Seminar to Advanced Macroeconomics

A Cookbook Introduction to the Analysis of Macroeconomic Time Series

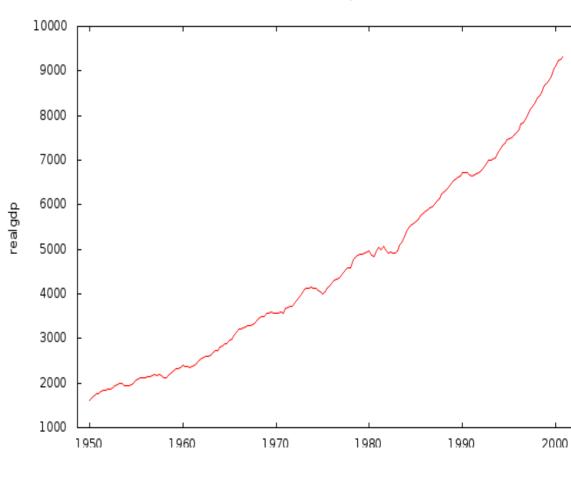
Outline

- Motivation: Identification of trend and cycles
- A quick look at GDP time series
- Trends and their extraction
- Seasonal Adjustment Methods

Motivation

- GDP in the U.S.
 - Clear trend
 - Business cycles?
 - Periods of faster and slower growth
- Extracting the cycles
 - Trend \rightarrow Growth
 - Cycle → monetary and fiscal policy (Taylor rule, golden rules for budget balances)

GDP of the U.S. economy, 1950-2000



A Quick Look at GDP Time Series

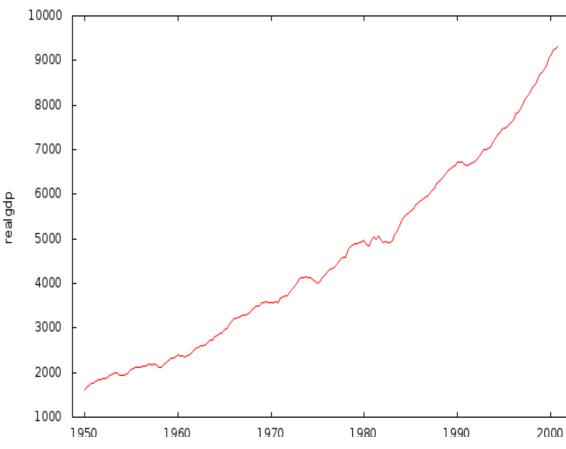
• Data file:

Gretl sample files \rightarrow Greene \rightarrow U.S. Macro

- Exponential growth
 - $gdp_{t+k}=gdp_{t0}(1+\beta)^k$
- Natural approach: Deterministic trend with OLS on logs

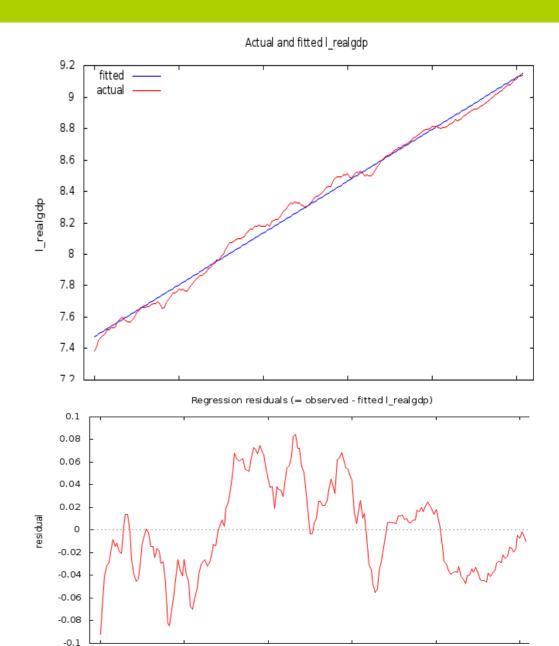
•
$$\log(gdp_{t+k}) = \alpha + \beta^*k + u_t$$

