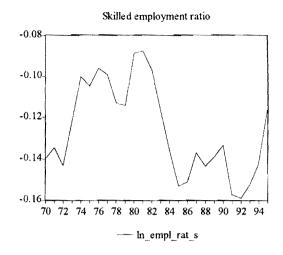


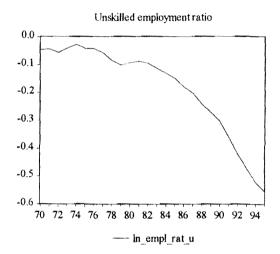


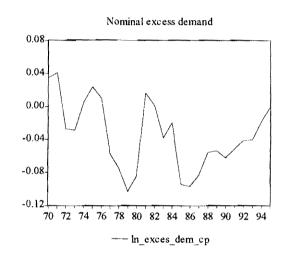


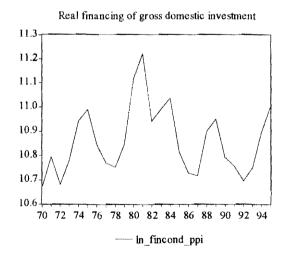








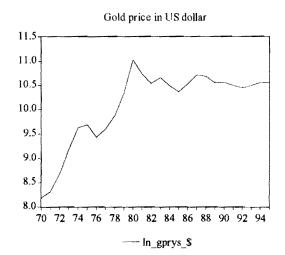


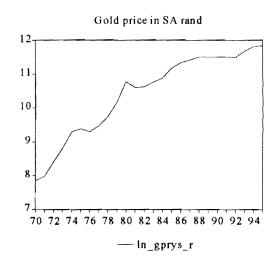


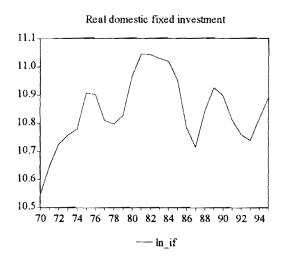


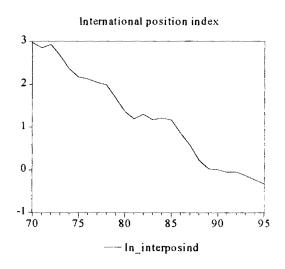
ranger and the state of the contract of the state of the

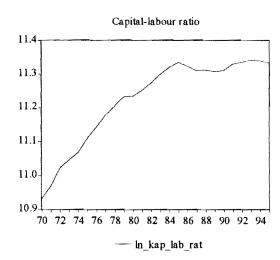


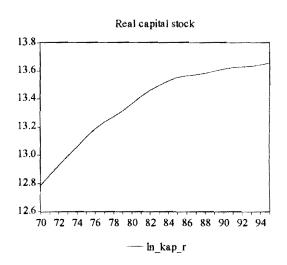




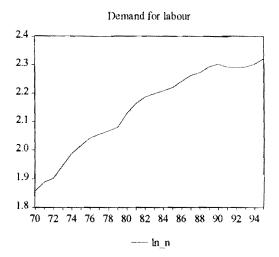


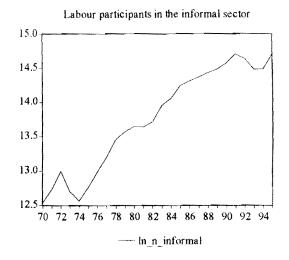


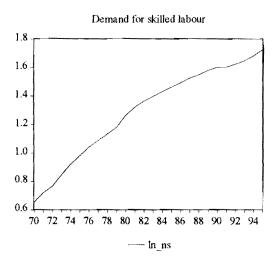


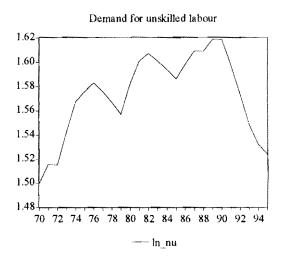


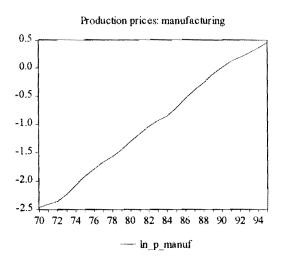


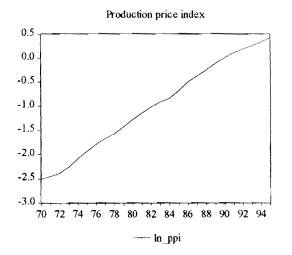






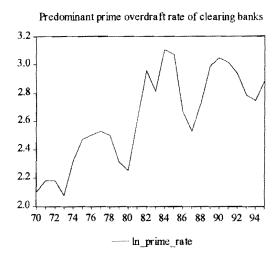


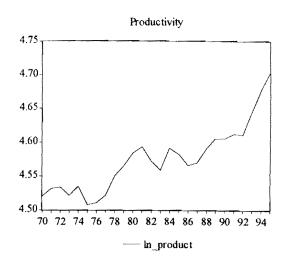


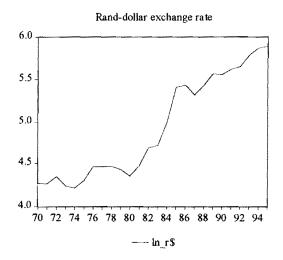


 $oldsymbol{1}$  ,  $oldsymbol{1}$  , oldsymbol

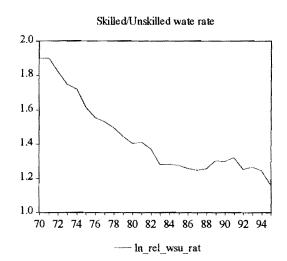


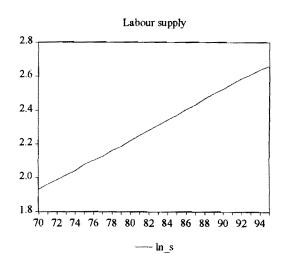




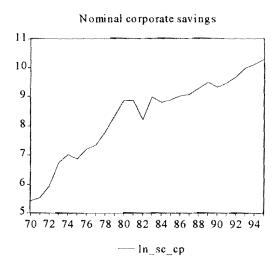


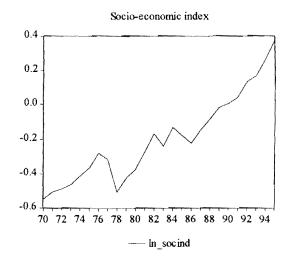


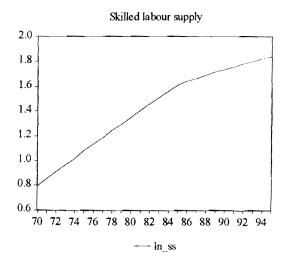


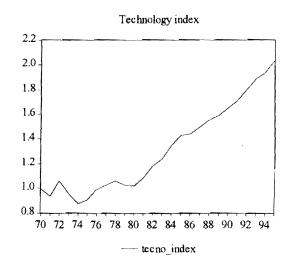


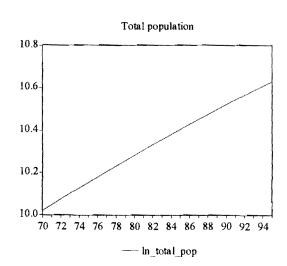


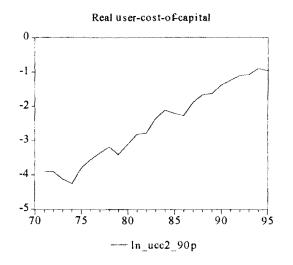






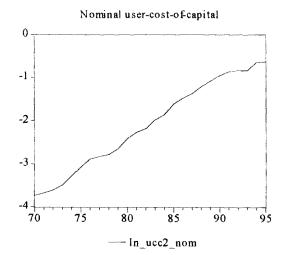


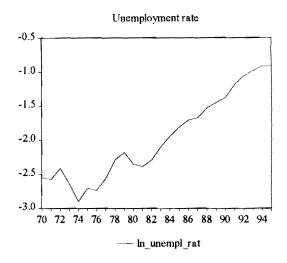


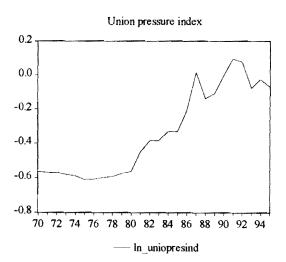


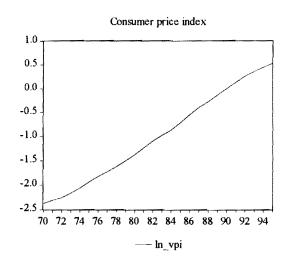
the second of the second of

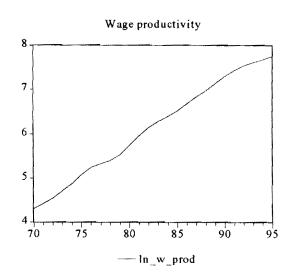


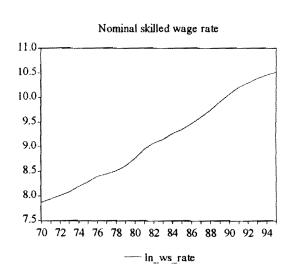




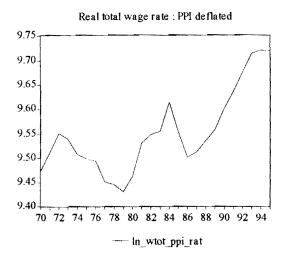


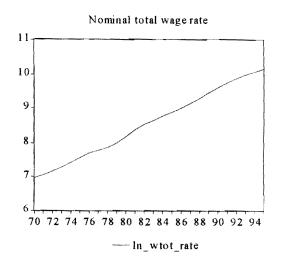


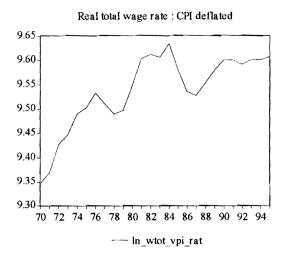


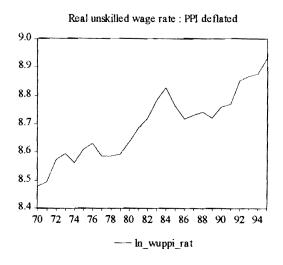


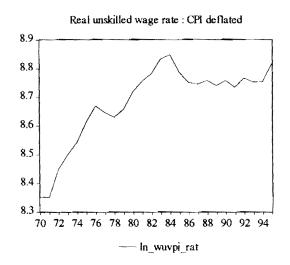


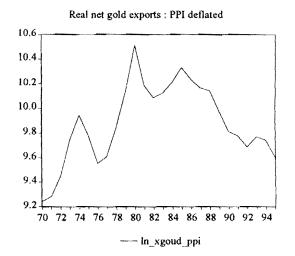








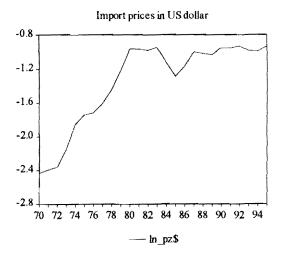




for the state of t









# APPENDIX 10: EVALUATING THE RESPONSE PROPERTIES OF THE MODEL: METHODOLOGY

The dynamic simulation properties of every individual model are investigated and it is tested for stability and robustness simultaneously.

For this purpose, dynamic, *ex-post* and partial simulations are performed on the estimated production function. Two types of dynamic simulations are performed: first, an *ex-post* dynamic simulation on the production function (without any shocks applied) and second, *ex-post* dynamic simulations where each of the explanatory variables are independently increased (shocked).

The initial dynamic simulation (without applying any shocks) is carried out as an indication of the goodness of fit of the model. The response characteristics of the model are evaluated by subsequently subjecting the model to sensitivity testing by changing (shocking) each of the explanatory variables (one at a time). This is done by increasing them by 10 percent. For a model to be stable and robust, shocks applied to it should result in consistent long-run multiplier effects. A consistent long-run multiplier effect means that the difference between the shocked simulated value and the simulated value without a shock must ideally result in approximately 10 percent of the original coefficient of the shocked variable (i.e. the multiplier effect). A shock in short-term explanatory variables should cause an initial movement away from the long-run equilibrium growth path, but eventually the model should converge to the original equilibrium growth path.

Having applied a 10 percent shock to each of the explanatory variable, the dependent variable converged around a new equilibrium, with the adjustment equal to 10 percent of the estimated coefficient of the shocked long-run explanatory variable. In the case of shocked short-run explanatory variables, the dependent variable converged to the original equilibrium growth path.

The results of the adjustment process towards either a new long-run equilibrium (in accordance with the elasticities of the respective cointegration relationships) or the baseline equilibrium (in the case of short-run explanatory variables) are both tabled and graphically illustrated. Vertical axes measure the difference between the outcome of the baseline estimation and the estimation subjected to the exogenous shock, as a percentage of the level of the dependent variable. The speed of adjustment in respective cases is apparent from the graphs. In all instances the adjustment process is completed within the sample range.



# APPENDIX 11: KMENTA'S TAYLOR APPROXIMATION OF THE EXPANSION OF THE CES FUNCTIONAL FORM INTO THE TRANSLOG FUNCTIONAL FORM

By using a Taylor series, the CES functional form can be expanded into the Translog functional form. This expansion was first suggested by Kmenta (1967), and has been utilised by Sargan (1971) and Griliches and Ringstad (1971).

Kmenta based his research on the CES function, which was first introduced by Arrow, Chenery, Minhas and Solow in 1961:

$$Q = \gamma \left(\delta V_1^{-\theta} + (1 - \delta) V_2^{-\theta}\right)^{\frac{-\nu}{\theta}}$$
(A1)

where Q = production;  $\gamma$  = efficiency parameter ( $\gamma$  > 0);  $V_i$  = production factors (inputs) with i = 1,2;  $\theta$  = substitution parameter (-1 <  $\theta$  <  $\infty$ ); and  $\delta$  = distribution (capital intensity) parameter (0 <  $\delta$  < 1).

By assuming a Taylor series expansion of about  $\theta = 0$ , Kmenta obtains a linear approximation of the CES function, which can be written as:

$$\ln Q = \ln \gamma + \nu \delta \ln V_1 + \nu (1 - \delta) \ln V_2 - \frac{1}{2} \nu \theta \delta (1 - \delta) [\ln (V_1 / V_2)]^2$$
(A2)

Provided  $\theta$  is near zero (i.e. provided the elasticity of substitution,  $\sigma = 1/(1 + \theta)$ , is near unity), equation A2 provides a close and convenient approximation of equation A1. This assumption has to be tested in practice.

Kmenta's Taylor approximation to the CES function (equation A2) is, however, of the same form as the Translog production function with imposed restrictions of homotheticity and separability, i.e. a Translog production function exhibiting a constant elasticity of substitution.

The Translog production function of Christensen, Jorgenson and Lau approximates the logarithm of output by a quadratic in the logarithms of the inputs:

$$lnQ = ln\alpha_0 + \sum_{i=1}^{2} \alpha_i \ln V_i + \frac{1}{2} \sum_{i=1}^{2} \sum_{j=1}^{2} \beta_{ij} \ln V_i \ln V_j$$

$$= ln\alpha_0 + \alpha_1 \ln V_1 + \alpha_2 \ln V_2 + \beta_{11} (\ln V_1)^2 + \beta_{22} (\ln V_2)^2 + \beta_{12} \ln V_1 \ln V_2 .$$
(A3)

Restricting the Translog production function to homotheticity and a constant elasticity of substitution, i.e. imposing the restrictions of  $\sum \beta_{ij} = \sum \beta_{ji} = 0$  and  $\beta_{11} = \beta_{22} = -\frac{1}{2}\beta_{12}$ , the equation becomes:

If  $\theta = 0$  and  $\sigma = I$ , then the Translog production function of the form:  $\ln Q = \ln \alpha_0 + \alpha_1 \ln V_1 + \alpha_2 \ln V_2 + \beta_{11} (\ln V_1)^2 + \beta_{22} (\ln V_2)^2 + \beta_{12} \ln V_1 \ln V_2,$ reduces to the Cobb-Douglas production function of the form:  $\ln Q = \ln \alpha_0 + \alpha_1 \ln V_1 + \alpha_2 \ln V_2.$ 



$$lnQ = ln\alpha_0 + \alpha_1 \ln V_1 + \alpha_2 \ln V_2 - \frac{1}{2} \beta_{12} (\ln V_1 - \ln V_2)^2$$
(A4)

which is of the same form as Kmenta's Taylor approximation to the CES function<sup>2</sup>.

Hence, if the Translog production function (equation A3) is estimated, the hypothesis of a CES may be tested by checking whether the estimated coefficients obey the restrictions:  $\beta_{11} = \beta_{22} = -\frac{1}{2}\beta_{12}$  (Thomas 1993: 329-331).

Griliches and Ringstad (1971) performed this test by using cross-sectional inter-firm data for different Norwegian industries. In order to address the problem of a high degree of multicollinearity between the variables  $lnV_1$ ,  $lnV_2$  and  $[ln(V_1/V_2)]^2$ , they rearranged Kmenta's Taylor approximation as:

$$ln\left(\frac{Q}{V_2}\right) = ln\gamma + (\nu - 1)lnV_2 + (\nu\delta)\ln\left(\frac{V_1}{V_2}\right) - \frac{1}{2}\nu\theta\delta(1 - \delta)\left[\ln\left(\frac{V_1}{V_2}\right)\right]^2. \tag{A5}$$

Estimation of equation A5 resulted in estimates for  $\gamma$ ,  $\nu$ ,  $\delta$  and  $\theta$  and therefore provides information on the properties such as the returns to scale ( $\nu$ ), capital intensity of production ( $\delta$ ) and elasticity of substitution ( $\sigma = 1/(1 + \theta)$ ). The closeness of  $\theta$  to zero serves as a test whether the Kmenta approximation to the CES function is valid (see equation A2).

As mentioned above this transformation of the CES functional form into the Translog functional form, will be used to test the validity of the Cobb-Douglas functional form as a representation of production technology in South Africa.

Consider the CES equation originally formulated by Arrow in 1961:

$$y^{\rho} = \alpha_0 + \sum_{i=1}^n \alpha_i x_i^{\rho}$$

with x a vector of inputs or prices, depending on whether a direct or dual form is being considered, y is output, cost or profit.

If  $\alpha_0 = 0$  and  $\Sigma \alpha_i = 1$  in the CES formula, then a first order expansion in  $\rho$ , provides a linear-in-parameters Translog form

$$\ln y = \sum \alpha_i \ln x_i + \frac{\rho}{2} \left[ \sum \alpha_i (\ln x_i)^2 - \sum \alpha_i \ln x_i \right]^2$$

with second-order terms 
$$\alpha_{ii} = \frac{\rho}{2} \alpha_i (1 - \alpha_i)$$
 and  $\alpha_{ij} = \frac{\rho}{2} \alpha_i \alpha_j$   $i \neq j$ 

Fuss, McFadden and Mundlak (1979: 239) give the following general derivation of the Kmenta Taylor expansion from the Translog function.



#### APPENDIX 12: REPRESENTATIONS OF TECHNICAL PROGRESS

Consider a two-factor Cobb-Douglas production structure:

$$Q_t = AK_t^{\alpha}L_t^B$$

with Q output, K capital and L labour.

Production models have to allow technology to improve over time in order to explain growth in output with stable populations and growth in the presence of diminishing returns to scale production structures.

A number of alternative representations of technical progress have been proposed.

#### A12.1 Neutral technical progress

One way of handling technical progress is to make the efficiency parameter A, vary over time so that:

$$Q_t = A(t)K_t^{\alpha}L_t^{\beta}. \tag{A12.1}$$

Before equation A12.1 can be estimated, some form has to be given to the function A(t). The form most frequently used in practice is:

$$A(t) = Ae^{gt}$$

where A and g are constants, so that equation  $A12.1^{1}$  becomes

$$Q_t = Ae^{gt}K_t^{\alpha}L_t^B. (A12.2)$$

A, also known as Hicks-neutral productivity, represents the value of A(t) at time = 0. Partially differentiating equation A12.2 with respect to t yields

$$\frac{\partial Q_t}{\partial t} = gQ_t.$$

Hence,

$$\frac{\partial Q_t}{\partial t} / Q_t = g .$$

A popular alternative application of this is the labour-augmenting or Harrod-neutral technical progress where  $L_t = A(t)L_{at}$ , with  $L_{at}$  actual labour input. In its simplest case  $A(t) = Ae^{gt}$ , reflecting an exponential time trend (Hall and Nixon 1997; Turner, Richardson and Rauffet 1993). An extension of this where labour is set to be human capital-augmented, i.e.  $A(t) = Ae^{\varphi(S)}$  with  $\phi(S)$  reflecting the efficiency of a unit of labour with S years of schooling relative to one with no schooling  $(\phi(S) = 0)$  (Hall and Jones 1996).



Thus g measures the proportionate change in output per time period when input levels are held constant. Put differently, g is the proportionate change in output that results from technical progress.

It is convenient to estimate equation A12.2 in logarithmic form as it simply requires the inclusion of a time trend in the usual Cobb-Douglas estimation equation:

$$\log Q_t = \log A + gt + \alpha \log K_t + \beta \log L_t. \tag{A12.3}$$

Although production functions with neutral specifications of technical progress are most commonly used in practice, they have definite limitations.

#### (i) Constancy of g

The implication of a constant g that technical progress occurs at a constant rate may not be realistic.

#### (ii) Neutrality of the technical progress

The neutrality of the technical progress implies that it has no effect on the marginal rate of substitution of capital for labour<sup>2</sup> and hence for a given ratio of factor prices does not influence the proportions in which capital and labour inputs are combined.<sup>3</sup> Thus, such technical progress does not affect the capital or labour intensity of the productive process, while it is most probable that technical innovations will either be labour-saving or capital-saving.

A non-neutral technical progress will permit the  $\alpha/\beta$ -ratio to vary over time.

#### (iii) Technical progress exogenous and disembodied

Technical progress in this case is exogenous because it has been superimposed on the system. (A is simply assumed to grow over time for no stated reason.)<sup>4</sup>

Disembodied technical progress is a form of exogenous technical progress, implying that all existing factors of production are transformed, no matter how long these factors have

$$MRS = \left(\frac{\beta}{\alpha}\right) \left(\frac{K_t}{L_t}\right)$$

The MRS remains constant for any K/L ratio. This occurrence is known as the Hicks-neutral technical progress. It is also Harrod-neutral, since for any Q/K ratio, the marginal product of capital is left unchanged.

· ·

For equation A12.2 it is still true that  $\partial Q_t/\partial K_t = \alpha(Q_t/K_t)$  and  $\partial Q_t/\partial L_t = \beta(Q_t/L_t)$  so that

This form of technical progress means that the isoquants are all shifted towards the origin, but their slopes at the point where they meet any ray from the origin (i.e. for any K/L ratio) remain unchanged.

Extensions of this, such as the modelling of effects of learning and of research and development expenditure, have been developed.



been in existence. This is clearly unrealistic, certainly as far as capital inputs are concerned. The occurrence of some new invention does not normally mean that all existing capital machinery, no matter of what age, can now be fully adapted to take advantage of the new technique (Thomas 1993).

#### A12.2 Alternative representations of technical progress

#### (i) Vintage models

The estimation of the so-called vintage models was an attempt to deal with the disembodied nature of production functions with a neutral technical progress specification.

The first rigorous attempt to formulate a model of embodied technical progress was that of Solow (1960). In the Solow model, technical progress proceeds at a constant rate g, but affects only newly produced capital goods. Separate production functions exist for machines of different vintages. Thus, if  $Q_v$  is the output produced by machines of vintage v (i.e. constructed in year v),  $K_v$  is the number of machines of that vintage, and  $L_v$  is the labour employed on such machines. Then (assuming a basic Cobb-Douglas form) the production function for machines of vintage v is

$$Q_{\nu} = Ae^{g\nu}K_{\nu}^{\alpha}L_{\nu}^{B} \tag{A12.4}$$

where g is the rate of technical progress. Capital stock of vintage v,  $K_v$ , is dependent on investment in machines in the year v and the rate (assumed constant) at which machines depreciate. Total output is the sum of all outputs obtained from machines of all vintages and Solow was able to derive an aggregate production function in which the normal capital stock variable was replace by an index of "effective capital stock". This index was a weighted sum of all machines with weights declining with age.

#### (ii) Developments based on cointegration

Budd and Hobbis (1989) have applied cointegration analysis to the UK production function and in particular to address the problem of how best to represent technical progress. They argue that there are two main sources of technical advance. First, it may come through domestic research effort which they proxy by the number of new patents taken out in the US by UK residents. Secondly, new technology can be imported from abroad and this flow is proxied by imports of new machinery and by royalty payments to foreign countries. These flows are converted into "net stock of technology" variables by a simple cumulating process in which given depreciation rates are assumed.

