Advanced Macroeconomics II

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An Economic Application of the Simple RBC Model

Estimating Economic Effects of Political Movements in China

Kwan and Chow, Journal of Comparative Economics 23, 192-208 (1996)

- Investment is determined by a central planner maximizing a multiperiod objective function.
- Political events are modeled by exogenous changes in the shocks to productivity and to investment.
- Model parameters are estimated with MLE.
- Effects of the events are measured by comparing the time paths generated by the model with and without the changes in the shocks.

Introduction: the question

Problem

What were the economic effects of the Great Leap Forward Movement in 1958-1962 and the Cultural Revolution in 1966-1969 in China?

Methodology Compare the historical time paths of the economy with the paths that would have prevailed absent the above events.

Tools

- An RBC model to explain the growth of Chinese economy.
- Solving the model with numerical method.
- Estimation: MLE
- Simulation.

Model and Data

Preference: from a social planner

$$E_t \sum_{i=1}^{\infty} \beta^{t+i} \log c_{t+i} \tag{1}$$

Technology:

Aggregate real output

$$Q_t = A_t K_t^{1-\alpha} L_t^{\alpha} \tag{2}$$

Denoting $q_t = Q/L$, k = K/L, net investment per laborer i = I/L,

$$q_t = A_t k_t^{1-\alpha} \tag{3}$$

$$q_t = c_t + i_t \tag{4}$$

$$k_{t+1} = k_t + i_t \tag{5}$$

$$\ln A_{t+1} = \gamma + \ln A_t + \eta_{t+1} \tag{6}$$

Endowment and **information** as so defined in the previous baseline model.

Model and Data

Treatment on data

- Q: National income (Statistical Yearbook of China (SYC) 1994) devided by price deflator (national income in current price to national income in 1952 price).
- $Q_t = C_t + I_t$, in Chinese official statistics.
- Initial estimate of capital K=2213(unit: 100 million yuan), from the estimate of Chow (1993b, p.821), In 1952, k=K/L. In later years, as defined by $k_{t+1}=k_t+i_t$. (An approximation)

$$K_{t+1} = K_t + I_t$$

$$\frac{K_{t+1}}{L_t} = \frac{K_t}{L_{t-1}} \frac{L_{t-1}}{L_t} + \frac{I_t}{L_t}, \text{ or } \frac{K_{t+1}}{L_{t+1}} \frac{L_{t+1}}{L_t} = \frac{K_t}{L_t} + \frac{I_t}{L_t}$$

$$k_{t+1} = k_t \frac{1}{1 + n_t} + i_t, \text{ or } k_{t+1}(1 + n_t) = k_t + i_t$$

 Admitted shortcomings: treating technology, population, and labor force as exogenous. An important step however...

Statistical Estimation I: solving the model

Detrending:

Convert the variables to stationary processes to avoid unit root problem in Eq. (6)

How? - To detrend all variables along their balanced growth paths.

$$z_t \equiv A_t^{1/\alpha}, \quad \bar{k}_{t+1} \equiv k_{t+1}/z_t, \quad \bar{c}_t \equiv c_t/z_t, \quad \bar{z}_t \equiv z_t/z_{t-1}.$$
 (7)

Hence,

$$\begin{array}{lll} \ln A_{t+1} &=& \gamma + \ln A_t + \eta_{t+1} \\ \alpha \ln z_{t+1} &=& \gamma + \alpha \ln z_t + \eta_{t+1} \\ \ln z_{t+1} &=& \gamma / \alpha + \ln z_t + \eta_{t+1} / \alpha \text{ (unit root)} \end{array}$$

Define

$$\mu \equiv \gamma/\alpha$$
, $\varepsilon_{t+1} \equiv \eta_{t+1}/\alpha$

$$\ln \bar{z}_{t+1} = \ln z_{t+1}/z_t = \mu + \varepsilon_{t+1}$$

Statistical Estimation I: solving the model

Detrending:

Combining Eq. (3) to (5) to get the intertemporal budget constraint with only capital and consumption,

$$c_t + k_{t+1} = A_t k_t^{1-\alpha} + k_t,$$
 (8)

transforming

$$c_{t}/z_{t} + k_{t+1}/z_{t} = z_{t}^{\alpha-1} k_{t}^{1-\alpha} + (k_{t}/z_{t-1}) z_{t-1}/z_{t}$$

$$\bar{c}_{t} + \bar{k}_{t+1} = \bar{k}_{t}^{1-\alpha} \bar{z}_{t}^{\alpha-1} + \bar{k}_{t} \bar{z}_{t}^{-1}$$
(9)

