# Estimating Economic Effects of Political Movements in China<sup>1</sup>

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### **Abstract**

To measure partially the economic effects of political movements in China a simple econometric model is constructed. Consumption, or equivalently investment, is determined by a central planner trying to maximize a multiperiod objective function. Political events are modeled by exogenous changes in the shocks to productivity and to investment which affect the time paths of major economic variables. Effects of the events are measured by comparing the time paths generated by the model with and without the changes in the shocks. The dynamic optimization model is estimated using data from 1952 to 1993. The results indicate that without the Great Leap Forward Movement output and consumption per capita in China would have been 2.0 times as great in 1993. Without the Cultural Revolution output and consumption per capita would have been 1.2 times as great.

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## 1. Introduction

What were the economic effects of the Great Leap Forward Movement in 1958-1962 and the Cultural Revolution in 1966-1969 in China? In other words if these two events had not occurred what would have been the time paths of the major economic variables such as consumption, real output and capital stock in the years following 1958? To answer this question one has to compare the historical time paths of these variables with the paths which would have prevailed absent the above events. We first construct an econometric model to explain the growth of the Chinese economy which incorporates the shocks from these two political events. Then the shocks are removed and the hypothetical time paths of the major economic variables are generated from the model. Comparing the hypothetical time paths with the time paths incorporating the shocks provides an answer to our question.

The econometric model has only one sector and includes aggregate output, consumption, investment, physical capital stock and total labor force as major variables. Aggregate output is produced by physical capital and labor according to a Cobb-Douglas production function. Output is divided into consumption and net investment (measured by "accumulation" in Chinese official statistics). Capital stock increases by the flow of investment. To determine investment we assume that actual investment equals planned investment plus an error. Planned investment is determined by the assumption that a central planner maximizes a multiperiod objective function with consumption per laborer as argument. The error may be affected by political events. The logarithm of total factor productivity follows a random walk with drift in normal years. In abnormal years such as during the Great Leap and the Cultural Revolution the residual of the random walk process can also be affected. Thus the effects of political events are modeled by changes in the error of the investment function and in the residual of the random walk process for productivity. Having estimated such a model one can remove the changes in order to measure the economic effects of the two political events. Section 2 specifies the model and the data. Section 3 presents the method of estimation and the parameter estimates. Section 4 reports on the time paths of major variables obtained by simulating the model absent the shocks from the two political events and provides measures of economic losses attributable to them. Section 5 concludes.

### 2. Model and Data

The econometric model consists of four equations. A Cobb-Douglas production function determines aggregate real output Q by physical capital stock K and labor L with constant return to scale. Denoting Q/L and K/L by q and k respectively and net investment per laborer by i, we can write the production function, the output identity, the capital accumulation equation and the equation explaining total factor productivity A as follows.

$$q_{t} = A_{t}k_{t}^{1-\alpha} \tag{1}$$

$$q_t = c_t + i_t \tag{2}$$

$$k_{t+1} = k_t + i_t (3)$$

$$\ln A_{t+1} = \gamma + \ln A_t + \eta_{t+1} \tag{4}$$

where  $\eta$  is a random shock to the logarithm of total factor productivity A. Note that the capital accumulation equation is obtained by dividing the original identity in aggregate variables by labor L in two adjacent periods and is therefore only an approximation.

The data for aggregate output Q are national income used (*Statistical Yearbook of China* 1994, abbreviated SYB, p.40) divided by the implicit price deflator of national income. The price deflator is the ratio of national income in current prices (SYB, p.33; measured in 100 million yuan) to national income in 1952 prices; the latter equals 589 (national income in 1952 in 100 million yuan) times the index of real national income (SYB, p.34; = 100 in 1952) divided by 100. In Chinese official statistics national income used equals consumption plus accumulation (net investment) in current prices. In our model this identity is assumed to hold in constant prices. We have estimated real national income used Q, real consumption C and real net investment I by dividing their current values (SYB, p. 40) by the above price deflator. Labor L is total labor force (SYB, p.88). Given K = 2213 (100 million yuan) in 1952 (an estimate from Chow (1993b, p. 821)), we estimate k in 1952 by K/L and k in later years by equation (3).