Capital stock figures are adjusted to take into account that proportion which consists of recently imported machinery (which may be assumed to be technically superior, otherwise it would not have been imported). "Quality-adjusted" capital stock, K is

.



defined as  $K^* = K(M/K)^{\gamma}$  where M is the net stock of recently imported machinery and K is the unadjusted capital stock.

A technology index, T, is then defined as  $T = aP^{\theta}R^{\phi}$  where P and R are the patents and royalties variables respectively.

A Cobb-Douglas production function of the kind  $Q = A * TK *^{\alpha} L^{B}$  is assumed which, after substitution becomes

$$Q = AL^{\beta} [K(M/K)^{\gamma}]^{\alpha} P^{\theta} R^{\phi} \text{ with } A = A * a.$$
 (A12.5)

If the technology variables are unimportant, then  $\gamma = \theta = \phi = 0$  and the function reduces to the standard Cobb-Douglas form:

$$Q = AK^{\alpha}L^{\beta}. \tag{A12.6}$$

Budd and Hobbis tested standard forms based on equation A12.6 (with and without a time trend included to represent disembodied technical progress) for cointegration. These production functions invariably failed the cointegration tests and frequently had implausible coefficients. It was possible to find a cointegrating vector only when the technology variables were added to the equations. Hence Budd and Hobbis conclude that the long-run production function is of the form (A12.5) rather than (A12.6) and that technical progress is better represented by their proxy variables than by the traditional time trend.



# APPENDIX 13: THE NAIRU/NAWRU CONCEPT: THEORETICAL AND MEASUREMENT ISSUES

#### A13.1 Basic concept and theoretical model

According to modern labour market literature, NAIRU is defined as the rate of unemployment at which inflation stabilises in the absence of wage-price surprises. Conventional thinking about the equilibrium unemployment rate assumes that, in the long-run NAIRU, is determined solely by supply-side factors of the labour market. However, Pichelmann and Schuh introduced an alternative hypothesis where hysteresis-mechanisms could lead to permanent shifts of equilibrium unemployment over time, implying that an unique long-run NAIRU may not even exist.

Consider the following simple formal exposition of equilibrium unemployment given by Pichelmann and Schuh (1997):

Firms set prices as a mark-up over expected wages:

$$p - w^e = \beta_0 - \beta_1 u \tag{A13.1}$$

with  $\beta_0$  denoting "price-push" factors (e.g. oil shocks, productivity slowdown) and with the mark-up depending (at least in the short-run) upon the state in the labour market.

Workers demand wages in relation to existing prices:

$$w - p^e = \gamma_0 - \gamma_1 u {A13.2}$$

with  $\gamma_1 > 0$  because of union bargaining or efficiency wage considerations, and with  $\gamma_0$  denoting "wage-push" factors (e.g. unemployment benefits, union power).

Solving for u yields

$$u = (\beta_0 - \gamma_0)/(\beta_1 + \gamma_1) - 1/(\beta_1 + \gamma_1)[p - p^e + w - w^e]. \tag{A13.3}$$

When there are no wage-price surprises

$$u = (\beta_0 - \gamma_0)/(\beta_1 + \gamma_1) = u^*$$

with  $u^*$  denoting the no-surprise equilibrium rate of unemployment.

Assuming real wages to equal expected real wages and expected inflation equals last year's inflation, give the underlying inflation-unemployment trade-off (standard Phillips-Curve):

$$\Delta \pi = \theta_1 (u^* - u) . \tag{A13.4}$$



Inflation will only remain unchanged in this setting, when actual unemployment equals  $u^*$ . Thus, the effects of aggregate monetary and fiscal policy, as well as of other types of demand shocks, are in the long-run constrained by this fundamental supply-side relationship. From this perspective, the only sustainable way to bring down unemployment is to reduce  $u^*$ .

The model is usually closed by introducing a conventional downward sloping aggregate demand schedule whereby lower prices elicit higher demand via real balance effects (and/or lower interest rates, improved competitiveness). Thus, demand disturbances may lead to cyclical unemployment, i.e., deviations of actual unemployment from its equilibrium level as defined above.

In the simple framework outlined above, movements in unemployment can be caused by shifts in aggregate demand giving rise to cyclical unemployment, and by shifts in the price or wage-setting schedules which change equilibrium unemployment.

The textbook theory claims that negative (positive) demand disturbances may temporarily push actual unemployment above (below) its equilibrium level, but over the medium term the ensuing process of disinflation (inflation) will inevitably drive unemployment back to equilibrium. The conventional approach then continues to argue that the degree of nominal inertia is simply not high enough to explain the sustained increase in unemployment in Europe. Thus, the theory concluded there must have been unfavourable shifts in the fundamental supply-side determinants of the NAIRU. The policy implication then, of course, is to press for supply-side reform.

However, despite considerable efforts, it has been hard to identify changes in the basic determinants of equilibrium unemployment large enough to account for the observed trend increase in actual unemployment. Consequently, the alternative hypothesis has been put forward that unemployment may be strongly dependent on its own history (hysteresis). According to this view, current equilibrium unemployment is not independent of past actual unemployment, because of endogenous mechanisms that tend to translate movements in actual unemployment into changes of equilibrium unemployment. The presence of such mechanisms blurs the simpleminded distinction between demand and supply factors because demand shocks eventually have longer-term supply-side consequences.

The distinguishing feature of a process characterised by hysteresis is that the behaviour of the process cannot be described by reference to state variables alone. Instead, in addition to state variables the past history of the process has to be invoked in order to explain its behaviour.

The following simple technical exposition is used to show how traditional economic thinking about the trade-off between unemployment and inflation is altered when the evolution of unemployment is subject to hysteresis effects. As a starting point, consider the following general formulation of the Phillips curve:

$$\pi = \pi^e + \theta_1(u^* - u) \tag{A13.5}$$



where  $\pi$  and  $\pi^e$  denote, respectively, the actual and expected rates of inflation and u is the rate of unemployment. Equilibrium unemployment (NAIRU) corresponds to the steady-state situation when actual inflation is equal to expected inflation, so that  $u = u^*$ . Then,  $u^*$  itself is usually assumed to be determined by a set of structural factors affecting the demand and supply-side of the labour market, but to be invariant with respect to business cycle conditions. Thus, denoting the relevant explanatory variables by X:

$$u^* = bX . (A13.6)$$

The possibility of hysteresis arises when equilibrium unemployment in a given period also depends on actual unemployment in the past, as e.g. in

$$u_{t}^{*} = \alpha u_{t-1} + bX . {A13.7}$$

In a steady state, where actual inflation is equal to expected inflation and unemployment is constant, equilibrium unemployment is now given by:

$$u^* = bX/(1-\alpha)$$
. (A13.8)

Therefore, when last period's actual unemployment is fully translated into equilibrium unemployment in the next period  $(\alpha = 1)$ , then steady-state equilibrium is no longer uniquely defined. Any change in actual unemployment, e.g. brought about by macroeconomic policies, would also alter the NAIRU by the same amount; such a situation has been labelled as pure hysteresis. When actual unemployment feeds only partly into future equilibrium unemployment  $(0 < \alpha < 1)$  there is persistence in unemployment in the sense that the NAIRU evolves only slowly towards its steady state (long-run) level ("speed-limits").

The conventional way to introduce hysteresis mechanisms<sup>1</sup> into the analysis is by adding into the wage equation an additional term denoting the change in unemployment:

$$w - p^e = \gamma_0 - \gamma_1 u - \gamma_2 \Delta u \tag{A13.9}$$

where in the case of pure hysteresis it is only the change term that matters. Thus, most explanations for hysteresis generating mechanisms focus on the behaviour of labour market participants, changes in their productive capacity caused by unemployment, and on the resulting consequences for wage bargaining and the matching process between workers and jobs.

#### A13.2 Measurement issues

The general consensus among economists is that there exists, at least in the long-run, a unique "equilibrium unemployment rate" (i.e. the "NAIRU") which is consistent with stable inflation. In practice, rules for the conduct of monetary policy or programs to reduce unemployment are guided by empirical estimates of the NAIRU. The construction of estimates of the NAIRU,

See Pichelmann and Schuh (1997) for a survey on hysteresis mechanisms.



however, suffers from the fundamental problem that the NAIRU is an unobserved variable, so that there exist leeway for a broad range of plausible methodological approaches for the estimation of the equilibrium unemployment rate.

## A13.2.1 Time-series approaches

A widely used method to construct estimates of the NAIRU relies on time-series methods, which are based solely on data of the unemployment rate. Univariate methods proceed by decomposing the unemployment rate into a deterministic and a stochastic component. The deterministic component of the series is then interpreted as the equilibrium unemployment rate whereas the stochastic component represents the cyclical development of the unemployment rate. In order to obtain an estimate of the NAIRU, it has to be ensured that the deterministic part of the unemployment rate is uncorrelated to inflation. This approach has various advantages: it is easy to construct estimates of the NAIRU and theoretical issues (i.e. misspecification of the "model") can be avoided to a large extent.

The simplest univariate specification assumes that the unemployment rate is a realisation of a stationary process, with its expectations being the (time-constant) NAIRU:

$$u_t = \sum_{i=1}^{p} \phi_i u_{t-i} + e_t \tag{A13.10}$$

where the random variable  $e_t \approx iid$  with mean 0, satisfying:

$$E[u_t] = \overline{u} \tag{A13.11}$$

$$E[u_t - u_{t-1}] = 0. (A13.12)$$

This specification is in accordance to the basic model specified in section 1, which would, in the absence of any change of structural factors of the labour market, imply that the equilibrium unemployment rate is constant over time. A look at various countries' unemployment rates reveals, however, that actual unemployment rates exhibit considerable deviations from the long-run mean over time. According to the previous discussion this could be due to changes in the structural factors of the labour market, so that the equilibrium rate shifts from time to time. This possibility is taken into account when constructing estimates of the NAIRU which allow for "breaks" in the series, so that (A13.11) becomes:

$$E[u_t] = u_i \text{ if } t_{i-1} < t \le t_i; \quad i = 1, ..., I.$$
 (A13.13)

A problem arises within this model as the breaks are treated as being known with certainty. In practice however, it is difficult to determine the exact timing when the NAIRU might switch from one regime to another, so that an additional source of imprecision is added to these estimates.

However, since the 1980s a growing number of empirical studies suggest that the equilibrium unemployment rates may be described by non-stationary time-series, i.e. that they follow a stochastic trend so that:



$$E[u_t] = E[u_{i-1}] + \eta_t \tag{A13.14a}$$

or

$$u_t - u_{i-1} = a + \eta_t \tag{A13.14b}$$

with the parameter a representing the deterministic and  $\eta$  the stochastic component of the trend.

This implies that an "equilibrium" value to which the unemployment reverts in the long-run does not exist. This specification of the NAIRU concurs with the theoretical view, described in section 1, that hysteresis factors are at work in the labour market, i.e. that the NAIRU depends on the historical evolution of actual unemployment.

Structural Time Series Models as proposed by Harvey (1989) represent an appropriate methodological tool to construct estimates of the (unobserved) stochastic components of unemployment rates. The models assume that the NAIRU may be driven by simple but flexible stochastic processes. A standard specification of a univariate structural time-series model of the NAIRU is:

$$u_{t} = u_{t}^{*} + uc_{t} + i_{t}$$
 (A13.15a)  
 $u_{t}^{*} - u_{t-1}^{*} = a_{t-1} + e_{t}$  (A13.15b)  
 $a_{t} = a_{t-1} + \eta_{t}$  (A13.15c)

where u is actual unemployment,  $u^*$  represents the trend unemployment rate, which may then be interpreted as the NAIRU, uc is the cyclical unemployment rate, which follows a stochastic cycle.

## A13.2.2 Wage-price models

The fundamental drawback of the time-series approach for the estimation of the NAIRU is that it does not provide causal explanations for the development of the "equilibrium" unemployment rate. Estimates based on the univariate time-series methodology therefore form no sound basis for policy interventions as they leave the interactions between the economic variables indeterminate.

Econometric models based on the theoretical model describe in section 1 take the interdependence between economic variables into account. Empirical results obtained form these models thus allow for causal interpretations of the NAIRU estimates. In contrast to time-series models, movements of the NAIRU are explained by various labour market variables (i.e. wage or price pressure elements) which are inserted into the empirical models.

One commonly applied method to estimate the equilibrium unemployment rate is based on the wage equation described in section 1(A13.9), written in log-linear form:



$$\Delta(\log(w_t) - \log(p_t)) = \beta_0 + \beta_1(L)(\log(\pi_t) - \log(\pi_t^e)) - \beta_2(U_t - NAWRU_t) + \beta_3\Delta U_t + \beta_4\log(x_t) + \beta_5\log(A_w) + \omega_t$$
(A13.16)

where NAWRU represents the equilibrium unemployment rate, x depicts productivity and  $Z_w$  are variables representing wage pressure elements (such as unemployment benefits, taxation, labour market mismatch, employment protection, etc.) and the change of the unemployment rate is inserted to capture possible hysteresis effects. Thus, the NAWRU may move over time due to changes in wage pressure elements or as a consequence of hysteresis effects. As this specification relates wage inflation rather than price inflation to movements of the unemployment rate, corresponding estimates of the equilibrium unemployment rate are referred to as the NAWRU (non-accelerating wage rate of unemployment). By imposing the long-run homogeneity restriction, namely that real wage growth must be proportional to productivity growth, this specification implies that it is possible to analyse the long-run equilibrium properties of the equilibrium unemployment rate.

In order to construct an estimate of the NAWRU, a model for inflationary expectations has to be developed for the estimation of the equation. A commonly used approach is to use lagged inflation rates as a proxy for "price surprises". Alternatively, some consensus or median forecast of inflation can be used in order to depict "expected inflation". Another commonly used possibility to construct estimates of the NAIRU is to use the standard Phillips-curve relation (A13.4):

$$\pi_{t} - \pi_{t}^{e} = \beta(L)(u_{t-1} - NAIRU_{t}) + \delta(L)(\pi_{t-1} - \pi_{t-1}^{e}) + \gamma(L)X_{t} + e_{t}$$
(A13.17)

where  $\pi^e$  represents expected inflation, L is the lag operator and X represents additional regressors included in some empirical specifications. As in the case of the NAWRU-estimates described above, the need for a model of inflationary expectations arises. This specification forces the equilibrium unemployment rate, the NAIRU, to satisfy the steady-state condition that expected inflation must equal actual inflation. A drawback of the Phillips-Curve formulation is that "surprises" of nominal wage inflation (deviations of actual wages from their expected values) have to be treated as non-existent.

The discussion in section 1 indicated that both price setting and wage formation incorporate important information on the development of the NAIRU. It should thus be expected that estimates of the NAIRU can be improved by analysing both price and wage setting. This can be achieved by combining the wage setting curve (A13.16) with the price setting schedule (A13.17):

$$D(\log(p_t) - \log(w_t)) = a_0 - a_1(L)D(\log(w_t - \log(w_t^e)) + a_3\log(y_t) - a_4\log(Z_{p_t}) + e_t \quad \text{(A13.18)}$$

where y is the level of output market activity and  $Z_p$  captures "price pressure" variables as describe in section 1.

The NAIRU is then estimated by simultaneous estimation of equations (A13.16) and (A13.18) and solving for unemployment. This method allows the imposition of homogeneity restrictions



both on price setting and on wage formation and thus implies that the estimate of the NAIRU satisfies the necessary conditions for the labour market equilibrium as described in section 1. The estimation of a wage-price system allows for the distinction between the impact of structural factors on wage formation and price setting respectively.

Although there exists a great number of empirical estimates of the NAIRU, there is an apparent lack of discussion about the precision of these estimates. In fact, two fundamental types of uncertainty may contribute to the imprecise measurement of the equilibrium unemployment rate. The first source of uncertainty arises from the fact that the NAIRU is an unobserved variable that leaves room for a number of plausible empirical models for its measurement. Different specifications generally lead to different point-estimates of the level of the NAIRU. The exposition above provided an overview of different possible approaches to the measurement of the equilibrium unemployment, all of which concur with the theoretical model of the NAIRU. The most important difficulty in this context arises from the possibility that in the long-run, the level of the NAIRU may be indeterminate, rendering the NAIRU stochastic in nature. Examples are presented by Pichelmann and Schuh (1997).



#### APPENDIX 14: COST OF CAPITAL: A NEOCLASSICAL APPROACH

The pioneer of neoclassical theory, Jorgenson (1963), defines the cost of capital as the cost which the firm incurs as a consequence of owning an asset. The cost of capital transforms the acquisition price of an asset into an appropriate rental price. This cost depends on the rates of return and depreciation. The rate of return is the opportunity cost of holding capital goods rather than financial assets. Depreciation arises from the decline in the price of capital goods with age (Jorgenson 1993: 4).

The neoclassical theory of capital accumulation is formulated in two alternative yet equivalent ways. First, the firm may accumulate capital to supply the service to itself. The objective of the firm is to maximise its value, subject to its technical limitations. Secondly, the firm may rent the assets in order to obtain a capital service. In this case the objective of the firm is to maximise its current profit, defined as gross revenue less the cost of inputs less the rental value of capital. The rental can be calculated from the relationship between the price of new capital goods and the discounted value of future services received from these goods (Jorgenson 1993: 4).

According to Jorgenson (1993), in the absence of direct taxes, this relationship takes the form:  $q_{t} = \int e^{(-\tau)(s-t)c(s)-\delta(\theta-t)} ds;$ 

where r is the discount rate, q the price of capital goods, c the cost of capital services and  $\delta$  the rate of replacement (depreciation). The time of acquisition is given by t and time s is the time during which capital services are supplied (Jorgenson 1993: 4).

Differentiating this with respect to t gives  $c = q(r + \delta) - q$ , which is the rental price of capital services supplied by the firm to itself. Under static expectations about the price of investment goods, the rental price reduces to  $c = q(r + \delta)$ .

To extend the formula to allow for taxation, Jorgenson (1993) defines a depreciation formula D(s) to calculate the proportion of the original cost of an asset of age s, which may be deducted from taxes. Jorgenson also assumes a tax credit k that may be deducted from investment expenditure. If the tax rate is constant over time at rate u, the equality between the price of investment goods and the discounted value of capital services is:

$$q_t = \int e^{(-r)(s-t)} [(1-u)c(s)e^{-\delta(\vartheta-t)} + u(1-k)q(t)D(s)]ds + k q_t.$$

Allowing the present value of depreciation on one rand's worth of investment to be denoted by z gives:

1-1

$$z = \int e^{-rs} D(s) ds$$
.

The rental value of capital under static expectations then becomes:

$$c = q(r+\delta) \times \frac{(1-k)(1-uz)}{(1-u)}.$$



There are at least three depreciation formulae, which can be applied when calculating z. After calculating the cost of capital, it can be used as one of the most important determinants in a neoclassical investment function and subsequently a production function. By including this variable, any effect of a change in tax, interest rates or depreciation can be investigated. Through this variable, a tax reduction will, for example, influence investment behaviour and eventually aggregate supply (Jorgenson 1993: 4).

The effect of tax policy on investment behaviour enters the investment function through the rental value of capital. This results in a change in the desired level of capital. Such a change leads to net investment (or disinvestment), increasing (or decreasing) capital stock to its new desired level (Jorgenson 1993: 4).

The user-cost-of-capital (r) can be expressed as:

$$r = price \ of \ capital \left( \frac{(interest\ rate) - (inflation\ rate) + (rate\ of\ depreciation)}{1 - tax\ ratio} \right)$$

To summarise: this measure for user-cost-of-capital combines four effects. The first relates to an opportunity cost to invest and, based on the long-run nature of investment, is approximated by the yield on long-term government bonds. Second, fluctuations in the price of capital may lead to losses or gains for a firm when it sells its capital at the end of a period. A capital gain (loss) would reduce (increase) the user-cost-of-capital. These gains/losses can be approximated by the rate of inflation times the price of capital. A third cost for the owner of capital stems from the depreciation of capital, while the fourth component is taxes, resulting in a difference between the pre-tax and after-tax rates of return on capital.



## APPENDIX 15: LONG-RUN RESPONSE PROPERTIES

Figure A15.1(a) Long-run elasticities of the full supply-side model: 10% shock to real GDP at factor cost (actual and smoothed)

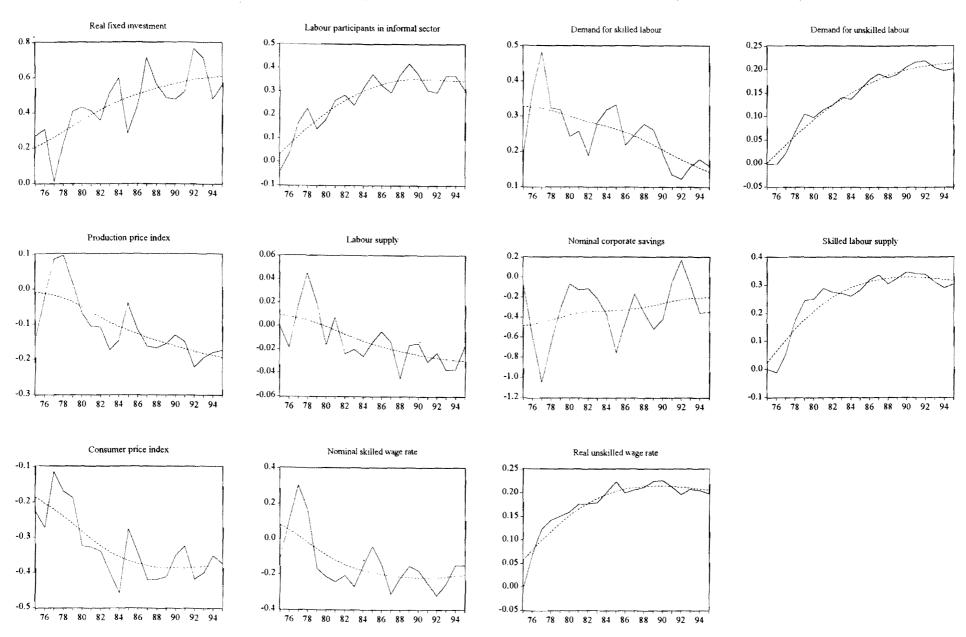




Figure A15.1(b) Long-run multipliers of the full supply-side model: 10 % shock to real GDP at factor cost in 1975 (actual and smoothed)

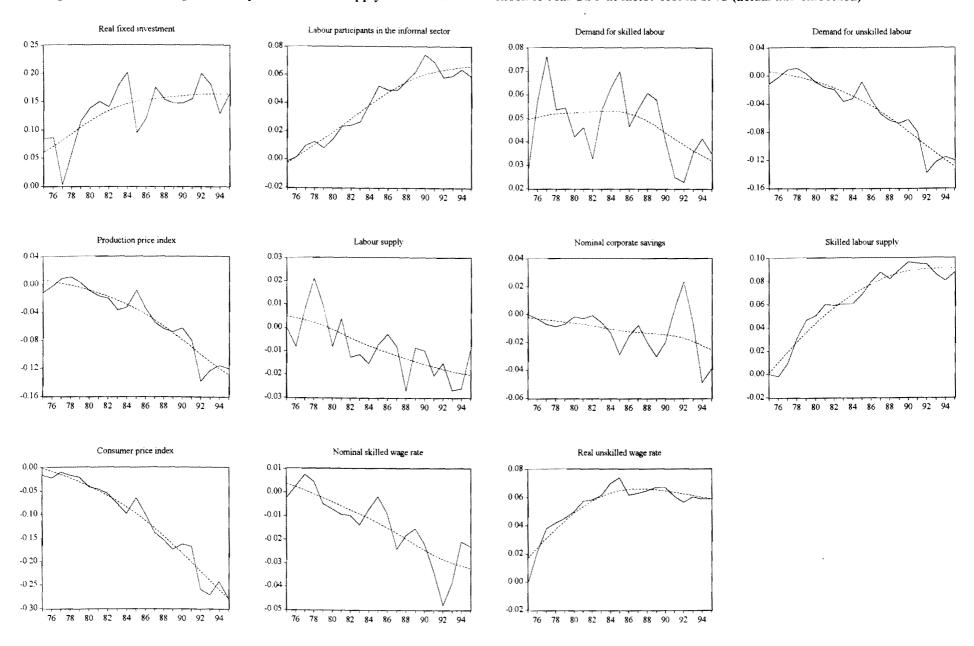




Figure A15.2(a) Long-run elasticities of the full supply-side model: 10% shock to real fixed investment (actual and smoothed)