GDP of the U.S. economy, 1950-2000



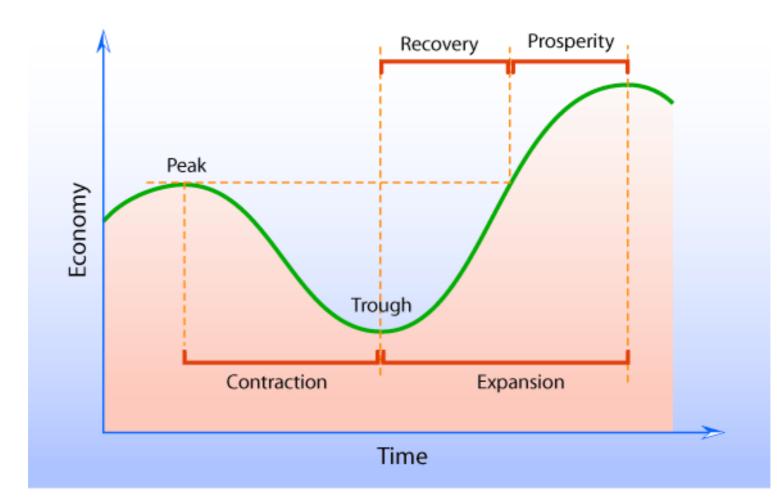
A Quick Look at GDP Time Series

- Linear trend
- Residuals = cycle.
- Some tests necessary:
- Is the trend adequate?
- Do the cycle correspond to our intuition about cycles?



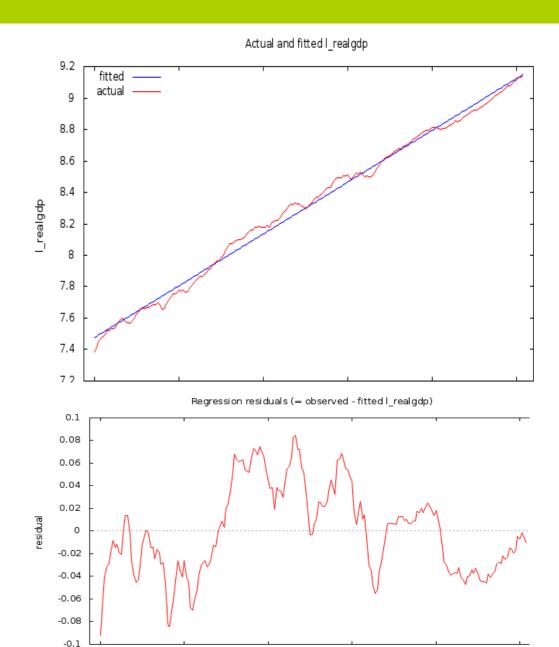
Ideal Business Cycle

(Picture from Wikipedia)



Nature of the Trend

- Intuition about cycles:
- Oscilating... → cycles are temporary.
- It has to be mean reverting, zero mean, no heteroscedasticity
- So called stationarity

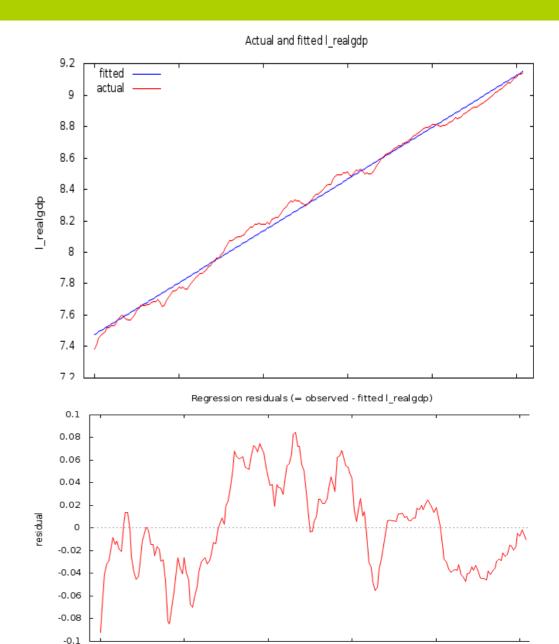


Intermezzo: Stationarity I

- A stationary time series = one whose statistical properties such as mean, variance, autocorrelation, etc. are all constant over time.
- Most statistical forecasting methods assume that the time series can be made approximately stationary through the use of mathematical transformations.
- A stationarized series is relatively easy to predict: you simply predict that its statistical properties will be the same in the future as they have been in the past!
- The predictions for the stationarized series can then be "untransformed," by reversing whatever mathematical transformations were previously used, to obtain predictions for the original series.