We assume that the Chinese economy evolves as if there were a central planner who, knowing the parameters of the model as we have specified, tries to maximize the following objective function at the beginning of each period t:

$$E_t \sum_{i=1}^{\infty} \beta^{t+i} \log c_{t+i} \tag{5}$$

subject to the constraints in (1) - (4). This dynamic optimization problem can be solved by defining the control variable as either consumption per laborer c, or investment per laborer i, or even next-period capital stock, as they are related by the identities (2) and (3). This maximization assumption might be questioned. A critic might argue that economic planners in China are not so rational as to have a specific objective function. She would say, just look at what happened to rational economic planning during the Great Leap and the Cultural Revolution. Our response is that during these abnormal periods there were exogenous shocks to the production and investment processes in China (caused to a large extent by the behavior of Chairman Mao!) which the economic planners could not control. However, given these shocks the planners still attempted in each period to maximize the above objective function from that period onward.

Among the possible shortcomings of this model are the treatment of technology, population and labor force as exogenous and the failure to incorporate possible effects through effects on human capital formation. Inspite of these possible shortcomings we believe that the present study is an important step towards measuring the economic effects of the two major political events and can serve as a benchmark for incorporating other important effects in future research.

# 3. Statistical Estimation

As discussed in the last section, the observed Chinese time series data on output, consumption and capital are interpreted as the outcome of a dynamic optimization process. The solution to the dynamic optimization problem will depend on the parameters  $(\alpha, \beta, \gamma)$  and the process governing the evolution of productivity. When we estimate the parameters by the method of maximum likelihood, we are in fact searching for a set of parameters for which the solution to the dynamic optimization problem and the observed series are as close as possible. A dynamic optimization problem is thus embedded within each evaluation of the likelihood function. More precisely, calculating the likelihood value for a given parameter setting proceeds in two stages. First, an optimal decision function for investment is determined by assuming that the central planner in China maximizes the objective function (5) subject to the constraints of the model (1) - (4) at each period t. Second, the optimal decision function is combined with the original model to form an econometric model for which the likelihood value can be calculated.

The dynamic optimization problem as stated in (1) - (5) can be converted into an equivalent version involving only stationary processes. The idea is to detrend all variables along their balanced growth paths. Define

$$z_{t} = A_{t}^{1/\alpha}, \quad \bar{k}_{t+1} = k_{t+1} / z_{t}, \quad \bar{c}_{t} = c_{t} / z_{t}, \quad \bar{z}_{t} = z_{t} / z_{t-1}.$$
 (6)

Replacing i by q - c and q by the production function, we can write the capital accumulation equation as

$$k_{t+1} = k_t + A_t k_t^{1-\alpha} - C_t,$$

implying

$$k_{t+1} / z_t = (k_t / z_{t-1}) z_{t-1} / z_t + k_t^{1-\alpha} z_t^{\alpha-1} - c_t / z_t,$$

or, in terms of the detrended variables defined in (6),

$$\bar{k}_{t+1} = \bar{k}_t \, \bar{z}_t^{-1} + \bar{k}_t^{1-\alpha} \, \bar{z}_t^{\alpha-1} - \bar{c}_t. \tag{7}$$

Similarly the productivity equation (4) can be written as

$$\ln \bar{z}_{t} = \mu + \varepsilon_{t}, \tag{8}$$

where

$$\mu = \gamma / \alpha, \quad \epsilon_{_{\it f}} = \eta_{_{\it f}} / \alpha.$$

Since  $z_t$  is exogenous, we may replace the objective function (5) by

$$E_t \sum_{i=1}^{\infty} \beta^{t+i} \ln \overline{C}_{t+i} \tag{9}$$

Maximizing (9) subject to (7) - (8) is equivalent to the non-stationary version in (1) - (5). We approach the dynamic optimization problem by first substituting (7) into (9) to eliminate the detrended consumption variable, and then define the control variable to be  $\ln \overline{k}_{t+1}$ , and two state variables  $\ln \overline{z}_t$  and  $\ln \overline{k}_t$ . With state and control so defined, we obtain numerically an approximate solution in the form of a log-linear first-order difference equation:

$$\ln \bar{k}_{t} = g + G_{1} \ln \bar{z}_{t-1} + G_{2} \ln \bar{k}_{t-1}$$
 (10)

The coefficients  $(g,G_1,G_2)$  may be regarded as reduced form parameters, as they are implicit functions of the three structural parameters  $(\alpha,\beta,\gamma)$ . The solution procedure and numerical algorithm can be found in the appendix.