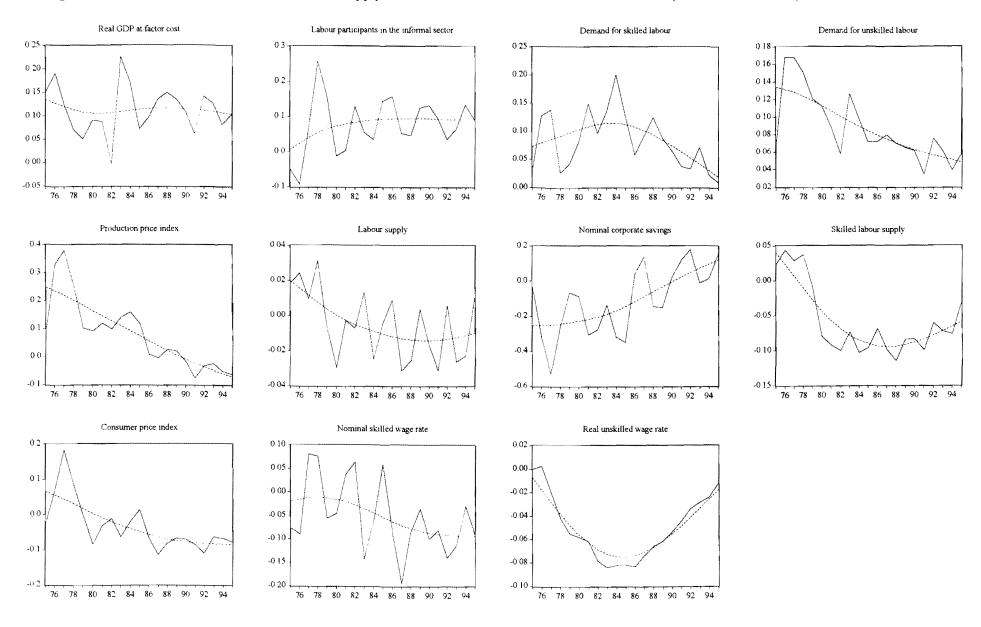




Figure A15.2(b) Long-run multipliers of the full supply-side model: 10% shock to real fixed investment in 1975 (actual and smoothed)

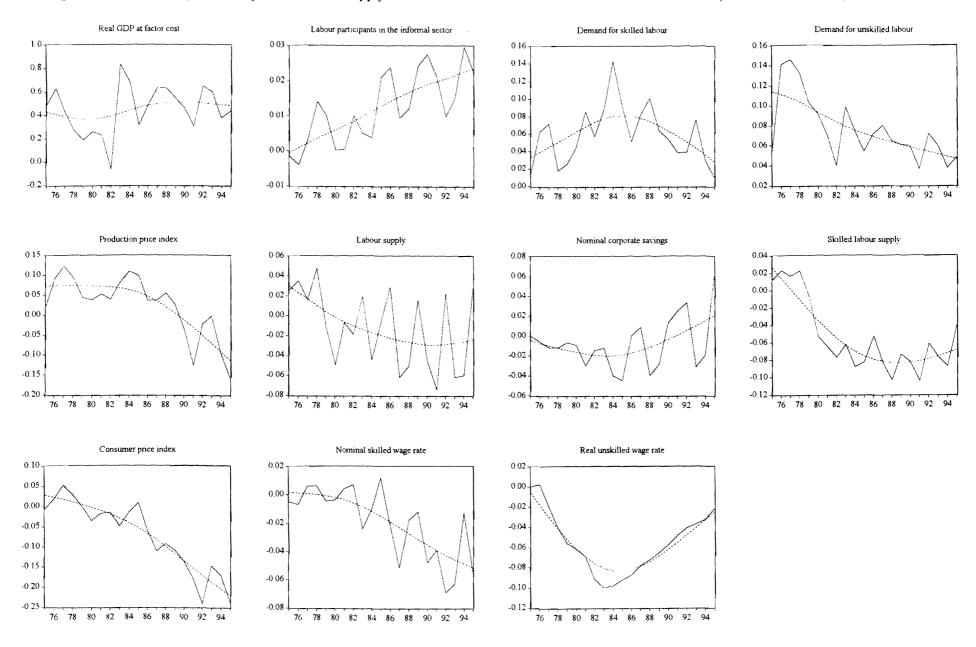




Figure A15.3(a) Long-run elasticities of the full supply-side model: 10% shock to nominal corporate savings (actual and smoothed)

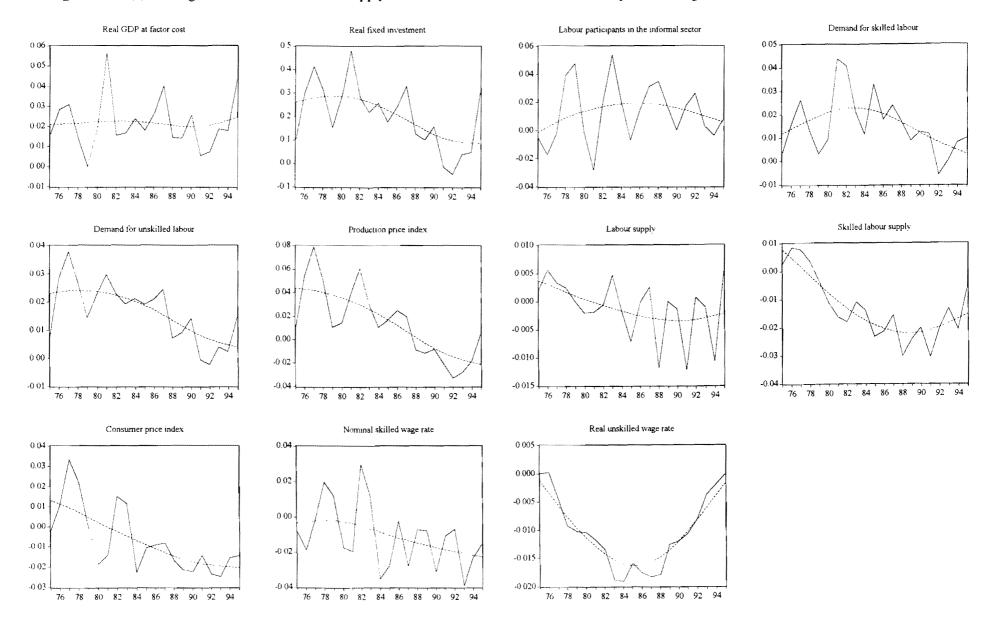




Figure A15.3(b) Long-run multipliers of the full supply-side model: 10% shock to nominal corporate savings in 1975 (actual and smoothed)

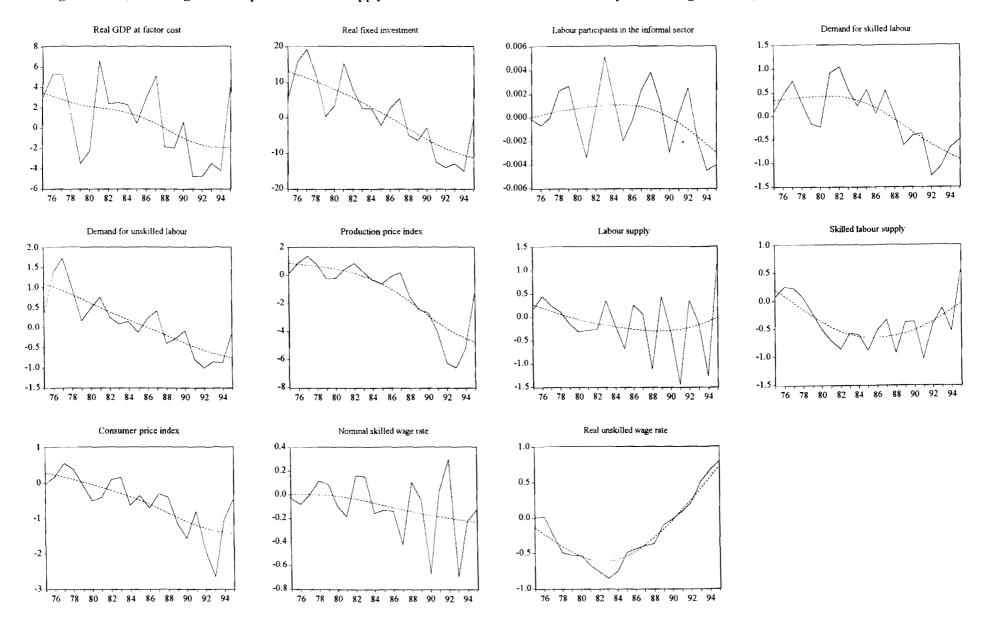




Figure A15.4(a) Long-run elasticities of the full supply-side model: 10% shock to demand for skilled labour (actual and smoothed)

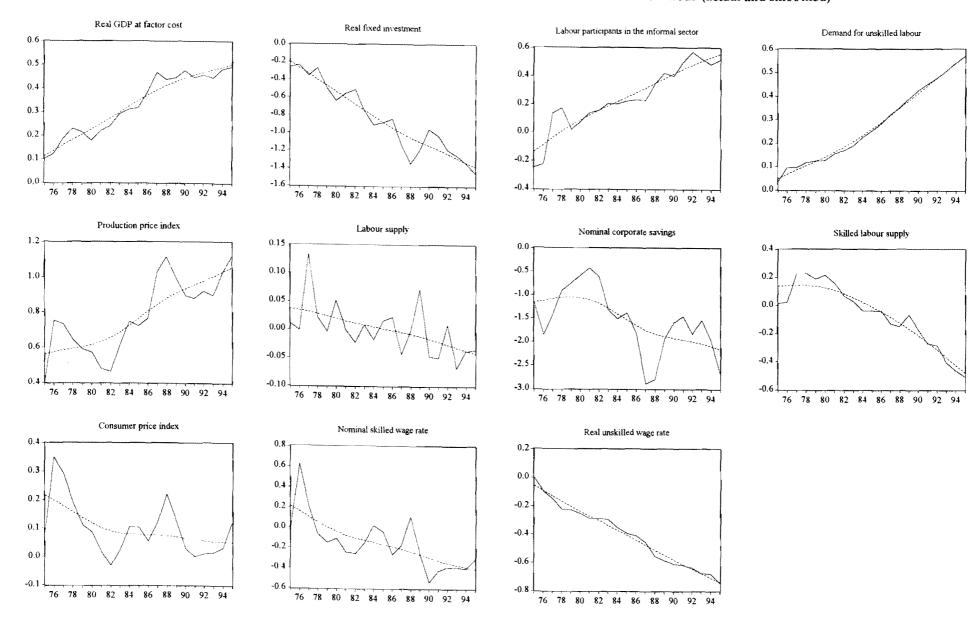




Figure A15.4(b) Long-run multipliers of the full supply-side model: 10% shock to demand for skilled labour in 1975 (actual and smoothed)

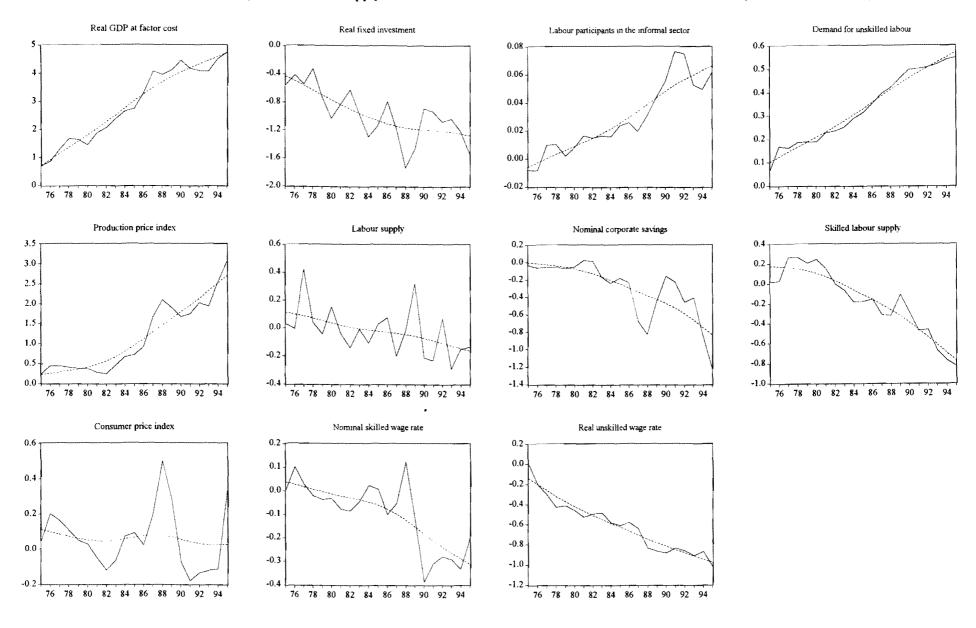




Figure A15.5(a) Long-run elasticities of the full supply-side model: 1% shock to demand for unskilled labour (actual and smoothed)

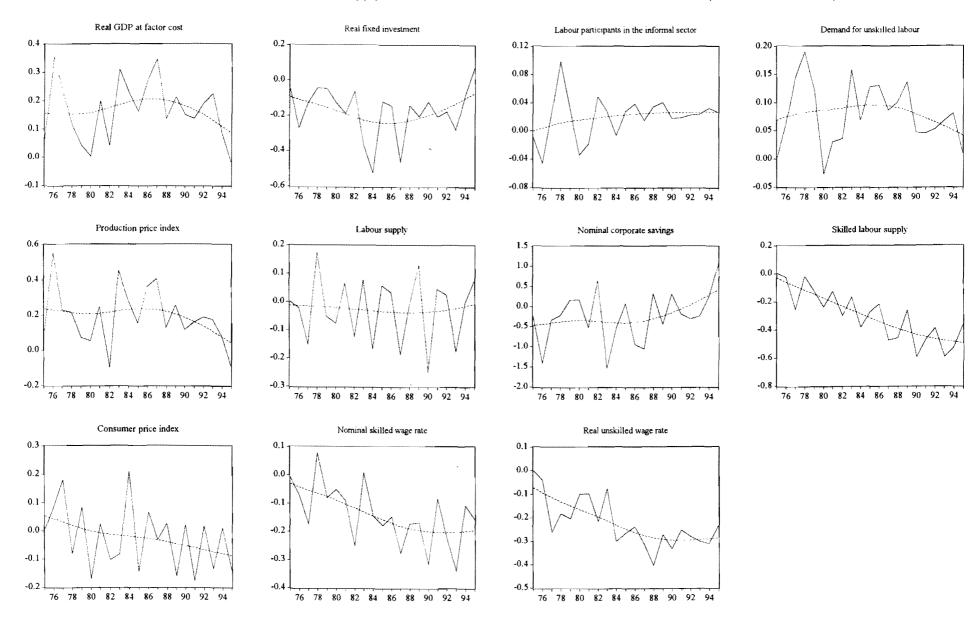




Figure A15.5(b) Long-run multipliers of the ful supply-side model: 1% shock to demand for unskilled labour in 1975 (actual and smoothed)

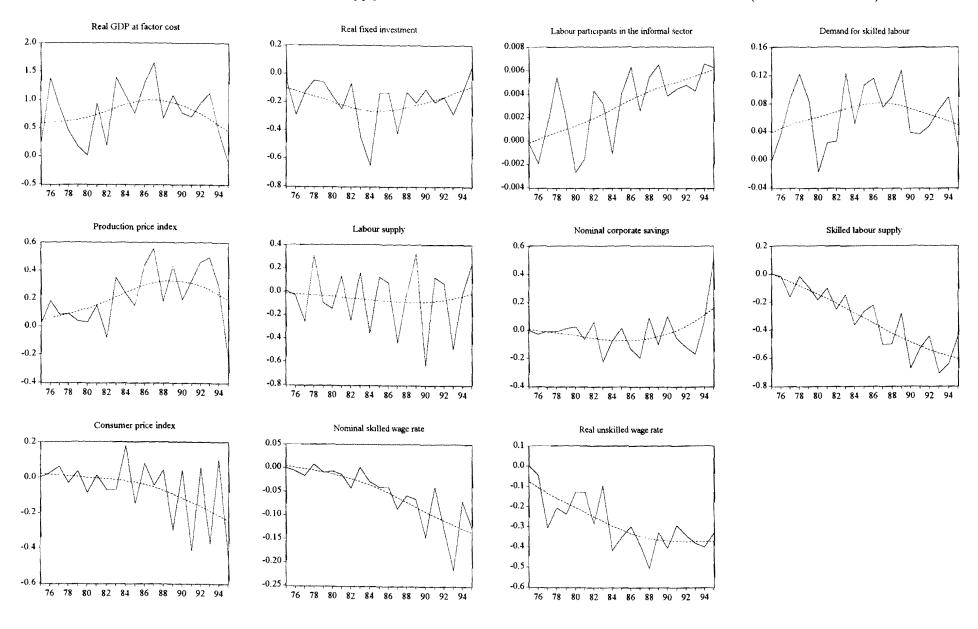


Figure A15.6(a) Long-run elasticities of the full supply-side model: 10% shock to labour supply (actual and smoothed)

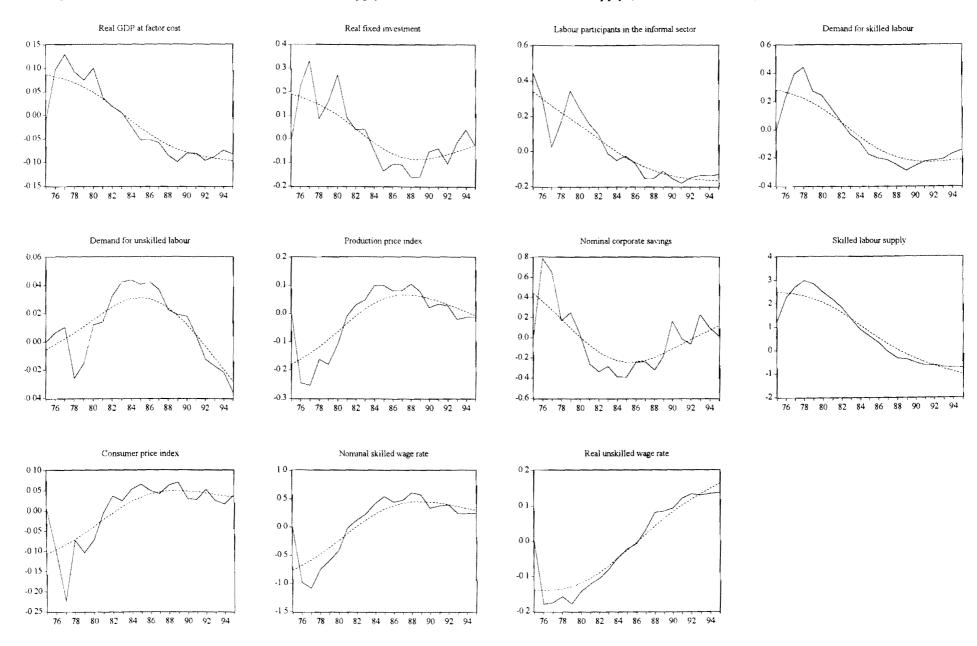




Figure A15.6(b) Long-run multipliers of the full supply-side model: 10% shock to labour supply in 1975 (actual and smoothed)

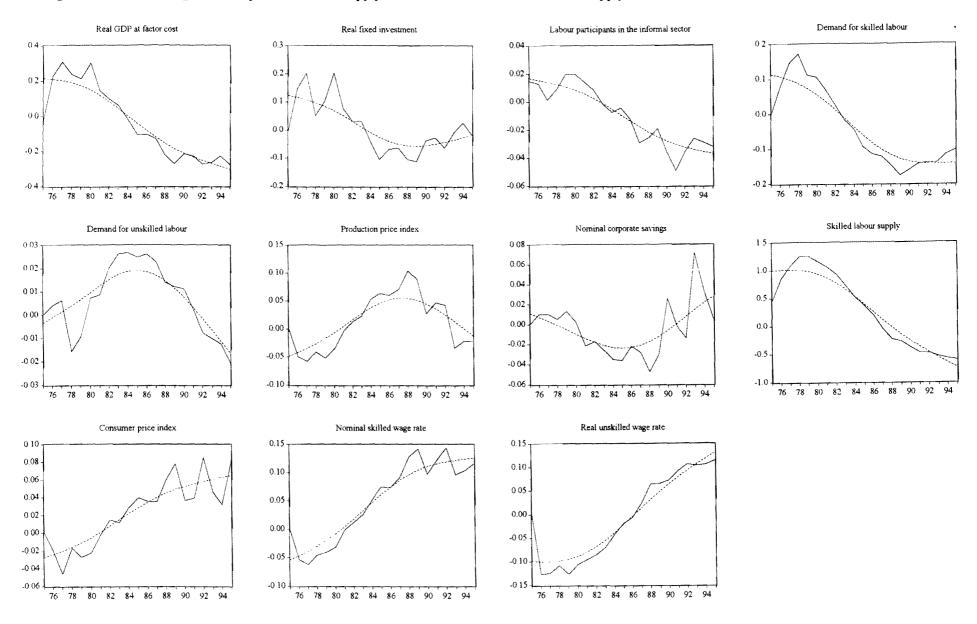




Figure 15.7(a) Long-run elasticities of the full supply-side model: 10% shock to skilled labour supply (actual and smoothed)

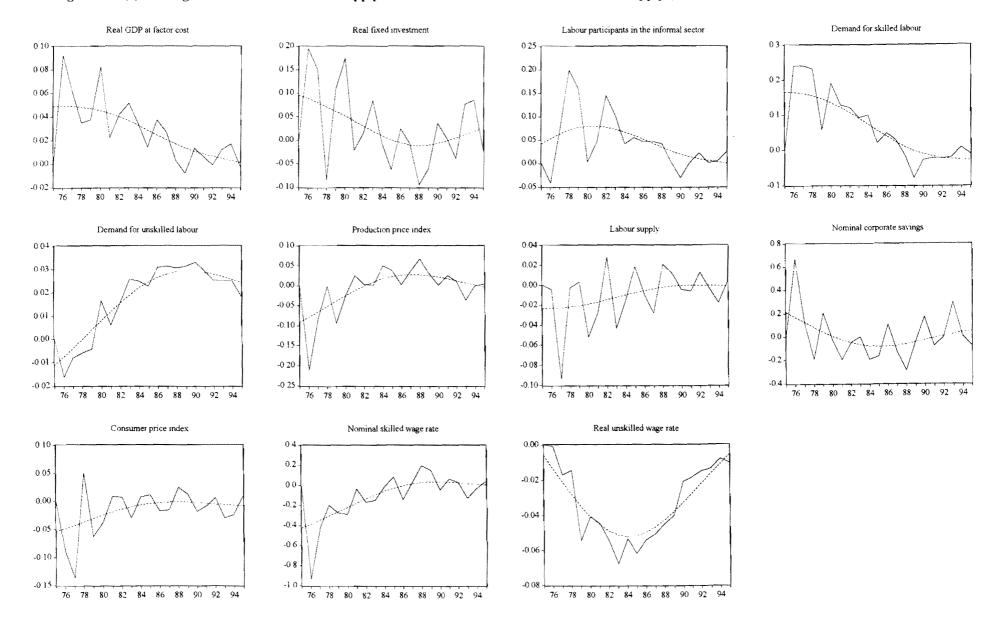




Figure A15.7(b) Long-run multipliers of the full supply-side model: 10% shock to skilled labour supply in 1975 (actual and smoothed)

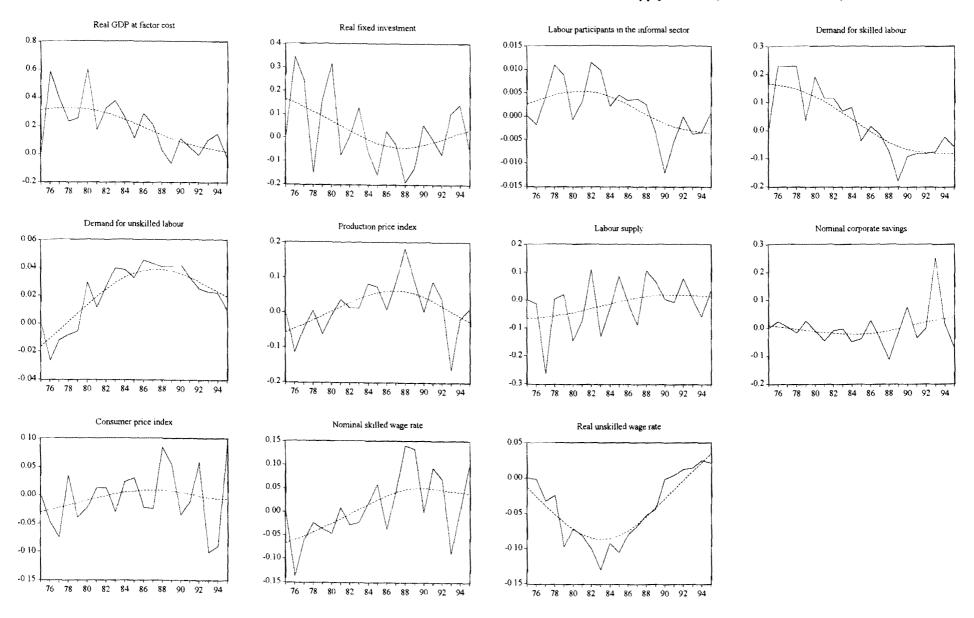




Figure 15.8(a) Long-run elasticities of the full supply-side model: 10% shock to nominal skilled wage rate (acutal and smoothed)

