Time Series Econometrics For the 21st Century

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Overview

- Most U.S. undergraduate economic students do not pursue PhDs
- Many work for firms and government
- They see, work with, and analyze time-series data
- Time-series tools are useful for such work

Core Models

• Autoregressive (AR)

$$y_t = \alpha + \phi_1 y_{t-1} + \dots + \phi_p y_{t-p} + e_t$$

Regression

$$y_t = \alpha + \delta_0 x_t + e_t$$

Distributed Lag (DL)

$$y_t = \alpha + \delta_0 x_t + \delta_1 x_{t-1} + \dots + \delta_q x_{t-q} + e_t$$

Autoregressive-Distributed Lag (ADL)

$$y_t = \alpha + \phi_1 y_{t-1} + \dots + \phi_p y_{t-p} + \delta_0 x_t + \delta_1 x_{t-1} + \dots + \delta_q x_{t-q} + e_t$$

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Core Insights for Core Time-Series Models

- The coefficients can be estimated by OLS
- Interpropriate standard error (robust or HAC) is important
- The AR model helps understand serial correlation.
- The coefficients in the DL and ADL can be interpreted as multipliers.
- Multipliers are structural under exogeneity
- Lags may be selected by AIC, not testing.
- Spurious regression
- Parameter change.
- Use ADL model for one-step-ahead point forecasts.
- Point forecasts should be combined with interval forecasts.
- Multi-step forecasts can use a multi-step ADL model
- Ø Multi-step point forecasts should be accompanied by fan charts.

Standard Errors

- Three methods popular for time-series
 - Classical (homoskedastic)
 - Robust (Heteroskedastic)
 - HAC (Newey-West)
- Classical (old-fashioned) are not used in contemporary economics.
 - Should only be taught as a stepping stone
- Robust
 - Appropriate for AR and ADL
 - Inappropriate for non-dynamic regression or DL
- HAC
 - Important for regression or DL

Illustration

- Weekly retail gasoline and crude oil prices
- Three standard errors: old-fashioned, robust, and HAC (appropriate)

$$\begin{array}{rcrcrc} \textit{gas}_t = & 0.029 & + & 0.269 & \textit{oil}_t + \widehat{e}_t \\ & (0.046) & (0.011) \\ & (0.046) & (0.015) \\ & (0.073) & (0.021) \end{array}$$

t-statistics versus standard errors

- Always report coefficient estimates & standard errors
- Never coefficient estimates & t-statistics
- Standard errors convey degree of uncertainty always important
- t-statistics concern testing hypothesis of a zero coefficient
 rarely of key interest
- De-emphasize significance testing in favor of measurement and analysis

Autoregressive Models

- Useful for understanding dynamics
- Illustration: U.S. quarterly real GDP growth rates, post-war

$$GDP_{t} = 1.93 + 0.34 \quad GDP_{t-1} + 0.13 \quad GDP_{t-2}$$

$$(0.32) \quad (0.06) \quad (0.06)$$

$$- 0.09 \quad GDP_{t-3} + \hat{e}_{t}$$

$$(0.06)$$

Illustration

- Weekly return on S&P 500
- Useful to illustrate test of efficient market hypothesis
- Also illustrates importance of correct (robust) standard errors, otherwise test will falsely reject.

$$\begin{array}{rcccc} \textit{return}_t = & 0.16 & - & 0.032 & \textit{return}_{t-1} + & 0.037 & \textit{return}_{t-2} + \widehat{e}_t \\ & (0.04) & (0.029) & (0.025) \end{array}$$

Distributed Lag Models

- Useful for understanding multipliers
- Illustration: retail gasoline and crude oil prices

$$gas_{t} = -0.009 + 0.243 \quad oil_{t} + 0.112 \quad oil_{t-1} + 0.063 \quad oil_{t-2} \\ (0.057) \quad (0.016) \quad (0.012) \quad (0.011) \\ + 0.064 \quad oil_{t-3} + 0.030 \quad oil_{t-4} + 0.032 \quad oil_{t-5} + 0.018 \quad oil_{t-6} \\ (0.013) \quad (0.010) \quad (0.011) \quad (0.012) \end{cases}$$