Having derived the planned capital stock as described by (10), we allow actual capital stock to differ from planned capital by an error e due partly to failure of the planner to execute the plan and partly to failure of our simple model to capture the complicated economy completely. The econometric model to be estimated consists of two equations, an equation for  $\ln \overline{z}_i$  and an equation for  $\ln \overline{k}_i$ , which can be written as a system of two regression equations:

$$y_t = \Gamma x_t + \xi_t \tag{11}$$

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

$$\Gamma = \begin{bmatrix} \dot{\mu} & 0 & 0 \\ g & G_1 & G_2 \end{bmatrix} \qquad \xi_t = \begin{bmatrix} \varepsilon_t \\ e_t \end{bmatrix}$$
 (12)

With n observations (11) can be stacked up as

$$Y = X\Gamma + \Xi \tag{13}$$

with the transpose of (11) being the t th row of (13).

Assuming normal and serially uncorrelated residuals, and  $\xi$ , having covariance matrix  $\Sigma$ , we can use the well-known concentrated log-likelihood function (see Chow(1983), pp.170-171))

$$\ln L = const - (n/2) \ln n^{-1} (Y - X\Gamma)' (Y - X\Gamma)$$
(14)

and the maximum likelihood estimate of  $\Sigma$  is given by

$$\hat{\Sigma} = n^{-1} (Y - X\Gamma)' (Y - X\Gamma) \tag{15}$$

The observed variables are  $\ln q_t$  and  $\ln k_t$ , with  $\ln z_t$  defined as  $[\ln q_t - (1-\alpha)\ln k_t]/\alpha$ . Substituting this definition for  $\overline{z}_t = z_t/z_{t-1}$  in equations (8) and (10), we have

$$\ln q_t - (1 - \alpha) \ln k_t = \gamma + \ln q_{t-1} - (1 - \alpha) \ln k_{t-1} + \eta_t \tag{8'}$$

$$\ln k_{t} = g + [\ln q_{t-1} - (1-\alpha) \ln k_{t-1}] / \alpha$$

$$+ G_{1} [\Delta \ln q_{t-1} - (1-\alpha) \Delta \ln k_{t-1}] / \alpha$$

$$+ G_{2} [\ln k_{t-1} - (\ln q_{t-2} - (1-\alpha) \ln k_{t-2}) / \alpha] + e_{t}$$
(10')

The Jacobian of the transformation from  $\eta_t = \alpha \varepsilon_t$  and  $e_t$  to  $\ln q_t$  and  $\ln k_t$  is

$$J = \begin{vmatrix} 1 & -(1-\alpha) \\ 0 & 1 \end{vmatrix} = 1$$

which is implicit in the likelihood (14).

To calculate likelihood value for the parameters  $(\alpha, \beta, \mu)$  we use these parameters and the data on output and capital to compute z from the production function,  $\overline{z}_t$  and  $\overline{k}_t$  from equation (6), and the coefficients in  $\Gamma$  using equation (10). Thus the likelihood function (14) can be computed from the parameters and the data. We maximize the likelihood function in a sequential manner, i.e.  $\max_{\alpha} \max_{\beta,\mu} \ln L(\alpha,\beta,\mu)$ . The maximization with respect to  $(\beta,\mu)$  is performed by the MAXLIK package in GAUSS; and the line-search in  $\alpha$  is done by Brent's method (see Press et al. (1992, p.402-405)). The point estimate and standard error of  $\gamma$  can be recovered from that of  $\alpha$  and  $\mu$  via (8). To make sure that we have indeed located the global maximum, we have also used the simulated annealing algorithm as implemented by Goffe, Ferrier, and Rogers (1991) to maximize the likelihood function. The sample period is from 1954 to 1993.

The maximum likelihood estimates of  $(\alpha, \beta, \gamma)$ , with standard errors given in parentheses, are

$$(\alpha, \beta, \gamma) = [.7495 (.0108), .9999 (.0001), .0218 (.0025)]$$
 (16)  
mean log likelihood = 6.6120, sample size = 40.