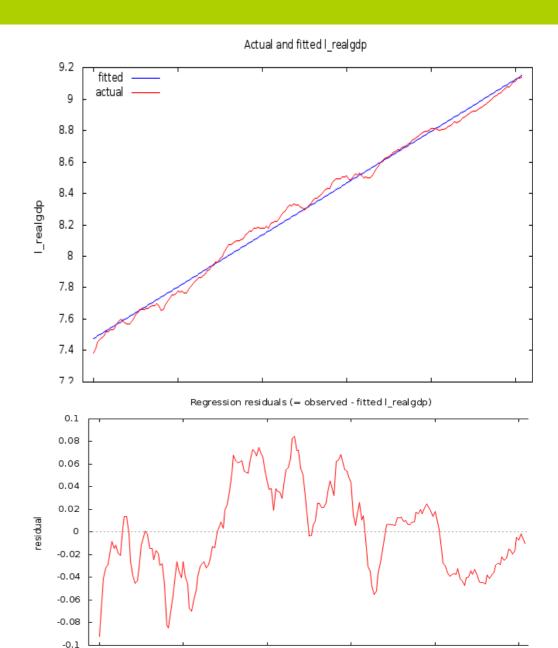
Nature of the Trend

- If stationarity of residuals rejected => trend stationarity does not correspond fully to the true model of the time series
- The trend is somehow stochastic



Nature of the Trend: Structural Breaks

- Sometimes stationarity of cycles rejected because of structural breaks.
- How to detect them?
- And how to deal with them?



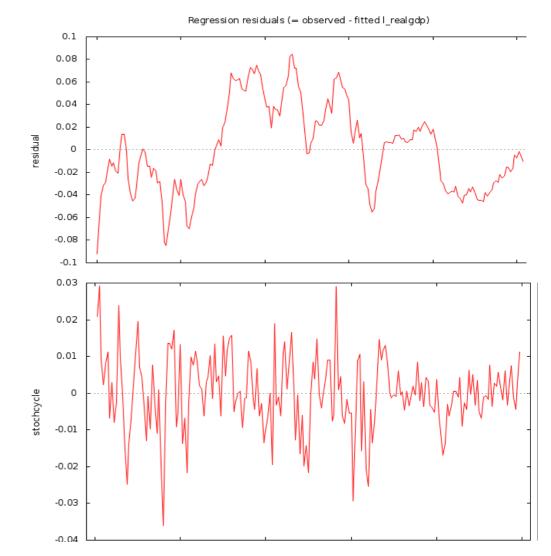
Nature of the Trend

Stochastic trend

Difference-stationarity

- First differences to detect periods with growth below mean
- Or seasonal differences: (y_t-y_{t-4}).

Cycle of the U.S. economy, 1950-2000



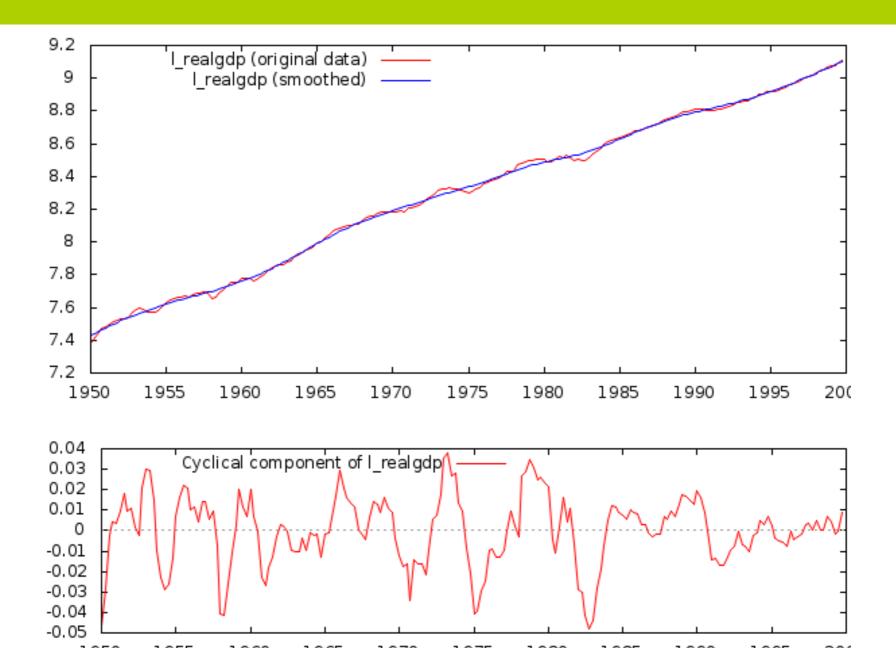
Filtering methods

- Currently most popular approach to separate trend and cycle from original time series.
- Based on frequency domain → it simply extracts frequencies corresponding to the frequencies of business cycles (from 6 to 32 quarters)
- Transformation of spectrum