Auto-Regressive Distributed Lag Models

Phillips Curve: U.S. quarterly inflation and unemployment rate

$$\begin{split} \Delta Inf_t &= \begin{array}{cccc} 0.44 & - & 0.34 \\ (0.42) & (0.11) \end{array} \begin{array}{c} \Delta Inf_{t-1} - & 0.39 \\ (0.09) \end{array} \begin{array}{c} \Delta Inf_{t-2} \\ (0.09) \end{array} \\ \\ &- & \begin{array}{c} 0.02 \\ (0.11) \end{array} \begin{array}{c} \Delta Inf_{t-3} - & 0.17 \\ (0.07) \end{array} \begin{array}{c} \Delta Inf_{t-4} - & 1.53 \\ (0.56) \end{array} \begin{array}{c} UR_{t-1} \\ (0.56) \end{array} \\ \\ &+ & \begin{array}{c} 1.58 \\ (1.06) \end{array} \begin{array}{c} UR_{t-2} + & 0.11 \\ (1.03) \end{array} \begin{array}{c} UR_{t-3} - & 0.23 \\ (0.47) \end{array} \begin{array}{c} UR_{t-4} + \widehat{e}_t \end{array}$$

Model Selection

- Economic theory does not inform about lag structure
- In practice, choice implies a bias-variance trade
- Akaike Information Criterion (AIC) is a simple practical tool to compare models
- \bullet Testing (t and F) is appropriate for assessing economic hypotheses
- Testing is inappropriate for model selection

Spurious Regression

- Could be the single most important insight we can teach our students
- Spurious regressions are commonplace
- Teach students to recognize serial correlation and exercise caution when interpreting regressions

Illustration: Two Annual Series

$$y_{1t} = -2.95 + 0.95 \quad y_{2t} + \widehat{e}_t, \qquad R^2 = 0.54 \ (0.52) \qquad (0.07)$$



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January 2017 14 / 22



Both series were generated as independent random walks

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Trump-Era Exchange Rate Theory

- Outsourcing causes displaced workers to be discouraged and leave labor force, and the production shift alters the exchange rate
- $ExchangeRate_t = 10.2$ LaborParticipation_t $R^2 = 0.35$ (0.56)



How to Detect a Spurious Regression

- Dependent variable is highly serially correlated
- Simple regression (no lagged dependent variable)
- Solution: Include at least one lagged dependent variable

$$Ex_{t} = \begin{array}{cccc} 1.43 & Ex_{t-1} - & 0.59 & Ex_{t-2} + & 0.18 & Ex_{t-3} \\ (0.05) & & (0.08) & & (0.09) \end{array}$$
$$- \begin{array}{cccc} 0.05 & Ex_{t-4} - & 0.013 & LaborParticipation_{t} \\ (0.05) & & (0.032) \end{array}$$

Structural Change

Example: U.S. Real GDP Growth Rates

	Mean	Standard Deviation	AR(1) Coefficient
1947-1956	4.0	5.3	0.44
1957-1976	3.6	4.2	0.30
1977-1996	3.2	3.5	0.31
1997-2016	2.3	2.5	0.41

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Forecasting

- Using *h*-step ADL for multi-step point forecasts
- Example: Inflation given unemployment rates
- Estimates

$$\pi_{t+h} - \pi_t = \widehat{\alpha} + \widehat{\phi}_1 \Delta \pi_t + \dots + \widehat{\phi}_p \Delta \pi_{t-p+1} + \widehat{\delta}_1 UR_t + \dots + \widehat{\delta}_q UR_{t-q+1}.$$

Point Forecasts

$$\widehat{\pi}_{n+h} = \pi_n + \widehat{\alpha} + \widehat{\phi}_1 \Delta \pi_n + \dots + \widehat{\phi}_p \Delta \pi_{n-p+1} + \widehat{\delta}_1 U R_n + \dots + \widehat{\delta}_q U R_{n-q+1}$$

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Forecast Intervals

- Uncertainty should be emphasized
- Noise typically dominates signal
- Emphasis on forecasts intervals
- Simple normal approximation interval

$$\widehat{\pi}_{n+h} \pm \widehat{s}_{n+h} z_{1-\alpha/2}$$

• \hat{s}_{n+h} is the standard error of the forecast

•
$$\widehat{s}_{n+h} \approx \left(n^{-1}\sum_{t=1}^{n} \widehat{e}_t^2\right)^{1/2}$$

Fan Charts

- Multi-step forecasts are elegantly presented using fan charts
- Illustration: U.S. quarterly inflation using estimated Phillips curve



Conclusion

- Time-series should be part of econometrics curriculum
- Emphasis on core models used in applications