The estimate 0.7495 for labor elasticity of production is reasonable. It is somewhat higher than the estimate of about 0.4 reported in Chow(1993b, especially Table VII); but the latter study uses a deterministic trend for log total factor productivity and a sample period ending in 1980 whereas the current estimate is based on a stochastic trend and a sample period extending to 1993. The estimate 0.9999 for the annual discount factor is also reasonable in view of the high value which Chinese planners are supposed to place on future consumption or current investment at the expense of current consumption. This parameter is considered difficult to estimate statistically and is often imposed *a priori* in empirical studies of real business cycles in the United States. The positive drift of log total factor productivity of 0.0218 is also reasonable as the sample includes the post-reform years 1978-1993. It is consistent with Chow (1993b) which found no positive deterministic trend in total factor productivity during the sample period from 1952 to 1980 but a positive trend from 1979 on. Unlike Chow

(1993b), the present study not only extends the sample period to 1993 but in estimating model parameters does not exclude any observations which are considered abnormal. This extension is possible partly because a stochastic trend is used for log total productivity rather than a linear deterministic trend as in Chow (1993b).

For sensitivity analysis we present below estimates for the remaining two parameters when the labor elasticity parameter is fixed *a priori* at other values sometimes chosen in growth accounting exercises (see e.g. Li, Gong and Zheng (1995)).

α	β		γ̂		mean log likelihood
0.4	0.9627	(0.0050)	0.0046	(0.0011)	5.9754
0.5	0.9715	(0.0037)	0.0083	(0.0017)	6.2012
0.6	0.9817	(0.0024)	0.0132	(0.0024)	6.3869
0.7	0.9940	(0.0015)	0.0194	(0.0033)	6.5456

# 4. Measuring the Effects of Two Political Events

To estimate the economic effects of the Great Leap Forward alone we change the estimated residuals of the two reduced form equations in the years 1958-1962 to the mean values of the corresponding residuals in the remaining years; see Figures 1 and 2. Columns 2 and 3 of Table 1 present actual output per laborer  $q_t$  (which can be generated by our model if the estimated residuals are used in the two equations) and simulated output  $q_t$ \* which is generated by our model if the estimated residuals in the years 1958-1962 are changed to the mean values of the remaining years. The remaining columns of Table 1 are the corresponding actual and simulated series for consumption, capital stock and log productivity.

From Table 1 and Figure 3 we observe that simulated output (which would have obtained absent the Great Leap) is about 2 times actual output in 1993. This result is derived from two sources. First, simulated total factor productivity in 1993 is about 0.56 higher than the actual value in logarithm, or about 1.74 times the actual value. Second, simulated capital stock per laborer in 1993 is 1.72 times the actual value, as can be readily computed from the relevant entries in Table 1. According to our model and commonly used models of real business cycles for the U.S. economy shifts in productivity due to shocks are permanent. Observe in Table 1 and Figure 4 that simulated log productivity in 1962 is .9058, or .5587 higher than actual log productivity. The last figure equals 2.2874 - 1.7286, the difference between simulated and actual log productivity in 1993. Such a parallel shift in log productivity due to the Great Leap is clearly shown in Figure 6. This is a characteristic of our model as equation (4) has a unit root which implies a permanent shift in total factor productivity when its residual changes. The permanent shift in productivity in turn implies that output, consumption and capital will all undergo a permanent level shift. There is no effect on the steady state growth rate of each variable.

To see the extent of the permanent level shift, we generate 500 residuals of zero mean and covariance matrix given in (15) and append them to the observed residuals

as well as our modified residuals. Output, consumption and capital are calculated according to these two extended residuals series. Examining the last 100 entries reveals that the steady state has been attained, as evident by the balanced growth of the three variables. Taking the ratio of the two output series gives the permanent level effect which we report in Table 2 (the row labeled as "steady state").

To assess the effect of the Cultural Revolution and the combined effect of the two Movements, we have performed similar simulation exercise as described above by removing residuals of the turbulent years. Table 2 provides a short summary for comparison with the Great Leap case; tables similar to Table 1 are available on request. For example, the output level by 1992 would have been 2.7 times higher than otherwise if both political movements had never occurred. To show the degree of sensitivity of our results, Tables 3-4 give similar comparisons when other parameter values reported in Section 3 are used.

