# Notes on Econ 102 Sect. 2, Fall 2005 

 José-Víctor Ríos-Rull, , University of PennsylvaniaChapter 4 Growth Accounting<br>Evolved from notes written by Jesús Fernández-Villaverde

October 3, 2005

## 4 Growth Accounting

- Output is produced by inputs capital $K$ and labor $L$, in combination with the available technology $A$
- Want: decompose the growth rate of output into the growth rate of capital input, the growth rate of labor input and technological progress. This exercise is called growth accounting.
- Why?


## Aggregate production function

- Maps inputs into output:

$$
Y=F(A, K, L)
$$

$A$ is called total factor productivity (TFP).

- Cobb-Douglas example:

$$
Y=A K^{\alpha} L^{1-\alpha}
$$

- Interpretation.


## Discrete vs. Continuous Time

- In discrete time a variable is indexed by time: $x_{t}$.
- In continuous time a variable is a function of time: $x(t)$.
- We observe the world only in discrete time...
- but it is often much easier to work with continuous time!


## From Growth Rates and Logarithms

- Remember that

$$
g_{x}(t-1, t)=\frac{x_{t}-x_{t-1}}{x_{t-1}}
$$

## From Growth Rates and Logarithms

- Remember that and that

$$
\begin{aligned}
g_{x}(t-1, t) & =\frac{x_{t}-x_{t-1}}{x_{t-1}} \\
1+g_{x}(t-1, t) & =\frac{x_{t}}{x_{t-1}}
\end{aligned}
$$

## From Growth Rates and Logarithms

- Remember that and that

$$
\begin{aligned}
g_{x}(t-1, t) & =\frac{x_{t}-x_{t-1}}{x_{t-1}} \\
1+g_{x}(t-1, t) & =\frac{x_{t}}{x_{t-1}}
\end{aligned}
$$

- Take logs

$$
\log \left[1+g_{x}(t-1, t)\right]=\log \left(\frac{x_{t}}{x_{t-1}}\right)
$$

## From Growth Rates and Logarithms

- Remember that and that

$$
\begin{aligned}
g_{x}(t-1, t) & =\frac{x_{t}-x_{t-1}}{x_{t-1}} \\
1+g_{x}(t-1, t) & =\frac{x_{t}}{x_{t-1}}
\end{aligned}
$$

$$
\log \left[1+g_{x}(t-1, t)\right]=\log \left(\frac{x_{t}}{x_{t-1}}\right)
$$

- Taylor series expansion of $\log (1+y)$ around $y=0$ :

$$
\left.\log (1+y)\right|_{y=0}=\ln 1+\frac{1}{1!} y+\text { higher order terms } \simeq y
$$

## From Growth Rates and Logarithms

- Remember that

$$
\begin{aligned}
g_{x}(t-1, t) & =\frac{x_{t}-x_{t-1}}{x_{t-1}} \\
1+g_{x}(t-1, t) & =\frac{x_{t}}{x_{t-1}}
\end{aligned}
$$ and that

- Take logs

$$
\log \left[1+g_{x}(t-1, t)\right]=\log \left(\frac{x_{t}}{x_{t-1}}\right)
$$

- Taylor series expansion of $\log (1+y)$ around $y=0$ :

$$
\left.\log (1+y)\right|_{y=0}=\ln 1+\frac{1}{1!} y+\text { higher order terms } \simeq y
$$

- Then:

$$
\begin{aligned}
\log \left[1+g_{x}(t-1, t)\right] & \simeq g_{x}(t-1, t) \simeq \log \left(\frac{x_{t}}{x_{t-1}}\right) \\
g_{x}(t-1, t) & \simeq \log x_{t}-\log x_{t-1}=\Delta \log x_{t}
\end{aligned}
$$

## From Growth Rates and Logarithms

- Remember that

$$
\begin{aligned}
g_{x}(t-1, t) & =\frac{x_{t}-x_{t-1}}{x_{t-1}} \\
1+g_{x}(t-1, t) & =\frac{x_{t}}{x_{t-1}}
\end{aligned}
$$

and that

- Take logs

$$
\log \left[1+g_{x}(t-1, t)\right]=\log \left(\frac{x_{t}}{x_{t-1}}\right)
$$

- Taylor series expansion of $\log (1+y)$ around $y=0$ :

$$
\left.\log (1+y)\right|_{y=0}=\ln 1+\frac{1}{1!} y+\text { higher order terms } \simeq y
$$