Statistical Estimation I: solving the model

The dynamic optimization problem under the transformed system

$$\max_{(c_t, k_{t+1})_{t=0}^{\infty}} E_0 \left[\sum_{t=0}^{\infty} \beta^t \ln \bar{c}_t \right]$$

[Why can we do this from maximizing equation (1)?]

$$s.t.\bar{c}_t + \bar{k}_{t+1} = \bar{k}_t^{1-\alpha} \bar{z}_t^{\alpha-1} + \bar{k}_t \bar{z}_t^{-1}$$

 $\ln \bar{z}_{t+1} = \mu + \varepsilon_{t+1}$

Statistical Estimation I: solving the model

FONCs:

$$E_t \left[\beta \frac{\bar{c}_t}{\bar{c}_{t+1}} R_{t+1} \right] = 1 \tag{10}$$

$$R_{t+1} \equiv \left(1 + (1-\alpha)\bar{k}_{t+1}^{-\alpha}\bar{z}_{t+1}^{\alpha}\right)/\bar{z}_{t+1} \tag{11}$$

$$\bar{c}_t + \bar{k}_{t+1} = \bar{k}_t^{1-\alpha} \bar{z}_t^{\alpha-1} + \bar{k}_t \bar{z}_t^{-1}$$
 (12)

$$\ln \bar{z}_{t+1} = \mu + \varepsilon_{t+1} \tag{13}$$

Statistical Estimation I: solving the model

Steady states (sorry for the abuse of notation):

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; =
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Statistical Estimation I: solving the model

Log-linearization: (denote $\hat{x}_t \equiv \ln(\bar{x}_t/\ddot{x})$)

Statistical Estimation I: solving the model

Postulate a linear recursive law of motion for \hat{c}_t and \hat{k}_t

$$\hat{c}_t = v_{ck}\hat{k}_t + v_{cz}\hat{z}_t
\hat{k}_{t+1} = v_{kk}\hat{k}_t + v_{kz}\hat{z}_t$$

Insert the return equation to Euler equation to delete \hat{r}_t , and insert the law of motion to the log-linearized budget constraint and Euler equation.

Statistical Estimation I: solving the model

Undetermined coefficients:

$$V_{ck}$$
 V_{cz} V_{kk} V_{kz} ??

After getting these coefficients in terms of the parameters and steady state values, write the system as

$$\begin{bmatrix} \hat{k}_t \\ \hat{z}_t \end{bmatrix} = \begin{bmatrix} v_{kk} & v_{kz} \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \hat{k}_{t-1} \\ \hat{z}_{t-1} \end{bmatrix} + \begin{bmatrix} e_t \\ \varepsilon_t \end{bmatrix}.$$

To utilize the transformed data, write the system in log term,

$$\left[\begin{array}{c} \ln \bar{z}_t \\ \ln \bar{k}_t \end{array}\right] = \left[\begin{array}{c} \mu \\ g \end{array}\right] + \left[\begin{array}{cc} 0 & 0 \\ G_1 & G_2 \end{array}\right] \left[\begin{array}{c} \ln \bar{z}_{t-1} \\ \ln \bar{k}_{t-1} \end{array}\right] + \left[\begin{array}{c} \varepsilon_t \\ e_t \end{array}\right]$$

[What are g, G_1 and G_2 ?]



Statistical Estimation II: Estimating the model

With $y_t = (\ln ar{z}_t, \ln ar{k}_t)'$ and $x_t = (1, \ln ar{z}_{t-1}, \ln ar{k}_{t-1})'$, and

$$\Gamma = \left[egin{array}{ccc} \mu & 0 & 0 \\ g & G_1 & G_2 \end{array}
ight]$$
 , $\xi_t = \left[egin{array}{c} arepsilon_t \\ e_t \end{array}
ight]$

$$y_t = \Gamma x_t + \xi_t$$

With T observations, $Y=(y_1,...,y_T)'$, $X=(x_1,...x_T)'$ and $\Xi=(\xi_1,...\xi_T)'$, the stacked form is

$$Y = X\Gamma' + \Xi$$

Statistical Estimation II: Estimating the model

Estimation procedure

Assumption: $\xi_t \sim i.i.d.N(0, \Sigma)$, $\Sigma = \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & \sigma_2^2 \end{bmatrix}$.

Likelihood function:

$$L = \frac{1}{(2\pi)^T} |\Sigma|^{-T/2} \exp[-\frac{1}{2} \Sigma_{t=1}^T (y_t - \Gamma x_t)'(\Sigma)^{-1} (y_t - \Gamma x_t)]$$

Using the concentrated log-likelihood function

$$\ln L = const - (T/2) \ln |T^{-1}(Y - X\Gamma')'(Y - X\Gamma')|$$

Statistical Estimation II: Estimating the model

What's crucial in the estimation?