Absent the Cultural Revolution output in China in 1992 would have been 1.20 times as large as the actual figure. This estimate might be considered too small. The possibility of under-estimation is mainly due to the omission of the effect on human capital formation in our model. Given that human capital is not considered and within the confine of our model, the measured effect appears reasonable. The disruption of the Cultural Revolution in the production of physical output in China is recognized to be much smaller than the disruption of the Great Leap. The relative magnitudes of 1.2 and 2.0 seem quite plausible. The Cultural Revolution is known for its effect on the production of human capital when many schools and universities were closed or ceased to function properly. The estimate of 1.2 can serve as a benchmark for studying the effects of the Cultural Revolution through its effect on the accumulation of human capital.

## 5. Conclusions

We have constructed a very simple econometric model to measure the effects of two major political events in China. The model is based on a dynamic optimization framework. It is assumed that an economic planner in China tries to maximize a multiperiod object function in making consumption and investment decisions. The values of the parameters of the optimization model as estimated by maximum likelihood are reasonable. The dynamic optimization framework is useful for studying economic behavior and the effects of political events in China as in other countries.

Concerning the effects of the Great Leap and the Cultural Revolution, our results indicate that absent the former output and consumption per laborer in 1990 would have been 2.0 times as large as the observed, that absent the latter output and consumption would have been 1.2 times as large and that if neither had occurred output and consumption would have been 2.7 times the actual amounts.

Two referees have kindly made comments to the above paper which require responses. The first raises a philosophical question as to why the estimates provided are interesting. He/she asks:

"How can we say that if these events did not happen we would be two times better off today? If they had not happened, then there would have been no Mao, and with no Mao, maybe no revolution, and no Soviet type economy - then where would be Chinese economy be today? And if they had not happened, then maybe there would have been no reform and where would China be today? In other words, how can we take one event out of its environment and say if this or that did not happen, then this is what would have happened. These events were part of that environment, i.e., a Leninist Party and democratic-centralism, with crazy economic campaigns from time to time. This is a very interesting game to play and estimate to make, but what do they do for us?

A partial response is that Mao could have had a serious illness in 1957 and passed away, and we ask what would have happened to China's economy then? The reader should decide how interesting such a question is.

The second referee raises three technical points. First, one could get a closed form solution to the model if we assumed the entire stock of capital depreciate completely in a year, and why not use such a model. Our response is that the assumption is poor and that there is nothing wrong in using a numerical rather than an analytical solution to obtain equation (10) for estimation purpose. Second, there may be a unit root problem in econometric analysis. The answer is that there is no unit-root problem if we difference the data, as we did by using  $\ln z_t - \ln z_{t-1}$  instead of  $\ln z_t$ . There is no unit root in the differenced data. Third why not use VAR? We are open minded about different modelling approaches. The advantage of a structural, dynamic optimization approach is that we have a deeper understanding of economic behavior, e.g., we have estimated  $\beta$  = .9999 to show the preference of the Chinese planners for future consumption as compared with current consumption. A VAR approach would not provide very different estimates for the two equations in the model, as the equation for  $\ln z_{t+1}$  -  $\ln z_t$  fits the data well. except that there may be more lagged variables in the equation for  $\ln k_{t+1}$ . We will report on the results from different models, such as introducing more lagged variables in the equation for  $\ln k_{t+1}$  and using a state-space model together with a Kalman filter, in another paper to satisfy curious readers.

# **Appendix**

A standard dynamic optimization problem is to choose a sequence of q by 1 control vectors  $\{u_t, t = 0,1,2,...\}$  to solve

$$\max E_0 \sum_{t=0}^{\infty} \beta^t r(x_t, u_t)$$
 (A1)

subject to

$$x_{t+1} = f(x_t, u_t) + \varepsilon_{t+1}.$$
 (A2)

where  $E_0$  is the conditional expectation operator given information at time 0,  $x_t$  is a p by 1 vector of state variables, and  $\varepsilon_t$  is an iid random vector with mean zero and covariance matrix  $\Sigma$ . Our problem is to solve

$$\max E_0 \sum_{t=0}^{\infty} \beta^t \ln \{ \overline{k}_t^{1-\alpha} \overline{z}_t^{\alpha-1} - \overline{k}_{t+1} + \overline{k}_t \overline{z}_t^{-1} \}$$
 (A3)

subject to

$$\ln \overline{z}_{t+1} = \mu + \varepsilon_{t+1}.$$
(A4)

