Macroeconomics II

Lesson 01 — Preliminaries. Stylized Facts and Measurement of Business Cycles

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Preliminaries

The Course

- Macroeconomics II
- Check the syllabus
- Instructors: Alessandro Di Nola and Rafa Serrano-Quintero.

Me

- o rafael.serrano@ub.edu
- o Office hours: Send me an email and we can arrange a meeting.

Syllabus Highlights

What you have to do

- 1. Three problem sets
- 2. Midterm exam. 50% of the grade of the final exam. Other 50% will be Alessandro's part.

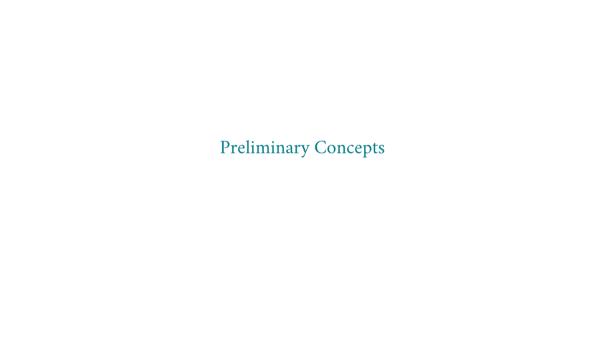
Materials

- o In class notes and slides. Most of the work will be in the blackboard.
- Useful references:
 - o Cooley (1995)

King and Rebelo (1999)

o Judd (1998)

Ljungqvist and Sargent (2004)



Preliminary Concepts

Modern macroeconomics is

- o dynamic: things change and are related over time.
- **stochastic:** there are random events. Expectations matter.
- We denote X_t the realization of variable X at time t.
- Models are stochastic and people try to guess what will happen in the future. We call that expectations.

Notation and Expectations

- If we do not know anything about the current state of the system, we call it **unconditional expectation** and express it as $\mathbb{E}(X_{t+1})$.
- If we condition on what we know at time t, we call the **conditional expectation** and we express it as $\mathbb{E}(X_{t+1}|\Omega_t)$ where Ω_t is what we know at time t. Usually, we shorten it by writing $\mathbb{E}_t(X_{t+1})$.
- Note that $\mathbb{E}_t(X_t) = X_t$ and $\mathbb{E}_t(X_{t-k}) = X_{t-k}$ for all k > 0.

Theorem 1 (Law of Iterated Expectations)

Let Y and Z be two arbitrary random variables, then

$$\mathbb{E}(Y) = \mathbb{E}(\mathbb{E}(Y|Z))$$

i.e. the expected value of the conditional expected value of Y given Z is the same as the expected value of Y.

Notation and Expectations

• The LIE (Theorem 1) has the following implication

$$\mathbb{E}_t(\mathbb{E}_{t+1}(X_{t+2})) = \mathbb{E}_t(X_{t+2})$$

- Rational expectations (Muth 1961) is a stronger than simple expectations. Two
 conditions on expectations of future realizations of variables
 - 1. Correct on average.
 - 2. Unpredictable forecasting errors given current info.
- This implies the agents "know the model and use it to act."
- Does this imply agents do not make mistakes? No!!

Stochastic Processes

Definition 2 (Markov Property)

A stochastic process has the Markov property if

$$\Pr[X_t = x | X_0, \dots, X_{t-1}] \equiv \Pr[X_t = x | X_{t-1}]$$

i.e. that the conditional probability of future states depends upon the present state only. If you know X_{t-1} knowing X_0, \ldots, X_{t-2} does not give you extra information.

Definition 3 (ARMA(p, q) Process)

An ARMA(p, q) process can be expressed as

$$X_t = c + \sum_{i=1}^p \rho_i X_{t-i} + \sum_{i=1}^q \theta_i \varepsilon_{t-i}$$

where c is a constant, ρ_i are the autorregressive parameters, and θ_i are the moving average parameters.