- Then:

$$
\begin{aligned}
\log \left[1+g_{x}(t-1, t)\right] & \simeq g_{x}(t-1, t) \simeq \log \left(\frac{x_{t}}{x_{t-1}}\right) \\
g_{x}(t-1, t) & \simeq \log x_{t}-\log x_{t-1}=\Delta \log x_{t}
\end{aligned}
$$

- Remember from calculus that validity of Taylor series expansion is local: $g$ small!


## Moving between Continuous and Discrete Time

- Let $x(t)$ be a variable that depends of $t$. We write

$$
\dot{x}(t) \equiv \frac{d x(t)}{d t} .
$$

## Moving between Continuous and Discrete Time

- Let $x(t)$ be a variable that depends of $t$. We write

$$
\dot{x}(t) \equiv \frac{d x(t)}{d t}
$$

- Time derivative: $\quad \dot{x}(t)=\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}$


## Moving between Continuous and Discrete Time

- Let $x(t)$ be a variable that depends of $t$. We write

$$
\dot{x}(t) \equiv \frac{d x(t)}{d t}
$$

- Time derivative: $\dot{x}(t)=\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}$
- Take logs

$$
\frac{d \log [x(t)]}{d t}=\frac{\dot{x}(t)}{x(t)}=g_{x}(t)
$$

## Moving between Continuous and Discrete Time

- Let $x(t)$ be a variable that depends of $t$. We write

$$
\dot{x}(t) \equiv \frac{d x(t)}{d t}
$$

- Time derivative: $\dot{x}(t)=\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}$
- Take logs

$$
\frac{d \log [x(t)]}{d t}=\frac{\dot{x}(t)}{x(t)}=g_{x}(t)
$$

- Then:

$$
\frac{\dot{x}(t)}{x(t)}=\frac{\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}}{x(t)}
$$

## Moving between Continuous and Discrete Time

- Let $x(t)$ be a variable that depends of $t$. We write $\quad \dot{x}(t) \equiv \frac{d x(t)}{d t}$.
- Time derivative: $\dot{x}(t)=\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}$
- Take logs

$$
\frac{d \log [x(t)]}{d t}=\frac{\dot{x}(t)}{x(t)}=g_{x}(t)
$$

$$
\frac{\dot{x}(t)}{x(t)}=\frac{\lim _{\Delta t \rightarrow 0} \frac{x(t+\Delta t)-x(t)}{\Delta t}}{x(t)}
$$

- When $\Delta t$ is small (let's say a year):

$$
g_{x}(t)=\frac{\dot{x}(t)}{x(t)} \simeq \frac{x(t+1)-x(t)}{x(t)}=g_{x}(t-1, t) \simeq \Delta \log x_{t}
$$

Growth Rates of Ratios: Suppose $k(t)=\frac{K(t)}{L(t)}$. What is $g_{k}(t)$ ?

Growth Rates of Ratios: Suppose $k(t)=\frac{K(t)}{L(t)}$. What is $g_{k}(t)$ ?

1 Take logs

$$
\log (k(t))=\log (K(t))-\log (L(t))
$$

Growth Rates of Ratios: Suppose $k(t)=\frac{K(t)}{L(t)}$. What is $g_{k}(t)$ ?

1 Take logs

$$
\log (k(t))=\log (K(t))-\log (L(t))
$$

2 Differentiate with respect to time

$$
\begin{aligned}
\frac{d \log ((k(t))}{d t} & =\frac{d \log (K(t))}{d t}-\frac{d \log (L(t))}{d t} \\
\frac{\dot{k}(t)}{k(t)} & =\frac{\dot{K}(t)}{K(t)}-\frac{\dot{L}(t)}{L(t)} \\
g_{k}(t) & =g_{K}(t)-g_{L}(t)
\end{aligned}
$$

Growth Rates of Ratios: Suppose $k(t)=\frac{K(t)}{L(t)}$. What is $g_{k}(t)$ ?