Our problem can be mapped into the standard form by defining the states and control as

$$x_t \equiv (x_{1t}, x_{2t})' = (\ln \overline{z}_t, \ln \overline{k}_t)', \quad u_t = \ln \overline{k}_{t+1}.$$
 (A5)

The objective function and the constraint are respectively

$$r(x_t, u_t) = \ln\{\exp((1 - \alpha)(x_{2t} - x_{1t})) - \exp(u_t) + \exp(x_{2t} - x_{1t})\}$$
 (A6)

and

$$f(x_t, u_t) = Ax_t + Cu_t + b, \tag{A7}$$

where A is a 2 by 2 zero matrix, C = (0,1), and  $b = (\mu,0)$ . The steady state  $(\overline{\mu}, \overline{x})$  can be found by solving a deterministic, time invariant version of the first order conditions. For our choice of state and control as in (A5), the steady state values are

$$\overline{u} = -\frac{1}{\alpha} \ln[\beta^{-1} \exp(\mu) - 1] + \frac{1}{\alpha} \ln(1 - \alpha) + \mu, \quad \overline{x}_1 = \mu, \quad \overline{x}_2 = \overline{u}.$$
 (A8)

Only in exceptional cases would one be able to find an analytical solution for the optimal control function. In most applications one has to rely on numerical approximation. One convenient way to do so has been developed in Chow (1992, 1993a). We now describe briefly the solution procedure. Consider the first order conditions:

$$r_2(x_t, u_t) + \beta f_2'(x_t, u_t) E_t \lambda_{t+1} = 0,$$
(A9)

$$r_{1}(x_{i}, u_{i}) + \beta f_{1}'(x_{i}, u_{i}) E_{i} \lambda_{i+1} = \lambda_{i},$$
(A10)

$$x_{t+1} = f(x_t, u_t) + \varepsilon_{t+1}. \tag{A11}$$

where the subscripts 1 and 2 of the functions r and f denote derivatives with respect to the first and second arguments respectively.  $\lambda$  is a vector of random Lagrange multipliers. We proceed by linearizing the non-linear functions in (A9) - (A11) around the steady state ( $\bar{x}, \bar{u}$ ):

$$f = Ax + Cu + b;$$
  $r_1 = K_{11}x + K_{12}u + k_1;$   $r_2 = K_{21}x + K_{22}u + k_2$  (A12)

Given the linear functions above, if  $\lambda$  is assumed to be linear, say equal to Hx + h, substituting these functions in the first order conditions will yield a linear control function

$$u = Gx + g \tag{A13}$$

where

$$G = -(K_{22} + \beta C HC)^{-1}(K_{21} + \beta C HA)$$
(A14)

$$g = -(K_{22} + \beta C HC)^{-1} (k_2 + \beta C (Hb + h))$$
(A15)

and the coefficient matrices of the Lagrangean function are respectively

$$H = K_{11} + K_{12}G + \beta A' H(A + CG)$$
(A16)

$$h = (K_{12} + \beta A' HC)g + k_1 + \beta A' (Hb + h)$$
(A17)

Iterating the matrix equation system (A14) - (A17) until convergence gives the required matrices G, g, H and h. We have accelerated such a direct iteration scheme by incorporating a modified version of the doubling algorithm described in Anderson and Moore (1979, p.159). A detail discussion of the algorithm and numerical examples will be reported elsewhere.

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Table I: Great Leap Forward effect