Stochastic Processes

- ARMA processes can sometimes be approximated with sufficiently long AR processes.
- An AR(p) process does not strictly have the Markov property. However

$$\begin{pmatrix} s_t \\ s_{t-1} \\ \vdots \\ s_{t-p+1} \end{pmatrix} = \begin{pmatrix} \rho_1 & \rho_2 & \cdots & \rho_p \\ 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix} \begin{pmatrix} s_{t-1} \\ s_{t-2} \\ \vdots \\ s_{t-p} \end{pmatrix} + \begin{pmatrix} \varepsilon_t \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

which reduces to

$$\mathbf{s}_t = \Lambda \mathbf{s}_{t-1} + \varepsilon_t$$

Impulse Response Functions

If we give a shock to some exogenous variable. How do the endogenous variables react?

$$IRF(h) = \mathbb{E}_t(X_{t+h}) - \mathbb{E}_{t-1}(X_{t+h}|\varepsilon_t = \varepsilon)$$

Example 4

Take the AR(1) process

$$X_t = \rho X_{t-1} + \varepsilon_t$$

suppose $\varepsilon \sim \mathcal{N}(0,1)$ and suppose at time t there is a shock of $\varepsilon_t = 1$. Compute the IRF after h periods.

Impulse Response Functions

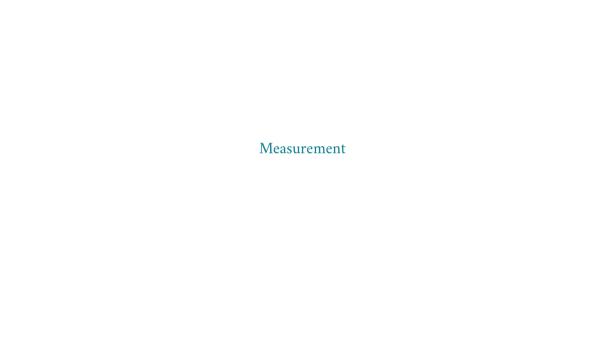
Take a multivariate process and assume **X** is a 2×1 vector

$$\mathbf{X}_t = \mathbf{A}\mathbf{X}_{t-1} + \mathbf{B}\boldsymbol{\varepsilon}_t$$

Assume also that off-diagonals elements of **A** and **B** are not zero. The IRF is now a vector

$$IRF_1(h) = \mathbf{A}^{h-1}\mathbf{B}_1$$
$$IRF_2(h) = \mathbf{A}^{h-1}\mathbf{B}_2$$

Measurement and Stylized Facts of Business Cycles



Trend and Cycle

- o In Macro I you have studied the macroeconomics of the long-run. Why countries grow.
- Now we will focus on booms and busts or expansions and recessions.
- We first need to separate trend and cycle. Then, identify what are expansion and recession periods and define them formally.

Trend and Cycle

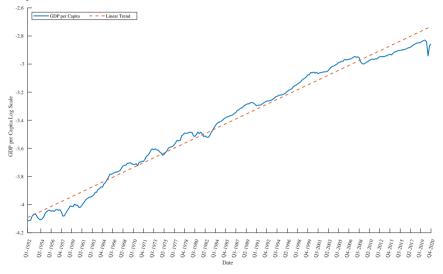


Figure 1: U.S. GDP per capita in Log Scale

Trend and Cycle

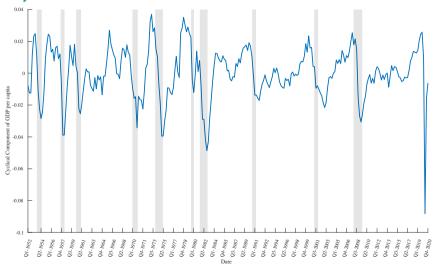


Figure 2: Cycles of U.S. GDP per capita in Log Scale

Trend and Cycle — Identification

- How can we go from trend to cycle? How do we remove this?
- Let y_t be GDP per capita and decompose it into a trend and a cycle component.

$$y_t = y_t^c + y_t^g$$

- How can we estimate these two? Several options:
 - Spectral analysis (see Hamilton 1994, Chapter 6)
 - Describe y_t as a weighted sum of periodic functions like $\cos(\omega t)$ and $\sin(\omega t)$. Identify peaks and troughs and call that a cycle.
 - Detrending methods.
 - o Linear detrending, first differences...
 - Filters: band-pass (Baxter and King 1999), Hodrick and Prescott (1997).