1 Take logs

$$
\log (k(t))=\log (K(t))-\log (L(t))
$$

2 Differentiate with respect to time

$$
\begin{aligned}
\frac{d \log ((k(t))}{d t} & =\frac{d \log (K(t))}{d t}-\frac{d \log (L(t))}{d t} \\
\frac{\dot{k}(t)}{k(t)} & =\frac{\dot{K}(t)}{K(t)}-\frac{\dot{L}(t)}{L(t)} \\
g_{k}(t) & =g_{K}(t)-g_{L}(t)
\end{aligned}
$$

3 Growth rate of a ratio $=$ the difference of the growth rates:

$$
g_{k}(t)=g_{K}(t)-g_{L}(t)
$$

Growth Rates of Ratios: Suppose $k(t)=\frac{K(t)}{L(t)}$. What is $g_{k}(t)$ ?

1 Take logs

$$
\log (k(t))=\log (K(t))-\log (L(t))
$$

2 Differentiate with respect to time

$$
\begin{aligned}
\frac{d \log ((k(t))}{d t} & =\frac{d \log (K(t))}{d t}-\frac{d \log (L(t))}{d t} \\
\frac{\dot{k}(t)}{k(t)} & =\frac{\dot{K}(t)}{K(t)}-\frac{\dot{L}(t)}{L(t)} \\
g_{k}(t) & =g_{K}(t)-g_{L}(t)
\end{aligned}
$$

3 Growth rate of a ratio $=$ the difference of the growth rates:

$$
g_{k}(t)=g_{K}(t)-g_{L}(t)
$$

4 A constant ratio that both variables grow at same rate:

$$
g_{k}(t)=0 \Rightarrow g_{K}(t)=g_{L}(t)
$$

Growth Rates of Weighted Products. Suppose $Y(t)=K(t)^{\alpha} L(t)^{1-\alpha}$ What is $g_{Y}(t)$ ?
[1] Take logs $\quad \log (Y(t))=\alpha \log (K(t))+(1-\alpha) \log (L(t))$
[2] Differentiate

$$
\begin{aligned}
\frac{d \log (Y(t))}{d t} & =\alpha \frac{d \log (K(t))}{d t}+(1-\alpha) \frac{d \log (L(t))}{d t} \\
\frac{\dot{Y}(t)}{Y(t)} & =\alpha \frac{\dot{K}(t)}{K(t)}+(1-\alpha) \frac{\dot{L}(t)}{L(t)} \\
g_{Y}(t) & =\alpha g_{K}(t)+(1-\alpha) g_{L}(t)
\end{aligned}
$$

Growth rate $=$ weighted sum, (weights equal to share parameters)

## Growth Accounting

- Observations in discrete time.
- Production Function: $Y(t)=F(A(t), K(t), L(t))$
- Differentiating with respect to time and dividing by $Y(t)$

$$
\frac{\dot{Y}(t)}{Y(t)}=\frac{F_{A} A(t)}{Y(t)} \frac{\dot{A}(t)}{A(t)}+\frac{F_{k} K(t)}{Y(t)} \frac{\dot{K}(t)}{K(t)}+\frac{F_{L} L(t)}{Y(t)} \frac{\dot{L}(t)}{L(t)}
$$

- Useful benchmark: Cobb-Douglas $Y(t)=A(t) K(t)^{\alpha} L(t)^{1-\alpha}$.
- Why? (factor shares)
- Taking logs and differentiating with respect to time gives

$$
g_{Y}(t)=g_{A}(t)+\alpha g_{K}(t)+(1-\alpha) g_{L}(t)
$$

- $g_{A}$ is called TFP growth or multifactor productivity growth.