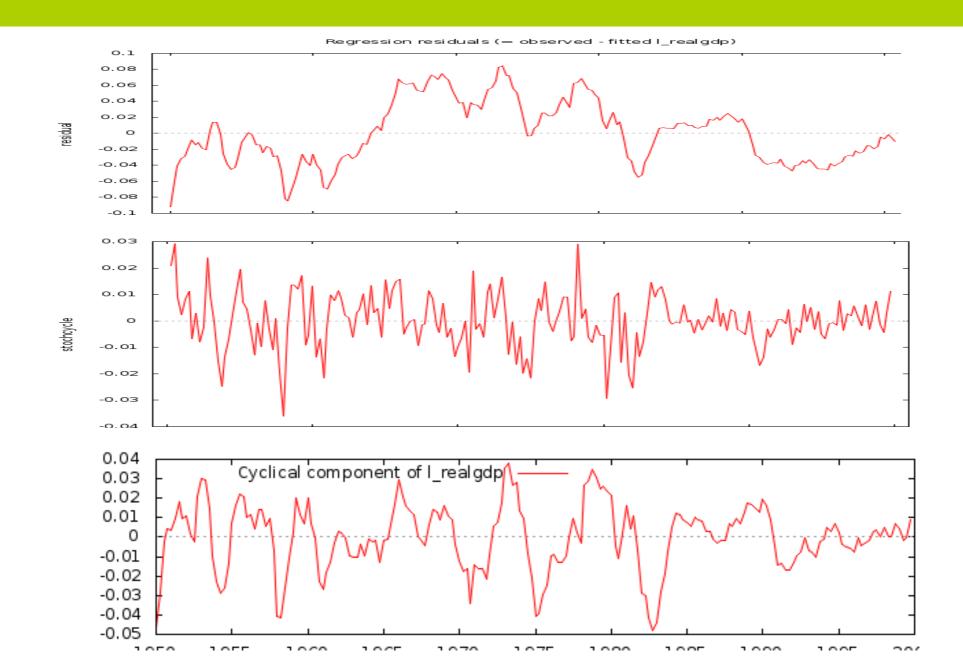
Hodrick-Prescott Filter

- Formula:
- T-1min $\sum_{t=1}^{t} (y_t - \tau_t)^2 + \lambda \sum_{t=2}^{t} [(\tau_{t+1} - \tau_t) - (\tau_t - \tau_{t-1})]^2.$ • $y_t \dots$ actual value of gdp, $\tau_t \dots$ trend in time t min
- λ ... smoothing parameter, for quarterly data set usually to 1600, for annual data 10.
- The value has been set more less arbitrarily: authors commented it that the value to derive "trend, which students of business cycles would draw into the plot of GDP."

Hodrick-Prescott Filter - Results



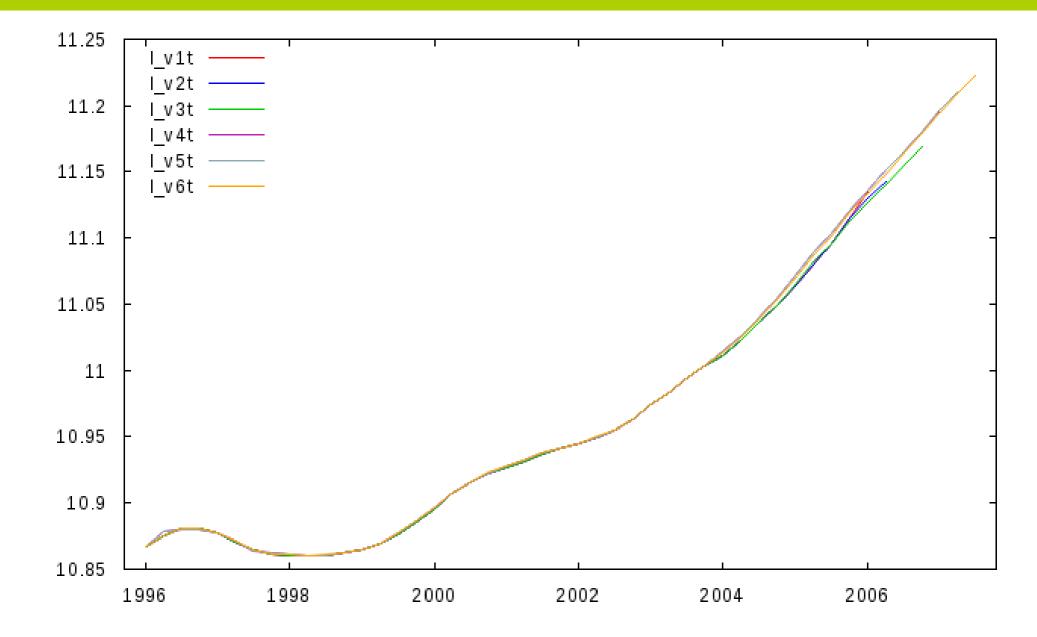
Comparing All the Cycles: The Great Moderation