Data construction for y_t and x_t !

We have in the beginning: k_t , q_t , i_t , c_t . For y_t , we need $\ln \bar{k}_t = \ln(k_t/z_{t-1})$

$$\ln z_t = \left[\ln q_t - (1-\alpha) \ln k_t\right]/\alpha$$

So the data itself is a function of the underlying parameter sets (α, γ, β) . Each time of likelihood calculation given parameter set needs a lot of computation.

Statistical Estimation II: Estimating the model

$$\ln A_t = \alpha \ln z_t = \ln q_t - (1 - \alpha) \ln k_t.$$

Then the productivity process

$$\begin{array}{lcl} \ln A_t & = & \gamma + \ln A_{t-1} + \eta_t \\ \ln \bar{z}_t & = & \ln z_t - \ln z_{t-1} = 1/\alpha \left[\Delta \ln q_t - (1-\alpha) \Delta \ln k_t \right] \end{array}$$

Statistical Estimation II: Estimating the model

What's crucial in the estimation?

Writing out the second equation using $\ln ar{k}_t = \ln(k_t/z_{t-1})$

$$\begin{array}{lll} \ln \bar{k}_t &=& g + G_1 \ln \bar{z}_{t-1} + G_2 \ln \bar{k}_{t-1} + e_t \\ \ln k_t &=& g + \left[\ln q_{t-1} - (1-\alpha) \ln k_{t-1} \right] / \alpha \\ &+ G_1 \left[\Delta \ln q_{t-1} - (1-\alpha) \Delta \ln k_{t-1} \right] / \alpha \\ &+ G_2 \left[\ln k_{t-1} - (\ln q_{t-2} - (1-\alpha) \ln k_{t-2}) / \alpha \right] + e_t \end{array}$$

So the data itself is a function of the underlying parameter set (α, γ, β) . Each time of likelihood calculation given parameter set needs the computation of productivity.

Statistical Estimation II: Estimating the model

Implementing the MLE, you will obtain the parameter estimate

$$ilde{ heta}_{ extit{MLE}} = \left(ilde{lpha}, ilde{\gamma}, ilde{eta}
ight)$$
 .

Now each period, your estimated model gives predictions on productivity and capital.

$$\begin{split} \ln \tilde{A}_t &= \tilde{\gamma} + \ln A_{t-1} \\ &= \tilde{\gamma} + \ln q_{t-1} - (1-\alpha) \ln k_{t-1} \end{split}$$

The residual term is

$$\begin{split} \tilde{\boldsymbol{\eta}}_t &= & \ln A_t - \ln \tilde{A}_t \\ &= & \ln q_t - (1-\alpha) \ln k_t - \left[\tilde{\boldsymbol{\gamma}} + \ln q_{t-1} - (1-\alpha) \ln k_{t-1} \right]. \end{split}$$

Likewise, you obtain $\ln \tilde{k}_t$ and \tilde{e}_t .

Simulation: A counter-factual study

 Assume the model is true, the actual data are generated exactly by the dynamic equation with exogenous stochastic shocks

$$y_t = f(y_{t-1}, \theta) + \varepsilon_t$$

• With the estimated parameters

$$y_t = \tilde{f}(y_{t-1}, \tilde{\theta}) + \tilde{\varepsilon}_t$$

- In a counter-factual study, we ask what if $\tilde{\varepsilon}_t$ is different, i.e. $\check{\varepsilon}_t$? -> y_t changes to \check{y}_t .
- The difference, $y_t \check{y}_t$, is the effect of $(\tilde{\varepsilon}_t \check{\varepsilon}_t)$.
- The "counter-factual" \check{y}_t tells what y_t would have been, had $\check{\varepsilon}_t$ happened.

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962. 1) Productivity

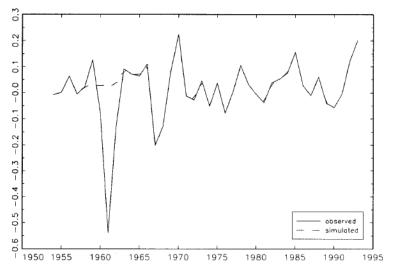


Fig. 1. Observed and simulated residual 1.



Simulation: A counter-factual study on the Great Leap Forward in 1958-1962. 2) Capital

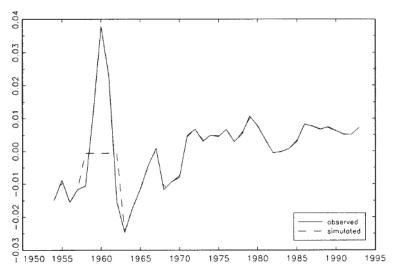


Fig. 2. Observed and simulated residual 2.