The Cobb-Douglas Production Function and its properties

- Under competition factors of production are paid their marginal productivities.
$w=\frac{\partial F(K, L)}{\partial L}=(1-\alpha) A K^{\alpha} L^{-\alpha} \quad r=\frac{\partial F(K, L)}{\partial K}=\alpha A K^{\alpha-1} L^{1-\alpha}$
- Labor and capital shares of output in any year $t$ are

$$
\begin{aligned}
\frac{W_{t} L_{t}}{Y_{t}} & =\frac{(1-\alpha) A_{t} K_{t}^{\alpha} L_{t}^{-\alpha} L_{t}}{Y_{t}}=1-\alpha \\
\frac{r_{t} K_{t}}{Y_{t}} & =\frac{\alpha A_{t} K_{t}^{\alpha-1} L_{t}^{1-\alpha} K_{t}}{Y_{t}}=\alpha
\end{aligned}
$$

- Cobb-Douglas production functions are consistent with the Kaldor facts.


## Doing the Accounting

- Pick $\alpha$ (capital share).
- Measure $g_{Y}, g_{K}$ and $g_{L}$ from the data.
- $g_{A}$ is the residual $g_{A}(t)=g_{Y}(t)-\alpha g_{K}(t)-(1-\alpha) g_{L}(t)$
- Therefore $g_{A}$ is also called the Solow residual.
- Severe problems if missmeasurement ( $g_{K}$ is hard to measure).


## Data for the US

- We pick $\alpha=\frac{1}{3}$

| Per. | $g_{Y}$ | $\alpha g_{K}$ | $(1-\alpha) g_{L}$ | TFP $\left(g_{A}\right)$ |
| :--- | :--- | :--- | :--- | :--- |
| $48-98$ | 2.5 | $0.8(32 \%)$ | 0.2 | $(8 \%)$ |
| $48-73$ | 3.3 | $1.0(30 \%)$ | 0.2 | $(6 \%)$ |
| $48-1.4)$ | $2.1(64 \%)$ |  |  |  |
| $73-95$ | 1.5 | $0.7(47 \%)$ | $0.3(20 \%)$ | $0.6(33 \%)$ |
| $95-98$ | 2.5 | $0.8(32 \%)$ | $0.3(12 \%)$ | $1.4(56 \%)$ |

- Key observation: Productivity Slowdown in the 70's
- Note: the late 90 's look much better


## Reasons for the Productivity Slowdown

1. Sharp increases in the price of oil in 70's
2. Structural changes: more services and less and less manufacturing goods produced
3. Slowdown in resources spent on R\&D in the late 60's.
4. TFP was abnormally high in the 50 's and 60 's
5. Information technology (IT) revolution in the 70's

## Growth Accounting for Other Countries

- One key question: was fast growth in East Asian growth miracles mostly due to technological progress or mostly due to capital accumulation?
- Why is this an important question?

| Country | Per. | $g_{Y}$ | $\alpha$ | $\alpha g_{K}$ | $(1-\alpha) g_{L}$ | $g_{A}$ |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| Germany | $60-90$ | 3.2 | 0.4 | $59 \%$ | $-8 \%$ | $49 \%$ |
| Italy | $60-90$ | 4.1 | 0.38 | $49 \%$ | $3 \%$ | $48 \%$ |
| UK | $60-90$ | 2.5 | 0.39 | $52 \%$ | $-4 \%$ | $52 \%$ |
| Argentina | $40-80$ | 3.6 | 0.54 | $43 \%$ | $26 \%$ | $31 \%$ |
| Brazil | $40-80$ | 6.4 | 0.45 | $51 \%$ | $20 \%$ | $29 \%$ |
| Chile | $40-80$ | 3.8 | 0.52 | $34 \%$ | $26 \%$ | $40 \%$ |
| Mexico | $40-80$ | 6.3 | 0.63 | $41 \%$ | $23 \%$ | $36 \%$ |
| Japan | $60-90$ | 6.8 | 0.42 | $57 \%$ | $14 \%$ | $29 \%$ |
| Hong Kong | $66-90$ | 7.3 | 0.37 | $42 \%$ | $28 \%$ | $30 \%$ |
| Singapore | $66-90$ | 8.5 | 0.53 | $73 \%$ | $31 \%$ | $-4 \%$ |
| South Korea | $66-90$ | 10.3 | 0.32 | $46 \%$ | $42 \%$ | $12 \%$ |
| Taiwan | $66-90$ | 9.1 | 0.29 | $40 \%$ | $40 \%$ | $20 \%$ |