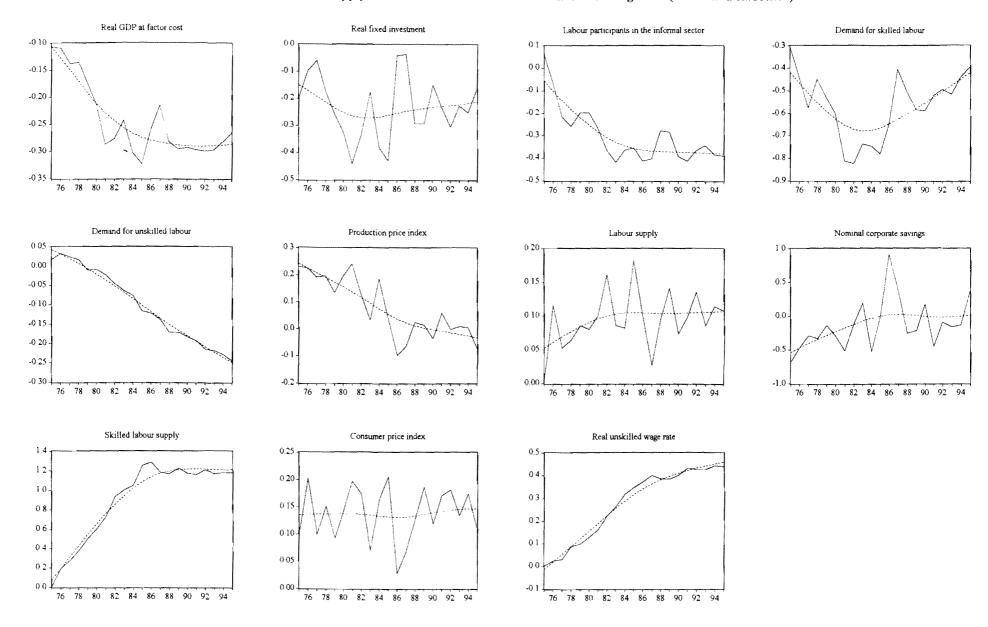




Figure A15.8(b) Long-run multipliers of the full supply-side model: 10% shock to nominal skilled wage rate in 1975 (actual and smoothed)

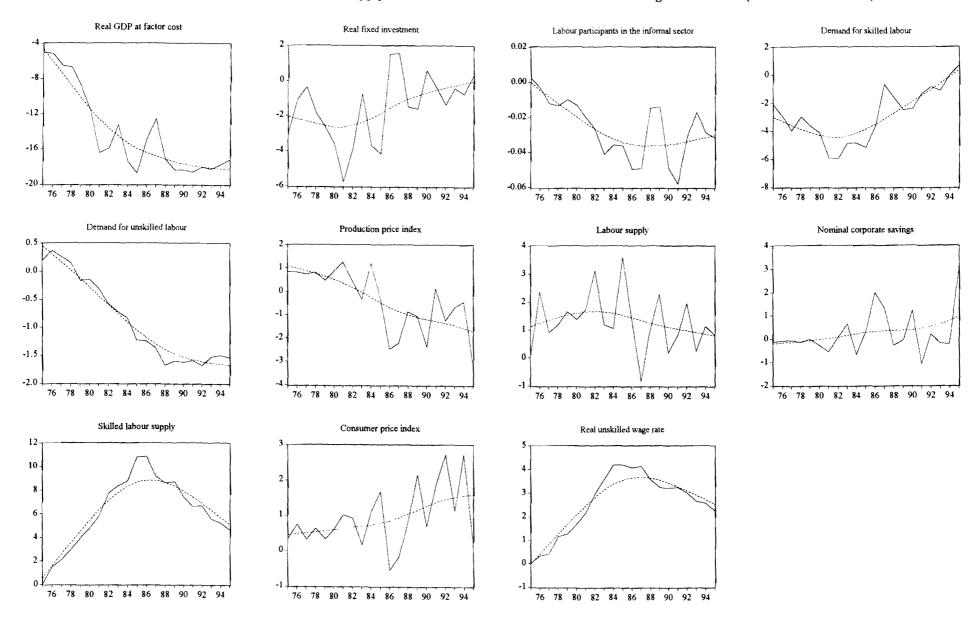
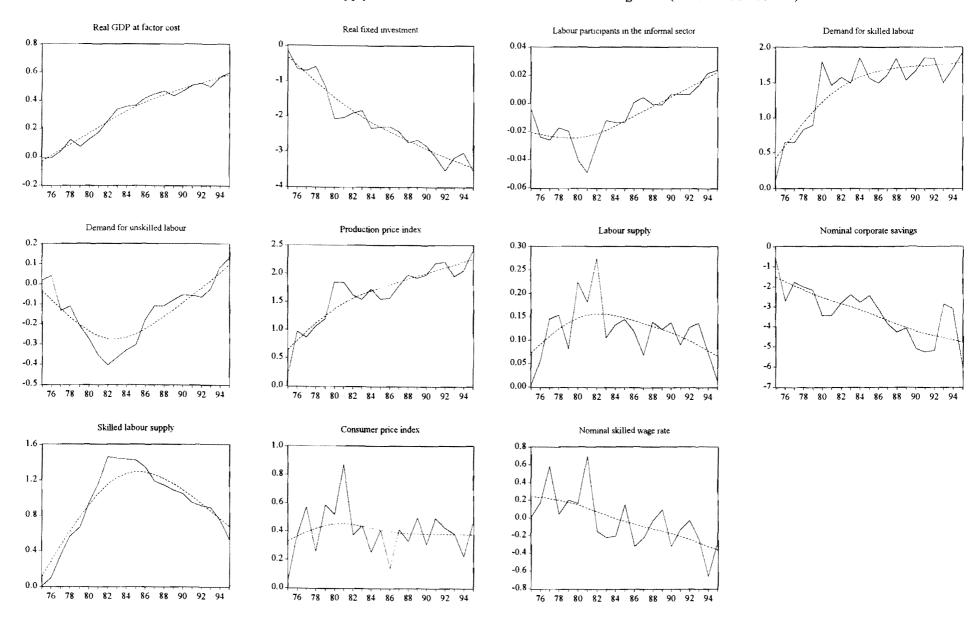




Figure 15.9(a) Long-run elasticities of the full supply-side model: 1% shock to real unskilled wage rate (actual and smoothed)



.

\_



Figure A15.9(b) Long-run multipliers of the full supply-side model: 1% shock to real unskilled wage rate in 1975 (actual and smoothed)

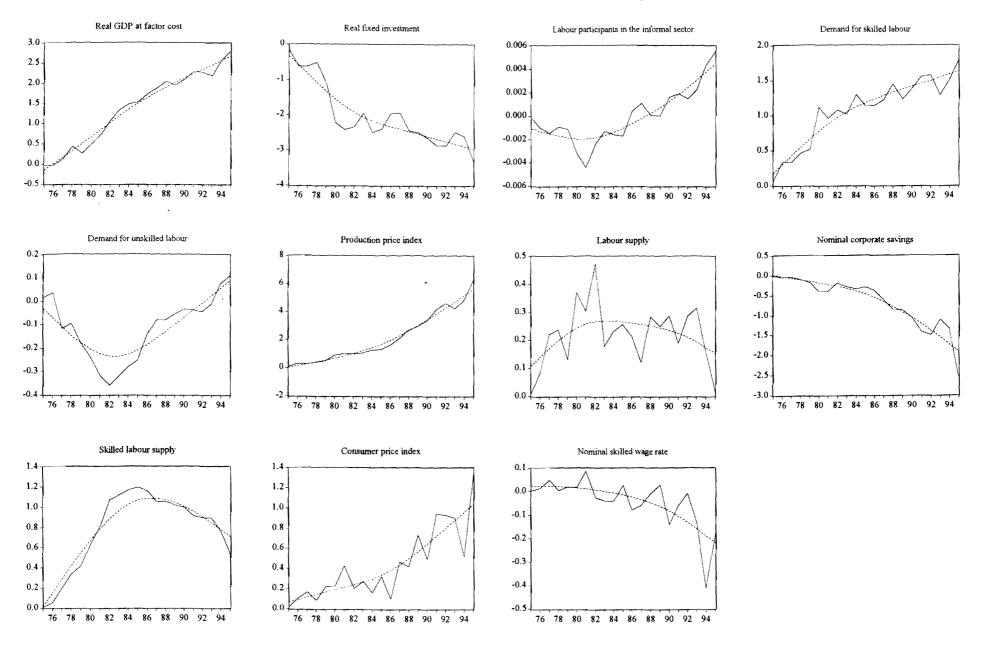
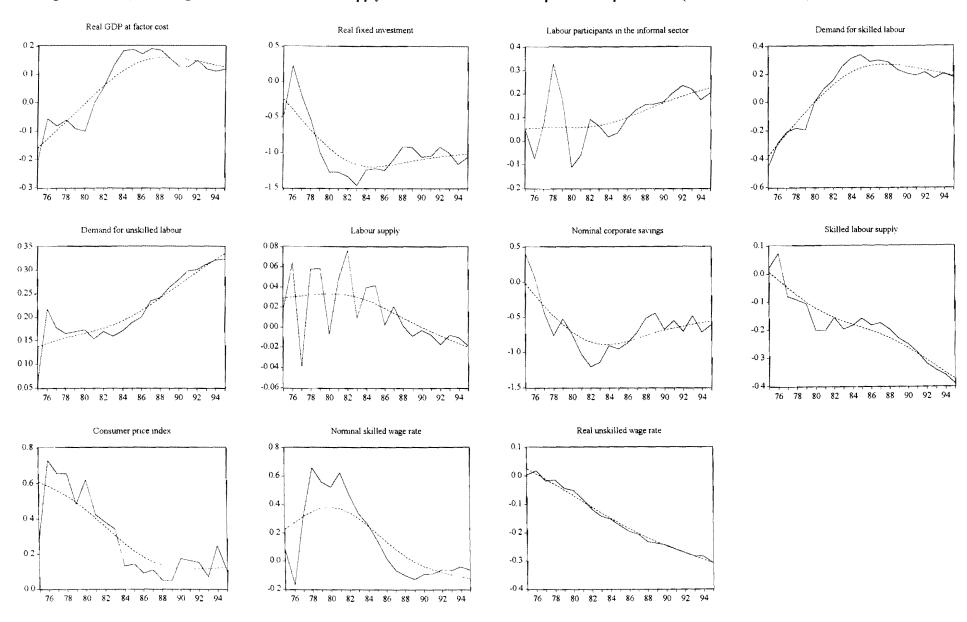




Figure A15.10(a) Long-run elasticities of the full supply-side model: 10% shock to production price-index (actual and smoothed)



-

\_



Figure A15.10(b) Long-run multipliers of the full supply-side model: 10% shock to production price index in 1975 (actual and smoothed)

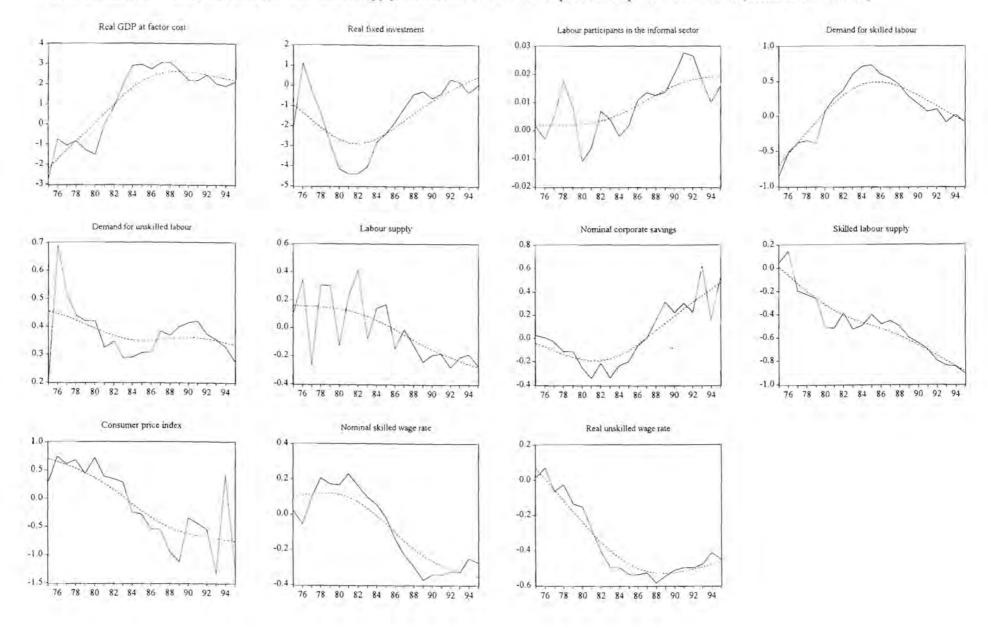




Figure 15.11(a) Long-run elasticities of the full supply-side model: 10% shock to consumer price index (actual and smoothed)

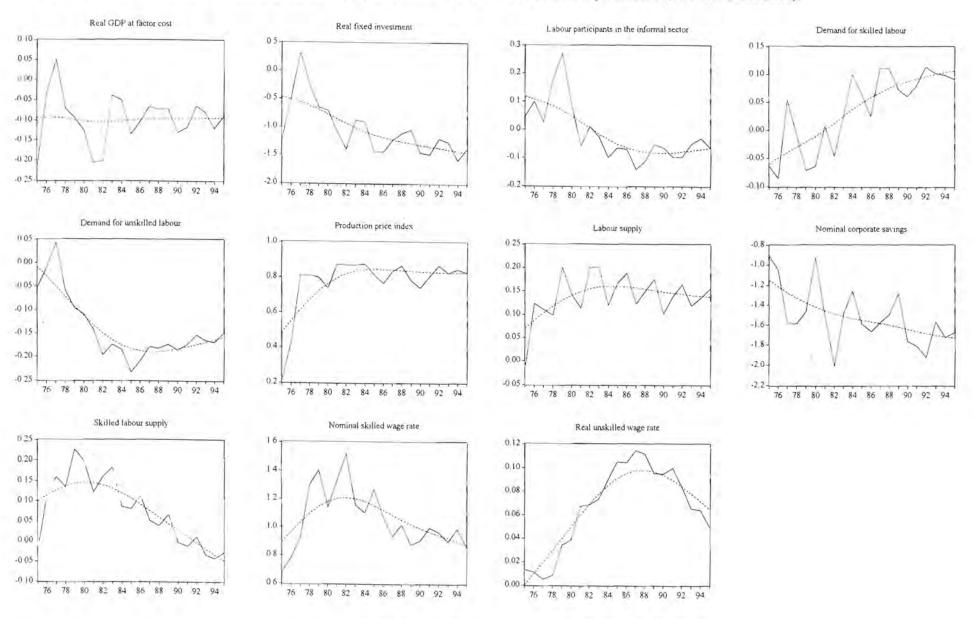




Figure A15.11(b) Long-run multipliers of the full supply-side model: 10% shock to consumer price index in 1975 (actual and smoothed)

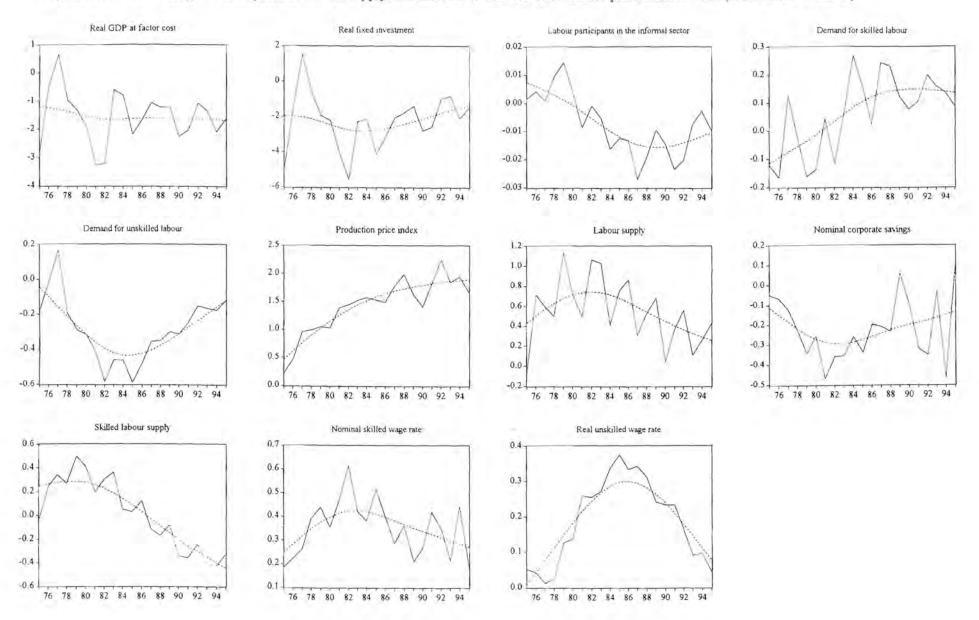




Figure A15.10(b) Long-run multipliers of the full supply-side model: 10% shock to production price index in 1975 (actual and smoothed)

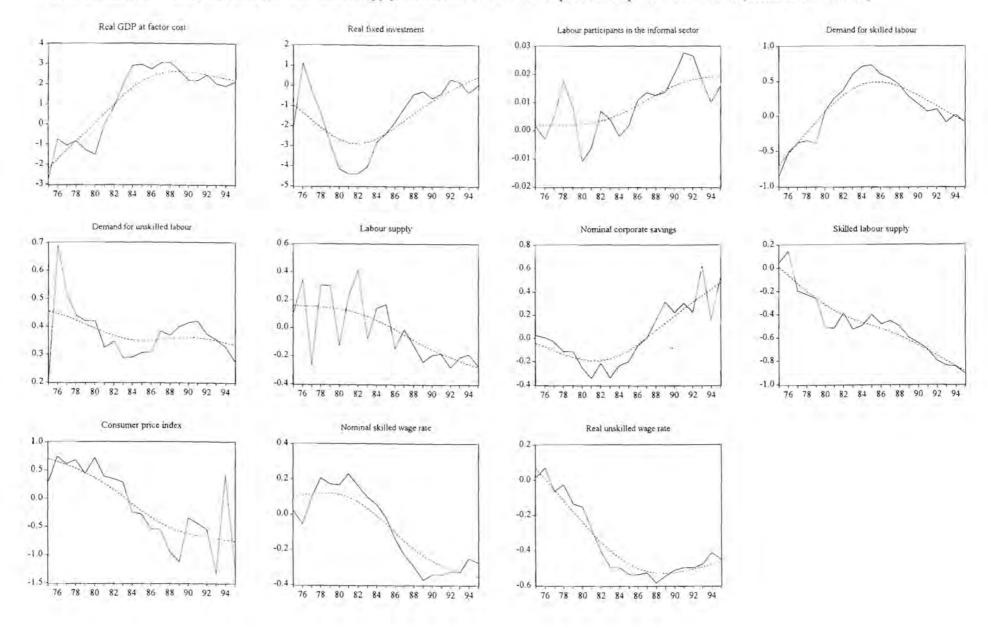




Figure 15.11(a) Long-run elasticities of the full supply-side model: 10% shock to consumer price index (actual and smoothed)

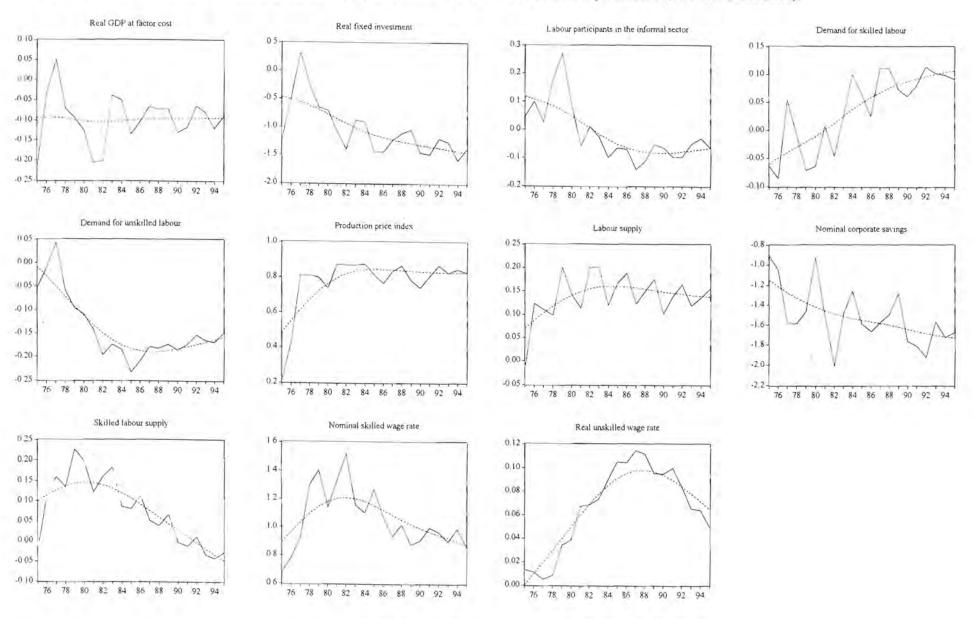
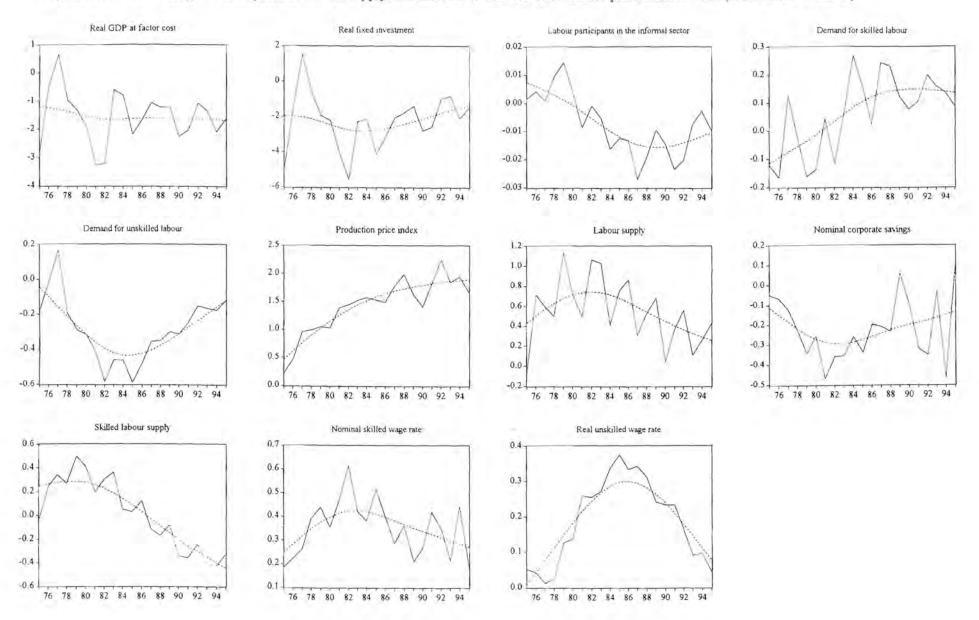




Figure A15.11(b) Long-run multipliers of the full supply-side model: 10% shock to consumer price index in 1975 (actual and smoothed)