Table I: Great Leap Forward effect output consumption				capital stock log productivity				
	out			simulated		simulated		simulated
year	observed	simulated	observed		10.676	10.676	0.48132	0.48132
1952	2.9283	2.9283	2.3011	2.3011	11.303	11.303	0.56285	0.56285
1953	3.2227	3.2227	2.4780	2.4780 2.4794	12.048	12.048	0.57888	0.57888
1954	3.3276	3.3276	2.4794		12.896	12.896	0.60262	0.60262
1955	3.4661	3.4661	2.6715	2.6715	13.691	13.691	0.67214	0.67214
1956	3.7717	3.7717	2.8500	2.8500	14.612	14.612	0.69025	0.69025
1957	3.9038	3.9038	2.9310	2.7747	15.585	15.741	0.73053	0.73336
1958	4.1304	4.1525	2.7289	2.9482		16.946	0.84648	0.77647
1959	4.7393	4.4162	2.6635	3.1322	16.986	18.230	0.80816	0.81959
1960	4.6947	4.6959	2.8339	3.3274	19.062	19.598	0.42542	0.86270
1961	3.2774_	4.9924	2.6465	3.5344	20.923	21.056	0.34707	0.90581
1962	3.0530	5.3069	2.7342	4.2926	21.554		0.43750	0.99625
1963	3.3543	5.8781_	2.7680	4.5729	21.873	22.070	0.43730	1.0714
1964	3.6400	6.4285_	2.8314	4.8570	22.459	23.376		1.1413
1965	3.9385	7.0076	2.8712	5.1104	23.268	24.947	0.58258 0.68628	1.2450
1966	4.4182	7.9173	3.0654	5.6422	24.335	26.844		1.1153
1967	3.9337	7.0975	3.0963	5.3530	25.688	29.119	0.55660	1.0409
1968	3.6809	6.6849	2.9024	4.9747	26.525	30.864	0.48213	1.1236
1969	4.0273	7.3599	3.0919	5.4272	27.303	32.574	0.56482	1.3130
1970	4.9087	9.0246	3.2916	6.2019	28.239	34.507	0.75429	1.3256
1971	5.0405	9.3203	3.3235	6.2964	29.856	37.329	0.76684	
1972	5.1180	9.5159	3.5018	6.5478	31.573	40.353	0.76810	1.3268
1973	5.4831	10.249	3.6789	6.9632	33.189	43.321	0.82451	1.3833
1974	5.4627	10.262	3.6958	6.9540	34.993	46.607	0.80751	1.3663
1975	5.8133	10.974	3.8447	7.3186	36.760	49.915	0.85738	1.4161
1976	5.6731	10.759	3.9225	7.3361	38.729	53.570	0.81990	1.3786
1977	5.8764	11.194	3.9762	7.4969	40.479	56.993	0.84404	1.4028
1978	6.5737	12.576	4.1718	8.0845	42.379	60.690	0.94469	1.5034 1.5492
1979	6.9773	13.402	4.5635	8.8103	44.781	65.181	0.99046	
1980	7.1944	13.873	4.9267	9.4212	47.195	69.773	1.0080	1.5667
1981	7.2277	13.990	5.1806	9.8049	49.463	74.225	1.0008	1.5596
1982	7.6748	14.908	5.4636	10.415	51.510	78.410	1.0507	1.6094
1983	8.2453	16.072	5.7936	11.144	53.721	82.903	1.1119	1.6706
1984	9.0219	17.643	6.1797	12.032	56.173	87.831	1.1907	1.7494
1985	5 10.490	20.579	6.8177	13.562	59.015		1.3291	1.8878
1986	6 11.107	21.854	7.2579		62.687	100.46	1.3711	1.9298
198'	7 11.438	22.570	7.5423		66.536		1.3855	1.9443
198	8 12.408	24.554	8.1293		70.431	115.44	1.4528	2.0115
1989	9 12.492	24.785	8.2737		74.710		1.4447	
199	0 12.409	24.684	8.3370		78.928		1.4243	1.9830
199	1 12.806	5 25.536	8.6106		83.000		1.4432	2.0019
199	<b>2</b> 14.512	29.005	9.5145	19.074	87.196		1.5559	2.1147
199	<b>3</b> 17.491	35.036			92.194	158.65	1.7286	2.2874
1								
mean	6.7323	3 12.244	4.4410	7.9904	39.606		0.90151	
std de					22.816	42.633	0.35652	0.47905
Jett at	2.07.1							

TABLE 2 Simulation/Observed Level in 1992

	Great Leap	<b>Cultural Revolution</b>	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

 $<sup>(\</sup>hat{\alpha}, \hat{\beta}, \hat{\gamma}) = (0.7495, 0.9999, 0.0218)$ 

TABLE 3
Simulation/Observed Level in 1992

	Great Leap	<b>Cultural Revolution</b>	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

 $<sup>^{</sup>a}$  α fixed at 0.5,  $\hat{\beta} = 0.9715, \hat{\gamma} = 0.0083$ .

TABLE 4
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

<sup>&</sup>lt;sup>a</sup> α fixed at 0.6,  $\hat{\beta} = 0.9817, \hat{\gamma} = 0.0132$ .