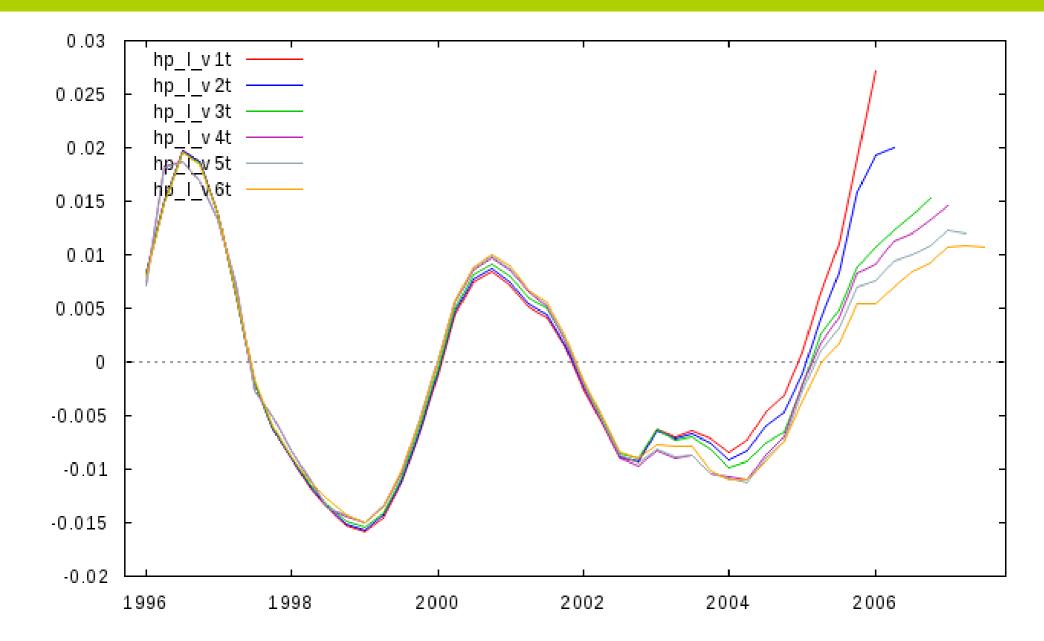
Final Note: Economic Data and Revisions

- Macro data are often revised.
- Example: GDP
- Estimate First calculation Revision Final number 2 years after.
- Causes: all information have to be gathered; minor changes in seasonal adjustment...
- Real-time vs. Ex-post data

Final Note: Economic Data and Revisions

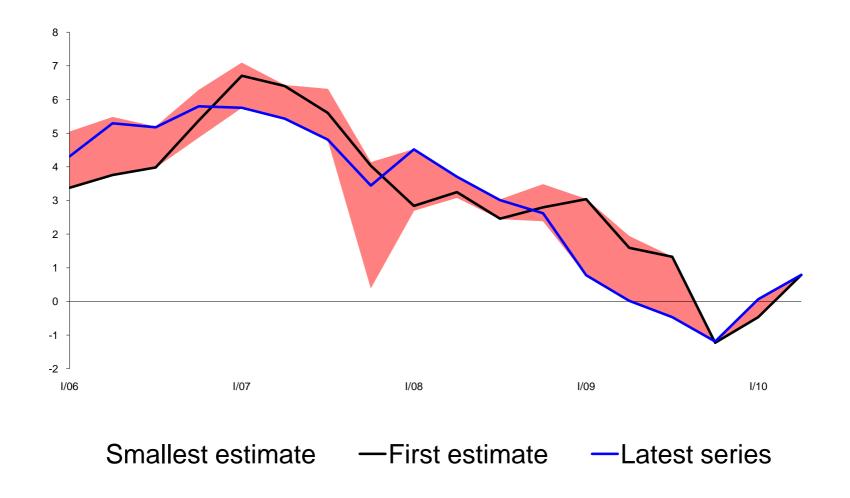


Final Note: Economic Data and Revisions



Final Note: Economic Data and Revisions Magnitude of revisions of household consumption

Constant prices; seasonaly adjusted, annual % change





- Linear trend on logs of data (Log-linear model)
- Cycles stationarity, economic intuition
- Differencing, filtering.
- Hodrick-Prescott filter, smoothing parameter
- Ex-post vs. real-time data, revisions