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

Question: What if the Great Leap Forward had not happened?

- Assume $\varepsilon_t = \left[\begin{array}{cc} \eta_t & e_t \end{array}\right]' = 0_{2\times 1}$, for t=1958,...,1962.
- ullet Keep $ilde{arepsilon}_t = \left[egin{array}{cc} ilde{\eta}_t & ilde{e}_t \end{array}
 ight]'$ for all other t.
- ullet Up to 1957: $ilde{f}(y_{t-1}, ilde{ heta})+ ilde{arepsilon}_t=y_t$, realized data.
- In 1958: $\check{y}_t \equiv \tilde{f}(y_{t-1}, \tilde{\theta}) + 0$, the simulated data begin to diverge from the realized ones.
 - From 1959 to 1962: $\check{y}_t \equiv \check{f}(\check{y}_{t-1}, \tilde{\theta}) + 0$.
- From 1963 onwards: $\check{y}_t \equiv \check{f}(\check{y}_{t-1}, \tilde{\theta}) + \tilde{\varepsilon}_t$.

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

Question: What if the Great Leap Forward had not happened?

Answer

- Compare \check{y}_t and y_t , i.e., $\left[\ln \check{A}_t \ln \check{k}_t \right]'$ and $\left[\ln A_t \ln k_t \right]'$.
- ullet \check{A}_t actually helps to pin down \check{q}_t , as

$$\ln \check{q}_t = \ln \check{A}_t + (1-lpha) \ln \check{k}_t.$$

- Convert the variables to their levels from log terms.
- What about consumption? Simply use the budget constraint in level.

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

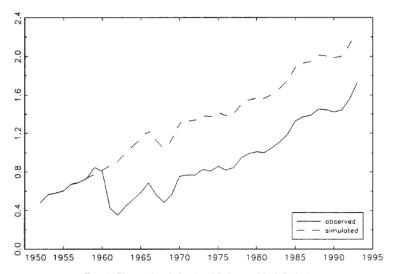


Fig. 6. Observed and simulated Solow residual (in log).

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

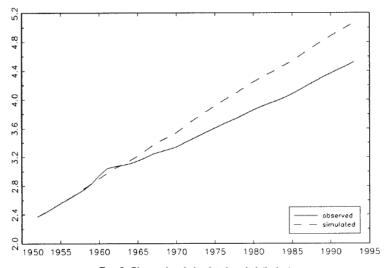


Fig. 5. Observed and simulated capital (in log).

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

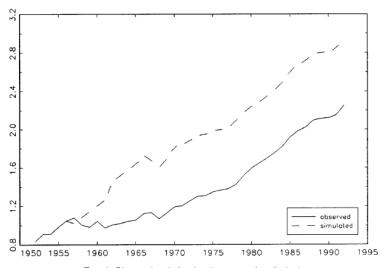


Fig. 4. Observed and simulated consumption (in log).

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

TABLE 3
SIMULATION/OBSERVED LEVEL IN 1992

	Great leap	Cultural revolution	Both
Output	2.0031	1.2033	2.7130
Consumption	2.0047	1.2022	2.7261
Capital	1.7208	1.1537	2.1687
Steady state	2.1074	1.2204	2.9238

Note. $(\hat{\alpha}, \hat{\beta}, \hat{\gamma}) = (0.7495, 0.9999, 0.0218).$

Simulation: A counter-factual study on the Great Leap Forward in 1958-1962.

Question

How a new steady state is computed with the simulated data?

Simulation: a counter-factual study on the Great Leap Forward in 1958-1962.

Robustness check

How sensitive are your results to the parameter estimates, and modeling assumptions?

Simulation: a counter-factual study on the Great Leap Forward in 1958-1962.

TABLE 4 Simulation/Observed Level in 1992

	Great leap	Cultural revolution	Both
Output	2.5446	1.2355	3.6549
Consumption	2.5680	1.2349	3.7277
Capital	1.9708	1.1643	2.5461
Steady state	3.2856	1.3111	5.2465

Note. α fixed at 0.5, $\hat{\beta} = 0.9715$, $\hat{\gamma} = 0.0083$.

Simulation: a counter-factual study on the Great Leap Forward in 1958-1962.

TABLE 5 Simulation/Observed Level in 1992

	Great leap	Cultural revolution	Both
Output	2.2907	1.2217	3.2082
Consumption	2.3008	1.2207	3.2459
Capital	1.8614	1.1597	2.3796
Steady state	2.6306	1.2648	3.9152

Note. α fixed at 0.6, $\hat{\beta} = 0.9817$, $\hat{\gamma} = 0.0132$.