## APPENDIX 16: POLICY SCENARIOS

## Figure A16.1(a) Individual scenario 0: Baseline and shocked simulation paths

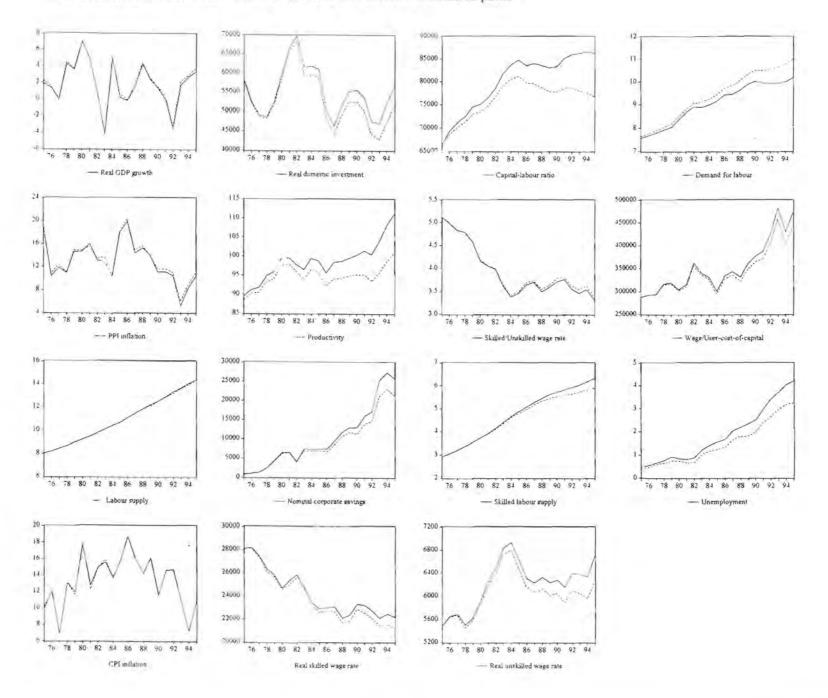




Figure A16.1(b) Individual scenario 0; Percentage differences between shocked and baseline simulation paths

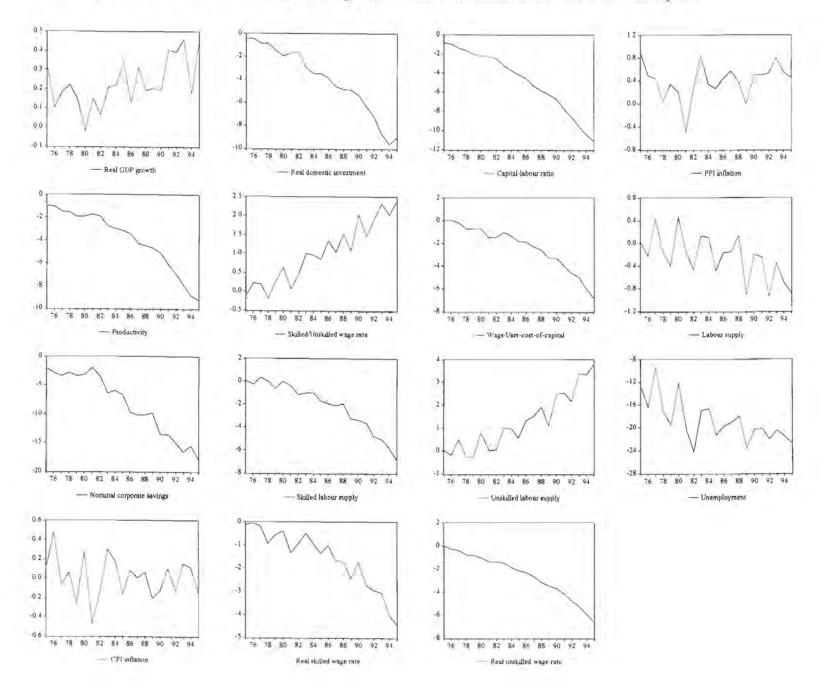




Figure A16.2(a) Individual scenario I: Baseline and shocked simulation paths

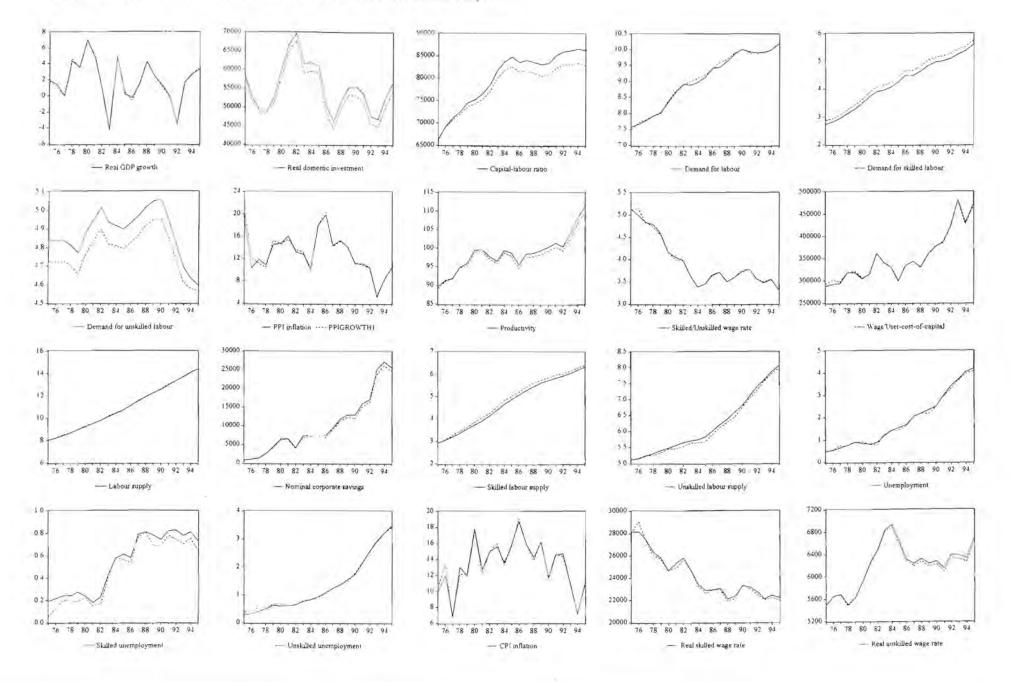




Figure A16.2(b) Individual scenario 1: Percentage differences between shocked and baseline simulation paths

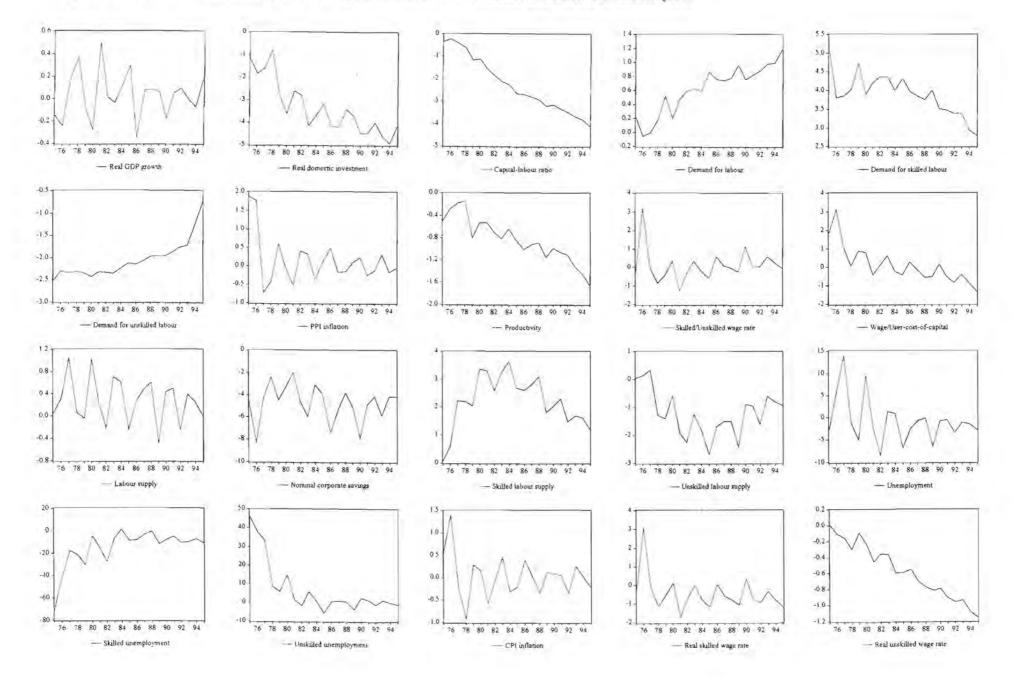




Figure A16.3(a) Individual scenario 2: Baseline and shocked simulation paths

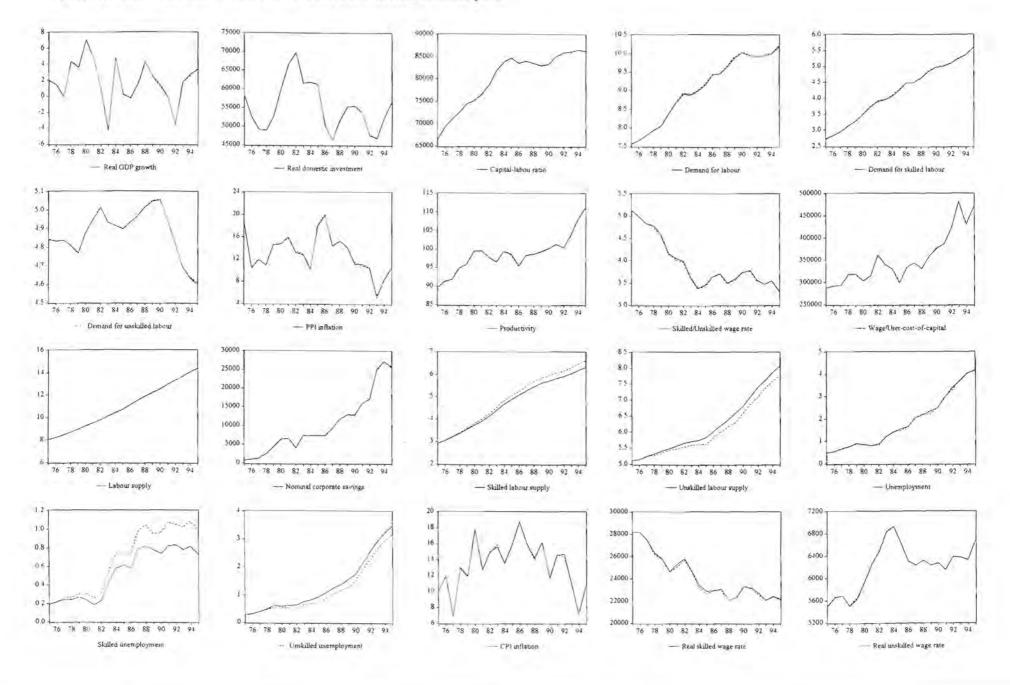
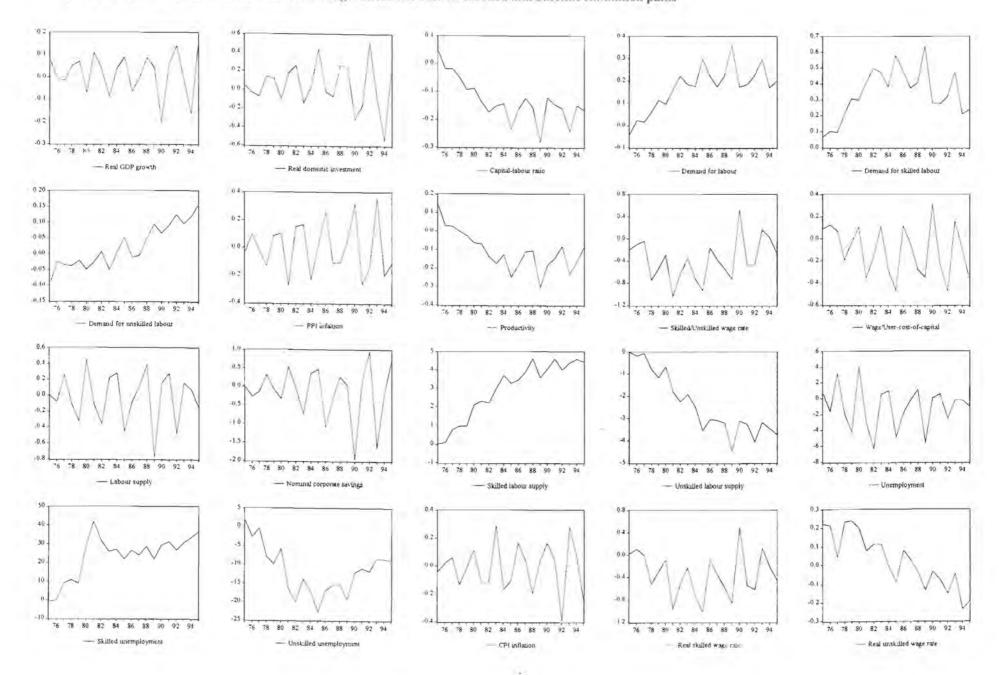


Figure A16.3(b) Individual scenario 2: Percentage differences betwee shocked and baseline simulation paths





FigureA16.4(a) Individual scenario 3: Baseline and shocked simulation paths

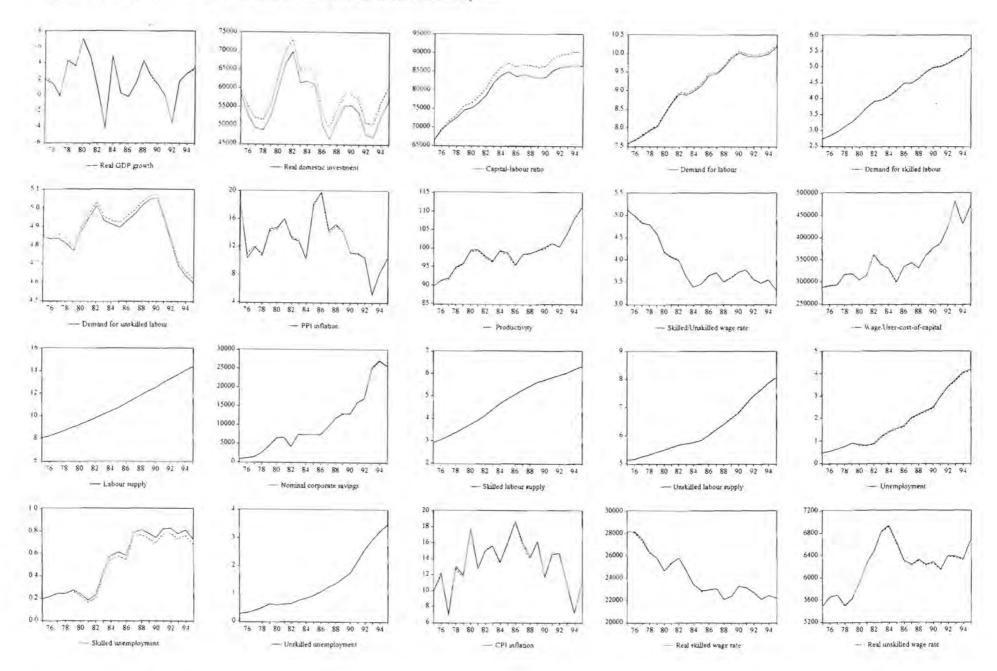




Figure A16.4(b) Individual scenario 3: Percentage differences between shocked and baseline simulation paths

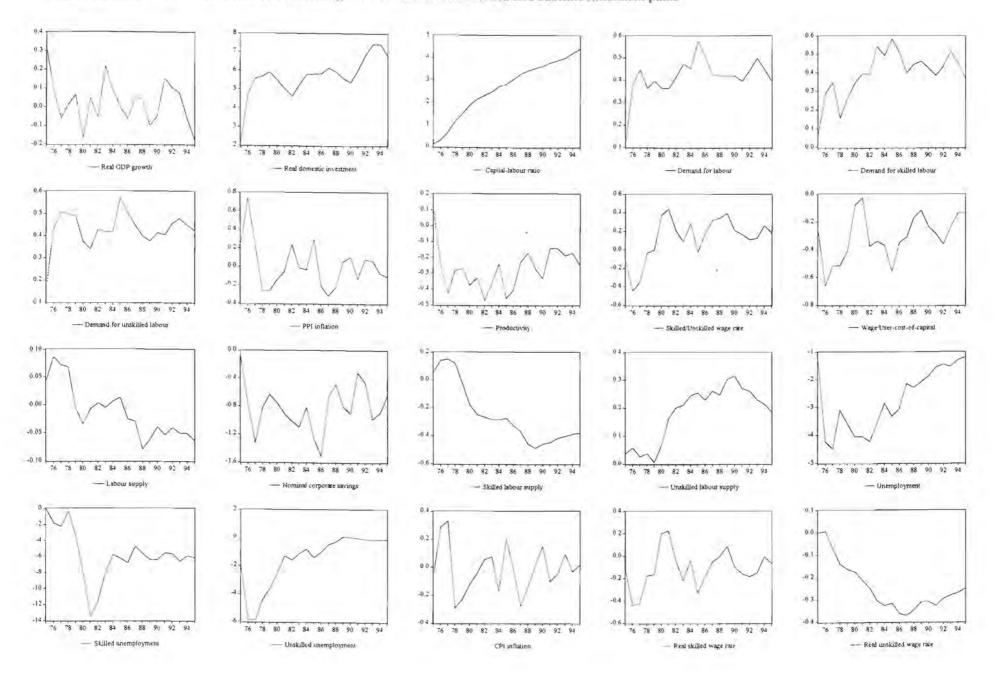




Figure A16.5(a) Individual scenario 4: Baseline and shocked simulation paths

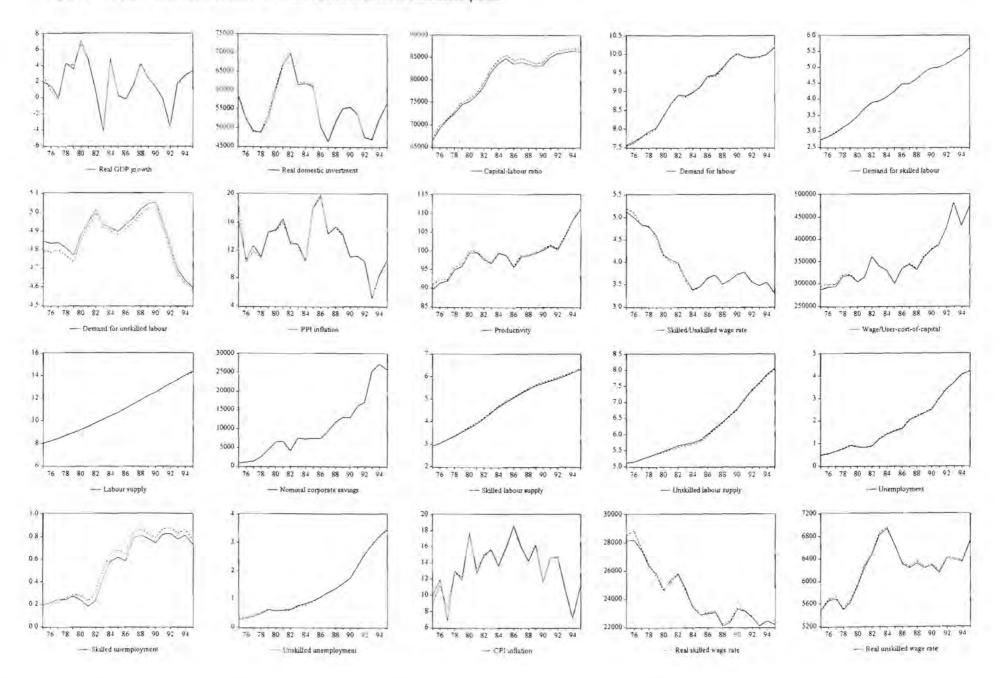




Figure A16,5(b) Individual scenario 4: Percentage differences between baseline and shocked simulation paths

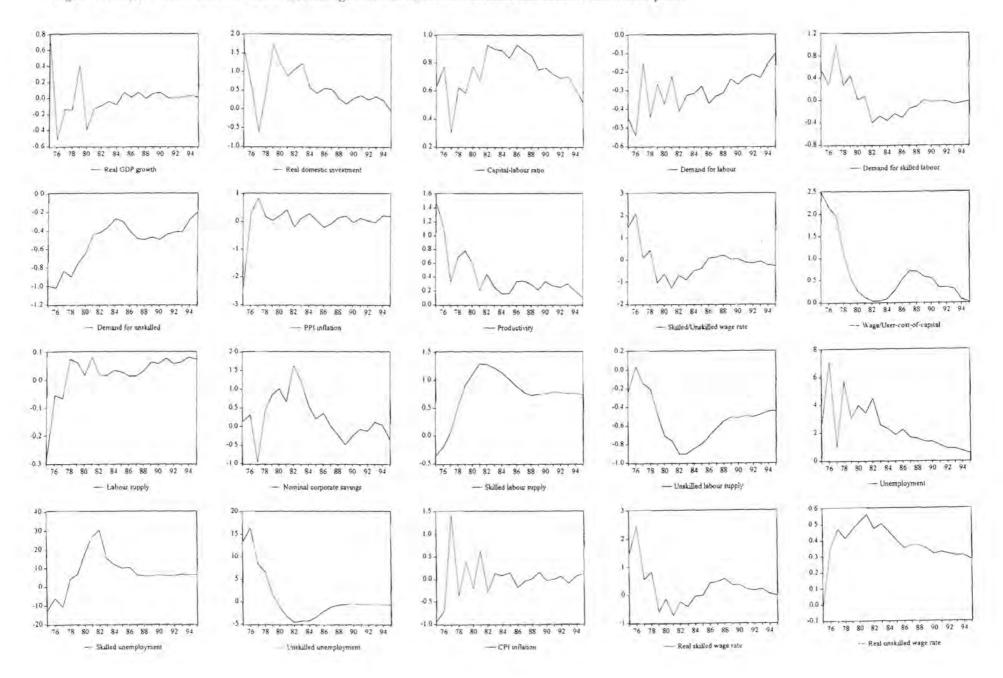




Figure A16.6(a) Individual scenario 5: Baseline and shocked simulation paths

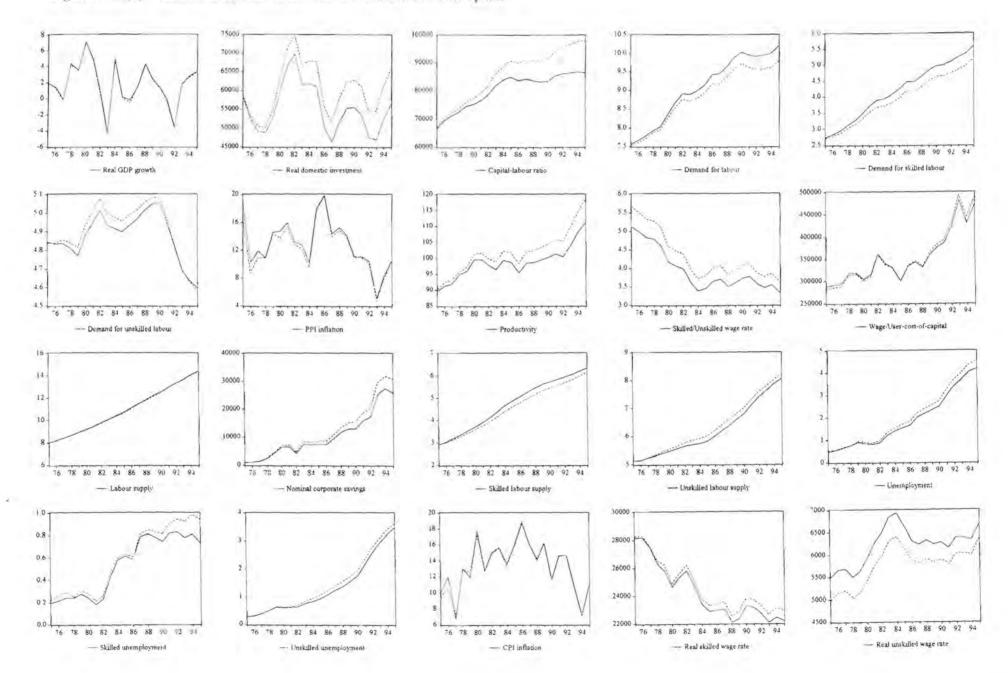




Figure A16.6(b) Individual scenario 5: Percentage differences between baseline and shocked simulation paths