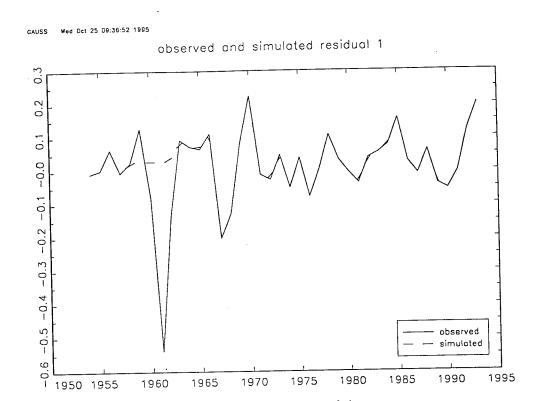


Figure 1: Observed and simulated residual 1

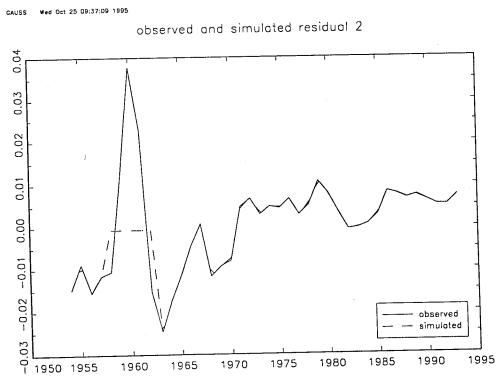


Figure 2: Observed and simulated residual 2

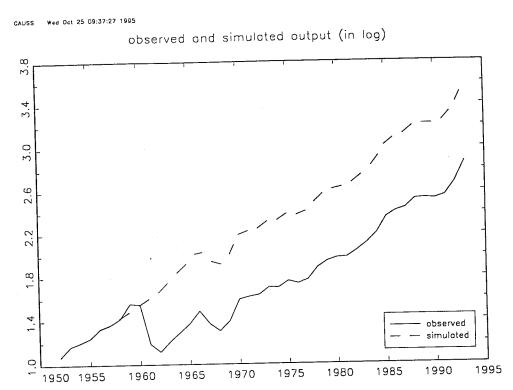


Figure 3: Observed and simulated output (in log)

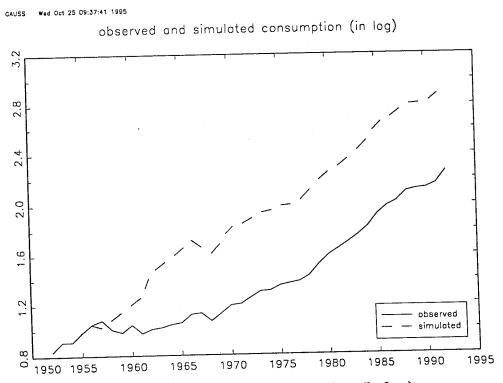


Figure 4: Observed and simulated consumption (in log)

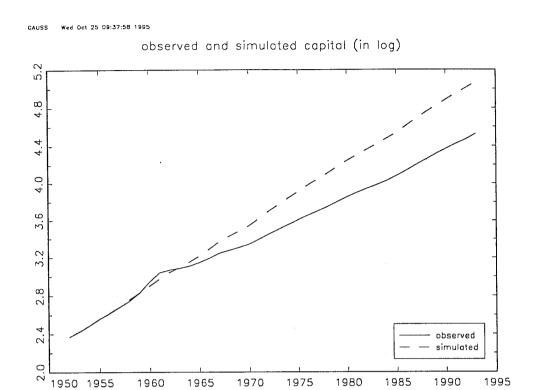


Figure 5: Observed and simulated capital (in log)

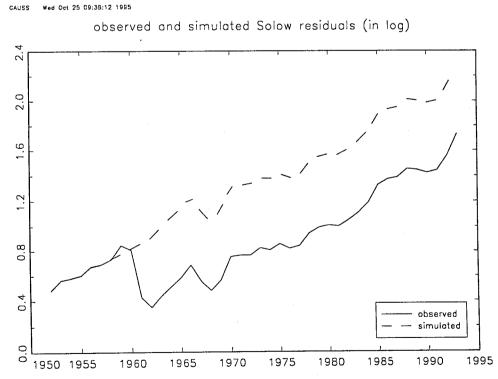


Figure 6: Observed and simulated Solow residual (in log)