Hodrick-Prescott Filter

The method separates the trend component from the solution of the problem

$$\min_{\{y_t^g\}_{t=0}^{\infty}} \quad \sum_{t=0}^{\infty} (y_t - y_t^g)^2 + \lambda \left[(y_{t+1}^g - y_t^g) - (y_t^g - y_{t-1}^g) \right]^2$$

where λ is a smoothing parameter. The cycle is thus

$$y_t^c = y_t - y_t^g$$

Widely used method but with many critiques. From Hamilton (2018):

- 1. Introduces spurious dynamic relations w/o basis in the underlaying DGP and filtered values at the end are very different from the middle.
- 2. Choosing λ from a statistical formalization implies \neq to the typically applied.
- 3. Not good for prediction, double-sided.

Hodrick-Prescott Filter

Alternatives

- Hamilton (2018) proposes instead a regression approach.
- How different is the value at date t + h from the expected value at t?
- Suggests to estimate by OLS

$$y_{t+h} = \beta_0 + \beta_1 y_t + \beta_2 y_{t-1} + \beta_3 y_{t-2} + \beta_4 y_{t-3} + \nu_{t+h}$$

The residuals

$$\hat{v}_{t+h} = y_{t+h} - \hat{\beta}_0 - \hat{\beta}_1 y_t - \hat{\beta}_2 y_{t-1} - \hat{\beta}_3 y_{t-2} - \hat{\beta}_4 y_{t-3}$$

Filtering Techniques

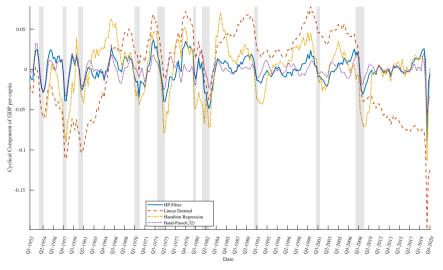


Figure 3: Cycles of U.S. GDP per capita in Log Scale



Business Cycle Statistics

When we study empirical properties of business cycles we typically look at three moments of the data

- 1. **Volatility:** Relative standard deviation sd(x)/sd(gdp). Amplitude of fluctuations of x_t relative to GDP.
- 2. Cyclicality: the correlation between x_t and GDP $\rho(x_t, GDP)$
 - If $\rho(x_t, GDP) > 0$ then x_t is **procyclical**.
 - If $\rho(x_t, GDP) < 0$ then x_t is **countercyclical**.
 - If $\rho(x_t, GDP) = 0$ then x_t is acyclical.
- 3. **Persistence:** autocorrelation $\rho(x_t, x_{t-1})$

Business Cycle Statistics

Table 1: Business Cycle Statistics — Sims (2017)

Series	Std. Dev.	Rel. Std. Dev.	Corr w/y_t	Autocorr	Corr w/Y_{t-4}	Corr w/Y_{t+4}
Output	0.017	1.00	1.00	0.85	0.07	0.11
Consumption	0.009	0.53	0.76	0.79	0.07	0.22
Investment	0.047	2.76	0.79	0.87	-0.10	0.26
Hours	0.019	1.12	0.88	0.90	0.29	-0.03
Productivity	0.011	0.65	0.42	0.72	-0.50	0.35
Wage	0.009	0.53	0.10	0.73	-0.10	0.10
1+ Interest Rate	0.004	0.24	0.00	0.42	0.27	-0.25
Price Level	0.009	0.53	-0.13	0.91	0.09	-0.41
TFP	0.012	0.71	0.76	0.75	-0.34	0.34

Business Cycle Stylized Facts

Volatility:

○ Investment \gg GDP. Consumption \ll GDP . Hours \approx GDP. Prices \ll GDP.

Cyclicality:

- Consumption, investment, hours are procyclical.
- Real wages and real interest rate acyclical. If used other expectations, real interest rate
 is slightly countercyclical. Moder macro models struggle to match this feature of the
 data.

Persistence: typically, all variables are strongly persistent.