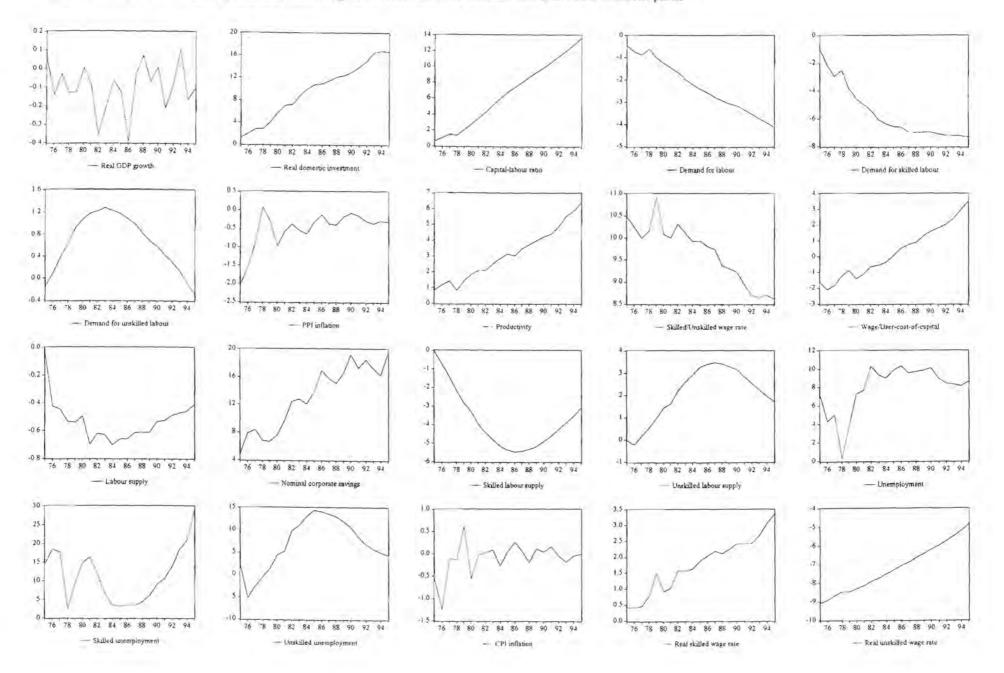




Figure A16.7(a) Individual scenario 6: Baseline and shocked simulation paths

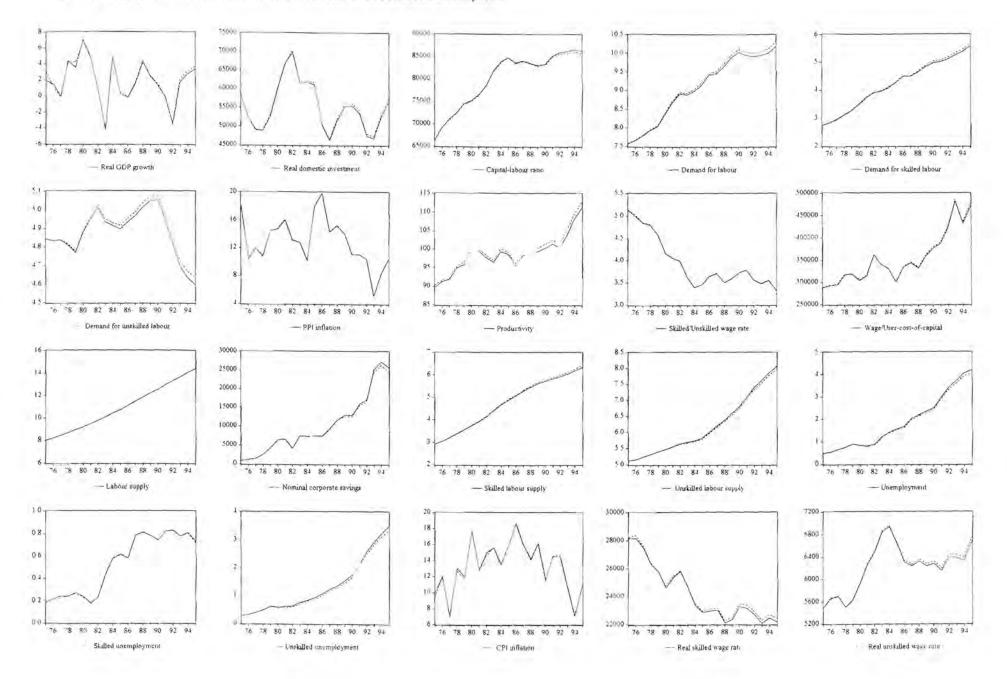




Figure A16.7(b) Individual scenario 6: Percentage differences between baseline and shocked simulation paths

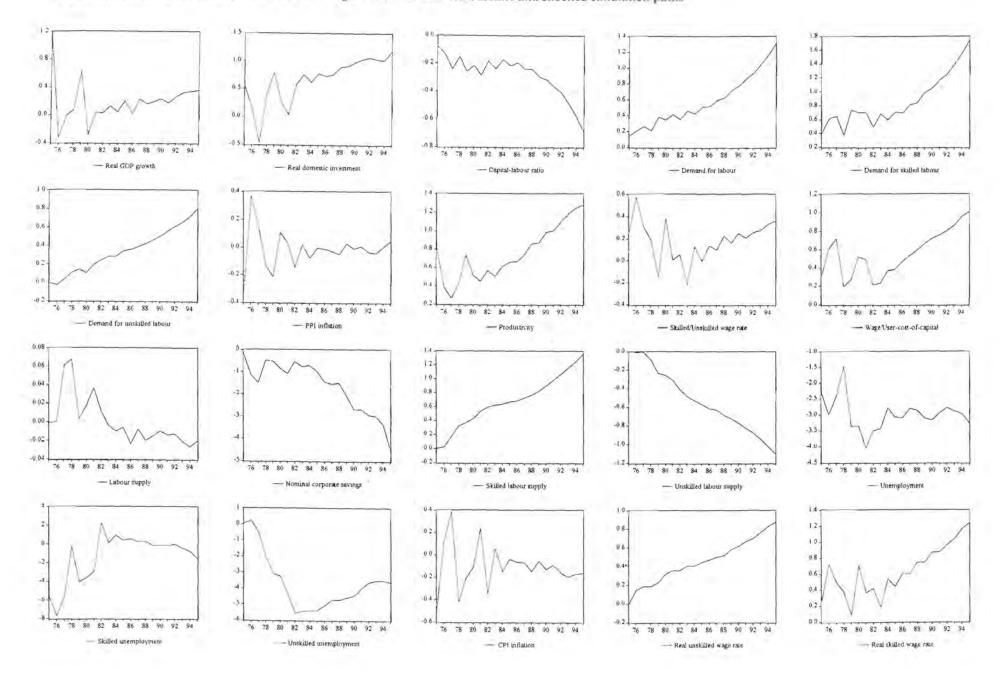




Figure A16.8(a) Individual scenario 7: Baseline and shocked simulation paths

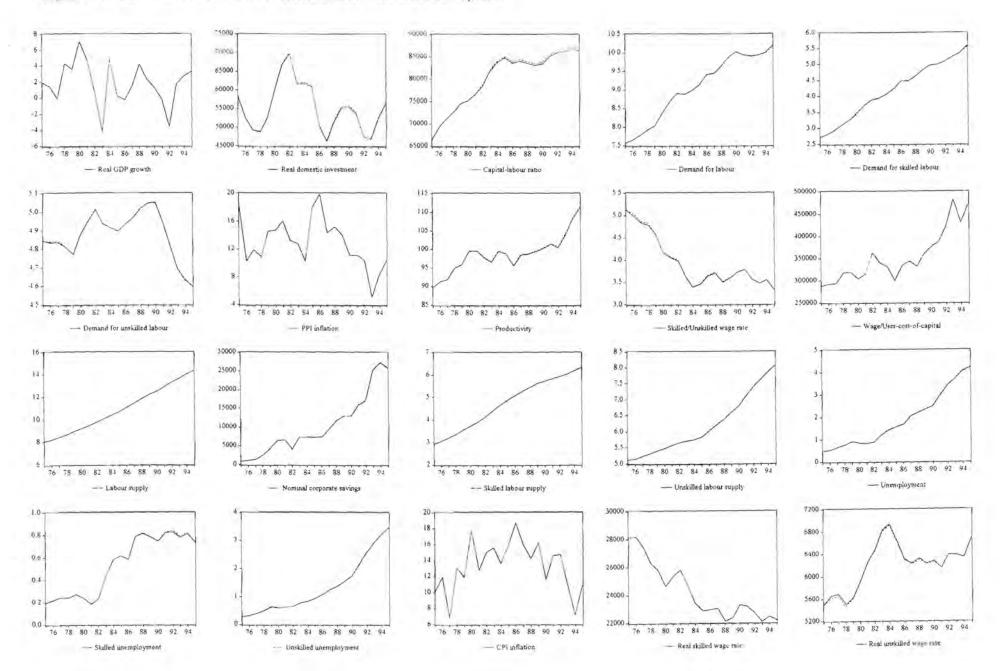




Figure A16.8(b) Individual scenario 7: Percentage differences between baseline and shocked simulation paths

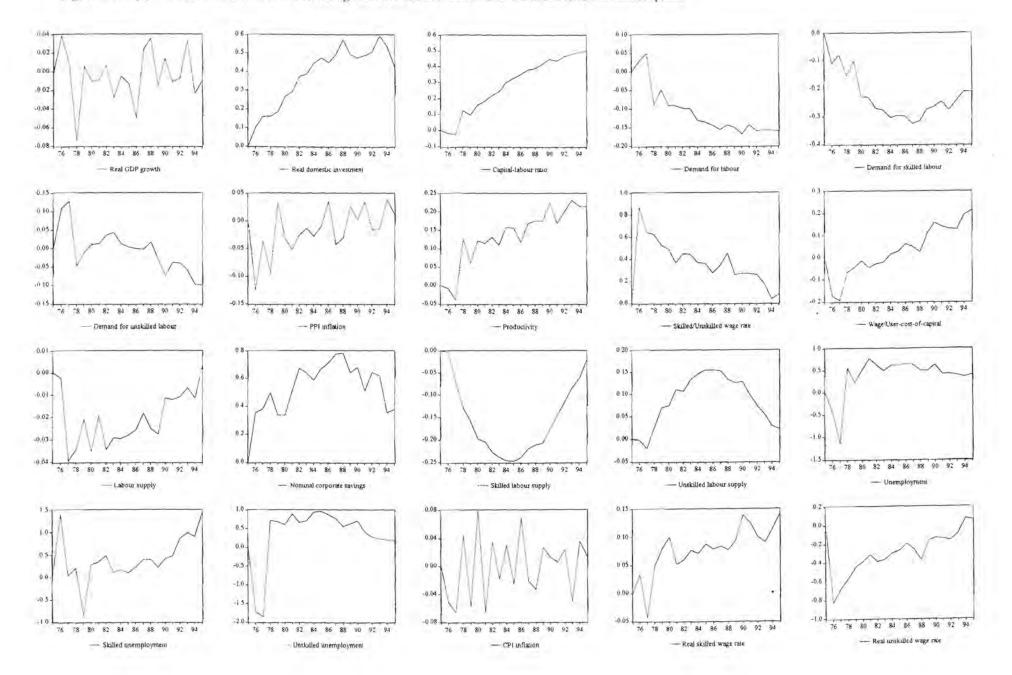




Figure A16.9(a) Individual scenario 8: Baseline and shocked simulation paths

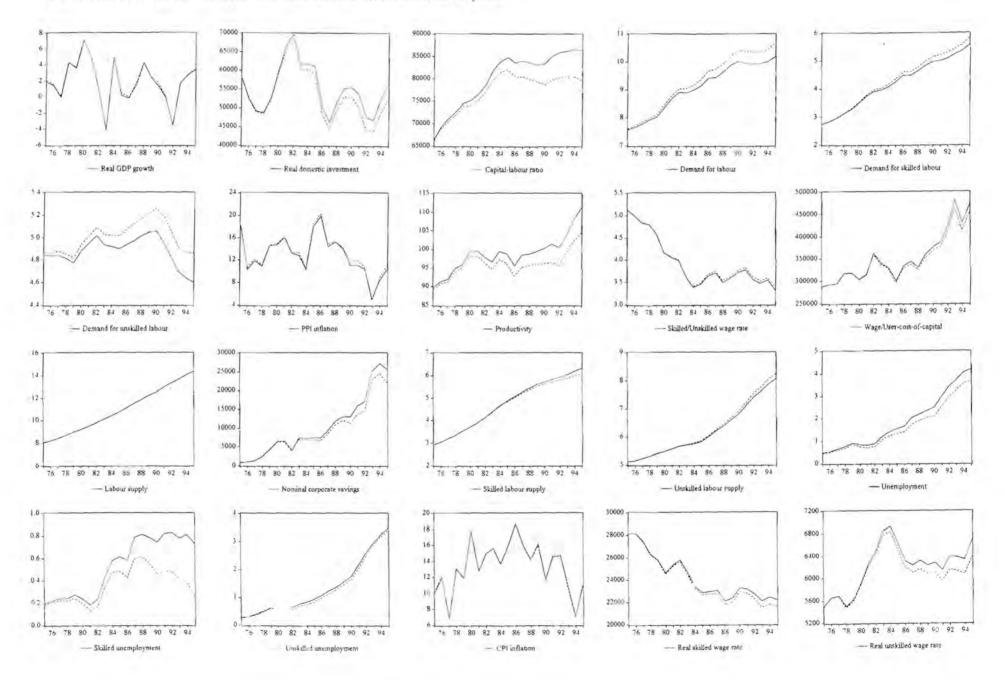




Figure A16.9(b) Individual scenario 8: Percentage differences between baseline and shocked simulation paths

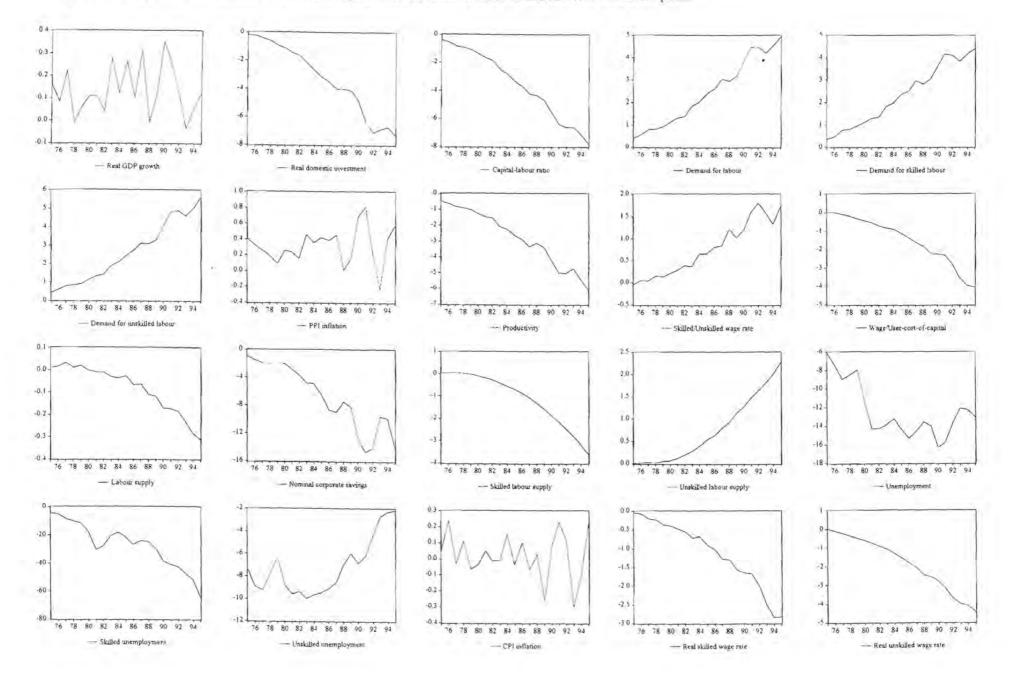




Figure A16.10(a) Individual scenario 9: Baseline and shocked simulation paths

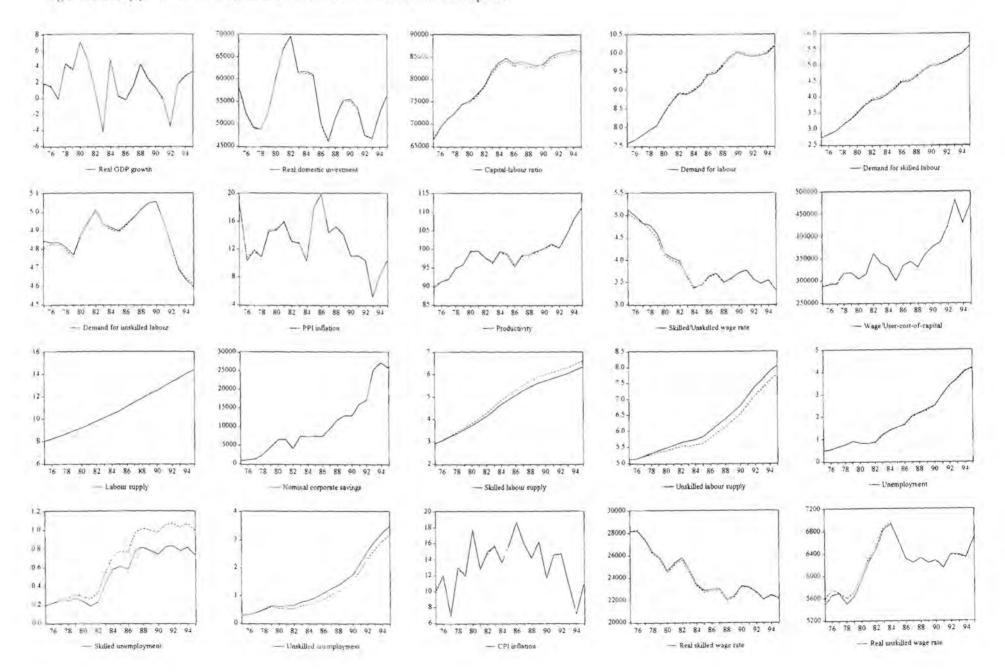




Figure A16.10(h) Individual scenario 9: Percentage differences between baseline and shocked simulation paths

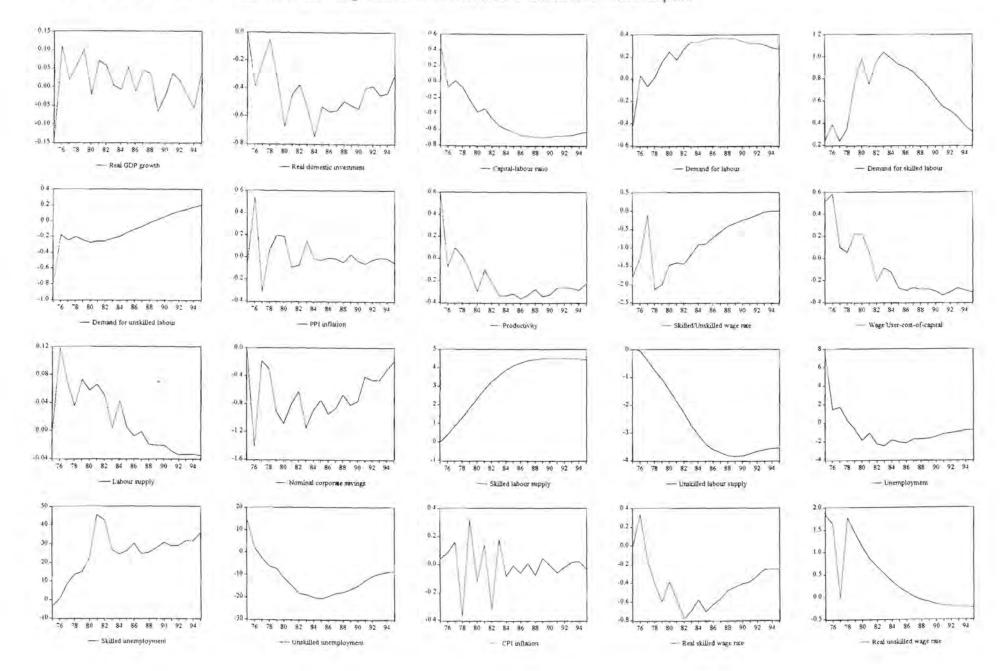




Figure A16.11(a) Combined scenario 1: Baseline and shocked simulation paths

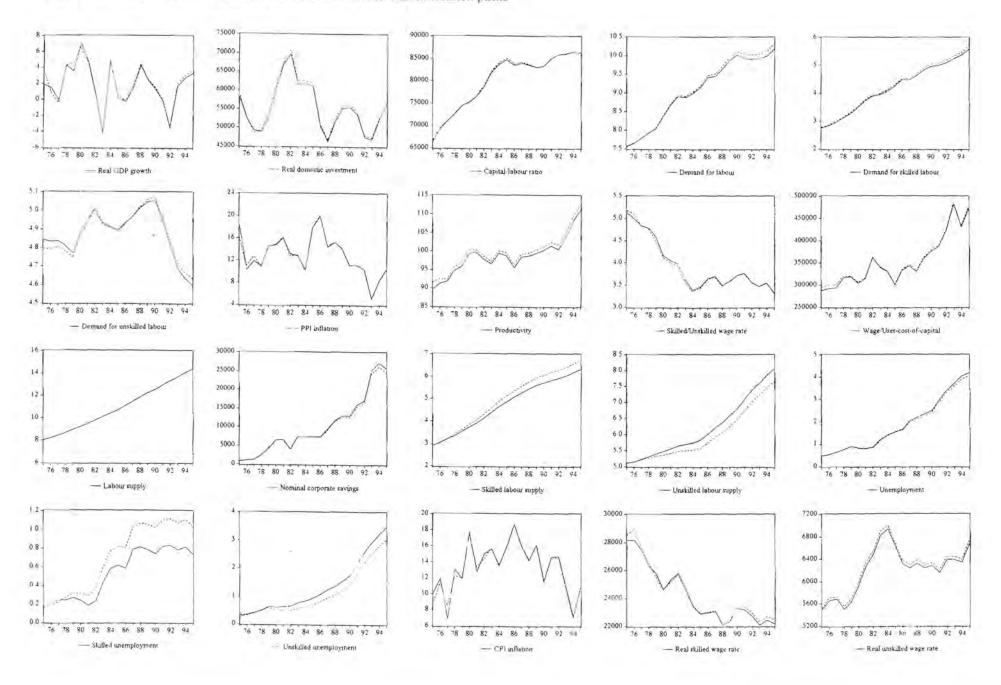




Figure A16.11(b) Combined scenario 1: Percentage differences between baseline and shocked simulation paths

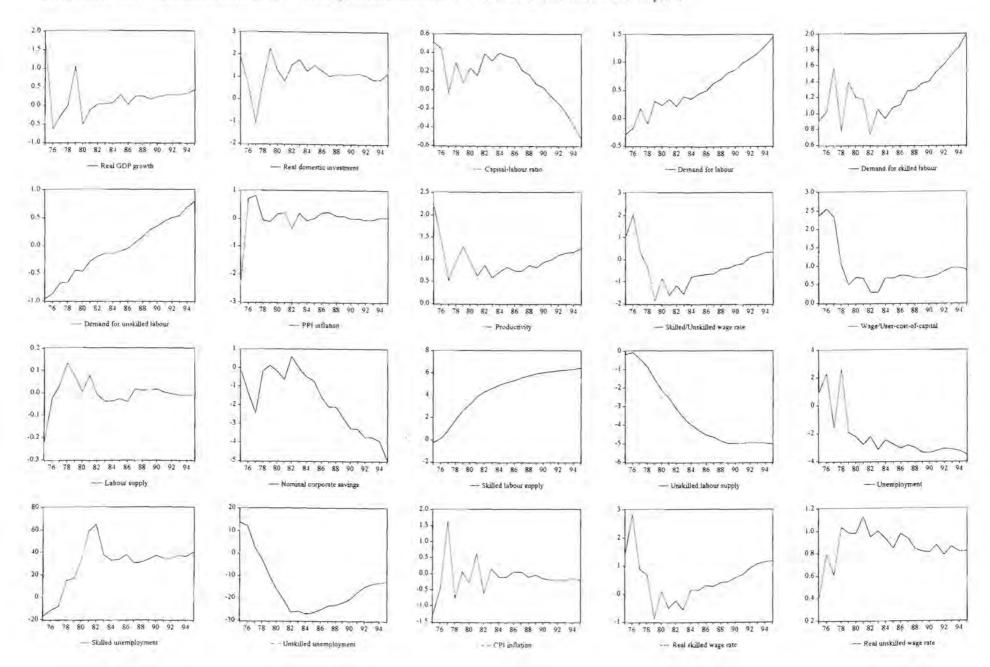




Figure A16.12(a) Combined scenario 2: Baseline and shocked simulation paths

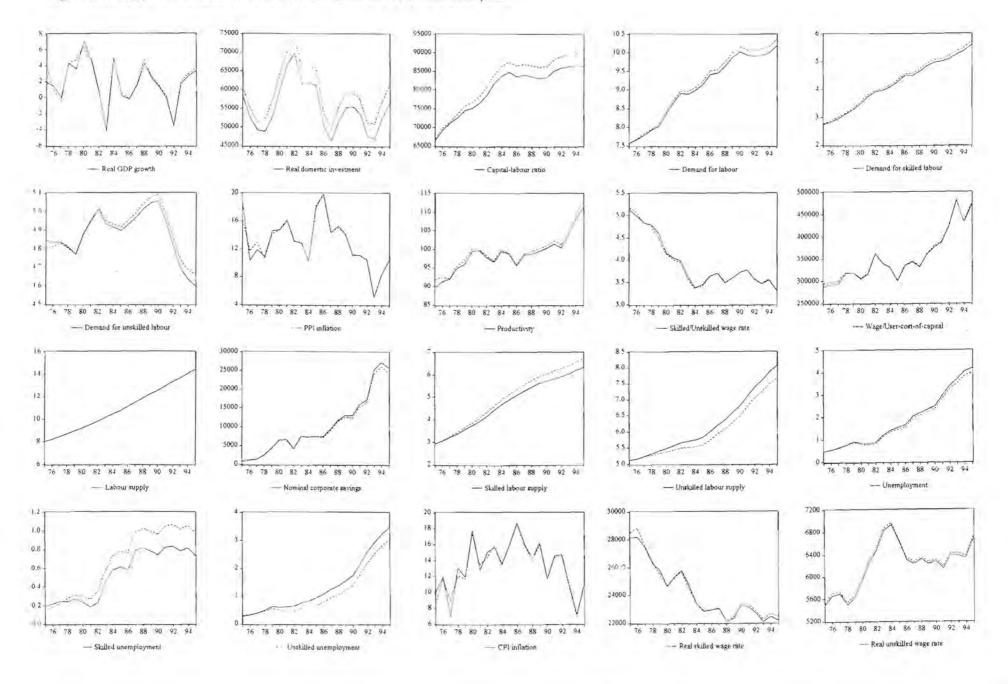




Figure A16.12(b) Combined scenario 2: Percentage differences between baseline and shocked simulation paths

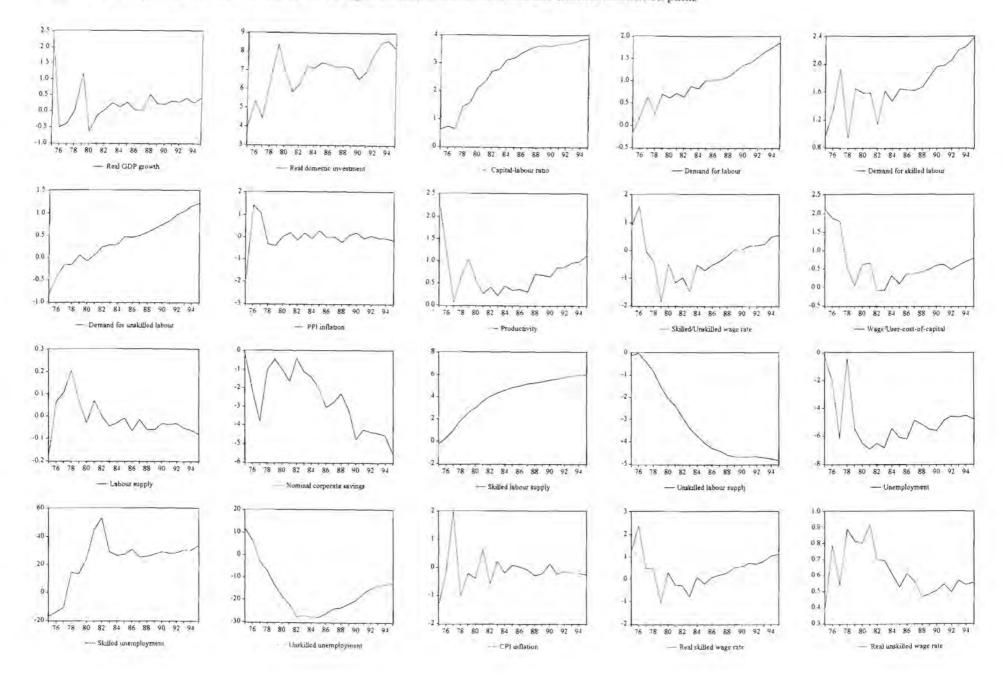




Figure A16.13(a) Combined scenario 3: Baseline and shocked simulation paths

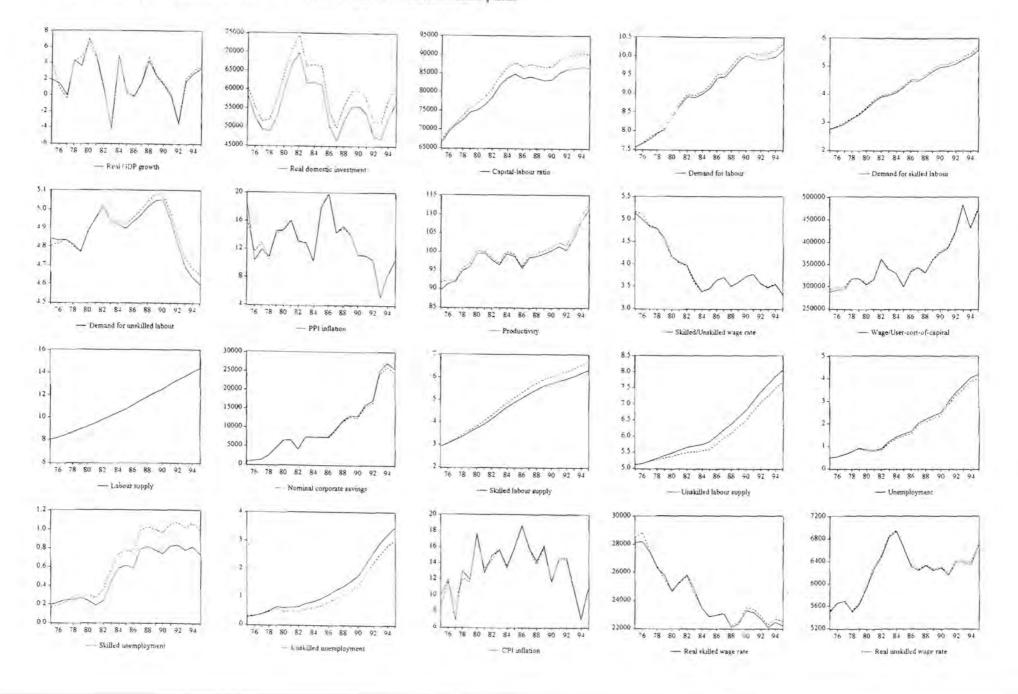




Figure A16.13(b) Combined scenario 3: Percentage differences between baseline and shocked simulation paths

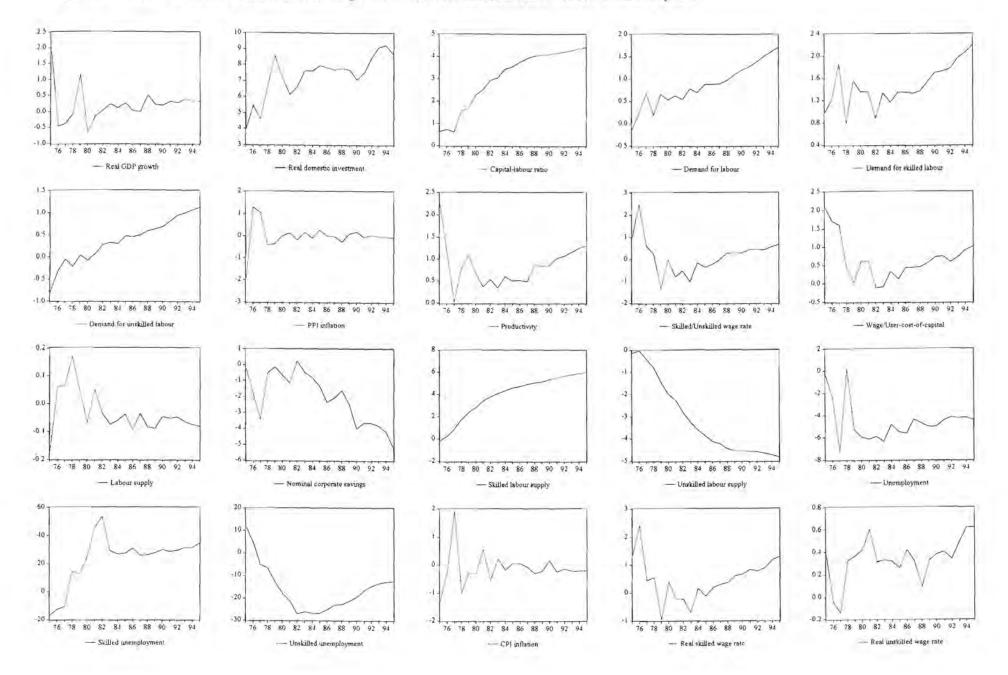




Figure A16.14(a) Combined scenario 4: Baseline and shocked simulation paths

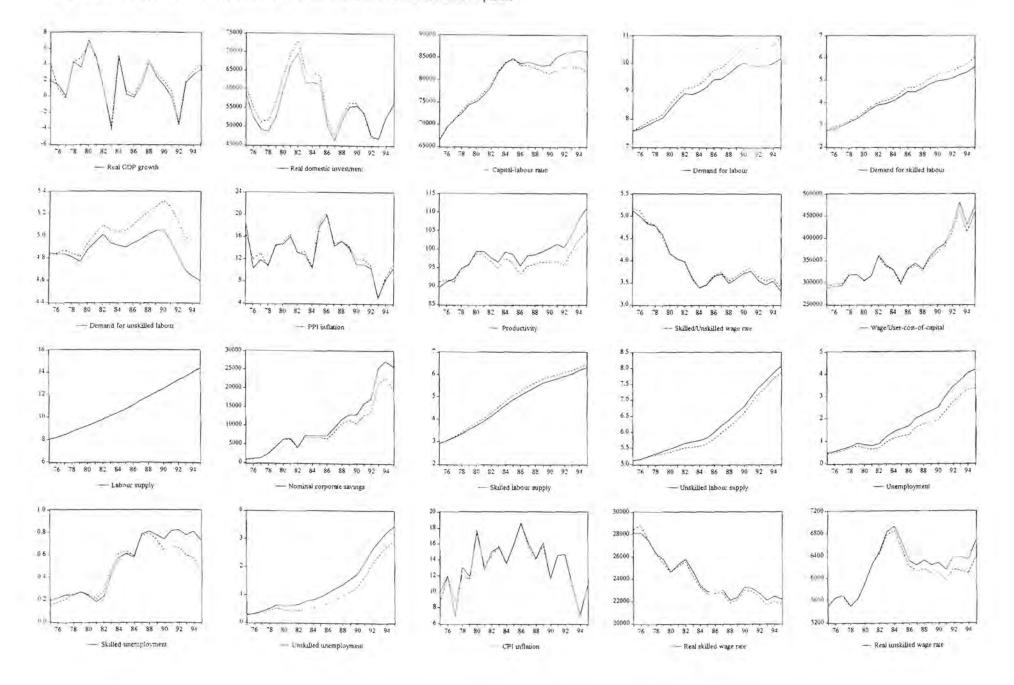




Figure A16.14(b) Combined scenario 4: Percentage differences between baseline and shocked simulation paths

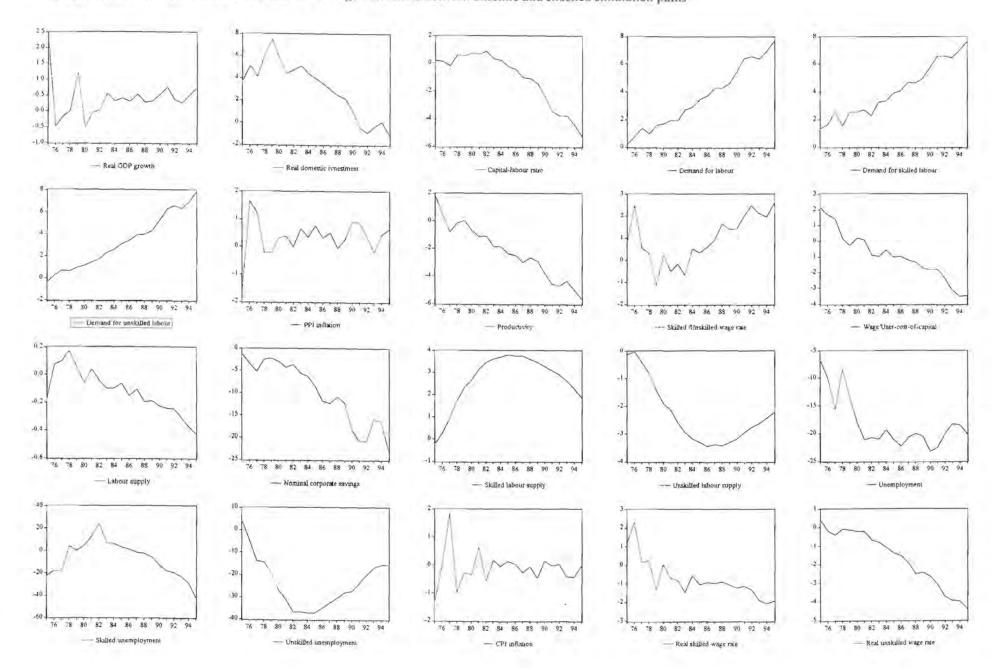




Figure A16.15(a) Combined scenario 5: Baseline and shocked simulation paths

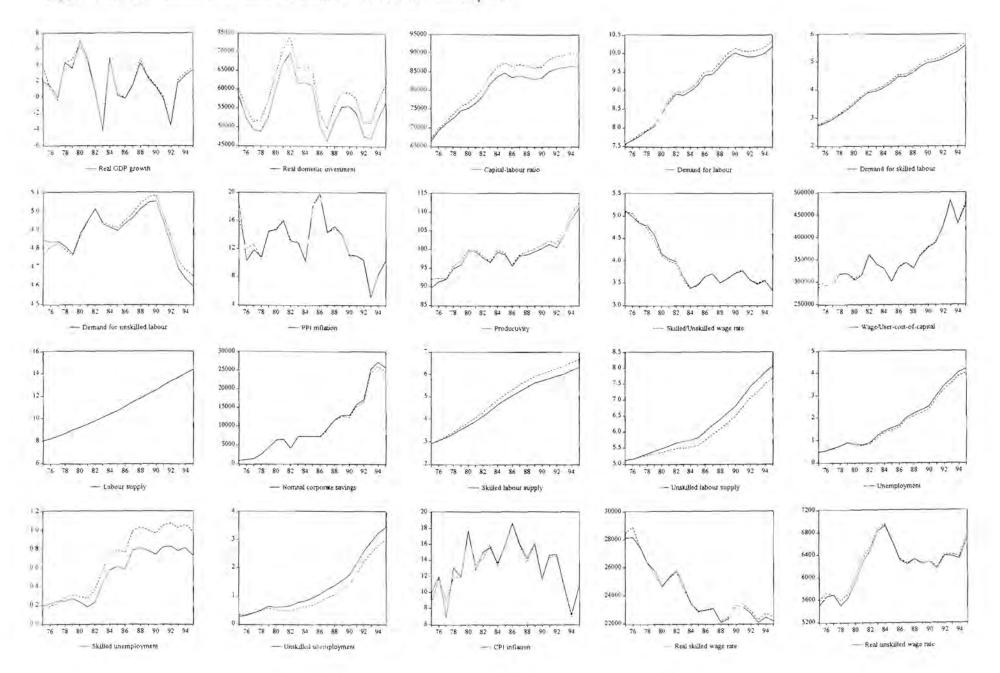




Figure A16.15(b) Combined scenario 5: Percentage differences between baseline and shocked simulation paths

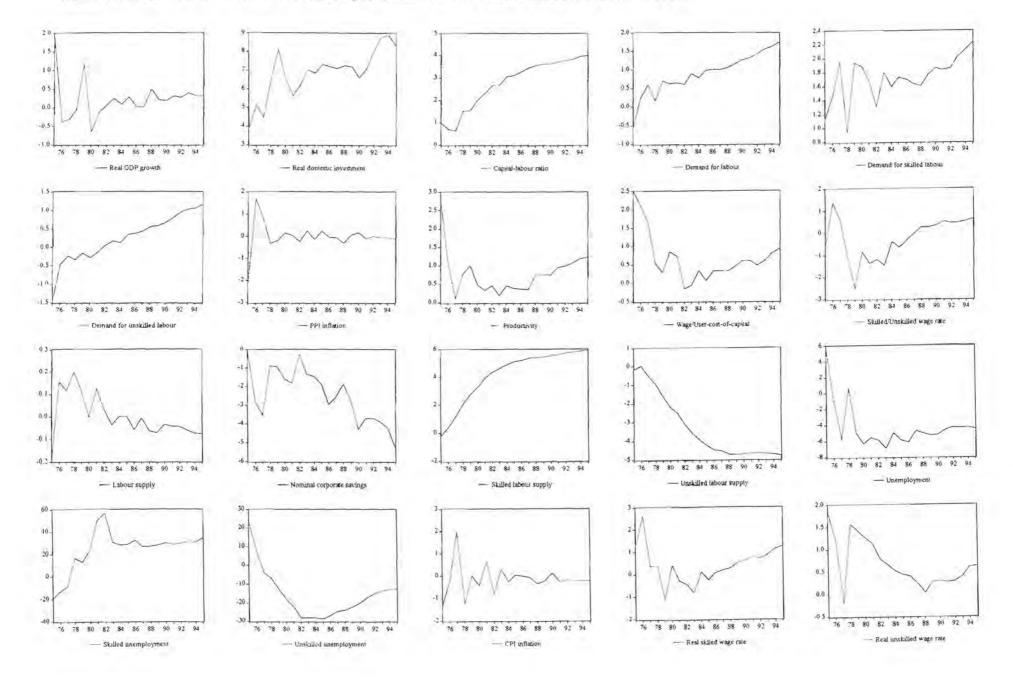




Figure A16.16(a) Combined scenario 6: Baseline and shocked simulation paths

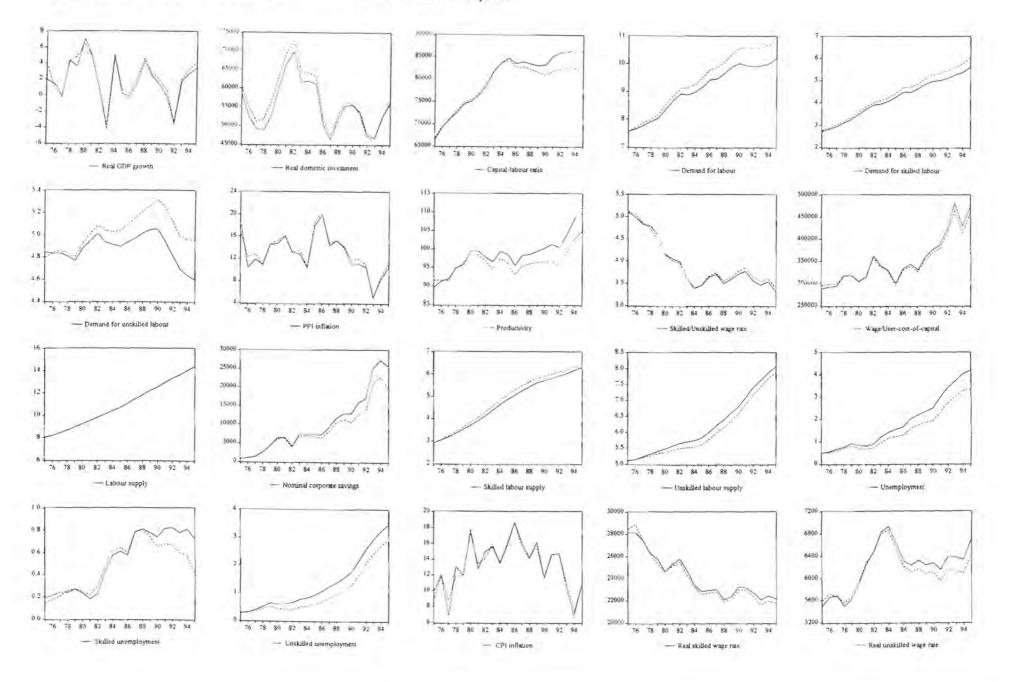
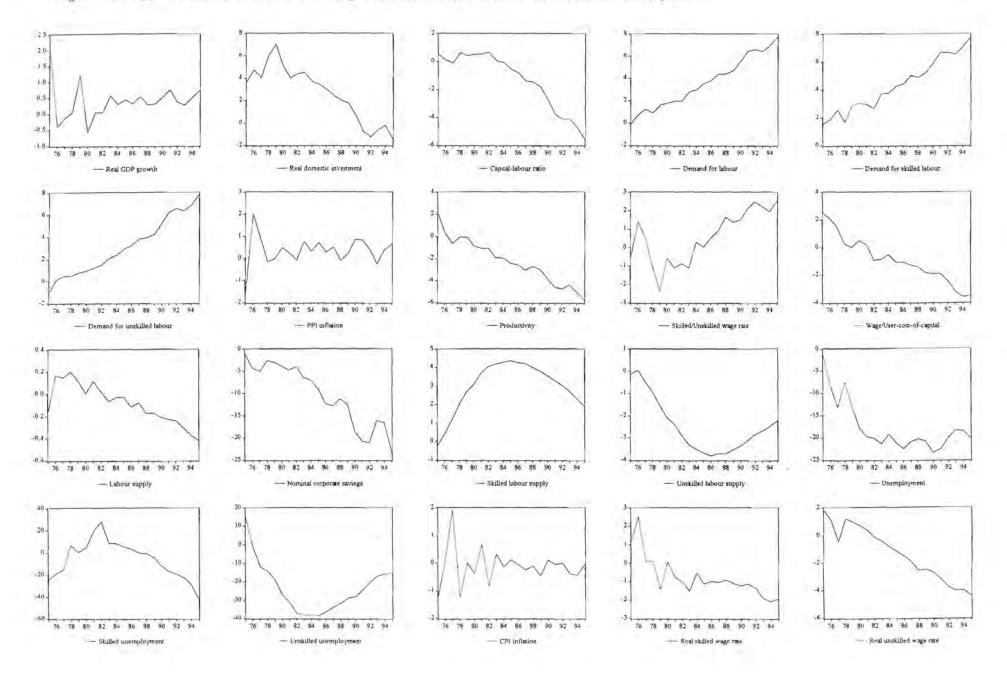




Figure A16.16(b) Combined scenario 6: Percentage differences between baseline and shocked simulation paths





## APPENDIX 17: ORDER OF INTEGRATION

The augmented Dickey-Fuller unit root test is employed to test whether the data series are stationary or not. The testing strategy suggested by Dolado et al. (1990) and applied by Strum and de Haan (1995) is followed.

The augmented Dickey and Fuller (1981) unit root test is specified as the ordinary least squares estimation of:

$$\Delta Y_{t} = \eta_{0} + \eta_{1} \text{Trend} + \eta_{2} Y_{t-1} + \sum_{i=1}^{m} \eta_{2+i} Y_{t-i} + \varepsilon_{t} . \tag{A17.1}$$

where  $Y_t$  is the series being tested, m is the number of lags in the testing equation and  $\epsilon_t$  the residual. The null hypothesis of non-stationarity is rejected in favour of a stationary data series, if the Dickey-Fuller test statistic is significantly less than the relevant critical value.

The test is implemented through the usual t-statistic of  $\hat{\eta}_2$ , denoted as  $\tau_{\tau}$ . Under the null hypothesis,  $\tau_{\tau}$  will not follow the standard t-distribution and the adjusted critical values computed by MacKinnon (1991) are used for evaluation. If  $\tau_{\tau}$  is significant, the null of non-stationarity is rejected and the data series is stationary.

If  $\tau_t$  is insignificant, the joint null hypothesis of  $\eta_1 = \eta_2 = 0$ , using the *F*-statistic denoted as  $\Phi_3$  is tested. The relevant critical values from Dickey and Fuller are used to evaluate the test statistic  $\Phi_3$ . If  $\Phi_3$  is significant, the unit root test is repeated, now using the critical values of the standard *t*-distribution.

If the trend is not significant in the maintained model, the next step is to estimate equation 1 without a trend ( $\eta_1 \equiv 0$ ). The unit root test is carried out, now denoting the *t*-statistic of  $\hat{\eta}_2$  as  $\tau_{\mu}$  and using the relevant critical values from MacKinnon. If the null hypothesis is rejected, the data series is stationary.

If the null of non-stationarity is not rejected, the joint null hypothesis of  $\eta_0 = \eta_2 = 0$ , using the F-statistic denoted as  $\Phi_1$  is tested and the critical values reported by Dickey and Fuller are employed. If  $\Phi_1$  is significant, the unit root test is repeated, using the standard normal distribution.

If  $\Phi_1$  is insignificant, the Dickey-Fuller  $\tau$  test is carried out without a constant in the testing equation, testing the joint hypothesis of  $\eta_0 = \eta_2 = 0$ . If the test statistic ( $\tau$ ) is significantly less than the relevant MacKinnon critical value, the null hypothesis of non-stationarity is rejected and the data series is stationary.

The number of lags used in the estimation equation is determined in similar way as Perron (1989). Perron suggested starting with eight lags. If the last (eighth) lag is insignificant at a 10 percent level of significance (using the standard normal distribution), it is omitted. The test is repeated with seven lags and the last lag tested for significance. The process is repeated until there are no more lags left, in which case the test has reduced to the standard Dickey-Fuller (DF) test. The large level of significance (10 percent) is taken because, as Perron (1989: 1384) pointed out, "...including too



many regressors of lagged first-differences does not affect the size of the test but only decreases its power. Including too few lags may have a substantial effect on the size of the test". Furthermore, Molinas (1986) noticed that "...a rather large number of lags has to be taken in the ADF test in order to capture the essential dynamics of the residuals."

The following reasons necessitate the careful evaluation of the ADF integration test results:

- (i) In the DFτ test the presumption is that the data generating process (DGP), stationary or not, has a zero mean (no constant or drift term), and no trend: in effect there are no deterministic components to the series. The implication is that the initial y value is zero, since a DGP without deterministic components and a unit root (i.e. either a true stationary or differenced stationary data series) has its mean determined by its initial observation. Where the mean of a non-stationary series is not zero, i.e. where the DGP has a drift component (constant), the DFτ test will be inappropriate it will lead to an over-rejection of the null of non-stationarity (stationarity will too easily be accepted) and therefore generate misleading results. Therefore, if the initial value of y is not certain (i.e. whether the DGP is subject to drift or not), the DFτμ test need to be employed.
- (ii) Under the  $DF\tau_{\mu}$  test:  $\Delta y_t = \mu_b + (\rho_b 1)y_{t-1} + \mu_t$ , (only testing for a unit root and drift term), if the series has a unit root  $(\rho_b = 1)$ ,  $y_t$  will follow a stochastic trend, drifting upwards or downwards depending on the sign of  $\mu_b$ .

Alternatively, if the series is stationary  $(\rho_b - 1)$ ,  $y_t$  will have no stochastic trend, and will be stationary around a constant mean of  $\mu_b/(1-\rho_b)$ . If, however, the true DGP is a trend-stationary process (stationary process with a deterministic trend component): i.e.  $y_t = \alpha + \beta t + \mu_t$ ,  $y_t$  will also follow a trend (now of a stationary deterministic and not a non-stationary stochastic nature), drifting upwards or downwards depending on the sign of  $\beta$ . The  $DF\tau_\mu$  test will mistaken the stationary deterministic trend for a non-stationary stochastic trend and render the  $y_t$  process non-stationary. The  $DF\tau_\mu$  test fails to distinguish between a series with a stochastic and a series with a deterministic trend. The  $\tau_\tau$  version of the Dickey-Fuller test ( $DF\tau_\tau$ ) allows for the presence of a time trend. However, there are two points to note here:

- If the DFτ<sub>τ</sub> test is inappropriately avoided, there would be a tendency to overaccept the null of non-stationarity for series, which are in fact stationary.
- If the DFτ<sub>τ</sub> test is inappropriately used, there would be a tendency to under-accept
  the null of non-stationarity for series, which are in fact non-stationary.

DF \tau test:  $\Delta y_t = (\rho_a - 1)y_{t-1} + \mu_t$  test whether  $\rho = 1$  ( $\rho - 1 = 0$ ) with  $\mu_t \sim \text{IID}(0, \sigma^2)$ .

If  $\rho = 1$  ( $y_t$  is a non-stationary data series):  $\Delta y_t = 0 + \mu_t$  (indicating that  $\rho - 1 = 0$  and, although not explicitly testing for it, indicating that  $\Delta y_t$  has a zero mean). Now, if  $y_t$  is subject to drift (constant) and the  $DF\tau$  test is applied, still only testing for  $\rho - 1 = 0$ ,  $\Delta y_t = \delta + \mu_t$  will indicate that  $\rho - 1 = \delta \neq 0$ . If, by example,  $\delta = -1.2 < 0$ , then the  $DF\tau$  test will wrongly indicate that  $y_t$  is stationary due to the outcome of  $\rho - 1 = \delta = -1.2$ .



- (iii) Unit root tests have problems with their "power": in small samples they are likely to accept the null-hypothesis of a unit root when the true DGP is in fact stationary, though close to having a unit root. In effect, standard unit root test will under-reject the presence of a unit root (accept stationarity). With small samples, therefore, special care must be exercised in the application of unit root tests. Effectively, in small (finite) samples some unit root processes appear to behave like trend-stationary processes, while some trend-stationary processes appear to behave like random walk (non-stationary) processes. Unit root tests are likely to be fooled in either event: either rejecting non-stationarity where non-stationarity should be accepted, or accepting non-stationarity where non-stationarity should be rejected. This is obviously a very important problem in the current scenario since the sample size used to test for the presence of a unit root only consists of 25 observations.
- (iv) Where (unknown) structural breaks are present in the data, application of "standard" unit root tests would again lead to an under-rejection of the null of the presence of a unit root (i.e. accept stationarity). In the South African context the periods of sanctions, disinvestment, outflow of human capital and changes in monetary policy are of importance and need to be taken in consideration.

An important consequence of the above is that inferences concerning the stationarity properties of data should always be based on a range of evidence, of which the formal ADF (and other) tests are merely one component. Economic theory, structural properties of the data, a priori information concerning political and other exogenous shocks to the economy over the sample period, data plots, residual plots, plots of the autocorrelation function, and spectral analysis all form part of the evidence required to draw conclusions on the stationarity of data. These inference guidelines should always be taken in consideration when testing for the presence of a unit root.

Tables A17.1, A17.2 and A17.3 report the outcomes of the ADF tests for all relevant data series employed in the estimations. The series tested are given in the first column. The second column reports whether a trend and a constant (Trend), only a constant (Constant), or neither (None) is included. In the third column, the number of lags recorded is reported. The next column shows the ADF t-statistic, called  $\tau_{\tau}$  when a trend and constant are included,  $\tau_{\mu}$  when only a constant is included, and  $\tau$  when neither occurs. The last column reports the F-statistic,  $\Phi_3$  ( $\Phi_1$ ), testing whether the trend (constant) is significant under the null hypothesis of no unit root.

According to table A17.1 there are a couple of variables that seem to be stationary in levels. However, apart from the fact that the test results are not conclusive, taking the problems and inference guidelines associated with the testing for a unit root in consideration, it is obvious that these variables cannot be stationary in their levels. The only variable that can indeed be rendered stationary in levels is  $ln\_cu$ . Table A17.2 indicates that all other variables, except  $ln\_ss$  are integrated of order 1. Although  $ln\_ss$  tested as stationary in second differences, consideration of the inference guidelines makes it is plausible that the variable is integrated of order 1 rather than order 2. The ADF test may recognise the structural break as a trend within a trend, rendering  $ln\_ss$  as an l(2) variable. Due to the nature of construction of the dummy variables, they are all integrated of order 0 and are therefore all stationary in levels.



Table A17.1. Augmented Dickey-Fuller tests for non-stationarity, levels, 1970-1995

Series	Model	Lags	$\tau_{\rm c}, \tau_{\rm \mu}$ , $\tau$	$\Phi_3, \Phi_1$
ln_bbp_90p	Trend	0 (insign)	-1.68	3.21
	Constant	0	-2,30	5.29*
	None	0	4.16	
In bbp min 90p	Trend	3	-3.73*	3.46
	Constant	3	-3.72*	3.46
	None	0 (insign)	-1.28	
In bbpfact 90p	Trend	0 (insign)	-1.71	4.04
	Constant	0	-2.70	7.31*
	None	0	4.07	
ln c90p lrem5	Trend	3	-4.77**	5.46
	Constant	4	0.15	1.37
	None	4	2.66	
ln_cu	Trend	8	-3.87*	
	Constant			
	None			
In educind	Trend	0	-2.81	4.70
/D-2328117	Constant	0	0.45	0.20
	None	0	-2.60*	
In empl rat	Trend	4	-0.24	3.50
Mand, 47.	Constant	4	2.04	2.17
	None	4	3,34	
In empl rat s	Trend	6	-1,24	1.76
	Constant	6	-1,79	2.14
	None	4	0,13	
In_empl_rat_u	Trend	0 (insign)	0.57	23.26
	Constant	0	5.83	33.96
	None	0	9,59	
In_exces_dem_cp	Trend	0	-2,35	3.37
	Constant	0	-2.65	7.04*
	None	0	-1.77	
In_fincond_ppi	Trend	t	-3.05	3.26
	Constant	1	-3.15*	5,13
	None	4	0.13	
In_gostc_cp	Trend	1	-5.09**	8.77*
	Constant	2	-0.38	2.09
	None	2	4.39	
In_gprys_\$	Trend	4	-1.19	2.52
	Constant	4	-2.58	2.72
	None	2	0.89	

<sup>\*(\*\*)</sup> Significant at a 5(1) percent level.

(insign) Insignificant lag.

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_{\tau})$ , and -3.00(-3.75) when only a constant is included  $(\tau_{\mu})$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.1 (cont.). Augmented Dickey-Fuller tests for non-stationarity, levels, 1970-1995

Series	Model	Lags	$\tau_{r}, \tau_{\mu}, \tau$	$\Phi_3, \Phi_1$
ln_gprys_r	Trend	4	0.27	4.31
	Constant	4	-3.76**	4.10
	None	I	1.77	
ln if	Trend	2	-2.32	7.90*
	Constant	2	-2.27	9.42**
	None	2	0.50	
In interposind	Trend	6	-3.24	3.44
	Constant	0 (insign)	-0.74	0.54
	None	0	-3.91*	
ln_kap_lab_rat	Trend	4	-2.36	8.56*
	Constant	0	-6.32**	39.97**
	None	0.	4.51	
In kap r	Trend	2	-1.76	146.30**
	Constant	2	-3.40	187.24**
	None	2	0.69	130,44
ln n	Trend	4	0.80	3.95
2-30	Constant	0	-3.74*	14.02**
	None	0	6.10	
In n informal	Trend	1	-2.80	3.28
13.6	Constant	0 (insign)	-1.11	1.24
	None	0	3.01	
liv ns	Trend	0 (insign)	-1.23	17.33
·	Constant	0	-6.01**	36.22**
	None	0	6.68	
In nu	Trend	4	1.03	3.89
-	Constant	1	-2.08	8.88**
	None	1	-0.21	
ln_p_manuf	Trend	1	-4.01*	18.01**
	Constant	i.	-0.61	11.64
	None	1	-1.84	
In_ppi	Trend	1	-1.91	6,09
7	Constant	2	-2.30	6.41*
	None	1	-2.07*	
In_prime_rate	Trend	3	-3.46	4,23
	Constant	2	-1.48	2.90
	None	0 (insign)	0.65	
ln_product	Trend	1	-1.82	2,82
200	Constant	0 (insign)	1.07	1.13
	None	0	2.24	

(insign) Insignificant lag.

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included ( $\tau_c$ ), and -3.00(-3.75) when only a constant is included ( $\tau_\mu$ ), and -1.95(-2.66) when neither is included ( $\tau$ ). The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.1 (cont.). Augmented Dickey-Fuller tests for non-stationarity, levels, 1970-1995

Series	Model	Lags	$\tau_{\tau}, \tau_{\mu}, \tau$	$\Phi_3, \Phi_1$
In pz\$-	Trend	2	-2.07	6.36
	Constant	2	-3.53*	8.93**
	None	2	-2.97**	
ln r\$	Trend	3	-2,23	2,77
	Constant	0 (insign)	0.24	0.06
	None	0	2.74	
In rel wscost u	Trend	6	1.82	2.05
	Constant	4	-1.54	2.03
	None	4	-3.89**	L Y
In rel wsu rat	Trend	0 (insign)	-1.53	2.78
	Constant	6	-1.36	1.09
	None	6	-1,65	3775
In s	Trend	3	-1.27	2.12
72-5	Constant	3	-1.43	2,19
	None	3	-0.79	
ln_ss	Trend	1	0.33	31.27
	Constant	1	-2.54	46.06**
	None	1	-0.03	
ln_sc_cp	Trend	0	-1.99	2.06
2-1-1	Constant	0 (insign)	-0.90	0.81
	None	0	2.47	
In socind	Trend	.0	-1.50	1.82
-	Constant	6	3.43	3.06
	None	5	-0.33	
tecno index	Trend	0	-2.01	6.14
	Constant	0	1.83	3.36
	None	0	4,02	
In total pop	Trend	4	0.36	12671.24
A MODELLE OF	Constant	4	-2.94	15012.84**
	None	8	-0.85	
ln_ucc2_90p	Trend	1	-5.06**	8.54*
	Constant	4	-0.78	1.83
	None	7	-1.56	
ln_ucc2_nom	Trend	6	-1.75	1.01
	Constant	0 (insign)	-1.07	1.14
	None	5	-1.53	12.00
In unempl rat	Trend	4	-2.58	3.33
	Constant	4	0.76	1.70
	None	6	-4.02**	

(insign) Insignificant lag.

a At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included ( $\tau_c$ ), and -3.00(-3.75) when only a constant is included ( $\tau_\mu$ ), and -1.95(-2.66) when neither is included ( $\tau$ ). The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.1 (cont.). Augmented Dickey-Fuller tests for non-stationarity, levels, 1970-1995

Series	Model	Lags	$\tau_{\rm r}, \tau_{\mu}, \tau$	$\Phi_3, \Phi_1$
In uniopresind	Trend	7	-1.99	2.36
-	Constant	5	2.74	2.32
	None	5	-1.70	
ln_vpi	Trend	4	-0.54	4.02
	Constant	4	-2.11	5.06
	None	4	-3.06**	
In w prod	Trend	2	-0.52	10.84
	Constant	2	-1.74	15.00**
	None	2	2.38	
In ws rate	Trend	L	-2.73	6.88
2/2	Constant	1	-0.35	5.12
	None	1	1.99	
In wtot ppi rat	Trend	3	-3.24	3.74
	Constant	1	-0.84	1.88
	None	1	0.73	
ln_wtot_rate	Trend	1	-3.59	9.79*
	Constant	2	-0.41	5.20
	None	2	2.41	
ln wtot vpi rat	Trend	1	-3.23	5.56
	Constant	5	-2.20	2.15
	None	0	1.82	
In wuppi rat	Trend	3	-3.34	2.96
	Constant	6	-0.18	1.04
	None	6	2.61	
ln wuvpi rat	Trend	0	-1,92	3.87
Tien wit letter	Constant	0	-2.82	7.96**
	None	0	2.39	
ln_xgoud_ppi	Trend	4	0.15	2.47
	Constant	5	-1,28	1.86
	None	5	-0.35	- T
In_xgoud_px	Trend	0 (insign)	-1.29	1.72
	Constant	1	-2.07	2,88
	None	0 (insign)	-0.18	

(insign) Insignificant

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_c)$ , and -3.00(-3.75) when only a constant is included  $(\tau_{\mu})$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.2. Augmented Dickey-Fuller tests for non-stationarity, first differences, 1970-

Series	Model	Lags	$\tau_t, \tau_\mu, \tau$	$\Phi_3, \Phi_1$
Δln bbp 90p	Trend	0	-3.81*	7.38*
	Constant	0	-3.65*	13.29*
	None	0	-2.63*	
Δln bbp min 90p	Trend	0	-3.78*	7.32*
	Constant	0	-3.90**	15.25**
	None	0	-3.84**	
ΔIn bbpfact 90p	Trend	0	-3.85*	7.55*
and Table and Table	Constant	0	-3.58*	12.85**
	None	0	-2.67**	
Δln c90p lrem5	Trend	3	-3.26***	3.68
Tee skTusmin	Constant	3	-3.32*	4.78***
	None	0	-2.82**	
ΔIn educind	Trend	0	4.16	9.87
and Tanna in the	Constant	0	-4.45**	19.83**
	None	0	-3.28**	
Δln empl rat	Trend	3	-4.75**	7.27*
	Constant	0	-2.98*	8.87**
	None	0	-2.21*	
Δln empl rat s	Trend	5	-0.54	2.58
	Constant	3	-3.44*	4.76***
	None	3	-3.53**	
∆ln empl rat u	Trend	0	-3.29***	
	Constant			
	None			
Δln exces dem cp	Trend	8	-2.77	4.47
	Constant	3	-4.26**	7.05*
	None	3	-4.35**	
∆ln fincond ppi	Trend	3	-3.49	5.00
	Constant	3	-3.74*	6.48*
	None	3	-3.85**	
Δln_gostc_cp	Trend	1	-5.06**	9.24*
	Constant	1	-5.20**	14.49**
	None	0	-1.62***	3200
Δln gprys \$	Trend	3	-4.53**	6.58
OF-1-7-4	Constant	0	-3.21*	10.32**
	None	0	-3.02**	1000,000

<sup>\*(\*\*)</sup> Significant at a 5(1) percent level.

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_r)$ , and -3.00(-3.75) when only a constant is included  $(\tau_\mu)$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level. (insign) Insignificant



Table A17.2 (cont.). Augmented Dickey-Fuller tests for non-stationarity, first differences, 1970-1995

Series	Model	Lags	$\tau_{\tau}, \tau_{\mu}, \tau$	$\Phi_3, \Phi_1$
Δln_gprys_r	Trend	3	-6.81**	12.34**
201.5	Constant	0	-3.39*	11.52**
	None	0	-2.45*	
Δln if	Trend		-4.49**	7.33*
	Constant	1	-4.30**	9.90**
	None	1	-4.35**	
Δln interposind	Trend	0	-3.45***	5.95***
Am_interposina	Constant	0	-3.50*	12.22**
	None	0	-2.26*	12.22
Δln kap lab rat	Trend	7	-1.98	2.54
Am_kap_iao_iai	Constant	5	-1.33	2.09
	None	5	-2.54*	2.03
Alla Iron e	Trend	1 1	-3.48***	7.13***
∆ln_kap_r	Constant	2		6.43
	100 200 200 200 200		-1.12	0.43
	None	2	-2.34*	7.00
Δln_n	Trend	8	-1.51	7.00
	Constant	8	2.22	3.92
	None	5	-1.76***	
∆ln_n_informal	Trend	- II.	-4.36*	6.77***
	Constant	- 1	-4.57**	10.63**
	None	1	-3.33**	
Δln_ns	Trend	0	-2.95	4.55
	Constant	0	-2.07	4.29***
	None			
Δln_nu	Trend	0	-3.06	4.70
	Constant	0	-2.47	6.09*
	None	0	-2.55*	
Δln_p_manuf	Trend	0	-2.27	3.17
32	Constant	0	-2.53	6.38*
	None			
Δln_ppi	Trend	1	-4.57**	7.50*
-F P	Constant	1	-3.75**	7.12*
	None			
Δln prime rate	Trend	4	-5.23**	8.24*
b.i.i.e_i.i.e	Constant	4	-5.13**	9.67**
	None	3	-3.95**	6181
Δln product	Trend	0	-3.24***	5.42
Zin_product	Constant	0	-2.98	8.89**
	None	0	-2.56*	0.02

<sup>\*(\*\*)</sup> Significant at a 5(1) percent level.

(insign) Insignificant

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_r)$ , and -3.00(-3.75) when only a constant is included  $(\tau_\mu)$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.2 (cont.). Augmented Dickey-Fuller tests for non-stationarity, first differences, 1970-1995

Series	Model	Lags	$\tau_{\rm r}, \tau_{\rm p}$ , $\tau$	$\Phi_3, \Phi_1$
Δln_pz\$	Trend	1	-4.57**	6.96
	Constant	2	-1.76	5.28*
	None	2	-1.68***	
Δln_r\$	Trend	1	-4.12*	6.07***
	Constant	0	-3.81**	14.49**
	None	0	-3.12**	
Δln rel wscost u	Trend	3	-4.31*	8.30*
	Constant	3	-3.93**	9.05**
	None	0	-3.41**	5755
Δln rel wsu rat	Trend	0	-4.66**	10,87**
	Constant	0	-3.88**	15.05**
	None	0	-2.56*	92.34
Δln s	Trend	2	-0.90	12.08
	Constant	0	-6.36**	40.47**
	None	9	-7.14**	
Δln ss	Trend	0	-2.96	4.49
J. 11	Constant	0 (insign)	-0.94	0.88
	None	0 (insign)	-1.33	0.00
Δln sc_cp	Trend	0	-5.64**	15.94**
oop	Constant	0	-5.72**	32.75**
	None	0	-4.53**	52.75
Δln socind	Trend	5	-4.12*	7.89*
	Constant	3	-3.33*	5.86*
	None	0	-3.79**	2.89
Δtecno index	Trend	0	-5.22**	13.66**
- steno - mayo	Constant	0	-4.45**	19.82**
	None	0	-2.92**	27,54
Δln total pop	Trend	8	-6.86**	596.94**
	Constant	7	0.04	73.39
	None	7	-4.04**	14.46
Δln ucc2 90p	Trend	3	-3.96*	7.91*
	Constant	3	-4.05**	10.11**
	None	0	-3.39**	2018
Δln ucc2 nom	Trend	0	-4.18*	9.12*
	Constant	0	-3.99**	15.93**
	None		4.4.4	*415.5
Δln unempl rat	Trend	3	-5.10**	-9.64*
anompt_tat	Constant	3	-5.39**	11.46**
	None	0	-3.42**	11719

<sup>\*(\*\*)</sup> Significant at a 5(1) percent level.

a At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_r)$ , and -3.00(-3.75) when only a constant is included  $(\tau_\mu)$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level.



Table A17.2 (cont.). Augmented Dickey-Fuller tests for non-stationarity, first differences, 1970-1995

Series	Model	Lags	Tr, Tu, T	$\Phi_3, \Phi_1$
Δln_uniopresind	Trend Constant	4 0	0.76 -4.96**	6.48 24.61**
	None	0	-4.72**	
Δln_vpi	Trend	5	-1.89	3.54
	Constant None	0	-2.15	4.61***
Δln_w_prod	Trend	ı	-4,80**	9.31*
	Constant None	1	-4,31**	11,37**
Δln ws rate	Trend	0	-2.12	2.66
	Constant None	0	-2.36	5,55*
Δln wtot ppi rat	Trend	6	-4.28*	4,38
	Constant	3	-2.89***	4.20***
	None	3	-2.57*	
Δln_wtot_rate	Trend	1	-3.21	3.80
	Constant None	1	-3.37*	5.91*
Δln_wtot_vpi_rat	Trend	0	-3.39	5.75
	Constant	0	-3.23*	10.47**
	None	0	-3.07**	
Δln_wuppi_rat	Trend	0	-3.91	7,69*
	Constant	0	-4.01**	16.05**
	None	0	-3.41**	
Δln_wuvpi_rat	Trend	0	-3.66*	6.73
	Constant	0	-3.19*	10.20**
	None	0	-2.67**	
Δln_xgoud_ppi	Trend	3	-4.81**	6.82
	Constant	0	-3.27*	10.70**
	None	0	-3,34**	
∆ln_xgoud_px	Trend	0	-3.73	7.01
	Constant	0	-3.58*	12.82**
	None	0	-3.66**	

lag.

At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included  $(\tau_t)$ , and -3.00(-3.75) when only a constant is included  $(\tau_\mu)$ , and -1.95(-2.66) when neither is included  $(\tau)$ . The standard t-distribution critical value is -1.708(-2.485).

b At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).

<sup>\*\*\*</sup> Significant at a 10 percent level. (insign) Insignificant



Table A17,3. Augmented Dickey-Fuller tests for non-stationarity, second differences, 1970-

Series	Model	Lags	$\tau_{\tau}, \tau_{\mu}, \tau$	$\Phi_3, \Phi_1$
ΔΔln_ss	Trend	1	-5.36**	16.00**
	Constant	1	-5.37**	24.29**
	None	1	-4.78**	

- a At a 5(1) percent significance level the MacKinnon critical values are -3.60(-4.38) when a trend and a constant are included ( $\tau_{\tau}$ ), and -3.00(-3.75) when only a constant is included ( $\tau_{\mu}$ ), and -1.95(-2.66) when neither is included ( $\tau$ ). The standard t-distribution critical value is -1.708(-2.485).
- At a 5(1) percent significance level the Dickey-Fuller critical values (for 25 observations) are 7.24(10.61) when a trend and a constant are included ( $\Phi_3$ ), and 5.18(7.88) when only a constant is included ( $\Phi_1$ ).
- \*\*\* Significant at a 10 percent level. (insign) Insignificant lag.